The Dynamic Effects of Computerizing VAT Invoices in China*

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Abstract

This study examines the short- and long-run effects of computerizing VAT transactions on large manufacturing firms in China. In the short run, computerization increased tax revenues by reducing VAT evasion from exaggerated deductions. In the long run, computerization increased tax revenue, but the magnitude of gains is smaller than in the short run due to a decline in revenue. The results suggest that the revenue decline is at least partly driven by a fall in real output. A back-of-the-envelope calculation shows that the computerization of VAT accounts for 13.7% of VAT growth during 1998 to 2007 and 11.7% of China’s total 2000 VAT revenue.

Keywords: State Capacity, Tax Enforcement, Tax Compliance

JEL: H25, H26, O12

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1 Introduction

Increasing public revenues is a primary concern for developing middle-income economies, which have much lower tax revenues as a share of GDP than rich countries because of relatively limited state capacity and rampant tax evasion (e.g., Besley and Persson, 2013; Burgess and Stern, 1993). One of the most important sources of tax revenues in such contexts is the Value Added Tax (VAT). The popularity of the VAT partly comes from its self-enforcing properties – i.e., upstream firms are incentivized to understate their sales, but downstream firms are incentivized to overstate their inputs (Gordon and Li, 2009; Kleven et al., 2016a). Such third-party information can help increase compliance and reduce evasion (Gordon and Li, 2009; Kleven et al., 2016b). Policymakers face two challenges in implementing the VAT. First, tax officials in most developing economies do not have the capacity to link and validate VAT transactions, which limits the extent of self-enforcement (Pomeranz, 2015). Second, firms can respond to an increase in VAT enforcement by changing their economic or reporting behavior and offset the tax increase. For example, increased VAT can cause firms to evade other less well-enforced taxes or reduce real output.

This paper addresses these questions by examining the effect of computerization and linking VAT invoices on VAT revenues and firm behavior.

We study the context of China, where VAT transactions were computerized in 2001. Before this, it was easy to falsify quantities on the hand-written invoices and evasion was rampant. Because large or sudden changes in sales triggered audits, evasion was usually accomplished by exaggerating the amount of deductibles rather than understating sales. The tremendous number of transactions made it infeasible for tax authorities to manually and systematically link the downstream input and upstream sales receipts. The computerization of VAT was designed to increase tax revenues and reduce evasion. The reform digitized invoices and linked transactions nationwide, and strengthened the information chain for VAT enforcement.

China is an ideal context to study the effect of linking VAT transactions on tax revenues and firm behavior for several reasons. First, the rapid implementation of a nationwide roll-out in 2001 facilitates empirical identification. Second, the presence of a functional tax administration and the ability to credibly punish evaders means that our estimates of the increase in third-party information are not confounded by other deficiencies in administrative capacity. Similarly, the fact that changes in revenues trigger audits helps us understand the mechanism driving our results. Finally,

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1 According to the United Nations, 166 member nations have a VAT in 2018. Kopczuk and Slemrod (2006) argue that VAT is easier to enforce than sales tax, to which it is otherwise equivalent. Besley and Persson (2009, 2010) make a point of using the ratio of income tax revenues to GDP as a measure of bureaucratic capacity, with the underlying idea that VAT requires much less capacity to administer than other types of taxes.

2 It also reduced reporting and documentation costs for firms and may have allowed tax officials to better target audits by providing better information. These effects are part of the reduced form estimates. We discuss in the paper why our results are unlikely to be driven by changes in reporting costs.
China is similar to other developing-middle income economies in its need to increase tax revenues and digitize tax collection. At the same time, China is the world’s second-largest economy and VAT is the most important source of tax revenues (e.g., 47.61% of total tax revenues for China in 2002)\textsuperscript{3} Thus, our findings shed light on a relevant context.

Our study faces two empirical challenges. The first is the measurement of VAT since administrative tax data are not available from this period. The second is causal inference. Changes in VAT before and after computerization may be spuriously correlated with changes in macroeconomic growth. Our analysis addresses both of these challenges.

To measure VAT, we construct a firm-level panel using the Annual Survey of Industrial Production (ASIP), 1998-2007. The ASIP includes the universe of large manufacturing firms in China. The surveys are conducted by National Bureau of Statistics enumerators, who transcribe information from firm records.\textsuperscript{4} An important advantage of the ASIP data is that in addition to VAT payment, it reports a breakdown of gross VAT and VAT deductibles such that we are able to examine the mechanisms underlying changes in VAT. Our main analysis uses a balanced panel of firms that exist throughout the period that we study to avoid the confounding influences of firm entry and exit.

To estimate the causal effect of computerization, we exploit two sources of variation. First, we compare outcomes before and after computerization. Second, we exploit sector-level variation in the amount of non-deductible inputs as a share of sales, henceforth, NDS. Firms with higher NDS had stronger incentives to falsify claims (i.e., there is little need to evade VAT if all sales can be offset by legitimately deductible inputs). Thus, higher NDS firms are more intensely treated by the reform. Our approach is similar in spirit to a difference-in-differences strategy, except that the cross-sectional variation is a continuous measure. The baseline specification uses a parsimonious set of controls and includes firm fixed effects to account for all time-invariant differences across firms (e.g., firm size), and year fixed effects to account for all economy-wide changes over time (e.g., macroeconomic growth). To allow for firms of different sizes to evolve differentially over time, it also controls for the interaction of both the firm’s average sales and average VAT in the pre-reform period and year fixed effects.

We find that computerization significantly increased VAT in the five-year post-period. The increase in VAT was driven by a reduction of deductible inputs. Computerization had little effect on reported VAT sales. These results are consistent with a pre-period firm evasion strategy of exaggerating inputs. We find no evidence that the increase in VAT resulted in substitution away from paying other taxes or tax avoidance behavior such as shifting towards exports or lower NDS.

\textsuperscript{3}Source: China Tax Policy Department, Ministry of Finance 2007.
\textsuperscript{4}The ASIP has been used by recent studies on corporate income tax (Cai and Liu, 2009) and payroll tax (Li et al., 2021).
sectors. The results are driven by firms with low trade exposure and broadly similar between state-owned and privately owned firms.

To estimate the magnitude of the increase, we use the full sample of ASIP firms, which yields qualitatively similar estimates to the smaller balanced panel. A back-of-the-envelope calculation using the full sample estimates implies that computerization accounted for 13.7% of VAT revenue growth from 1998-2007. These results are economically meaningful when one considers that nearly half of total tax revenues for China in the early 2000s came from VAT. In other words, the computerization of VAT constituted one of the largest sources of revenue growth for the Chinese government during this period.

Our estimates are likely to understate the true effect of computerization on tax revenues for two reasons. First, we do not have a perfect control group. In our sample, even the lowest NDS firms had some non-deductible inputs and were therefore partially treated. Second, computerization likely generated enforcement spillovers along transaction chains to low NDS firms.

The main challenge to the causal interpretation of our baseline estimates is the concern of unobserved factors which influence VAT differently for high and low NDS sectors. We address this concern in several ways. First, we document that there are no pre-trends for any of the main outcome variables. Second, we consider other changes around the time of computerization. The two main concerns are China’s entry into the World Trade Organization (WTO) in 2001 and the enterprise reforms that partially privatized historically state-controlled firms during 1998-2003. We address these potentially confounding influences in several ways. For entry into the WTO, we control for sector-by-year import tariffs and export rebates, pre-period sector-specific export growth rates, and sector-year import and export flows. We also show that firms which are more exposed to trade do not drive the main results. For the enterprise reforms, we show that the results are robust to controlling for firm ownership in the base year interacted with year fixed effects, and that the effects are similar for state and privately owned firms. More generally, since most major economic reforms are implemented at the province level, we also control for province-year fixed effects. Our results are very robust. We conduct many additional sensitivity exercises, such as addressing potential measurement error in our sector-level NDS variable by showing that the results are qualitatively similar if we instrument for the Chinese measure with NDS calculated from data from other countries. We also perform random permutation tests. See Section 4.4 for these and other robustness results.

In addition to examining the average impact of computerization during the post-period, 2002-
2007, we also examine dynamic effects. This is interesting because the long-run response can be different from the short-run response. In the short run, firms can easily change reporting behavior, but cannot easily change real economic behavior such as adjusting production inputs. In the long run, firms can change both types of behavior. We find that VAT revenues experience the largest increase in the first four years after the reform and then decline slightly afterward. The decline is statistically insignificant. After the reform, both gross VAT and VAT deductible inputs decline monotonically over time. Computerization also reduced total sales and deductible intermediate inputs and increased total factor productivity (TFPR) in the long run.

The last part of the paper provides indirect evidence on the extent to which firms are changing their economic behavior or reporting behavior. The two changes are not mutually exclusive. For economic behavior, we show that the empirical results are consistent with a simple model where firms adjust inputs in the long run. For reporting behavior, we first consider the possibility that firms gradually reduce reported sales over time to avoid audits by separately examining sales that are eligible and ineligible for VAT. We find that computerization also reduces both types of sales, even the one that would not reduce VAT liability\footnote{One possible explanation is that firms understate total revenues beyond what is part of VAT in order to evade corporate income tax – i.e., Enterprise Income Tax (EIT). We rule this out by showing that computerization has no effect on EIT.} Second, we consider the most common method of two-way collusion in the presence of stronger third-party information that has been documented in other middle-income economies: flying invoices (Waseem, 2020). We find no evidence of this in our context. We conclude that the negative effect of computerization on production (sales and inputs) are consistent with changes in economic behavior and not unlikely to be entirely driven by misreporting.

Taken together, this study shows that digitizing and linking VAT invoices generated large tax revenue gains for the Chinese government. In the long run, firms responded by scaling down production, but not enough to offset the large tax revenue gains from the increased compliance.

Our study is the first to estimate the dynamic long-run effects of an increase in VAT compliance and to present evidence of changes in real economic behavior\footnote{Harju et al., 2016 use the bunching behavior of small Finnish firms around VAT compliance cost thresholds and the fact that bunching does not vary by firm size to infer that there is a reduction in the growth of real output.} We are most closely related to studies of tax compliance in developing countries, where firms have been found to over- and under-pay VAT. For example, a recent study by Almunia et al. (2021) finds that approximately 25% of firms in Uganda misreport sales and inputs so that they over-pay VAT, while the remaining majority misreport to evade VAT. In contexts where firms are believed to systematically under-pay, earlier studies have found that third-party information can improve compliance. Naritomi (2019) finds that providing rewards for consumers to whistle-blow increases VAT in Brazil. Pomeranz (2015) finds that VAT paper trails prevent evasion in Chile. Eissa and Zeitlin (2014) find that the
introduction of electronic billing machines increased VAT in Rwanda. In evaluating the impact of computerization on tax revenues, we are closely related to Jensen et al. (2022), which finds that mapping and revenue management software increased property tax revenues in Ghana. Waseem (2020) exploits variation in tax cuts in Pakistan to find that firms evade VAT by overstating inputs, understating sales and using invoice mills.

In studying tax compliance in China, we complement Fisman and Wei (2004), which detects evasion using customs data discrepancies and Li et al. (2021), which finds evidence that firms offset VAT increases by evading payroll taxes. In contrast to Li et al. (2021), we find no evidence that the increase in VAT caused by the computerization of VAT was offset by a reduction in payroll taxes. Our finding that computerizing VAT transactions has such a strong effect on compliance in China, where tax officials can audit and punish evading firms, is consistent with Almunia and Lopez-Rodriguez (2018), which finds that the paper trail and other monitoring efforts are complements in Spain.

This paper is organized as follows. Section 2 discusses the relevant institutional background. Section 3 describes the data. Section 4 presents the empirical strategy and estimates of the average effects of computerization on VAT. Section 5 presents the dynamic effects of computerization and their interpretation. Section 6 concludes.

## 2 Background

This section summarizes the enforcement environment prior to the computerization reform. We draw on government documents as well as interviews of tax officials and firm managers.

### 2.1 VAT

China first introduced the VAT in 1994. By 2002, it had become 46.5% of total tax revenue, the largest single source. All formal manufacturing firms were required to register within the VAT.

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9For an overview of the larger literature, see Andreoni et al. (1998), Slemrod and Yitzhaki (2002), and Saez et al. (2012).
10Several studies use notches in the tax system to infer behavior. For example, in Pakistan, see Best et al. (2016); Kleven and Waseem (2013); in Costa Rica, see Bachas and Soto (2018); in China, see Chen et al. (2021b).
11We proxy for payroll taxes with the wage bill, which is the main determinant of payroll taxes, because payroll tax data are not available until 2001. Li et al. (2021) instrument for county-level VAT revenue with the decline in county tax revenue driven by the abolition of Agricultural Taxes in 2004; the 2SLS estimate of the instrumented VAT revenue on payroll tax revenue is negative. The main differences between their data and ours are the source of the variation and sample. Although their main sample includes large manufacturers, their results are driven by privately owned, small and cash-constrained firms. Our results are similar across firm ownership and size. These results are in Subsection 4.5.
The final VAT bill was 17% of the VAT tax base, which equals the difference between total VAT-eligible sales and total eligible input deductions. In our study period, full deductions were awarded for purchases of manufactured inputs, repair inputs, retail inputs and wholesale inputs. No deductions were given for fixed asset purchases, capital depreciation, abnormal losses, rent, fringe benefits, interests from bank loans and operating expenses (overhead). Labor was not deductible. For any deductible imported inputs, firms could report purchases using VAT completion receipts issued by the customs office (State Council 1993). Exports were partially exempt from VAT due to rebates that vary by sector and year. There were no other notable changes in the VAT formula or exemptions during the period of our study.

A typical VAT-relevant transaction was a firm-to-firm sale of some input. An official handwritten invoice with carbon copies was generated: one copy for the buying firm and the other one for the selling firm. Registered firms could obtain these invoices from the local tax office and firms paid VAT and obtained deductions monthly. Each month, a firm representative went to the local tax office and gave the tax official invoices for all VAT-eligible sales from the past month. These are used to calculate VAT obligations, which are paid on the same day. During the same visit, the representative submitted the firm’s invoices for VAT-eligible deductibles. The deductions were calculated and paid back on the same day.

In the pre-computerization period, VAT fraud was prevalent (Lu 1997; Jin 2002). VAT invoices were handwritten and lacked effective anti-counterfeit technology. Manually cross-checking invoices for the tremendous number of transactions was prohibitively costly. As a result, firms could exaggerate deductible inputs and be fairly certain that they would not be caught. Similarly, firms could use real invoices of canceled transactions to file for deductions because tax authorities would not know that the other party had never filed the sales.

The main tool for enforcing tax payment was audits. Audits were not random and were triggered by sudden changes in sales or ownership. One trigger for audits for these firms is revenue changes, and in particular, revenue declines since they are a sign of evasion as well as a cause for political concern. Another trigger for audits was reporting too many deductibles relative to sales. In 2000, the audit rate was 17.9%. In comparison, it was 1.12% in the United States (Internal Revenue Service 2001). There were more than 10,041 criminal cases in China that year and

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12 Firms register either as a “small VAT taxpayer” or a “general VAT taxpayer”. Within the manufacturing sector, firms with less than one million RMB (120,772 USD) in annual sales were categorized as “small” and larger firms were categorized as “general” (Ministry of Finance 1993). Because our dataset contains only firms much larger than this cutoff (i.e., annual revenues exceeding five million RMB, or 603,864 USD), we focus the rest of our discussion on general VAT taxpayers. Note that we use the 1 USD = 8.28 RMB conversion rate from 2000 in this paper.

13 A reduced rate of 13% was made for basic staples or household necessities such as food, fuel, electricity, books, newspapers and magazines, and primary agricultural products (State Council 1993); a reduced rate of 10% was applied to procured waste goods; a reduced rate of 7% was applied to transportation costs. These exceptions do not affect our study, which examines very large manufacturing firms.

14 The Appendix provides a detailed list of deductible and non-deductible items.
over 60% of these cases were related to over-reporting input VAT (page 632 of China’s Tax Audit Yearbook, 2003). Very few cases ultimately resulted in penalties because of the difficulty of conclusively proving evasion, or estimating its extent, without a clear paper trail. Hence, computerization complemented audits in enforcing VAT. Throughout the period of study, the procedures for audits and punishments for VAT evasion were unchanged, although computerization likely improved audit efficacy.

2.1.1 Computerization Reform

The goal of computerization (known in China as the second phase of the Golden Tax Project) was to improve VAT enforcement with a fully digitized invoice system coupled with a national database of firm VAT filings. The two most important components of this reform were: 1) replacing handwritten invoices with digitally encrypted invoices, and 2) digitally linking transactions (Jin, 2002).

Under the new system, firms had to use new smart cards to complete transactions. The IC cards contained basic information of the holding firm. During a transaction, both parties would insert their cards into a computer, which created a unique transaction record. This record took the form of an 84-digit code that encrypted the invoice ID, invoice code, invoice date, buyer’s tax ID, seller’s tax ID, value and VAT of the transaction. This information would then be stored on both cards and the seller would print the deductible amount on a paper invoice for the buyer.

Under the new system, it was prohibitively difficult to generate false transactions, as the digital records came with a unique ID and QR code. It also became more difficult to change the amount on an invoice, as the value of the transaction was now encoded within the transaction ID. With nationwide linking, it became even harder to unilaterally falsify transactions or amounts without cross-firm collusion.

After the reform, firms continued to pay VAT during monthly visits to local offices of the State Administration of Tax (SAT). They presented their IC cards as well as printed copies of the encrypted deduction invoices. As before, the net payment was calculated and made at the tax office the same day of the visit. The deduction invoices were cross-checked with a national database of transactions, effectively linking all transactions (Xu, 2003).

Computerization increased the fixed cost of reporting because it forced firms to adopt new computing equipment and reduced the marginal cost of reporting because it was much easier for

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15 To the best of our knowledge, there are no disaggregated data on audits available to researchers. See the Appendix for a discussion about the aggregate audit data.

16 The technology has continuously improved. For example, in later years, the government extended the 84-digit code into a 108-digit code. In 2011, another improvement permitted the encryption of Chinese characters as well as numerals, so the government added additional information to deduction invoices in a few designated sectors (e.g., gold, gasoline, rare earth, etc.) in three provinces (Shanghai, Shaanxi, and Shenzhen). The information includes the seller’s name, the buyer’s name, the product name and the quantity sold.
firms to record digital transactions. After we present the main results, we discuss why it is unlikely that our results are driven by the change in reporting costs.

Policymakers claim that the technology had a remarkable deterrence effect on VAT fraud in China, mainly by reducing exaggerated deductions. For example, in February 2001, the number of invoices that tax authorities identified as “problematic” over the total number of invoices audited was 8.51%. By August 2002, it had dropped to 0.062% (Jin, 2002). On January 20, 2003, the former deputy head of SAT reported that the over-invoicing problem had been effectively resolved. Beyond tax officials, it is widely believed by firms and other policymakers that fake invoices have almost completely disappeared.

In interviews conducted by the authors, tax officials discussed the few remaining evasion strategies after the reform. First, a subset of receipts (including procurement receipts for agricultural goods, customs VAT completion receipts, transportation receipts, and waste goods) were not covered by computerization and it was still possible to falsify these invoices. Second, firms could buy genuine invoices on a black market from final consumers who did not wish to claim deductions. Finally, entire value chains could opt for off-book cash transactions, though an immediate switch to cash transactions would trigger an audit. Outside of the Chinese context, studies of other middle-income countries have also highlighted the use of flying invoices. We discuss this more in Section 5.2.

For our empirical strategy, it is important to keep two points in mind. First, tax officials claim that before computerization, the firms with a high level of non-deductible inputs as a share of sales (NDS) were more incentivized to evade and found it easier to evade VAT. Such firms had fewer legitimate deductibles with which to lower VAT obligations (as a share of sales). They also found it easier to evade since a large amount of deductibles relative to sales triggered audits. This means that firms with high NDS, which had less real deductibles to report, had more scope for exaggeration. The conventional wisdom that firms with low NDS evaded VAT more prior to computerization is important to keep in mind for understanding our empirical strategy later in the paper. Second, computerization made it difficult for all firms to evade. Thus, the fact that high NDS firms evaded more prior to computerization (see the discussion in the previous section) meant that high NDS firms were more intensely treated by computerization. Our empirical strategy uses this cross-sectional variation in intensity.

