Unions and Inequality Over the Twentieth Century: New Evidence from Survey Data

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July 2018

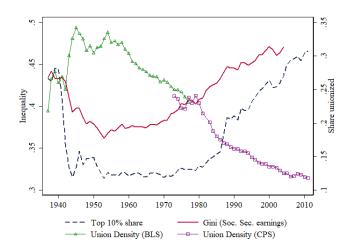
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Unions and inequality, past and present

- Large literature argues for causal effect of unions on inequality (Card (2001); DiNardo *et al.* (1996); Western and Rosenfeld (2011)).
 - ▶ Even larger literature emphasizes market forces as determinants of inequality (Autor *et al.*, 2008).
- Time series variation over the 20th century quite *suggestive* of inverse relationship between union density and income inequality.

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Top income shares and union density



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Limitations of this picture

- Prevailing (macro) view is that increasing inequality and de-unionization are jointly driven by market forces (Acemoglu et al. (2001); Dinlersoz and Greenwood (2012); Açıkgöz and Kaymak (2014))
- But data limitations have prevented investigation of alternative explanations
 - Census Bureau doesn't have a consistent question on union membership until 1970s CPS.
 - Earlier data come from aggregate union reports (with major data quality issues)
 - ▶ Pre-CPS, cannot look at union wage premium, demographic selection into unions, or state-level trends
- This project tries to address these limitations.

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Our paper

- Develop new household-level dataset from over 500 Gallup and ANES opinion polls beginning in 1936
 - ▶ Include union membership, demographics, political views
 - Span period of unions' rise, heyday, and decline
- Document empirical patterns of pre-CPS union membership
 - ▶ Selection into unions by race and education
 - ▶ Union wage premium over time
- Directly estimate unions' effect on inequality
 - ▶ Distributional regressions (RIFs) separately by year (micro effect).
 - ▶ Plenty of reasons to think micro effect misses many macro effects (unemployment, tech. change, returns to capital/skilled labor).
 - □ Time-series regressions as in Goldin and Katz (2009)
 - □ State-year panel regressions
 - □ No clean instrument, but pattern of results consistent.

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Preview of Results

- Highlights from descriptive results
 - ► Stable union premium over time (10-20%)
 - ▶ *u*-shaped selection pattern: mid-century union members were relatively less skilled than either today or pre-WW2.
 - ▶ Consistent with a causal theory of unions-inequality relationship
- Distributional regressions, time-series regressions, and state-year panel regressions all suggest increased union density lowers inequality.

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Outline

- Background and Data
 - Gallup opinion polls
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 - Selection into unions by education
 - Selection into unions by race
- 3 Estimates of the union premium over the last nine decades
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 - Unconditional quantile regression analysis
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What do Unions do (to inequality)?

- Traditional mechanism: union wage premium lowers skill differentials.
 - ▶ Old view: union premium a monopoly distortion. (Rees 1963).
 - ▶ More recent view: union premium reflects product market rent-sharing. (e.g. Abowd and Lemieux 1993).
 - ► Even more recent: union premium counteracts monopsony power (Benmelech et al. 2018)
- Early literature (Lewis 1960) emphasized inequality between covered and uncovered sectors.
 - ▶ Widespread belief that unions *increased* inequality.
 - ▶ Freeman and Medoff 1984: within-sector compression bigger than between-sector difference.
- But unions can affect income distribution beyond effect on its members wages ("threat" effect, negative spillovers, political economy mechanisms, fairness norms).

