# Capital and Labor Income Pareto Exponents in the United States, 1916-2019 

Ji Hyung Lee ${ }^{1} \quad$ Yuya Sasaki ${ }^{2} \quad$ Alexis Akira Toda ${ }^{3} \quad$ Yulong Wang ${ }^{4}$<br>${ }^{1}$ University of Illinois, Urbana-Champaign<br>${ }^{2}$ Vanderbilt University<br>${ }^{3}$ University of California San Diego<br>${ }^{4}$ Syracuse University

Seminar @UCSD<br>June 3, 2022

## Pareto tail

- Random variable $X>0$ has Pareto upper tail if $\mathrm{P}(X>x) \sim x^{-\alpha}$ for large $x$, where $\alpha$ : Pareto exponent
- Discovered by Pareto (1896) for income, but holds for other variables:
- city size (Gabaix, 1999), $\alpha \sim 1$,
- firm size (Axtell, 2001), $\alpha \sim 1$,
- COVID cases (Beare and Toda, 2020), $\alpha \sim 1$ Picture
- household wealth (Klass et al., 2006; Vermeulen, 2018), $\alpha \sim 1.5$
- household consumption (Toda and Walsh, 2015), $\alpha \sim 4$,
- total income (Feenberg and Poterba, 1993; Atkinson and Piketty, 2010), $\alpha \sim 1.5-3$,
- capital income (de Vries and Toda, 2021), $\alpha \sim 1.5$, etc.


## Importance of income and wealth Pareto exponent

- Convenient descriptive statistic for top tail inequality (small $\alpha \Longrightarrow$ high inequality)


## Importance of income and wealth Pareto exponent

- Convenient descriptive statistic for top tail inequality (small $\alpha \Longrightarrow$ high inequality)
- Theory of optimal taxation (Saez, 2001): optimal top tax rate is $\tau=\frac{1-g}{1-g+\alpha e}$, where
- $g \in[0,1]$ : marginal utility weight on top earners,
- e: elasticity of top income w.r.t. tax rate


## Importance of income and wealth Pareto exponent

- Convenient descriptive statistic for top tail inequality (small $\alpha \Longrightarrow$ high inequality)
- Theory of optimal taxation (Saez, 2001): optimal top tax rate is $\tau=\frac{1-g}{1-g+\alpha e}$, where
- $g \in[0,1]$ : marginal utility weight on top earners,
- e: elasticity of top income w.r.t. tax rate
- Calibration of macroeconomic models
- average wealth of agents above some threshold is $\frac{\alpha}{\alpha-1}$ times threshold
- hence wealthy agents have substantial impact on aggregate quantities
- (see Beare and Toda (2022) for determining $\alpha$ in economic models and Gouin-Bonenfant and Toda (2022) for numerically solving models)


## Accurately estimating income $\alpha$ is challenging

- Limitation in data availability
- micro survey data (CPS, SCF, etc.) have small sample size $n=10^{3} \sim 10^{4}$
- survey data suffer from low or inaccurate response
- micro administrative data hard to access (IRS Public Use File available only for 2009-2014 with $\$ 10,000$ per year, noise added to data to protect confidentiality)


## Accurately estimating income $\alpha$ is challenging

- Limitation in data availability
- micro survey data (CPS, SCF, etc.) have small sample size $n=10^{3} \sim 10^{4}$
- survey data suffer from low or inaccurate response
- micro administrative data hard to access (IRS Public Use File available only for 2009-2014 with $\$ 10,000$ per year, noise added to data to protect confidentiality)
- Limitation in applicability of statistical methods
- common methods (Hill, 1975; Gabaix and Ibragimov, 2011) assume availability of micro data
- maximum likelihood can be applied to grouped data if income thresholds observable, but IRS data provides income thresholds only for total income
- Hence existing estimates (i) rarely distinguish capital/labor income, (ii) are likely inaccurate, or (iii) are non-systematic


## What we do

- Estimate capital and labor income Pareto exponents in U.S., 1916-2019, using best data and best estimation method
- distinguishing capital/labor matters because taxed differently
- We use tabulated summaries from IRS Statistics of Income
- administrative data from tax returns (likely accurate)
- publicly available for 1916-2019
- large sample size: $n=10^{6} \sim 10^{8}$
- We apply minimum distance method of Toda and Wang (2021) based on extreme value theory
- can be applied to grouped data
- no need to observe income thresholds
- suffices to observe group averages