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18 The government aimed to resolve these remaining loopholes through improved enforcement technology. The third phase of the Golden Tax Project was piloted in Chongqing in 2013 and applied nationwide in 2016. It is outside of the scope of our study.
2.1.2 Timing

Computerization took some time to implement. Starting from January 1, 2000, transactions exceeding 100,000 RMB (12,077 USD) were invoiced using the new encryption software and such large transactions were common for the large manufacturing firms that we study. Handwritten invoices for these large transactions were banned at the end of 2001. Though provinces varied in implementation speed, numerous tax officials at the central, provincial and local levels stated that the system was operational nationwide by July 1, 2001. We interpret 2002 as the first year that the reform became relevant for all provinces (State Administration of Taxation, 2000). In robustness exercises, we control for province-year fixed effects to account for the slight variation in the timing of the roll-out of the program.

2.1.3 Imports, Exports, Other Changes

The VAT payment rules that we have described thus far apply to almost all goods in China. Two notable special cases are imports and exports (State Council, 1993). Import tariffs existed in China throughout this period and were deductible in the same manner as the original input value. Exports were awarded VAT rebates throughout the period of our study. Chinese export rebates are typically less than the total sum owed – i.e., firms pay some VAT on exports (Chandra and Long, 2013). Both import tariffs and export rebates vary across sectors (products) and over time.

In principle, tax officials are supposed to cross-validate trade flows by linking the customs and VAT data. However, as with other VAT transactions, this rarely occurred prior to computerization due to the administrative burden. Computerization should therefore also increase third-party information for importing and exporting firms.

There were no other changes to VAT in the period that we study, but changes in several other policies did take place. The two main policies that could confound our results are China’s entry to the World Trade Organization (WTO) in 2001, and the privatization of state-owned firms, often referred to as “enterprise reforms” during 1998-2003. We discuss these policies in detail and show that they are unlikely to confound our estimates in Section 4.4.

3 Data

We use data from the Annual Survey of Industrial Production (ASIP), 1998-2007. These data cover large manufacturing firms and have been widely used by studies of Chinese firms. Our main sample is a balanced panel of firms from 1998-2007 with annual revenues above 5 million RMB.
We also present results using the full sample of firms. The data contain a rich set of variables related to firm production and inputs. An important advantage for our paper is that it records net VAT payment, gross VAT-eligible sales and VAT-eligible deductions for intermediate inputs. Thus, we can examine the mechanisms through which computerization improves enforcement as well as its effect on VAT payment. The VAT payment variable is inclusive of rebates, such as those awarded for exports.

The ASIP is conducted each year by the National Bureau of Statistics (NBS). NBS officials visit each firm and copy data from firm records to the survey. The data collected in ASIP cannot be used in any legal action against firms (such as tax violations). The production and balance sheet data (e.g., total output, intermediate inputs, labor inputs) are in a different module of the ASIP than VAT and are typically transcribed from a different firm account book than the one that includes VAT payments. These data have been accepted by a large number of existing studies, including those studying firm productivity (e.g., Hsieh and Klenow, 2009) and tax compliance (Cai and Liu, 2009; Li et al., 2021).

All of the values in the paper are reported in real terms. To avoid outliers, our sample excludes observations with the top and bottom 1% values of VAT and sales each year. We use 4-digit Chinese Industry Classification sector definitions. The baseline sample fixes firm sectors to be the sector when the firm first appears in the sample.

Given our prior that firms were over-reporting VAT deductibles before computerization, one may be concerned that using the firm data in ASIP to calculate the non-deductible share (NDS) would introduce measurement error – i.e., confound real differences in NDS and evasion. To address this, we calculate sector-level NDS. We calculate pre-computerization sector-level NDS with the 1997 Chinese Input-Output (I-O) Table obtained from the China’s National Bureau of Statistics (1999). The data for the I-O table are collected in an independent process by a different group from the ASIP and are mainly used to tabulate national statistics and compute national GDP. The statistical office that collects the ASIP data and the one that constructs the I-O tables do not collaborate. The firm-level information used for the construction of the I-O table is not shared with the tax department and it cannot be used as evidence of tax evasion (China’s National Bureau of Statistics, 2009). Nevertheless, we address the possibility that the I-O table data are also confounded after we present the main results and show that our findings are qualitatively robust to using NDS calculated from the I-O tables from other countries as instrumental variables.

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19 The panel is not perfectly balanced because some variables are missing for some years. All firms in the sample have non-missing values for the key variables for at least nine of the ten years that we study.

20 We use deflators provided by the Penn World Tables. To the extent that one is concerned about region-specific changes in prices, we show that our results are robust to controlling for province-year fixed effects in Section 4.4.

21 The results are similar without dropping the outliers, but slightly less precise. They are available upon request.

22 In the 1997 Chinese I-O Table, there are 125 total listed inputs, 85 of which are VAT-deductible under Chinese tax law. The transaction-level data used to build the I-O table are not available to researchers.
Our empirical strategy exploits time variation in the introduction of computerization and cross-sectional variation in NDS. The logic for the latter follows from the background discussion that firms with fewer real deductibles (as a share of sales) were able to exaggerate their reported deductibles more prior to the reform. Thus, these firms were therefore treated more intensely by the reform.

The cross-sectional measure of intensity, \( \tilde{NDS}_s \), is denoted as:

\[
\tilde{NDS}_s = \left( \frac{\text{NonDeductible Inputs}}{\text{Total Sales}} \right)_s.
\]

This term is the ratio of total non-deductible inputs to total output in sector \( s \). To construct NDS by sector, we map each sector in the input-output tables into two groups, deductible or non-deductible, according to Chinese tax law. In practice, we consider inputs from agricultural, mining and manufacturing industries as materials, and thus deductible. We treat inputs from service industries, overhead, labor inputs, and value-added as non-deductible.

The correlation coefficient between our main measure of pre-computerization NDS calculated from the I-O tables and the measure calculated from ASIP is 0.34 and statistically significant at the 1% level. See Appendix Figure A.1.

### 3.1 VAT over time by NDS

To illustrate the variation behind our empirical strategy, Figure[1] plots average VAT over time for firms with above and below the sample median of NDS. Since average VAT payments are higher in the high-share group (2.19 million RMB, or 264,492.75 USD) than the low-share group (1.88 million RMB, or 227,053 USD), we normalize the 1998 data to zero for both groups. Consistent with high macroeconomic growth, the figure shows that VAT increased throughout the entire sample period for both groups. The trend was similar between the two groups prior to the reform and diverged after 2001, when the high NDS group, which was more intensely treated, experienced a larger increase.

Conceptually, our estimated effect of computerization on VAT compares the average difference between the two lines after the reform to the average difference before the reform. The similarity

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\[ \text{In a standard input-output table, the sum of all input values should be equal to the value of output. Therefore, to obtain the final measure, we can equivalently sum the fractions of inputs from deductible industries to obtain a single fraction for each industry that represents the share of inputs deductible under Chinese VAT rules. This object can be characterized by the following equation, where } D \text{ represents the set of deductible industries:} \]

\[
\tilde{NDS}_s = 1 - \sum_{d \in D} \text{Input fraction}_{sd}.
\]

Appendix Table [A.1] lists the fifty sectors with the highest and lowest values for \( \tilde{NDS}_s \).
in the pre-reform lines supports the parallel trends assumption of our empirical strategy. The
timing of the divergence supports our interpretation that the second difference captures the effects
of computerization rather than other changes that occurred before or afterward.

We can also check the logic of our empirical strategy and NDS measure by examining the
relationship between sector NDS from the I-O data and changes in effective VAT rate, measured
by VAT divided by sales. If high NDS firms were more intensely treated, then such firms should
experience systematically higher increases in VAT share. This is true in the data. A binned scatter
plot of these two variables is shown in Appendix Figure A.1 sub-Figure (a).

3.2 Correlates of NDS

Since NDS is not randomly assigned, one of the main concerns for our identification strategy is
that NDS is correlated with omitted variables, which would affect VAT (and the other outcomes
of interest) through channels other than computerization. Table 1 documents the differences between
high- and low-share sectors by estimating the correlation coefficient of NDS and a number of pre-
reform firm characteristics averaged at the sector level. For brevity, we focus on variables which we
later examine as outcomes. These cross-sector correlation coefficients show that firms in sectors
with high NDS on average pay higher VAT, pay higher VAT as a share of sales and pay fewer VAT
deductions. On average, firms in high NDS sectors have lower sales, fewer intermediate inputs and
are more productive. In Section 4.4 we show that the main results are robust to controlling for
these baseline characteristics interacted with year fixed effects.

4 The Average Effect on VAT

4.1 Baseline

The baseline estimate exploits two sources of variation: time variation from the 2001 introduction
of computerization and cross-sector variation in the intensity of the treatment effect. The latter
is motivated by the discussion in Section 2.1.1 that firms with higher NDS were more intensely
treated by computerization. The baseline equation can be written as the following.

\[ VAT_{ist} = \alpha + \beta \tilde{NDS}_s \times Post_t + \Gamma X_{ist} + \tau_t + \phi_i + \epsilon_{ist}, \]  

where VAT paid by firm \( i \), in sector \( s \), and year \( t \), \( VAT_{ist} \), is a function of: the interaction of a
dummy which takes the value of one if it is 2002 or later, \( Post_t \), and the measure of intensity at the
sector level, \( \tilde{NDS}_s \); firm fixed effects, \( \phi_i \); and year fixed effects, \( \tau_t \). We choose 2002 as the start of
the post-reform period because hand-written invoices were not banned until the end of 2001. When
we examine the dynamic effects, we allow the effects to differ for each year. Note that sector fixed effects are absorbed by firm fixed effects, as we code each firm’s sector as the one it has when first observed. The standard errors are clustered at the sector level.

\( X_{ist} \) is a vector of controls. For parsimony, the baseline only includes two measures of firm size to account for the possibility that tax policy varies by firm size (Bachas et al., 2019; Kleven et al., 2016a). The first is the average pre-reform sales and the second is the average pre-reform firm VAT. We control for the average value of each variable over 1998-2001 to avoid endogeneity and their interactions with year fixed effects to allow the influences to be completely flexible over time.

We are interested in the estimate of \( \beta \). If the reform increased compliance and VAT, then \( \beta > 0 \). Our identification strategy assumes parallel trends – i.e., absent the reform, the outcomes of interest across sectors with different NDS would have evolved along parallel trends (conditional on the controls). The descriptive statistics in the previous section support this assumption. We provide additional support after presenting the main results.

Table 2 examines the effect of computerization on VAT. The sample means of the dependent variables are stated at the top of the table. This section focuses on Panel A. Column (1) shows that the effect on gross VAT or VAT-eligible sales is negative, but statistically indistinguishable from zero. Column (2) shows that the reform reduces deductions. The estimate is statistically significant at the 1% level and is larger in magnitude than the estimated reduction in gross VAT in column (1). The reform, on average, reduced VAT deductions by 6,281 × 1,000 RMB (6.281 million RMB, or 765,975 USD) for a firm in a sector with no deductibles (i.e., the NDS is 100% of sales) relative to a firm in a sector where all sales are deductible (i.e., the NDS is 0% of sales). In terms of magnitudes, a back-of-the-envelope calculation shows that a firm with the sample mean NDS of 0.4042 would have experienced a 40.99% (6,281 × 0.4042/6,194 = 0.4099) decline in VAT deductions after computerization.

Column (3) shows that the reform increased VAT payment. The estimate is statistically significant at the 1% level. Our estimates imply that the elasticity of VAT sales with respect to the VAT rate is -0.92\(^{24}\). In terms of magnitudes, the treatment effect constitutes 33.4% of the increase in firm VAT from 1998 to 2007 in the balanced panel (see Appendix Section E). Later, in Section 4.2, we examine the full sample of firms to compute the contribution of the reform to the Chinese economy\(^{25}\).

Column (4) examines VAT as a share of sales. The denominator is total firm revenues, which is reported in a different module of the firm survey. Since not all sales are VAT eligible, we use total sales in the denominator to normalize by firm size. If VAT payments increase because of

\[^{24}\text{See Appendix Section E for calculations.}\]
\[^{25}\text{Estimates using log outcome variables are comparable and available upon request. We choose to not use levels without logs in the main estimates because it is easier to conduct the back-of-the envelope calculations later in the paper.}\]
a change in enforcement, we may expect it to increase as a share of sales. If, instead, we are capturing spurious trends due to general macroeconomic growth, then total firm revenues would also increase and we would find no effect on VAT as a share of total sales. Indeed, the coefficient is positive and statistically significant at the 1% level.

The results show that the reform increased VAT paid by firms and that the increase is driven by a decline in deductions. This is consistent with conventional wisdom that the reform mainly impacted firms by removing their ability to falsify invoices for deductions.

Another explanation for our results is that VAT increased because computerization lowered the marginal cost of reporting deductible inputs (Best et al., 2015). This is complementary to our interpretation of intentional evasion, but it is hard to see why high NDS firms are less lazy. If anything, one would think that such firms, which use fewer legitimate deductibles in the production process would be more inclined to skip filing deductible inputs. We discuss Panels B through E later in Section 4.4.2.

There are three factors to keep in mind for the interpretation of our results. The first is that we assume that the treatment effect is increasing with NDS. We can examine this assumption by creating dummy variables for the quartiles of NDS and estimating a specification like the baseline except that the main explanatory variables are the quartile dummy variables interacted with Postt. We show that the VAT and VAT/Sales results are increasing with quartiles of NDS in Appendix Table A.2.

Second, we do not have a pure control group. Sectors with low NDS may still have evaded VAT prior to the reform (albeit less than sectors with higher NDS). Thus, the reform will also increase their VAT compliance (though less than for sectors with higher NDS). The second caveat arises from the presence of cross-sector transactions. This is particularly relevant when the transactions become linked: higher compliance in sectors with high NDS will lead to higher compliance in sectors with low NDS. Both caveats will lead to our results to be an underestimate.

The third is the concern that despite the best efforts of the National Bureau of Statistics, the raw data used to generate the 1997 Chinese I-O tables are confounded by evasion and measurement error. To address this issue, we construct NDS measures using Mexican and U.S. input-output tables and use these to instrument for our main measure after we present the main results (see Section 4.4.2).

Finally, there is the concern of omitted variables – that the treatment is correlated with other factors which influence VAT. We address this issue in Section 4.4.

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26Note that an ostensibly reasonable alternative strategy is to use exporting sectors, which are commonly thought of as “exempt” to VAT, as a control group. However, in practice, the VAT rebates for exporting sectors vary over time and across sectors, and there is no one sector that is always VAT exempt. However, in a similar spirit, we estimate heterogeneous treatment effects of the reform according to the degree of exports in Subsection 4.5.
4.2 All Firms

To estimate the aggregate impact of computerization, we estimate the baseline with the full sample of firms that includes entry and exit. We replace firm fixed effects with sector fixed effects. We find qualitatively and quantitatively similar results. The coefficient on VAT in Table 6 column (1) is 1,634 and statistically significant at the 1% level. It is similar to the baseline coefficient of 1,839. We use the full sample estimates to calculate that the computerization of VAT contributed to 13.7% of average VAT growth from 1998 to 2007 and 11.7% of China’s VAT revenues in 2000.27

4.3 Tax Avoidance and Substitute Evasion

Firms may have responded to the increase in VAT by shifting into production with lower VAT burden. One way to do this is to increase exports, which were eligible for VAT rebates. In column (2) of Table 6 we find that export shares did not increase due to the treatment.28 This is consistent with conventional wisdom about the high fixed costs of exporting. Firms could have also shifted to sectors with lower VAT burdens. We investigate this by examining an indicator for whether a firm switched sectors after the reform.29 Table 6 column (3) shows that computerization increased the probability of changing sectors in the balanced panel of firms. The result is similar if we use a Logit specification (column 4).30

Besides VAT, the two other important taxes paid by firms are the enterprise (corporate) tax and the payroll tax. Firms may evade these taxes more when they find it more difficult to evade VAT. During this period, the enterprise tax was levied on profits, with rates between 15% and 33%, depending on firm ownership (Cai and Liu 2009; Chen et al. 2021a). In columns (5) and (6) of Table 6 we examine enterprise tax. In column (6), we add province-year fixed effects, which absorb changes in enterprise tax rates awarded to western provinces as part of the 2001 “Develop the West” campaign. In column (7), we study payroll tax by examining the firm wage bill. Payroll

27 See Appendix Section E for calculations.
28 This result holds if the sample is restricted to firms who ever exported prior to 2001. This result is available upon request.
29 Note that our baseline measure of sector is time-invariant; we use each firm’s first observed sector. The baseline non-deductible measure is assigned using this time-invariant measure. The 4-digit sector ID is assigned by the NBS official based on the share of production of each of the top three products of the firm.
30 The change in firm size and sectors can reflect real and/or reporting changes. We discuss this more in Section 5.2
For Japan, Onji (2009) finds that firms avoid VAT by splitting into smaller firms so that they fall to a threshold with a lower tax rate. This is unlikely to occur for our sample of large manufacturing firms. Falling below the threshold for VAT would trigger immediate official scrutiny for such large firms.
31 Off-setting tax evasion has been documented by several recent studies. Li et al. (2021) find that counties that lost more revenues from the abolition of the Agricultural Tax in 2005 experienced increases in VAT and reductions in payroll taxes. They interpret this as evidence of offsetting tax evasion. In other contexts, Carrillo et al. (2017) and Slemrod et al. (2017) find that more accurate reporting of firm revenues did not substantially increase business income tax because firms offset the increase in revenues by reporting more costs.
tax rates were unchanged during the period of our study. They were levied on workers’ wages and required employers to contribute 20% for pensions and 6-10% for health care. Direct data on payroll tax payments are only reported after 2001. The estimates show that computerization had no effect on enterprise or payroll tax. Thus, we find no effect on tax substitution.

4.4 Robustness

4.4.1 Omitted Variables

WTO Entry The main caveat for identification is omitted variables. An important and potentially confounding event was China’s entry into the WTO in 2001 and the ensuing changes in tariffs and trade flows. Though the economy-wide effect of entry into the WTO is absorbed by year fixed effects, the event could still confound our estimates if entry differentially changed VAT rebates, sales, or productivity for firms with high versus low NDS. We investigate this possibility in several ways. First, recall our earlier result that computerization has little effect on exporting firms. This is reassuring and supports our interpretation that the baseline is not confounded by trade exposure. Second, we control for import tariffs, export VAT rebates and export duties for each sector and year\[32\] Table 3, Panel A, column (2) shows that the results are very similar in magnitude to the baseline. Note that the number of observations changes slightly because of the limited availability of the tariff data. Third, we address sector-specific differences in export growth by controlling for average export growth rates in each sector in the pre-reform years interacted with year fixed effects. The coefficient in Panel A, column (3) is very similar to the baseline. Third, we control for the total amount of imports and exports in each four-digit Chinese Industrial Code sector and year\[33\] The estimate in Panel A, column (4) is very robust to these additional controls. Panel A column (1) reports the baseline for comparison.

State and Private Ownership Another relevant policy change during our period was the privatization of state-owned firms, often referred to as “enterprise reforms”, which took place during 1998-2003. The manufacturing sector transitioned from mostly publicly (state) owned to partly privately (not state) owned and some state firms closed down entirely.\[34\] To avoid potentially confounding effects from firm entries and exits, our analysis focuses on a panel of firms that exist throughout the period of our study.

We also address this by controlling for ownership interacted with year fixed effects. This addresses the possibility that privately owned firms and state-owned firms may have evolved dif-

\[32\]Rebate data are from Garred (2018). We use the method presented in Fan et al. (2015, 2018) to obtain output and input tariffs.

\[33\]These data are reported by China’s General Administration of Customs, 1998-2007.

\[34\]See, for example, Hsieh and Song (2017) for a detailed discussion.
ferently over time or that state-owned firms were affected by the reform differently from privately owned firms. For example, the enterprise reforms are usually considered to have taken place during 1998-2003. During this period, many state-owned firms were restructured, shut down or privatized (e.g., Hsieh and Song, 2017). One may naturally wonder whether such restructuring confounds the VAT reform that we study. Panel A, column (5) shows that the estimates when controlling for ownership-year fixed effects are very similar to the baseline.

**Competition** Table 3 Panel A, column (6) shows that our estimates are very similar if we control for the competitiveness of the sector measured using the Herfindahl-Hirschman Index (HHI) variable interacted with year fixed effects, which has been shown by Cai and Liu (2009) to influence corporate income tax evasion.

**Province-Specific Policies** To address the possibility that there are province-specific policy changes or differences in the implementation of the reform (e.g., some provinces used the linked transaction database before others), or changes in province-specific economic conditions, we control for province-year fixed effects. For example, Chen (2017) argued that the abolition of agricultural taxes in 2005 led tax authorities to supplement their lost income with other tax sources such as VAT. In 2004, the central government changed how it split the burden of VAT export rebates with province and local governments (Chandra and Long, 2013; Bai and Liu, 2017). Both reforms are national policy changes, but may have different effects across provinces depending on the degree to which the province relied on agricultural taxes or VAT.