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Gallup survey data

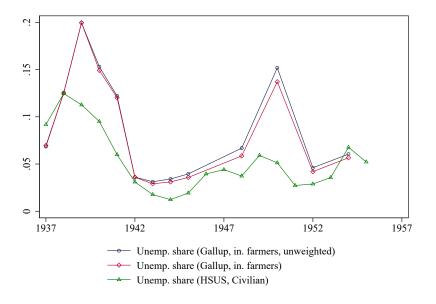
- Over 500 surveys, $N \approx 980,000$
- Repeated cross-section at \approx monthly frequency from 1936-1986
- Nationally representative*, cell-weighted by race-region Sampling
- Include union membership, gender, race, education, age, occupation, and a variety of political opinions

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Typical Gallup codebook (1946)

21. It has been suggested that the Government take over most of the wheat and fats in this country to feed people in officer nations during the next 60 days. This would leave very little for people in this country during that time. Would you favor or oppose such a plan? 21. Favor 22. In politics, as of today, do you consider yourself a Republican, Democrat, Socialist, or Independent? 24. Rep. 2 Dem. 3 Socialist 3 Independent	25. What is the lastade or class you completed in chool? 1
23. What is your occupation? (Record SPECIFIC occupation, not just industry or name of organization worked for). Write in description of occupation: Lell Columbia. (S. 45, 45. (If housewife, widow or student, record occupation of head of family. If retired or unemployed, give former occupation).	26a. Are you (or is your husband), a member of a labor union? If YES, ask: \(\)_{\bar{\bar{\bar{\bar{\bar{\bar{\bar
24a. Do you remember FOR CERTAIN whether or not you voted in the 1944 presidential election? 1 Yes, voted 2 No, didn't vote 3 No, too young to vote 4 Don't remember 1F YES, VOTED, ask: 1b. Did you vote for Dewey, Roosevelt or Thomas? 4 Dewey 6 Roosevelt 6 Thomas 7 Other	If YES, ask: b. Is the telephone listed either under your name or the name of a member of your immediate family? 5. If Yes ASK ONLY OF MEN: 28. Did you serve in any branch of the armed forces in World of War II? 51 II Yes 2 No

Check: Unemployment Reasonably Close to HSUS



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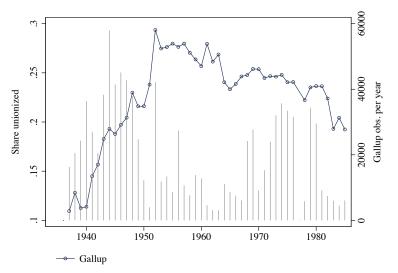
Typical Gallup union question

- "Are you (or is your husband) a member of a labor union?"
 - ▶ Most (but not all) years choices are: "neither," "yes, I am," "yes, he is," "yes, both are."
 - ▶ We harmonize so that union household coded as one if *either* is a member, zero otherwise.
- Implied unit is an individual or couple though we use it to proxy household union status.
- Limitation: Gallup does not ask industry, so we cannot break into public vs. private-sector union.
- Use other surveys, especially ANES, with union household question: "Does anyone in the household belong to a labor union?" Other Data

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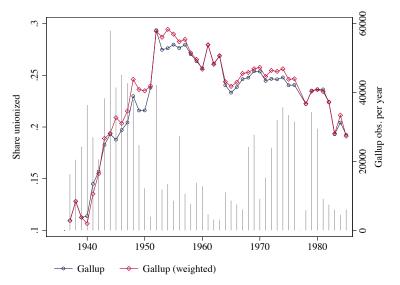
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Comparing our new series to existing series



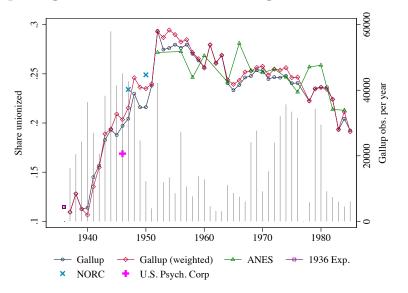
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Comparing our new series to existing series



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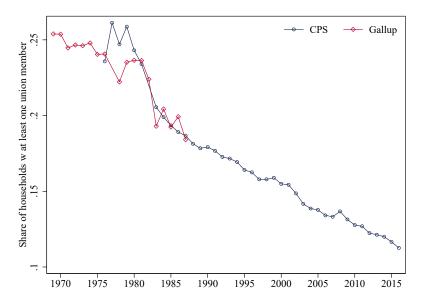
Comparing our new series to existing series