## What we find

- Based on $\alpha$, sample period can be divided into three sub periods, pre-1940, 1940-1985, and post-1985
- Post-1985, capital $\alpha \approx 1.2$, labor $\alpha \approx 2.0$
- $\alpha$ lower than existing estimates, hence higher top tail inequality (likely due to underreporting in survey)
- No systematic trend post-1985, so rise in income inequality measured by top income shares (Piketty and Saez, 2003) is inequality between rich and poor, not among rich


## General framework

- Income $\left\{Y_{i}\right\}_{i=1}^{n}$, unobserved by researcher
- Top order statistics $Y_{(1)} \geq Y_{(2)} \geq \cdots \geq Y_{(n)}$
- Partial sums of order statistics $S_{m}:=\sum_{i=1}^{m} Y_{(i)}$
- Observables are $\left\{n_{k}, S_{n_{k}}\right\}_{k=1}^{K}$, where $K$ is number of income groups and $n_{1}<n_{2}<\cdots<n_{K} \leq n$


## Example: U.S. 2019 tax returns data



## Details on data

- Primary data is Statistics of Income (SOI) Individual Tax Returns Publication 1304 from IRS
- Excel spreadsheets available since 1993 (Table 1.4 at https:///wur.irs.gov/statistics/ soi-tax-stats- indi vidual-income-tax-returns-complete-report-publication- 1304 -basic-tables -
- SOI 2019
- Before 1993, only PDFs of scanned copies of SOI are available (https:///wuw.irs.gov/statistics/soi-tax-stats-archive)

```
- SOI }191
```

- Before 1993, manually input adjusted gross income (AGI), AGI thresholds, salaries and wages, and number of returns into spreadsheets
- Human errors inevitable (for each year, we typed 10-digit numbers 100 times); checked accuracy by comparing column sums of spreadsheet to sums reported in SOI tables


## Definitions of incomes

- We define
- Total income $:=$ "adjusted gross income (AGI)"
- Labor income $:=$ "salaries and wages"
- Capital income $:=$ non-labor income $=$ AGI - labor income
- This definition of capital income is broad and includes clearly non-capital income such as "state income tax refunds", "alimony

```
received'', ''unemployment compensation''
```

- Hence also consider adding up capital income components

```
SUCh aS ''taxable interest'', ''tax-exempt interest'', ''ordinary dividends',,
''qualified dividends'', ''business or profession'', ''capital gain distributions
reported on Form 1040'', ''sales of capital assets reported on Form 1040, Schedule D'',
''sales of property other than capital assets'', ''taxable Individual Retirement
Arrangement (IRA) distributions'', ''pensions and annuities'', ''total rent and
royalty'', ''partnership and S corporation'', ''estate and trust''
```


## Capital income $\approx$ non-labor income for AGI $>25 \mathrm{k}$



## Sample size

- Our unit of analysis is tax unit
- individuals or married couples filing jointly, with dependents if any
- We only observe tax filers
- non-filer could have income below filing requirement or work in informal sectors using cash and evade taxes
- Sample size (number of potential tax units) necessary for estimation (definition of top fractiles)
- To estimate sample size, we collect data on
- number of total returns ( $T$ ),
- number of joint returns ( $J$ ),
- number of adults $(A)$,
- number of married couples ( $M$ )


## Number of adults and tax returns

- If $\{$ Adults $\}=\{$ Tax filers $\}$, then $A=T+J$
- Post-1950, $(T+J) / A \approx 0.9$, so missing about $10 \%$ of adults
- Pre-1945, missing 90-99\% of adults due to high exemptions (Tax Reform Act of 1942)



## Income fractiles and joint returns

- Low (high) income earners tend to file separately (jointly)
- Hence can estimate sample size as $n=A-J$
- Still need to estimate $J$ for pre-1950



## Married couples and joint returns

- Post-1950, married/adults $(M / A)$ and joint/adults $(J / A)$ have common trends
- Regress $\log (J / A)$ on $\log (M / A)\left(R^{2}=0.989\right)$ post-1950 and use OLS estimates to construct $\widehat{J}$ pre-1950



## Potential tax units

- Use $n=A-J(\widehat{J})$ (upper bound)
- Using $n=A-M$ (lower bound, assuming all married couples file jointly, as Piketty and Saez (2003) do) has no material impact robustiness