Another potential concern arises from the granting of preferential corporate income tax rates to Western provinces in 2001. To the extent that firms substitute tax evasion between VAT and corporate income tax, this could confound our estimates.

One may also be concerned that prices change differentially across provinces since we deflate the variables with a national deflator.

To address the concern of province-year-specific confounders, we control for province-year fixed effects. Panel A, column (7) shows that our results are robust.

**VAT Pilot Provinces** As we discussed in Section 2, further changes in VAT policy made in 2009 (increasing the number of inputs that qualified for deductions) were piloted in three northeastern provinces (Liaoning, Heilongjiang and Jilin) starting in 2004 (Cai and Harrison, 2011; Liu and Mao, 2019; Liu and Lu, 2015). To investigate whether our main results are confounded by the

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35 We categorize official state-owned firms, collective ventures, and joint ventures as state-owned firms. We categorize private enterprises and limited-liability companies as private firms.

36 The Western provinces are Chongqing, Sichuan, Guizhou, Yunnan, Tibet, Shaaxi, Gansu, Ningxia, Qinghai, Xinjiang, Inner Mongolia and Guangxi.
pilot, we omit all observations from these provinces starting in 2004. Panel B, column (1) of Table 3 shows that the resulting estimate is very similar to the baseline.

**Local Enforcement** We also consider the possibility that the distance between a firm and the county seat (where tax officials have offices) influences its propensity and ability to evade VAT (Fan et al., 2020). We address the concern that distance is correlated with NDS by controlling for the interaction of the (travel) distance to the county seat and year fixed effects. The estimates in Panel B, column (2) are similar to the baseline.

Relatedly, in Panel B, column (3), we control for the pre-period share of county revenues from VAT to address the possibility that counties relying more on VAT revenues in the pre-period may have been different in unobservable ways. The coefficient and precision are similar to the baseline.

**Within-Sector Variation** In Panel B, column (4), we address the possibility that the I-O tables may have used sector-level averages to impute input and output composition. However, some sectors may have had higher or lower within-sector variation in NDS share. To account for the possibility that high- or low-variation sectors are not systematically different, we control for the within-sector standard deviation of firm NDS values. The baseline result remains precise and of similar magnitude.

**Correlates of NDS** Earlier in the paper, we document several correlates in Table 1. To ensure that these characteristics do not confound our results, we control for these firm-level characteristics interacted with year fixed effects in Table 3, Panel B, column (5). Specifically, we compute each firm’s 1998-2000 average value of VAT, VAT deductions, sales, intermediate inputs and TFPR and separately interact each with year fixed effects. We find that our main result remains precise, positive, and similar in magnitude.

**4.4.2 Measurement Error in NDS**

The baseline estimates calculate NDS with data from the 1997 Chinese I-O Table to capture real differences across sectors that would affect the incentives to evade VAT. We assumed that this measure avoided measurement error from the effects of evasion under pre-period Chinese tax rules. For measurement error in NDS to overturn our main finding, it would need to distort the ranking of NDS across sectors. This seems prima facie unlikely. Nevertheless, to be cautious, we check the validity of our finding by using NDS calculated from Mexican and U.S. Input-Output Tables as instrumental variables. The logic is that Mexican and U.S. NDS across sectors will reflect

37We have fewer observations in this exercise due to missing addresses for some firms.
real differences across sectors, but not capture the effects of evasion under pre-period Chinese tax rules.  

We begin by using NDS constructed from the 2000 I-O tables of Mexico, another middle-income economy, as an instrument. Table 2 Panels B and C present the 2SLS and reduced form results. They are similar to the baseline in sign and statistical precision for VAT deductions, VAT payment and VAT as a share of sales. The magnitudes are, if anything, larger. Next, we use data from the 2007 U.S. I-O Tables to construct an alternative instrument. We choose the year of 2007 because it reports data at a disaggregated level (405 sectors), which allows for better mapping to the Chinese tables. Table 2 Panels D and E present the 2SLS and reduced form estimates. The signs are consistent with the main results. The magnitudes are larger and statistical precision varies across columns. These estimates indicate that the main OLS results are not likely due to measurement error in the explanatory variable.

In Appendix Table A.3, we test whether the 2SLS and reduced form estimates using Mexican and U.S. NDS are confounded by trade exposure between China and these two countries. To do so, we sort sectors by trade exposure, as measured by the ratio of total bilateral trade between Mexico and China (imports plus exports) to total production in Mexico for that sector and year. Then, we omit from the sample the top 25% most exposed sectors. We repeat this process for U.S. to China trade flows. We find that the main results are robust to omitting sectors highly exposed to bilateral trade.

### 4.4.3 Clustered Standard Errors

The baseline estimates cluster the standard errors at the 4-digit sector level. Table 3, Panel B, column (6) presents clustered standard errors at the 2-digit sector level. Because there are just 29 2-digit sectors in our sample, we estimate wild bootstrapped standard errors (Cameron et al., 2008; Roodman, 2019). The p-values show that the estimates remain statistically significant at the 1% level.

### 4.4.4 Random Permutations of NDS

To test whether our baseline estimates are spuriously generated by the distribution of sectoral NDS, we perform a random permutation test in Appendix Subsection F.1. The baseline results are not likely to be driven by the distribution of sectoral NDS.

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38 We plot pre-2001 ASIP NDS against the U.S. and Mexican I-O measures in Appendix Figure A.1 sub-Figures (c) and (d). The correlation coefficients are respectively 0.34 and 0.22, both with \( p < 0.01 \).

39 The correlation coefficient between Mexican and Chinese NDS is 0.366, with a standard error of 0.0689. See the Appendix Section F for a detailed discussion of implementation.
4.5 Heterogeneous Treatment Effects

Exporting vs Non-exporting firms Exporting firms are eligible for VAT rebates before and after the reform. Thus, computerization should have limited effect on exporting firms. To investigate this, we compute the export share for each firm-year by dividing total exports by total revenues. We define non-exporters as firms that always have no exports and exporters as firms that ever have an export share of 50% or greater. Columns (1) and (2) in Panel A of Table 4 show that the positive effect of computerization on VAT is much larger and statistically significant for non-exporting firms. The coefficient in column (1) is 1,892 and is statistically significant at the 1% level. For exporters, it is much smaller in magnitude and statistically insignificant.\(^{40}\)

In Panel A, columns (3) and (4), we divide the sample according to the pre-period sectoral import input share. We divide firms into those that import above and below the pre-2001 median. We find that computerization increased VAT for both types of firms, but the increase is much larger for firms that import relatively little.

The estimates show that the reform increased VAT more for firms with less trade exposure and go against the concern that our findings are confounded by changes in global trade patterns triggered by China’s entry into the WTO. If that were the case, omitting the sectors that import or export more should weaken our results.

State-Owned versus Private Firms Given the difference in the amount of government attention and political connections between state-owned and privately owned firms, we divide the sample by ownership using the same ownership definition as in the earlier robustness exercises. The estimates in Table 4, Panel A, columns (5) and (6) show that the estimates are larger for privately owned firms than for state-owned firms. This could mean that private firms evaded more VAT prior to the computerization, or that state-owned firms had more leeway to evade after the reform due to their political connections to local tax officials.

Firm Size In Table 4, Panel B, columns (1) and (2), we allow the effects of computerization to vary by firm size. This is motivated by recent studies which find that firm size influences evasion\(^{41}\) (Bachas et al., 2019; Kleven et al., 2016a). We divide the sample into firms with average 1998-2001 total revenues that are above and below the sample median of this measure. The impact of computerization is similar for large and small firms.

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\(^{40}\)Export shares are calculated using our data. Imported input shares are calculated using Chinese Customs Administration data. Note that the sample median export share is zero, which is why the subsample in Table 4 column (2) is much larger than that of column (3). During the time period of study, the customs data were not linked automatically to the computerized VAT data (State Administration of Taxation, 2004). Cross-checking across the VAT and customs tax systems began in 2017 (State Administration of Taxation, 2017).

\(^{41}\)We also divide the sample according to the share of fixed assets for the median firm in a sector and find no difference. These results are not presented for brevity.
very large. If evasion differs for very small firms, we would not observe that in our data.

**Pre-period County VAT Revenue Share**  In Table 4, Panel B, columns (3) and (4), we examine whether the treatment effect differs by the extent of pre-period county VAT revenue share. We divide the baseline sample by the 2001 share of county-level revenue from VAT. We find that counties with lower pre-period VAT revenue shares exhibit a moderately larger treatment effect.

**Distance to the End Consumer**  Earlier studies of other contexts have found that the self-enforcing incentives of VAT differ for upstream firms than those closer to the consumer (Almunia and Lopez-Rodriguez, 2018; Mittal and Mahajan, 2017; Naritomi, 2019; Pomeranz, 2015, e.g.).

We use a sector-level measure of distance from the final consumer and test whether computerization increases VAT more for downstream firms. In Table 4, Panel B, columns (5) and (6), we find that downstream firms exhibit a slightly larger treatment effect, but the difference with upstream firms is not statistically significant. One reason for this muted difference could be that the overall variation in distance from the final consumer may not be large among large manufacturers. Our sample of large manufacturing firms does not include firms that typically interact with consumers directly, like retailers or service firms, and the pairwise correlation between sectoral “upstream-ness” and sectoral NDS is small and imprecise, at 0.0316 with \( p = 0.2461 \).

## 5  The Dynamic Effects on VAT and Firm Behavior

This section investigates the dynamic effects of computerization, which can change over time. In the short run, computerization will impact a firm’s reporting behavior by forcing firms to reduce falsified deductions. This leads to higher VAT payments. In the longer run, both reporting and economic behavior can change. The firm can change its economic behavior by adjusting inputs or outputs to avoid VAT. It can also learn new ways of evading VAT via two-sided collusion or start to evade other taxes (e.g., payroll taxes, corporate income taxes). This section uses the dynamic estimates to investigate the changes in economic behavior and reporting behavior in the longer run. The two types of changes in firm behavior are not mutually exclusive.

### 5.1 Real Economic Changes

To guide the investigation of economic effects, we develop a simple model of dynamic firm behavior that generates empirically testable predictions. The model also motivates the firm outcomes that we examine other than VAT.

\[ \text{Also, see Slemrod (2007) for a discussion.} \]
5.1.1 Conceptual Framework

To understand the potential implications of the increase in taxes on the economic behavior of the firm, we develop a simple theoretical framework. The formal model is presented in Appendix Section G. The intuition is summarized here.

Demand is downward-sloping and short-run supply is upward-sloping. Pre-tax prices and tax-exclusive prices are trivially equal in period 0, $q_0 = p_0$. When the tax, $\tau$, is imposed, the supply shifts upwards by the amount of the tax, since the marginal cost of production has increased by $\tau$. This shift increases the pre-tax equilibrium price to $q_1 > q_0$. The tax-exclusive price received by producers is $p_1$, with $q_1 = (1 + \tau) p_1$. Figure 2 shows that the tax-exclusive price will decrease to $p_1 < p_0$.

In the long run, the supply curve becomes more elastic, because we assume that capital (i.e., intermediate inputs) can only be adjusted in the long run. For simplicity, Figure 2 illustrates a perfectly elastic long-run supply curve. Since $q_0 = p_0$ is optimal, we simply rotate the supply curve around the initial point where supply and demand intersect. As with the short-run, the long-run response to the increase in taxes can be illustrated by shifting the supply curve up by the amount of the tax. The long-run tax-inclusive price will be $q_2 > q_1 > q_0$, while the long-run tax-exclusive price will be $p_2 = p_0$. Figure 2 illustrates the key intuition.

The simple model also predicts that labor input will decline over time. The intuition for this result comes from the observation that the short-run elasticity of labor is smaller than the long-run elasticity of labor (because capital can also be adjusted in the long run) holding pre-tax prices fixed. This effect implies that labor should react even more in the long run to the tax change than in the short run. In our setting, there is also an offsetting effect, since the increase in pre-tax prices calls for larger inputs, all things being equal. If demand is elastic, prices react little to changes in output, so that the first effect dominates. It follows with additional algebra that other inputs also decline over time.

Several empirically testable implications emerge from the model. First, tax revenues will increase from period 0 to period 1, and then decline in period 2 to a level between that of periods 0 and 1, such that $0 = \text{taxes}_0 < \text{taxes}_2 < \text{taxes}_1$.43 Second, the pre-tax price, which is algebraically equivalent to $TFPR$ as formulated in [Hsieh and Klenow (2009)], increases every period, $q_2 > q_1 > q_0$. Third, if the elasticity of demand, $\sigma$, is greater than 1, sales decline each period, $q_2 y_2 < q_1 y_1 < q_0 y_0$. Fourth, labor and intermediate inputs decline each period, $l_0 > l_1 > l_2$ and $k_0 \geq k_1 > k_2$.

The baseline model assumes a Cobb-Douglas production function with two factors, labor and intermediate inputs, and perfect competition. We provide several extensions to show that all of the

\footnotesize
\begin{equation}
\text{taxes}_t = \tau p_t y_t = \tau q_t y_t / (1 + \tau).
\end{equation}

\normalsize
main insights carry through with imperfect competition, endogenous input prices, or with three factors of production (labor, capital, and deductible inputs). See Appendix Section G.

Note that our model uses logged quantities for tractability. However, we have thus far presented our results in levels to simplify the accounting exercises in Section 4 and will continue to do so for comparability. The results are robust to using logged dependent variables; these are available upon request.

5.1.2 Dynamic Estimates

The first prediction of the simple model is that tax revenues should increase after the reform, but the long-run level – though still positive – will be slightly lower than the short-run level. We examine this by estimating an OLS equation similar to the baseline, except that we divide the seven-year post-reform period into three sub-periods: 2002-2003, 2004-2005, and 2006-2007. Table 5, column (3) shows that the reform increases VAT payment in all three post-reform periods. The revenues rise from the first to the second period, which most likely reflects the phasing in of computerization. Then, consistent with the theory, it declines in the third period. However, the decline is statistically insignificant (see the p-value at the bottom of the table) and the magnitude of the coefficient in the third period (2,047) is still large and very similar to the one in the peak period (2,267). These results show that the decline in VAT predicted by the theory is small in magnitude and the positive impact on VAT revenue persists over time.

As with the earlier estimates, we also examine gross VAT and eligible deductions in columns (1) and (2). The estimates for gross VAT are negative, but statistically indistinguishable from zero. The estimates for deductions are negative, statistically significant starting in 2002 and larger in magnitude than the decline in sales. Column (4) examines VAT as a share of sales. The temporal patterns are similar. The magnitudes of the coefficients increase from the first to the second period, and then remain nearly the same in the third period.

Table 7 examines other firm outcomes. Recall that these outcomes are reported in a different module of the ASIP and recorded from a different set of firm accounting records than the VAT variables. Column (1) examines total sales, which are the annual revenues of the firms. As the sample mean on top of the table shows, this is on average four times larger than VAT eligible sales since it includes revenues from items that are not part of VAT.

In column (2), we find that the component of sales that is not part of the VAT base, which we call “ineligible sales”, also declines.

Note that because our empirical strategy relies on cross-sector as well as time variation, the results, taken literally, will also reflect the ability of factors to reallocate across sectors. For simplicity, our baseline model does not take this additional mechanism into account. The extension is straightforward and available upon request. All of the insights carry through.

We compute ineligible sales in two steps. First, we take Gross VAT, which is separately reported in the firm data,
Column (3) examines the number of employees as a proxy for labor input because the large amount of non-wage compensation in large Chinese firms makes the wage bill a poor proxy for labor input. Columns (4)-(5) examine intermediate inputs, first in levels, and then as a share of total inputs. As with sales, this includes intermediate inputs that are not eligible for VAT.

Column (6) examines intermediate inputs which are deductible from VAT as a share of total inputs. The results show that the reform reduced sales, intermediate inputs and deductible input share. The estimate on number of employees is negative but not statistically significant.

The coefficients for sales and intermediate inputs in columns (1) and (4) are statistically significant in periods 2 and 3. They are also increasing in magnitude over time. The p-values at the bottom of the table show that the increases from period 1 to 3 and from periods 2 to 3 are statistically significant at the 5% level. These results are consistent with the fact that it takes time to adjust real production and that it becomes easier to adjust in the long run.

In column (7), we examine total factor productivity of revenue (TFPR) that accounts for endogenous markups (De Loecker and Warzynski, 2012). We find that the reform increases productivity. Since productivity is intuitively output normalized by inputs, this conceptually reflects the fact that the reform reduced inputs more than it reduced output. The results are consistent with model predictions.

Figure 3 presents the year-by-year effects. For brevity, we focus on the outcomes with clear predictions from the model. We re-estimate the baseline equation except that we replace the interaction term, $\tilde{NDS}_s \times Post_t$, with the interaction of $\tilde{NDS}_s$ and each year dummy variable. 2001 is the reference year. The coefficients and standard errors are presented in Appendix Table A.4.

Figure 3a shows that there is no pre-trend for VAT. All of the coefficients prior to 2001 are statistically zero. After 2001, the effect of computerization increases over time until around 2005, after which there is a slight decline. These estimates are consistent with the predictions of the model. Moreover, they show that the long-run decline in VAT is quantitatively unimportant.

Figures 3b to 4f examine total sales, the number of employees, total intermediate inputs and TFPR. For each outcome, the estimate is statistically zero prior to computerization and we see a change in the direction predicted by the model after the reform – output, the number of employees and intermediate inputs decline, although the estimates for employees are imprecise. TFPR starts to increase in 2003. The lack of pre-trends supports the parallel trends assumption.

We subject the dynamic effects to the same large set of controls from Section 4.4. Appendix Tables A.6 and A.7 show that these additional controls do not substantially change the magnitudes or the temporal patterns of the estimates. We also show that the estimates are qualitatively similar and divide by 17%. This yields the implied sales tax base. Then, we subtract this implied sales tax base from total sales. The result represents the portion of sales that is not included in the VAT tax base.

46 Appendix Table A.5 presents the average effects of computerization on firm outcomes for comparison.
if we instrument for the Chinese NDS with Mexican and U.S. NDS in Appendix Tables A.8 and A.9. The 2SLS estimate has three endogenous variables (the interactions of Chinese NDS with each of the three time periods) and three instruments (the interactions of Mexican or U.S. NDS with each of the three time periods).

The dynamic estimates are consistent with real economic changes. It is interesting to note that VAT deductions, sales, and intermediate inputs in Tables 5 and 7 all decline steadily in the six years after computerization. Since computerization had an immediate effect in raising the difficulty of falsifying VAT claims, the continued change over time cannot be explained by simple misreporting. We discuss this more in the next section.

5.2 Reporting Changes

This section investigates the possibility that firms developed new strategies to evade VAT after computerization.

Under-reporting Sales

After computerization increased the difficulty of exaggerating deductibles, firms may have shifted to understating sales. As we discussed earlier, sudden declines in sales triggered audits. If firms gradually reduce their reported sales over time to avoid audits, then one may observe patterns similar to the dynamic estimates on sales shown in Table 7 column (1).

To investigate this possibility, we separately examine sales that are eligible and ineligible for VAT. Firms have an incentive to under-report eligible sales, but not ineligible sales. In fact, firms had strong incentives to re-classify eligible sales as ineligible sales to reduce VAT and maintain steady overall revenue which reduces the chances of audits. Table A.10 column (2) shows that ineligible sales decrease and the decline grows in magnitude and becomes statistically precise five to six years after computerization. This goes against the possibility that the decline in total sales in column (3) is entirely due to reporting changes.

Two-sided Collusion

As we discussed earlier in Section 2 after the computerization of VAT invoices, evasion required firms in the same value chain to collude. The officials and firm managers that we interviewed did not provide examples of this in China. But earlier studies have documented the possibility of conducting two-sided collusion with flying invoices in other contexts (Waseem, 2020). Intermediate firms would report some of their sales to a retailer as sales to an exporter. Though the retailer loses tax credit from this loss of reported purchases, keeping some input “off

\footnote{See also Naritomi (2019) and Pomeranz (2015).}
the books” would allow it more scope to understate its revenues. For the exporter, additional input costs reduce its tax burden.

This strategy could generate a decline in VAT over time, though it would not directly generate our other results, such as the decrease in sales, ineligible sales or the increase in productivity. If flying invoices were used to evade VAT, then computerization should increase inputs more for firms that export more. We examine this with the data, and consider firms with export shares of greater than 0, 0.5, and 0.9. The results presented in Table 8 show that computerization does not increase inputs for exporting firms. It is therefore unlikely that flying invoices were a major form of adaptation in this context.