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Comparison with CPS over more recent years



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Estimating selection over time

- Covariate X of interest: education and race
- We estimate, separately by survey source and year y:

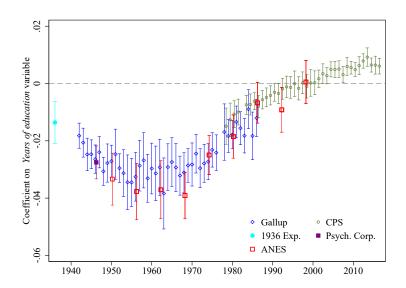
$$union_{hst} = \beta_y X_{hst} + \gamma D_{hst} + \mu_s + \nu_t + e_{hst},$$

where D_{hst} are basic demographics (age & its square, gender), X_{hst} is the covariate of interest, μ_s and ν_t are state and survey-date fixed effects.

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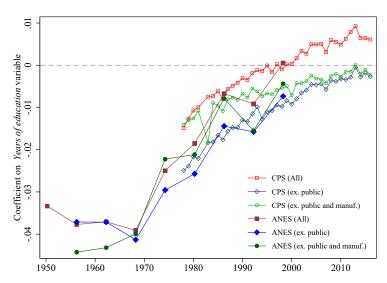
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Selection into unions by education



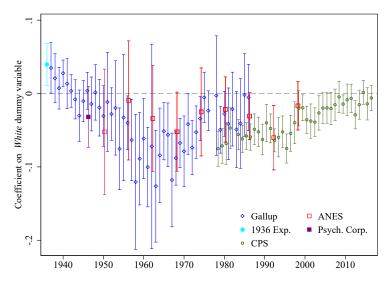
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Not driven by rise of public sector nor fall of manufacturing alone



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Selection into unions by race



Driven by Great Migration? Recall we include state FE, and in Appendix, we show result looks the same if we simply drop the South.

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Unpacking Selection

- Aggregate u shape in selection driven by inverse-u in aggregate density, not just time-quadratic.
- Same relationship between density and selection exists at state-year level, conditional on state + year FE.
- When union density is high union members are much more low-skilled.

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Selection and Drivers of Inequality

As union density increases, the marginal member is increasingly negatively selected

- Consistent with causal effect of unions on inequality.
 - ▶ Inequality increases as low-skilled no longer receive union premia
- Inconsistent with some SBTC models (e.g. Acemoglu *et al.* (2001))
 - ▶ Predict unions should become *less* skilled over time, as high skill workers opt-out of union

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Household income premium

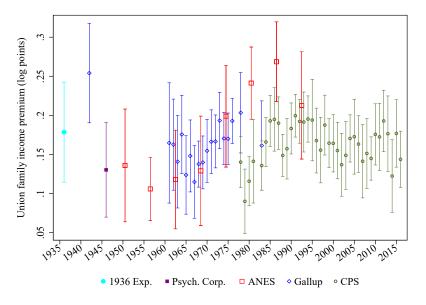
- Polling data typically ask about *household* income, whereas union wage usually estimated on *individual earnings*.
- We estimate the following household income function, separately by year and data source:

$$\log(y_{hst}^{HH}) = \beta Union_h^{HH} + \gamma_1 Black_h^R + \gamma_2 Female_h^R + f(age^R) + g(education_h^R) + k(employment_h) + \lambda_s + \mu_t + e_{hst},$$

- where h denotes household, R respondent, s state, and t survey number.
- In Appendix we show results controlling for occupation of household head, but categories vary across surveys.
- Note an implicit assumption is assortative matching across households. All covars have expected signs (see Appendix).

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Union Family Income Premium, 1936–2016.



