

Convicts and Comrades*

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Abstract

What role did coerced labor play in establishing the first labor unions? This paper introduces a model where certain firms employ convict labor, reproducing the empirical patterns observed in the data. As a result, workers face reduced wages and migrate to other firms, while firms see heightened profits. In response, workers organize, form unions, and initiate strikes. I use an instrumental variable approach to demonstrate that, at the turn of the 20th century, Black convict labor significantly boosted white union growth, strikes, and membership in the Southern United States. My empirical findings further suggest that this influence has persisted as counties with a history of heavy dependence on convict labor continue to display higher rates of union membership in the present day.

Key words: Wages, Coerced Labor, Labor Unions

JEL codes: D5, J3, J47, J5

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1 Introduction

Workers often turn to unions to address power imbalances with employers. Individually, a worker may lack substantial bargaining power; however, collectively, they can negotiate better wages and improved working conditions, thereby rebalancing power. In the United States, the value of this collective strength is evident, as union members typically earn 10%–20% more than non-union members (Farber et al., 2021).

Certain events, such as the rise of globalization and automation, can drive workers to unionize due to changes in their working conditions (Acemoglu and Restrepo, 2018; Autor and Salomons, 2018; Autor et al., 2020). In this paper, I identify coerced labor as a novel factor that stimulates labor organization by suppressing labor income.¹ The post-slavery period in American history, marked by the emergence of convict labor, provides a unique setting to study the impact of depressed wages on worker unionization.²

Convict labor arose in the aftermath of the Civil War as the Southern United States sought alternative labor sources in the wake of emancipation. This practice of employing incarcerated individuals, which continued until the mid-20th century, is contentious in United States history; some deem it rehabilitative, others exploitative, undermining free labor (Lichtenstein, 1996).³ The use of convict labor posed challenges for free workers, who were alarmed by competition from low-cost coerced labor and potential wage impacts (Hiller, 1914). I argue that these concerns pushed wage workers to collectively counteract wage suppression, making the institution of convict labor a critical factor in forming the first labor unions. Understanding the origins and effects of unionization unravels the complexities of workers’ socio-economic challenges, forming a crucial foundation for today’s labor movements and illuminating critical aspects of labor market dynamics.

My paper offers the first theoretical and empirical analysis of the impact of convict labor at the dawn of the 20th century on the development of the first labor unions in the American South. Motivated by the observable decline in wage growth and rise in unions following the introduction of convict labor in the southern United States, I develop a theoretical model that illustrates the specific dynamics of late 19th-century convict labor employment. This model introduces a novel general equilibrium framework to the literature, capturing the labor market impacts of convict labor and informing my empirical analysis.

The model categorizes labor into two types: convict and free labor. Convict laborers

¹Coerced or unfree labor refers to work performed under duress or threats, encompassing all forms of slavery, penal labor, and institutions like debt slavery, serfdom, and labor camps.

²This era is ideal for investigating the influence of suppressed wages on unionization, especially as it lacks minimum wage laws and numerous other labor policies that might otherwise complicate the analysis.

³Historically, free labor denoted work by those not legally bound to serve others.

receive no salary, while free laborers are compensated with wages. Convict labor is a limited resource; every convict in prison is employed as a laborer (U.S. Bureau of Labor, 1886). Thus, the model imposes a stringent capacity constraint on a firm's use of convict labor, a unique aspect of its novelty. Consequently, any increase in the exogenously imposed capacity constraint affects the labor market and union formation. The model predicts a decline in the equilibrium wage for free laborers as the capacity for convict labor expands, which in turn enhances firm profitability. Finally, the model demonstrates that greater use of convict labor increases union activities among free laborers due to its effects on suppressing wages.

To empirically investigate this relationship, I have digitized the Bureau of Labor Reports from 1886 to 1940 for the first time, measuring the extent of convict labor employment across counties to assess its impact on union formation. To address potential unobserved factors that could influence the use of convict labor and union formation, I employ an instrumental variable estimation method in my analysis. In this approach, I use the locations and capacities of prisons built before the introduction of convict labor as instruments for convict labor exposure. The pre-existing prison locations and capacities positively correlate with the number of convict laborers following the enactment of convict labor laws.

The instrument's validity relies on the critical assumption that the influence of prisons (locations) built before the era of convict labor and their capacities on union formation is solely due to their subsequent effect on convict labor usage. To support the credibility of this assumption, I provide evidence showing that exposure to prison locations and capacities from the pre-convict labor era does not correlate with changes in union formation before the introduction of convict labor, indicating that the instruments are not associated with pre-existing trends. Additionally, I show that these prisons had no connection to labor market outcomes before the implementation of convict labor laws, alleviating concerns that the prisons directly impacted local labor markets. Supplementary analyses further reinforce the instrument's validity.

Analyses indicate that a 10% increase in the convict labor share, on average, leads to a 9.5% increase in union assemblies and a 16% rise in strikes between 1886-1905. This surge in convict labor had profound implications for union activity. Not only did it lead to a significant increase in the number of active unions, but it also boosted their longevity. The rise in the establishment of new union assemblies further emphasizes this trend. As convict labor became more prevalent, the number of new unions grew, and their longevity extended. The ripple effect of this dynamic interaction persists; counties with a historically 10% higher prevalence of convict laborers still witness a 2.2% higher union membership today. This finding, documented for the first time in this paper, highlights the deep-rooted influence of convict labor on the institutional development of unions.

In addition, convict labor significantly impacts local economies, particularly affecting the working conditions of free laborers in areas with a high presence of convict labor. Empirical findings show a decline in per-worker wage growth and increases in manufacturing output value and gross profits, consistent with the model’s predictions. The growing disparity between free laborer wages and firm profits confirms the assertions of [Alston and Ferrie \(1993, 1999\)](#) and [Althoff and Reichardt \(2022\)](#) that the American South’s oppressive institutions predominantly served the elite’s interests.

The main factor driving these results is labor market competition. Convict labor depresses wages, compelling wage workers to unite. Supporting this, counties without unions experience a statistically significant decrease in average manufacturing wage growth following an increase in convict labor, whereas counties with active unions do not see a statistically significant change in wage growth. An additional factor is convict labor’s role in fostering employment in crafts and operative roles, which have historically been more susceptible to unionization. However, the shift towards these roles is less pronounced in areas with active unions, as unions help mitigate wage reductions for laborers.

Finally, this paper presents a fresh perspective on racial segregation in the United States, a persistent issue rooted deeply in the country’s post-Civil War history. It emphasizes the substantial damage inflicted by the convict labor system, shedding light on its role in aggravating the conditions for Black convict laborers and inadvertently deteriorating the labor market for white individuals. This system fuelled union formation as a protest against Black convict labor, possibly reinforcing enduring racial divisions and suggesting contemporary U.S. segregation issues may have origins in this historical context.

Related Literature. This paper is the first to quantify the impact of convict labor on union formation, contributing to several strands of economic literature. Despite extensive research on labor unions, the history of the American labor movement has received less attention. My work addresses this gap, building on foundational studies like [Bernstein \(1954\)](#); [Commons et al. \(1918\)](#); [Foner \(1978\)](#); [Friedman \(1999\)](#); [Naidu and Yuchtman \(2016\)](#); [Schmick \(2018\)](#); [Taft \(1964\)](#). It offers the first comprehensive theoretical and empirical evidence of a key factor driving the formation and development of labor unions. Additionally, this paper adds to the debate on the drivers of inequality, emphasizing how companies’ ability to access low-cost coerced labor contributes to this issue ([Card et al., 2013](#); [Song et al., 2019](#); [Autor et al., 2020](#)).

I build upon the existing literature on coercive institutions, including studies on the lasting impacts of forced labor ([Dell, 2010](#); [Nunn and Wantchekon, 2011](#); [Lowe and Montero, 2016](#)), and its consequences for institutional development ([Nunn, 2008](#); [Acemoglu et al., 2012](#); [Bugge and Nafziger, 2021](#)). This paper extends this research by exploring the previously

unexplored relationship between convict labor and union formation. Moreover, it features the lasting path dependence of union membership in these labor markets, even post-era of convict labor. This underscores the lasting impact of historical institutional choices on today’s labor market outcomes and institutional development. I enrich this literature by establishing the first causal link between convict labor and unions’ rise and enduring presence.

Furthermore, this paper leads an examination of the impact of convict labor on the emergence of labor unions, expanding upon the literature on post-slavery prisoner employment. While [Muller \(2018, 2021\)](#); [Muller and Schrage \(2021\)](#) and [Rubio \(2022\)](#) explores the racial disparity in incarceration rates and [Poyker \(2019\)](#) investigates the short-term economic consequences of prison-made products, this paper emphasizes the broader and enduring influence of convict labor on laborers’ strategies and organization.⁴ Furthermore, [Archibong and Obikili \(2020\)](#) explores the implications of labor demand shocks on incarceration rates, underlining the value of prison labor to colonial Nigeria’s exploitative system. In this context, I establish that convict labor as a post-slavery coercive institution primarily benefits the elites at the expense of the working class.

The paper is structured as follows: Section 2 provides the institutional context, Section 3 covers the data and stylized facts, Section 4 introduces the model, Section 5 details the identification strategy and reports convict labor’s effects on union formation and the labor market and Section 6 concludes.

2 Institutional Context

2.1 Crime and Punishment in the American South

Before the Civil War, crime was not a concern in the Southern U.S. The honor culture’s informal system of addressing grievances made personal violence a key aspect of crime. Only the most dangerous criminals were subjected to the centralized state penitentiary system, and even they often avoided incarceration. Southern courts operated on the principle that it was better for 99 guilty men to go free than for one innocent man to suffer ([Ayers, 1984](#)).

Unlike Northerners, Southern states were uniquely anxious about building prisons before the Civil War. Persistent debates centered on republican principles, specifically the state’s role in social governance. [Ayers \(1984\)](#) explains that for many Southerners, “repub-

⁴[Muller \(2018\)](#) contrasts with [Rubio \(2022\)](#), showing that Black incarceration rates were lowest in the cotton belt. Former slaveholders preferred exploiting Black agricultural labor locally rather than losing workers to the convict lease system, using data on prisoners’ county of conviction (not their county of incarceration). [Muller \(2018\)](#) reveals that primarily industrial and extractive employers, not agricultural ones, utilized convict labor.

licanism” meant freedom from external control, and they saw centralized power as more harmful than beneficial. The majority of Southerners were against establishing penitentiaries. For instance, in both Alabama and North Carolina, the penitentiary system was decisively rejected by voters (Hawks, 1971).⁵ The Southern reluctance to establish prisons was influenced by the slave system, which made a white criminal class undesirable (Hindus, 2017). Southerners believed that prisons endangered American freedom and the principles of the Revolution. Thus, prisons were rare in the pre-Civil War South, with little objective need for their construction. Despite the population’s opposition, Southern states eventually built penitentiaries in the early 19th century (Ayers, 1984).⁶

In the era of slavery, virtually no free black Southerners were imprisoned, and no slaves were incarcerated (Ayers, 1984). Slaves accused of lesser crimes were typically tried informally in extra-legal plantation “courts” (Hindus, 2017). The overwhelming majority of Southern inmates during the antebellum period were white (Ayers, 1984). The Civil War significantly transformed Southern society and its criminal justice system. For the first time, freed slaves fell under the jurisdiction of local governments (Ayers, 1984). The market economy began to influence previously untouched Southern regions. Severe poverty at the end of the 19th century dismantled the South’s race-based social structure (Ayers, 1984; Byrne, 1979). Southern legal institutions then focused on preserving white supremacy, leading to a racialized incarceration system and convict labor that persisted into the 20th century (Ayers, 1984; Blackmon, 2009).

Following the Civil War, the transition to convict labor corresponded with an increase in Black incarceration. For instance, in Georgia, fifteen years after the war, Blacks were imprisoned at over 12 times the rate of whites (Muller, 2018). However, convict labor was not just an attempt to reestablish slavery but mirrored continuities in race relations and changes in the Southern economy (Ayers, 1984; Gottschalk, 2006). Millions of freed slaves came under state penal control, and the South, facing labor shortages due to industrial growth, found former slaves to be the easiest demographic to exploit.⁷

⁵Traditional public punishments were seen as the most republican form of justice because of their transparency. Before prisons, criminal law relied on fines, whipping, branding, and execution (Ayers, 1984).

⁶Before the Civil War, Southern legislators enacted prison laws despite public opposition. Ayers notes their mixed motives: some believed they knew what was best, while many elites had a “class control” motive but used their penitentiary support to show benevolence (Ayers, 1984).

⁷Gottschalk (2006) sees convict labor as a crucial bridge between a slavery-based agricultural economy and the New South’s industrialization.

2.2 Convict Labor in the United States

The 13th Amendment's provision for slavery or involuntary servitude as a punishment for crime led to the creation of local laws and institutions designed to limit the freedom and economic opportunities of free Black people. Black people could be imprisoned for vagrancy or other petty offenses and then funneled into the convict labor system to be exploited as cheap labor.

The use of convict labor grew after the formation of the National Prison Association in 1870. It was widely adopted in almost all states by the end of President Rutherford Hayes' term in 1881 (Wines, 1871). In 1870, approximately 10.6% (3,500 out of 32,901) of the nation's prisoners worked as convict laborers, all located in New York. By 1886, this number had significantly risen to 50,000 prisoners working in the convict labor system.⁸

Laws regarding prisoner labor varied in terms of profitability for the state and other parties involved and the working conditions for prisoners. For example, in Southern states, prisoners worked 16 hours a day, while in the North, the workday was limited to 8 hours. The U.S. Commissioner of Labor reported that prison labor was significantly cheaper than other forms of labor, estimated to be only 19% of the cost of free labor (U.S. Bureau of Labor, 1886, 1906, 1925). As a result, convict labor became a lucrative source of income for state governments, with prisoners working long hours under different systems such as contract, piece-price, state-account, state-use, public-works systems, and convict leasing.⁹

The convict lease system was widely adopted in the Southern states after the American Civil War and persisted well into the 20th century. Under this system, convicts were rented out to private companies, which were given complete control over the prisoners, including responsibilities for guarding, feeding, and disciplining them. Convict leasing was considered the most profitable form of convict labor for the Southern states (U.S. Bureau of Labor, 1886, 1914).¹⁰ However, convicts were also contracted out to private firms under contract and piece price systems.

As a significant portion of the workforce, the convicts typically worked without pay. The inexpensive nature of prison labor pressured free workers and boosted business profitability. This reliance on prison labor threatened free workers by driving down their wages. Therefore, the relationship between convict labor and free labor competition became increasingly relevant at the end of the 19th century (Mohler, 1925).

The introduction of convict labor marked the beginning of a slow, complex transition

⁸The Online Appendix Figure D1 depicts the prison labor utilization from 1870 to 1940 in the Southern United States.

⁹For detailed information on these systems, see the Supplementary Appendix Section K.

¹⁰For instance, in 1910, Florida alone secured \$346,000 (\$10,896,086 in 2023) net profits from the convict lease system (Mohler, 1925).

of Black individuals in the industrial labor force.¹¹ For over a century, the workforce of the Southern manufacturing sector was predominantly white, a pattern dating back to antebellum times. During the economic downturn of the 1840s, a limited number of enslaved individuals were leased to textile mills, but this practice was short-lived. They were quickly withdrawn in favor of agricultural labor during the booming 1850s (Starobin, 1970). As the regional industry began to recover and expand after the Civil War, it primarily relied on white workers, a choice deeply influenced by racial favoritism (Wright, 2013). This indicates that integrating Black labor into the industrial workforce lagged nearly a century behind emancipation.

2.3 Origins of the American Labor Movement

In the 19th century, American labor unionism emerged, with the Knights of Labor (KOL) becoming the third national labor federation in the United States.¹² While it shared similarities with previous federations, such as the emphasis on political action and independent producers, the Knights differentiated itself by including a broader range of workers. The Knights were the leading labor society of the 1880s, reaching its peak with a fifth of the workforce in 1886 (Friedman, 1988; Voss, 1993).¹³

Throughout this period, one of the biggest challenges free labor faced was competition with convict labor, as it burdens their wages and job security. In the words of one union representative, “The prison labor system has unquestionably been the means of lowering the wage rate for thousands of wage-earners, and in some instances, its competition has practically driven an industry from the field. It is because of this competition and for humanitarian reasons that trade unions have been opposed to its existence” (Frey, 1913). During that period, free laborers grappled with rising merchant capitalists who aimed to cut costs, especially labor wages, using cheap prison labor. This reliance on convict labor posed an economic challenge for free laborers. They saw it as not only threatening their livelihoods but also as the government’s abuse of trust. Consequently, free labor sought to limit the

¹¹In the South, racial discrimination in labor markets was severe, with jobs often divided by race. The integration of free Black labor into industry lagged nearly a century after slavery’s abolition. For instance, in states like Florida, Georgia, and Louisiana, no Black men were among thousands of telegraph operators, and fewer than 50 out of nearly 15,000 locomotive operators were Black. Southern textile industries, which employed many convict laborers, remained predominantly white. In South Carolina, Black workers made up only 5% of the workforce and were confined to manual labor roles until the Civil Rights movement.

¹²The National Trades’ Union, the first U.S. labor federation founded in 1834, collapsed in the 1837 financial crisis. The National Labor Union, established in 1866 to unite local trade unions, dissolved by 1873 due to unmet goals.

¹³The AFL-CIO, the largest U.S. union federation with over 12 million members, represents a smaller percentage of American workers today compared to the KOL in 1886, which had up to 20% of all workers affiliated. (Kaufman, 2001).

impact of convict labor on fair competition by restricting their employment (Mohler, 1925).¹⁴

Qualitative evidence suggests that labor unions advocated protecting the rights and wages of free workers by supporting restrictions on convict labor and working to ensure fair wages and conditions for the free workers. Specifically, the Knights made a strong statement against convict competition at their 1886 assembly in Richmond, where they called for states to enact laws limiting the employment of convicts. The Knights’ demands also included preventing the shipment of prison-made goods, branding such goods, limiting convicts’ access to powerful machinery, abolishing the contract, lease, and piece-price systems. Their aim was to shield free labor from this unfair competition (U.S. Bureau of Labor, 1886). The preamble of the KOL “declared to the world” that its main objective is “to prohibit the hiring out of convict labor” (Ely, 1886).

3 Data and Stylized Facts

3.1 Novel Data on Convict Labor in the U.S.

To investigate empirically the effect of convict labor on the formation of first labor unions, I gathered data from several sources. First is a series of reports published by the U.S. Bureau of Labor on convict labor. To address the debate over convict and free labor competition, the Bureau of Labor surveyed penal institutions roughly every decade, producing reports on convict labor that detailed prisoner employment numbers and U.S. correctional facility outputs. Bureau of Labor staff conducted on-site surveys using prisons’ accounting records. The data encompassed various penal institutions, from prisons and labor camps to juvenile reformatories and county jails employing prisoners. It also detailed the prison’s name, location, employment practices, and inmate breakdown by gender and responsibilities.

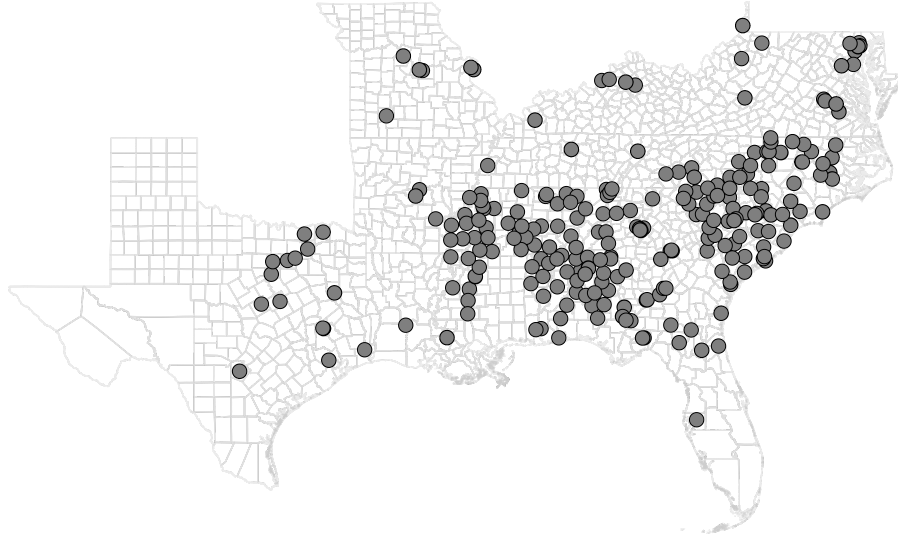
I digitized seven Bureau of Labor reports from 1886, 1895, 1905, 1914, 1923, 1932, and 1940.¹⁵ Notably, my paper is the first to digitize this dataset comprehensively. I first geographically map all prisons and labor camps from the data.¹⁶ Then, I link all facilities across the years to their 1880 counties using their geographic coordinates. In counties with multiple prisons, I aggregated the convict numbers. The data lack specific industry codes but do include the types of goods produced. I attribute two-digit Standard Industrial Classification

¹⁴Ayers (1984) reports examples of strikes against convict labor. In one instance, convict labor united miners with their community and miners from other areas. In Anderson County, Tennessee, people from all professions supported miners against the coal company and the state.

¹⁵Post the 1941 convict labor abolition, the Bureau stopped collecting data. Despite collecting data from these seven reports, I focus my analysis on 1886-1905 due to the outcome variable data constraints.

¹⁶I initially identified prisons using historical documents for current/historical addresses and then sourced GPS coordinates from Google Maps.

Figure 1. Geographic Distribution of Convict Labor Prisons



Note: The figure is generated using data from the Bureau of Labor reports spanning the years 1886, 1895, 1905, 1914, 1923, 1932, and 1940. Each dot in the figure represents a prison in this time frame.

(SIC) codes to each item based on the 1987 SIC System, then aggregate the values to generate industry-level output.¹⁷ Convict labor was used across various industries, significantly focusing on manufacturing.¹⁸

The analysis concentrates on the Southern states where convict labor was most prevalent following the Civil War.¹⁹ Figure 1 shows the geographic distribution of convict labor prisons in the Southern U.S. The data encompass 266 correctional facilities employing convict labor and details nearly 700,000 employed convicts for the available years.²⁰ I analyze the locations of incarceration, noting that convicts were utilized as laborers in these areas rather than in the places of their conviction. Muller (2018) shows that convicts were often incarcerated far from where they were convicted (see the Online Appendix Figure D4). He also finds that industrial employers, rather than agricultural ones, typically used convict labor.

3.2 KOL Assemblies, Strikes, and Membership

The outcome variables of interest that I utilize to measure union formation are (i) the number of Knights of Labor assemblies, (ii) union membership, and (iii) the number of strikes. I use

¹⁷All dollar amounts are normalized to 1880 dollars.

¹⁸The Online Appendix Figure D2 details industry-wise convict labor totals, 1886-1940.

¹⁹The Southern states include Alabama, Arkansas, Florida, Georgia, Kentucky, Louisiana, Maryland, Mississippi, Missouri, North Carolina, South Carolina, Tennessee, Texas, Virginia, and West Virginia.

²⁰The Online Appendix Figure D3 shows average convict counts per prison, 1886-1940.

data on the Knights of Labor assemblies from 1869-1919 from [Garlock \(1982\)](#) and [Garlock \(1984\)](#). [Garlock \(1982, 1984\)](#) sourced information on the Knights of Labor’s local assemblies from the Journal of United Labor and General Assembly proceedings. These data reveal the location, operation years, and some membership stats.²¹ I geocoded assembly locations to 1880 counties and tallied the assemblies per county annually.²² The Online Appendix Figure D5 displays the local Knights of Labor assemblies’ spatial distribution in the southern United States. The data cover more than 6,600 assemblies across 472 counties.

I utilize strike data from the *Third Annual Report of the Commissioner of Labor* (U.S. Bureau of Labor, 1888) and the *Tenth Annual Report of the Commissioner of Labor* (U.S. Bureau of Labor, 1896). The *Third Report*, a second nationwide strike survey, covers 1881-1886, while the *Tenth Report* spans 1886-1894. The Bureau of Labor sourced these data from trade journals, newspapers, and direct on-site visits, interacting with managers, workers, and union officers. This rich dataset details each strike’s location, industry, dates, causes, outcomes, union involvement, and worker impact—like occupation, wages, and affected worker numbers. It also provides establishment information, such as size, wages, working hours, closures, and financial losses. The report further breaks down pre- and post-conflict employment, wage, and work-hour data by gender.²³ I geocoded the strike locations by their 1880 counties and tallied the annual cumulative strikes per county. The data, merged with prison information based on report year and county, covers approximately 1,400 strikes across 172 counties in the southern United States from 1881 to 1894.²⁴

3.3 Other Data Sources

The 2020 U.S. labor union membership data comes from SimplyAnalytics, derived from the MRI-Simmons LOCAL consumer survey.²⁵ This survey assesses American consumer behaviors, including labor union memberships. MRI-Simmons then refines the data using geo-behavioral modeling to estimate union memberships by county. I merge this with their respective 1880 counties using an 1880-2010 FIPS code crosswalk ([Eckert et al., 2020](#)) and combine it with prison data based on 1880 county boundaries.

The control variables are obtained from the decennial Census records and the Historical, Demographic, Economic, and Social Data ([Haines et al., 2005](#)). The county-level data on

²¹Local chapters regularly reported membership, summarized annually ([Kaufman, 2001](#)).

²²Data gaps on assemblies post-1919 and missing convict employment data from 1914 restrict my analysis to Bureau of Labor reports from 1886-1905.

²³The Tenth Report lacks wage data and differs in observation unit from the Third. Later reports only offer aggregate data, which I omit.

²⁴See the Online Appendix Figure D6 for the geographical distribution of these strikes.

²⁵I am utilizing 2020 data due to the absence of county-level union data from the BLS.

manufacturing wage growth, employment share, output, profits, and the agricultural sector come from the ICPSR 2896 (Haines et al., 2005). For calculating migration outcomes, I utilize the linked full-count census (Helgertz et al., 2023; Ruggles et al., 2021). To quantify the shift in labor distribution, I use the linked full-count census data, focusing on the occupation variable (Helgertz et al., 2023; Ruggles et al., 2021). The data for the county-level vote shares in presidential elections are from ICPSR 08611 (Clubb et al., 1987). The Supplementary Appendix Section J details the remaining data and the construction of variables.

3.4 Stylized Facts

The implementation of convict labor intensified the competition within the local labor market, which could potentially have a negative impact on the overall demand for industry labor. Hiring out of convict labor was a privilege exclusive to firms with political connections (Gildemeister, 1977; McKelvey, 1934). These connected firms gained a competitive edge by exploiting the cost-efficient convict labor, which was cheaper than free labor as there was no obligation to pay inmates. This competition often forced regular firms to lower wages (U.S. Bureau of Labor, 1925). As a result, the equilibrium wages for labor across all firms have experienced a downturn, as formalized in Section 4.

Later in Section 5.4.3, I show that the introduction of convict labor did not significantly impact local labor supply, as migration trends remained unaffected, neither attracting nor driving away residents.²⁶ However, with the introduction of convict labor, there was a decline in labor demand, while supply remained unchanged. Consequently, this dynamic applied a downward force on the per-worker local wage.

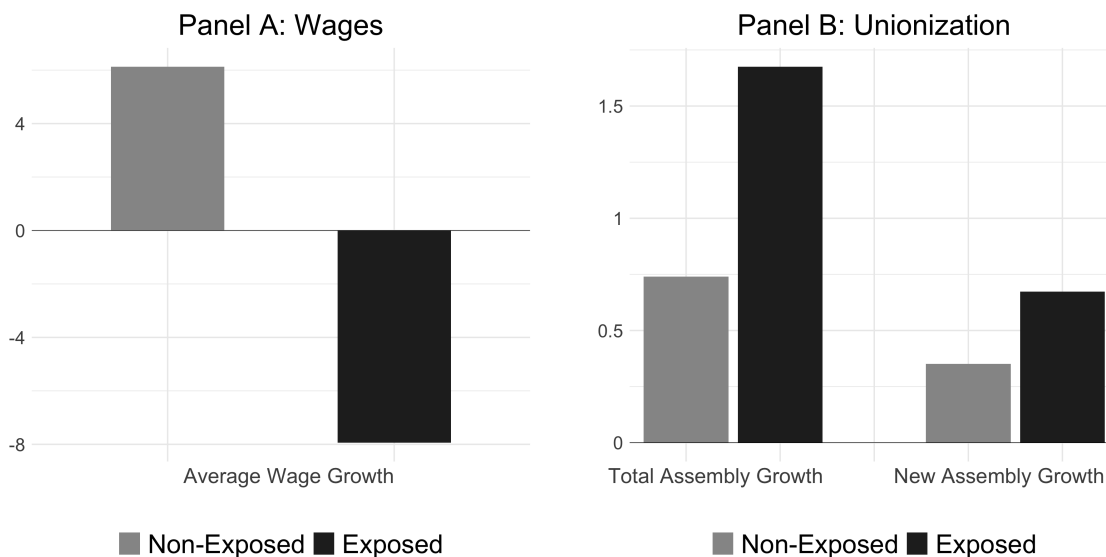
I start by examining the link between convict labor and local economies. Upon the introduction of convict labor, I observe a decline in the average wage growth, focusing on the southern parts of the United States that heavily relied on this type of labor. This change is paralleled by a rise in the formation of unions within these counties, a sign of escalating labor unrest.