6 Conclusion

Tax enforcement is a central concern for all governments, particularly those of developing countries. The perennial questions for policymakers are how to improve enforcement and how much firms can offset the enforcement by changing their reporting or economic behaviors.

The results of this paper show that computerizing VAT transactions led to tremendous gains in Chinese government tax revenues. The evidence suggests that firms responded in the long run by reducing output and moving into sectors with lower VAT liability, but the tax revenue gains persisted.

It is important to keep in mind that the estimates of this study are specific to our context. In particular, there are two features of our environment worth noting. The first is that we study very large manufacturing firms. Unlike small firms, they cannot easily exit the formal sector to avoid taxes. The second is that Chinese tax authorities are able to punish firms for tax evasion. The increase in third-party information would be much less useful if it were not backed by credible enforcement. Therefore, our results are most relevant for large firms in middle-income countries that have some degree of state capacity.

For policymakers, the results suggest that technological advances can have significant impact on state capacity. 75% of tax-related World Bank projects in the 1990s involved record computerization, and 12.8% of the World Bank’s Global Tax Program budget in 2021 was still devoted to digitization (World Bank, 2012, World Bank, 2022). Our findings suggest that these initiatives are likely to generate tax revenue gains.

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48 Bruhn and McKenzie (2014) review this literature. Also, see dePaula and Scheinkman (2010), which theorizes that taxation could increase formalization in developing countries.
### Table 1: Correlates of NDS and Pre-Reform Firm Characteristics

<table>
<thead>
<tr>
<th>Pre-Reform Sector Mean</th>
<th>Sector-Level Non-Deductible Share</th>
</tr>
</thead>
<tbody>
<tr>
<td>VAT</td>
<td>0.1924***</td>
</tr>
<tr>
<td>VAT Gross</td>
<td>-0.0764</td>
</tr>
<tr>
<td>VAT Deductions</td>
<td>-0.159**</td>
</tr>
<tr>
<td>Sales</td>
<td>-0.1443**</td>
</tr>
<tr>
<td>Employees</td>
<td>0.0071</td>
</tr>
<tr>
<td>Wage Bill</td>
<td>0.0325</td>
</tr>
<tr>
<td>Intermediate Inputs</td>
<td>-0.1774**</td>
</tr>
<tr>
<td>Export Share</td>
<td>-0.0778</td>
</tr>
<tr>
<td>TFPR DLW</td>
<td>0.2412***</td>
</tr>
</tbody>
</table>

*Notes: This table presents the standardized bivariate correlation coefficients between the non-deductible share and the sector mean of key variables measured in 1998-2000. *** p<0.01, ** p<0.05, * p<0.1*
Table 2: The Average Effect of Computerization on VAT

<table>
<thead>
<tr>
<th>Dep Var Mean</th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>VAT Gross (1,000 RMB)</td>
<td>7,758</td>
<td>6,194</td>
<td>2,043</td>
<td>0.0418</td>
</tr>
<tr>
<td>VAT Deductions (1,000 RMB)</td>
<td>-3,974</td>
<td>-6,281***</td>
<td>1,839***</td>
<td>0.0228***</td>
</tr>
<tr>
<td>VAT (1,000 RMB)</td>
<td>(2,492)</td>
<td>(2,144)</td>
<td>(568.2)</td>
<td>(0.00579)</td>
</tr>
<tr>
<td>VAT/Sales</td>
<td>0.628</td>
<td>0.503</td>
<td>0.702</td>
<td>0.570</td>
</tr>
</tbody>
</table>

A. OLS

| Non-deductible share × Post-2002 | -6,133 | -13,313** | 4,938** | 0.0563** |
|                                | (6,634) | (6,108) | (2,242) | (0.0230) |
| Observations | 180,026 | 180,026 | 180,026 | 180,026 |
| R-squared | 0.628 | 0.503 | 0.702 | 0.570 |

B. Mexico: 2SLS

| U.S. Non-deductible share × Post-2002 | -1,258 | -2,732** | 1,013*** | 0.0115*** |
|                                       | (1,331) | (1,088) | (385.8) | (0.00385) |
| Observations | 180,026 | 180,026 | 180,026 | 180,026 |
| R-squared | 0.628 | 0.503 | 0.702 | 0.570 |

C. Mexico: Reduced Form

| Non-deductible share × Post-2002 | -13,545*** | -16,575*** | 2,218 | 0.0500*** |
|                                 | (5,031) | (4,190) | (1,562) | (0.0143) |
| Observations | 180,103 | 180,103 | 180,103 | 180,103 |
| Kleibergen-Paap F-statistic | 32.53 | 32.53 | 32.53 | 32.53 |

D. U.S. 2SLS

| Non-deductible share × Post-2002 | -3,428*** | -4,194*** | 561.3 | 0.0126*** |
|                                 | (1,304) | (1,064) | (384.3) | (0.00331) |
| Observations | 180,103 | 180,103 | 180,103 | 180,103 |
| R-squared | 0.628 | 0.503 | 0.701 | 0.570 |

E. U.S. Reduced Form

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1 In Panels B through E, we use Mexican and U.S. non-deductible share x post-2002.
Table 3: The Average Effect of Computerization on VAT – Robustness to Additional Controls

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Dependent Variable:</strong></td>
<td>VAT (1,000 RMB)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Panel A</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Baseline</td>
<td></td>
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<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Export Rebates, Import and Export Duties</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Export Growth 1998-2000 x Year</td>
<td></td>
<td></td>
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<td></td>
</tr>
<tr>
<td>Sector-Year Imports and Exports</td>
<td></td>
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</tr>
<tr>
<td>Ownership Category x Year FE</td>
<td></td>
<td></td>
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<td></td>
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<tr>
<td>HHI 1998-2000 x Year FE</td>
<td></td>
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<td></td>
</tr>
<tr>
<td>Province FE x Year FE</td>
<td></td>
<td></td>
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<td></td>
<td></td>
</tr>
<tr>
<td><strong>Non-deductible share × Post-2002</strong></td>
<td>1,839*** (568.2)</td>
<td>1,712*** (540.8)</td>
<td>1,858*** (562.1)</td>
<td>1,793*** (546.1)</td>
<td>1,844*** (553.7)</td>
<td>1,848*** (567.2)</td>
<td>1,767*** (561.2)</td>
</tr>
<tr>
<td><strong>Observations</strong></td>
<td>180,103</td>
<td>180,012</td>
<td>180,103</td>
<td>180,103</td>
<td>180,102</td>
<td>180,103</td>
<td>180,075</td>
</tr>
<tr>
<td><strong>R-squared</strong></td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.706</td>
</tr>
<tr>
<td><strong>Panel B</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Liaoning, Jiling, Heilongjiang</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Firm Distance from County Seat x Year</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2001 Share of County Revenues from VAT</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Sectoral SD of Firm NDS x Year FE</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Pre-Reform Firm Correlates x Wild Bootstraps</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Non-deductible share × Post-2002</strong></td>
<td>1,834*** (575.0)</td>
<td>1,990*** (576.0)</td>
<td>1,906*** (581.6)</td>
<td>1,624*** (578.6)</td>
<td>2,039*** (578.5)</td>
<td>1,839*** (0.008)</td>
<td></td>
</tr>
<tr>
<td><strong>Observations</strong></td>
<td>177,026</td>
<td>158,820</td>
<td>135,378</td>
<td>158,812</td>
<td>158,865</td>
<td>180,103</td>
<td></td>
</tr>
<tr>
<td><strong>R-squared</strong></td>
<td>0.703</td>
<td>0.710</td>
<td>0.708</td>
<td>0.710</td>
<td>0.711</td>
<td>0.702</td>
<td></td>
</tr>
</tbody>
</table>

**Notes:** The sample is a balanced panel of firms covering 1998-2007. Column (2) pre-reform correlates are firm-level averages from 1998-2000 of VAT, VAT deductibles, sales, intermediate inputs, and TFPR DLW. Each variable is separately interacted with year fixed effects. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. Standard errors are reported in parentheses, while p-values are reported in brackets. *** p<0.01, ** p<0.05, * p<0.1
Table 4: The Heterogeneous Effects of Computerization on VAT

<table>
<thead>
<tr>
<th>Panel A</th>
<th>Dep Var Mean</th>
<th>Non-Exporters (exports/total sales never &gt; 0%)</th>
<th>Exporters (exports/total sales ever &gt;= 50%)</th>
<th>Imported input share &lt; median</th>
<th>Imported input share &gt;= median</th>
<th>State Owned</th>
<th>Privately Owned</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1) (2) (3) (4) (5) (6)</td>
<td>(1,000 RMB)</td>
<td>(1,000 RMB)</td>
<td>(1,000 RMB)</td>
<td>(1,000 RMB)</td>
<td>(1,000 RMB)</td>
<td>(1,000 RMB)</td>
</tr>
<tr>
<td></td>
<td>Dep Var Mean</td>
<td>1790</td>
<td>1626</td>
<td>2077</td>
<td>2010</td>
<td>1706</td>
<td>2233</td>
</tr>
<tr>
<td></td>
<td>Non-deductible share × Post-2(2(2)**</td>
<td>1,892***</td>
<td>972.7</td>
<td>2,647***</td>
<td>1,082**</td>
<td>1,381***</td>
<td>2,159***</td>
</tr>
<tr>
<td></td>
<td>Observations</td>
<td>81,825</td>
<td>54,898</td>
<td>88,542</td>
<td>91,561</td>
<td>65,406</td>
<td>64,718</td>
</tr>
<tr>
<td></td>
<td>R-squared</td>
<td>0.709</td>
<td>0.623</td>
<td>0.718</td>
<td>0.684</td>
<td>0.759</td>
<td>0.754</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Panel B</th>
<th>Dep Var Mean</th>
<th>Sales &lt; median</th>
<th>Sales &gt;= median</th>
<th>2001 County VAT Revenue Share &lt; median</th>
<th>2001 County VAT Revenue Share &gt;= median</th>
<th>Distance to final consumer &lt; median</th>
<th>Distance to final consumer &gt;= median</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1) (2) (3) (4) (5) (6)</td>
<td>(1,000 RMB)</td>
<td>(1,000 RMB)</td>
<td>(1,000 RMB)</td>
<td>(1,000 RMB)</td>
<td>(1,000 RMB)</td>
<td>(1,000 RMB)</td>
</tr>
<tr>
<td></td>
<td>Dep Var Mean</td>
<td>1742</td>
<td>2340</td>
<td>2106</td>
<td>2001</td>
<td>2036</td>
<td>2051</td>
</tr>
<tr>
<td></td>
<td>Non-deductible share × Post-2(2(2)**</td>
<td>2,676***</td>
<td>2,243**</td>
<td>2,305***</td>
<td>1,516***</td>
<td>2,657***</td>
<td>1,355**</td>
</tr>
<tr>
<td></td>
<td>Observations</td>
<td>89,289</td>
<td>90,814</td>
<td>72,439</td>
<td>105,855</td>
<td>93,358</td>
<td>86,207</td>
</tr>
<tr>
<td></td>
<td>R-squared</td>
<td>0.686</td>
<td>0.707</td>
<td>0.709</td>
<td>0.713</td>
<td>0.725</td>
<td>0.695</td>
</tr>
</tbody>
</table>

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1
Table 5: The Dynamic Effects of Computerization on VAT

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>VAT Gross</td>
<td>VAT Deductions</td>
<td>VAT</td>
<td>VAT/Sales</td>
</tr>
<tr>
<td>Non-deductible share × 2002-2003 (β1)</td>
<td>-1,341</td>
<td>-2,743**</td>
<td>1,203**</td>
<td>0.0158***</td>
</tr>
<tr>
<td></td>
<td>(1,432)</td>
<td>(1,081)</td>
<td>(551.9)</td>
<td>(0.00492)</td>
</tr>
<tr>
<td>Non-deductible share × 2004-2005 (β2)</td>
<td>-4,583*</td>
<td>-5,612***</td>
<td>2,267***</td>
<td>0.0261***</td>
</tr>
<tr>
<td></td>
<td>(2,347)</td>
<td>(2,105)</td>
<td>(594.4)</td>
<td>(0.00696)</td>
</tr>
<tr>
<td>Non-deductible share × 2006-2007 (β3)</td>
<td>-5,962</td>
<td>-10,388***</td>
<td>2,047***</td>
<td>0.0264***</td>
</tr>
<tr>
<td></td>
<td>(4,299)</td>
<td>(3,774)</td>
<td>(720.8)</td>
<td>(0.00693)</td>
</tr>
<tr>
<td>Observations</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.628</td>
<td>0.503</td>
<td>0.702</td>
<td>0.570</td>
</tr>
<tr>
<td>H0: β1=β2 (p-value)</td>
<td>0.0620</td>
<td>0.0770</td>
<td>0.00700</td>
<td>0.0190</td>
</tr>
<tr>
<td>H0: β2=β3 (p-value)</td>
<td>0.571</td>
<td>0.0250</td>
<td>0.556</td>
<td>0.919</td>
</tr>
<tr>
<td>H0: β1=β3 (p-value)</td>
<td>0.216</td>
<td>0.0250</td>
<td>0.147</td>
<td>0.0510</td>
</tr>
</tbody>
</table>

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1
Table 6: The Average Effect of Computerization on Corporate Income Tax, Moving Sectors, Export Share

<table>
<thead>
<tr>
<th>Dependent Variable</th>
<th>VAT (Full Sample)</th>
<th>Export Share</th>
<th>Indicator for Sector Change</th>
<th>Indicator for Sector Change (Logit)</th>
<th>Enterprise Income Tax (1000 RMB)</th>
<th>Control for Province-Year FE</th>
<th>Wage Bill (1,000 RMB)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
<td>(5)</td>
<td>(6)</td>
<td>(7)</td>
</tr>
<tr>
<td>Dep Var Mean</td>
<td>1538</td>
<td>0.212</td>
<td>0.084</td>
<td>0.204</td>
<td>603.98</td>
<td>603.98</td>
<td>3847.5</td>
</tr>
<tr>
<td>Non-deductible share × Post-2002</td>
<td>1,634***</td>
<td>-0.0477</td>
<td>0.0793***</td>
<td>1.423***</td>
<td>308.1</td>
<td>409.7</td>
<td>338.1</td>
</tr>
<tr>
<td></td>
<td>(408.7)</td>
<td>(0.0493)</td>
<td>(0.0346)</td>
<td>(0.419)</td>
<td>(275.1)</td>
<td>(267.7)</td>
<td>(897.7)</td>
</tr>
<tr>
<td>Observations</td>
<td>711,643</td>
<td>180,103</td>
<td>161,512</td>
<td>66,933</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.612</td>
<td>0.834</td>
<td>0.240</td>
<td>0.459</td>
<td>0.466</td>
<td>0.733</td>
<td></td>
</tr>
</tbody>
</table>

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1
Table 7: The Dynamic Effects of Computerization on Firm Behavior

<table>
<thead>
<tr>
<th>Dep Var Mean</th>
<th>Dependent Variable</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1) (2) (3) (4) (5) (6) (7)</td>
</tr>
<tr>
<td>Sales (1,000 RMB)</td>
<td>58545</td>
</tr>
<tr>
<td>Ineligible Sales (1,000 RMB)</td>
<td>-11,020</td>
</tr>
<tr>
<td>Employees (persons)</td>
<td>(7,901)</td>
</tr>
<tr>
<td>Intermediate Inputs (1,000 RMB)</td>
<td>-26,978**</td>
</tr>
<tr>
<td>Non-deductible share × 2004-2005 (β2)</td>
<td>(13,228)</td>
</tr>
<tr>
<td>Non-deductible share × 2006-2007 (β3)</td>
<td>-50,019**</td>
</tr>
<tr>
<td>Observations</td>
<td>180,103</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.769</td>
</tr>
<tr>
<td>H0: β1=β2 (p-value)</td>
<td>0.101</td>
</tr>
<tr>
<td>H0: β2=β3 (p-value)</td>
<td>0.0830</td>
</tr>
<tr>
<td>H0: β1=β3 (p-value)</td>
<td>0.0710</td>
</tr>
</tbody>
</table>

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1
Table 8: The Average Effect of Computerization on Inputs Among Exporters

<table>
<thead>
<tr>
<th></th>
<th>Export Share &gt; 0</th>
<th>Export Share &gt; 0.5</th>
<th>Export Share &gt; 0.9</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Total Inputs</td>
<td>Intermediate Inputs</td>
<td>Total Inputs</td>
</tr>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
</tr>
<tr>
<td>Dep Var Mean</td>
<td>47378</td>
<td>40590</td>
<td>47378</td>
</tr>
<tr>
<td></td>
<td>(18,412)</td>
<td>(17,582)</td>
<td>(19,630)</td>
</tr>
<tr>
<td>Observations</td>
<td>65,255</td>
<td>65,255</td>
<td>34,248</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.843</td>
<td>0.831</td>
<td>0.846</td>
</tr>
</tbody>
</table>

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1
Figure 1: VAT Levels over Time for Firms with NDS Above and Below the Sample Median

Notes: For each of the two groups of firms, the 1998 mean is subtracted from the yearly value.
Figure 2: Illustration of Short- and Long-run Responses to VAT

Short-run supply

Long-run supply

Demand \( (slope = -1/\sigma) \)
Figure 3: The Yearly Effect of Computerization on VAT, Revenue, Employment, Intermediate Inputs, Productivity
References


Roodman, David, “BOOTTEST: Stata module to provide fast execution of the wild bootstrap with null imposed,” 2019.


APPENDIX

A  Other Enforcement Mechanisms (Audits)

The Chinese State Administration for Taxation (SAT) is a large bureaucracy. This fact can be observed in data reported by the China Tax Audit Yearbook Committee (2007). On average, there are 12,688 tax officials per province during the period that we study. Notably, it did not decline after computerization. On average, the SAT conducted 22,999 audits per province per year.\textsuperscript{49}

The China Tax Audit Yearbook Committee (2007) also reports the number of audits that indicate problematic behavior and the number of cases that are fined. On average, the percentage of problematic cases before computerization is 7.7% of all filings (China Tax Audit Yearbook Committee, 2007). However, very few of these are prosecuted and the fines are moderate. For example, in 1997, the penalty as a share of the SAT’s estimate of unpaid taxes was below 5%, which was lower than the interest rate (China Tax Audit Yearbook Committee, 2002); in 2002, among the 112,984 tax-fraud cases investigated by tax officials, only 2,658 cases were prosecuted (China Tax Audit Yearbook Committee, 2003). This is reportedly due to the difficulty of providing conclusive evidence and the inability of the tax authority to accurately estimate the amount of evasion without the true transaction amounts or linked transactions.

Additionally, we test whether the allocation of tax officials changed after computerization in Appendix Table A.11. We obtain data for the number of tax personnel in each province and year from the Tax Yearbook of China, 1998-2007. There are only a few missing observations. We regress the number of tax officials on province average NDS, as well as other variables that might affect the probability a firm would be audited: ruggedness, the geographic size of the province, the total province population, and the number of firms in a province.\textsuperscript{50} The data are at the province and year level. To focus on cross-province variation, we control for year fixed effects. Column (1) shows pre-2002 correlations. The coefficient for province average NDS is negative, which means that provinces with firms in higher VAT share sectors had fewer tax personnel who could conduct manual audits. Other factors also correlate with the number of tax personnel in the way that one would expect. Provinces that are larger, have a higher population, and have a greater firm density have more tax officials.\textsuperscript{51}

\textsuperscript{49}The personnel data are reported by the Tax Yearbooks of China; the audit rate data are reported by the Tax Audit Yearbooks of China.

\textsuperscript{50}Ruggedness is computed using ArcGIS by the authors. The size of the province is reported by the China Statistical Yearbook, 2000. We calculate VAT Share and the number of firms per province and year using the full sample (not the balanced panel that we use for the regressions) of our main dataset.

\textsuperscript{51}Since we also control for province size, we interpret the coefficients on population and firms as the effect of population and firm densities.
Note that the R-squared in column (1) is 0.875, implying that our crude controls explain 87.5% of the cross-province variation in tax personnel. Moreover, the standardized coefficient for NDS, which estimates the effect in terms of standard deviations, shows that a one standard deviation increase in VAT share reduces the number of tax officials by 0.28 standard deviations. This is sizable in terms of magnitude.

In column (2), we examine the post-computerization period. The estimates share signs, though the negative correlation between province NDS and tax officials somewhat attenuates after 2002 to a standardized beta coefficient of $-0.07$.

These results are consistent with the anecdotal evidence that the low level of enforcement prior to computerization focused on sectors with high deductible shares, i.e., low VAT share, and that there may have been some change in the allocation of tax officers after computerization was introduced.