## Estimation

- We now have data on $\left\{\left(n_{k}, S_{n_{k}}\right)\right\}_{k=1}^{K}$ and $n$, where $n_{1}<n_{2}<\cdots<n_{K} \leq n$
- We apply the minimum distance method of Toda and Wang (2021) (TW) to estimate income Pareto exponents
- Here are basic idea of TW method
- Letting $J:[0,1] \rightarrow \mathbb{R}$ a bounded and almost everywhere continuous function, the asymptotic behavior of weighted sums of order statistics

$$
\frac{1}{n} \sum_{i=1}^{n} J\left(\frac{i}{n+1}\right) Y_{(n-i+1)}
$$

is known (Stigler, 1974)

- Let $p_{k}=n_{k} / n$ be top fractile


## Estimation

- Basic idea (continued)
- Letting $J(x)=\mathbb{1}\left(1-p_{n_{k+1}}<x \leq 1-p_{n_{k}}\right)$, we have

$$
\frac{1}{n} \sum_{i=1}^{n} J\left(\frac{i}{n+1}\right) Y_{(n-i+1)}=\frac{S_{n_{k+1}}-S_{n_{k}}}{n}
$$

- Hence if we consider self-normalized quantity

$$
s:=\left(\frac{S_{n_{2}}-S_{n_{1}}}{S_{n_{L+1}}-S_{n_{L}}}, \ldots, \frac{S_{n_{L}}-S_{n_{L-1}}}{S_{n_{L+1}}-S_{n_{L}}}\right),
$$

asymptotic behavior depends only on Pareto exponent $\alpha$ if $\left\{Y_{i}\right\}$ have Pareto upper tail and $n_{L+1} \ll n$

- Can estimate $\alpha$ by minimizing quadratic distance of $s$ from theoretical value implied by Pareto distribution


## Is Pareto tail reasonable?

- If income has Pareto upper tail with exponent $\alpha$, top $p$ fractile income share is $S(p) \propto p^{1-1 / \alpha}$
- Hence top fractiles and shares should be linear in log-log scale



## Assumptions

- For estimation, we need partial sums of order statistics
- In IRS data, income groups are defined by order of AGI
- We assume AGI and capital income are ordered in the same way across income groups (e.g., tax filers in group $k$ have higher capital income than those in group $k+1$ for $k=1, \ldots, L$ )
- Reasonable for capital income if $L$ not too large because AGI $\approx$ capital income for top earners average income
- Unreasonable for labor income because labor income $\ll$ AGI for top earners


## Choice of income groups

- Literature typically uses top $5 \%$ observation for estimation
- We choose largest $L$ such that $n_{L+1} / n \leq 0.01$ (top $1 \%$ ) to be conservative, given large sample size (and need $L+1 \geq 3$ )
- Results are robust to different cutoffs



## AGI and capital income Pareto exponents

- Capital $\alpha \approx 1.2$ pre-1940 and post-1985, inverse U-shape in 1940-1985, AGI $\alpha$ similar pattern
- Standard error omitted because order of magnitude $\left(10^{8} \times 0.01\right)^{-1 / 2}=10^{-3}$



## No rise in top tail inequality post-1985

- Top income shares have risen post-1985
- If income Pareto, then $S(p)=(p / q)^{1-1 / \alpha} S(q)$
- $S(p)$ constructed from $(p, q)=(0.001,0.01)$ and $\alpha=1.5$ is similar to actual $S(p)$ post-1985, confirming stable top tail inequality



## Labor income Pareto exponents

- Several issues when estimating labor income exponents
- Not necessarily reasonable to assume same ordering of AGI and labor income across income groups
- Size distribution of labor income available only for subset of 1927-1978
- For a particular year (1968) the joint distribution of AGI and labor income is available, compare top labor income shares


## Top labor income shares, 1968



## Top labor income shares, 1968



## Joint distribution of AGI and labor income



## Joint distribution of AGI and labor income



## Labor income Pareto exponents

- These figures suggest that AGI and labor income are highly correlated, but excluding top incomes
- We simply report two numbers using labor income ranked exactly (1927-1978) and ranked by AGI (1934-2019)



## Comparison to existing estimates

- We compare to existing estimates,
- maximum likelihood (only AGI due to applicability of estimation method),
- Feenberg and Poterba (1993) (only AGI due to applicability of estimation method),
- Atkinson and Piketty (2010) (only AGI due to applicability of estimation method),
- de Vries and Toda (2021) (capital and labor income exponents)