3.4.1 Wages and Union Formation

The primary catalyst driving the research outlined in this paper is the observation that the Southern United States experienced a continuous decrease in labor demand following the implementation of convict labor laws. This systemic shift reverberated throughout the economy, leading to depressed average wage growth. Figure 2 demonstrates this interplay,

²⁶Online Appendix Table D17 confirms that migration trends remained stable, unaffected by convict labor.

Figure 2. Post-Legislation Growth: Average Wages, Total and New Knights of Labor Assemblies



Note: The figure shows growth in average wages, and total and new Knights of Labor Assemblies post-convict labor introduction (1880), compared to their level before slavery’s abolishment (1860). Panel A, using data from ICPSR 2896 (Haines et al., 2005), shows average wage (Bars 1-2) growth post-convict labor legislation in Exposed and Non-Exposed counties. Panel B, using data from Garlock (1982, 1984), presents growth rates of total (Bars 1-2) and new (Bars 3-4) assemblies post vs. pre-convict-labor in Exposed and Non-Exposed counties of the Southern U.S.

mapping the cross-sectional expansion of convict labor alongside trends in average wage growth.

Panel A of Figure 2 illustrates the trends in average wage growth in the Southern United States manufacturing industries. The figure focuses on two key regions: the Exposed South (in black), where industries faced competition from convict labor, and the Non-Exposed South (in gray), unaffected by such competition. After the introduction of convict labor, the Exposed South experienced about 10% decline in average manufacturing wage growth from its pre-legislation levels. In contrast, the Non-Exposed South saw over a 5% rise. This trend hints at a correlation between diminished wage growth and the onset of convict labor in exposed counties.

Collectively, these trends show that the introduction of convict labor paralleled a marked shift in the local labor market dynamics of Southern United States counties affected by this competitive labor market shift.

Next, I depict the patterns related to labor discord and the progression of unionization in the United States. Panel B of Figure 2 shows the escalation in the quantity of total and newly formed Knights of Labor assemblies (per 100,000 of the population) mapped across the era of convict labor implementation. It distinguishes between the Exposed South, where industries contended with convict labor, and the Non-Exposed South, which remained

unaffected. Post-legislation, the Exposed South saw over double the growth in total KOL assemblies and new assemblies compared to Non-Exposed counties. This indicates that the introduction of convict labor laws paralleled a surge in union formation, particularly in areas of the South competing with convict labor.

During this period, labor unions, especially the Knights of Labor, played a pivotal role in orchestrating strikes, thereby underlining the intricate connection between these organizations and collective labor actions. As the most prominent union of the time, the Knights helmed most of these protests, emphasizing their central role in mobilizing labor forces (Bittarello, 2019).

4 A Model of Convict Labor and Unions

To formalize the effect of convict labor on labor market outcomes and unionization, I introduce a novel general equilibrium framework capturing the labor market impacts of convict labor. I construct a model in which the representative firm utilizes two types of labor: convict and free labor. The model reveals that an increase in the convict labor capacity leads to lower wages for free laborers, and an increase in union and strike activities.

This model’s broad framework is robust to the free workers’ endogenous labor supply (see the Supplementary Appendix Section B). In the model’s extension with endogenous labor supply, a wage reduction decreases free labor supply. This indicates that convict labor lowers both the equilibrium wage rate and work hours of free labor.²⁷

In the Supplementary Appendix Section E, I expand the model to a two-firm framework, where only *elite firms* use convict labor.²⁸ This approach, informed by Census data showing multiple firms in a county, highlights labor reallocation and demonstrates benefits for firms without access to convict labor. The model accurately represents both types of firms within a county. Additionally, when the elite firm component is examined in isolation, it yields predictions similar to those of a representative firm. This model’s broad framework is robust to general firm-type distribution for elite and regular firms (see the Supplementary Appendix Section I).

²⁷Refer to the Online Appendix Section A for detailed proofs about the model section.

²⁸An elite firm is one with political connections that grant access to low-cost convict labor. To ensure the model accurately reflects historical contexts, it incorporates elite firms, defined by political ties rather than size or efficiency, and regular firms. This inclusion hints at the potential for elite firms to use their monopsony power, derived from political connections, to influence labor market dynamics and potentially depress wages for free workers, a factor separate from the use of convict labor. However, the observation of declining wages for free labor and increased unionization following the introduction of convict labor suggests that these elite firms had not exercised their monopsony power before the use of convict labor.

4.1 Convict and Free Labor

This paper examines an era in the Southern U.S. when coercive institutions led to Black individuals constituting most convict labor while white individuals dominated free labor. These coercive laws and institutions allowed imprisoned Black individuals to be exploited as a source of cheap labor. Consequently, these laws segregated the two labor groups and prohibited unionization among convict laborers (Hiller, 1914).²⁹

During the period in question, convict labor is a limited resource—there are no convicts in prisons who are not contributing labor (U.S. Bureau of Labor, 1886). This introduces a stringent capacity constraint, and any exogenous increase in prison capacity may significantly impact the wages of free labor and stimulate union formation, among other effects. The model’s introduction of this capacity constraint represents a novel approach to understanding convict labor employment during this period. In terms of compensation, convict laborers do not receive salary payments, denoted as $w_c = 0$ in the model. On the contrary, free labor receives a wage denoted by $w > 0$.

The model posits that convict and free laborers are perfect substitutes, an assumption supported by Bureau of Labor reports (U.S. Bureau of Labor, 1886, 1925). These reports indicate that employers could use convict labor in any role generally occupied by free labor.³⁰ Nevertheless, there may be concerns about specific tasks where convict labor does not perfectly substitute free labor. To address this, in the Supplementary Appendix Section H, I introduce a constant elasticity of substitution model, theoretically and empirically exploring this case. The findings from this model qualitatively align with the model’s predictions presented here.

4.2 Representative Firm

The representative firm employs both types of workers – convict and free labor. The representative firm’s production function is:

$$y = z(\theta l_c + l_f)^\alpha k^\beta, \tag{1}$$

²⁹As explained in Section 2.2, restrictive laws and limited opportunities hindered Black labor mobility, with many industries, like manufacturing, remaining predominantly white. Thus, the model omits a free Black labor force in the main industry, acknowledging their limited representation even a century post-emancipation while allowing for their employment in the alternative industry.

³⁰The model assumes no distinction between convict and free labor in the labor market, supported by the era’s dominance of unskilled workers and Bureau of Labor reports indicating employers could interchangeably use convict labor (U.S. Bureau of Labor, 1886, 1925). While potential differences in monitoring costs to prevent shirking might suggest some market segmentation, the lack of data prevents distinguishing these costs, leading to the assumption that such differences are minimal.

where y is the output of the representative firm and $\alpha + \beta < 1$.³¹ The representative firm hires l_c units of convict labor and l_f units of free labor and acquires capital in the measure of k . The total factor productivity of the firm is denoted by z . The model permits the potential for productivity disparity between free and convict laborers through the θ multiplier preceding l_c . This firm independently determines the quantity of each labor type to employ and the capital to procure. The number of convict laborers hired, l_c , is subject to a capacity constraint $l_c \leq \bar{l}$. As the wage rate for convict labor is zero, the firm's convict labor capacity constraint always binds, i.e., $l_c = \bar{l}$.³²

Wages of free labor, w , will be determined endogenously in general equilibrium. Rental rate $r > 0$ of capital is exogenous, with full depreciation of capital.³³ Solving the maximization problem of the representative firm, demands for l_f and k become:

$$l_f = \left(\frac{w_f}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} - \theta \bar{l} \quad (2)$$

$$k = \left(\frac{\alpha}{w_f}\right)^{\frac{\alpha}{1-\alpha-\beta}} \left(\frac{r}{\beta}\right)^{\frac{\alpha-1}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} \quad (3)$$

4.3 Households and General Equilibrium Wages

There are N households in the economy whose total labor supply equals \bar{L} , and they inelastically supply their labor services for the representative firm.³⁴ The total demand for free labor in the economy is expressed as l_f . The general equilibrium of the economy is determined by:

$$\bar{L} = \left(\frac{w}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} - \theta \bar{l}, \quad (4)$$

³¹Decreasing returns to scale are employed in this model to reflect historical accuracy and empirical observations, enabling an analysis of profit distribution and the broader economic impact of convict labor on the representative firm.

³²I assume $w_c = 0$ without loss of generality. If $w_c = c$, the convict labor capacity constraint of the firm still binds as long as c is sufficiently small. In this case, the total labor cost of the firm would become $c\bar{l}$, and the optimization program would remain unaltered since $c\bar{l}$ is a constant lump sum. The Bureau of Labor reports (U.S. Bureau of Labor, 1886, 1925) implies that the unit convict labor costs (w_c) were sufficiently small, such that convict labor supply was employed in production to its full extent.

³³The model's constant interest rate assumption is backed by literature supporting a unified capital market in the Southern US (Rockoff, 1990; Bodenhorn, 1992; Rosenbloom, 1996). Additionally, incorporating banking density into the analysis leaves the results unchanged (see the Supp. Appendix Table K22).

³⁴In the Online Appendix Section B, I relax this assumption to permit endogenous labor supply. A wage reduction due to an increase in the convict labor use decreases the free labor supply, lowering equilibrium wage and work hours.

which characterizes the equilibrium wage:

$$w = \alpha \left(\frac{\left(\frac{\beta}{r}\right)^{\frac{1-\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}}}{\bar{L} + \theta \bar{l}} \right)^{\frac{1-\alpha-\beta}{1-\beta}}. \quad (5)$$

Equation (5) shows that an increase in the capacity constraint for convict labor results in a corresponding decrease in the wages of free labor ($\frac{\partial w}{\partial l} < 0$). Furthermore, when \bar{l} is sufficiently large, the productivity of convict labor and its capacity become strategic complements. As a result, enhancing convict labor productivity further depresses equilibrium wage level ($\frac{\partial^2 w}{\partial l \partial \theta} > 0$). This leads to the following proposition:

Proposition 1: *Increasing the capacity constraint of convict labor reduces the equilibrium wage rate. Furthermore, if \bar{l} exceeds a threshold \bar{l}^* , an increase in θ further lowers the equilibrium wage rate.*

4.4 Equilibrium Profit of the Firm

As the capacity for convict labor expands and equilibrium wages decrease, it is reasonable to anticipate a corresponding impact on firms' profits. The profits of the representative firm are given by:

$$\Pi = (\bar{L} + \theta \bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{1}{1-\beta}} z^{\frac{1}{1-\beta}} \left(\frac{r}{\beta}(1 - \alpha - \beta) + \alpha \frac{r}{\beta} \frac{\theta \bar{l}}{\bar{L} + \theta \bar{l}}\right). \quad (6)$$

The capacity constraint of convict labor increases the equilibrium profits of the representative firm, i.e., $\frac{\partial \Pi}{\partial l} > 0$, because the firm can access convict labor input at no cost. This gives rise to the ensuing proposition:

Proposition 2: *An increase in the capacity for convict labor increases the profits of the representative firm.*

This rise in firm profits, captured by my model, formalizes the argument why convict labor found such a high demand and support by firm owners in counties where convict labor became available.

4.5 Union Membership and Strikes

Suppose N households with a total inelastic labor supply of \bar{L} collectively invest J to establish a union.³⁵ This union offers free laborers of the representative firm the opportunity to initiate a strike at the beginning of a production cycle against the hiring of convict laborers. It compels the firm to curtail convict labor employment to zero with some probability. Should the strike succeed, the value or benefit (Ω) accrued by the employee is as follows. I now write $w(l_c)$ to emphasize the dependence of the equilibrium wage on the capacity constraint:

$$\Omega = (w(l_c = 0) - w(l_c = \bar{l})) \bar{L}, \quad (7)$$

and this value determines the free laborer's willingness to coordinate and pay the joint unionization cost J . A clear insight from equation (7) suggests that as \bar{l} grows, the likelihood of joining the union also rises. This is because the wage gap, $w(l_c = 0) - w(l_c = \bar{l})$ is increasing in \bar{l} , a conclusion that follows the findings from the earlier section.

Free labor's willingness to form a union, requiring the payment of cost J , should align with equilibrium beliefs that strikes stand a chance of succeeding. Assume that the owner of the representative firm can prevent a strike with a probability of $1 - p$, and with a probability of p , the strike succeeds. In such an instance, free labor fulfills its demands, and the employment of convict labor in the representative firm is completely eliminated. However, if the strike proves unsuccessful, the laborers suffer a lump sum utility loss, δ . This suggests that free labor would opt to join the union and initiate strikes if and only if:

$$J < p\Omega - (1 - p)\delta = p(w(l_c = 0) - w(l_c = \bar{l})) \bar{L} - (1 - p)\delta. \quad (8)$$

Considering the right-hand side of equation (8) rises with \bar{l} , we can infer that a greater capacity for convict labor directly motivates increased unionization and strike activities. This leads to the following proposition:

Proposition 3: *An increase in the capacity for convict labor results in increased unionization and strike actions.*

The growth in the number of union assemblies is embodied in the probability of unionization within this model. Notably, the effects on the extensive and intensive margins move

³⁵The model frames the cost of unionization as a macro-level phenomenon. These macro determinants include the economic conditions that reduce this cost, such as employment rates, wage levels, and industry characteristics. Empirical evidence supports the impact of such macro determinants on union dynamics, with research linking union growth to economic cycles, inflation, and wage trends (Ashenfelter and Pencavel, 1969; Bain and Elsheikh, 1976; Stepina and Fiorito, 1986).

in tandem. These dual channels are currently encapsulated within the J term. The model reveals that the presence of convict labor exerts a downward pressure on the wage rates of free labor while concurrently amplifying their unionization and strike actions.

Furthermore, when \bar{l} reaches a significant magnitude, the productivity of convict labor and its capacity begin to function as strategic complements. This implies that a boost in convict productivity leads to a further drop in the equilibrium wage. Consequently, the right-hand side of equation (8) amplifies with increased convict labor productivity, thereby strengthening the impact of convict labor capacity on unionization.³⁶

This model guides the empirical analysis and illustrates the core mechanism underlying the results: the escalated labor market competition free labor faces due to convict labor.

5 Effect of Convict Labor on Union Formation

To quantify the impact of introducing convict labor on the process of union formation, I employ the following regression equation:

$$y_{cst} = \beta Convict_{cst} + \mathbf{X}_{cs,1860}\gamma + \delta_{st} + \varepsilon_{cst} \quad (9)$$

where, y_{cst} denotes union formation metrics: Knights of Labor assemblies, strikes, and union members for county c and state s in year t . $Convict_{cst}$ denotes prisoners employed per hundred thousand residents for county c and state s in year t . Given the skewed distribution of $Convict_{cst}$, I use an inverse hyperbolic sine transformation. $\mathbf{X}_{cs,1860}$ are pre-determined control variables for county c and state s measured in 1860.

More specifically, I consider several factors that could be linked to county-level convict labor utilization. Economically, I control for variables like population density, enslaved population share, average farm output value per acre, and improved farmland acres, as affluent or populated counties might differ in convict labor use. For trade aspects, I control for coastal proximity, elevation, railway access, water body navigability, and distance to the nearest large urban center. I also include in the Democratic Party’s vote share to reflect regional perspectives on race and convict labor. I employ state \times year fixed effects, δ_{st} , to partial out time-invariant and aggregate trend variables. Standard errors are clustered at the county level. The key parameter, β , estimates the link between the share of convict labor and union formation metrics.

³⁶The likelihood of striking increases with unionization, based on the assumption that workers will strike once unionized, making union membership a precursor to striking. Furthermore, Kennan (1986) views strikes as a method for workers to secure higher wages from profitable employers. The use of convict labor escalates the tendency to strike by expanding the wage gap and increasing firm profits, thus enlarging the available benefits for workers.

5.1 Identification

The estimation, as delineated in equation (9), alongside the specified set of controls and fixed effects, potentially accounts for a significant proportion of conceivable endogeneity sources. Yet, we cannot fully affirm causality if convict labor correlates with unobserved variables, which could concurrently exert influence over union formation measures.

An endogenous independent variable, correlated with the error term of the regression equation, can introduce bias into OLS estimates. For example, during this period in the Southern U.S., areas with a higher prevalence of convict labor had a higher share of formerly enslaved populations. These areas displayed a heightened capacity and inclination for oppression, effectively curbing labor organization. Policies, such as employing convicts as strikebreakers, were strategically used in these areas to hinder labor initiatives (Friedman, 2000). In this case, the underlying economic context suggests that the correlation between the outcomes of union formation and convict labor understates the true effect (Jiang, 2017).

I employ an instrumental variable approach to address this potential pitfall, capitalizing on the variation within a county’s interaction with pre-convict-labor prisons and their respective capacities. This approach finds a plausibly exogenous cross-sectional variation from the prisons that were built before the enactment of the convict labor legislation. The data required for this analysis, which includes information on the locations of pre-convict labor prisons, the years of their establishment, and prisoner populations, were collected from references like [North American Review \(1866\)](#), [Prison Association of New York \(1870\)](#), [Wines \(1871\)](#), and 1870 U.S. Population Census.³⁷

Subsequently, I utilize the locations and capacities of pre-convict-labor prisons as instruments for the number of prisoners employed per hundred thousand inhabitants in a county. The equation for the first stage is as follows:

$$Convict_{cst} = \bar{\beta}IV_{cs} + \mathbf{X}_{cs,1860}\bar{\gamma} + \bar{\delta}_{st} + \epsilon_{cst} \quad (10)$$

where IV_{cs} is either a binary indicator for exposure to a pre-convict-labor prison or the capacity of the pre-convict-labor prison in county c and state s .

The estimation equation for the second stage takes the following form:

$$y_{cst} = \beta\widehat{Convict}_{cst} + \mathbf{X}_{cs,1860}\gamma + \delta_{st} + \epsilon_{cst} \quad (11)$$

where the instruments measure the exposure of each county to a pre-convict-labor prison or the capacity of the pre-convict-labor prison.

³⁷I assign geographic coordinates to these prison locations and merge with their corresponding counties. The Online Appendix Figure D7 shows the geographic distribution of these prisons for the southern U.S.

The instrumental variable approach assumes that a county’s exposure to pre-convict-labor prisons and the number of convicts in those prisons are valid predictors for the number of convicts after 1886. The validity of this assumption is supported by the Online Appendix Table D4, which demonstrates a positive correlation between the presence of pre-convict-labor prisons and their capacities and higher levels of prisoner employment (per 100,000 inhabitants) after the introduction of convict labor laws.³⁸ The robustness of the results to the addition of controls confirms the strength and relevance of the instrument. Throughout the columns, the coefficients display the anticipated signs and are statistically significant. The coefficients presented in columns (3) and (6) with the complete set of control variables and fixed effects demonstrate that counties with exposure to pre-convict-labor prisons (and capacities) exhibit higher prisoner employment. The Kleibergen-Paap F-statistics also surpass the commonly accepted threshold for weak instruments (Kleibergen and Paap, 2006).

The exclusion restriction requires that the impact of pre-convict-labor prisons (and their capacities) on union formation can only be mediated by their effect on the number of convict laborers after the implementation of convict laws. Therefore, the identification assumption relies on the absence of potentially unobserved variables that are correlated with pre-convict-labor prisons (and their capacities) and the formation of labor unions.

The estimation results would be biased if pre-convict-labor prisons were strategically built in locations with a higher likelihood of future union formation. Historians list several factors that were considered when selecting prison locations before 1870, including proximity to urban centers, access to railroads or navigable rivers, and proximity of materials for construction of a prison (Lewis, 1922; McKelvey, 1936). To account for these factors, I control for population density, proximity to the coast, elevation, access to railway transportation and navigable water bodies, and the distance to the nearest urban center. The Supplementary Appendix Tables K7 to K10 show the estimation results where I gradually add these control variables. The outcome remains unchanged even with the gradual inclusion of controls, and the effect sizes across all columns are similar in magnitude and statistically significant.

Moreover, I regress the instruments on an extended set of population, economic, trade and commerce, geographic and climate, and socio-political control variables in the Online Appendix Table D7. Apart from access to railways and population density, I find no statistically significant correlations between pre-existing characteristics and pre-convict-labor prisons’ presence and capacities (I control for these variables). This suggests that the location of these prisons was not systematically related to observable county characteristics,

³⁸The Online Appendix Table D5 and Table D6 show the reduced form regressions of the main outcomes on each instrument: pre-convict labor prison locations and capacities. Furthermore, the Supp. Appendix Tables K5 and K6 show positive correlations between pre-convict-labor prisons (and capacities) and increases in the number of prisons and convicts employed following the introduction of convict labor.

which makes the selection on unobservables less likely.

The use of pre-convict labor prison locations and capacities as an instrumental variable is economically rational due to their exogenous selection conditional on population density and railway access, independent of future convict labor laws or slavery’s abolition. These prisons, established without foresight into their eventual role in profit-driven manufacturing or convict labor employment, predate the 13th Amendment and related legislation. Furthermore, [Ayers \(1984\)](#) states that Southerners had a unique political anxiety about building prisons before the Civil War, using them only for the most hardened criminals, who were overwhelmingly white, with no slaves imprisoned. This changed after the abolition of slavery and the introduction of convict labor, leading to a predominantly Black prison population ([Ayers, 1984](#); [Muller, 2018](#)). Therefore, the pre-convict labor prison locations, chosen without regard to future labor market dynamics, offer an exogenous setting for assessing convict labor’s future economic impact.

To address concerns that the findings could be attributed to selection bias due to unobserved variables, I adopt a method suggested by [Oster \(2019\)](#) to strengthen the robustness of the results. [Oster \(2019\)](#) emphasizes the significance of coefficient stability and movements in R^2 . Using her recommended bounding value of 1.3 times increase in R^2 , beyond which the results can be considered robust, I apply this criterion to the preferred specifications in the main tables. The analysis indicates that to explain away the findings, there needs to be a substantial degree of selection bias on both observed and unobserved variables, with a relative degree of selection δ between 2.79 and 34.39. The results indicate that the estimated selection on unobservables is substantially larger than the suggested upper bound of $\delta = 1$ in empirical studies. This finding indicates that unobserved selection needs to be excessively large to explain the results. The analysis supports the relevance assumption and the exclusion restriction, reinforcing the instrument’s validity.

In Section 5.3, I execute further checks to mitigate any lingering concerns regarding factors that might have influenced the locations of prisons before the introduction of convict labor. Further additional analysis aims to demonstrate that these pre-convict labor prisons did not impact local labor market outcomes through channels unrelated to the application of convict labor.

5.2 Results

5.2.1 Unions, Strikes, and Membership

Table 1 exhibits the outcomes of estimating equation (11), with each column representing a distinct regression with a different set of control variables. The table demonstrates the 2SLS

Table 1: Effect on KOL Assemblies and Strikes 1886–1905

	<i>Dependent variable:</i>							
	Number of KOL Assemblies				Number of Strikes			
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Convict Labor per 100,000 (IHS)	2.273** (0.927)	2.176** (0.905)	1.154*** (0.417)	1.166*** (0.418)	0.968** (0.447)	0.936** (0.437)	0.449** (0.202)	0.450** (0.201)
KP F-stat	128.98	126.88	129.3	126.59	128.98	126.88	129.3	126.59
Panel B: Early Prison Capacities IV								
Convict Labor per 100,000 (IHS)	2.282** (0.980)	2.187** (0.962)	1.114*** (0.430)	1.126*** (0.431)	1.021** (0.492)	0.992** (0.483)	0.487** (0.234)	0.488** (0.234)
KP F-stat	181.46	177.92	191.08	186.55	181.46	177.92	191.08	186.55
State \times Year FE	\times	\times	\times	\times	\times	\times	\times	\times
Geographic Controls		\times	\times	\times		\times	\times	\times
Economic Controls			\times	\times			\times	\times
Socio-political Controls				\times				\times
Mean Outcome	1.23	1.23	1.23	1.23	0.28	0.28	0.28	0.28
Clusters	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906

Note: Equation (11) displays regression results with coefficients and standard errors (in parentheses) for each column. The unit of observation is the 1880 county. Outcome variables encompass local KOL assemblies and strikes, with the convict labor presence, measured as convicts per 100,000 inhabitants (IHS), serving as the independent variable. Economic controls include population density, the share of enslaved individuals, average farm output value per acre, and improved farmland acreage. Geographic controls capture distance to the coast, elevation, railway and water body access, and proximity to the nearest urban center. The Democratic Party’s vote share acts as the socio-political control. Early prison locations (Panel A) and capacities (IHS) (Panel B) serve as instrumental variables. The model incorporates state \times year fixed effects and clusters standard errors at the county level.

estimations of convict labor per 100,000 inhabitants, transformed via inverse hyperbolic sine, regressed on the number of local KOL assemblies and strikes in a given county, respectively. Panel A displays estimations instrumenting convict labor with pre-convict-labor prison locations, while Panel B utilizes pre-convict-labor prison capacities as instruments. Columns (1) and (5) present the specifications with only state-by-year fixed effects. Controls for pre-existing geography, economic conditions, and socio-political variables are incrementally added across columns (2) and (6), (3) and (7), and (4) and (8), respectively, for both outcome variables. Both panels show similar and statistically significant effect sizes across columns (1)–(4) and (5)–(8).

The more conservative and preferred specifications in columns (4) and (8) incorporate a complete set of controls and state \times year fixed effects. From Panel A, I find that a 10% rise in convict labor share results in about a 9.5% surge in local KOL assemblies and a 16% increase in strikes. Likewise, Panel B results reveal a similar pattern; a 10% growth in convict labor triggers a 9.2% increase in local KOL assemblies and a 17% rise in strikes. Notably, the instruments yield remarkably similar estimates, despite exploiting different margins (extensive vs. intensive).³⁹

³⁹As Tables 1 and 2 show, the IV coefficients are magnified, ranging from 1.5 to 4.5 times the size of

Table 2: Effect on Membership 1886–1905 and 2020

	<i>Dependent variable:</i>							
	Early Membership (IHS)				Membership 2020 (IHS)			
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Convict Labor per 100,000 (IHS)	0.243*** (0.061)	0.228*** (0.060)	0.185*** (0.049)	0.187*** (0.049)	0.323*** (0.044)	0.254*** (0.039)	0.214*** (0.038)	0.225*** (0.040)
KP F-stat	128.98	126.88	129.3	126.59	128.82	126.84	129.21	126.29
Panel B: Early Prison Capacities IV								
Convict Labor per 100,000 (IHS)	0.238*** (0.061)	0.224*** (0.059)	0.179*** (0.048)	0.181*** (0.048)	0.318*** (0.041)	0.250*** (0.037)	0.211*** (0.036)	0.222*** (0.038)
KP F-stat	181.46	177.92	191.08	186.55	181.2	177.85	190.87	186.04
State × Year FE	×	×	×	×	×	×	×	×
Geographic Controls		×	×	×		×	×	×
Economic Controls			×	×			×	×
Socio-political Controls				×				×
Mean Outcome	20.83	20.83	20.83	20.83	3,542	3,542	3,542	3,542
Clusters	1,302	1,302	1,302	1,302	1,296	1,296	1,296	1,296
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906

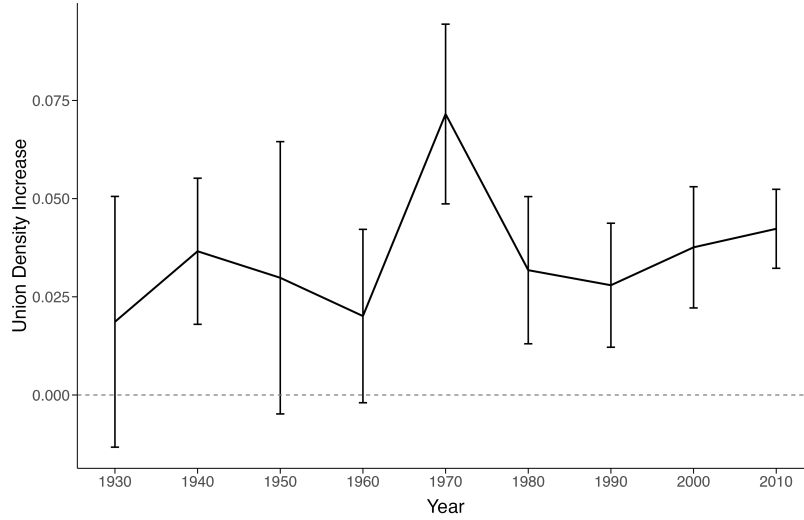
Note: Equation (11) displays regression results with coefficients and standard errors (in parentheses) for each column. The unit of observation is the 1880 county. Outcome variables encompass KOL membership in the 19th century and in 2020, with the convict labor presence, measured as convicts per 100,000 inhabitants (IHS), serving as the independent variable. Economic controls include population density, the share of enslaved individuals, average farm output value per acre, and improved farmland acreage. Geographic controls capture distance to the coast, elevation, railway and water body access, and proximity to the nearest urban center. The Democratic Party’s vote share acts as the socio-political control. Early prison locations (Panel A) and capacities (IHS) (Panel B) serve as instrumental variables. The model incorporates state × year fixed effects and clusters standard errors at the county level.

In the Online Appendix Table D8, I demonstrate that the introduction of convict labor not only increases the count of KOL assemblies but also extends their longevity and catalyzes new assembly initiations. The layout of this table mirrors Table 1, with each column signifying a separate regression analysis featuring distinct control variables. Columns (1)-(3) detail the assemblies’ average survival rate, while columns (4)-(6) record the expansion in new assembly establishments. The findings suggest that with every 10% rise in convict labor, there is a 3% increment in the average duration of an assembly’s survival and a 7% surge in the number of newly formed assemblies.

