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

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June 2018

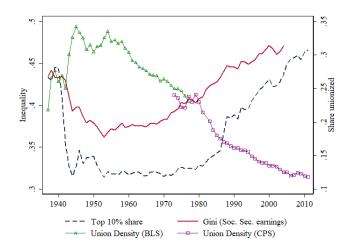
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Unions and inequality, past and present

- Large literature argues for causal effect of unions on inequality (Card [2001]; DiNardo et al. [1996]; Western and Rosenfeld [2011]).
- Time series variation over the 20th century quite *suggestive* of inverse relationship between union density and income inequality.

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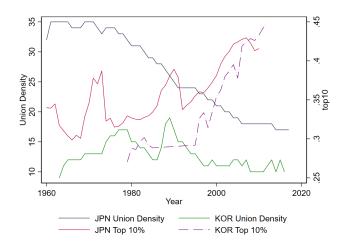
Top income shares and union density



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Not Simply A US Phenomena

• Jaumotte and Osorio [2015] show cross country panel effect of union density on measures of income inequality.



Limitations of this picture

- Prevailing view is that increasing inequality and de-unionization are jointly driven by technological forces (Acemoglu et al. [2001])
- 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 using recentered influence functions (RIFs) within cross-sections
 - ► Time-series regressions as in Goldin and Katz [2009]
 - ► State-year panel regressions

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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

- 1 Background and Data
 - Gallup opinion polls
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- 2 Estimates of the union premium over the last nine decades
 - Did unions benefit certain groups more?
 - Does union sector exhibit lower residual income variance?
- 3 Who joined unions? And why?
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 - State-year panel regressions



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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

, 19a. Will you tell me who Richard M. Nixon is?	- L πep. L νem. L υtner L υπαετιαεα
IF ANY ANSWER, ask b and c: b.tin general, is your opinion of Vice President Nixon favorable or unitavorable? 6/-38 4 Favorable 2 Unfavorable 3 No Opinion	27a. In the election last November, did things come up which kept you from voting, or did you happen to vote? '□ Yes '□ No, didn't'□ No, too '□ Don't Co'. '43 voted vote young remember IF YES, VOTED, ask b and c:
c. Why do you feel that way?	b. Did you vote for Eisenhower or Stevenson? ' Eisenhower ' Stevenson Other
Now, here's another interesting and different kind of question: 20. Do you have a television set in your home? 39 10 Yes 10 Yes	col. 441 Yes ² No 28a. In politics, as of TODAY, do you consider yourself a Democrat, Republican, or Independent?
IF "YES," ask Questions 21, 22, 23, and 24: 21. Will you please tell me the names of your favorite tele-	¹□ Democrat ²□ Republican ³□ Independent col. □, □, □, Other IF INDEPENDENT, ask:
vision shows?	b. As of today, do you lean a little more to the Democratic party or a little more to the Republican party? ** Dem. ** Rep. ** Other

kept you from voting, or o	iid you happen t	o vote?	
¹□ Yes ²□ No, didn't	³ □ No, too	"□ Don't	
:u/.43 voted vote	young	remember	
IF YES, VOTED, ask b and c:			
b. Did you vote for Eisenhower or Stevenson?			
4□ Eisenhower 5□ Stevenson 6□ Other			
c. Was this the first Presiden	tial election you	have voted in?	
201.44¹□ Yes 2□	No		
28a. In politics, as of TODAY Democrat, Republican, or l	7, do you cons Independent?	ider yourself a	
¹□ Democrat ²□ Re	publican ⁸ [Independent	
Col. 75 Other			
 As of today, do you lean party or a little more to t 	a little more to he Republican p	the Democratic arty?	
⁶ □ Dem. ⁶ □ Rep. ⁷ □ C)ther	⁸ □ Undecided	

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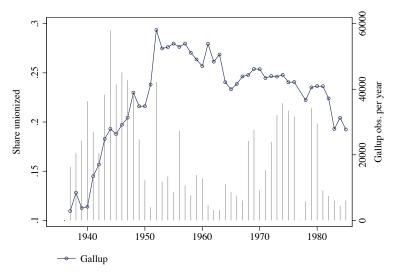
Typical Gallup union question