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Summary of union premium results

- Evidence suggests that union households were far better off than their education or demographics would suggest. Premium reasonably steady (given wide std. errors) over the 1936 to 2016 period.
- How to compare family premium with traditional wage premium?
 - ► Family premium *not* driven by more workers per household.
 - ▶ Union households have slightly more workers, mechanically, but ANES data suggest little or no bias.
- Also show significant non-wage benefits: union household more likely to have paid vacations back in 1940s (see Appendix), especially low-status households.
- Rents: Union households more likely to say hard to find as good a job as this one.

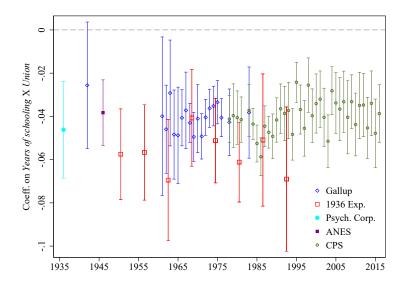
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Did unions benefit certain groups more?

• We interact the $Union^{HH}$ var with, respectively, years of education and then a white dummy.

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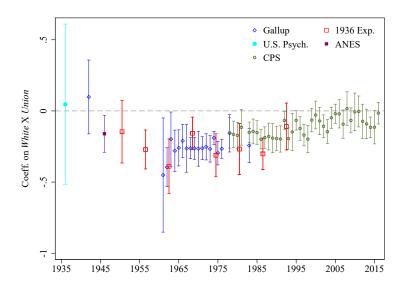
Differential premium by years of education



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Differential premium by race



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Unions and inequality

- Have shown in the micro data that mechanisms by which unions might reduce inequality active in the historical period.
 - ▶ High union density associated with lower-skilled union membership and unions increase relative wages of low-skilled workers.
- Now look directly at income distribution
 - ▶ Distributional regression exercise, as in Firpo et al. 2009.
 - ► Aggregate time series analysis.
 - State-year panel data analysis.

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Unconditional quantile regression analysis

- Adapt methodology from Firpo et al. (2009).
- Let v(F) be some distributional statistic (e.g. 90-10, Gini); u a dummy for union status; y family income; X covariates.
- If so, can decompose distribution F(y, X, u) as

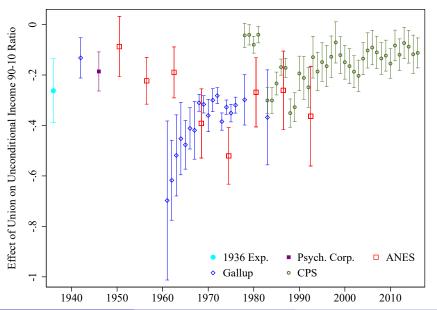
$$Pr(u = 1)F(y, X|u = 1) + [1 - Pr(u = 0)]F(y, X|u = 0).$$

- OLS coefficient from regression of $RIF(v, F, y_i)$ on union u_i gives estimate of $\frac{dv(F)}{dPr(u=1)}$
- Interpretation: How much would inequality fall if you increased the share of union members, holding the joint distribution of wages and covariates constant.

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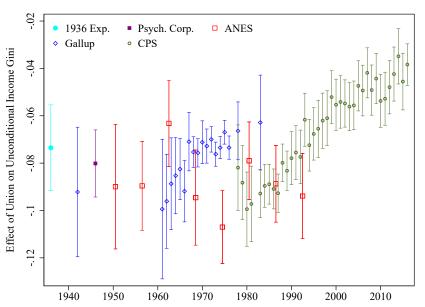
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Results for 90-10 HH income ratio



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Results for Gini coefficient



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Time-series analysis

- We use national time-series at either the annual or decadal level.
 - See Katz Murphy (1992), Autor et al. (1998), Goldin and Katz (2009) and Autor et al. (2008) in analysis of skill shares on skill premia (SBTC and polarization literature).
- We essentially adopt their specifications, but add union density (averaged between BLS and Gallup measures) as an additional explanatory variable and explore other outcome variables.
 - ▶ College premium (decadal until CPS in 1964).
 - ▶ Log 90/10 ratio (decadal until CPS in 1964).
 - ▶ Gini coefficient from Social Security earnings (annual, 1937-2004).
 - ▶ Top10 share (annual since 1937).