Finally, I explore whether firms strategically placed convict labor in union-active areas to suppress unionization after convict labor laws were enacted. If changes in union formation directly influence the number of convict laborers between report years, it suggests this strategy. However, analysis in the Supp. Appendix Table K11, which examines the link between

the OLS coefficients (the Online Appendix Tables D1 and D2). This amplification is attributed to the potential downward bias due to areas with a higher share of enslaved having a higher capacity for oppression. Supporting this, the difference between IV and OLS coefficients for manufacturing outcomes is slight (the Online Appendix Tables D3 and D9). Yet, potential structural factors, like classical measurement errors due to unseen industry displacement rate differences, could introduce a downward bias in OLS estimates, possibly explaining some observed data noise. Furthermore, the Supp. Appendix Tables K1 to K4 present various outcome variables, supplemented by analyses with transformed variables, alternate measurements, and Conley standard errors.

Figure 3. Effect on Union Membership at State Level 1930–2010



Note: The figure presents the coefficients of a regression analysis, where the dependent variable is the union density for each decade, and the independent variable is the state-level share of convict labor in 1886. This analysis is based on the following regression equation: $y_{sd} = \beta Convict_{s,1886} + \mathbf{X}_{s,1860}\gamma + \delta_d + \varepsilon_{sd}$, where y_{sd} denotes union density for state t in decade d . Union density data is obtained from Farber et al. (2021).

union formation and changes in convict labor, does not support such a strategic effect.

Table 2 displays the analysis results examining the impact of convict labor on 19th century and 2020 union membership. In Panel A, pre-convict-labor prison locations instrument convict labor, while Panel B utilizes pre-convict-labor prison capacities for the same. Columns (4) and (8) showcase the most comprehensive specifications incorporating a complete control set and state \times year fixed effects for both panels.

Panels A and B’s fourth column results suggest a 10% convict laborer increase prompts a 1.8% surge in 19th century union membership. Simultaneously, column (8) from both panels indicates that a 10% increase in convict laborers equates to 2.2% more union members, underlining convict labor’s enduring impact on current union membership, thus emphasizing its critical role in institutional development.

Table 2 provides evidence of the relationship between convict labor and union membership across the 19th century and in 2020. In Figure 3, I analyze data from Farber et al. (2021) to explore the association between state-level convict labor share and union density for each decade between 1930 and 2010. The regression coefficients reveal a positive association, indicating that the convict labor consistently influenced union density over the 20th century. This section confirms Section 4’s predictions: increased convict labor increases union activities; memberships, and strikes.

5.2.2 Wages, Output, and Profits

The model in Section 4 postulates that an increase in convict labor leads to decreased wages and a corresponding rise in profits. To investigate these propositions, I will now examine the effects of convict labor on industry conditions.

The influence of convict labor extends substantially across local economies, particularly impacting the working conditions of free labor in counties where convict labor is widespread. As shown in the Online Appendix Table D9, counties with convict labor presence show a decline in per-worker wage growth, reduced growth in employment share, and increased manufacturing output value and gross profits in their manufacturing industries.⁴⁰ Furthermore, the decrease in wages for free laborers and the increase in firms' output value and gross profits suggests that convict labor exacerbates income disparity between workers and the firm owners.⁴¹ On the other hand, I observe no statistically significant shifts in farming outcomes. This lack of substantial changes, detailed in the Online Appendix Table D10, is noteworthy for two main reasons: first, prisoners engaged in agriculture did not utilize pre-convict labor prison facilities. Second, this period aligns with a historical phase in the agricultural industry when labor unions were conspicuously absent (Friedman, 2000).

As the model predicts, wage suppression appears to be the key mechanism driving the relationship between convict labor and the increase in union formation. Convict labor puts downward pressure on wages, which then stimulates wage laborers to rally together, forming unions to counter this impact. Section 5.4.1 will explore this mechanism further.

Next, I explore the interplay between industries that directly compete with convict labor sourcing data from Hornbeck and Rotemberg (2021).⁴² In counties where industries compete with convict labor, especially in the nature of their produce, the local economic landscape also experiences a significant shift. This change includes a decline in the share of manufacturing employment (see the Supp. Appendix Table K13). Also, union activity changes significantly in areas where industries are in direct competition with convict labor. This competition stirs up union activity, leading to a noticeable increase in the formation of unions. Correspondingly, all aspects related to union formation witness growth in these counties (see the Supp. Appendix Table K14). A rise in labor unrest is also noticeable, evident in more

⁴⁰This decrease in employment share growth can be explained by the endogenous labor supply model in Section B. This model demonstrates that \bar{l} lowers the equilibrium wage rate of free labor, which subsequently reduces the equilibrium hours worked by free labor.

⁴¹This corroborates the findings of Alston and Ferrie (1993, 1999) and Althoff and Reichardt (2022) that oppressive institutions in the American South benefited the elite.

⁴²For these analyses, I use an analogous estimation strategy to equation (11); however, as I use the sum of pre-convict labor prisons and their capacities to instrument for the IHS transformation of the sum of the interaction of a continuous measure for both competing division and the number of convict laborers per 100,000 inhabitants and their sum.

frequent, protracted, and intense strikes (see the Supp. Appendix Table [K15](#)).

The productivity of convict labor, on the other hand, further spurs union formation. In areas where convict labor proves more productive, all indicators associated with union formation escalate (see the Supp. Appendix Table [K16](#)).

5.3 Robustness

First, I start by performing further sensitivity analyses to reinforce the validity of the identification strategy. I use a nearest-neighbor matching approach to address remaining concerns about factors influencing pre-convict labor prison locations. I identify control counties with similar observable characteristics to treated counties. The algorithm selects the closest control counties for each county with a pre-convict labor prison based on predetermined county-level characteristics. The Online Appendix Table [D11](#) displays estimation results from matching counties considering population density, distance to coast, elevation, railway and water body access, and proximity to the nearest urban center. Second, I randomize pre-convict-labor prison locations by estimating the likelihood of a county having such a prison using pre-defined control variables. I derive predicted probabilities and randomly assign counterfactual prisons based on them. I then compute counterfactual counties' prison capacities using the same variables and obtain predicted capacities. This procedure is iterated 1,000 times. The Online Appendix Figure [D8](#) presents t-statistic densities from placebo regressions, which, with low values, fail to refute the null hypothesis for all outcome variables.

An additional source of concern could be that pre-convict-labor prisons may affect local labor market outcomes through means other than the use of convict labor. To address this potential issue, I demonstrate that exposure to early prisons (and capacities) does not correlate with the change in several labor market covariates in an exposed county before the implementation of convict-labor laws (see the Online Appendix Table [D12](#)). Moreover, the Online Appendix Table [D13](#) shows no correlation between exposure to pre-convict labor prisons (and capacities) and measures of union formation before convict labor legislation, indicating no association with pre-existing trends.

To address potential cultural influences within the county that may lean towards unionization, I include additional control variables in the Supp. Appendix Tables [K17](#) and [K18](#). These controls account for the proportion of European, Italian, German, Scandinavian, Scot-Irish, and French immigrants, considering their European roots and potential inclination towards labor organization. Additionally, I factor in the share of West Indies immigrants to account for their history with more brutal slavery practices. These tables demonstrate that the results remain statistically significant even when factoring in immigrant demographics.

Next, to account for *rugged individualism*, a characteristic potentially linked to both elevated crime rates among whites (resulting in more early prisons) and their tendency to establish labor unions in opposition to a primarily black convict labor force post-Civil War, I sourced data from [Bazzi et al. \(2020\)](#). This involves controlling for whether a county was on the frontier and total frontier experience (the Supp. Appendix Tables [K19](#) and [K20](#)). The results remain substantially unchanged. To account for the impact of race-motivated unionization factors, I included racial terror lynching prevalence as a control, digitizing data from [Seguin and Rigby \(2019\)](#) on lynching events between 1877 and 1950. Aggregated at the county level to compute lynching density per 100,000 individuals up to the first Bureau Labor report year, this metric is incorporated as a control variable in the Supp. Appendix Table [K21](#), with results showing no significant change.

To examine the impact of regional capital market characteristics, I incorporate the number of banks per 100,000 inhabitants as an additional control variable for outcomes related to union formation, specifically, KOL assemblies, strikes, and membership rates in the 19th century and 2020.⁴³ The results, displayed in the Supplementary Appendix Table [K22](#), show that adding bank density as a control does not meaningfully change the coefficient estimates.

The monotonicity assumption assumes that the relationship between the instruments (pre-convict-labor prisons and their capacities) and the outcome (convict labor per 100,000) is such that the presence of a pre-convict-labor prison (or an increase in its capacity) should lead to a rise in the use of convict labor in a given county. This can be tested by examining the first-stage estimates of the instrumental variable regression. The monotonicity assumption is considered satisfied if the estimates are non-negative for any subsample. The results of these tests are presented in the Online Appendix Table [D14](#). I show that the first stage is satisfied in subsamples of the data.

Second, I conduct robustness checks to reaffirm the findings of Section [5.2](#). In the Supp. Appendix Tables [K23](#) to [K25](#), I present coefficients for various transformed outcome variables. I also evaluate the impacts of convict labor using alternative definitions, such as total convict laborers, number of prisons using them, and value of convict-made goods (the Supp. Appendix Tables [K26](#) to [K28](#)). I further analyze effects using standardized outcome variables in the Supp. Appendix Table [K29](#). Even with normalized outcomes, the effect persists, indicating it is not solely due to population size. [Ayers \(1984\)](#) further notes that urban areas objected to penitentiaries as much as rural areas, confirming that objections were not influenced by population size. Next, I address potential spatial correlation through

⁴³To build a proxy for capital markets, I digitized the national bank locations data from the Annual Report of the Comptroller (U.S. Office of the Comptroller, 1880). I digitized the locations of private banks from the Bankers' Directory of the U.S. ([Rand and McNally, 1880](#)). Then, I aggregate the count of national and private banks at the county level, computing the bank density per 100,000 individuals.

Conley standard errors (Conley, 1999) in the Supp. Appendix Table K30. Finally, to account for the spatial dependence across neighboring counties or counties within a state, I use two-dimensional clustering by both county and state in the Supp. Appendix Tables K31 and K32. All checks maintain the statistical significance of the results.

Up to this point, I have presented the results using one merging method, in which I linked the outcome variables to various measures of convict labor using report years from the Bureau of Labor reports. Yet, the findings remain robust across different merging techniques.⁴⁴ The Supp. Appendix Table K33 presents regression outcomes for the inverse hyperbolic sine transformation of convict labor per 100,000 residents. This focuses on several key metrics: local Knights of Labor assemblies, 19th-century strikes and members, and 2020 union memberships. Using control variables from Section 5 and incorporating state \times year fixed effects, the results align closely and significantly with Tables 1 and 2, reinforcing the consistency of the findings.

5.4 Mechanisms

5.4.1 Competition in the Labor Market

Why does a rise in convict labor lead to increased unionization? The main mechanism at play appears to be labor market competition. The rise in convict labor depresses wages, necessitating wage laborers to organize collectively to counteract this effect. Simultaneously, these organized efforts also aim to eliminate the use of convict labor in the economy.

To understand whether wage suppression is the driving mechanism, examining the counterfactual impacts of convict labor on wages in the absence of unions is necessary. In an attempt to achieve this, I divide the sample into two categories: counties with union presence and those without. The Online Appendix Table D15 shows the influence of convict labor on average manufacturing wage growth across the entire sample, a subset with KOL presence and a subset without union presence.

In counties without active unions, there is a statistically significant decline in average manufacturing wage growth following the increase in convict labor. However, the average manufacturing wage growth in counties with union presence does not show a statistically significant change. If anything, it tends to be in a positive direction. These results indicate a more pronounced effect of convict labor on wage growth decline when unions are absent. This provides suggestive evidence that the mechanism is via wage reduction and solidifies the primary mechanism outlined in the model in Section 4.

⁴⁴For a detailed breakdown of merging methods, see the Supp. Appendix Section N.

5.4.2 Employment in Crafts and Operating Occupations

The prevalence of convict labor inadvertently stimulates an increase in free labor employment within alternative occupations, such as skilled crafts or operating roles. This shift in labor distribution, coupled with a rise in craft and operator workers, might facilitate the process of union organization.⁴⁵

Employing the comprehensive linked full-count Census data for the first time in this context, I calculate the number of individuals who transitioned from laborer to craft and operative occupations between the successive census years of 1880 to 1910 (Helgertz et al., 2023). In the Online Appendix Table D16, I disclose the results illustrating convict labor’s influence on various occupational transitions within the entire sample and the subset of the sample with a presence of the Knights of Labor and without. These factors include the number of laborers transitioning to craftsmen or operative roles.

The findings suggest that convict labor bolsters the transition of laborers to craftsmen and operatives, an effect that is intensified in the absence of unions. Statistically significant transitions from labor to craftsman positions occur notably in the absence of unions, while unions’ presence does not generate a similar impact. The shift from labor roles to operative positions is less pronounced in the presence of unions than without unions, despite both instances being statistically significant. This suggests an indirect influence where the prevalence of convict labor may inadvertently boost employment within the craft or operative roles, thereby aiding the process of union organization.

5.4.3 Unlikely Mechanisms

In this section, I evaluate possible mechanisms that, while feasible, are unlikely to explain the observed results. I probe whether increased labor supply, political influence, or the KOL’s proto-socialist nature account for the findings.

A plausible theory might link a rise in local labor supply to increased union formation. To investigate this, I use several migration metrics to examine how convict labor impacts local labor supply. These metrics cover overall inbound and outbound migration rates, specific to the white population and those employed within the agricultural and manufacturing sectors. I determine these rates by quantifying the number of individuals per 100,000 residents who migrated into or out of each county in each category. Using linked full-count census data, I track individuals transitioning between counties from 1880 to 1910 for each census year (Helgertz et al., 2023). A statistically significant shift in the migration of white

⁴⁵Unionization within craft and operative occupations was a relatively straightforward process. By the late-nineteenth century, trade unions had established a significant foothold across numerous skilled occupations in the United States (Friedman, 1998, 1999).

individuals—who comprised the majority of the free labor workforce—would impact the local labor supply. However, convict labor had no statistically significant effects on migration patterns, indicating it did not substantially alter the local labor dynamics (Online Appendix Table D17). Additionally, there is no evidence of spillover effects to neighboring counties (Online Appendix Table D18).

An alternative mechanism could be political, where convict labor might influence political outcomes and potentially increase union formations. To explore this, I analyze election results in the Online Appendix Table D19, examining vote share, winning margins, and outcomes (win or loss) for the Democratic Party in presidential and congressional elections prior to party realignment. While there is a slight decrease in the Democratic Party’s vote share and winning margin, these changes do not significantly affect overall election outcomes. Thus, political outcomes likely do not drive the results.

Finally, the Knights of Labor, often seen as forerunners to proto-socialist groups, may have aimed to improve convict laborers’ conditions. However, their main focus was on limiting competitive threats against free laborers. As stated in their preamble, one of their goals was to end the lease system, which allowed convicts to be hired out for work (Ely, 1886). Moreover, resistance to convict labor stemmed from two main concerns: moral abuses associated with it and the economic impact of this competition disrupting and replacing free labor with inmate labor. It is crucial to clarify that the moral abuses discussed here do not involve the mistreatment of convict laborers. Instead, the issue centers on the unfair competition perceived by working citizens as morally wrong (Commons, 1910; Hiller, 1914).⁴⁶ Thus, categorizing the Knights as a proto-socialist entity does not adequately explain the observed impact of convict labor on union formation.

6 Conclusion

Following the abolition of slavery in 1865, the Southern United States experienced a marked rise in the use of convict labor to compensate for the lost labor force. Introducing low-cost convict labor posed direct competition to free workers, jeopardizing their rights and employment opportunities. As a result, wages declined, forming the first labor unions.

This paper offers the first theoretical and empirical study on how the U.S. convict labor system influenced the first labor union formation. Using pre-legislation prison locations and capacities as an instrumental variable, I analyze the impact of post-legislation con-

⁴⁶For example, one argument posits that by allowing prison labor to influence fair competition, the government fails its duties, leading to significant societal harm and forcing law-abiding citizens to undervalue their work (Hiller, 1914).

vict laborers. Evidence reveals a direct relationship between the rise in union assemblies, members, strikes, and counties with larger convict labor shares. Furthermore, convict labor employment consistently influenced union density over the 20th century. By 2020, counties with higher convict laborer concentrations had elevated union memberships. These findings underscore the deep-seated implications of historical and institutional contexts on today's outcomes, emphasizing the lasting consequences of leveraging convict labor as a low-cost workforce on the labor market.

Finally, this paper provides a fresh perspective on U.S. racial conflict's deep-seated roots post-Civil War. It underscores the convict labor system's detrimental impact, highlighting its worsening of conditions for Black convicts and adverse effects on the white labor market. This system spurred union creation in opposition to Black convict labor, potentially perpetuating racial divides in the United States. Future research should investigate the impact of convict labor on firm dynamics, specifically comparing firms that could hire convict labor with those that could not. Convict labor likely benefited some firms over others, affecting them differentially and potentially increasing the labor intensity of particular firms, which may have influenced technological development in the Southern United States. Additionally, it is crucial to explore the role of convict labor in the market integration of the South, both domestically and internationally, through its contribution to the developing manufacturing industry. Finally, examining how convict labor perpetuated racial divides in the Southern United States will provide insights into its broader social implications.

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Online Appendix to:
Convicts and Comrades
by Hazal Sezer

A Model Solution and Proofs

A.1 Labor Demand and Capital

Representative firm's production function is $y = z(\theta l_c + l_f)^\alpha k^\beta$ where $\alpha + \beta < 1$. Solving the maximization problem of the firm and taking the first-order conditions:

$$\begin{aligned} \max_{\{l_f, k\}} \Pi &= y - wl_f - rk \\ \max_{\{l_f, k\}} \Pi &= z(\theta \bar{l} + l_f)^\alpha k^\beta - wl_f - rk, \quad \bar{l} = l_c \end{aligned}$$

Taking the first-order condition w.r.t. l_f :

$$\begin{aligned} \frac{\partial \Pi}{\partial l_f} &\Rightarrow \alpha z k^\beta (\theta \bar{l} + l_f)^{\alpha-1} - w = 0 \\ (\theta \bar{l} + l_f)^{\alpha-1} &= \frac{w}{\alpha z k^\beta} \\ \theta \bar{l} + l_f &= \left(\frac{w}{\alpha z k^\beta} \right)^{\frac{1}{\alpha-1}} \end{aligned}$$

Simplifying l_f becomes:

$$l_f = \left(\frac{w}{\alpha z k^\beta} \right)^{\frac{1}{\alpha-1}} - \theta \bar{l}$$

Taking the first-order condition w.r.t. k :

$$\frac{\partial \Pi}{\partial k} \Rightarrow \beta z (\theta \bar{l} + l_f)^\alpha k^{\beta-1} - r = 0$$

Simplifying k becomes:

$$k = \left(\frac{r}{\beta z (\theta \bar{l} + l_f)^\alpha} \right)^{\frac{1}{\beta-1}}$$

Plugging k into l_f :

$$\begin{aligned}
l_f &= \left(\frac{w}{\alpha z \left(\frac{r}{\beta z (\theta \bar{l} + l_f)} \right)^{\frac{\beta}{\beta-1}}} \right)^{\frac{1}{\alpha-1}} - \theta \bar{l} \\
l_f &= \left(\frac{w}{\alpha} \left(\frac{\beta}{r} \right)^{\frac{\beta}{\beta-1}} z^{\frac{1}{\beta-1}} (\theta \bar{l} + l_f)^{\frac{\alpha \beta}{\beta-1}} \right)^{\frac{1}{\alpha-1}} - \theta \bar{l} \\
l_f &= \left(\frac{w}{\alpha} \right)^{\frac{1}{\alpha-1}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{(\beta-1)(\alpha-1)}} z^{\frac{1}{(\beta-1)(\alpha-1)}} (\theta \bar{l} + l_f)^{\frac{\alpha \beta}{(\beta-1)(\alpha-1)}} - \theta \bar{l} \\
(l_f + \theta \bar{l})^{\frac{1-\alpha-\beta}{(\beta-1)(\alpha-1)}} &= \left(\frac{w}{\alpha} \right)^{\frac{1}{\alpha-1}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{(\beta-1)(\alpha-1)}} z^{\frac{1}{(\beta-1)(\alpha-1)}} \\
l_f + \theta \bar{l} &= \left(\left(\frac{w}{\alpha} \right)^{\frac{1}{\alpha-1}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{(\beta-1)(\alpha-1)}} z^{\frac{1}{(\beta-1)(\alpha-1)}} \right)^{\frac{(\beta-1)(\alpha-1)}{1-\alpha-\beta}}
\end{aligned}$$

Plugging l_f into k and k into l_f , l_f and k become:

$$l_f = \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} - \theta \bar{l} \quad (\text{A.1})$$

$$k = \left(\frac{r}{\beta} \right)^{\frac{\alpha-1}{1-\alpha-\beta}} \left(\frac{\alpha}{w} \right)^{\frac{\alpha}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} \quad (\text{A.2})$$

Total labor supply of free-laborers:

$$\bar{L} = \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} - \theta \bar{l} \quad (\text{A.3})$$

A.2 Equilibrium Wage Rate

Simplifying equation (A.3), we find the equilibrium wage rate:

$$\begin{aligned}
\left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} &= \frac{\bar{L} + \theta \bar{l}}{\left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}}} \\
w &= \alpha \left(\frac{\bar{L} + \theta \bar{l}}{\left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}}} \right)^{\frac{1-\alpha-\beta}{\beta-1}} \quad (\text{A.4})
\end{aligned}$$

Proof of Proposition 1. *The effect of capacity constraint on equilibrium wage rate is negative, $\frac{\partial w}{\partial \bar{l}} < 0$:*

$$\frac{\partial w}{\partial \bar{l}} = \alpha \theta \frac{(1 - \alpha - \beta)}{\beta - 1} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\beta}} z^{\frac{1}{1-\beta}} (\bar{L} + \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}-1} \quad (\text{A.5})$$

When \bar{l} is sufficiently large, the productivity of convict labor and its capacity becomes strategic complements. As a result, enhancing convict productivity further depresses equilibrium wages, $\frac{\partial^2 w}{\partial \bar{l} \partial \theta} > 0$:

$$\begin{aligned} \frac{\partial^2 w}{\partial \bar{l} \partial \theta} = & \alpha \frac{1 - \alpha - \beta}{\beta - 1} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\beta}} z^{\frac{1}{1-\beta}} \\ & \left((\bar{L} + \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}-1} \left(1 + \frac{\theta \bar{l}}{(\bar{L} + \theta \bar{l})} \left(\frac{1 - \alpha - \beta}{\beta - 1} - 1 \right) \right) \right) \end{aligned} \quad (\text{A.6})$$

For equation (A.6) to be positive, it must be that $\left| \left(\frac{1-\alpha-\beta}{\beta-1} - 1 \right) \frac{\theta \bar{l}}{\theta \bar{l} + \bar{L}} \right| > 1$. The first term consistently exceeds one, while the second remains below one. Thus, equilibrium wages decrease with increased convict productivity when \bar{l} is sufficiently large, i.e., above the threshold \bar{l}^* that uniquely solves $\left| \left(\frac{1-\alpha-\beta}{\beta-1} - 1 \right) \frac{\theta \bar{l}}{\theta \bar{l} + \bar{L}} \right| = 1$. ■

A.3 Equilibrium Profit of the Firm

Proof of Proposition 2. *The profit of the representative firm is $\Pi = z (\theta \bar{l} + l_f)^\alpha k^\beta - w l_f - r k$. Solving for the equilibrium level, plug l_f in Π :*

$$\Pi = z^{\frac{1-\beta}{1-\alpha-\beta}} \left(\frac{w}{\alpha} \right)^{\frac{\alpha(\beta-1)}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\alpha\beta}{1-\alpha-\beta}} k^\beta - w \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} + \theta \bar{l} w - r k$$

Plugging k in Π :

$$\begin{aligned} \Pi = & \left(\frac{w}{\alpha} \right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} - w \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} + \theta \bar{l} w \\ & - r \left(\frac{w}{\alpha} \right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{1-\alpha}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} \end{aligned}$$

Defining parts of the equation:

$$\begin{aligned}
(1) &\equiv \left(\frac{w}{\alpha}\right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} \\
(2) &\equiv -w \left(\frac{w}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}} \\
(3) &\equiv \theta \bar{l} w \\
(4) &\equiv -r \left(\frac{w}{\alpha}\right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{1-\alpha}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}}
\end{aligned}$$

Plugging w in Π for all parts of the equation:

$$\begin{aligned}
(1) &\equiv (\bar{L} + \theta \bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} z^{\frac{1}{1-\beta}} \\
(2) &\equiv -\alpha (\bar{L} + \theta \bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} z^{\frac{1}{1-\beta}} \\
(3) &\equiv \alpha \theta \bar{l} (\bar{L} + \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} z^{\frac{1}{1-\beta}} \\
(4) &\equiv -r (\bar{L} + \theta \bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{1}{1-\beta}} z^{\frac{1}{1-\beta}}
\end{aligned}$$

Simplifying $\Pi = (1) + (2) + (3) + (4)$ becomes:

$$\Pi = (\bar{L} + \theta \bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{1}{1-\beta}} z^{\frac{1}{1-\beta}} \left(\frac{r}{\beta}(1 - \alpha - \beta) + \alpha \frac{r}{\beta} \frac{\theta \bar{l}}{\bar{L} + \theta \bar{l}}\right) \quad (\text{A.7})$$

The impact of capacity constraint on equilibrium profit of the representative firm is positive ($\frac{\partial \Pi}{\partial \bar{l}} > 0$). The first part of equation A.7 demonstrates an increasing trend with respect to \bar{l} . To see that $\frac{\theta \bar{l}}{(\bar{L} + \theta \bar{l})}$ is also increasing in \bar{l} , applying the product rule:

$$\underbrace{(\theta(\bar{L} + \theta \bar{l}) - \theta \theta \bar{l})}_{= \theta \bar{L} > 0 \Rightarrow \uparrow \text{ in } \bar{l}} \underbrace{\left(\frac{1}{(\bar{L} + \theta \bar{l})^2}\right)}_{\uparrow \text{ in } \bar{l}}$$

Therefore, it holds that ($\frac{\partial \Pi}{\partial \bar{l}} > 0$). ■

A.4 Union Membership and Strikes

Benefit of workers from striking: Should the strike succeed, the value or benefit (Ω) accrued by the employee is:

$$\Omega = (w(l_c = 0) - w(l_c = \bar{l})) \bar{L} \quad (\text{A.8})$$

Probability of workers joining a union and striking: Assuming a strike at the representative firm can be prevented with probability $1 - p$ or succeed with p , successful strikes lead to the elimination of convict labor and unsuccessful ones to a utility loss, δ . Free labor will strike if:

$$J < p\Omega - (1 - p)\delta = p((w(l_c = 0) - w(l_c = \bar{l})) \bar{L} - (1 - p)\delta) \quad (\text{A.9})$$

Proof of Proposition 3. *The right-hand side of equation (A.9) increases with \bar{l} :*

$$J < p \underbrace{\left(\underbrace{w(l_c = 0)}_{\frac{\partial w(l_c=0)}{\partial l} = 0} - \underbrace{w(l_c = \bar{l})}_{\frac{\partial w(l_c=\bar{l})}{\partial l} < 0} \right)}_{\uparrow \text{ in } \bar{l}} \bar{L} - (1 - p)\delta$$

To see that $(w(l_c = 0) - w(l_c = \bar{l}))$ is increasing in \bar{l} , let us plug in $w(l_c = 0)$ and $w(l_c = \bar{l})$:

$$w(l_c = 0) - w(l_c = \bar{l}) = \alpha \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\beta}} z^{\frac{1}{1-\beta}} \left(\bar{L}^{\frac{1-\alpha-\beta}{\beta-1}} - (\bar{L} + \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}} \right) \quad (\text{A.10})$$

Taking the derivative of this expression w.r.t. \bar{l} , we find:

$$\frac{\partial w(l_c = 0) - w(l_c = \bar{l})}{\partial \bar{l}} = -\frac{1 - \alpha - \beta}{\beta - 1} \theta \alpha \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\beta}} z^{\frac{1}{1-\beta}} (\bar{L} + \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}-1} > 0$$

Therefore, an increase in convict labor capacity results in increased levels of union membership and strike occurrences. ■

B Model Extension: Household Preferences and Endogenous Labor Supply

In this section, the model is expanded to include a labor-leisure trade-off for workers. The aspect of the model pertaining to the representative firm remains unaltered; it builds upon the aspects related to households and the general equilibrium. The model features an economy populated by N households, endowed with a maximum amount of \hat{l} units of labor services. Each household exhibits consumption-labor preferences:

$$u(c, l) = c - \frac{1}{1 + \mu} l^{1+\mu}, \quad (\text{B.1})$$

where c denotes the consumption, l denotes the hours of labor supply, and μ denotes the wage elasticity of labor supply of the free households. Solving the maximization problem of consumers:

$$\begin{aligned} \max_{\{c, l\}} U &= u(c, l) \quad s.t. \quad c = wl \\ \max_{\{c, l\}} U &= wl - \frac{1}{1 + \mu} l^{1+\mu} \end{aligned}$$

$$\frac{\partial U}{\partial l} \Rightarrow -l^\mu + w = 0$$

Equilibrium hours of labor supply provided by households become:

$$l = w^{\frac{1}{\mu}} \quad (\text{B.2})$$

Now, let us use this last expression together with the equation that characterizes the general equilibrium of the economy:

$$\begin{aligned} Nw^{\frac{1}{\mu}} &= l_{f,e} \\ \underbrace{Nw^{\frac{1}{\mu}}}_{\text{Free Labor Supply}} &= \underbrace{\left(\frac{w}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z^{\frac{1}{1-\alpha-\beta}}}_{\text{Free Labor Demand}} \underbrace{-\theta \bar{l}}_{\text{Convict Labor Demand (Capacity)}} \quad (\text{B.3}) \\ &\quad \underbrace{\hspace{15em}}_{\text{Total Labor Demand}} \end{aligned}$$

Key properties of equation (B.3):

Equilibrium uniqueness. The left-hand-side (LHS) of (B.3) monotonically increases in w while the right-hand-side (RHS) monotonically decreases in w . Therefore, the equilibrium is unique.