- "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."
 - ▶ For now, we harmonized so that union household coded as one if *either* is a member, zero otherwise.
 - ► Sometimes asks which union federation (AFL, CIO, other).
- 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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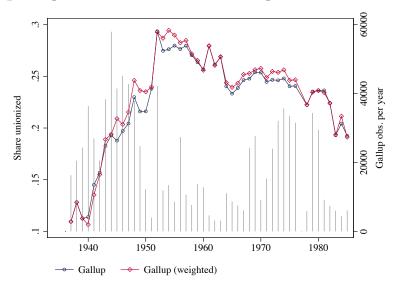
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Comparing our new series to existing series



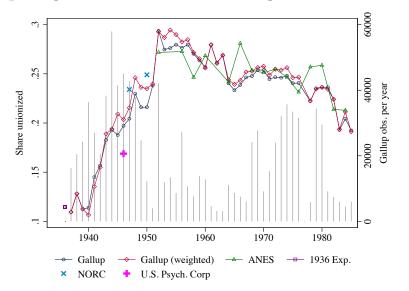
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Comparing our new series to existing series



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Comparing our new series to existing series



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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:

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

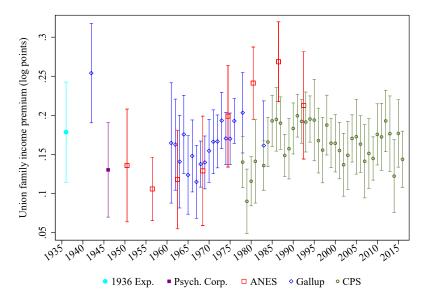
where notation follows naturally.

- 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.
- Also show significant non-wage benefits: union household more likely to have paid vacations back in 1940s (see Appendix), especially low-status households.

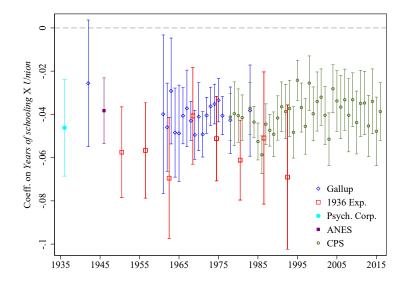
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Did unions benefit certain groups more?

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

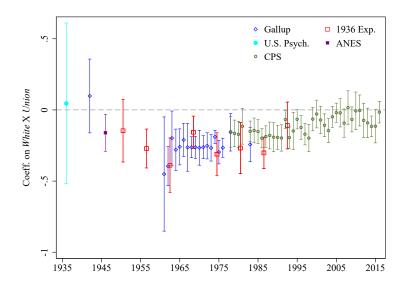
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Differential premium by years of education



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Differential premium by race



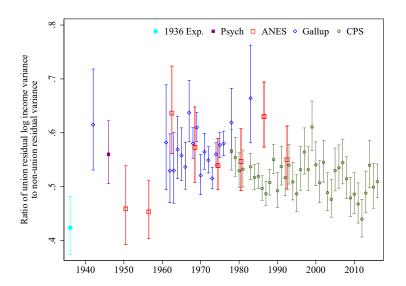
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Does union sector have lower residual income variance?

- 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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Estimating selection over time

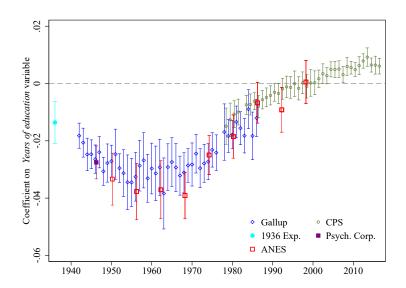
- Covariate X of interest: education and race
- We estimate, separately by year survey source and year y:

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

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.

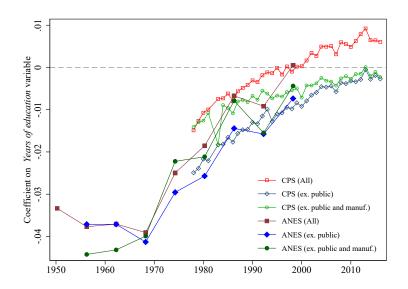
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Selection into unions by education



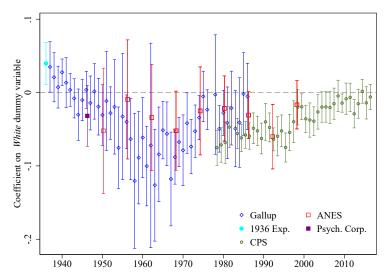
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Not driven by public sector alone



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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.
- When unions are strong they 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 impact story
 - ▶ Inequality increases as low-skilled no longer receive union premia
- Inconsistent with many 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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Unions and inequality