## Maximum likelihood

- For AGI, we observe income thresholds $\infty=t_{0}>t_{1}>\cdots>t_{K}>0$
- Can apply ML using conditional tail probability $\mathrm{P}\left(Y \geq y \mid Y \geq t_{L}\right)=\left(y / t_{L}\right)^{-\alpha}$



## Feenberg and Poterba (1993)

- Feenberg and Poterba (1993) find two income thresholds $y_{1}<y_{2}$ that bracket the top 0.5\%
- Estimate $\widehat{\alpha}=\log \left[\left(1-F\left(y_{1}\right)\right) /\left(1-F\left(y_{2}\right)\right)\right] / \log \left(y_{2} / y_{1}\right)$ using Pareto CDF



## Atkinson and Piketty (2010)

- If income Pareto, top $p$ fractile income share is $S(p) \propto p^{1-1 / \alpha}$
- Estimate $\widehat{\alpha}=\left(1-\frac{\log [S(q) / S(p)]}{\log (q / p)}\right)^{-1}$ using top income shares of Piketty and Saez (2003)



## de Vries and Toda (2021)

- de Vries and Toda (2021) apply Hill (1975) estimator to micro data (Luxembourg Income Study, which is CPS for U.S.)
- Estimates from survey data biased upwards, suggesting low response/underreporting by rich



## Concluding remarks

- First systematic estimates of capital and labor income Pareto exponents in U.S., 1916-2019
- Post-1985, exponents stable at capital $\alpha \approx 1.2$, labor $\alpha \approx 2.0$
- $\alpha$ lower than existing estimates (higher top tail inequality)
- No systematic trend post-1985, so rise in income inequality is inequality between rich and poor, not among rich


## References

Atkinson, A. B. and T. Piketty, eds. (2010): Top Incomes: A Global Perspective, New York, NY: Oxford University Press.
Axtell, R. L. (2001): "Zipf Distribution of U.S. Firm Sizes," Science, 293, 1818-1820.
Beare, B. K. and A. A. Toda (2020): "On the Emergence of a Power Law in the Distribution of COVID-19 Cases," Physica D: Nonlinear Phenomena, 412, 132649.

- (2022): "Determination of Pareto Exponents in Economic Models Driven by Markov Multiplicative Processes," Econometrica, forthcoming.
de Vries, T. and A. A. Toda (2021): "Capital and Labor Income Pareto Exponents across Time and Space," Review of Income and Wealth.
Feenberg, D. R. and J. M. Poterba (1993): "Income Inequality and the Incomes of Very High-Income Taxpayers: Evidence from Tax Returns," Tax Policy and the Economy, 7, 145-177.
Gabaix, X. (1999): "Zipf's Law for Cities: An Explanation," Quarterly Journal of Economics, 114, 739-767.
Gabaix, X. and R. Ibragimov (2011): "Rank-1/2: A Simple Way to Improve the OLS Estimation of Tail Exponents," Journal of Business \& Economic Statistics, 29, 24-39.
Gouin-Bonenfant, É. and A. A. Toda (2022): "Pareto Extrapolation: An Analytical Framework for Studying Tail Inequality," Quantitative Economics, forthcoming.
Hill, B. M. (1975): "A Simple General Approach to Inference about the Tail of a Distribution," Annals of Statistics, 3, 1163-1174.
Klass, O. S., O. Biham, M. Levy, O. Malcai, and S. Solomon (2006): "The Forbes 400 and the Pareto Wealth Distribution," Economics Letters, 90, 290-295.
Pareto, V. (1896): La Courbe de la Répartition de la Richesse, Lausanne: Imprimerie Ch. Viret-Genton.
Piketty, T. and E. Saez (2003): "Income Inequality in the United States, 1913-1998," Quarterly Journal of Economics, 118, 1-41.


## References

SaEz, E. (2001): "Using Elasticities to Derive Optimal Income Tax Rates," Review of Economic Studies, 68, 205-229.

Stigler, S. M. (1974): "Linear Functions of Order Statistics with Smooth Weight Functions," Annals of Statistics, 2, 676-693.
Toda, A. A. and K. Walsh (2015): "The Double Power Law in Consumption and Implications for Testing Euler Equations," Journal of Political Economy, 123, 1177-1200.