**B VAT Deductibles**

The regulation that governs VAT remittance rules during the study period is the *Provisional Regulations of the People’s Republic of China on Value-Added Tax* (State Council Order 134, published in December 1993). The rules were effective between Jan 1, 1994, and Jan 1, 2009, when these Regulations were amended for the first time. The Regulations specifies the deductible items for VAT, which are not exactly the same as in other countries. The general principle is that any purchases that come with VAT special invoices, regardless of whether they originate from a domestic or international seller, can be deducted from the VAT duty. Full deductions are allowed for manufactured inputs, repair inputs, retail inputs, and wholesale inputs. Partial deductions are allowed for some “necessity goods” (including agricultural products, oils, gas, books, fertilizers, and salt) at a rate of 13%, for old and waste materials at a rate of 10%, and for transportation costs at a rate of 7%. No deductions are allowed for labor costs, capital (fixed asset) purchases, capital depreciation, abnormal losses, rent, fringe benefits, interests from bank loans, and overhead/operating expenses. Three Northeastern provinces, namely Liaoning, Jilin, and Heilongjiang, experimented with variants of VAT reforms in eight sectors in 2004 that allowed for deductions of fixed asset purchases. However, this did not affect other regions until 2009.

**C Data**

The unit of observation in the *Annual Survey of Industrial Production* (ASIP) is the firm. Because of varying English translations, these data are referred to also as the *Census of Manufacturing Firms* or the *Annual Survey of Manufacturers*. Subsidiaries are coded as separate entities as long
as they are unique legal units\textsuperscript{52} The inclusion and exclusion criteria for non-state-owned firms are asymmetric. The dataset includes all state-owned manufacturing firms (regardless of size) and non-state manufacturing firms with sales greater than five million RMB (603,865 USD).

The five-million RMB revenue threshold for non-state-owned firms is not systematically imposed: we observe non-state-owned firms below this threshold (with no apparent pattern in firm attributes). To avoid selective sampling, we impose a uniform cutoff and drop all observations with less than five million RMB (603,865 USD) in revenues.

Otherwise, we follow the standard procedure for cleaning the ASIP data, as first done by Cai and Liu (2009). We drop observations for which any reported sub-component of assets is greater than total assets, as well as observations for which the start month does not fall between 1 and 12. We also drop observations for which the founding year of the firm is greater than the year of the survey. We remove the influence of extreme outliers, which are likely to represent coding errors in these self-reported data. We drop the top and bottom 1\% of observations for the variables VAT and sales.

We construct measures of NDS for several countries using the World Input-Output Tables (Di-etztenbacher et al., 2013; Timmer et al., 2016). We use input-output tables from the year 2000 to construct these alternative country NDS, as that is the first available year for which the tables are reported with 56 sectors. For years prior to 2000, the World Input-Output Tables are reported at an aggregation of 35 sectors. We create a correspondence between the 56 input sectors and whether each sector would legally be considered a non-deductible input type under Chinese tax law in the year 2000. Then, we compute the sector-and-country-specific share of each industry’s inputs that are not deductible.

### D Productivity Estimation

We estimate total factor revenue productivity (TFPR DLW) using the method of De Loecker and Warzynski (2012), which is itself based on the method of Ackerberg et al. (2015).

Productivity is obtained as a residual from a value-added production function for each two-digit sector in the Chinese Industrial Codes. We deflate the nominal values of output and inputs separately using sector-level price indices. For the production function, we assume a translog form, \( y = \beta_1 l + \beta_{1k} k + \beta_2 l^2 + \beta_{2k} k^2 + \beta_{1lk} lk \) as do De Loecker and Warzynski (2012) for their specification IV. They show that productivity is highly robust to the choice of production function. We replicate our main results using alternative choices of production function and these results are available upon request.

\textsuperscript{52}For regulatory reasons, most subsidiaries are separate legal entities in China.
We use log real capital as our measure of capital as in Brandt et al. (2012), intermediate inputs as our measure of material inputs, and number of employees as our measure of employment.

E Magnitude Calculations

We benchmark our treatment effect in two ways. First, we multiply the full sample coefficient on VAT, 1,634, by the difference in NDS between the mean and minimum of the full sample, or \((0.397 - 0.244) = 0.153\), which yields an average treatment effect of 250.0 thousand RMB. The average in-sample firm VAT bill increased from 3,590 to 5,416 thousand RMB from 1998 to 2007, an increase of 1,826 thousand RMB. Our treatment effect represents 13.7% of the average growth in VAT over our sample.

Another way to benchmark the treatment effect is to compute the share of China’s 2000 total VAT revenue it represents. To do so, we multiply the average full sample treatment effect by the average number of firms in the full sample per year, 711,643 firms / 10 years = 71,164. We obtain 250.0 \times 71,164 = 17.79 billion RMB. We then divide this value by China’s total VAT revenue in 2000, 455.3 billion RMB, yielding 3.95%. However, this value should be scaled by the share of China’s manufacturing sector included in our sample. To obtain this value, we divide the total manufacturing value-added in our full-ASIP cross-section, which equals 1,488 billion RMB, by total value-added GDP from manufacturing in 2001, 4,385 billion RMB. We find that our sample covers 33.9% of manufacturing in China, as 4,385,430 / 1,487,844 = 0.339. We find that our VAT treatment effect represents 11.65% of all VAT revenues in China in 2000.

We also compute these figures for the balanced panel results. We find that the average treatment effect is 1,839 \times (0.398 - 0.244) = 284.3 thousand RMB. In the balanced sample, the average VAT bill increased from 1,492 to 2,342 thousand RMB from 1998 to 2007, an increase of 850 thousand RMB. Therefore, the balanced panel treatment effect represents 33.4% of the increase in VAT over our sample. In the balanced panel, there are 18,010 firms per year. The total treatment effect is 284.3 \times 18,010 = 5,120,497.5 thousand RMB, and divided by China’s 2000 VAT revenues of 455.31 billion RMB, is 1.12%.

To obtain the elasticity of firm sales with respect to VAT/Sales (the effective VAT rate), we first compute the treatment effect of computerization on sales and on VAT rate. For sales, we take the average balanced panel coefficient on sales, multiply it by the difference between the average and minimum NDS in the sample, and divide by average pre-period sales. We find that computerization decreased firm sales by \(-29,501 \times (0.398 - 0.244) / 44,301 = -0.077\), or \(-7.7\%\) percent. The same computation for effective VAT rate yields \(0.0228 \times (0.398 - 0.244) / 0.042 = 0.083\), or 8.3%. Dividing the treatment effects, we find that firm sales declined by \(-8.3 / 7.7 = -0.92\) percent for every percent increase in the effective VAT rate.
F 2SLS Estimates

To calculate the Mexican NDS, we use data from the 2000 World Input-Output Table reports (Di- etzenbacher et al., 2013; Timmer et al., 2016). We use the year 2000 as it is the earliest available year with a richer disaggregation of 56 sectors. For years prior to 2000, the World Input-Output Tables are reported at an aggregation of 35 sectors. We create a correspondence between the 56 input sectors and whether each sector would provide a non-deductible input type under Chinese tax law in the year 2000.

In practice, we consider inputs from agricultural, mining, and manufacturing industries to be materials, and thus deductible under Chinese VAT rules. We treat inputs from service industries, overhead, labor inputs, and value-added to be non-deductible. To obtain the final measure, we sum the input shares from deductible industries to obtain a single fraction for each industry that represents the share of inputs deductible under Chinese VAT rules. This object can be characterized by the following equation, where \( D \) represents the set of deductible industries. For each Mexican output sector \( s \), we compute:

\[
\tilde{NDS}_s^{\text{MEX}} = 1 - \sum_{d \in D} \text{Input fraction}_{sd},
\]

where \( \text{Input fraction}_{sd} \) is the share of inputs required for one unit of production in sector \( s \) from all other sectors \( d \), and \( D \) is the set of the industries providing VAT-deductible inputs.

For the U.S. NDS, we use a very similar procedure. We use data from the 2007 Detailed Input U.S. Tables (U.S. Bureau of Economic Analysis, 2007). To construct our measure of U.S. NDS, we again map each sector in the input-output tables into two groups, deductible or non-deductible, according to the same rules as above. We then construct the following object, where \( D \) represents the set of deductible industries:

\[
\tilde{NDS}_s^{\text{US}} = 1 - \sum_{d \in D} \text{Input fraction}_{sd}.
\]

We also note that the Mexican and U.S. NDS may measure Chinese NDS with error, which if classical, will attenuate the results.

F.1 Random Permutations of NDS

To test whether our baseline estimates are spuriously generated by the distribution of sectoral NDS, we perform a random permutation test. We generate counterfactual sector-level NDS shares using the distribution of values in the true data and re-estimate Baseline Equation (3). We perform 500 iterations for the outcome variables: VAT, sales, employees, intermediate inputs and TFPR.
(DLW) and plot histograms of the coefficients of \( \widetilde{NDS_s \times Post_t} \) in Appendix Figure A.2. We then perform an analogous permutation for the timing of treatment years. For each of the five outcomes above, we randomly reassign the introduction of computerization across the years in our sample and estimate the baseline regression for 500 iterations. The resulting distributions of counterfactual coefficients are presented in Appendix Figure A.3. We find that our baseline estimates are highly unlikely to be generated by random chance.

G Simple Model

G.1 Benchmark

We present a simple model that generates all of the main dynamic effects. In the simple benchmark case, we begin by considering one sector, populated by identical, perfectly competitive firms. We assume that all firms in the given sector have the Cobb-Douglas technology, \( k^\alpha l^{1-\alpha} \), and factor prices of \( k \) and \( l \) are given by \( r \) and \( w \). The pre-tax price of output (paid by the buyer) is \( q \), and the tax-exclusive price of the output (received by the producer) is \( p \), with \( q = (1 + \tau) p \). Demand for the output of the sector is given by \( y = q^{-\sigma} \) where \( \sigma > 0 \) is the elasticity of demand.

We assume that there are three periods. In period 0, there is no tax on the sector, \( \tau_0 = 0 \). The tax is introduced in period 1, and \( \tau_2 = \tau_1 \). Period 1 represents "short run", when only one factor, \( l \), can be adjusted freely. Period 2 represents "long run", when both factors can be adjusted. We assume that neither \( k \) nor \( l \) can be deducted from VAT, so that VAT is a pure sales tax. In addition, we assume that the sector is "small", so that \( r \) and \( w \) are not affected by the introduction of taxes on the given sector. Sector prices \( q \) and \( p \) will naturally be affected by taxation.

There are a few important points regarding these assumptions. (i) It is straightforward to write a full GE model with multiple sectors, so that taxes on sector \( i \) are economy-wide and affect \( r, w \). It requires much more algebra, but the results are the same as in this model, just less transparent. (ii) It is similarly straightforward to add intermediate inputs that can be deducted from the VAT, so that technology is \( k^\alpha l^{1-\alpha-\beta} x^{\beta} \), where \( x \) is the deductible input. All the results from the simpler model below will hold, but again there will be more algebra, and, moreover, one must take a stand on whether \( x \) is adjusted in the long or short run. After we present the baseline model, we will show that all of the main insights follow through with extensions, and demonstrate that the results follow through under monopolistic competition.

Also note that while we refer to \( k \) as capital in the model, it does not correspond to the "assets" in the data (which do not change much), but rather to inputs that firms can change over time (e.g., intermediate inputs). Later, we extend this model to three factors, one of which can be adjusted in period 1 and 2, another in period 2 only, and a third that can never be changed. All the key results
G.1.1 Period 0

Consider the firm’s cost minimization problem in period 0:

\[
C_0(y) = \min_{k,l} rk + wl, \\
\text{s.t. } y = k^{\alpha}l^{1-\alpha}.
\]

The first order conditions are [k]: \( r = \eta k^{\alpha-1}l^{1-\alpha} \) and [l]: \( w = \eta (1-\alpha) k^{\alpha}l^{-\alpha} \). These conditions yield the optimal capital-labor ratio \( k_0 = \frac{\alpha w}{r} \). Marginal costs are therefore \( C'_0(y) = \frac{\eta}{\alpha (1-\alpha)} \), and in equilibrium, we have \( C'_0(y_0) = \frac{\eta}{\omega} \), where \( \omega \) does not depend on anything under the firm’s control.

When firms are perfectly competitive, their tax-exclusive price is equal to their marginal cost, so that \( p_0 = C'_0(y_0) \). Consumer demand gives \( y_0 = q^{-\sigma} = p_0^{-\sigma} \). We substitute this object into the expression above to obtain \( y_0^{-1/\sigma} = C'_0(y_0) \).

The solution to this equation characterizes the output in period 0. In particular, we have \( y_0 = \omega^{-\sigma} \). Since \( y_0 = k_0^{\alpha}l_0^{1-\alpha} = \left( \frac{k_0}{l_0} \right)^\alpha l_0 = \left( \frac{\alpha w}{r} \right)^\alpha l_0 \), we also obtain an expression for labor, \( l_0 = \omega^{-\sigma} \left( \frac{\alpha w}{r} \right)^\alpha \).

We derive \( k_0 \) and \( p_0 \) using the above equations.

G.1.2 Short-run Equilibrium

Suppose a VAT is introduced. Under our assumptions, firms cannot deduct anything, so the VAT is equivalent to a sales tax. Suppose that in the short run, the firm cannot adjust \( k \), so that \( k_1 = k_0 \).

Then we have

\[
C_1(y) = \min_{l} rl_0 + wl, \\
\text{s.t. } y = k_0^{\alpha}l_1^{1-\alpha},
\]

which gives [l]: \( w = \eta (1-\alpha) k_0^{\alpha}l_1^{-\alpha} \). Therefore, marginal costs are \( C'_1(y) = \eta = \frac{w}{(1-\alpha)k_0^{\alpha}} \).

Perfect competition gives \( p_1 = C'_1(y) \), and demand is determined by the pre-tax price \( q_1 = (1+\tau) p_1 \), so the equilibrium condition is \( y_1^{-1/\sigma} = q_1 = (1+\tau) C'_1(y_1) \).

We wish to derive the effect of taxation on inputs, prices, sales, tax revenues, and TFPR. The sales that we observe in the data are \( qy \); tax revenues are \( \tau py \); and TFPR is \( \frac{qy}{k_0^{\alpha}l_1^{1-\alpha}} = q \).

**Lemma 1.** In the short run, \( y_1 < y_0, p_1 < p_0, l_1 < l_0, q_1 > q_0, \text{TFPR}_1 > \text{TFPR}_0, \text{and taxes}_1 > \text{taxes}_0 = 0. \) If \( \sigma > 1, \) then sales$_1 < \text{sales}_0$. 

7
Since TFPR is just other factors, their marginal costs rise. This implies that \( p \) rises, and therefore, \( q \) rises even further.

\[
y_1 < y_0 \text{ implies } l_1 < l_0, \quad C'_1(y_1) < C'_0(y_0), \quad p_1 < p_0. \quad \text{From } y_1 = q_1^{-\sigma}, \text{ we obtain } q_1 > q_0.
\]

Tax revenues are \( \tau p_1 y_1 = \tau (1 + \tau)^{-\sigma} p_1^{1-\sigma} > 0 \), so tax revenues increase. Sales are \( q_1 y_1 = q_1^{1-\sigma} \), they decline if \( \sigma > 1 \). Labor goes down \( l_1 < l_0 \). Capital does not change \( k_1 = k_0 \). TFPR is equal to \( q \) in this model, so TFPR goes up. \( \blacksquare \)

For the next section, we need to derive \( l_1 \). From the previous equation, we obtain \( \left[ k_0 (l_1^{1-\alpha}) \right]^{-1/\alpha} = \left( 1 + \tau \right) \frac{w}{(1-\alpha)k_0^{1-\alpha}}. \)

### G.1.3 Long-run Equilibrium

Now consider the long-run equilibrium, when capital can also be adjusted. Therefore \( C_2(y) = C_0(y) \) (the cost function is the same) and in the long run we have \( \frac{k_2}{l_2} = \frac{\alpha}{1-\alpha} \frac{w}{\tau} = \frac{k_0}{l_0} \). It follows that \( C'_2(y_2) = C'_0(y_0) > C'_1(y_1) \), so \( p_2 = p_0 > p_1 \).

Since \( q_2 = (1 + \tau) p_2 \), \( q_1 = (1 + \tau) p_1 > p_0 \), and \( q_0 = p_0 \), we also find that \( q_2 > q_1 > q_0 \) and \( TFPR_2 > TFPR_1 > TFPR_0 \).

**Remark 2.** Intuitively, since not all factors can be adjusted immediately, the marginal costs fall: there is too much capital relative to labor in the short run, so the marginal cost of labor (the only factor that can be adjusted in period 1) is low. Therefore, the tax-exclusive price falls, although less than one for one with the tax rate, so that pre-tax price \( q \) increases. Over time, as firms adjust other factors, their marginal costs rise. This implies that \( p \) rises, and therefore, \( q \) rises even further. Since TFPR is just \( q \), the same is true about TFPR.

Demand is \( y_2 = [(1 + \tau) p_2]^{-\sigma} < [(1 + \tau) p_1]^{-\sigma} < y_1 \), so \( y_2 < y_1 < y_0 \).

Sales are \( qy = q_1^{1-\sigma} \). Therefore, if \( \sigma > 1 \), \( q_2^{1-\sigma} < q_1^{1-\sigma} < q_0^{1-\sigma} \), and \( sales_2 < sales_1 < sales_0 \).

Tax revenues are \( \tau py = \tau_0 qy = \frac{\tau}{1+\tau} \times sales_0 \). Since \( \tau_0 = 0 \), \( \tau_1 = \tau_2 > 0 \), this gives us, if \( \sigma > 1 \), that \( 0 = taxes_0 < taxes_2 < taxes_1 \).

**Remark 3.** The intuition behind these results comes from the previous remark and the assumption that \( \sigma > 1 \). As \( q \) increases in each period, \( y \) must fall in each period. If demand is elastic, \( y \) falls faster than \( q \) rises, which implies that sales, \( qy \), fall. Since tax revenues are \( \frac{\tau}{1+\tau} \times sales_0 \), it first increases between periods 0 and 1 (since taxes are increased from 0 to \( \tau \)) and then falls between periods 1 and 2 (since sales fall between periods 1 and 2).

Finally, we examine how labor responds. We already know \( l_0 > l_1 \) and \( l_0 > l_2 \), so the remaining comparison of interest is between \( l_1 \) and \( l_2 \). In both cases, we have \( y^{-1/\sigma} = (1 + \tau) C'(y) \). Thus,
we have
\[ l_1^{(\alpha-1)/\sigma-\alpha} = (1 + \tau) \frac{w}{(1 - \alpha)} k_0^{\alpha/\sigma-\alpha}, \]
\[ l_2^{(\alpha-1)/\sigma-\alpha} = (1 + \tau) \frac{w}{(1 - \alpha)} k_2^{\alpha/\sigma-\alpha}. \]

It holds that \( k_2 < k_0 \) (since \( k_2/l_2 = k_0/l_0 \) and \( k_2 (k_2/l_2)^{\alpha-1} = y_2 < y_0 = k_0 (k_0/l_0)^{\alpha-1} \)). Therefore, if \( \sigma > 1 \), we have \( k_2^{\alpha/\sigma-\alpha} > k_0^{\alpha/\sigma-\alpha} \) and therefore \( l_2^{(\alpha-1)/\sigma-\alpha} > l_1^{(\alpha-1)/\sigma-\alpha} \). Since \( \alpha < 1 \), this implies that \( l_2 < l_1 \). Therefore, \( l_0 > l_1 > l_2 \).

**Remark 4.** The intuition for this result comes from the following observation. We know from the Le Chatelier Principle (Samuelson, 1949) that the short-run elasticity of labor should be smaller than the long-run elasticity of labor (because capital can also be adjusted in the long run) holding pre-tax prices fixed. This effect implies that labor should react even more in the long run to the tax change than in the short run. However, there is an offsetting effect in this setting, as the pre-tax price increase drives higher input purchases. Final demand determines which of these forces dominates: if demand is highly elastic, the first force is stronger, and labor will fall monotonically.

### G.1.4 Empirical Implications

This model has several empirically testable implications. First, tax revenues will increase from period 0 to period 1, and then decline in period 2 to a level between the levels of period 0 and one: \( 0 = \text{taxes}_0 < \text{taxes}_2 < \text{taxes}_1 \). Second, the pre-tax price, or \( \text{TFPR} \), increases every period, \( q_2 > q_1 > q_0 \). Third, sales decline each period, \( q_2 y_2 < q_1 y_1 < q_0 y_0 \). Fourth, labor inputs decline each period, \( l_0 > l_1 > l_2 \) and \( k_0 \geq k_1 > k_2 \). The empirical analysis will examine whether these implications are borne out in the data.