Equilibrium existence. We have the following limit properties:

$$\begin{aligned} \lim_{w \rightarrow 0} \text{LHS}(w) &= 0, & \lim_{w \rightarrow 0} \text{RHS}(w) &= \infty \\ \lim_{w \rightarrow \infty} \text{LHS}(w) &= \hat{l}, & \lim_{w \rightarrow \infty} \text{RHS}(w) &= -\theta \bar{l} \end{aligned}$$

which jointly imply that the equilibrium exists. Equilibrium uniqueness and equilibrium existence together show that the equilibrium is characterized by a unique wage rate.

The effect of \bar{l} on the equilibrium wage rate is negative – as in the benchmark model in Section 4 where the labor supply is exogenously determined. To recognize this effect, it is essential to understand that an increase in \bar{l} uniformly lowers the right-hand side (RHS) of equation (B.3). Since $\text{RHS}(w)$ declines monotonically in w , while $\text{LHS}(w)$ monotonically increases in w , this implies that a rise in \bar{l} unambiguously lowers the equilibrium wage, w , of the free labor.

The properties obtained above extend (generalize) the previously formalized impact of convict labor capacity result, that it reduces the unit wage rate of free labor, to a set-up where the supply of labor is determined endogenously. Additionally, it provides the insight that the equilibrium labor supply (work hours, voluntary (un)employment) of free-labor adjusts in equilibrium. To understand this, refer to equation (B.2), which demonstrates that the supply of free labor decreases as the wage rate drops. Specifically:

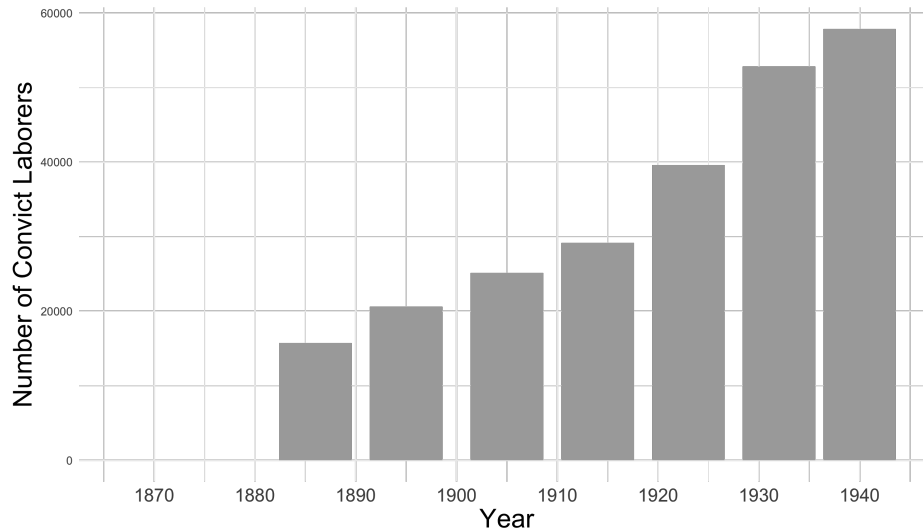
$$\frac{\partial l}{\partial w} = \frac{1}{\mu} w^{\frac{1}{\mu}-1} \quad (\text{B.4})$$

which implies that \bar{l} reduces the equilibrium wage rate of the free labor and in turn also the hours-work of the free labor in equilibrium.

Union Membership and Strikes: The model’s union membership and strike facet remains constant. As before, an increase in the capacity for convict labor results in a surge in unionization and strikes.

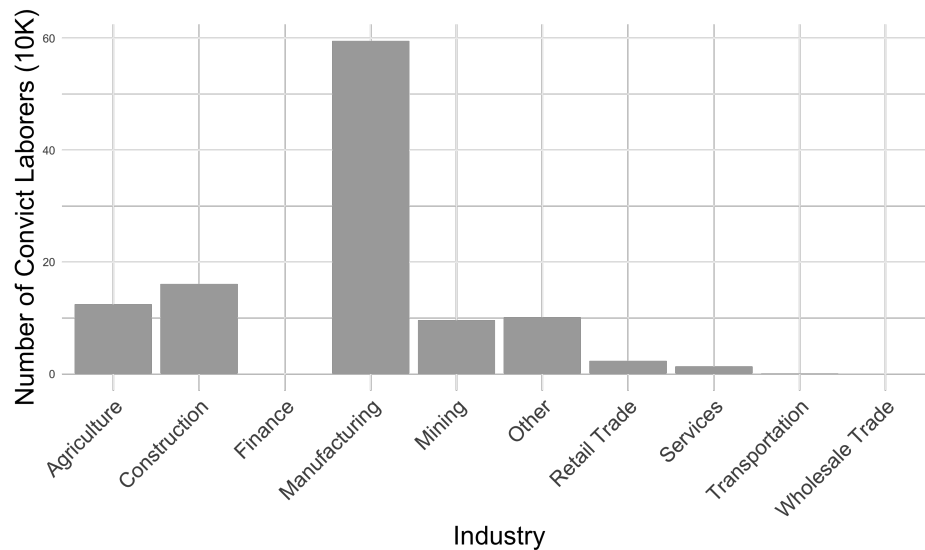
D Figures and Tables

Figure D1. Convict Labor Employment in the South: 1870-1940



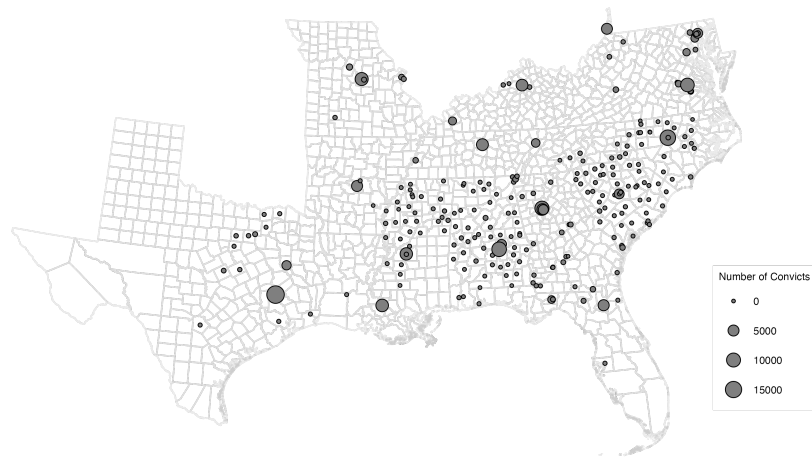
Note: Drawing upon data from the Bureau of Labor reports for the years 1886, 1895, 1905, 1914, 1923, 1932, and 1940, the figure visually portrays the total count of convict laborers in each reported year in the southern United States.

Figure D2. Employment of Convict Labor Across Industries: 1886-1940



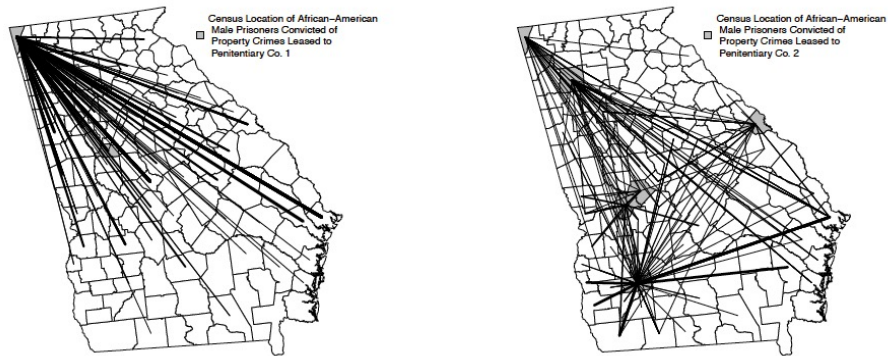
Note: Drawing from the Bureau of Labor reports for the years 1886, 1895, 1905, 1914, 1923, 1932, and 1940, the figure aggregates and illustrates the total number of convict laborers across industries over this period.

Figure D3. Average Number of Convicts 1886-1940



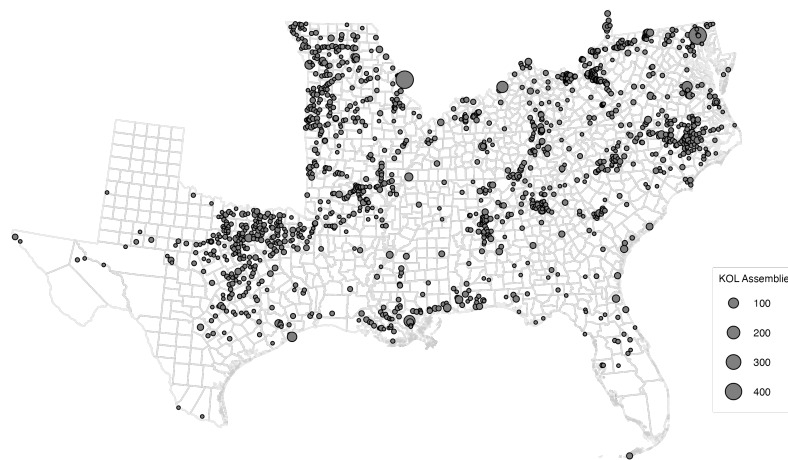
Note: The figure uses data from Bureau of Labor reports for 1886, 1895, 1905, 1914, 1923, 1932, and 1940 to visually display the average number of convicts employed at various prisons. In this illustration, each dot's size reflects the average number of convicts employed at the corresponding prison.

Figure D4. Conviction vs Incarceration Locations



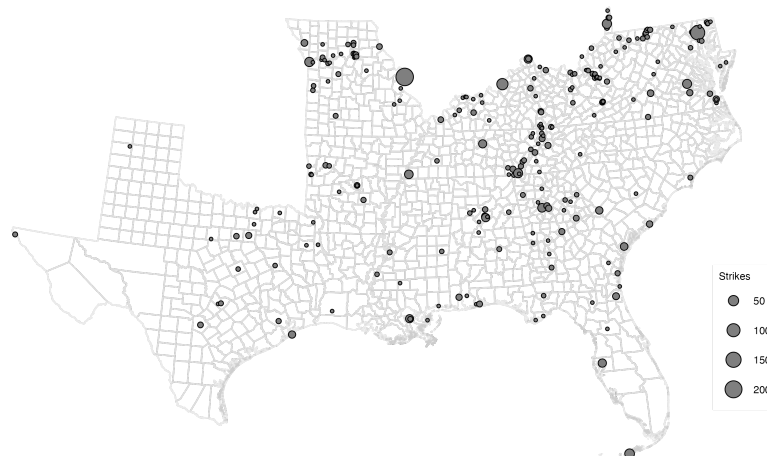
Note: The figure is obtained from Muller (2018). It presents the origin and destination counties of Black male prisoners convicted of property crimes and leased to Georgia Penitentiary Companies.

Figure D5. Geographic Distribution of KOL Assemblies



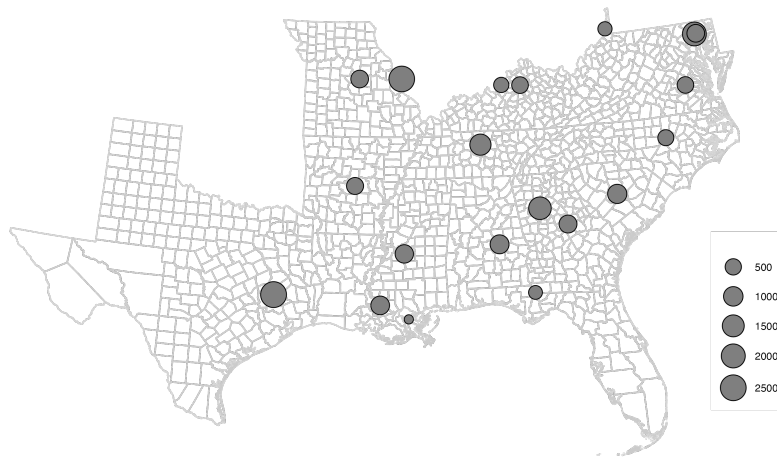
Note: Using data sourced from [Garlock \(1982\)](#) and [Garlock \(1984\)](#), the figure visually illustrates various local KOL assemblies, with each dot marking a distinct assembly.

Figure D6. Geographic Distribution of Strikes



Note: The data from ([U.S. Bureau of Labor, 1888](#)) and ([U.S. Bureau of Labor, 1896](#)) serve as the basis for the figure. This visual representation plots each strike as a dot, with each dot's value corresponding to the total number of strikes in a specific locality.

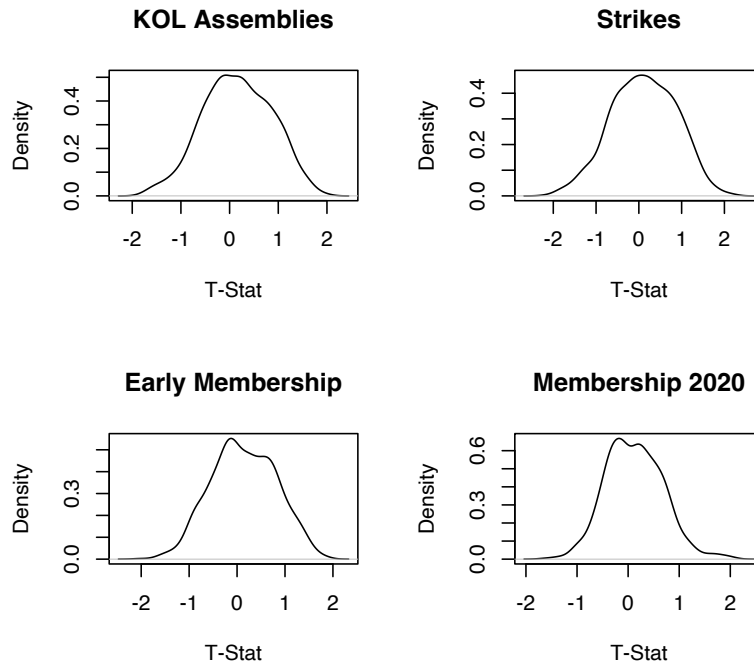
Figure D7. Geographic Distribution of Pre-Convict-Labor Prisons



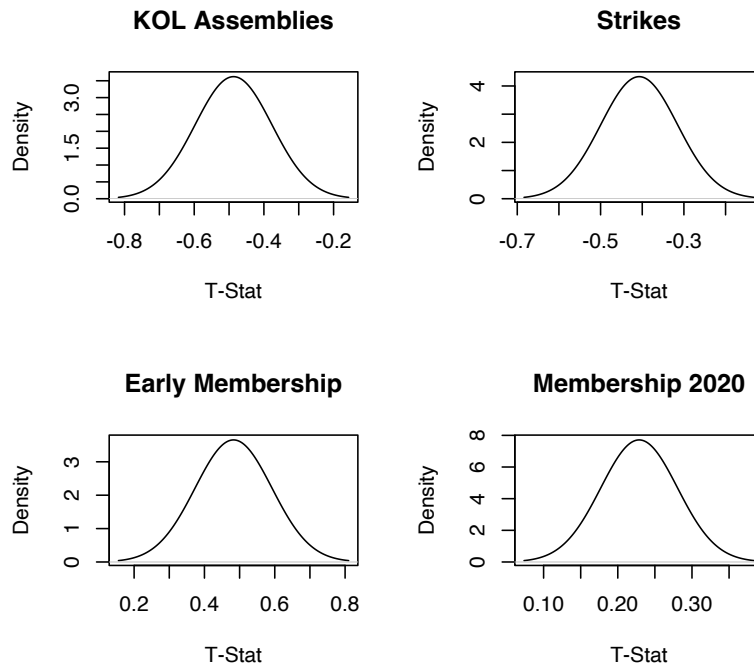
Note: By leveraging data from [North American Review \(1866\)](#), [Prison Association of New York \(1870\)](#), and [Wines \(1871\)](#), the figure offers a visual depiction of pre-convict-labor prisons. Each dot in the figure represents an individual correctional facility, with its size indicating its capacity.

Figure D8. Density Plot for t-statistics of Placebo Regressions

Panel A: Early Prison Dummy IV



Panel B: Early Prison Capacities IV



Note: The figure displays the density distributions of t-statistics obtained from the placebo regressions analogous to equation (11). I use counterfactual pre-convict-labor prison locations to generate graphs 1-4 in Panel A, and capacities to generate graphs 5-8 in Panel B. The actual t-statistics from Table 1 for the number of KOL assemblies and the number of strikes are 2.79 and 2.24, and 2.61 and 2.08 for the early prison dummy and capacities as instruments, respectively. Similarly, the actual t-statistics from Table 2 for early membership and membership in 2020 are 3.82 and 6.63, and 3.77 and 5.84 for the early prison dummy and capacities as instruments, respectively.

Table D1: Effect on KOL Assemblies and Strikes 1886-1905

	<i>Dependent variable:</i>					
	Number of KOL Assemblies			Number of Strikes		
	(1)	(2)	(3)	(4)	(5)	(6)
Convict Labor per 100,000 (IHS)	0.253*** (0.094)	0.232** (0.092)	0.236*** (0.091)	0.114*** (0.041)	0.109*** (0.040)	0.108*** (0.039)
Economic Controls	×	×	×	×	×	×
Geographic Controls		×	×		×	×
Socio-political Controls			×			×
State × Year FE	×	×	×	×	×	×
Mean Outcome	1.23	1.23	1.23	0.28	0.28	0.28
Clusters	1,302	1,302	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (9). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. Outcome variables include the count of local KOL assemblies and strikes. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D2: Effect on Membership 1886-1905 and 2020

	<i>Dependent variable:</i>					
	Early Membership (IHS)			Membership 2020 (IHS)		
	(1)	(2)	(3)	(4)	(5)	(6)
Convict Labor per 100,000 (IHS)	0.052*** (0.019)	0.049*** (0.018)	0.050*** (0.018)	0.142*** (0.023)	0.133*** (0.022)	0.144*** (0.023)
Economic Controls	×	×	×	×	×	×
Geographic Controls		×	×		×	×
Socio-political Controls			×			×
State × Year FE	×	×	×	×	×	×
Mean Outcome	20.83	20.83	20.83	3,525	3,525	3,525
Clusters	1,302	1,302	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (9). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. The outcome variables are the IHS-transformed counts of 19th-century Knights of Labor members and current union members. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D3: Effect on Manufacturing Outcomes 1880-1900

	<i>Dependent variable:</i>					
	Avg. Wage Growth			Employment Share Growth		
	(1)	(2)	(3)	(4)	(5)	(6)
Convict Labor per 100,000 (IHS)	-1.696** (0.735)	-1.543** (0.732)	-1.467** (0.723)	-1.412*** (0.372)	-1.366*** (0.362)	-1.349*** (0.364)
Economic Controls	×	×	×	×	×	×
Geographic Controls		×	×		×	×
Socio-political Controls			×			×
State × Year FE	×	×	×	×	×	×
Mean Outcome	26.4	26.4	26.4	94.8	94.8	94.8
Clusters	1,154	1,154	1,154	1,231	1,231	1,231
Observations	2,265	2,265	2,265	2,439	2,439	2,439

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (9). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. The outcome variables are the average manufacturing wage growth and the manufacturing employment share. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D4: First Stage: Effect on Convict Labor per 100,000 Inhabitants

	<i>Dependent variable:</i>					
	Convict Labor per 100,000 (IHS)					
	(1)	(2)	(3)	(4)	(5)	(6)
Early Prison Dummy	6.677*** (0.588)	6.694*** (0.589)	6.663*** (0.592)			
Early Prison Capacities (IHS)				0.941*** (0.068)	0.943*** (0.068)	0.939*** (0.069)
Economic Controls	×	×	×	×	×	×
Geographic Controls		×	×		×	×
Socio-political Controls			×			×
State × Year FE	×	×	×	×	×	×
Mean Outcome	76.07	76.07	76.07	76.07	76.07	76.07
Clusters	1,302	1,302	1,302	1,302	1,302	1,302
F-stat	89.45	81.67	80.6	92.24	84.22	83.1
Observations	3,906	3,906	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (10). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the dependent variable. Independent variables include pre-convict labor prison locations and capacities. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The regression includes state × year fixed effects, with standard errors clustered at the county level.

Table D5: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership (IHS)	Membership 2020 (IHS)
	(1)	(2)	(3)	(4)
Early Prison Dummy	7.772*** (2.509)	2.997** (1.290)	1.244*** (0.307)	1.496*** (0.284)
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses). The observation unit is the county (1880), with locations of pre-convict labor prisons as the independent variable. The outcome variables are the count of local KOL assemblies, strikes, IHS-transformed 19th-century Knights of Labor members and current union members. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D6: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership (IHS)	Membership 2020 (IHS)
	(1)	(2)	(3)	(4)
Early Prison Capacities (IHS)	1.057*** (0.377)	0.458** (0.208)	0.170*** (0.043)	0.208*** (0.038)
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses). The observation unit is the county (1880), with capacities of pre-convict labor prisons as the independent variable. The outcome variables are the count of local KOL assemblies, strikes, IHS-transformed 19th-century Knights of Labor members and current union members. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D7: Effect of Covariates on the Presence of Early Prisons

	<i>Dependent variable:</i>	
	Early Prison Dummy	Early Prison Capacities (IHS)
	(1)	(2)
White Population Share	0.940 (1.793)	7.556 (13.121)
Free Colored Population Share	0.861 (1.840)	6.886 (13.532)
Enslaved Population Share	0.945 (1.808)	7.609 (13.249)
Population Density	0.003*** (0.001)	0.016* (0.008)
Number of Manufacturing Establishments	0.0002 (0.0002)	0.001 (0.001)
Manufacturing Output Value per Laborer	-0.00000 (0.00000)	-0.00001 (0.00001)
Farm Output Value per Acre of Farmland	0.001 (0.001)	0.009 (0.006)
Number of Improved Acres of Farmland	-0.00000 (0.00000)	-0.00000 (0.00000)
Number of Farms	0.00001 (0.00002)	0.0001 (0.0001)
Elevation	0.0002 (0.0004)	0.001 (0.003)
Distance to Coast	0.00001 (0.00004)	0.00003 (0.0003)
Access to Railways	0.020* (0.010)	0.159** (0.076)
Access to Navigable Water Bodies	0.012 (0.010)	0.085 (0.072)
Distance to the Nearest Urban Center	-0.00003 (0.00005)	-0.0003 (0.0003)
Vote Share for the Democratic Party	-0.0001 (0.0002)	-0.001 (0.001)

Note: Each column presents coefficients and standard errors (in parentheses) from distinct regressions, where the pre-convict-labor prison dummy and capacities are regressed on all 1860 covariates. Each row discloses the coefficients of the incorporated covariates. The model includes state fixed effects, and standard errors are clustered at the county level.

Table D8: Effect on Survival of KOL 1886-1905

Early Prison Dummy and Capacities IV	<i>Dependent variable:</i>			
	Avg. Survival (1)	New KOL (2)	Avg. Survival (3)	New KOL (4)
Convict Labor per 100,000 (IHS)	0.161*** (0.056)	0.171** (0.069)	0.162*** (0.057)	0.156** (0.063)
KP F-stat	126.59	126.59	186.55	186.55
Controls	×	×	×	×
State × Year FE	×	×	×	×
IV: Early Prison Dummy	×	×		
IV: Early Prison Capacities			×	×
Mean Outcome	0.52	0.24	0.52	0.24
Clusters	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variable is the average survival period of all KOL assemblies and the number of newly formed KOL assemblies. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D9: Effect on Manufacturing Outcomes 1880-1900

Panel A: Early Prison Dummy IV	<i>Dependent variable:</i>			
	Avg. Wage Growth (1)	Employment Share (2)	Output Value (3)	Gross Profit (4)
Convict Labor per 100,000 (IHS)	-1.581* (0.816)	-1.920*** (0.670)	181,093** (77,583)	31,254** (13,983)
KP F-stat	86.79	86.05	86.03	86.03
Panel B: Early Prison Capacities IV				
Convict Labor per 100,000 (IHS)	-1.382* (0.771)	-1.914*** (0.629)	185,888** (88,966)	31,160** (15,330)
KP F-stat	121.15	120.26	120.4	120.4
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	26.4	94.8	970,522	79,779
Clusters	1,154	1,231	1,298	1,298
Observations	2,265	2,439	2,526	2,526

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the average manufacturing wage growth, manufacturing employment share, manufacturing output value, and gross profits. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D10: Effect on Farming Outcomes 1880-1900

	<i>Dependent variable:</i>								
	Δ Acimp	Δ Acumimp	Δ Farmval	Δ Equipval	Δ Farmout	Δ Farms	Δ Farm Tenant	Δ Sharecrop	Δ Acres
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Convict Labor per 100,000 (IHS)	3,228.285 (3,089.867)	50,837.530 (42,596.680)	74,527.800 (73,407.690)	2,223.608 (1,694.205)	31,445.680 (24,330.500)	-8.361 (13.766)	-34.577 (63.487)	-762.436 (608.406)	9,924.631 (8,938.648)
KP F-stat	86.07	86.07	86.07	86.07	86.07	86.07	86.07	86.07	86.07
Panel B: Early Prison Capacities IV									
Convict Labor per 100,000 (IHS)	2,683.423 (2,846.639)	43,392.750 (38,852.110)	63,282.970 (65,887.090)	1,888.140 (1,534.479)	26,862.420 (22,227.830)	-10.008 (13.734)	-37.125 (63.651)	-774.254 (613.703)	8,363.240 (8,236.992)
KP F-stat	120.49	120.49	120.49	120.49	120.49	120.49	120.49	120.49	120.49
Controls	x	x	x	x	x	x	x	x	x
State \times Year FE	x	x	x	x	x	x	x	x	x
Mean Outcome	30,139	409,882	575,443	20,157	234,323	120.64	470.48	1,730	78,527
Clusters	1,284	1,284	1,284	1,284	1,284	1,284	1,284	1,284	1,284
Observations	2,543	2,543	2,543	2,543	2,543	2,543	2,543	2,543	2,543

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the changes in acres of improved and unimproved farmland, farm value, farm equipment value, farm output value, number of farms, number of farm tenants and sharecroppers, and total acres of farmland. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state \times year fixed effects, and standard errors are county-level clustered.

Table D11: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership (IHS)	Membership 2020 (IHS)
	(1)	(2)	(3)	(4)
Panel A: Early Prison Dummy IV				
Convict Labor per 100,000 (IHS)	0.633*** (0.245)	0.196* (0.113)	0.175*** (0.042)	0.253*** (0.049)
KP F-stat	105.78	105.78	105.78	105.43
Panel B: Early Prison Capacities IV				
Convict Labor per 100,000 (IHS)	0.590** (0.245)	0.246** (0.121)	0.166*** (0.042)	0.246*** (0.047)
KP F-stat	130.76	130.76	130.76	130.44
Controls	x	x	x	x
State \times Year FE	x	x	x	x
Mean Outcome	2.45	0.77	57.98	6,760
Clusters	295	295	295	294
Observations	788	788	788	785

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variable is the IHS-transformed 19th-century KOL membership. The control group is identified using a nearest-neighbor matching procedure based on population density, the distance to the coast, elevation, access to railways and navigable water bodies, and the distance to the nearest urban center. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state \times year fixed effects, and standard errors are county-level clustered.

Table D12: Effect on the Change in Manufacturing Covariates 1860-1870

<i>Dependent variable:</i>				
	Δ Farm Output Value	Δ M. Establishments	Δ M. Output Value	Δ M. Total Employed
Panel A	(1)	(2)	(3)	(4)
Early Prison Dummy	-5.680 (9.554)	158.062 (110.824)	-11.668 (120.981)	1.778 (71.921)
Panel B	(5)	(6)	(7)	(8)
Early Prison Capacities (IHS)	-0.769 (1.276)	21.902 (15.148)	-2.830 (14.796)	0.533 (9.431)
Controls	×	×	×	×
State FE	×	×	×	×
Mean Outcome	-14.53	189.72	268.1	151.88
Clusters	1,074	1,074	1,074	1,074
Observations	1,176	1,176	1,176	1,176

Note: Each column reports coefficients and standard errors from a separate regression. The observation unit is the county (1880), with pre-convict labor prisons (Panel A) and their capacities (Panel B) as the independent variables. Outcome variables are the changes in the average farm output value, the number of manufacturing establishments, the manufacturing output value, and the total number of manufacturing employees from 1860-1870. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state-fixed effects and standard errors are county-level clustered.