Toda, A. A. and Y. Wang (2021): "Efficient Minimum Distance Estimation of Pareto Exponent from Top Income Shares," Journal of Applied Econometrics, 36, 228-243.

Vermeulen, P. (2018): "How Fat is the Top Tail of the Wealth Distribution?" Review of Income and Wealth, 64, 357-387.

## Statistics of Income 2019



## Statistics of Income 1919

TABLE 2.-PERSONAL RETURNS-DISTRIBUTION BY INCOME GLASSES, FOR THE UNITED STATES; showing for each elass of income, the number of returns, net income, personal exernption, dividends, tax paid, and percentages.
[Income returned for the calendar year ended Dec, 31, 1919.]

| Income class, | Number ofreturns. | Net Income. | Exmptiona frum normal tax. |  |  | Normal tax. | Surtax. | Total tax. | Average amount of tax per individual. | Average rate of tax per cent. |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  | Pereonal exemptlon. | Dividends. | Interest on Government obligations. 1 |  |  |  |  |  |
| \$1,000 to $\$ 2,000$ | 539, 946 | * $229,331,321$ | \$1,057, 674, 000 | \$30, 0000,938 |  |  |  |  |  |  |
| \$1,000 to 82,0000 | 1,388,820 | 1,999,781, 1800 | (1,352, 960, 2700 | 13, ${ }^{1379} 9^{\prime} 107$ | 60,422 | \$24,690,200 |  | \$24,696,200 | \$17.78 | 1. 13 |
| $82,000 ~ t o ~$ <br> 82,000 <br> 8,0001 <br> 8.000 | 1 1, 02402,269 | 1, 207, 889,326 | 1,200,453,500 | 47,771,042 |  |  |  |  |  |  |
| \$3,000 to \$4,000 | 1, $27.0,785$ | 2, $90.743,612$ | $1,86 \%, 881,109$ $61,970,509$ | 30,704,159 | 80, 239 | 28,257,861 |  | 28,257, 861 | 2.16 | 1.60 |
| $\$ 3,000$ to $84,009$. $\$ 4,000$ to 80,000 | 715,029 8,909 | $2,462,185,504$ $30,937,500$ | 1,476,25, 40.400 | $53,715,381$ $33,191,326$ | 200,523 | 37,658, 889 |  | 37,658, 780 | 2.67 | 1.53 |
| \$4,000 to 35,000 | 129, 195 | 1,920,396,985 | 807, 802,600 | 72, 322,462 | [ $40,40 \cdot 4$ | 38,200,6\%7 |  | 38,250,067 | 39.13 | 1.90 |
| \$5,000 to \$0,000. | 107,005 | 913, 242, 237 | 347, 919,700 | 67, 0000,834 | 3,092, 505 | 20,030,659 | \$503,926 | 20, 387,885 | 124.77 | 2.28 |
| \$ 87,000 to 87,000 | 109,644 | 704, 23, $3 \times 2$ | 225, 877,000 | $71,044,145$ | 3, 0667, 587 | 17,210,561 | 2,005,798 | 19,217,369 | 175.29 | 2.73 |
| 87,000 to 83,000 | 73,719 | 549,005, 475 | 150, 724,460 | 65, 148, 884 | $2,870,648$ | 15,515,478 | 2, 777, 868 | 18,293, 100 | 248.15 | 3.33 |
| 88,000 to 8,000 | 50,468 | 427,391,591 | 102, 799,600 | 60,061,704 | $2,446,223$ | 13, 688,701 | 3, 171, 449 | 16, 860,150 | 333.26 | 3.94 |
| \$0,000 to $810,000$. | 37,967 | 360, 2644,588 | 76,802, 600 | $56,962,70$ | ${ }^{2}$, 2898,801 | 12,747, 816 | ${ }_{3}^{3,581,610}$ | 14, 329.416 | ${ }^{4330.09}$ | 4.53 |
| \$10,000 to 811,000 | 28,409 | 208,741,306 | $57,115,600$ | 31,548,012 | 1, $1,790,9071$ | 11, 1154,245 |  | 14, ${ }^{14}$ | 519.63 | 4.96 |