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Unions and inequality: Annual time-series

	Dependent variable:					
	Coll. p	remium	90/10	ratio		
	(1)	(2)	(3)	(4)		
Union Density	-1.360**	-1.536**	-1.449***	-1.456**		
	[0.559]	[0.741]	[0.451]	[0.678]		
Mean, dept. var	0.512	0.512	1.376	1.376		
Gallup edu. control?	No	No	No	No		
Addit. controls? Cubic polynomial? Observations	No	Yes	No	Yes		
	Yes	Yes	Yes	Yes		
	49	49	49	49		

All specifications include skill shares. Additional Controls are: Federal minimum wage, national unemployment rate, and top marginal tax rate. Newey-West Standard Errors.

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Unions and inequality: Annual time-series

	Dependent variable:					
	Gini	coeff.	Top 10	share		
	(1)	(2)	(3)	(4)		
Union Density	-0.141**	-0.116**	-35.45**	-16.83		
	[0.0553]	[0.0552]	[14.83]	[12.13]		
Mean, dept. var	0.410	0.410	35.848	35.848		
Gallup edu. control?	Yes	Yes	Yes	Yes		
Addit. controls? Cubic polynomial? Observations	No	Yes	No	Yes		
	Yes	Yes	Yes	Yes		
	65	65	70	70		

All specifications include skill shares. Additional Controls are: Federal minimum wage, national unemployment rate, and top marginal tax rate. Newey-West Standard Errors.

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State-year panel regressions

- In general, the inequality literature hasn't taken a state-year panel approach.
 - ▶ In the SBTC literature, concern is the college-educated will migrate to state-years with high college premia, leading to reverse causality.
 - ► Similarly, unions target places with high profit margins, low-skill workers might migrate to places with unions.
- Nonetheless, we try to exploit variation within states across time:
 - ▶ Absorbing state and year FE, and controlling for policy environment, business cycle, industry mix, and skill shares.

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Regressing state-year inequality measures on state-year union density

	Coll. premium		90/10 ratio		
	(1)	(2)	(3)	(4)	
Household union share			-0.296** [0.118]	-0.244** [0.0953]	
Mean, dept. var.	0.490	0.497	1.386	1.398	
Industry shares State-spec. quad.	No No	Yes Yes	No No	$\begin{array}{c} { m Yes} \\ { m Yes} \end{array}$	
Income covars.	No	Yes	No	Yes	
Policy covars. Observations	No 1640	Yes 1505	No 1640	Yes 1505	

All regs have state and South X year fixed effects. Industry shares are 1-digit employment shares. Income covars are Log GDP/capita and share of hh filing taxes. Policy covars are minimum wage and "policy liberalism" index. SEs clustered by state. *p < 0.1, **p < 0.05, ***p < 0.01

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Regressing state-year inequality measures on state-year union density

	Top 10) Share	Gini coeff.		
	(1)	(2)	(3)	(4)	
Household union share	-0.0635*** [0.0235]	-0.0738*** [0.0235]	-5.342*** [2.065]	-3.147** [1.399]	
Mean, dept. var.	0.376	0.378	36.61	36.96	
Industry shares	No	Yes	No	Yes	
State-spec. quad.	No	Yes	No	Yes	
Income covars.	No	Yes	No	Yes	
Policy covars.	No	Yes	No	Yes	
Observations	1640	1505	3107	2723	

All regs have state and South X year fixed effects. Industry shares are 1-digit employment shares. Income covars are Log GDP/capita and share of hh filing taxes. Policy covars are minimum wage and "policy liberalism" index. SEs clustered by state. *p < 0.1, **p < 0.05, ***p < 0.01

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How much of the change in inequality can unions explain?

Share explained by Δ union density, using β values from...