Table D13: Effect on KOL Assemblies before Convict Labor 1870-1880

<i>Dependent variable:</i>						
Number of KOL Assemblies						
	(1)	(2)	(3)	(4)	(5)	(6)
Early Prison Dummy	0.168 (0.145)	0.163 (0.143)	0.161 (0.141)			
Early Prison Capacities (IHS)				0.029 (0.025)	0.028 (0.025)	0.028 (0.025)
Economic Controls	×	×	×	×	×	×
Geographic Controls		×	×		×	×
Socio-political Controls			×			×
State \times Year FE	×	×	×	×	×	×
Mean Outcome	0.03	0.03	0.03	0.03	0.03	0.03
Clusters	1,302	1,302	1,302	1,302	1,302	1,302
Observations	15,624	15,624	15,624	15,624	15,624	15,624

Note: Each column reports coefficients and standard errors (in parentheses) from a separate regression. The observation unit is the county (1880), with pre-convict labor prisons and their capacities as the independent variables. The outcome variables are the number of KOL assemblies before convict labor laws were enacted. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state \times year fixed effects and standard errors are county-level clustered.

Table D14: Testing Monotonicity Assumption

<i>Dependent variable:</i>										
Convict Labor per 100,000 (IHS)										
	High Density (1)	High Farm OV (2)	High Acimp (3)	High Coast (4)	High Ele (5)	High Rail (6)	High Water (7)	High Urban (8)	High DemVote (9)	High S.Share (10)
Panel A										
Early Prison Dummy	6.351*** (0.969)	6.852*** (0.740)	6.648*** (0.714)	7.705*** (0.604)	8.206*** (0.143)	6.427*** (0.725)	6.388*** (0.859)	6.212*** (1.398)	6.210*** (0.874)	6.559*** (0.852)
Early Prison Capacities (IHS)	0.896*** (0.134)	0.985*** (0.083)	0.930*** (0.084)	1.100*** (0.074)	1.143*** (0.091)	0.899*** (0.090)	0.924*** (0.096)	0.865*** (0.144)	0.905*** (0.103)	0.910*** (0.102)
Observations	300	1,317	1,560	1,878	1,494	840	1,173	1,515	2,379	1,680
	Low Density (11)	Low Farm OV (12)	Low Acimp (13)	Low Coast (14)	Low Ele (15)	Low Rail (16)	Low Water (17)	Low Urban (18)	Low DemVote (19)	Low S.Share (20)
Panel B										
Early Prison Dummy	7.131*** (0.791)	6.863*** (0.846)	6.847*** (1.003)	6.263*** (0.794)	6.487*** (0.660)	7.231*** (0.957)	7.044*** (0.775)	6.728*** (0.629)	7.854*** (0.492)	6.653*** (0.738)
Early Prison Capacities (IHS)	0.988*** (0.085)	0.963*** (0.103)	0.997*** (0.111)	0.895*** (0.086)	0.918*** (0.077)	1.034*** (0.106)	0.962*** (0.102)	0.957*** (0.080)	1.071*** (0.061)	0.956*** (0.082)
Observations	3,606	2,589	2,346	2,028	2,412	3,066	2,733	2,391	1,527	2,226
Controls	×	×	×	×	×	×	×	×	×	×
State × Year FE	×	×	×	×	×	×	×	×	×	×
Mean Outcome	76.07	76.07	76.07	76.07	76.07	76.07	76.07	76.07	76.07	76.07
Clusters	1,202	863	782	676	804	1,022	911	797	509	742

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (10). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the dependent variable. Independent variables include pre-convict labor prison locations and capacities. Panel A displays data for the economic, geographic, and socio-political controls, filtered for the subsample falling above the median. Conversely, Panel B illustrates the same controls, but for the subsample that falls below the median. The regression includes state × year fixed effects, with standard errors clustered at the county level.

Table D15: Presence of the Knights of Labor and Average Manufacturing Wage Growth 1880-1910

<i>Dependent variable:</i>			
Average Manufacturing Wage Growth			
	All (1)	KOL Presence (2)	No KOL Presence (3)
Panel A: Early Prison Dummy IV			
Convict Labor per 100,000 (IHS)	-1.581* (0.816)	1.047 (1.998)	-3.240*** (1.061)
KP F-stat	86.79	271.02	24.88
Panel B: Early Prison Capacities IV			
Convict Labor per 100,000 (IHS)	-1.382* (0.771)	1.138 (1.970)	-3.246*** (1.047)
KP F-stat	121.15	326.9	33.81
Controls	×	×	×
State × Year FE	×	×	×
Mean Outcome	26.4	44.34	21.77
Clusters	1,154	411	1,097
Observations	2,265	465	1,800

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons and their capacities. Outcome variable is the average manufacturing wage growth. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D16: Presence of the Knights of Labor and Effect on Craftsmen and Operatives 1880-1910

	<i>Dependent variable:</i>					
	Labor – Craft	Labor – Operative	Labor – Craft	Labor – Operative	Labor – Craft	Labor – Operative
	All		KOL Presence		No KOL Presence	
	(1)	(2)	(3)	(4)	(5)	(6)
Panel A: Early Prison Dummy IV						
Convict Labor per 100,000 (IHS)	0.421** (0.196)	4.166** (1.928)	0.075 (0.045)	3.716** (1.820)	0.718** (0.356)	3.959** (1.954)
KP F-stat	151	151	217.25	217.25	91.12	91.12
Panel B: Early Prison Capacities IV						
Convict Labor per 100,000 (IHS)	0.373** (0.163)	4.234* (2.177)	0.086* (0.050)	3.869* (2.107)	0.626** (0.292)	3.983* (2.187)
KP F-stat	225.21	225.21	322.59	322.59	128.42	128.42
Controls	×	×	×	×	×	×
State × Year FE	×	×	×	×	×	×
Mean Outcome	0.54	4.6	0.21	9.54	0.61	3.6
Clusters	1,278	1,278	421	421	1,275	1,275
Observations	2,510	2,510	423	423	2,087	2,087

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons and their capacities. Outcome variable is the number of laborers transitioning into craftsmen and operatives. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D17: Effect on Migration 1886-1905

	<i>Dependent variable:</i>							
	Out-Migration				In-Migration			
	All	White	Agriculture	Manufacturing	All	White	Agriculture	Manufacturing
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Panel A: Early Prison Dummy IV								
Convict Labor per 100,000 (IHS)	12.771 (13.557)	12.898 (13.566)	3.949 (5.082)	0.170 (0.127)	126.437 (86.205)	126.236 (86.227)	55.914 (42.088)	0.269 (0.269)
KP F-stat	151.15	151.15	151.15	151.15	151.15	151.15	151.15	151.15
Panel B: Early Prison Capacities IV								
Convict Labor per 100,000 (IHS)	10.461 (12.549)	10.581 (12.559)	3.168 (4.620)	0.177 (0.120)	109.943 (79.426)	109.725 (79.444)	47.834 (38.779)	0.266 (0.242)
KP F-stat	225.36	225.36	225.36	225.36	225.36	225.36	225.36	225.36
Controls	×	×	×	×	×	×	×	×
State × Year FE	×	×	×	×	×	×	×	×
Mean Outcome	312.06	309.01	79.03	1.8	1,031.66	1,028.93	462.31	1.29
Clusters	1,271	1,271	1,271	1,271	1,271	1,271	1,271	1,271
Observations	2,503	2,503	2,503	2,503	2,503	2,503	2,503	2,503

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). The outcome variables represent the number of incoming and outgoing migrants across the entire population, specifically among white individuals, as well as those employed in the agriculture and manufacturing sectors. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D18: Spillovers to Neighboring Counties

	<i>Dependent variable:</i>			
	KOL Assemblies (1)	Strikes (2)	Early Membership (IHS) (3)	Membership 2020 (IHS) (4)
Panel A: Early Prison Dummy IV				
Convict Labor per 100,000 (IHS)	0.257 (0.205)	0.159 (0.110)	0.049 (0.038)	0.066* (0.035)
KP F-stat	126.59	126.59	126.59	126.59
Panel B: Early Prison Capacities IV				
Convict Labor per 100,000 (IHS)	0.238 (0.196)	0.149 (0.101)	0.046 (0.038)	0.067** (0.033)
KP F-stat	186.55	186.55	186.55	186.55
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.14	0.22	17.86	3694.31
Clusters	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the number of KOL assemblies, strikes, IHS-transformed 19th-century KOL, and 2020 union membership. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table D19: Effect on Election Outcomes 1886-1905

	<i>Dependent variable:</i>					
	Presidential			Congressional		
	Vote Share (1)	Win Margin (2)	Win (3)	Vote Share (4)	Win Margin (5)	Win (6)
Panel A: Early Prison Dummy IV						
Convict Labor per 100,000 (IHS)	-0.635*** (0.230)	-1.270*** (0.461)	-0.001 (0.010)	-0.583** (0.269)	-1.165** (0.539)	-0.009 (0.009)
KP F-stat	125.11	125.11	125.11	269.98	269.98	269.98
Panel B: Early Prison Capacities IV						
Convict Labor per 100,000 (IHS)	-0.629*** (0.227)	-1.258*** (0.454)	-0.001 (0.010)	-0.592** (0.269)	-1.183** (0.538)	-0.008 (0.009)
KP F-stat	195.98	195.98	195.98	309.61	309.61	309.61
Controls	×	×	×	×	×	×
State × Year FE	×	×	×	×	×	×
Mean Outcome	61.05	22.1	0.71	61.83	23.67	0.7
Clusters	1,290	1,290	1,290	1,286	1,286	1,286
Observations	3,797	3,797	3,797	3,403	3,403	3,403

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). The outcome variables encompass the Democratic Party's vote share, victory margin, and election results for both presidential and congressional elections. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The model includes state × year fixed effects, and standard errors are county-level clustered.

Supplementary Appendix to:
Convicts and Comrades
by Hazal Sezer

E Model with Elite and Regular Firms

To formalize the effect of convict labor on labor market outcomes and unionization, I introduce a novel general equilibrium framework capturing the labor market impacts of convict labor. I construct a two-firm model in which the *elite firm* utilizes convict labor, while the *regular firm* operates without it. I employ this two-firm framework to underscore labor reallocation and to show that firms unable to use convict labor also benefit from this policy. Given the Census data indicating multiple firms in a county during this period, this model effectively represents both firm types within a county. For insights into a representative firm, one can examine the elite firm component of the model; when isolated, it provides predictions of a representative firm.

The model reveals that an increase in convict laborers leads to lower wages for free laborers, a reallocation of free laborers from elite to regular firms, along with an increase in union and strike activities. This model’s broad framework is robust to general firm-type distribution for elite and regular firms and the free workers’ endogenous labor supply.

E.1 Convict and Free Labor

This paper examines an era in the Southern U.S. when oppressive laws led to Black individuals constituting most convict labor while white individuals dominated free labor. These coercive laws and institutions allowed imprisoned Black individuals to be exploited as a source of cheap labor. Consequently, these laws segregated the two labor groups and prohibited unionization among convict laborers (Hiller, 1914).

During the period in question, convict labor is a limited resource—there are no convicts in prisons who are not contributing labor (U.S. Bureau of Labor, 1886). This introduces a stringent capacity constraint, and any exogenous increase in prison capacity may significantly impact the wages of free labor and stimulate union formation, among other effects. The model’s introduction of this capacity constraint represents a novel approach to understanding convict labor employment during this period. In terms of compensation, convict laborers do not receive salary payments, denoted as $w_c = 0$ in the model. On the contrary, free labor receives a wage denoted by $w > 0$.

The model posits that convict and free laborers are perfect substitutes, an assumption supported by Bureau of Labor reports (U.S. Bureau of Labor, 1886, 1925). These reports indicate that employers could use convict labor in any role generally occupied by free labor. Nevertheless, there may be concerns about specific tasks where convict labor does not perfectly substitute free labor. To address this, in the Supplementary Appendix Section H, I introduce a constant elasticity of substitution model, theoretically and empirically exploring this case. The findings from this model qualitatively align with the predictions of the model presented here.

E.2 Elite and Regular Firms

There are two types of firms: elite and regular. Elite firms employ both types of workers, convict and free labor, while regular firms only hire free labor.⁴⁷ Both types of firms produce a homogeneous product without quality differentiation and compete in the labor market.

The elite firm's (e) production function is:

$$y_e = z_e(\theta l_{c,e} + l_{f,e})^\alpha k_e^\beta, \quad (\text{E.1})$$

where y_e is the output of the elite firm and $\alpha + \beta < 1$. The elite firm hires $l_{c,e}$ units of convict labor and $l_{f,e}$ units of free labor and acquires capital in the measure of k_e . The total factor productivity of the elite firm is denoted by z_e . The model permits the potential for productivity disparity between free and convict laborers through the θ multiplier preceding $l_{c,e}$. The elite firm independently determines the quantity of each labor type to employ and the capital to procure. The number of convict laborers hired, $l_{c,e}$, is subject to a capacity constraint $l_{c,e} \leq \bar{l}$. As the wage rate for convict labor is zero, the elite firm's convict labor capacity constraint always binds, i.e., $l_{c,e} = \bar{l}$. I assume $w_c = 0$ without loss of generality. If $w_c = c$, the convict labor capacity constraint of the elite firm still binds as long as c is sufficiently small. In this case, the total labor cost of the elite firm would become $c\bar{l}$, and the optimization program would remain unaltered since $c\bar{l}$ is a constant lump sum. The Bureau of Labor reports (U.S. Bureau of Labor, 1886, 1925) implies that the unit convict labor costs (w_c) were sufficiently small by that time, such that convict labor supply was employed in production to its full extent.

The regular firm's (r) production function is:

$$y_r = z_r l_{f,r}^\alpha k_r^\beta \quad (\text{E.2})$$

⁴⁷An elite firm is one with political connections that grant access to low-cost convict labor.

where y_r is the output of the regular firm and $\alpha + \beta < 1$. The regular firm hires $l_{f,r}$ units of free labor and acquires capital in the measure of k_r . The total factor productivity of the regular firm is denoted by z_r .

Wages of free labor, w , will be determined endogenously in general equilibrium. Rental rate $r > 0$ of capital is exogenous, with full depreciation of capital. Solving the maximization problem of the elite firm, demands for $l_{f,e}$ and k_e become:

$$l_{f,e} = \left(\frac{w}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} - \theta \bar{l} \quad (\text{E.3})$$

$$k_e = \left(\frac{r}{\beta}\right)^{\frac{\alpha-1}{1-\alpha-\beta}} \left(\frac{\alpha}{w}\right)^{\frac{\alpha}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} \quad (\text{E.4})$$

By solving the optimization problem for the regular firm, the demands for $l_{f,r}$ and k_r are derived as:

$$l_{f,r} = \left(\frac{w}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \quad (\text{E.5})$$

$$k_r = \left(\frac{r}{\beta}\right)^{\frac{\alpha-1}{1-\alpha-\beta}} \left(\frac{\alpha}{w}\right)^{\frac{\alpha}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \quad (\text{E.6})$$

E.3 Households and General Equilibrium Wages

There are N households in the economy whose total labor supply equals \bar{L} , and they inelastically supply their labor services for the elite and regular firms. The total demand for free labor in the main industry is expressed as $l_{f,e} + l_{f,r}$. The general equilibrium of the economy is determined by:

$$\bar{L} = \left(\frac{w}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right) - \theta \bar{l}, \quad (\text{E.7})$$

which characterizes the equilibrium wage:

$$w = \alpha \left(\frac{\left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)}{\bar{L} + \theta \bar{l}} \right)^{\frac{1-\alpha-\beta}{1-\beta}}. \quad (\text{E.8})$$

Equation (E.8) shows that an increase in the capacity constraint for convict labor results in a corresponding decrease in the wages of free labor ($\frac{\partial w}{\partial \bar{l}} < 0$). Furthermore, when \bar{l} is sufficiently large, the productivity of convict labor and its capacity become strategic com-

plements. As a result, enhancing convict labor productivity further depresses equilibrium wage level $\left(\frac{\partial^2 w}{\partial l \partial \theta} > 0\right)$. This leads to the following proposition:

Proposition E1: *Increasing the capacity constraint of convict labor reduces the equilibrium wage rate. Furthermore, if \bar{l} exceeds a threshold \bar{l}^* , an increase in θ further lowers the equilibrium wage rate.*

E.4 Labor Reallocation

A direct implication of equations (E.4) and (E.6) is that a decline in wages is associated with an increased demand for capital across all firms. However, diminishing wages for free labor and elite firms employing convict labor reallocates free laborers from elite to regular firms. Based on this, I formalize the following:

Proposition E2: *Increasing the capacity constraint for convict labor causes a reallocation of free labor from the elite to the regular firm.*

The labor reallocation dynamic in my model parallels that of [Hubmer and Restrepo \(2022\)](#). The authors find that larger firms automate more, reducing the overall labor share. Consequently, the median firm sees a rise in labor share due to workers migrating from large to median firms. In this model, elite firms replace free labor with convict labor – free labor shifts to regular firms as wages decline, amplifying their labor intensity. The key distinction from [Hubmer and Restrepo \(2022\)](#) lies in the underlying substitution mechanism: while [Hubmer and Restrepo \(2022\)](#) focus on labor substitution with capital, my model emphasizes the substitution of free labor with convict labor.

E.5 Equilibrium Profits

As the capacity for convict labor expands and equilibrium wages decrease, it is reasonable to anticipate a corresponding impact on firms' profits. The profits of the elite firm are given by:

$$\begin{aligned} \Pi_e = & (\bar{L} + \theta \bar{l})^{\frac{\alpha}{1-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right)^{\frac{\alpha}{\beta-1}} \left(\frac{\beta}{r} \right)^{\frac{1}{1-\beta}} \left(\frac{r}{\beta} (1 - \alpha - \beta) \right) \\ & + \alpha \theta \bar{l} (\bar{L} + \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right)^{\frac{1-\alpha-\beta}{1-\beta}}. \end{aligned} \quad (\text{E.9})$$

The capacity constraint of convict labor positively influences the equilibrium profits of the elite firm, i.e., $\frac{\partial \Pi_e}{\partial l} > 0$, because the elite firm can access convict labor input at no cost.

The profits of the regular firm are given by:

$$\Pi_r = (\bar{L} + \theta \bar{l})^{\frac{\alpha}{1-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right)^{\frac{\alpha}{\beta-1}} \left(\frac{\beta}{r} \right)^{\frac{1}{1-\beta}} \left(\frac{r}{\beta} (1 - \alpha - \beta) \right). \quad (\text{E.10})$$

The equilibrium profits of the regular firm increase with the capacity constraint of convict labor, i.e., $\frac{\partial \Pi_r}{\partial \bar{l}} > 0$. This occurs as free labor shifts from the elite firm to the regular one due to the elite's advantage of accessing convict labor without incurring any costs. This gives rise to the ensuing proposition:

Proposition E3: *An increase in the capacity for convict labor increases profits for both elite and regular firms.*

This rise in firm profits, captured by my model, formalizes the argument why convict labor found such a high demand by elite firm owners and overall support by all firm owners, including those who do not hire convicts, in counties where convict labor became available. The model suggests that regular firm owners also benefited from convict labor due to reduced free labor wages and the general equilibrium reallocation of free labor from elite firms.

E.6 Capital–Labor Cost Ratio

Next, I explore the effects of convict labor on the capital–labor cost ratio, assessing its impact across the entire county, encompassing elite and regular firms. The capital-to-labor cost ratio for the county is denoted as $CI \equiv \frac{rk_e + rk_r}{w_f \gamma \bar{L} + w_c \bar{l}}$. Since $w_c = 0$, the ratio simplifies to:

$$CI \equiv \frac{rk_e + rk_r}{w \bar{L}}. \quad (\text{E.11})$$

The capital-to-labor cost ratio rises with the capacity constraint of convict labor, as both k_e and k_r are increasing in \bar{l} , and w is decreasing in \bar{l} . The underlying rationale is that increased capital demand by all firms in a county raises total capital costs, while fixed free labor supply and falling wages lower overall labor costs. Drawing from this, I formalize:

Proposition E4: *A rise in convict labor capacity increases the capital-to-labor cost ratio throughout the region.*

E.7 Union Membership and Strikes

Suppose that N households, whose inelastic labor supply adds up to \bar{L} , are able to address their collective action problem by investing a total amount J to establish a union. This

union offers free laborers of an elite firm the opportunity to initiate a strike at the beginning of a production cycle against the hiring of convict laborers, which will also be supported by a strike by free laborers at the regular firm — as their wage rate also falls due to the hiring of convicts at the elite firm. It compels the elite firm to curtail convict labor employment to zero with some probability. Should the strike succeed, the value or benefit (Ω) accrued by the employee is as follows. I now write $w(l_c)$ to emphasize the dependence of the equilibrium wage on the capacity constraint:

$$\Omega = (w(l_c = 0) - w(l_c = \bar{l})) \bar{L}, \quad (\text{E.12})$$

and this value determines the free laborer's willingness to coordinate and pay the joint unionization cost J . A clear insight from equation (E.12) suggests that as \bar{l} grows, the likelihood of joining the union also rises. This is because the wage gap, $w(l_c = 0) - w(l_c = \bar{l})$ is increasing in \bar{l} , a conclusion that aligns with findings from the earlier section.

Free labor's willingness to form a union, requiring the payment of cost J , should align with equilibrium beliefs that strikes stand a chance of succeeding. Assume that the owner of the elite firm can prevent a strike with a probability of $1 - p$, and with a probability of p , the strike succeeds. In such an instance, free labor fulfills its demands, and the employment of convict labor in the elite firm is completely eliminated. However, if the strike proves unsuccessful, the laborers suffer a lump sum utility loss, δ . This suggests that free labor would opt to join the union and initiate strikes if and only if:

$$J < p\Omega - (1 - p)\delta = p(w(l_c = 0) - w(l_c = \bar{l})) \bar{L} - (1 - p)\delta. \quad (\text{E.13})$$

Considering the right-hand side of equation (E.13) rises with \bar{l} , we can infer that a greater capacity for convict labor directly spurs increased unionization and strike activities. This leads to the following proposition:

Proposition E5: *An increase in the capacity for convict labor results in heightened unionization and strike actions.*

The growth in the number of union assemblies is embodied in the probability of unionization within this model. Notably, the effects on the extensive and intensive margins move in tandem. These dual channels are currently encapsulated within the J term. The model reveals that the presence of convict labor exerts a downward pressure on the wage rates of free labor while concurrently amplifying their unionization and strike actions.

Furthermore, when \bar{l} reaches a significant magnitude, the productivity of convict labor

and its capacity begin to function as strategic complements. This implies that a boost in convict productivity leads to a further drop in the equilibrium wage. Consequently, the right-hand side of equation (E.13) amplifies with increased convict labor productivity, thereby heightening the impact of convict labor capacity on unionization.

This model guides the empirical analysis and illuminates the core mechanism underlying the results: the intensified labor market competition free labor faces due to convict labor.

F Model Solution and Proofs: Elite and Regular Firms

F.1 Labor Demand and Capital

Elite firm $l_{f,e}$ and k_e : Elite firm's production function is $y_e = z_e (\theta l_{c,e} + l_{f,e})^\alpha k_e^\beta$, $\alpha + \beta < 1$. Solving the maximization problem of the elite firm and taking the first-order conditions:

$$\begin{aligned} \max_{\{l_{f,e}, k_e\}} \Pi_e &= y_e - w l_{f,e} - r k_e \\ \max_{\{l_{f,e}, k_e\}} \Pi_e &= z_e (\theta \bar{l} + l_{f,e})^\alpha k_e^\beta - w l_{f,e} - r k_e, \quad \bar{l} = l_{c,e} \end{aligned}$$

Taking the first-order condition w.r.t. $l_{f,e}$:

$$\begin{aligned} \frac{\partial \Pi_e}{\partial l_{f,e}} &\Rightarrow \alpha z_e k_e^\beta (\theta \bar{l} + l_{f,e})^{\alpha-1} - w = 0 \\ (\theta \bar{l} + l_{f,e})^{\alpha-1} &= \frac{w}{\alpha z_e k_e^\beta} \\ \theta \bar{l} + l_{f,e} &= \left(\frac{w}{\alpha z_e k_e^\beta} \right)^{\frac{1}{\alpha-1}} \end{aligned}$$

Simplifying $l_{f,e}$ becomes:

$$l_{f,e} = \left(\frac{w}{\alpha z_e k_e^\beta} \right)^{\frac{1}{\alpha-1}} - \theta \bar{l}$$

Taking the first-order condition w.r.t. k_e :

$$\frac{\partial \Pi_e}{\partial k_e} \Rightarrow \beta z_e (\theta \bar{l} + l_{f,e})^\alpha k_e^{\beta-1} - r = 0$$

Simplifying k_e becomes:

$$k_e = \left(\frac{r}{\beta z_e (\theta \bar{l} + l_{f,e})^\alpha} \right)^{\frac{1}{\beta-1}}$$

Plugging k_e into $l_{f,e}$:

$$\begin{aligned} l_{f,e} &= \left(\frac{w}{\alpha z_e \left(\frac{r}{\beta z_e (\theta \bar{l} + l_{f,e})^\alpha} \right)^{\frac{\beta}{\beta-1}}} \right)^{\frac{1}{\alpha-1}} - \theta \bar{l} \\ l_{f,e} &= \left(\frac{w}{\alpha} \left(\frac{\beta}{r} \right)^{\frac{\beta}{\beta-1}} z_e^{\frac{1}{\beta-1}} (\theta \bar{l} + l_{f,e})^{\frac{\alpha\beta}{\beta-1}} \right)^{\frac{1}{\alpha-1}} - \theta \bar{l} \\ l_{f,e} &= \left(\frac{w}{\alpha} \right)^{\frac{1}{\alpha-1}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{(\beta-1)(\alpha-1)}} z_e^{\frac{1}{(\beta-1)(\alpha-1)}} (\theta \bar{l} + l_{f,e})^{\frac{\alpha\beta}{(\beta-1)(\alpha-1)}} - \theta \bar{l} \\ (l_{f,e} + \theta \bar{l})^{\frac{1-\alpha-\beta}{(\beta-1)(\alpha-1)}} &= \left(\frac{w}{\alpha} \right)^{\frac{1}{\alpha-1}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{(\beta-1)(\alpha-1)}} z_e^{\frac{1}{(\beta-1)(\alpha-1)}} \\ l_{f,e} + \theta \bar{l} &= \left(\left(\frac{w}{\alpha} \right)^{\frac{1}{\alpha-1}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{(\beta-1)(\alpha-1)}} z_e^{\frac{1}{(\beta-1)(\alpha-1)}} \right)^{\frac{(\beta-1)(\alpha-1)}{1-\alpha-\beta}} \end{aligned}$$

Plugging $l_{f,e}$ into k_e and k_e into $l_{f,e}$, $l_{f,e}$ and k_e become:

$$l_{f,e} = \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} - \theta \bar{l} \quad (\text{F.1})$$

$$k_e = \left(\frac{r}{\beta} \right)^{\frac{\alpha-1}{1-\alpha-\beta}} \left(\frac{\alpha}{w} \right)^{\frac{\alpha}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} \quad (\text{F.2})$$

Regular firm $l_{f,r}$ and k_r : Regular firm's production function is $y_r = z_r l_{f,r}^\alpha k_r^\beta$, $\alpha + \beta < 1$.