- Have shown in the micro data that mechanisms by which unions might reduce inequality active in the historical period.
- 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

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

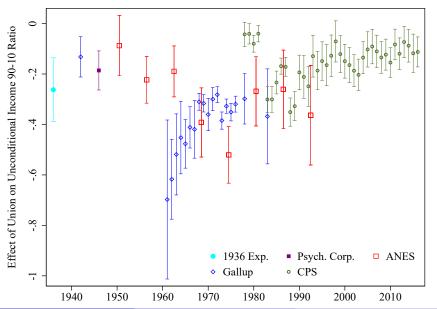
- OLS coefficient from regression of $RIF(v, F, y_i)$ on union u_i gives estimate of $\frac{dv(F)}{dPr(u=1)}$
- Interpretation: How much would inequality fall if you increased the share of union members, holding the joint distribution of wages and covariates constant.

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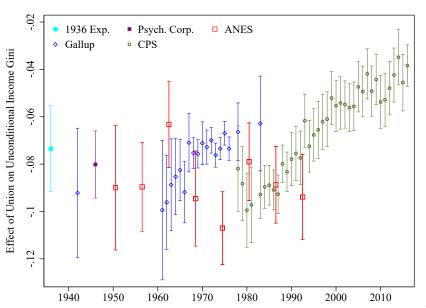
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Results for 90-10 HH income ratio



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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 as an additional explanatory variable and explore other outcome variables.
 - ▶ College premium (decadal until CPS in 1979).
 - ▶ 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	-1.790*** [0.375]	-1.630*** [0.520]	-2.935*** [0.502]	-2.575*** [0.575]		
Mean, dept. var	0.502	0.502	1.620	1.620		
R-squared	0.943	0.944	0.976	0.984		
Gallup edu. control?	No	No	No	No		
Addit. controls?	No	Yes	No	Yes		
BLS IV?	No	Yes	No	Yes		
Cubic polynomial?	Yes	Yes	Yes	Yes		
Observations	43	43	52	52		

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Unions and inequality: Annual time-series

	Dependent variable:					
	Gini	coeff.	Top 10 share			
	(1)	(2)	(3)	(4) -6.985 [12.22]		
Union density	-0.100** [0.0447]	-0.0471 [0.0545]	-35.21*** [11.13]			
Mean, dept. var	0.410	0.410	36.319	36.319		
R-squared	0.985	0.985	0.950	0.964		
Gallup edu. control?	Yes	Yes	Yes	Yes		
Addit. controls?	No	Yes	No	Yes		
BLS IV?	No	Yes	No	Yes		
Cubic polynomial?	Yes	Yes	Yes	Yes		
Observations	65	65	73	73		

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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	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	1640	1505	

All regs have state and year fixed effects, 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			-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			3107	2723	

All regs have state and year fixed effects, SEs clustered by state. p < 0.1, p < 0.05, p < 0.01

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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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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
	Garrap	Consus	Сеньць	Garrap	Сепвав
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

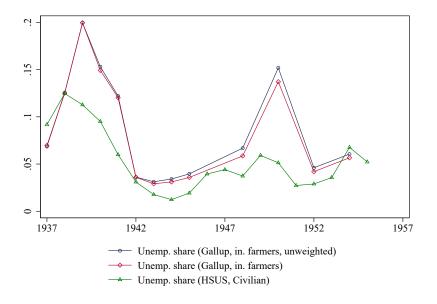
	1950		1960		1970		1980	
	Census	Gallup	Census	Gallup	Census	Gallup	Census	Gallup
South Share —South	0.242	0.117	0.259	0.138	0.271	0.247	0.296	0.256
Female	0.516	0.505	0.521	0.518	0.529	0.507	0.529	0.503
Age	44.61	44.31	45.07	47.64	45.94	46.35	45.20	46.13
Black	0.200	0.0849	0.182	0.147	0.160	0.129	0.159	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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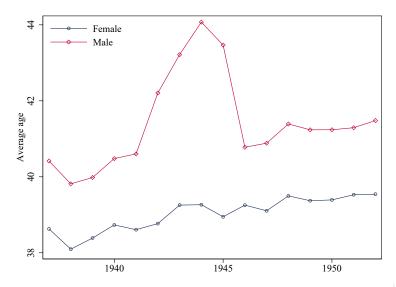
Unemployment Reasonably Close to HSUS



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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.

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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). Consus 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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