In the following sections, we show that these results hold when we introduce a third deductible good, allow for monopolistic competition, and endogenize input prices.

### G.2 Intermediate Goods

Suppose we have technology \( k^\alpha l^{1-\alpha-\beta} x^\beta \) where \( x \) can be deducted from the VAT. Let the price of \( x \) be \( z \). Firm profits without VAT are \( qy - rk - wl - zx \) and profits with VAT tax \( \tau \) are \( (1 - \tau) [qy - zx] - rk - wl = (1 - \tau) qy - rk - wl - (1 - \tau) z x \).

Note that we have changed the pricing convention. Before, we used \( (1 + \tau) p = q \), where \( p \) is the tax-exclusive price. Now we use \( p = (1 - \tau) q \), where \( q \) is the pre-tax price. The connection to the data is clearer with this notation, since we directly observe \( q \).
G.2.1 Period 0

Consider the cost function in period 0:

\[ C_0(y) = \min_{k,l,x} rk + wl + zx, \]
\[ \text{s.t. } y = k^{\alpha} l^{1-\alpha-\beta} x^\beta. \]

The first order conditions are:
\[ [k] : r = \omega \alpha k^{\alpha-1} l^{1-\alpha-\beta} x^\beta, \]
\[ [l] : w = \omega (1-\alpha-\beta) k^\alpha l^{-\alpha-\beta} x^\beta, \]
\[ [x] : z = \omega \beta k^{\alpha} l^{1-\alpha-\beta} x^{\beta-1}. \]

The optimal input ratios are therefore:
\[ \frac{k_0}{l_0} = \frac{\alpha w}{1-\alpha-\beta r} \]
\[ \frac{x_0}{l_0} = \frac{\beta w}{1-\alpha-\beta z}. \]

Marginal costs are:
\[ C'_0(y_0) = \omega = \frac{w}{(1-\alpha-\beta) k^{\alpha} l^{1-\alpha-\beta} x^\beta}. \]

Competitive firms set the tax-exclusive price to equal its marginal cost. Since there are no taxes in period 0, we have \( q_0 = \omega \). Then, the first order conditions immediately imply:
\[ \frac{k_0}{l_0} = \frac{\alpha w}{1-\alpha-\beta r} \]
\[ \frac{x_0}{l_0} = \frac{\beta w}{1-\alpha-\beta z}. \]

Finally, quantities are determined by the downward sloping demand curve, \( y_0 = q_0^{\sigma} \).

This equation gives:
\[ \frac{k_0}{l_0} = \left( \frac{\alpha w}{1-\alpha-\beta} \right)^{(\alpha-1)} \]
\[ \frac{x_0}{l_0} = \left( \frac{\beta w}{1-\alpha-\beta} \right)^{(\beta-1)}. \]

Substitution yields:
\[ l_0 = \left( \frac{w}{1-\alpha-\beta} \right)^{\sigma} \left( \frac{\alpha w}{1-\alpha-\beta} \right)^{(\alpha-1)} \left( \frac{\beta w}{1-\alpha-\beta} \right)^{(\beta-1)}. \]

It then follows that:
\[ k_0 = \frac{\alpha w}{1-\alpha-\beta r} l_0, \]
\[ x_0 = \frac{\beta w}{1-\alpha-\beta z} l_0. \]

G.2.2 Period 2

We analyze period 2 before period 1, since period 2 is almost identical to period 0. With VAT, the firm’s profits are:
\[ (1-\tau) [gy - zx] - rk - wl = (1-\tau) qy - rk - wl - (1-\tau) z x, \]
so the cost min-
The minimization problem is:

\[
C_2(y) = \min_{k,l,x} rk + wl + (1 - \tau) zx, \\
\text{s.t. } y = k^\alpha l^{1 - \alpha - \beta} x^\beta.
\]

The tax-exclusive price is equal to the marginal cost, so that \((1 - \tau) q_2 = C'_2(y_2) = \omega_2 \implies q_2 = \frac{C'_2(y_2)}{1 - \tau} = \frac{\omega_2}{1 - \tau} \omega_2 = \frac{\omega_2}{1 - \alpha - \beta \{1 - \tau\} z} \omega_2 = \frac{\omega_2}{1 - \alpha - \beta \{1 - \tau\} z}\]. The optimal input ratios are therefore \(k_2 = \frac{\alpha}{1 - \alpha - \beta} \frac{w}{r} \) and \(l_2 = \frac{\beta}{1 - \alpha - \beta \{1 - \tau\} z} \frac{w}{z} \). Solving for wage, we obtain \(\omega_2 = (1 - \tau)^{\beta} \omega_0\).

Finally, \(y_2 = q_2^{-\sigma} = \left(\frac{\omega_2}{1 - \tau}\right)^{-\sigma} \left(\frac{k_2}{l_2}\right)^{\alpha} \left(\frac{x_2}{l_2}\right)^{\beta} l_2 = (1 - \tau)^{\sigma} \left[\frac{w}{(1 - \alpha - \beta) \left(\frac{k_2}{l_2}\right)^{\alpha} \left(\frac{x_2}{l_2}\right)^{\beta}}\right]^{-\sigma} \implies l_2 = (1 - \tau)^{\sigma} \left(\frac{w}{1 - \alpha - \beta}\right)^{-\sigma} \left(\frac{k_2}{l_2}\right)^{\alpha(\sigma - 1)} \left(\frac{x_2}{l_2}\right)^{\beta(\sigma - 1)} \right]^{-\sigma}

\[l_2 = (1 - \tau)^{\sigma(1 - \beta) + \beta} \left(\frac{w}{1 - \alpha - \beta}\right)^{-\sigma} \left(\frac{\alpha}{1 - \alpha - \beta \frac{w}{r}}\right)^{\alpha(\sigma - 1)} \left(\frac{\beta}{1 - \alpha - \beta \frac{w}{z}}\right)^{\beta(\sigma - 1)}
\]

Similarly, we have

\[k_2 = \frac{\alpha}{1 - \alpha - \beta \frac{w}{r}} l_2 = (1 - \tau)^{(1 - \beta) + \beta} k_0, \]
\[x_2 = \frac{\beta}{1 - \alpha - \beta \{1 - \tau\} z} l_2 = (1 - \tau)^{(\sigma - 1)(1 - \beta) + \beta} x_0. \]

This result generates clear predictions about the long run.

**Lemma 5.** Suppose \(\sigma > 1\). Then,

1. \(TFPR_2 > TFPR_0\),
2. \(sales_2 < sales_0\),
3. \(k_2 < k_0, x_2 < x_0, l_2 < l_0, \omega_2 < \omega_0\),
4. \(0 = taxes_0 < taxes_2\).
Proof. 1. In our model, \( TFPR \equiv \frac{q_y}{k^{\alpha l^{1-\alpha}-\beta x^\beta}} = q \). We have

\[
q_2 = \frac{\omega_2}{1 - \tau} = \frac{(1 - \tau)^\beta \omega_0}{1 - \tau} = (1 - \tau)^{(\beta - 1)} q_0 > q_0.
\]

2. In our model, \( sales = qy = q^{1-\sigma} \). We have, when \( \sigma > 1 \),

\[
q_2^{1-\sigma} = \left[ (1 - \tau)^{(\beta - 1)} q_0 \right]^{1-\sigma} = (1 - \tau)^{(1-\beta)(\sigma-1)} q_0^{1-\sigma} < q_0^{1-\sigma}.
\]

3. We have

\[
\frac{k_2}{k_0} = \frac{l_2}{l_0} = (1 - \tau)^{\sigma(1-\beta)+\beta} < 1
\]

and

\[
\frac{x_2}{x_0} = (1 - \tau)^{(\sigma-1)(1-\beta)+\beta} < 1.
\]

Note that the latter follows from \( \sigma > 1 \) and we showed the result about \( \omega \) earlier.

4. Note that in our model, collected taxes are \( taxes = \tau [qy - zx] \). So

\[
taxes_2 = \tau [q_2 y_2 - zx_2] = \tau [q_2 y_2 - \beta q_2 y_2] = \tau (1 - \beta) q_2 y_2 > 0 = taxes_0.
\]

\[\blacksquare\]

G.2.3 Period 1

Now consider the period 1 problem. We assume that intermediate goods can be adjusted in period 1, which simplifies the analysis. \(^{53}\)

We have

\[
C_1(y) = \min_{l,x} rk_0 + w l + (1 - \tau) z x,
\]

s.t. \( y = k_0^{\alpha l^{1-\alpha}-\beta x^\beta} \).

Which gives

\[
[l] : w = \omega (1 - \alpha - \beta) k_0^{\alpha l^{1-\alpha}-\beta x^\beta},
\]

\[
[x] : (1 - \tau) z = \omega \beta k_0^{\alpha l^{1-\alpha}-\beta x^\beta - 1}.
\]

We have

\[
\frac{x_1}{l_1} = \frac{\beta}{1 - \alpha - \beta} \frac{w}{(1 - \tau) z}.
\]

\(^{53}\)If they cannot, there is a lot more algebra involved although the result about taxes will hold under additional assumptions about the parameters.
As before, we have

\[ q_1 = \frac{C'_1(y_1)}{1 - \tau} = \frac{\omega_1}{1 - \tau}. \]

Hence, we have

\[ w l_1 = (1 - \alpha - \beta) (1 - \tau) q_1 y_1, \]
\[ (1 - \tau) z x_1 = \beta (1 - \tau) q_1 y_1. \]

The marginal costs are

\[ \omega_1 = C'_1(y_1) = \frac{1}{1 - \alpha - \beta} \frac{w}{k_0 l_1^{\alpha - \beta} x_1^\beta}. \]

We find \( l_1 \) as before, using the demand curve:

\[ y_1 = \left[ \frac{\omega_1}{1 - \tau} \right]^{-\sigma}, \]
\[ k_0 l_1^{\alpha - \beta} \left( \frac{x_1}{l_1} \right)^\beta = (1 - \tau)^\sigma \left[ \frac{1}{1 - \alpha - \beta} \frac{w}{k_0 l_1^{\alpha - \beta} \left( \frac{x_1}{l_1} \right)^\beta} \right]^{-\sigma}. \]

Therefore,

\[ l_1^{1 - \alpha + \sigma \alpha} = (1 - \tau)^\sigma \left( \frac{w}{1 - \alpha - \beta} \right)^{-\sigma} k_0^{\alpha (\sigma - 1)} \left( \frac{x_1}{l_1} \right)^{\beta (\sigma - 1)}, \]
\[ = (1 - \tau)^\sigma \left( \frac{w}{1 - \alpha - \beta} \right)^{-\sigma} k_0^{\alpha (\sigma - 1)} \left( \frac{\beta}{1 - \alpha - \beta} \frac{w}{z (1 - \tau)} \right)^{\beta (\sigma - 1)}, \]
\[ = (1 - \tau)^{\sigma + \beta (1 - \sigma)} \left( \frac{w}{1 - \alpha - \beta} \right)^{-\sigma} k_0^{\alpha (\sigma - 1)} \left( \frac{\beta}{1 - \alpha - \beta} \frac{w}{z} \right)^{\beta (\sigma - 1)}. \]

This equation gives the following useful intermediate result.

**Lemma 6.** Suppose \( \sigma > 1 \). Then

1. \( l_0 > l_1 > l_2 \),
2. \( y_0 > y_1 > y_2 \),
3. \( \omega_1 < \omega_2 < \omega_0 \) and \( \omega_0 < \frac{\omega_1}{1 - \tau} < \frac{\omega_2}{1 - \tau} \).
Proof. 1. The previous equation should also hold in period 2 when capital stock is set at its optimal value \(k_2\), i.e.

\[
l_2^{1-\alpha+\sigma\alpha} = (1 - \tau)^{\sigma+\beta(1-\sigma)} \left( \frac{w}{1 - \alpha - \beta} \right)^{-\sigma} k_2^{\alpha(\sigma-1)} \left( \frac{\beta}{1 - \alpha - \beta} \right)^{\beta(\sigma-1)}
\]

which implies

\[
\left( \frac{l_2}{l_1} \right)^{1+(\sigma-1)\alpha} = \left( \frac{k_2}{k_0} \right)^{\alpha(\sigma-1)}
\]

\[
l_2 = \left( \frac{k_2}{k_0} \right)^{\frac{\alpha(\sigma-1)}{1+\alpha(\sigma-1)}}.
\]

Since \(k_2 < k_0\), this implies \(l_2 < l_1\).

Similarly, the analogous equation should hold in period 0 (when \(\tau = 0\)) so that

\[
\left( \frac{l_1}{l_0} \right)^{1+(\sigma-1)\alpha} = (1 - \tau)^{\sigma+\beta(1-\sigma)} = (1 - \tau)^{(1-\beta)+\beta}
\]

\[
l_1 = (1 - \tau)^{\frac{\alpha(1-\beta)+\beta}{1+\alpha(\sigma-1)}} < 1.
\]

Therefore \(l_1 < l_0\).

2. For output, we have

\[
y_1 = \left( \frac{l_1}{l_0} \right)^{1-\alpha} \left( \frac{x_1}{x_0} \right)^{\beta}
\]

\[
= (1 - \tau)^{\frac{\alpha(1-\beta)+\beta}{1+\alpha(\sigma-1)}} (1-\alpha)^{-\beta}
\]

\[
= (1 - \tau)^{\frac{\alpha-1-\beta}{\alpha(\sigma-1)}} < 1.
\]

Therefore, \(y_1 < y_0\).

Using the fact that \(\frac{y_1}{l_1} = \frac{y_2}{l_2}\), we have

\[
\frac{y_2}{y_1} = \frac{k_2^\alpha l_2^{1-\alpha}}{k_0^\alpha l_1^{1-\alpha}}.
\]

Since we showed already that \(\frac{k_2}{k_0} < 1\) and \(\frac{l_2}{l_1} < 1\), this implies that \(y_2 < y_1\).
3. For marginal costs, we have

\[
\frac{\omega_1}{\omega_2} = \frac{\frac{1}{1-\alpha-\beta} \frac{w}{k_0 l_1 - \alpha \frac{y_1}{l_1}}}{1-\alpha-\beta} \frac{\frac{k_2}{l_2}}{\frac{k_0}{l_1}} = \frac{\left(\frac{k_2}{l_2}\right)^{\alpha}}{\frac{k_0}{l_1}} = \left(\frac{k_2}{k_0}\right)^{\frac{\alpha}{1+\alpha(\sigma-1)}}
\]

Thus, \( \omega_1 < \omega_2 \). We showed already that \( \omega_2 < \omega_0 \), which implies \( \omega_1 < \omega_0 \).

Moreover,

\[
\frac{\omega_1}{\omega_0} = \frac{\frac{1}{1-\alpha-\beta} \frac{w}{k_0 l_1 - \alpha \frac{y_1}{l_1}}}{1-\alpha-\beta} \frac{\frac{l_0}{l_2}}{\frac{l_0}{l_1}} = \frac{\left(\frac{l_1}{l_0}\right)^{\alpha}}{1-\alpha-\beta} \frac{\alpha}{\sigma-1}
\]

or

\[
\frac{\omega_1}{\omega_0} = \frac{1}{1-\tau}(1-\tau)^{\frac{\sigma}{1+\alpha(\sigma-1)}} \frac{\alpha}{1-\beta} - \frac{\frac{l_1}{l_0}}{1-\alpha-\beta} \frac{\alpha}{\sigma-1}
\]

which implies that \( \frac{\omega_1}{\omega_0} > \omega_0 \).

With this lemma, we can extend all the results of the simple model.\( \blacksquare \)

**Lemma 7.** Suppose \( \sigma > 1 \). Then

1. \( T F P R_2 > T F P R_1 > T F P R_0 \),

2. \( sales_0 > sales_1 > sales_2 \),

3. \( 0 = taxes_0 < taxes_2 < taxes_1 \).

**Proof.** 1. Since \( T F P R = q = \frac{\omega}{1-\tau} \), from the previous lemma we have

\[
q_0 < q_1 < q_2.
\]

2. Sales are \( qy = q^{1-\sigma} \), so with \( \sigma > 1 \) we have, from the previous equation

\[
sales_0 > sales_1 > sales_2.
\]

3. Taxes revenues are \( \tau (qy - z) \). Since

\[
\frac{z x_1}{q_1 y_1} = \frac{z x_2}{q_2 y_2} = \beta,
\]

it becomes

\[
taxes = (1 - \beta) \tau \times sales.
\]
Since $\tau_0 = 0$, and $sales_1 > sales_2$, we get

$$0 = taxes_0 < taxes_2 < taxes_1.$$ 

G.3 Monopolistic competition

Here, we extend the analysis to allow firms to have market power and set prices. We will focus on the benchmark economy without intermediate goods for simplicity.

Firms will be monopolistically-competitive, as in the Dixit-Stiglitz model. There is a continuum of firms and each firm produces a differentiated good. Consumers buy all these goods, so their budget constraint is

$$\int_0^1 q(i) c(i) di = wl + m,$$

where $m$ is non-labor income.

Consumer preferences in each period are given by

$$Y^{1-1/\sigma} \frac{1}{1-1/\sigma} - l,$$

where

$$Y = \left( \int_0^1 y(i)^{1-1/\varepsilon} di \right)^{\frac{\varepsilon}{1-\varepsilon}}.$$ 

Here, $\varepsilon > 1$ is the elasticity of substitution between goods.

Standard results imply that demand for good $i$ is determined by equation

$$y(i) = \left( \frac{q(i)}{Q} \right)^{-\varepsilon} Y,$$

where the aggregate price satisfies

$$Q = \left( \int_0^1 q(i)^{1-\varepsilon} di \right)^{\frac{1}{1-\varepsilon}}.$$ 

The aggregate demand can be found from

$$\max_{Y,l} \frac{Y^{1-1/\sigma}}{1-\sigma} - l,$$

We assume that the variety set is $[0,1]$ because we assume that $y = Y$ and $q = Q$. 

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which gives

\[ Y^{-1/\sigma} = Q/w. \]

Wage \( w \) can be taken to be a numeraire, and it is without loss of generality to set \( w = 1 \).

### G.3.1 Firm's Problem

We work in "partial" equilibrium so that the interest rate \( r \) is fixed (equivalent to a GE model in which there are international capital markets with a rental rate of capital given by \( r \)). We relax this assumption in another extension. In equilibrium, firm \( i \) takes \( Q, Y, \) and \( r \) as given (\( w = 1 \) always) and chooses \( q(i) \) to maximize its profits, taking into account consumer demand. So the firm in period 0 solves

\[
\max_{q,y,l,k} qy - wl - rk,
\]

s.t.

\[
y = \left( \frac{q}{Q} \right)^{-\epsilon} Y,
\]

\[
y = k^\alpha l^{1-\alpha}.
\]

We have

\[
[l]: w = \omega (1 - \alpha) k^\alpha l^{-\alpha},
\]

\[
[k]: r = \omega \alpha k^{\alpha-1} l^{1-\alpha},
\]

\[
[y]: q = \lambda + \omega,
\]

\[
[q]: qy = \lambda \epsilon \left( \frac{q}{Q} \right)^{-\epsilon} Y.
\]

The first two equations give us the usual conditions

\[
k_0^l = \frac{\alpha w}{1-\alpha r},
\]

\[
\omega_0 = \frac{w}{(1-\alpha) k_0^l l^{-\alpha}} = \frac{w}{(1-\alpha) \left( \frac{\alpha w}{1-\alpha r} \right)^{\alpha}}.
\]

Note that \( \omega_0 \) has the same meaning as before: the marginal cost of producing an extra unit of a good.

In equilibrium, since all firms are identical, we have

\[
q = Q, y = Y.
\]

Therefore, the last two optimality conditions become

\[
q_0 = \lambda_0 + \omega_0,
\]

\[
q_0 = \lambda_0 \epsilon.
\]
This gives us

\[ q_0 = q_0 \varepsilon - \omega_0 \varepsilon = \frac{\varepsilon}{\varepsilon - 1} \omega_0. \]

This equation is the standard condition that the optimal price is equal to a markup \( \frac{\varepsilon}{\varepsilon - 1} > 1 \) times the marginal cost, \( \omega_0 \). As \( \varepsilon \to \infty \), goods become more and more substitutable and we converge to the perfect competition case considered in the benchmark model.