Solving the maximization problem of the regular firm:

$$\begin{aligned} \max_{\{l_{f,r}, k_r\}} \Pi_r &= y_r - w l_{f,r} \\ \max_{\{l_{f,r}, k_r\}} \Pi_r &= z_r l_{f,r}^\alpha k_r^\beta - w l_{f,r} - r k_r \end{aligned}$$

Taking the first-order condition w.r.t. $l_{f,r}$:

$$\frac{\partial \Pi_r}{\partial l_{f,r}} \Rightarrow \alpha z_r k_r^\beta (l_{f,r})^{\alpha-1} - w = 0$$

Simplifying $l_{f,r}$ becomes:

$$l_{f,r} = \left(\frac{w}{\alpha z_r k_r^\beta} \right)^{\frac{1}{\alpha-1}}$$

Taking the first-order condition w.r.t. k_r :

$$\frac{\partial \Pi_r}{\partial k_r} \Rightarrow \beta z_r l_{f,r}^\alpha k_r^{\beta-1} - r = 0$$

Simplifying k_r becomes:

$$k_r = \left(\frac{r}{\beta z_r l_{f,r}^\alpha} \right)^{\frac{1}{\beta-1}}$$

Plugging $l_{f,r}$ into k_r and k_r into $l_{f,r}$, $l_{f,r}$ and k_r become:

$$l_{f,r} = \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \quad (\text{F.3})$$

$$k_r = \left(\frac{r}{\beta} \right)^{\frac{\alpha-1}{1-\alpha-\beta}} \left(\frac{\alpha}{w} \right)^{\frac{\alpha}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \quad (\text{F.4})$$

Total labor supply of free-laborers: Adding $l_{f,e}$ and $l_{f,r}$ and simplifying, the total labor supply of free-laborers become:

$$\bar{L} = \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right) - \theta \bar{l} \quad (\text{F.5})$$

F.2 Equilibrium Wage Rate

Simplifying equation (F.5), we find the equilibrium wage rate:

$$\begin{aligned} \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} &= \frac{\bar{L} + \theta \bar{l}}{\left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right)} \\ w &= \alpha \left(\frac{\bar{L} + \theta \bar{l}}{\left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right)} \right)^{\frac{1-\alpha-\beta}{\beta-1}} \end{aligned} \quad (\text{F.6})$$

Proof of Proposition E1. *The effect of capacity constraint on equilibrium wage rate is negative, $\frac{\partial w}{\partial \bar{l}} < 0$:*

$$\frac{\partial w}{\partial \bar{l}} = \frac{\theta(-\alpha - \beta + 1)}{\beta - 1} (\bar{L} + \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}-1} \frac{\alpha}{\left(\frac{\beta}{r}\right)^{\frac{\beta}{\beta-1}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{1-\alpha-\beta}{\beta-1}}} \quad (\text{F.7})$$

When \bar{l} is sufficiently large, the productivity of convict labor and its capacity becomes strategic complements. As a result, enhancing convict productivity further depresses equilibrium wages, $\frac{\partial^2 w}{\partial \bar{l} \partial \theta} > 0$:

$$\frac{\partial^2 w}{\partial \bar{l} \partial \theta} = \frac{1 - \alpha - \beta}{\beta - 1} (\bar{L} + \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}-1} \left(\left(\frac{1 - \alpha - \beta}{\beta - 1} - 1 \right) \frac{\theta \bar{l}}{\theta \bar{l} + \bar{L}} + 1 \right) \frac{\alpha}{\left(\frac{\beta}{r}\right)^{\frac{\beta}{\beta-1}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{1-\alpha-\beta}{\beta-1}}} \quad (\text{F.8})$$

For equation (F.8) to be positive, it must be that $\left| \left(\frac{1-\alpha-\beta}{\beta-1} - 1 \right) \frac{\theta \bar{l}}{\theta \bar{l} + \bar{L}} \right| > 1$. The first term consistently exceeds one, while the second remains below one. Thus, equilibrium wages decrease with increased convict productivity when \bar{l} is sufficiently large, i.e., above the threshold \bar{l}^* that uniquely solves $\left| \left(\frac{1-\alpha-\beta}{\beta-1} - 1 \right) \frac{\theta \bar{l}}{\theta \bar{l} + \bar{L}} \right| = 1$. ■

F.3 Labor Reallocation

To demonstrate the shift of labor from elite to regular firms, it is necessary to show that $\frac{\partial l_{f,e}}{\partial l_{f,r}} < 0$. This involves substituting the equilibrium wage into the expressions for $l_{f,e}$, and $l_{f,r}$:

$$\begin{aligned} l_{f,e} &= (\bar{L} + \theta \bar{l}) \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right)^{-1} z_e^{\frac{1}{1-\alpha-\beta}} - \theta \bar{l} \\ l_{f,r} &= (\bar{L} + \theta \bar{l}) \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right)^{-1} z_r^{\frac{1}{1-\alpha-\beta}} \end{aligned}$$

Then $\frac{l_{f,e}}{l_{f,r}}$ becomes:

$$\frac{l_{f,e}}{l_{f,r}} = \frac{z_e^{\frac{1}{1-\alpha-\beta}}}{z_r^{\frac{1}{1-\alpha-\beta}}} - \frac{\theta \bar{l}}{\bar{L} + \theta \bar{l}} \frac{z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}}{z_r^{\frac{1}{1-\alpha-\beta}}}$$

Proof of Proposition E2. Taking the derivative of $\frac{l_{f,e}}{l_{f,r}}$ with respect to \bar{l} , we find the equilibrium level of labor demand ratio of both firms:

$$\frac{\partial \frac{l_{f,e}}{l_{f,r}}}{\partial \bar{l}} = - \underbrace{\frac{\theta \bar{L}}{(\bar{L} + \theta \bar{l})^2}}_{\downarrow \text{ in } \bar{l}} \frac{z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}}{z_e^{\frac{1}{1-\alpha-\beta}}}$$

Therefore, an increase in the capacity constraint of convict labor leads to a reallocation of free labor from elite to regular firms $\left(\frac{\partial \frac{l_{f,e}}{l_{f,r}}}{\partial \bar{l}} < 0 \right)$. ■

F.4 Equilibrium Profits of the Elite Firm

Proof of Proposition E3. The profit of the elite firm is $\Pi_e = z_e (\theta \bar{l} + l_{f,e})^\alpha k_e^\beta - w l_{f,e} - r k_e$. Solving for the equilibrium level, plugging $l_{f,e}$ in Π_e :

$$\Pi_e = z_e^{\frac{1-\beta}{1-\alpha-\beta}} \left(\frac{w}{\alpha} \right)^{\frac{\alpha(\beta-1)}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\alpha\beta}{1-\alpha-\beta}} k_e^\beta - w \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} + \theta \bar{l} w - r k_e$$

Plugging k_e in Π_e :

$$\begin{aligned} \Pi_e &= \left(\frac{w}{\alpha} \right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} - w \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} + \theta \bar{l} w \\ &\quad - r \left(\frac{w}{\alpha} \right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{1-\alpha}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} \end{aligned}$$

Defining parts of the equation:

$$(1) \equiv \left(\frac{w}{\alpha} \right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}}$$

$$(2) \equiv -w \left(\frac{w}{\alpha} \right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}}$$

$$(3) \equiv \theta \bar{l} w$$

$$(4) \equiv -r \left(\frac{w}{\alpha} \right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r} \right)^{\frac{1-\alpha}{1-\alpha-\beta}} z_e^{\frac{1}{1-\alpha-\beta}}$$

Plugging w in Π_e for all parts of the equation:

$$\begin{aligned}
(1) &\equiv (\bar{L} + \theta\bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{\alpha}{\beta-1}} \\
(2) &\equiv -\alpha (\bar{L} + \theta\bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{\alpha}{\beta-1}} z_e^{\frac{1}{1-\alpha-\beta}} \\
(3) &\equiv \alpha\theta\bar{l} (\bar{L} + \theta\bar{l})^{\frac{1-\alpha-\beta}{\beta-1}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{1-\alpha-\beta}{1-\beta}} \\
(4) &\equiv -r (\bar{L} + \theta\bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{1}{1-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{\alpha}{\beta-1}} z_e^{\frac{1}{1-\alpha-\beta}}
\end{aligned}$$

Simplifying $\Pi_e = (1) + (2) + (3) + (4)$ becomes:

$$\begin{aligned}
\Pi_e &= (\bar{L} + \theta\bar{l})^{\frac{\alpha}{1-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{\alpha}{\beta-1}} \left(\frac{\beta}{r}\right)^{\frac{1}{1-\beta}} \left(\frac{r}{\beta}(1-\alpha-\beta)\right) \\
&\quad + \alpha\theta\bar{l} (\bar{L} + \theta\bar{l})^{\frac{1-\alpha-\beta}{\beta-1}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{1-\alpha-\beta}{1-\beta}} \tag{F.9}
\end{aligned}$$

The impact of capacity constraint on equilibrium profits of the elite firm is positive ($\frac{\partial\Pi_e}{\partial\bar{l}} > 0$). The first part of equation F.9 demonstrates an increasing trend with respect to \bar{l} . Rewriting the second part of equation F.9:

$$\begin{aligned}
&= \alpha\theta\bar{l} (\bar{L} + \theta\bar{l})^{\frac{\alpha}{1-\beta}} \frac{1}{(\bar{L} + \theta\bar{l})} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{1-\alpha-\beta}{1-\beta}} \\
&= \frac{\theta\bar{l}}{(\bar{L} + \theta\bar{l})} \underbrace{(\bar{L} + \theta\bar{l})^{\frac{\alpha}{1-\beta}}}_{\uparrow \text{ in } \bar{l}} \alpha \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{1-\alpha-\beta}{1-\beta}}
\end{aligned}$$

To see that $\frac{\theta\bar{l}}{(\bar{L} + \theta\bar{l})}$ is also increasing in \bar{l} , applying the product rule:

$$\begin{aligned}
&\underbrace{(\theta(\bar{L} + \theta\bar{l}) - \theta\theta\bar{l})}_{= \theta\bar{L} > 0 \Rightarrow \uparrow \text{ in } \bar{l}} \underbrace{\left(\frac{1}{(\bar{L} + \theta\bar{l})^2}\right)}_{\uparrow \text{ in } \bar{l}}
\end{aligned}$$

Therefore, it holds that ($\frac{\partial\Pi_e}{\partial\bar{l}} > 0$).

F.5 Equilibrium Profits of the Regular Firm

This section provides the proof of Proposition E4. The profit of the regular firm is $\Pi_r = z_r l_{f,r}^\alpha k_r^\beta - w l_{f,r} - r k_r$. Solving for the equilibrium level, plugging $l_{f,r}$ in Π_r :

$$\Pi_r = z_r^{\frac{1-\beta}{1-\alpha-\beta}} \left(\frac{w}{\alpha}\right)^{\frac{\alpha(\beta-1)}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\alpha\beta}{1-\alpha-\beta}} k_r^\beta - w \left(\frac{w}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} - r k_r$$

Plugging k_r in Π_r :

$$\begin{aligned} \Pi_r &= \left(\frac{w}{\alpha}\right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} - w \left(\frac{w}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \\ &\quad - r \left(\frac{w}{\alpha}\right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{1-\alpha}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \end{aligned}$$

Defining parts of the equation:

$$\begin{aligned} (1) &\equiv \left(\frac{w}{\alpha}\right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \\ (2) &\equiv -w \left(\frac{w}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \\ (3) &\equiv -r \left(\frac{w}{\alpha}\right)^{\frac{-\alpha}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{1-\alpha}{1-\alpha-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \end{aligned}$$

Plugging w in Π_r for all parts of the equation:

$$\begin{aligned} (1) &\equiv (\bar{L} + \theta\bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{\alpha}{\beta-1}} \\ (2) &\equiv -\alpha (\bar{L} + \theta\bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{\alpha}{\beta-1}} \\ (3) &\equiv -r (\bar{L} + \theta\bar{l})^{\frac{\alpha}{1-\beta}} \left(\frac{\beta}{r}\right)^{\frac{1}{1-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{\alpha}{\beta-1}} \end{aligned}$$

Simplifying $\Pi_r = (1) + (2) + (3)$ becomes:

$$\Pi_r = (\bar{L} + \theta\bar{l})^{\frac{\alpha}{1-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{\alpha}{\beta-1}} \left(\frac{\beta}{r}\right)^{\frac{1}{1-\beta}} \left(\frac{r}{\beta}(1-\alpha-\beta)\right) \quad (\text{F.10})$$

The impact of capacity constraint on equilibrium profits of the regular firm is positive ($\frac{\partial \Pi_r}{\partial \bar{l}} > 0$):

$$\frac{\partial \Pi_r}{\partial \bar{l}} = \frac{\theta \alpha}{1 - \beta} (\bar{L} + \theta \bar{l})^{\frac{\alpha}{1-\beta}-1} z_r^{\frac{1}{1-\alpha-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right)^{\frac{\alpha}{\beta-1}} \left(\frac{\beta}{r} \right)^{\frac{1}{1-\beta}} \left(\frac{r}{\beta} (1 - \alpha - \beta) \right). \blacksquare$$

F.6 Capital–Labor Cost Ratio

The capital-to-labor cost ratio for the region is denoted as $CI \equiv \frac{rk_e + rk_r}{wL + w_c l}$. Since $w_c = 0$, the ratio simplifies to:

$$CI \equiv \frac{rk_e + rk_r}{w\bar{L}}.$$

Proof of Proposition E4. The capital-to-labor cost ratio rises with the convict labor capacity constraint, i.e., $\frac{\partial CI}{\partial \bar{l}} > 0$. We deduce this from the fact that as \bar{l} increases, both k_e and k_r increase and w decreases. \blacksquare

F.7 Union Membership and Strikes

Benefit of workers from striking: Should the strike succeed, the value or benefit (Ω) accrued by the employee is

$$\Omega = (w(l_c = 0) - w(l_c = \bar{l})) \bar{L} \tag{F.11}$$

Probability of workers joining a union and striking: Assuming a strike at an elite firm can be prevented with probability $1 - p$ or succeed with p , successful strikes lead to the elimination of convict labor and unsuccessful ones to a utility loss, δ . Free labor will strike if:

$$J < p\Omega - (1 - p)\delta = p((w(l_c = 0) - w(l_c = \bar{l})) \bar{L} - (1 - p)\delta) \tag{F.12}$$

Proof of Proposition E5. The right-hand side of equation (F.12) increases in correlation with \bar{l} :

$$J < p \underbrace{\left(\underbrace{w(l_c = 0)}_{\frac{\partial w(l_c=0)}{\partial \bar{l}}=0} - \underbrace{w(l_c = \bar{l})}_{\frac{\partial w(l_c=\bar{l})}{\partial \bar{l}} < 0} \right)}_{\uparrow \text{ in } \bar{l}} \bar{L} - (1 - p)\delta$$

To see that $(w(l_c = 0) - w(l_c = \bar{l}))$ is increasing in \bar{l} , let's first plug in $w(l_c = 0)$ and $w(l_c = \bar{l})$:

$$w(l_c = 0) - w(l_c = \bar{l}) = \alpha \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right)^{\frac{1-\alpha-\beta}{1-\beta}} \left(\bar{L}^{\frac{1-\alpha-\beta}{\beta-1}} - (\bar{L} + \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}} \right)$$

Taking the derivative of this expression w.r.t. \bar{l} , we find:

$$\frac{\partial w(l_c = 0) - w(l_c = \bar{l})}{\partial \bar{l}} = -\frac{1 - \alpha - \beta}{\beta - 1} \theta \alpha \left(\frac{\beta}{r} \right)^{\frac{\beta}{1-\beta}} \left(z_e^{\frac{1}{1-\alpha-\beta}} + z_r^{\frac{1}{1-\alpha-\beta}} \right)^{\frac{1-\alpha-\beta}{1-\beta}} (\bar{L} + \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}-1} > 0$$

Therefore, an increase in convict labor capacity results in increased levels of union membership and strike occurrences. ■

G Further Quantitative Analysis

G.1 Calibration

The calibration exercise seeks to evaluate the impact of convict labor on the labor market dynamics and assess the counterfactual scenario in which unions successfully eliminate the use of convict labor. Furthermore, this section demonstrates that when calibrated, my model aligns with the moments observed in the data related to the average wage and capital-labor cost ratio in response to the impact of convict labor capacity.

The parameters of the model are α , β , θ , r , \bar{l} , \bar{L} , and $z_e = z_r = z$. I set $z_e = z_r = z$ because differences in productivity between elite and regular firms are indiscernible in the data. Productivity of convict labor, θ , comes from the [U.S. Bureau of Labor \(1925\)](#), reporting the number of convict laborers and the number of free laborers needed to complete the same amount of work. The capital rental rate, r , is set as 0.0919, consistent with [Bodenhorn and Rockoff \(1992\)](#).⁴⁸ The number of free laborers is normalized to 1, whereas the quantity

⁴⁸Utilizing Table 5.9 from [Bodenhorn and Rockoff \(1992\)](#), which presents postbellum interest rates in the Southern United States, I calculate the average for 1870-1991 and 1891-1904, aligning with my analysis period in Section 5.2.

Table G1: Calibration

Parameter	Description	Value
α	Share of labor in production	0.194
β	Share of capital in production	0.766
θ	Productivity of convict labor	0.7
r	Rental rate of capital	0.0919
\bar{l}	Convict laborers	0.16
\bar{L}	Free laborers	1
$z_e = z_r = z$	Firm productivity	1

Note: The model’s parameters include α , β , θ , r , \bar{l} , \bar{L} , and $z_e = z_r = z = 1$. I sourced convict labor productivity, θ , from [U.S. Bureau of Labor \(1925\)](#). Main industry’s employment share, γ , matches the causal coefficient estimate. I set the capital rental rate, r , at 0.05 based on [Bodenhorn and Rockoff \(1992\)](#), normalized free laborers to 1, and aligned convict labor quantities to the observed convict-to-free labor ratio. I use α and β to target the two moments observed in the data.

of convict laborers in exposed counties is chosen to match the convict-to-free labor ratio, $\frac{\bar{l}}{\bar{L}}$, observed in the data presented in Section 3.

Calibration Strategy and Model-Data Match. I target two key moments in the data: the contraction in free labor wages and the rise in capital-labor cost ratio - in response to an average increase in convict labor observed in exposed counties. To achieve exact identification with the remaining three parameters and the two target moments, I fix $z = 1$. The remaining parameters — the labor and capital shares in production, α and β — are set at 0.194 and 0.766, respectively, to match the model with the two target moments:

$$m(\lambda) = \begin{bmatrix} \frac{w(\alpha, \beta, l_c = \bar{l})}{w(\alpha, \beta, l_c = 0)} - \frac{w^{exposed}}{w^{overall}} \\ \frac{CI(\alpha, \beta, l_c = \bar{l})}{CI(\alpha, \beta, l_c = 0)} - \frac{CI^{exposed}}{CI^{overall}} \end{bmatrix}$$

The objective function can be expressed as follows, with W denoting the identity matrix:

$$\min_{\lambda} (m(\lambda)'W^{-1}m(\lambda)) \tag{G.1}$$

To identify the target moments, I sourced average wage and capital-labor cost ratio data from [Haines et al. \(2005\)](#) for all counties in 1870 and calculated their average growth rates using the same dataset. I then utilize the causal coefficient estimates from Section 5.2 to determine the change in these variables in counties impacted by convict labor’s introduction.⁴⁹

⁴⁹For more details on the calibration strategy, see the Supp Appendix Section G.2.

Table G2: Model-Data Match

Variable	Model	Data
$\frac{w_{exposed}}{w_{overall}}$	0.982	0.982
$\frac{CI_{exposed}}{CI_{overall}}$	1.11	1.04
$\frac{\Pi_{exposed}^e}{\Pi_{overall}^e}$	1.35	–
$\frac{\Pi_{exposed}^r}{\Pi_{overall}^r}$	1.07	–
$\frac{\left(\frac{k_e}{l_{f,e}}\right)_{exposed}}{\left(\frac{k_e}{l_{f,e}}\right)_{overall}}$	1.23	–
$\frac{\left(\frac{k_r}{l_{f,r}}\right)_{exposed}}{\left(\frac{k_r}{l_{f,r}}\right)_{overall}}$	0.98	–

Note: The table displays ratios comparing the post-introduction convict labor effects in exposed counties to all pre-exposure period. Drawing from Haines et al. (2005), I determined 1870 averages and growth rates, evaluated convict labor’s impact with Section 5.2 coefficients, and estimated absent variables using calibrated parameters Table G1.

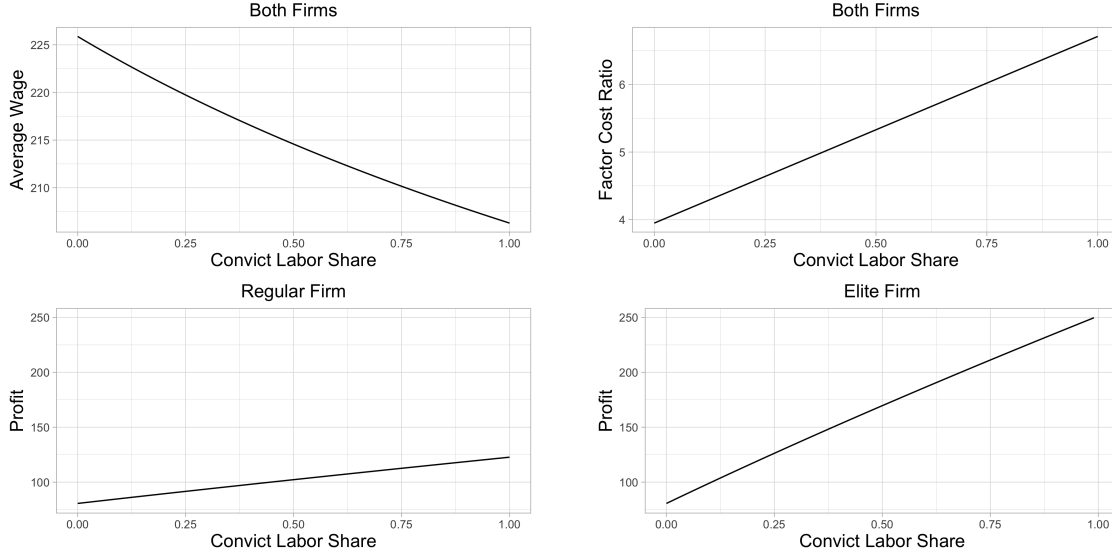
Table G1 reports the list of the model parameters and their values used in the calibration. The initial two rows of Table G2 show the model’s performance matching the two target moments. Evident from these rows, the model closely matches the data, demonstrating its capability to represent the two pivotal effects of convict labor on wage and capital–labor cost ratio. These effects are fundamental to my empirical analysis in Section 5.2. Later rows in Table G2 show how exposure to convict labor amplifies the regular firm’s physical labor intensity, increases the elite firm’s physical capital intensity, and increases profits for both firm types.

Additionally, the first calibrated metric regarding the average wage illustrates how the model aligns with an untargeted moment: the average wage rate of free labor before the rise of convict labor. The data, described in Section 3, indicates an annual wage of \$218 prior to the introduction of convict labor. In comparison, the model predicts an annual wage of \$221, offering a form of model validation. Second, the model corroborates another untargeted moment. It predicts the labor share, α , at 0.194, closely matching the empirical data figure of $\alpha = 0.18$.⁵⁰ This not only validates the model but also confirms its internal consistency.

Figure G1 illustrates the dynamics of key model variables over various levels of convict labor employment. This figure highlights several trends: With the rise in convict labor use,

⁵⁰I calculate the labor share by dividing workers’ total wages by the manufacturing industry’s output value, as detailed in Section 3.

Figure G1. Calibrated Metrics: Average Wage, Factor Cost Ratio, and Profits



Note: The figure illustrates the variations in per-worker wage, capital-labor cost ratio for both firms, physical capital intensity for elite and regular firms, and profits for both firms as the capacity constraint of convict labor shifts, based on equations E.8, E.11, E.9, E.10.

there is a decrease in that of elite firms. Additionally, the capital-labor cost ratio in the county and profits for both types of firms increase. Although both elite and regular firms see profit gains, the increase is significantly more pronounced for elite firms. This trend shows each firm type’s benefits from the growing use of convict labor. Crucially, there is a noticeable decline in wages per worker.

G.2 Further Details on the Calibration Strategy

To identify the target moments, I sourced and calculated the average growth rates of average wage and capital-labor cost ratios for all counties in 1870 from Haines et al. (2005). Then, using the causal coefficient estimates from Section 5.2, I determined the changes in these variables for counties impacted by the introduction of convict labor. Specifically, I employ the causal estimates detailed in the Supplementary Appendix Table K34 to calculate the ratios of $\frac{w_{exposed}}{w_{overall}} = 0.982$ and $\frac{CI_{exposed}}{CI_{overall}} = 1.04$. To achieve the decline in wage level and capital-labor cost ratio level, I follow the steps below:

Establishing baseline levels: The initial step is to determine the baseline levels for wages and capital-labor cost ratio before the introduction of convict labor. These are sourced from the county Census (Haines et al., 2005).

Calculating overall growth rates: After establishing the baseline levels, I calculate the overall growth rates in wages and capital-labor cost ratio. These growth rates reflect the

general trends in wages and capital-labor cost ratio before the introduction of convict labor, as derived from the county Census (Haines et al., 2005). Then, I apply these growth rates to the initial baseline wage and capital-labor cost ratio to find the wage and capital-labor cost ratio with growth rates, denoted as $w_{overall}$ and $CI_{overall}$.

Adjusting for convict labor impact: Next, I apply the causal coefficient estimates of the decline in the wage and capital-labor cost ratio growth from the Supplementary Appendix Table K34 to the growth rates before the introduction of convict labor. This involves adjusting the baseline growth rates to account for the impact of convict labor. The adjusted rates represent the change in wage growth and capital-labor cost ratio growth due to the introduction of convict labor. Applying these growth rates to the baseline wages and capital-labor cost ratio results in $w_{exposed}$ and $CI_{exposed}$.

Deriving ratios: Finally, I calculate the ratios $\frac{w_{exposed}}{w_{overall}} = 0.982$ and $\frac{CI_{exposed}}{CI_{overall}} = 1.04$. I jointly optimize the loss function for both ratios in calibrating the model.

H Constant Elasticity of Substitution

This model has a representative firm, employing convict and free labor. Convict labor is not paid, while free labor is ($w_f = w > 0$). The firm has a capacity limit for convict labor, $l_{c,e} \leq \bar{l}$.

The production function and the maximization problem:

$$\begin{aligned}
 Y &= (\eta \bar{l}^\rho + (1 - \eta) l_f^\rho)^{\frac{\alpha}{\rho}} & (H.1) \\
 \max & (\eta \bar{l}^\rho + (1 - \eta) l_f^\rho)^{\frac{\alpha}{\rho}} - w l_f \\
 \frac{\partial}{\partial l_f} & : \alpha (\eta \bar{l}^\rho + (1 - \eta) l_f^\rho)^{\frac{\alpha}{\rho} - 1} (1 - \eta) l_f^{\rho - 1} - w
 \end{aligned}$$

Wage becomes:

$$\begin{aligned}
 w &= \alpha Y^{\frac{\alpha - \rho}{\rho}} (1 - \eta) l_f^{\rho - 1} \\
 w &= \alpha (\eta \bar{l}^\rho + (1 - \eta) l_f^\rho)^{\frac{\alpha - \rho}{\rho}} (1 - \eta) l_f^{\rho - 1} & (H.2)
 \end{aligned}$$

Taking the logarithm:

$$\log w = \log(\alpha) + \frac{\alpha - \rho}{\rho} \log(\eta \bar{l}^\rho + (1 - \eta) l_f^\rho) + \log(1 - \eta) + (\rho - 1) \log(l_f)$$

Differentiate w.r.t. \bar{l} :

$$\frac{\partial \log w}{\partial \bar{l}} = \frac{\eta(\alpha - \rho)\bar{l}^{\rho-1}}{\eta(\bar{l}^\rho - l_f^\rho) + l_f^\rho} \quad (\text{H.3})$$

One can estimate the elasticity of substitution parameter, ρ , in the following way:

$$\begin{aligned} \log w &= \log(\alpha) + \frac{\alpha - \rho}{\rho} \log(\eta\bar{l}^\rho + (1 - \eta)l_f^\rho) \\ &\quad + \log(1 - \eta) + (\rho - 1) \log(l_f) \\ \frac{\partial \log w}{\partial \log Y} &= \left(\frac{\alpha - \rho}{\rho} \right) \end{aligned} \quad (\text{H.4})$$

Once $\frac{\partial \log w}{\partial \log Y}$ and α from the data are known, one can get a unique ρ from equation (H.4), since $\left(\frac{\alpha - \rho}{\rho}\right)$ is monotone in ρ . The estimation of ρ is carried out independently of η , simplifying the process as η is not necessary to determine ρ .

Calibration. First, I find $\alpha = 0.18$ by dividing the total wage by manufacturing output and taking the average for the years in the analysis (Haines et al., 2005). Then, I regress average wage growth on $\log(\text{manufacturing output})$ and $\log(\text{convict labor share})$. In this regression specification, I control for the convict labor share because the feasibility of estimating the CES-parameter ρ is conditional on convict labor being present in the area (i.e., $\bar{l} > 0$). We can note that the equation estimating the CES parameter would change, and ρ would altogether drop from the framework - leading to a standard Lucas-span-of-control production with l_f^α if convict labor is not available for the firms (i.e. if $\bar{l} = 0$).

The coefficient estimate β from the regression equation, although statistically insignificant, equals -0.8 , which gives the left-hand side of equation (H.4). Plugging these values in:

$$-0.8 = \left(\frac{0.18 - \rho}{\rho} \right)$$

which gives ρ equal to 0.9.

The estimated value of ρ is close to 1, suggesting a high degree of substitutability between convict and free labor. However, it is important to interpret this result cautiously due to the limitations of the available data. The proximity of ρ to 1 implies that convict labor and free labor can be considered nearly perfectly substitutable within the context of this estimation. The model presented in the main body is derived from, and is in accordance with, this estimation. It is important to note that since ρ is greater than α , the predictions of this model will be qualitatively the same as the model in the main text.

I Model Extension: Firm Type Distribution

In this section, I extend the model to allow for a general firm-type distribution for elite and regular firms: η to capture elite and $1 - \eta$ to capture regular and study the implications of η on $\partial w / \partial \bar{l}$. The equilibrium labor demand $l_{f,e}$ and $l_{f,r}$ remain the same as Section E. Combining $l_{f,e}$ and $l_{f,r}$, and firm weights η and $(1 - \eta)$ to find the total free-labor demand and equating it to the total labor supply of free-laborers, \bar{L} , characterizes the general equilibrium of the model:

$$\bar{L} = \left(\frac{w}{\alpha}\right)^{\frac{\beta-1}{1-\alpha-\beta}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} \left(\eta z_e^{\frac{1}{1-\alpha-\beta}} + (1-\eta) z_r^{\frac{1}{1-\alpha-\beta}}\right) - \eta \theta \bar{l} \quad (\text{I.1})$$

Wage rate of free-laborers, w :

$$w = \alpha \left(\frac{\bar{L} + \eta \theta \bar{l}}{\left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\alpha-\beta}} \left(\eta z_e^{\frac{1}{1-\alpha-\beta}} + (1-\eta) z_r^{\frac{1}{1-\alpha-\beta}}\right)} \right)^{\frac{1-\alpha-\beta}{\beta-1}} \quad (\text{I.2})$$

The effect of capacity constraint on equilibrium wage rate is negative, $\frac{\partial w}{\partial \bar{l}} < 0$:

$$\frac{\partial w}{\partial \bar{l}} = \frac{\eta \theta (-\alpha - \beta + 1)}{\beta - 1} (\bar{L} + \eta \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}-1} \frac{\alpha}{\left(\frac{\beta}{r}\right)^{\frac{\beta}{\beta-1}} \left(\eta z_e^{\frac{1}{1-\alpha-\beta}} + (1-\eta) z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{1-\alpha-\beta}{\beta-1}}}$$

When $\eta = 1$ —all firms are elite—convict labor negatively affects wages. But when $\eta = 0$ —no elite firms—convict labor does not impact wages ($\frac{\partial w}{\partial \bar{l}} = 0$).