| \$12,000 to \$13,000 | 22,841 | 262, $247,04,302$ | 45,680,400 | $48,218,235$ $44,089,629$ | 1, $1,711,3007$ | $10,344,370$ $8,512,024$ | $3,857,490$ $3,940,298$ | 13, $13,452,232$ | 721. 19 | 5.81 |
| \$ $\$ 13,000$ to $\$ 14,000$ | 15,248 | 205, 899,632 | 30,323,200 | 40, 977,190 | $1,209,502$ | 8, 998 , 003 | 4,017, 757 | 12,716, 36e | 833.97 | 6.18 |
| \$14,000 to 3 \$5,000 | 12,841 | 180, 215, 957 | 25,310,000 | 39, 885,022 | 1, 160, 745 | 7,984,373 | 4,100, 174 | 12,084,547 | 941.09 | 6.48 |
| \$15,000 to $320,000$. | 42, 028 | 724, 388,160 | 82.02t,000 | 16e, 3506,875 | 4, 896,6 6̄̃ | 32,286;924 | 2, 21918,502 | 33, 205,425 | - $1,265.95$ | 7.34 |
| \$20,000 to 525,000 | 22,665 | 504, 458, 801 | 43,590,200 | $132,167,485$ | 3,640,900 | 23,488,480 | 30,879,536 | 44,363,016 | 1,962.53 | 8. 79 |
| \$25,000 to S30,000. | 13,769 | 376, 457,979 | 26,438, 800 | $100,083,845$ | 2,723,665 | 17,675, 337 | 20,271,767 | 37,947,304 | 2,756.00 | 10.05 |
| \$ $\$ 30,000$ to $\$ 40,000$ | 15,410 | 530, 754,145 | 29,000,400 | 108,692, 831 | - | 25,093,240 | 38,081, 834 | \% 8 \%, 175, 124 | 4,090.62 | 11.90 |
| \$ 800,000 to 8800,000 | 8, | 370, 152,511 | $15,620,200$ $9,699,400$ |  | 1,920, 291 | 17, $13,764,138$ | 34, 364,546 | 48,629, 371 | 9,328.48 | 14.44 17.68 |
| \$ $\$ 00,1000$ to 870,000 | 3,193 | 200, 515,321 | 3,892,000 | 74, 81, 24.5 | 1,762, 463 | 9,818, 475 | 30, 320,310 | 40, 144,785 | 12,560. 85 | 19.44 |
| 870,000 to 880,000 | 2,237 | 167, 1052,648 | 4,377, 400 | 64,433,929 | 1,345, 842 | 7,8iti, 732 | 28,644,888 | $3 \mathrm{3}, 500,609$ | 16, 380.79 | ${ }^{21.85}$ |
| \$80,000 to 890,000 | 1,561 | $18,2,629,947$ | 2, 8177,800 | 49,816, 229 | 1, 135, 061 | C, 439, 7077 | 3, 313,745 | 32,040, 452 | 20, 818.71 | 24.55 |
| \$89,000 to $8100,000$. | 1,113 | $105,530,889$ $358,729,983$ | $2,412,260$ $5,323,200$ | - $31,354,249$ | 3, 757,971 | - $51,401,4685$ | 23,414,344 | 118, 205,305 | 25,618.51 | 27.02 33.12 |
| \$150,000 in \$200,000 | 1,682 | 187,816,010 | $1,940,209$ | $82,854,462$ | $2,244,253$ | 8,841, 502 | 68,085,248 | 76,972, 750 | 70, 487.87 | 40.98 |
| \$200,000 to \$250,000 | 522 | 115, 428,091 | 185, 600 | 18, 356,471 | 1, 8123,054 | 5,082,396 | 47,071, 376 | $52,754,132$ | 101,061. 56 | 4.71 |
| \$250,000 to $\$ 3000,000$ | 250 | 67,94, 6143 | 440,400 | 24, $40,485,908$ | 1, 804.400 | $3,255,588$ $4,615,053$ | 30, $214,89,076$ | 53, $50,824,129$ | 178, 3130.28 | 19.14 |

## Top relative income shares

- If income Pareto, top $p$ fractile income share is $S(p) \propto p^{1-1 / \alpha}$
- Hence top relative income share $S(p) / S(q)=(p / q)^{1-1 / \alpha}$ depends only on $\alpha$ and top relative fractile $p / q$



## Robustness to sample size

- We use $n=A-J$ for sample size
- Using $n=A-M$ has no material impact



## Robustness to choice of income groups

- We use top $1 \%$ for estimation



## Size distribution of COVID cases across U.S. counties

Physica D 412 (2020) 132649