The consumer’s optimality condition \( Y^{-1/\sigma} = Q/w \) (together with normalization \( w = 1, y = Y, q = Q \)) gives

\[ y_0 = q_0^{-\sigma} = \left( \frac{\varepsilon}{\varepsilon - 1} \right)^{-\sigma} \omega_0^{-\sigma}. \]

So the analysis goes through the same way as before, except now everything is multiplied by a markup.

Given that, we will verify that markup is the same in periods 1 and 2. In that case, then all the analysis thus far goes through without any changes.

Period 2’s problem is

\[
\max_{q, y,l,k} (1 - \tau) qy - wl - rk,
\]

s.t.

\[
y = \left( \frac{q}{Q} \right)^{-\varepsilon} Y,
\]

\[
y = k^\alpha l^{1-\alpha}.
\]

These give the optimality conditions.

We have

\[
[l] : w = \omega (1 - \alpha) k^{\alpha - l^{1-\alpha}}, [k] : r = \omega \alpha k^{\alpha - l^{1-\alpha}}, [y] : (1 - \tau) q = \lambda + \omega, [q] : (1 - \tau) qy = \lambda \varepsilon \left( \frac{q}{Q} \right)^{-\varepsilon} Y.
\]

So we have, as before, (the case \( \beta = 0 \)) from the first two equations:

\[ \omega_2 = \omega_0. \]

The last two give us

\[
q_2 = \frac{\varepsilon \omega_2}{\varepsilon - 1 - \tau}.
\]
This expression is the same as we had before, modulo a markup.

Finally, the period 1 problem is

$$\max_{q,y,l} (1 - \tau) qy - wl - rk_0.$$ 

with \([l] : w = \omega(1 - \alpha)k_0^{\alpha}l^{-\alpha}\), \([y] : (1 - \tau)q = \lambda + \omega\), and \([q] : (1 - \tau)qy = \lambda \epsilon \left( \frac{q}{\bar{q}} \right)^{-\epsilon} Y.$$ 

Note that again we have, \(q_1 = \frac{\epsilon}{\epsilon - 1} \frac{\omega_1}{1 - \tau}\).

Thus, the marginal costs are the same as in the baseline, and price is just a constant markup over those costs. This implies that all the steps in the proofs of the baseline economy go through with minimal modifications.

### G.4 Multiple Sectors, Fixed Capital

Now, we assume that there are two sectors, and that the capital stock is in fixed net supply. Other than that, we return to our baseline model of perfect competition. So consumers solve

$$\max \mu \frac{1}{\sigma} \frac{Y^{1 - 1/\sigma}}{1 - 1/\sigma} + (1 - \mu) \frac{1}{\sigma} \frac{Y^{1 - 1/\epsilon}}{1 - 1/\epsilon} - l,$$

s.t.

$$qy + QY = wl + r\bar{k} + \Pi,$$

where \(\bar{k}\) is the total capital stock and capital letters denote the “other sector”, not affected by taxes. Here, \(\mu \in (0, 1)\). The case \(\mu = 0\) corresponds to what we have done before: sector 1 is small, so nothing there affects taxes. Here, \(\Pi\) denotes firm profits. For simplicity, we assume that the production function is the same in the two sectors.

The capital stock is in fixed supply and is rented out by consumers to the firms at a rate \(r\). If the sector-level demands for capital are \(k\) and \(K\), then the market clearing condition for the capital stock is \(k + K = \bar{k}\).

Once again, everything is in units of labor, so we normalize \(w = 1\). The two sectors are identical in period 0, but the VAT tax will be applied to the first sector in period 1. Given our normalization, demand is again given by \(y = \mu q^{-\sigma}, Y = (1 - \mu) Q^{-\sigma}\).

#### G.4.1 Period 0
The analysis goes like before except now \( l_0 \) is given by

\[
\left( \frac{k_0}{l_0} \right)^{\alpha} l_0 = \mu \left[ \frac{w}{(1 - \alpha) \left( \frac{k_0}{l_0} \right)^{\alpha}} \right]^{-\sigma},
\]

or

\[
l_0 = \mu \left( \frac{w}{1 - \alpha} \right)^{-\sigma} \left( \frac{\alpha \ w}{1 - \alpha \ r_0} \right)^{\alpha(\sigma-1)},
\]

and

\[
k_0 = \frac{\alpha \ w}{1 - \alpha \ r_0} l_0 = \mu \left( \frac{w}{1 - \alpha} \right)^{-\sigma} \left( \frac{\alpha \ w}{1 - \alpha \ r_0} \right)^{\alpha(\sigma-1)+1}.
\]

Demand in the other sector is

\[
K_0 = (1 - \mu) \left( \frac{w}{1 - \alpha} \right)^{-\sigma} \left( \frac{\alpha \ w}{1 - \alpha \ r_0} \right)^{\alpha(\sigma-1)+1}.
\]

This allows us to find the rental rate \( r_0 \) from

\[
\mu \left( \frac{w}{1 - \alpha} \right)^{-\sigma} \left( \frac{\alpha \ w}{1 - \alpha \ r_0} \right)^{\alpha(\sigma-1)+1} + (1 - \mu) \left( \frac{w}{1 - \alpha} \right)^{-\sigma} \left( \frac{\alpha \ w}{1 - \alpha \ r_0} \right)^{\alpha(\sigma-1)+1} = \bar{k},
\]

\[
\left( \frac{w}{1 - \alpha} \right)^{-\sigma} \left( \frac{\alpha \ w}{1 - \alpha \ r_0} \right)^{\alpha(\sigma-1)+1} = \bar{k}.
\]

**G.4.2 Period 1**

In period 1, taxes are introduced but capital cannot be adjusted, so we simply assume that \( r_1 = r_0 \). Since the capital stock cannot move, the rental rate is strictly-speaking indeterminate, but small refinements of this setup should give \( r_1 = r_0 \).

Since \((r, w)\) are the same in period 1 as in period 0, the problems of the two sectors are unchanged. The whole characterization of the period 1 problem of the sector affected by the VAT goes without any changes. The labor demand in sector 1 is given by

\[
l_1^{1-\alpha+\sigma\alpha} = \mu \ (1 - \tau)^{\sigma} \left( \frac{w}{1 - \alpha} \right)^{-\sigma} k_0^{\alpha(\sigma-1)}.
\]
**G.4.3 Period 2**

We have, following the same steps as before, \[ l_2 = \mu (1 - \tau)^\sigma \left( \frac{w}{1-\alpha} \right) - \sigma \left( \frac{\alpha w}{1-\sigma} \right) \alpha(\sigma-1) \]

\[ = (1 - \tau)^\sigma \left( \frac{r_2}{r_0} \right) \alpha(\sigma-1) \]

and \[ k_2 = \frac{\alpha w}{1-\alpha} l_2 = \mu (1 - \tau)^\sigma \left( \frac{w}{1-\sigma} \right) \alpha(\sigma-1) + 1 \]

\[ = \left[ (1 - \tau)^\sigma \left( \frac{r_0}{r_2} \right) \alpha(\sigma-1) + 1 \right] \mu \left( \frac{w}{1-\alpha} \right) - \sigma \left( \frac{\alpha w}{1-\sigma} \right) \alpha(\sigma-1) + 1 = \left[ (1 - \tau)^\sigma \left( \frac{r_0}{r_2} \right) \alpha(\sigma-1) + 1 \right] k_0. \]

Capital in the other sector is \[ K_2 = (1 - \mu) \left( \frac{w}{1-\sigma} \right) - \sigma \left( \frac{\alpha w}{1-\sigma} r_2 \right) \alpha(\sigma-1) + 1, \]

so the market clearing condition is \[ \left[ (1 - \tau)^\sigma + (1 - \mu) \right] \left( \frac{w}{1-\alpha} \right) - \sigma \left( \frac{\alpha w}{1-\sigma} \right) \alpha(\sigma-1) + 1 - \hat{k}. \]

Equivalently,

\[ \left[ (1 - \tau)^\sigma + (1 - \mu) \right] \left( \frac{r_0}{r_2} \right) \alpha(\sigma-1) + 1 = \hat{k}, \]

or

\[ (1 - \tau)^\sigma \left( \frac{r_0}{r_2} \right) \alpha(\sigma-1) + 1 = \frac{(1 - \tau)^\sigma}{\mu (1 - \tau)^\sigma + (1 - \mu)}. \]

**Lemma 8.** Therefore, \((1 - \tau)^\sigma \left( \frac{r_0}{r_2} \right) \alpha(\sigma-1) + 1 \) is strictly increasing in \( \mu \), and

\[(1 - \tau)^\sigma \leq (1 - \tau)^\sigma \left( \frac{r_0}{r_2} \right) \alpha(\sigma-1) + 1 \leq 1. \]

The left and right relations hold with equality when \( \mu = 0 \) and \( \mu = 1 \), respectively.

**Lemma 9.** Additionally, suppose \( \sigma > 1 \). Then \( k_2 \leq k_0, l_2 \leq l_1, sales_2 \leq sales_1, taxes_2 \leq taxes_1, \) and \( T FPR_2 \geq T FPR_1 \), which hold as equalities only if \( \mu = 1 \). The inequalities reverse for sector 2.

**Proof.** The previous lemma and our equation for capital imply that \( k_2 \leq k_0 \). The labor supply \( l_1 \) and \( l_2 \) can be written (see Lemma 6) as

\[ l_1^{1-\alpha + \sigma \alpha} = \mu (1 - \tau)^\sigma \left( \frac{w}{1-\alpha} \right) k_0^\alpha(\sigma-1), \]

\[ l_2^{1-\alpha + \sigma \alpha} = \mu (1 - \tau)^\sigma \left( \frac{w}{1-\alpha} \right) k_2^\alpha(\sigma-1). \]

Therefore, \( l_2 \leq l_1 \) with strict inequality if \( \mu < 1 \). Since \( y_t = l_t^{\frac{\alpha}{1-\sigma}} l_t^{\frac{\sigma}{1-\sigma}} \), and both \( k \) and \( l \) decrease in period 2, \( y_2 \leq y_1 \). We have \( sales_t = q_t y_t = \mu^\alpha y_t^{\frac{\sigma}{1-\sigma}} \), therefore \( sales_2 \leq sales_1 \). Taxes are given by \( taxes_t = \tau \times sales_t \), so we get the result on sales. Since we can also write \( sales_t = \mu q_t^{1-\sigma} \) and \( T FPR_t = q_t \), we get that \( T FPR_2 \geq T FPR_1 \).
Since total capital is fixed, we must have $K_2 \geq K_0$ and the same steps prove reverse inequalities for sector 2 (which obviously does not have taxes).

This step completes the proof, since we already know what happens in period 1. Note that $\mu = 0$ is the same case as our baseline model (it is easier to see this if we redefine all variables as ratios to $\mu$ and look at the limit as $\mu \to 0$). In this case, sector 1 is small, so that any reallocation of capital from sector 1 to sector 2 has no effect on price $r$. The lemma above shows that all the insights continue to generalize in the 2 sector GE model where interest rate $r$ is endogenously determined and is affected by the reallocation. The mechanism is the same as in the benchmark case: as long as there is some reallocation in period 2 of capital due to re-optimization, capital $k_2$ will decrease in period 2, further depressing labor demand $l_2$ and output $y_2$, leading to lower sales and tax revenues in sector 1. In the limit case, $\mu = 1$, sector 2 is negligibly small and cannot absorb any capital. As a result, with fixed capital stock, rental rates $r_2$ must fall sufficiently to prevent any re-allocation of capital from sector 1, in which case, period 1 and period 2 become identical.
Table A.1: Sectors with Highest and Lowest NDS

<table>
<thead>
<tr>
<th>Sector Name</th>
<th>Lowest Non-Deductible Share</th>
<th>Highest Non-Deductible Share</th>
</tr>
</thead>
<tbody>
<tr>
<td>Beet Sugar</td>
<td>0.24</td>
<td>Tobacco Leaf Re - Baking</td>
</tr>
<tr>
<td>Cane Sugar</td>
<td>0.24</td>
<td>Other Tobacco Processing</td>
</tr>
<tr>
<td>Frozen Aquatic Products Processing</td>
<td>0.25</td>
<td>Cigarette Manufacturing</td>
</tr>
<tr>
<td>Dry Processing Of Aquatic Products</td>
<td>0.25</td>
<td>Caustic Soda Manufacturing</td>
</tr>
<tr>
<td>Electric Light Source Manufacturing</td>
<td>0.27</td>
<td>Inorganic Acid Manufacturing</td>
</tr>
<tr>
<td>Lamp Holder, Lampholders Manufacturing</td>
<td>0.27</td>
<td>Industrial Ceramics</td>
</tr>
<tr>
<td>Wire And Cable Manufacturing</td>
<td>0.30</td>
<td>Other Ceramics</td>
</tr>
<tr>
<td>Postal Machinery And Equipment Manufacturing</td>
<td>0.30</td>
<td>Other Stationery Manufacturing</td>
</tr>
</tbody>
</table>
| Construction Machinery Manufacturing            | 0.30                        |Note: Manufacturing sectors are defined by four-digit Chinese Industrial Codes. Non-deductible share is calculated from 1997 Chinese Input Output Tables. See the text for a detailed description.
Table A.2: The Effects of Computerization on VAT - NDS Quartiles

<table>
<thead>
<tr>
<th>Dep Var Mean</th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>VAT Gross (1,000 RMB)</td>
<td>7,758</td>
<td>6,194</td>
<td>2,043</td>
<td>0.0418</td>
</tr>
<tr>
<td>VAT Deductions (1,000 RMB)</td>
<td>(453.2)</td>
<td>(417.9)</td>
<td>(85.59)</td>
<td>(0.000571)</td>
</tr>
<tr>
<td>VAT (1,000 RMB)</td>
<td>-797.7*</td>
<td>-1,025***</td>
<td>152.3</td>
<td>0.00348***</td>
</tr>
<tr>
<td>(1,000 RMB)</td>
<td>(466.3)</td>
<td>(330.3)</td>
<td>(133.1)</td>
<td>(0.000989)</td>
</tr>
<tr>
<td>VAT/Sales</td>
<td>0.00104*</td>
<td>0.000571</td>
<td>0.00348***</td>
<td>0.00356***</td>
</tr>
<tr>
<td>(1,000 RMB)</td>
<td>(423.3)</td>
<td>(361.7)</td>
<td>(107.1)</td>
<td>(0.000921)</td>
</tr>
<tr>
<td>Observations</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.628</td>
<td>0.503</td>
<td>0.702</td>
<td>0.570</td>
</tr>
</tbody>
</table>

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1 In Panels B (D), the instrument is U.S. (Mexico) non-deductible share x post-2002.
Table A.3: The Average 2SLS Effect of Computerization on VAT – Robustness to Dropping Sectors Highly Exposed to Chinese Trade

<table>
<thead>
<tr>
<th>Dep Var Mean</th>
<th>Drop Sectors with top 25% (Total Mexico to China Trade Flows) / (Total Mexico Production)</th>
<th>2SLS</th>
<th>Reduced Form</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dependent Variable: VAT</td>
<td>(1)</td>
<td>(2)</td>
<td></td>
</tr>
<tr>
<td>Non-deductible share × Post-2002</td>
<td>6,221**&lt;sup&gt;1&lt;/sup&gt;, 1111.7***</td>
<td>1,767</td>
<td>374.0</td>
</tr>
<tr>
<td>Observations</td>
<td>148,242</td>
<td>148,242</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Dep Var Mean</th>
<th>Drop Sectors with top 25% (Total U.S. to ChinaTrade Flows) / (Total U.S. Production)</th>
<th>2SLS</th>
<th>Reduced Form</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dependent Variable: VAT</td>
<td>(1)</td>
<td>(2)</td>
<td></td>
</tr>
<tr>
<td>Non-deductible share × Post-2002</td>
<td>3,607**</td>
<td>(1,767)</td>
<td>(374.0)</td>
</tr>
<tr>
<td>Observations</td>
<td>137,593</td>
<td>137,593</td>
<td></td>
</tr>
</tbody>
</table>

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. The instrument is U.S. non-deductible share × Post-2002. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1.
Table A.4: The Year by Year Effects of Computerization on VAT

<table>
<thead>
<tr>
<th></th>
<th>Independent Variables</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>VAT (1)</td>
</tr>
<tr>
<td>Non-deductible Share × 1998</td>
<td>141.5</td>
</tr>
<tr>
<td></td>
<td>(296.4)</td>
</tr>
<tr>
<td>Non-deductible Share × 1999</td>
<td>157.5</td>
</tr>
<tr>
<td></td>
<td>(258.3)</td>
</tr>
<tr>
<td>Non-deductible Share × 2000</td>
<td>-4.821</td>
</tr>
<tr>
<td></td>
<td>(255.3)</td>
</tr>
<tr>
<td>Non-deductible Share × 2001</td>
<td>-</td>
</tr>
<tr>
<td>Non-deductible Share × 2002</td>
<td>1,031**</td>
</tr>
<tr>
<td></td>
<td>(417.0)</td>
</tr>
<tr>
<td>Non-deductible Share × 2003</td>
<td>1,520***</td>
</tr>
<tr>
<td></td>
<td>(580.6)</td>
</tr>
<tr>
<td>Non-deductible Share × 2004</td>
<td>2,247***</td>
</tr>
<tr>
<td></td>
<td>(592.4)</td>
</tr>
<tr>
<td>Non-deductible Share × 2005</td>
<td>2,431***</td>
</tr>
<tr>
<td></td>
<td>(517.2)</td>
</tr>
<tr>
<td>Non-deductible Share × 2006</td>
<td>2,158***</td>
</tr>
<tr>
<td></td>
<td>(641.8)</td>
</tr>
<tr>
<td></td>
<td>(695.9)</td>
</tr>
</tbody>
</table>

Observations: 180,103
R-squared: 0.702
2002-2007 Joint p-value: <0.001

Notes: This sample comprises of a balanced panel of firms during 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1
Table A.5: The Average Effect of Computerization on Firm Behavior

<table>
<thead>
<tr>
<th>Dependent Variable</th>
<th>Sales (1,000 RMB)</th>
<th>Indelible Sales (1,000 RMB)</th>
<th>Employees</th>
<th>Intermediate Inputs (1,000 RMB)</th>
<th>Intermediate Inputs as a Share of Total Input</th>
<th>TFPR DLW</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
<td>(5)</td>
<td>(6)</td>
</tr>
<tr>
<td>Dep Var Mean</td>
<td>58545</td>
<td>6776</td>
<td>3552</td>
<td>40590</td>
<td>0.835</td>
<td>0.764</td>
</tr>
<tr>
<td>Non-deductible share × Post-200</td>
<td>-29,501**</td>
<td>-8,114**</td>
<td>-35.17</td>
<td>-19,871**</td>
<td>-0.0333</td>
<td>-0.256***</td>
</tr>
<tr>
<td></td>
<td>(14,030)</td>
<td>(3,958)</td>
<td>(68.93)</td>
<td>(9,247)</td>
<td>(0.0310)</td>
<td>(0.0930)</td>
</tr>
<tr>
<td>Observations</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.769</td>
<td>0.306</td>
<td>0.817</td>
<td>0.788</td>
<td>0.639</td>
<td>0.145</td>
</tr>
</tbody>
</table>

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1
Table A.6: The Dynamic Effects of Computerization – Robustness to Additional Controls