		0 , 0,	
	Total Δ	RIF (high)	RIF (low)
			
Union density	0.113		
Male $90/10$ ratio	-0.326	0.128	0.0377
Gini coefficient	-0.0585	0.162	0.0892
—1970-2004			
Union density	-0.123		
Male $90/10$ ratio	0.465	0.0980	0.0288
Gini coefficient	0.0875	0.118	0.0651

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How much of the change in inequality can unions explain?

Share explained by Δ union density, using β values from...

	Total Δ	Time-series	State panel	
1940-1960				
Union density	0.113			
Skill premium	-0.0926	1.894	0.529	
Male $90/10$ ratio	-0.326	0.504	0.0844	
Gini coefficient	-0.0585	0.204	0.142	
Top ten percent	-12.77	0.137	0.0278	
—1970-2004				
Union density	-0.123			
Skill premium	0.212	0.904	0.252	
Male $90/10$ ratio	0.465	0.385	0.0645	
Gini coefficient	0.0875	0.149	0.104	
Top ten percent	12.13	0.158	0.0319	

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Concluding thoughts

- New data from political polls allows examination of unionized labor markets during heyday of union power
- We find descriptive results consistent with a causal effect of unions on inequality
 - union premium remains relatively stable over time
 - ightharpoonup selection increasingly less skilled as union density increases
- Direct estimates of unions' effect on inequality negative and significant across a variety of identification strategies
- With care, historical political polling data of considerable value for economic history

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THE END

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Magnitudes

- Implied effect on Gini roughly similar over all three methods. From 1970–2004, decline in union density could "explain" 9-20% of rise in Gini.
- Implied effect on 90/10 varies across methods: smaller for RIF and state-year (10–20% explained), but large for annual time-series (35%).
- Implied effect for college premium: 13% from state-year, 30% for annual time-series
- Implied state-year effect for top 10 percent income share explains between 2.5 and 5 of increase, implied annual effect explains between 30 and 60.
- 12\% decline between 1970-2004 roughly symmetric explanatory power as 11% increase between 1940-1960.
- One point to remember is that state-year reduction in inequality doesn't "aggregate up" to national reduction in inequality.

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Comparing Gallup and Census in 1940

	Gallup	Census	Census	Gallup	Census
Black	0.0290	0.0895	0.0906	0.0325	0.0357
Female	0.338	0.505	0.344	0.341	0.343
Age	40.45	39.61	40.06	40.40	40.55
HS Graduate	0.493	0.278	0.266	0.494	0.290
Northeast	0.0835	0.0660	0.0629	0.0946	0.0854
Mid Atlantic	0.262	0.253	0.241	0.297	0.327
East Central	0.207	0.187	0.186	0.235	0.252
West Central	0.176	0.127	0.129	0.200	0.175
South	0.118	0.258	0.263	0	0
Rocky Mountain	0.0751	0.0284	0.0308	0.0851	0.0418
Pacific Coast	0.0784	0.0754	0.0818	0.0888	0.111
College Graduate		0.0472	0.0499	0.0709	0.0543
Gender/HH adj?	No	No	Yes	No	No
Ex. S/SW?	No	No	No	Yes	Yes
Observations	148290	736832	736832	130400	544375

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	Gallup	Census	Census	Gallup	Census
Professional	0.0780	0.113	0.122	0.0793	0.129
Farmer	0.209	0.156	0.159	0.185	0.109
Propietors, managers, officials	0.0104	0.0928	0.0875	0.0106	0.0933
Clerks (white collar)	0.294	0.0535	0.0539	0.301	0.0609
Skilled workmen and foremen	0.0906			0.0953	
Sales workers		0.0462	0.0457		0.0499
Craftsmen		0.142	0.139		0.153
Operatives		0.146	0.147		0.159
Unskilled or semi-skilled labor	0.190			0.200	
Laborers		0.0932	0.0973		0.0944
Service workers (priv. HH)		0.0103	0.0105		0.00626
Other service workers		0.0477	0.0468		0.0508
No answer, N/A, etc.	0.0826	0.0999	0.0920	0.0836	0.0949
Survey wgts?	No	Yes	Yes	No	Yes
Gender wgts?	No	No	Yes	No	Yes
One obs. per HH?	N/A	No	Yes	N/A	Yes
Ex. S/SW?	No	No	No	Yes	Yes
Observations	148290	736832	736832	130400	544375