Profits of Elite and Regular Firms:

$$\begin{aligned} \Pi_e &= (\bar{L} + \eta \theta \bar{l})^{\frac{\alpha}{1-\beta}} z_e^{\frac{1}{1-\alpha-\beta}} \left(\eta z_e^{\frac{1}{1-\alpha-\beta}} + (1-\eta) z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{\alpha}{\beta-1}} \left(\frac{\beta}{r}\right)^{\frac{1}{1-\beta}} \left(\frac{r}{\beta}(1-\alpha-\beta)\right) \\ &\quad + \alpha \theta \bar{l} (\bar{L} + \eta \theta \bar{l})^{\frac{1-\alpha-\beta}{\beta-1}} \left(\frac{\beta}{r}\right)^{\frac{\beta}{1-\beta}} \left(\eta z_e^{\frac{1}{1-\alpha-\beta}} + (1-\eta) z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{1-\alpha-\beta}{1-\beta}} \end{aligned} \quad (\text{I.3})$$

The impact of capacity constraint on equilibrium profits of elite firms is positive ($\frac{\partial \Pi_e}{\partial \bar{l}} > 0$).

$$\Pi_r = (\bar{L} + \eta \theta \bar{l})^{\frac{\alpha}{1-\beta}} z_r^{\frac{1}{1-\alpha-\beta}} \left(\eta z_e^{\frac{1}{1-\alpha-\beta}} + (1-\eta) z_r^{\frac{1}{1-\alpha-\beta}}\right)^{\frac{\alpha}{\beta-1}} \left(\frac{\beta}{r}\right)^{\frac{1}{1-\beta}} \left(\frac{r}{\beta}(1-\alpha-\beta)\right) \quad (\text{I.4})$$

The impact of capacity constraint on equilibrium profits of the regular firms is positive ($\frac{\partial \Pi_e}{\partial t} > 0$).

Union Membership and Strikes: The union and strike aspect of the model is unchanged. As previously, expanding convict labor capacity leads to more union memberships and strikes.

J Additional Data

To account for the possibility that counties dependent on prison labor were fundamentally different from others, I control for trends in pre-existing differences in 1860 that might be related to the growth of prison labor. The baseline set of control variables includes population density, the share of enslaved individuals, the average value of agricultural output per acre of farmland, total improved acres of farmland, proximity to the coast, elevation, access to railway transportation and navigable waterways, distance to the nearest urban center with a population over 100,000, and the Democratic Party's vote share. Using United States shapefiles, I calculate the distance to the coast and the nearest urban center. The remaining control variables are obtained from the decennial Census records from the Integrated Public Use Microdata Series and the Historical, Demographic, Economic, and Social Data (ICPSR 2896) spanning the period 1850-1940 (Haines et al., 2005).

Average Wage Growth in Manufacturing: Considering the lack of direct data on county-level workers' wages, I adopt a different strategy to estimate the average manufacturing wage. This technique calculates the average wage by dividing the annual wage bill in manufacturing by the estimated number of sector workers. The data sources include ICPSR 2896 (Haines et al., 2005) for county-level data and Hornbeck and Rotemberg (2021) for county-industry-level data from 1860-1880.

Capital-Labor Cost Share: I derive the capital-labor cost ratio using county-level data from the ICPSR 2896 (Haines et al., 2005). The annual manufacturing capital spending represents the capital cost, and the annual wage bill in manufacturing signifies the labor cost. I compute the $\frac{K}{K+L}$ ratio from these values.

Election Outcomes: The election outcomes are sourced from the ICPSR 08611 (Clubb et al., 1987). The vote share of the Democratic Party in both presidential and congressional elections represents the percentage of votes they received in a particular election. I compute the win margin by subtracting the combined vote share of the other parties from the Democratic Party's vote share. Additionally, I define a dummy variable 'win' to depict the election outcome. This variable takes a value of 1 if the Democratic Party secures a vote

share exceeding 50 percent, indicating a win.

Farming Outcomes: In the analysis, I examine several farming outcomes including changes in farm value, farm equipment value, farm output value, number of farms, number of farm tenants and sharecroppers, and acres of farmland. For each outcome, I compute the change relative to the previous census year. This data comes from the ICPSR 2896 (Haines et al., 2005).

Manufacturing Employment Share: This measure represents the proportion of individuals employed in manufacturing out of the total number of employed individuals from the ICPSR 2896 (Haines et al., 2005).

Manufacturing Establishments: I sourced the data on the number of manufacturing establishments from ICPSR 2896 (Haines et al., 2005). This variable indicates the total count of manufacturing establishments in a specific county for each census year. The data for the 1910 census is not available.

Manufacturing Output Value: The information on the value of manufacturing output is obtained from ICPSR 2896 (Haines et al., 2005). This variable presents the cumulative value of manufacturing output in a county for each census year. The 1910 data is missing. For the balance table, I calculate the manufacturing output value per establishment. This is achieved by dividing the total manufacturing output value by the total number of manufacturing establishments within a county.

Migration Outcomes: For calculating migration outcomes, I utilize the linked full count census (Helgertz et al., 2023; Ruggles et al., 2021). Initially, I identify individuals who have changed their county of residence between consecutive census years. This data is then aggregated at the county level to compute the total in and out migration for the specific county. The linked full count census also includes demographic variables such as race and industry of occupation for each individual. To determine the total number of white individuals and those employed in the agriculture and manufacturing industries who have migrated, I perform the same calculation on these specific subsets of the population.

Occupation Outcomes: To quantify the shift in labor distribution, I use the linked full-count census data, specifically focusing on the occupation variable, occ1950 (Helgertz et al., 2023; Ruggles et al., 2021). According to this variable, values ranging from 500-595 indicate the individual is a craftsman, 600-690 suggest they are operatives, and 910-970 point to laborers. To compute the count of new craftsmen and operatives, I look for individuals who held occupations other than craftsman, operative, and laborer in the prior census year, and transitioned to either craftsman or operative in the subsequent year. Similarly, for determining the count of individuals transitioning from laborer roles to craft and operative

occupations, I identify those who were laborers in the previous census year and transitioned to craftsman or operative roles in the following year. Next, I aggregate these values at the county level. This gives me the total number of individuals who either transitioned into craftsman or operative roles or shifted from being laborers to either of these occupations within the county.

K Supplementary Tables

Table K1: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership (IHS)	Membership 2020 (IHS)
	(1)	(2)	(3)	(4)
Convict Labor (IHS)	0.429*** (0.154)	0.204*** (0.075)	0.081*** (0.027)	0.196*** (0.029)
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,525
Clusters	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (9). The observation unit is the county (1880), with convict labor presence (number of convicts employed - IHS) as the independent variable. The outcome variables are the count of local KOL assemblies, strikes, IHS-transformed 19th-century Knights of Labor members and current union members. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K2: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership (IHS)	Membership 2020 (IHS)
	(1)	(2)	(3)	(4)
Number of Prisons (IHS)	2.233** (1.036)	1.122** (0.568)	0.288*** (0.112)	0.818*** (0.149)
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,525
Clusters	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (9). The observation unit is the county (1880), with the number of prisons that employ convicts (IHS) as the independent variable. The outcome variables are the count of local KOL assemblies, strikes, IHS-transformed 19th-century Knights of Labor members and current union members. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K3: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership	Membership 2020
	(1)	(2)	(3)	(4)
Convict Labor per 100,000 (IHS)	0.236*** (0.091)	0.108*** (0.039)	8.138 (6.331)	1,189*** (272.175)
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (9). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. The outcome variables are the count of local KOL assemblies, strikes, 19th-century Knights of Labor members and current union members. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K4: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership (IHS)	Membership 2020 (IHS)
	(1)	(2)	(3)	(4)
Convict Labor per 100,000 (IHS)	0.236*** (0.087)	0.108*** (0.038)	0.050*** (0.018)	0.144*** (0.023)
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (9). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. The outcome variables are the count of local KOL assemblies, strikes, IHS-transformed 19th-century Knights of Labor members and current union members. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects and Conley standard errors.

Table K5: First Stage: Effect on Convict Labor

	<i>Dependent variable:</i>					
	Convict Labor (IHS)					
	(1)	(2)	(3)	(4)	(5)	(6)
Early Prison Dummy	5.915*** (0.498)	5.916*** (0.497)	5.896*** (0.499)			
Early Prison Capacities (IHS)				0.836*** (0.054)	0.836*** (0.055)	0.834*** (0.055)
Economic Controls	×	×	×	×	×	×
Geographic Controls		×	×		×	×
Socio-political Controls			×			×
State × Year FE	×	×	×	×	×	×
Mean Outcome	15.74	15.74	15.74	15.74	15.74	15.74
Clusters	1,302	1,302	1,302	1,302	1,302	1,302
F-stat	105.01	95.85	94.45	110.4	100.78	99.3
Observations	3,906	3,906	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (10). The observation unit is the county (1880), with convict labor presence (the number of convicts employed - IHS) as the dependent variable. Independent variables include pre-convict labor prison locations and capacities. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The regression includes state × year fixed effects, with standard errors clustered at the county level.

Table K6: First Stage: Effect on Prisons

	<i>Dependent variable:</i>					
	Number of Prisons (IHS)					
	(1)	(2)	(3)	(4)	(5)	(6)
Early Prison Dummy	0.884*** (0.077)	0.884*** (0.077)	0.878*** (0.077)			
Early Prison Capacities (IHS)				0.125*** (0.009)	0.125*** (0.009)	0.125*** (0.009)
Economic Controls	×	×	×	×	×	×
Geographic Controls		×	×		×	×
Socio-political Controls			×			×
State × Year FE	×	×	×	×	×	×
Mean Outcome	0.12	0.12	0.12	0.12	0.12	0.12
Clusters	1,302	1,302	1,302	1,302	1,302	1,302
F-stat	97.45	89.03	87.99	99.98	91.35	90.27
Observations	3,906	3,906	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (10). The observation unit is the county (1880), with the number of prisons that employ convict labor as the dependent variable. Independent variables include pre-convict labor prison locations and capacities. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The regression includes state × year fixed effects, with standard errors clustered at the county level.

Table K7: Effect on KOL Assemblies 1886-1905

	<i>Dependent variable:</i>									
	Number of KOL Assemblies									
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Convict Labor per 100,000 (IHS)	1.417** (0.553)	1.421** (0.552)	1.232*** (0.434)	1.232*** (0.433)	1.213*** (0.427)	1.209*** (0.426)	1.176*** (0.423)	1.166*** (0.421)	1.154*** (0.417)	1.166*** (0.418)
KP F-stat	122.77	125.81	128.3	128.85	130.18	129.35	129.16	129.49	129.3	126.59
Panel B: Early Prison Capacities IV										
Convict Labor per 100,000 (IHS)	1.370** (0.562)	1.376** (0.563)	1.187*** (0.443)	1.187*** (0.442)	1.169*** (0.436)	1.166*** (0.436)	1.134*** (0.435)	1.125*** (0.434)	1.114*** (0.430)	1.126*** (0.431)
KP F-stat	187.15	189.9	191.82	192.25	193.33	192.17	191.69	191.86	191.08	186.55
Population Density	×	×	×	×	×	×	×	×	×	×
Enslaved Population Share		×	×	×	×	×	×	×	×	×
Farm Output Value per Acre of Farmland		×	×	×	×	×	×	×	×	×
Number of Improved Acres of Farmland				×	×	×	×	×	×	×
Distance to Coast					×	×	×	×	×	×
Elevation						×	×	×	×	×
Access to Railways							×	×	×	×
Access to Navigable Water Bodies								×	×	×
Distance to the Nearest Urban Center									×	×
Vote Share for the Democratic Party										×
State × Year FE	×	×	×	×	×	×	×	×	×	×
Mean Outcome	1.23	1.23	1.23	1.23	1.23	1.23	1.23	1.23	1.23	1.23
Clusters	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variable is the number of local KOL assemblies. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K8: Effect on Strikes 1886-1905

	<i>Dependent variable:</i>									
	Number of Strikes									
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Panel A: Early Prison Dummy IV										
Convict Labor per 100,000 (IHS)	0.603** (0.286)	0.605** (0.286)	0.468** (0.206)	0.466** (0.206)	0.461** (0.204)	0.461** (0.204)	0.453** (0.204)	0.448** (0.204)	0.449** (0.202)	0.450** (0.201)
KP F-stat	122.77	125.81	128.3	128.85	130.18	129.35	129.16	129.49	129.3	126.59
Panel B: Early Prison Capacities IV										
Convict Labor per 100,000 (IHS)	0.635** (0.318)	0.638** (0.318)	0.504** (0.238)	0.503** (0.238)	0.497** (0.235)	0.497** (0.236)	0.490** (0.236)	0.487** (0.236)	0.487** (0.234)	0.488** (0.234)
KP F-stat	187.15	189.9	191.82	192.25	193.33	192.17	191.69	191.86	191.08	186.55
Population Density	×	×	×	×	×	×	×	×	×	×
Enslaved Population Share		×	×	×	×	×	×	×	×	×
Farm Output Value per Acre of Farmland		×	×	×	×	×	×	×	×	×
Number of Improved Acres of Farmland				×	×	×	×	×	×	×
Distance to Coast					×	×	×	×	×	×
Elevation						×	×	×	×	×
Access to Railways							×	×	×	×
Access to Navigable Water Bodies								×	×	×
Distance to the Nearest Urban Center									×	×
Vote Share for the Democratic Party										×
State × Year FE	×	×	×	×	×	×	×	×	×	×
Mean Outcome	0.28	0.28	0.28	0.28	0.28	0.28	0.28	0.28	0.28	0.28
Clusters	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variable is the number strikes. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K9: Effect on Early Membership

		<i>Dependent variable:</i>								
		Early Membership (IHS)								
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Convict Labor per 100,000 (IHS)	0.216*** (0.057)	0.216*** (0.057)	0.195*** (0.051)	0.195*** (0.051)	0.193*** (0.050)	0.192*** (0.050)	0.188*** (0.049)	0.186*** (0.049)	0.185*** (0.049)	0.187*** (0.049)
KP F-stat	122.77	125.81	128.3	128.85	130.18	129.35	129.16	129.49	129.3	126.59
Panel B: Early Prison Capacities IV										
Convict Labor per 100,000 (IHS)	0.209*** (0.055)	0.209*** (0.055)	0.188*** (0.049)	0.189*** (0.049)	0.186*** (0.049)	0.186*** (0.049)	0.182*** (0.048)	0.180*** (0.048)	0.179*** (0.048)	0.181*** (0.048)
KP F-stat	187.15	189.9	191.82	192.25	193.33	192.17	191.69	191.86	191.08	186.55
Population Density	×	×	×	×	×	×	×	×	×	×
Enslaved Population Share		×	×	×	×	×	×	×	×	×
Farm Output Value per Acre of Farmland		×	×	×	×	×	×	×	×	×
Number of Improved Acres of Farmland				×	×	×	×	×	×	×
Distance to Coast					×	×	×	×	×	×
Elevation						×	×	×	×	×
Access to Railways							×	×	×	×
Access to Navigable Water Bodies								×	×	×
Distance to the Nearest Urban Center									×	×
Vote Share for the Democratic Party										×
State × Year FE	×	×	×	×	×	×	×	×	×	×
Mean Outcome	20.83	20.83	20.83	20.83	20.83	20.83	20.83	20.83	20.83	20.83
Clusters	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variable is the IHS-transformed 19th-century KOL membership. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K10: Effect on Membership 2020

	<i>Dependent variable:</i>									
	Membership 2020 (IHS)									
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Convict Labor per 100,000 (IHS)	0.315*** (0.045)	0.314*** (0.046)	0.266*** (0.040)	0.269*** (0.040)	0.253*** (0.039)	0.250*** (0.039)	0.227*** (0.038)	0.223*** (0.039)	0.214*** (0.038)	0.225*** (0.040)
KP F-stat	122.59	125.69	128.16	128.7	130.08	129.21	129.03	129.36	129.21	126.29
Panel B: Early Prison Capacities IV										
Convict Labor per 100,000 (IHS)	0.309*** (0.041)	0.308*** (0.042)	0.260*** (0.037)	0.263*** (0.037)	0.248*** (0.037)	0.245*** (0.037)	0.223*** (0.036)	0.220*** (0.037)	0.211*** (0.036)	0.222*** (0.038)
KP F-stat	186.79	189.68	191.56	191.98	193.13	191.9	191.44	191.61	190.87	186.04
Population Density	×	×	×	×	×	×	×	×	×	×
Enslaved Population Share		×	×	×	×	×	×	×	×	×
Farm Output Value per Acre of Farmland		×	×	×	×	×	×	×	×	×
Number of Improved Acres of Farmland				×	×	×	×	×	×	×
Distance to Coast					×	×	×	×	×	×
Elevation						×	×	×	×	×
Access to Railways							×	×	×	×
Access to Navigable Water Bodies								×	×	×
Distance to the Nearest Urban Center									×	×
Vote Share for the Democratic Party										×
State × Year FE	×	×	×	×	×	×	×	×	×	×
Mean Outcome	3,542	3,542	3,542	3,542	3,542	3,542	3,542	3,542	3,542	3,542
Clusters	1,296	1,296	1,296	1,296	1,296	1,296	1,296	1,296	1,296	1,296
Observations	3,888	3,888	3,888	3,888	3,888	3,888	3,888	3,888	3,888	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variable is the IHS-transformed number of union members in 2020. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K11: Change Union Formation on Change Convict 1895-1905

	<i>Dependent variable:</i>					
	Change in Convict Labor 100K (IHS)					
	(1)	(2)	(3)	(4)	(5)	(6)
Change in KOL	0.003 (0.005)					
Change in KOL100K		-0.00000 (0.00002)				
Change in Strike			0.007 (0.013)			
Change in Strike100K				0.001 (0.001)		
Change in Member					0.0001 (0.0001)	
Change in Member100K						0.00002 (0.00003)
Controls	×	×	×	×	×	×
State x Year FE	×	×	×	×	×	×
Mean Outcome	30.41	30.41	30.41	30.41	30.41	30.41
Clusters	1302	1302	1302	1302	1302	1302
Observations	2,604	2,604	2,604	2,604	2,604	2,604

Note: Each column reports coefficients and standard errors (in parentheses) from a different regression. The observation unit is the county (1880), with change in convict labor presence (change in convicts per 100,000 inhabitants - IHS) as the dependent variable. Independent variables are the change in the number of KOL assemblies (and per 100,000 inhabitants), change in strikes (and per 100,000 inhabitants), and change in membership (and per 100,000 inhabitants). Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state \times year fixed effects, and standard errors are county-level clustered.

Table K12: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership	Membership 2020
	(1)	(2)	(3)	(4)
Convict Labor per 100,000 (IHS) \times Share of Enslaved Individuals	1.105** (0.453)	0.514** (0.256)	0.178*** (0.048)	0.221*** (0.037)
Controls	\times	\times	\times	\times
State \times Year FE	\times	\times	\times	\times
IV: Early Prison Dummy + Capacities	\times	\times	\times	\times
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
KP F-stat	100.16	100.16	100.16	99.8
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from a regression analogous to equation (11). The observation unit is the county (1880), with the IHS-transformed sum of the interaction of share of enslaved individuals and the number of convict laborers per 100,000 inhabitants and the sum of these two as the independent variable. It is instrumented by the sum of pre-convict labor prisons and their capacities. Outcome variables are the number of local KOL assemblies, strikes, 19th-century Knights of Labor members and current union members. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state \times year fixed effects, and standard errors are county-level clustered.

Table K13: Effect on Manufacturing Outcomes 1880-1900

	<i>Dependent variable:</i>	
	Avg. Wage Growth	Mfg. Emp. Share Growth
	(1)	(2)
Convict Labor per 100,000 (IHS) \times Competing Industries	-3.749 (2.587)	-6.805*** (2.412)
Controls	\times	\times
State \times Year FE	\times	\times
IV: Early Prison Dummy + Capacities	\times	\times
Mean Outcome	26.4	94.8
Clusters	1,154	1,231
KP F-stat	30.55	30.42
Observations	22,650	24,390

Note: Each column reports coefficients and standard errors (in parentheses) from a regression analogous to equation (11). The observation unit is the county (1880), with the IHS-transformed sum of the interaction of competing industries and the number of convict laborers per 100,000 inhabitants and the sum of these two as the independent variable. It is instrumented by the sum of pre-convict labor prisons and their capacities. Outcome variables are the average manufacturing wage growth and employment share growth. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state \times year fixed effects, and standard errors are county-level clustered.

Table K14: Effect on Union Formation 1880-1900

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership (IHS)	Membership 2020 (IHS)
	(1)	(2)	(3)	(4)
Convict Labor per 100,000 (IHS) × Competing Industries	1.550** (0.663)	0.744** (0.378)	0.249*** (0.070)	0.222*** (0.037)
Controls	×	×	×	×
State × Year FE	×	×	×	×
IV: Early Prison Dummy + Capacities	×	×	×	×
Mean Outcome	1.84	0.42	31.25	3525.44
Clusters	1,302	1,302	1,302	1,302
KP F-stat	70.23	70.23	70.23	70.23
Observations	2,604	2,604	2,604	2,604

Note: Each column reports coefficients and standard errors (in parentheses) from a regression analogous to equation (11). The observation unit is the county (1880), with the IHS-transformed sum of the interaction of competing industries and the number of convict laborers per 100,000 inhabitants and the sum of these two as the independent variable. It is instrumented by the sum of pre-convict labor prisons and their capacities. Outcome variables are the number of KOL assemblies and strikes and IHS-transformed 19th-century KOL membership and union membership in 2020. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K15: Effect on Strike Outcomes 1880-1900

	<i>Dependent variable:</i>			
	Strike	Duration	Aggressive	Hours Change
	(1)	(2)	(3)	(4)
Convict Labor per 100,000 (IHS) × Competing Division	0.666* (0.360)	1.207** (0.563)	0.031** (0.013)	-0.062* (0.036)
Controls	×	×	×	×
State × Year FE	×	×	×	×
IV: Early Prison Dummy + Capacities	×	×	×	×
Mean Outcome	0.08	0.42	0.01	-0.01
Clusters	1,302	1,302	1,302	1,302
KP F-stat	28.79	28.79	28.79	28.79
Observations	13,020	13,020	13,020	13,020

Note: Each column reports coefficients and standard errors (in parentheses) from a regression analogous to equation (11). The observation unit is the county (1880), with the IHS-transformed sum of the interaction of competing division and the number of convict laborers per 100,000 inhabitants and the sum of these two as the independent variable. It is instrumented by the sum of pre-convict labor prisons and their capacities. Outcome variables are the number, duration, and aggressiveness of strikes, and hours of employment change before and after the strike. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K16: Productivity of Convict Labor: Union Outcomes

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership (IHS)	Membership 2020 (IHS)
	(1)	(2)	(3)	(4)
Convict Labor per 100,000 (IHS) \times Productivity	0.123** (0.055)	0.059* (0.031)	0.020*** (0.005)	0.025*** (0.005)
Controls	\times	\times	\times	\times
State \times Year FE	\times	\times	\times	\times
IV: Early Prison Dummy + Capacities	\times	\times	\times	\times
Mean Outcome	1.23	0.28	20.84	3578.69
Clusters	1,192	1,192	1,192	1,187
KP F-stat	59.74	59.74	59.74	59.72
Observations	3,583	3,583	3,583	3,568

Note: Each column reports coefficients and standard errors (in parentheses) from a regression analogous to equation (11). The observation unit is the county (1880), with the IHS-transformed sum of the interaction of productivity of convict labor and the number of convict laborers per 100,000 inhabitants and the sum of these two as the independent variable. It is instrumented by the sum of pre-convict labor prisons and their capacities. Outcome variables are the number of KOL assemblies and strikes and IHS-transformed 19th-century KOL membership and union membership in 2020. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state \times year fixed effects, and standard errors are county-level clustered.

Table K17: Effect on KOL Assemblies 1886-1905

	<i>Dependent variable:</i>						
	Number of KOL Assemblies						
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Convict Labor per 100,000 (IHS)	0.887*** (0.331)	0.884*** (0.330)	0.866*** (0.326)	0.886*** (0.332)	0.863*** (0.316)	0.877*** (0.324)	0.845*** (0.327)
KP F-stat	125.19	125.17	123.42	124.88	129.49	124.86	124.26
Panel B: Early Prison Capacities IV							
Convict Labor per 100,000 (IHS)	0.769*** (0.272)	0.766*** (0.270)	0.748*** (0.267)	0.768*** (0.272)	0.753*** (0.266)	0.877*** (0.324)	0.731*** (0.269)
KP F-stat	195.76	195.89	193.4	196.01	196.7	124.86	194.91
Controls	×	×	×	×	×	×	×
European Share	×						
Italian Share		×					
German Share			×				
Scandinavian Share				×			
French Share					×		
West Indies Share						×	
Scot-Irish Share							×
State × Year FE	×	×	×	×	×	×	×
Mean Outcome	1.23	1.23	1.23	1.23	1.23	1.23	1.23
Clusters	1,180	1,180	1,180	1,180	1,180	1,180	1,180
Observations	3,546	3,546	3,546	3,546	3,546	3,546	3,546

Note: Each column reports coefficients and standard errors (in parentheses). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variable is the number of KOL assemblies. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. Additional controls are share of European, Italian, German, Scandinavian, French, Scot-Irish and West Indies immigrants. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K18: Effect on Strikes 1886-1905

	<i>Dependent variable:</i>						
	Number of Strikes						
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Convict Labor per 100,000 (IHS)	0.283** (0.132)	0.281** (0.130)	0.270** (0.128)	0.285** (0.132)	0.281** (0.131)	0.305** (0.135)	0.273** (0.131)
KP F-stat	125.19	125.17	123.42	124.88	129.49	124.86	124.26
Panel B: Early Prison Capacities IV							
Convict Labor per 100,000 (IHS)	0.275** (0.124)	0.273** (0.123)	0.263** (0.121)	0.276** (0.125)	0.274** (0.124)	0.295** (0.128)	0.265** (0.124)
KP F-stat	195.76	195.89	193.4	196.01	196.7	193.6	194.91
Controls	×	×	×	×	×	×	×
European Share	×						
Italian Share		×					
German Share			×				
Scandinavian Share				×			
French Share					×		
West Indies Share						×	
Scot-Irish Share							×
State × Year FE	×	×	×	×	×	×	×
Mean Outcome	0.28	0.28	0.28	0.28	0.28	0.28	0.28
Clusters	1,180	1,180	1,180	1,180	1,180	1,180	1,180
Observations	3,546	3,546	3,546	3,546	3,546	3,546	3,546

Note: Each column reports coefficients and standard errors (in parentheses). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variable is the number of strikes. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. Additional controls are share of European, Italian, German, Scandinavian, French, Scot-Irish and West Indies immigrants. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K19: Effect on KOL Assemblies, Strikes, and Membership 1886-1905, Membership 2020

	<i>Dependent variable:</i>							
	Number of KOL Assemblies		Number of Strikes		Early Membership (IHS)		Membership 2020 (IHS)	
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)		
Convict Labor per 100,000 (IHS)	1.166*** (0.418)	1.165*** (0.417)	0.450** (0.201)	0.449** (0.201)	0.187*** (0.049)	0.186*** (0.049)	0.225*** (0.040)	0.224*** (0.040)
KP F-stat	126.59	126.48	126.59	126.48	126.59	126.48	126.29	126.18
Panel B: Early Prison Capacities IV								
Convict Labor per 100,000 (IHS)	1.126*** (0.431)	1.124*** (0.431)	0.488** (0.234)	0.488** (0.234)	0.181*** (0.048)	0.180*** (0.048)	0.222*** (0.038)	0.221*** (0.038)
KP F-stat	186.55	186.42	186.55	186.42	186.55	186.42	186.04	185.91
Controls	×	×	×	×	×	×	×	×
Frontier County		×		×		×		×
State × Year FE	×	×	×	×	×	×	×	×
Mean Outcome	1.23	1.23	0.28	0.28	20.83	20.83	3,542	3,542
Clusters	1,302	1,302	1,302	1,302	1,302	1,302	1,296	1,296
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,888	3,888

Note: Each column reports coefficients and standard errors (in parentheses). The unit of observation is the 1880 county. Outcome variables encompass local KOL assemblies, strikes, membership in the 19th century and membership in 2020 with the convict labor presence, measured as convicts per 100,000 inhabitants (IHS), serving as the independent variable. Economic controls include population density, the share of enslaved individuals, average farm output value per acre, and improved farmland acreage. Geographic controls capture distance to the coast, elevation, railway and water body access, and proximity to the nearest urban center. The Democratic Party's vote share acts as the socio-political control. Early prison locations (Panel A) and capacities (IHS) (Panel B) serve as instrumental variables. For each outcome variable, the first column incorporates controls for Economic, Geographic, and Socio-political factors, while the second column additionally includes a dummy for the frontier county. The model incorporates state × year fixed effects and clusters standard errors at the county level.