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
<th>(9)</th>
<th>(10)</th>
<th>(11)</th>
<th>(12)</th>
<th>(13)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Baseline</td>
<td>Export</td>
<td>Rebates,</td>
<td>Import and</td>
<td>Export</td>
<td>Duty</td>
<td>Export</td>
<td>Growth</td>
<td>1998-2000</td>
<td>Year</td>
<td>Province</td>
<td>HHI 1998-</td>
<td>2000 x Year</td>
</tr>
<tr>
<td>Non-deductible share × 2002-2003</td>
<td>1.203**</td>
<td>1.113**</td>
<td>1.221**</td>
<td>1.193**</td>
<td>1.217**</td>
<td>1.215**</td>
<td>1.202**</td>
<td>1.210**</td>
<td>1.291**</td>
<td>1.139**</td>
<td>884.8</td>
<td>1,266**</td>
<td>1,203</td>
</tr>
<tr>
<td></td>
<td>(551.9)</td>
<td>(536.9)</td>
<td>(541.8)</td>
<td>(542.4)</td>
<td>(546.9)</td>
<td>(551.3)</td>
<td>(550.6)</td>
<td>(547.9)</td>
<td>(566.3)</td>
<td>(555.5)</td>
<td>(568.6)</td>
<td>(577.1)</td>
<td>(1008)</td>
</tr>
<tr>
<td>Non-deductible share × 2004-2005</td>
<td>2.267***</td>
<td>2.154***</td>
<td>2.287***</td>
<td>2.236***</td>
<td>2.279***</td>
<td>2.218***</td>
<td>2.290***</td>
<td>2.270***</td>
<td>2.424***</td>
<td>2.248***</td>
<td>1,828**</td>
<td>2.281***</td>
<td>2.267</td>
</tr>
<tr>
<td></td>
<td>(594.4)</td>
<td>(560.4)</td>
<td>(594.1)</td>
<td>(575.4)</td>
<td>(593.7)</td>
<td>(588.3)</td>
<td>(601.8)</td>
<td>(577.4)</td>
<td>(593.8)</td>
<td>(596.7)</td>
<td>(584.5)</td>
<td>(500.7)</td>
<td>(1004)</td>
</tr>
<tr>
<td>Non-deductible share × 2006-2007</td>
<td>2.047***</td>
<td>1.905***</td>
<td>2.066***</td>
<td>1.955***</td>
<td>2.048***</td>
<td>1.905***</td>
<td>2.057***</td>
<td>2.053***</td>
<td>2.247***</td>
<td>2.141***</td>
<td>1,560**</td>
<td>2.084***</td>
<td>2.047</td>
</tr>
<tr>
<td></td>
<td>(720.8)</td>
<td>(697.5)</td>
<td>(713.5)</td>
<td>(697.4)</td>
<td>(722.3)</td>
<td>(701.3)</td>
<td>(734.3)</td>
<td>(699.1)</td>
<td>(724.4)</td>
<td>(735.3)</td>
<td>(723.0)</td>
<td>(714.1)</td>
<td>(1036)</td>
</tr>
<tr>
<td>Observations</td>
<td>180,103</td>
<td>180,012</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,102</td>
<td>158,820</td>
<td>145,036</td>
<td>180,103</td>
<td>180,148</td>
<td>180,103</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
<td>0.711</td>
<td>0.702</td>
<td>0.702</td>
<td>0.702</td>
</tr>
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</table>

Panel B: Dependent Variable - Sales

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
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<th>(10)</th>
<th>(11)</th>
<th>(12)</th>
<th>(13)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(7,901)</td>
<td>(8,581)</td>
<td>(7,303)</td>
<td>(7,103)</td>
<td>(7,883)</td>
<td>(7,449)</td>
<td>(7,872)</td>
<td>(7,932)</td>
<td>(7,932)</td>
<td>(8,566)</td>
<td>(8,429)</td>
<td>(9,373)</td>
<td>(2,38)</td>
</tr>
<tr>
<td></td>
<td>(13,228)</td>
<td>(11,092)</td>
<td>(12,113)</td>
<td>(11,351)</td>
<td>(13,257)</td>
<td>(12,609)</td>
<td>(13,265)</td>
<td>(13,497)</td>
<td>(13,497)</td>
<td>(14,041)</td>
<td>(14,123)</td>
<td>(15,015)</td>
<td>(1004)</td>
</tr>
<tr>
<td>Observations</td>
<td>180,103</td>
<td>180,012</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,107</td>
<td>187,026</td>
<td>180,102</td>
<td>180,102</td>
<td>180,018</td>
<td>180,103</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.749</td>
<td>0.770</td>
<td>0.769</td>
<td>0.769</td>
<td>0.774</td>
<td>0.771</td>
<td>0.770</td>
<td>0.783</td>
<td>0.777</td>
<td>0.783</td>
<td>0.783</td>
<td>0.769</td>
<td>0.749</td>
</tr>
</tbody>
</table>

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects, and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Additional controls are stated in the column headings. Standard errors are clustered at the sector level. Standard errors are reported in parentheses, while p-values are reported in brackets. *** p<0.01, ** p<0.05, * p<0.1.
Table A.7: The Dynamic Effects of Computerization – Robustness to Additional Controls (Cont.)

<table>
<thead>
<tr>
<th>(1)</th>
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<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
<th>(9)</th>
<th>(10)</th>
<th>(11)</th>
<th>(12)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Baseline</td>
<td>Pre-Reform Firm Correlates x Year FE</td>
<td>Export Rebates, Import and Export Duties</td>
<td>Export Growth 1998-2000 x Year FE</td>
<td>HHI 1998-2000 x Year FE</td>
<td>Province FE x Year FE</td>
<td>Sector-Year Imports and Exports</td>
<td>Omit Liaoning, Jilin and Heilongjiang 2004-2007</td>
<td>Ownership Category x Year FE</td>
<td>Firm Distance from County Seat x Year FE</td>
<td>2001 Share of County Revenues from VAT</td>
<td>Sectoral SD of Firm NDS x Year FE</td>
</tr>
<tr>
<td>Observations</td>
<td>180,103</td>
<td>180,012</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,102</td>
<td>158,820</td>
<td>145,036</td>
<td>180,103</td>
<td>180,148</td>
<td>180,103</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.788</td>
<td>0.789</td>
<td>0.788</td>
<td>0.789</td>
<td>0.792</td>
<td>0.790</td>
<td>0.789</td>
<td>0.789</td>
<td>0.802</td>
<td>0.798</td>
<td>0.788</td>
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</table>

Panel D. Dependent Variable - TFPR (DLW)

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<tr>
<th>(1)</th>
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<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
<th>(9)</th>
<th>(10)</th>
<th>(11)</th>
<th>(12)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Non-deductible share × 2002-2003</td>
<td>0.0925</td>
<td>-0.0728</td>
<td>-0.0944</td>
<td>-0.132</td>
<td>-0.0921</td>
<td>0.0953</td>
<td>0.0908</td>
<td>0.101</td>
<td>-0.0716</td>
<td>-0.0844</td>
<td>-0.128</td>
</tr>
<tr>
<td>(0.276)</td>
<td>(0.294)</td>
<td>(0.275)</td>
<td>(0.274)</td>
<td>(0.277)</td>
<td>(0.275)</td>
<td>(0.276)</td>
<td>(0.276)</td>
<td>(0.289)</td>
<td>(0.289)</td>
<td>(0.280)</td>
<td>(0.285)</td>
</tr>
<tr>
<td>Non-deductible share × 2004-2005</td>
<td>2.435***</td>
<td>2.489***</td>
<td>2.421***</td>
<td>2.358***</td>
<td>2.431***</td>
<td>2.380***</td>
<td>2.415***</td>
<td>2.425***</td>
<td>2.505***</td>
<td>2.452***</td>
<td>2.348***</td>
</tr>
<tr>
<td>(0.678)</td>
<td>(0.645)</td>
<td>(0.665)</td>
<td>(0.660)</td>
<td>(0.678)</td>
<td>(0.685)</td>
<td>(0.682)</td>
<td>(0.682)</td>
<td>(0.641)</td>
<td>(0.633)</td>
<td>(0.707)</td>
<td>(0.592)</td>
</tr>
<tr>
<td>(1.215)</td>
<td>(1.166)</td>
<td>(1.183)</td>
<td>(1.123)</td>
<td>(1.211)</td>
<td>(1.212)</td>
<td>(1.220)</td>
<td>(1.223)</td>
<td>(1.181)</td>
<td>(1.151)</td>
<td>(1.225)</td>
<td>(1.059)</td>
</tr>
<tr>
<td>Observations</td>
<td>180,103</td>
<td>180,012</td>
<td>180,103</td>
<td>180,103</td>
<td>180,103</td>
<td>180,107</td>
<td>177,026</td>
<td>180,102</td>
<td>158,820</td>
<td>145,036</td>
<td>180,103</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.669</td>
<td>0.661</td>
<td>0.660</td>
<td>0.661</td>
<td>0.660</td>
<td>0.665</td>
<td>0.665</td>
<td>0.660</td>
<td>0.670</td>
<td>0.671</td>
<td>0.660</td>
</tr>
</tbody>
</table>

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects, and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Additional controls are stated in the column headings. Standard errors are clustered at the sector level. Standard errors are reported in parentheses, while p-values are reported in brackets. *** p<0.01, ** p<0.05, * p<0.1.
Table A.8: The Dynamic Effects of Computerization on VAT and Firm Behavior – 2SLS with Mexican NDS × Post Periods as Instruments

<table>
<thead>
<tr>
<th>Non-deductible share x 2002-2003</th>
<th>VAT (1)</th>
<th>Sales (2)</th>
<th>Ineligible Sales (3)</th>
<th>Employees (4)</th>
<th>Intermediate Inputs (5)</th>
<th>All Deductible (6)</th>
<th>TFPR DLW (7)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>3,897**</td>
<td>-13,624</td>
<td>7,800</td>
<td>-330.1*</td>
<td>-6,918</td>
<td>0.0728</td>
<td>-42,711*</td>
</tr>
<tr>
<td></td>
<td>(1,856)</td>
<td>(23,186)</td>
<td>(12,822)</td>
<td>(186.6)</td>
<td>(16,687)</td>
<td>(0.0832)</td>
<td>(21,772)</td>
</tr>
<tr>
<td>Non-deductible share x 2004-2005</td>
<td>5,951**</td>
<td>-40,106</td>
<td>3,551</td>
<td>-448.2</td>
<td>-28,726</td>
<td>0.0371</td>
<td>-76,245**</td>
</tr>
<tr>
<td></td>
<td>(2,457)</td>
<td>(40,661)</td>
<td>(18,860)</td>
<td>(288.2)</td>
<td>(27,429)</td>
<td>(0.127)</td>
<td>(37,198)</td>
</tr>
<tr>
<td>Non-deductible share x 2006-2007</td>
<td>4,972*</td>
<td>-92,153</td>
<td>-43,549*</td>
<td>-626.5</td>
<td>-48,398</td>
<td>-0.0138</td>
<td>-115,524**</td>
</tr>
<tr>
<td></td>
<td>(2,831)</td>
<td>(71,783)</td>
<td>(23,913)</td>
<td>(382.2)</td>
<td>(40,201)</td>
<td>(0.185)</td>
<td>(55,667)</td>
</tr>
</tbody>
</table>

| Observations                    | 180,026| 180,026 | 180,026              | 180,026      | 180,026                | 180,026           | 180,026     | 180,026 |
| Weak-ID F-statistic             | 0.053  | 0.127   | 0.004                | 0.019        | 0.073                  | 0.042             | 0.040       | 0.079   |
| H0: β1 = β2 (p-value)           | 0.104  | 0.322   | 0.766                | 0.379        | 0.220                  | 0.529             | 0.156       | 0.00300 |
| H0: β2 = β3 (p-value)           | 0.422  | 0.168   | 0.0420               | 0.111        | 0.289                  | 0.468             | 0.149       | 0.0100  |
| H0: β1 = β3 (p-value)           | 0.060  | 0.188   | 0.0460               | 0.189        | 0.189                  | 0.451             | 0.103       | 0.00100 |

| Non-deductible share x 2002-2003| 800.1**| -2,775  | 1,617                | -67.75*      | -1,407                 | 0.0150            | -8,753**    | 0.319*  |
|                                 | (320.9)| (4,741) | (2,573)              | (37.21)      | (3,418)                | (0.0172)          | (3,610)     | (0.171) |
| Non-deductible share x 2004-2005| 1,222***| -8,226  | 734.1                | -91.97       | -5,894                 | 0.00761           | -15,644**   | 1.819*** |
|                                 | (418.1)| (8,239) | (3,861)              | (59.79)      | (5,454)                | (0.0261)          | (6,582)     | (0.307) |
| Non-deductible share x 2006-2007| 1,018**| -18,862 | -8,910**             | -128.3       | -9,906                 | -0.00281          | -23,650**   | 2.317*** |
|                                 | (517.3)| (14,220)| (4,363)              | (78.07)      | (8,018)                | (0.0378)          | (10,573)    | (0.465) |

| Observations                    | 180,026| 180,026 | 180,026              | 180,026      | 180,026                | 180,026           | 180,026     | 180,026 |
| R-squared                       | 0.702  | 0.796   | 0.506                | 0.818        | 0.788                  | 0.638             | 0.503       | 0.655   |
| H0: β1 = β2 (p-value)           | 0.0930 | 0.315   | 0.764                | 0.395        | 0.197                  | 0.523             | 0.146       | <0.001  |
| H0: β2 = β3 (p-value)           | 0.416  | 0.147   | 0.0160               | 0.0980       | 0.289                  | 0.438             | 0.160       | 0.0320  |
| H0: β1 = β3 (p-value)           | 0.601  | 0.172   | 0.0180               | 0.196        | 0.175                  | 0.429             | 0.105       | <0.001  |

Notes: The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1
Table A.9: The Dynamic Effects of Computerization on VAT and Firm Behavior – 2SLS with U.S. NDS × Post Periods as Instruments

<table>
<thead>
<tr>
<th>Dependent Variables</th>
<th>VAT (1)</th>
<th>Sales (2)</th>
<th>Ineligible Sales (3)</th>
<th>Employees (4)</th>
<th>Intermediate Inputs (5)</th>
<th>Intermediate Inputs as a Share of Total Input (6)</th>
<th>TFPR DLW (7)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Non-deductible share × 2002-2003 (β1)</strong></td>
<td>1,995</td>
<td>-5,939</td>
<td>7,015</td>
<td>45.29</td>
<td>-26,755**</td>
<td>-0.0265</td>
<td>0.320</td>
</tr>
<tr>
<td>(1,471)</td>
<td>(18,265)</td>
<td>(10,201)</td>
<td>(130.2)</td>
<td>(13,209)</td>
<td>(0.0508)</td>
<td>(0.242)</td>
<td>(0.694)</td>
</tr>
<tr>
<td><strong>Non-deductible share × 2004-2005 (β2)</strong></td>
<td>2,709*</td>
<td>-68,169**</td>
<td>27,723</td>
<td>199.9</td>
<td>-54,854***</td>
<td>-0.149*</td>
<td>4.552***</td>
</tr>
<tr>
<td>(1,623)</td>
<td>(28,928)</td>
<td>(18,241)</td>
<td>(214.7)</td>
<td>(20,538)</td>
<td>(0.0865)</td>
<td>(0.314)</td>
<td>(1.354)</td>
</tr>
<tr>
<td><strong>Non-deductible share × 2006-2007 (β3)</strong></td>
<td>1,963</td>
<td>-105,635**</td>
<td>-10,076</td>
<td>267.0</td>
<td>-66,124**</td>
<td>-0.265**</td>
<td>7.262***</td>
</tr>
<tr>
<td>(1,865)</td>
<td>(51,421)</td>
<td>(21,958)</td>
<td>(289.9)</td>
<td>(29,397)</td>
<td>(0.1260)</td>
<td>(0.386)</td>
<td>(1.977)</td>
</tr>
</tbody>
</table>

**Notes:** The sample is a balanced panel of firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform firm sales and average pre-reform firm VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1
Table A.10: The Dynamic Effect of Computerization on Firm Behavior – Full Sample

<table>
<thead>
<tr>
<th>Dependent Variables</th>
<th>VAT Sales</th>
<th>Ineligible Sales</th>
<th>Employees</th>
<th>Intermediate Inputs</th>
<th>Intermediate Inputs as a Share of Total Input</th>
<th>TFPR DLW</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
<td>(5)</td>
<td>(6)</td>
</tr>
<tr>
<td>Dep Var Mean</td>
<td>1538</td>
<td>47339</td>
<td>6719</td>
<td>311.9</td>
<td>33861</td>
<td>0.847</td>
</tr>
<tr>
<td></td>
<td>Non-deductible share × 2002-2003</td>
<td>-11,220*</td>
<td>-639.6</td>
<td>-12.72</td>
<td>-7,319*</td>
<td>-0.00461</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(353.0)</td>
<td>(2,217)</td>
<td>(45.04)</td>
<td>(4,398)</td>
<td>(0.0141)</td>
</tr>
<tr>
<td></td>
<td>Non-deductible share × 2004-2005</td>
<td>-30,582***</td>
<td>-8,518**</td>
<td>-66.07</td>
<td>-19,831**</td>
<td>-0.0290</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(478.7)</td>
<td>(3,621)</td>
<td>(79.07)</td>
<td>(7,721)</td>
<td>(0.0309)</td>
</tr>
<tr>
<td></td>
<td>Non-deductible share × 2006-2007</td>
<td>-45,593***</td>
<td>-15,501***</td>
<td>-117.2</td>
<td>-24,205***</td>
<td>-0.0315</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(568.4)</td>
<td>(4,621)</td>
<td>(102.5)</td>
<td>(9,170)</td>
<td>(0.0449)</td>
</tr>
<tr>
<td>Observations</td>
<td>711,643</td>
<td>711,643</td>
<td>711,643</td>
<td>711,643</td>
<td>711,643</td>
<td>711,643</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.612</td>
<td>0.668</td>
<td>0.087</td>
<td>0.370</td>
<td>0.693</td>
<td>0.129</td>
</tr>
<tr>
<td>H0: β1=β2 (p-value)</td>
<td>0.00200</td>
<td>0.0150</td>
<td>0.00700</td>
<td>0.176</td>
<td>0.0170</td>
<td>0.224</td>
</tr>
<tr>
<td>H0: β2=β3 (p-value)</td>
<td>0.289</td>
<td>0.0630</td>
<td>0.0870</td>
<td>0.0970</td>
<td>0.248</td>
<td>0.876</td>
</tr>
<tr>
<td>H0: β1=β3 (p-value)</td>
<td>0.188</td>
<td>0.0130</td>
<td>&lt;0.001</td>
<td>0.104</td>
<td>0.0170</td>
<td>0.424</td>
</tr>
</tbody>
</table>

Notes: The sample includes all firms covering 1998-2007. All regressions include firm fixed effects, year fixed effects and the interactions of year fixed effects with average pre-reform sector sales and average pre-reform sector VAT. Standard errors are clustered at the sector level. *** p<0.01, ** p<0.05, * p<0.1
Table A.11: Correlates of Tax Officials Before and After 2002

<table>
<thead>
<tr>
<th></th>
<th>1998-2000 (1)</th>
<th>2001-2007 (2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Non-Deductible Share</td>
<td>-30.86***</td>
<td>-8.550**</td>
</tr>
<tr>
<td></td>
<td>(6.134)</td>
<td>(4.122)</td>
</tr>
<tr>
<td>Beta Coef.</td>
<td>-0.278</td>
<td>-0.0726</td>
</tr>
<tr>
<td>Ruggedness</td>
<td>-0.0317</td>
<td>-0.0193</td>
</tr>
<tr>
<td></td>
<td>(0.0591)</td>
<td>(0.0300)</td>
</tr>
<tr>
<td>Beta Coef.</td>
<td>-0.0279</td>
<td>-0.0178</td>
</tr>
<tr>
<td>Ln Area (Square km)</td>
<td>0.111***</td>
<td>0.0898***</td>
</tr>
<tr>
<td></td>
<td>(0.0401)</td>
<td>(0.0243)</td>
</tr>
<tr>
<td>Beta Coef.</td>
<td>0.160</td>
<td>0.135</td>
</tr>
<tr>
<td>Ln Population</td>
<td>0.558***</td>
<td>0.636***</td>
</tr>
<tr>
<td></td>
<td>(0.0768)</td>
<td>(0.0435)</td>
</tr>
<tr>
<td>Beta Coef.</td>
<td>0.584</td>
<td>0.685</td>
</tr>
<tr>
<td>Ln # Firms</td>
<td>0.120**</td>
<td>0.152***</td>
</tr>
<tr>
<td></td>
<td>(0.0588)</td>
<td>(0.0218)</td>
</tr>
<tr>
<td>Beta Coef.</td>
<td>0.202</td>
<td>0.284</td>
</tr>
<tr>
<td>Observations</td>
<td>121</td>
<td>186</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.875</td>
<td>0.923</td>
</tr>
</tbody>
</table>

Notes: This sample comprises of a panel of provinces. All regressions control for year fixed effects. The observations are at the province-year level. Robust standard errors are presented in the parentheses. *** p<0.01, ** p<0.05, * p<0.1. Data are reported by the *Tax Yearbook of China*. Standardized beta coefficients are presented in italics.
Figure A.1: Scatterplot of Firm NDS Measure and I-O Table NDS Measures

(a) Change in VAT/Sales, Before and After 2001

(b) Pre-2001 Firm and Chinese I-O NDS

(c) Pre-2001 Firm and U.S. I-O NDS

(d) Pre-2001 Firm and Mexican I-O NDS
Figure A.2: The Effect of VAT and Firm Outcomes with Random Permutations of NDS

(a) VAT

(b) Sales

(c) Employees

(d) Intermediate Inputs

(e) TFPR (DLW)
Figure A.3: The Effect of VAT and Firm Outcomes with Random Permutations of Treated Years

(a) VAT

(b) Sales

(c) Employees

(d) Intermediate Inputs

(e) TFPR (DLW)