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Comparing Gallup and Census, 1950–1980

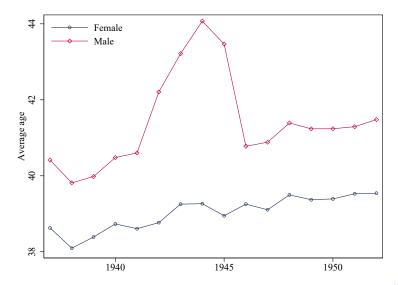
	1950		190	1960		1970		1980	
	Census	Gallup	Census	Gallup	Census	Gallup	Census	Gallup	
South Share —South	0.242	0.117	0.259	0.138	0.271	0.247	0.296	0.256	
Female	0.516	0.505	0.521	0.518	0.529	0.507	0.529	0.503	
Age	44.61	44.31	45.07	47.64	45.94	46.35	45.20	46.13	
Black	0.200	0.0849	0.182	0.147	0.160	0.129	0.159	0.160	
HS grad.	0.294	0.373	0.366	0.372	0.473	0.529	0.619	0.635	
-Non-South									
Female	0.515	0.504	0.517	0.512	0.528	0.506	0.528	0.503	
Age	46.67	43.75	45.96	45.87	46.27	45.38	45.28	44.10	
Black	0.0530	0.0454	0.0611	0.0586	0.0709	0.0614	0.0782	0.0874	
HS grad.	0.385	0.473	0.450	0.531	0.579	0.659	0.710	0.755	
Observ.	296223	182171	5388972	95064	2444218	138098	7475162	128507	





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Gallup data can pick up high-frequency changes in demographics



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Other surveys

- Gallup and ANES allow us to look across large spans of years.
- But we uncovered a few more sources of union micro data, which we use, mostly as a check on Gallup and ANES.
 - ▶ A 1936-1937 BLS Consumption survey asks if you have spent anyone in the household has spent money on union dues, which we use to generate a household union status.
 - ▶ The U.S. Psychological Corporation conducted a 1946 survey that asks union status, family income and standard covariates.
 - ► NORC occasionally has surveys with all these covariates as well.

Gallup sampling before 1950

- Berinsky (2006) provides great detail: quota-based sampling of voters.
- From 1950 onward, more effort to reach representative sample of Americans and to provide weights to correct. We construct our own weights for pre-1950 data.
- Before 1942, we can only adjust by region and race: WhitexSouth (4 cells). From 1942, we adjust by $White \times Educ \times South$ (16 cells). Census Comparison
- Match existing unemployment series (e.g. Roosevelt recession) quite well, and pick up WW2 deployment in age distribution.

 Gallup Unemp.

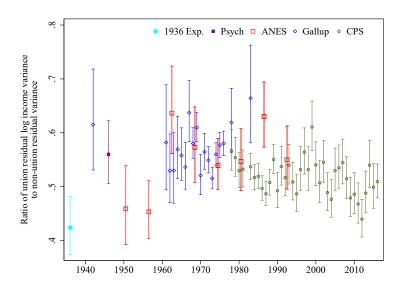
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Does union sector have lower residual income variance?

- A classic result (Card (2001)) from individual-earnings data in CPS is, relative to union sector:
 - greater total earnings variance in non-union sector
 - greater explained earnings variance in union-sector (covariates do more work)
 - but, nonetheless, greater residual earnings variance in non-union sector.
- As differential union effects by race and education suggest, we replicate (2) in our household data. Union sector has lower total income variance (not shown).
- Over our ninety-year period, ratio of union residual variance to non-union is significantly below one.

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Ratio of residual income variance: union v. non-union



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