Table K20: Effect on KOL Assemblies, Strikes, and Membership 1886-1905, Membership 2020

	<i>Dependent variable:</i>							
	Number of KOL Assemblies		Number of Strikes		Early Membership (IHS)		Membership 2020 (IHS)	
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)		
Convict Labor per 100,000 (IHS)	1.166*** (0.418)	1.159*** (0.415)	0.450** (0.201)	0.447** (0.201)	0.187*** (0.049)	0.186*** (0.049)	0.225*** (0.040)	0.221*** (0.040)
KP F-stat	126.59	128.35	126.59	128.35	126.59	128.35	126.29	128.14
Panel B: Early Prison Capacities IV								
Convict Labor per 100,000 (IHS)	1.126*** (0.431)	1.119*** (0.429)	0.488** (0.234)	0.486** (0.233)	0.181*** (0.048)	0.180*** (0.048)	0.222*** (0.038)	0.219*** (0.038)
KP F-stat	186.55	187.91	186.55	187.91	186.55	187.91	186.04	187.49
Controls	×	×	×	×	×	×	×	×
Total Frontier Experience		×		×		×		×
State × Year FE	×	×	×	×	×	×	×	×
Mean Outcome	1.23	1.23	0.28	0.28	20.83	20.83	3,542	3,542
Clusters	1,302	1,302	1,302	1,302	1,302	1,302	1,296	1,296
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,888	3,888

Note: Each column reports coefficients and standard errors (in parentheses). The unit of observation is the 1880 county. Outcome variables encompass local KOL assemblies, strikes, membership in the 19th century and membership in 2020 with the convict labor presence, measured as convicts per 100,000 inhabitants (IHS), serving as the independent variable. Economic controls include population density, the share of enslaved individuals, average farm output value per acre, and improved farmland acreage. Geographic controls capture distance to the coast, elevation, railway and water body access, and proximity to the nearest urban center. The Democratic Party's vote share acts as the socio-political control. Early prison locations (Panel A) and capacities (IHS) (Panel B) serve as instrumental variables. For each outcome variable, the first column incorporates controls for Economic, Geographic, and Socio-political factors, while the second column additionally includes a measure of total frontier experience. The model incorporates state × year fixed effects and clusters standard errors at the county level.

Table K21: Effect on KOL Assemblies, Strikes, and Membership 1886-1905, Membership 2020

	<i>Dependent variable:</i>							
	Number of KOL Assemblies		Number of Strikes		Early Membership (IHS)		Membership 2020 (IHS)	
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)		
Convict Labor per 100,000 (IHS)	1.166*** (0.418)	1.166*** (0.418)	0.450** (0.201)	0.450** (0.202)	0.187*** (0.049)	0.187*** (0.049)	0.225*** (0.040)	0.226*** (0.040)
KP F-stat	126.59	126.66	126.59	126.66	126.59	126.66	126.29	126.36
Panel B: Early Prison Capacities IV								
Convict Labor per 100,000 (IHS)	1.126*** (0.431)	1.126*** (0.431)	0.488** (0.234)	0.488** (0.234)	0.181*** (0.048)	0.181*** (0.048)	0.222*** (0.038)	0.223*** (0.038)
KP F-stat	186.55	186.71	186.55	186.71	186.55	186.71	186.04	186.21
Controls	×	×	×	×	×	×	×	×
Lynchings per 100,000		×		×		×		×
State × Year FE	×	×	×	×	×	×	×	×
Mean Outcome	1.23	1.23	0.28	0.28	20.83	20.83	3,542	3,542
Clusters	1,302	1,302	1,302	1,302	1,302	1,302	1,296	1,296
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,888	3,888

Note: Each column reports coefficients and standard errors (in parentheses). The unit of observation is the 1880 county. Outcome variables encompass local KOL assemblies, strikes, membership in the 19th century and membership in 2020 with the convict labor presence, measured as convicts per 100,000 inhabitants (IHS), serving as the independent variable. Economic controls include population density, the share of enslaved individuals, average farm output value per acre, and improved farmland acreage. Geographic controls capture distance to the coast, elevation, railway and water body access, and proximity to the nearest urban center. The Democratic Party's vote share acts as the socio-political control. Early prison locations (Panel A) and capacities (IHS) (Panel B) serve as instrumental variables. For each outcome variable, the first column incorporates controls for Economic, Geographic, and Socio-political factors, while the second column additionally includes lynching density per 100,000. The model incorporates state × year fixed effects and clusters standard errors at the county level.

Table K22: Effect on KOL Assemblies, Strikes, and Membership 1886-1905, Membership 2020

	<i>Dependent variable:</i>							
	Number of KOL Assemblies		Number of Strikes		Early Membership (IHS)		Membership 2020 (IHS)	
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)		
Convict Labor per 100,000 (IHS)	1.166*** (0.418)	1.165** (0.453)	0.450** (0.201)	0.471** (0.224)	0.187*** (0.049)	0.180*** (0.051)	0.225*** (0.040)	0.169*** (0.037)
KP F-stat	126.59	116.88	126.59	116.88	126.59	116.88	126.29	116.66
Panel B: Early Prison Capacities IV								
Convict Labor per 100,000 (IHS)	1.126*** (0.431)	1.122** (0.470)	0.488** (0.234)	0.511** (0.258)	0.181*** (0.048)	0.174*** (0.050)	0.222*** (0.038)	0.169*** (0.037)
KP F-stat	186.55	172.98	186.55	172.98	186.55	172.98	186.04	172.58
Controls	×	×	×	×	×	×	×	×
Banks per 100,000		×		×		×		×
State × Year FE	×	×	×	×	×	×	×	×
Mean Outcome	1.23	1.23	0.28	0.28	20.83	20.83	3,542	3,542
Clusters	1,302	1,302	1,302	1,302	1,302	1,302	1,296	1,296
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,888	3,888

Note: Each column reports coefficients and standard errors (in parentheses). The unit of observation is the 1880 county. Outcome variables encompass local KOL assemblies, strikes, membership in the 19th century and membership in 2020 with the convict labor presence, measured as convicts per 100,000 inhabitants (IHS), serving as the independent variable. Economic controls include population density, the share of enslaved individuals, average farm output value per acre, and improved farmland acreage. Geographic controls capture distance to the coast, elevation, railway and water body access, and proximity to the nearest urban center. The Democratic Party's vote share acts as the socio-political control. Early prison locations (Panel A) and capacities (IHS) (Panel B) serve as instrumental variables. For each outcome variable, the first column incorporates controls for Economic, Geographic, and Socio-political factors, while the second column additionally includes the total number of banks per 100,000 inhabitants in a county. The model incorporates state × year fixed effects and clusters standard errors at the county level.

Table K23: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies (IHS)	Strikes (IHS)	Early Membership (IHS)	Membership 2020 (IHS)
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)
Convict Labor per 100,000 (IHS)	0.091*** (0.023)	0.049*** (0.016)	0.187*** (0.049)	0.225*** (0.040)
KP F-stat	126.59	126.59	126.59	126.29
Panel B: Early Prison Capacities IV				
Convict Labor per 100,000 (IHS)	0.089*** (0.023)	0.048*** (0.016)	0.181*** (0.048)	0.222*** (0.038)
KP F-stat	186.55	186.55	186.55	186.04
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the IHS-transformed number of KOL assemblies, strikes, 19th-century KOL and 2020 union membership. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K24: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies (1)	Strikes (2)	Early Membership (3)	Membership 2020 (4)
Panel A: Early Prison Dummy IV				
Convict Labor per 100,000 (IHS)	1.166*** (0.418)	0.450** (0.201)	45.129** (22.546)	2,005*** (634.707)
KP F-stat	126.59	126.59	126.59	126.29
Panel B: Early Prison Capacities IV				
Convict Labor per 100,000 (IHS)	1.126*** (0.431)	0.488** (0.234)	40.982** (20.721)	1,973*** (618.722)
KP F-stat	186.55	186.55	186.55	186.04
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the number of KOL assemblies, strikes, 19th-century KOL and 2020 union membership. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K25: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies (log)	Strikes (log)	Early Membership (log)	Membership 2020 (log)
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)
Convict Labor per 100,000 (log)	0.083*** (0.022)	0.044*** (0.015)	0.183*** (0.049)	0.244*** (0.044)
KP F-stat	123.45	123.45	123.45	123.18
Panel B: Early Prison Capacities IV				
Convict Labor per 100,000 (log)	0.081*** (0.021)	0.044*** (0.015)	0.176*** (0.048)	0.241*** (0.041)
KP F-stat	184.13	184.13	184.13	183.66
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - log) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the natural logarithm transformed number of KOL assemblies, strikes, 19th-century KOL, and 2020 union membership. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K26: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies	Strikes	Early Membership (IHS)	Membership 2020 (IHS)
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)
Convict Labor (IHS)	1.318*** (0.458)	0.508** (0.222)	0.211*** (0.053)	0.254*** (0.043)
KP F-stat	139.57	139.57	139.57	139.46
Panel B: Early Prison Capacities IV				
Convict Labor (IHS)	1.268*** (0.469)	0.550** (0.256)	0.204*** (0.051)	0.250*** (0.041)
KP F-stat	230.9	230.9	230.9	230.63
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (the number of convicts employed - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the number of KOL assemblies, strikes, IHS-transformed 19th-century KOL, and 2020 union membership. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K27: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies (1)	Strikes (2)	Early Membership (IHS) (3)	Membership 2020 (IHS) (4)
Panel A: Early Prison Dummy IV				
Number of Prisons (IHS)	8.850*** (3.022)	3.412** (1.459)	1.417*** (0.390)	1.705*** (0.291)
KP F-stat	128.65	128.65	128.65	128.37
Panel B: Early Prison Capacities IV				
Number of Prisons (IHS)	8.487*** (3.051)	3.680** (1.676)	1.362*** (0.373)	1.672*** (0.287)
KP F-stat	188.55	188.55	188.55	188.15
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with the number of prisons employing convict labor as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the number of KOL assemblies, strikes, IHS-transformed 19th-century KOL, and 2020 union membership. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K28: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies (1)	Strikes (2)	Early Membership (IHS) (3)	Membership 2020 (IHS) (4)
Panel A: Early Prison Dummy IV				
Goods Value (IHS)	1.196*** (0.441)	0.461** (0.212)	0.191*** (0.047)	0.230*** (0.047)
KP F-stat	93.75	93.75	93.75	93.63
Panel B: Early Prison Capacities IV				
Goods Value (IHS)	1.155** (0.454)	0.501** (0.245)	0.185*** (0.046)	0.227*** (0.043)
KP F-stat	124.01	124.01	124.01	123.81
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with the total value of goods produced by convict laborers as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the number of KOL assemblies, strikes, IHS-transformed 19th-century KOL, and 2020 union membership. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K29: Normalized Outcomes

	<i>Dependent variable:</i>			
	KOL 100K IHS (1)	Strike 100K IHS (2)	Early Member 100K IHS (3)	Member 2020 100K IHS (4)
Panel A: Early Prison Dummy IV				
Convict Labor per 100,000 (IHS)	0.086*** (0.026)	0.043** (0.018)	0.195*** (0.051)	0.117*** (0.038)
KP F-stat	126.59	126.59	126.59	126.29
Panel B: Early Prison Capacities IV				
Convict Labor per 100,000 (IHS)	0.084*** (0.026)	0.042** (0.018)	0.188*** (0.050)	0.120*** (0.034)
KP F-stat	186.55	186.55	186.55	186.04
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	12.93	1.15	116.08	653,462
Clusters	1,302	1,302	1,302	1,296
Observations	3,906	3,906	3,906	3,888

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the IHS transformed number of KOL assemblies, strikes, 19th-century KOL, and 2020 union membership per 100,000 inhabitants. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K30: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies (1)	Strikes (2)	Early Membership (IHS) (3)	Membership 2020 (IHS) (4)
Panel A: Early Prison Dummy IV				
Convict Labor per 100,000 (IHS)	1.166*** (0.330)	0.450** (0.159)	0.187*** (0.044)	0.242*** (0.039)
F-stat	1,629	1,629	1,629	1,629
Panel B: Early Prison Capacities IV				
Convict Labor per 100,000 (IHS)	1.126*** (0.349)	0.488** (0.196)	0.181*** (0.044)	0.238*** (0.037)
F-stat	1,719	1,719	1,719	1,719
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	1.23	0.28	20.83	3,525
Clusters	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the number of KOL assemblies, strikes, IHS-transformed 19th-century KOL, and 2020 union membership. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and Conley standard errors.

Table K31: Effect on KOL Assemblies and Strikes 1886–1905

	<i>Dependent variable:</i>							
	Number of KOL Assemblies				Number of Strikes			
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Convict Labor per 100,000 (IHS)	2.273*** (0.670)	2.176*** (0.658)	1.154*** (0.256)	1.166*** (0.253)	0.968** (0.326)	0.936** (0.319)	0.449*** (0.142)	0.450*** (0.142)
KP F-stat	185.24	175.22	136.5	136.37	185.24	175.22	136.5	136.37
Panel B: Early Prison Capacities IV								
Convict Labor per 100,000 (IHS)	2.282*** (0.745)	2.187** (0.737)	1.114*** (0.277)	1.126*** (0.273)	1.021** (0.375)	0.992** (0.368)	0.487** (0.171)	0.488** (0.170)
KP F-stat	246.19	227.36	169.63	168.04	246.19	227.36	169.63	168.04
State × Year FE	×	×	×	×	×	×	×	×
Geographic Controls		×	×	×		×	×	×
Economic Controls			×	×			×	×
Socio-political Controls				×				×
Mean Outcome	1.23	1.23	1.23	1.23	0.28	0.28	0.28	0.28
Clusters	1,302	1,302	1,302	1,302	1,302	1,302	1,302	1,302
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses). The unit of observation is the 1880 county. Outcome variables encompass local KOL assemblies and strikes, with the convict labor presence, measured as convicts per 100,000 inhabitants (IHS), serving as the independent variable. Economic controls include population density, the share of enslaved individuals, average farm output value per acre, and improved farmland acreage. Geographic controls capture distance to the coast, elevation, railway and water body access, and proximity to the nearest urban center. The Democratic Party's vote share acts as the socio-political control. Early prison locations (Panel A) and capacities (IHS) (Panel B) serve as instrumental variables. The model incorporates state × year fixed effects and clusters standard errors at the county and state level.

Table K32: Effect on Membership 1886–1905 and 2020

	<i>Dependent variable:</i>							
	Early Membership (IHS)				Membership 2020 (IHS)			
Panel A: Early Prison Dummy IV	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Convict Labor per 100,000 (IHS)	0.243*** (0.040)	0.228*** (0.040)	0.185*** (0.040)	0.187*** (0.040)	0.323*** (0.031)	0.254*** (0.028)	0.214*** (0.026)	0.225*** (0.026)
KP F-stat	185.24	175.22	136.5	136.37	184.81	175.05	136.48	135.85
Panel B: Early Prison Capacities IV								
Convict Labor per 100,000 (IHS)	0.238*** (0.043)	0.224*** (0.044)	0.179*** (0.041)	0.181*** (0.041)	0.318*** (0.029)	0.250*** (0.028)	0.211*** (0.025)	0.222*** (0.024)
KP F-stat	246.19	227.36	169.63	168.04	245.54	227.12	169.61	167.28
State × Year FE	×	×	×	×	×	×	×	×
Geographic Controls		×	×	×		×	×	×
Economic Controls			×	×			×	×
Socio-political Controls				×				×
Mean Outcome	20.83	20.83	20.83	20.83	3,542	3,542	3,542	3,542
Clusters	1,302	1,302	1,302	1,302	1,296	1,296	1,296	1,296
Observations	3,906	3,906	3,906	3,906	3,906	3,906	3,906	3,906

Note: Each column reports coefficients and standard errors (in parentheses). The unit of observation is the 1880 county. Outcome variables encompass local KOL assemblies and strikes, with the convict labor presence, measured as convicts per 100,000 inhabitants (IHS), serving as the independent variable. Economic controls include population density, the share of enslaved individuals, average farm output value per acre, and improved farmland acreage. Geographic controls capture distance to the coast, elevation, railway and water body access, and proximity to the nearest urban center. The Democratic Party’s vote share acts as the socio-political control. Early prison locations (Panel A) and capacities (IHS) (Panel B) serve as instrumental variables. The model incorporates state × year fixed effects and clusters standard errors at the county level.

Table K33: Effect on KOL Assemblies, Strikes, and Membership

	<i>Dependent variable:</i>			
	KOL Assemblies (1)	Strikes (2)	Early Membership (IHS) (3)	Membership 2020 (IHS) (4)
Average Merge				
Panel A: Early Prison Dummy IV				
Convict Labor per 100,000 (IHS)	0.123*** (0.045)	0.048** (0.021)	0.083*** (0.026)	0.222*** (0.042)
KP F-stat	170.46	170.46	170.46	169.04
Panel B: Early Prison Capacities IV				
Convict Labor per 100,000 (IHS)	0.119*** (0.046)	0.052** (0.025)	0.078*** (0.023)	0.220*** (0.040)
KP F-stat	232.38	232.38	232.38	230.34
Cumulative Average Merge				
Panel C: Early Prison Dummy IV				
Convict Labor per 100,000 (IHS)	0.124*** (0.045)	0.048** (0.021)	0.084*** (0.025)	0.223*** (0.041)
KP F-stat	164.57	164.57	164.57	163.85
Panel D: Early Prison Capacities IV				
Convict Labor per 100,000 (IHS)	0.120*** (0.046)	0.052** (0.025)	0.078*** (0.023)	0.221*** (0.039)
KP F-stat	247.7	247.7	247.7	246.42
Maximum Value Merge				
Panel E: Early Prison Dummy IV				
Convict Labor per 100,000 (IHS)	0.117*** (0.042)	0.045** (0.020)	0.079*** (0.024)	0.210*** (0.040)
KP F-stat	169.21	169.21	169.21	167.85
Panel F: Early Prison Capacities IV				
Convict Labor per 100,000 (IHS)	0.113*** (0.044)	0.050** (0.024)	0.074*** (0.022)	0.209*** (0.038)
KP F-stat	228.95	228.95	228.95	227.02
Controls	×	×	×	×
State × Year FE	×	×	×	×
Mean Outcome	0.13	0.03	2.23	3,542
Clusters	1,302	1,302	1,302	1,296
Observations	36,456	36,456	36,456	36,288

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons and their capacities. Outcome variables are the number of KOL assemblies, strikes, IHS-transformed 19th-century KOL, and 2020 union membership. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

Table K34: Effect on Manufacturing Outcomes 1880-1900

	<i>Dependent variable:</i>				
	Avg. Wage Growth	Employment Share	Capital Cost Share Change	Output Value	Gross Profit
	(1)	(2)	(3)	(4)	(5)
Panel A: Early Prison Dummy IV					
Convict Labor per 100,000 (IHS)	-1.581* (0.816)	-1.920*** (0.670)	0.448** (0.204)	181,093** (77,583)	31,254** (13,983)
KP F-stat	86.79	86.05	86	86.03	86.03
Panel B: Early Prison Capacities IV					
Convict Labor per 100,000 (IHS)	-1.382* (0.771)	-1.914*** (0.629)	0.451** (0.204)	185,888** (88,966)	31,160** (15,330)
KP F-stat	121.15	120.26	120.24	120.4	120.4
Controls	×	×	×	×	×
State × Year FE	×	×	×	×	×
Mean Outcome	26.4	94.8	12.58	970,522	79,779
Clusters	1,154	1,231	1,237	1,298	1,298
Observations	2,265	2,439	2,463	2,526	2,526

Note: Each column reports coefficients and standard errors (in parentheses) from the regression displayed equation (11). The observation unit is the county (1880), with convict labor presence (convicts per 100,000 inhabitants - IHS) as the independent variable. It is instrumented by pre-convict labor prisons (Panel A) and their capacities (Panel B). Outcome variables are the average manufacturing wage growth, manufacturing employment share, the change in capital-labor cost-share, manufacturing output value, and gross profits. Economic controls encapsulate population density, enslaved individuals' share, and farmland statistics (output value per acre and improved acres). Geographic controls account for distance to the coast, elevation, transport access, and proximity to large urban centers. The vote share of the Democratic Party is the socio-political control. The model includes state × year fixed effects, and standard errors are county-level clustered.

K Types of Convict-Labor Systems in the United States

This section presents the definitions of the six convict labor systems: contract, piece-price, state-account, state-use, public-works-and-ways, and convict leasing.

Under the *contract system*, prison officers, under legal instruction, advertised for bids for the employment of the convicts of their respective institutions. The highest bidder would secure the contract. The contractor employed several convicts at a specific price per day. The prison or the state furnished power and sometimes machinery, but rarely tools. All convicts were employed within the walls of the prison.

The *piece-price system* was similar to the contract system, except that the contractor had nothing to do with the convicts. Under this system, the contractor had no position at the prison. The contractor furnished the prison officers with material ready for manufacturing. The prison officers agreed to return the completed work, for which the government received an agreed price per piece.

Under the *convict leasing system*, prisons and local sheriffs had the right to “lease” convicts to private individuals, firms, farms, and plantations. The lessee paid the prison and various public officials involved and were responsible for feeding, clothing, and housing the prisoners (Sellin, 1976).

Under the *state-account system*, the prison acted as a firm and sold goods on the market, thus assuming all business risks. All profits went directly to the states. However, this system had two major problems. The first problem was managerial: wardens were often subpar businessmen. Second, prisons needed to employ convicts even if there was no demand for the goods produced.

The *state-use system* is similar to the state-account system, except that the sale of goods was limited exclusively to state departments and agencies.

Under the *public works and ways (PWW) system*, as is evident from the name, prisoners constructed and repaired roads rather than producing goods for consumption.

M Jim Crow Laws

M.1 Enticement and Contract Enforcement Laws

It was illegal under the *enticement laws* for an employer to “entice” a worker under contract with another employer. These rules aimed to stop companies from aggressively bidding against one another for contract workers and were particularly geared toward white firms. Enticement laws have a long history in criminal and civil law, dating back to England in the fourteenth century. However, eight of the eleven Southern states that approved enticement laws did so in a way that turned the practice into a criminal infraction instead of a civil wrong.

The *contract-enforcement laws* enforced criminal penalties for violating an employment contract and were particularly aimed at black farm employees. According to such rules, a worker who signed a contract and then quit their position might be detained and charged with a crime. Ultimately, they had only two options: negotiate their contract or join the chain gang. Both inducement and contract enforcement legislation have the same economic goal: to reduce competition for farm labor during the contract season.

In the 1911 decision of *Bailey v. Alabama*, the United States Supreme Court ruled that the Alabama Act was unconstitutional. The court stated that the defendant was “stripped by the statute of the presumption of innocence, and exposed to conviction for fraud upon evidence only of breach of contract and failure to pay” and that there was “not a particle of evidence of any circumstance indicating that [the defendant] made the contract or received the money with any intent to injure or defraud his employer.” According to the Alabama Act, refusal to work and refund the money was *prima facie* evidence of intent to defraud.

The Peonage Act of 1867, which forbade involuntary servitude except to punish crime, stated, “The state... may not compel one man to labor for another in payment of a debt, by punishing him as a criminal if he does not perform the service or pay the debt.” The Alabama statute violated both of these provisions.

There is some proof that employment contracts were nonetheless upheld, notwithstanding the lack of evidence that the employee intended to defraud. The employee would frequently be ignorant that the legislation required proof of fraudulent intent or that the current law was unconstitutional. Additionally, employees were occasionally “reclaimed” by their employers—rather than by law enforcement officials—and coerced into working off debt without a warrant. Thus, the legislation gave cruel acts that were actually against the law the appearance of legitimacy.

M.2 Vagrancy Laws

Vagrancy laws served as the second law to enforce the labor-market cartel. Vagrancy laws essentially made it illegal to be out of work or unemployed. The traditional definition of a vagrant was “anyone roaming or strolling about in idleness, who can work, but has no property to support him; or any person leading an idle, immoral, profligate life, having no property to support him.” While vagrancy was typically regarded as a misdemeanor, this does not imply that the punishment was insignificant; misdemeanants were frequently put into state or county chain gangs.

Laws making unemployment a crime reduced the time spent looking for new, better-paying employment, at least in part. More importantly, these laws raised the cost of leaving the labor field altogether. One can discover instances of the law being applied to both goals. Black people who traveled, even to see family, risk being arrested. People who weren’t working or not in the labor force were frequently gathered up as vagrants and forced to work on nearby farms or public works projects.

M.3 Emigrant-Agent Laws

The *emigrant-agent legislation* was the last category of labor law. The Southern states created these regulations to stifle the actions of brokers who hired workers from one state for employment in another. Not the black employees themselves, but the white recruiters were subject to regulation under these regulations, which mandated that agents be licensed for up to \$5000 per county where the recruiting took place. Many states repeatedly raised the taxes and penalties in response to subsequent immigration since the laws were frequently implemented in response to a wave of out-migration.

By the time of the Great Migration of Black people to the North during World War I, Montgomery, Alabama, passed a law making it illegal to tempt workers to leave the city to pursue work elsewhere. Violators would be subject to a \$100 fine, up to six months of hard labor, or both. While these punishments might not seem severe, the ordinance had broad implications. For this law, anyone who printed, published, authored, delivered, posted, or distributed any newspaper, booklet, or advertisement encouraging citizens to emigrate was considered an emigrant agent. Blacks’ one defense against this ordinance was to subscribe to Northern periodicals exempt from local law. A radical black journal, *The Chicago Defender*, promoted visits to the Chicago stockyards.

The emigrant-agent regulations’ major impact on the economy was raising the cost of black laborers’ access to information about employment opportunities outside their immediate market. The city of Montgomery’s efforts to retain its labor force demonstrates how

limited a description of the local labor market may be. Although possible and often done, breaking the law came at a price for both the agents and the workers.

M.4 Convict-Lease System

The *convict-lease system*, which is a statutory practice of leasing out state or county convicts to private businesses, was used by Southern states in addition to the legislation already mentioned. Particularly in the 1880s, convict leasing was commonly practiced in the South. The lease system gave the lessee firm control over the day-to-day management of the prisoners, in contrast to other forms of convict labor where the government retains control. The lessee firm lacked any financial incentive to ensure the survival of prisoners beyond the expiration of their sentences or contracts. The lease system was worse than slavery in this regard. A slaveholder is incentivized to maintain the enslaved person's health since he obtains the fully capitalized worth of the enslaved person's output for the duration of his working life. These chain gangs' fatality statistics highlight the distinction: Mortality rates were as high as 45%.

The convict-lease system was significant because many victims were black misdemeanants, such as vagrants and those who broke the laws enforcing contracts. A combination of the likelihood of punishment and the seriousness of punishment can be used to estimate the cost of breaching any law. Although we lack a precise estimate of the probability of receiving a sentence for contract jumping or vagrancy, a 45% percent possibility of death as a punishment undoubtedly served as a potent deterrent for the black laborer.

N Merging Methods

I have digitized and compiled seven Bureau of Labor reports covering the years 1886-1940. These reports provide comprehensive data on penal institutions and the prisoners employed therein, including information on the facility's name, location, management, employment practices, as well as a breakdown of inmates by gender and prison responsibilities. To supplement this, I have also utilized data on Knights of Labor assemblies from 1869-1919, which include location, years of operation, and membership statistics. Additionally, I have incorporated information on strikes that occurred from 1881-1894, which includes location and dates. However, I face some limitations due to the absence of data on local Knights of Labor assemblies after 1919 and the lack of information on the number of convicts employed in the 1914 Bureau of Labor report. As a result, I focused on the Bureau of Labor reports from 1886-1905 and employed different merging methods, which are described below.

Report Year Merge: In this particular merge, also known as the “report year merge,” I adopt a discrete merging method to combine the Knights of Labor assemblies and strikes with prison data. The Bureau of Labor releases reports for the years 1886, 1895, 1905, 1914, 1923, 1932, and 1940. To accomplish the merge, I introduce a merge-year variable to the KOL and strikes data that takes on the values of 1886, 1895, and 1905 for the respective periods of 1886-1894, 1895-1904, and 1906-1913. Within each time bracket, I calculate the total number of assemblies and strikes. Ultimately, I merge the resulting data with prison data using the year of the report (merge-year) and the county FIPS codes.

Average Merge: I use a continuous merging method, known as the “average merge,” to combine the data. This method involves calculating the average number of prisons and employed convicts over the seven report years for each county. First, I assign year variables to the prison data, covering the period from 1869 to 1940. Then, the resulting averages are assigned to the corresponding year variables. Finally, I merge these data with the Knights of Labor and strikes data by utilizing the year and county FIPS codes.

Cumulative Average Merge: I utilize a continuous merging method called the “cumulative average merge” to combine The data. The process involved assigning a year variable to the prison data spanning from 1869 to 1940. I then compute the average number of prisons and convicts employed for each county for every year between the report years. Specifically, I calculate the average of these variables between 1886 and 1895 for the year 1890, and for years between 1869 and 1886, I average 0 and the 1886 value. Ultimately, I merge this data with the Knights of Labor and strikes data using year and county FIPS codes.

Maximum Value Merge: I utilize a continuous merging method known as the “maximum value merge” to merge The data. This method involves obtaining the highest value for the number of prisons and convicts employed in each county across the seven report years. I then assign year variables to the prison data spanning from 1869 to 1940, with each year having the maximum value of all the report years. Lastly, I merge this data with the Knights of Labor and strikes data, using year and county FIPS codes.