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References

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

Ji Hyung Lee¹ Yuya Sasaki² Alexis Akira Toda³ Yulong Wang⁴

¹University of Illinois, Urbana-Champaign

²Vanderbilt University

³University of California San Diego

⁴Syracuse University

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Pareto tail								

- Random variable X > 0 has Pareto upper tail if $P(X > x) \sim x^{-\alpha}$ for large x, where α : Pareto exponent
- Discovered by Pareto (1896) for income, but holds for other variables:
 - city size (Gabaix, 1999), $lpha \sim$ 1,
 - firm size (Axtell, 2001), $lpha \sim$ 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), $lpha \sim$ 1.5, etc.

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Importance of income and wealth Pareto exponent

• Convenient descriptive statistic for top tail inequality (small $\alpha \implies$ high inequality)

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Importance of income and wealth Pareto exponent

- Convenient descriptive statistic for top tail inequality (small $\alpha \implies$ 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

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 - $g \in [0,1]$: 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 α in economic models and Gouin-Bonenfant and Toda (2022) for numerically solving models)

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Accurately estimating income α 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)

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 - 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

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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

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- 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

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What we find

- Based on α , sample period can be divided into three sub periods, pre-1940, 1940–1985, and post-1985
- Post-1985, capital lpha pprox 1.2, labor lpha pprox 2.0
- α 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

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General framework

- Income $\{Y_i\}_{i=1}^n$, 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 {n_k, S_{n_k}}^K_{k=1}, where K is number of income groups and n₁ < n₂ < ··· < n_K ≤ n

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Example: U.S. 2019 tax returns data

1	ncome group	Adjusted gro	ss income (AGI)	Salaries	and wages
k	AGI threshold	# returns	Total income	# returns	Total income
18	\$1	9,866,880	24,439,988	6,672,531	23,927,191
17	\$5,000	9,925,940	74,584,857	7,622,306	58,927,624
16	\$10,000	11,087,737	138,230,399	8,277,447	100,631,554
15	\$15,000	10,039,446	175,255,963	7,931,946	134,897,400
14	\$20,000	9,493,968	213,660,160	7,855,283	173,142,941
13	\$25,000	9,289,939	254,877,708	7,943,835	212,428,275
12	\$30,000	16,090,602	560,073,192	14,045,867	471,544,226
11	\$40,000	12,503,041	560,258,808	10,931,707	465,547,848
10	\$50,000	22,238,948	1,366,892,948	18,976,338	1,071,062,478
9	\$75,000	14,118,568	1,222,947,425	12,033,727	921,390,540
8	\$100,000	21,997,582	3,004,363,636	19,028,674	2,209,484,837
7	\$200,000	7,297,883	2,090,808,696	6,414,121	1,429,162,189
6	\$500,000	1,162,371	781,920,814	1,010,488	449,489,139
5	\$1,000,000	254,197	305,561,848	214,955	141,101,999
4	\$1,500,000	103,075	176,961,208	85,285	72,754,006
3	\$2,000,000	143,514	425,088,995	117,168	145,270,762
2	\$5,000,000	34,738	237,781,553	28,162	66,367,353
1	\$10,000,000	20,876	590,230,011	16,866	102,518,828
AI	l returns, total	157,796,807	11,966,873,976	129,775,754	8,273,071,046

Estimation Results

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References

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://www.irs.gov/statistics/

Data

soi-tax-stats-individual-income-tax-returns-complete-report-publication-1304-basic-tables-> SOI 2019

- Before 1993, only PDFs of scanned copies of SOI are available (https://www.irs.gov/statistics/soi-tax-stats-archive) SOI 1919
- 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

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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 > 25k



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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)

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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)



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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



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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) (R² = 0.989) post-1950 and use OLS estimates to construct J pre-1950



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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 robustness





Estimation

- We now have data on $\{(n_k, S_{n_k})\}_{k=1}^K$ and n, where $n_1 < n_2 < \cdots < n_K \le n$
- 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 $J:[0,1] \to \mathbb{R}$ a bounded and almost everywhere continuous function, the asymptotic behavior of weighted sums of order statistics

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

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is known (Stigler, 1974)

• Let $p_k = n_k/n$ be top fractile

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Estimation

- Basic idea (continued)
 - Letting $J(x) = \mathbbm{1}(1 p_{n_{k+1}} < x \le 1 p_{n_k})$, 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

$$s := \left(\frac{S_{n_2} - S_{n_1}}{S_{n_{L+1}} - S_{n_L}}, \dots, \frac{S_{n_L} - S_{n_{L-1}}}{S_{n_{L+1}} - S_{n_L}} \right),$$

asymptotic behavior depends only on Pareto exponent α if $\{Y_i\}$ have Pareto upper tail and $n_{L+1} \ll n$

• Can estimate α by minimizing quadratic distance of s from theoretical value implied by Pareto distribution

Estimation 00000

Is Pareto tail reasonable?

- If income has Pareto upper tail with exponent α , 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,..., 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 $\ll {\rm AGI}$ for top earners

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Choice of income groups

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

 robustness

Estimation



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AGI and capital income Pareto exponents

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



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No rise in top tail inequality post-1985

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



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References

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

Results

Top labor income shares, 1968





Top labor income shares, 1968



Joint distribution of AGI and labor income



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Joint distribution of AGI and labor income



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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)



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References

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)

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References

Maximum likelihood

For AGI, we observe income thresholds

$$\infty = t_0 > t_1 > \cdots > t_K > 0$$

• Can apply ML using conditional tail probability $P(Y \ge y \mid Y \ge t_L) = (y/t_L)^{-\alpha}$



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Feenberg and Poterba (1993)

- Feenberg and Poterba (1993) find two income thresholds $y_1 < y_2$ that bracket the top 0.5%
- Estimate $\hat{\alpha} = \log[(1 F(y_1))/(1 F(y_2))]/\log(y_2/y_1)$ using Pareto CDF



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Atkinson and Piketty (2010)

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



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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



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Results 00000000000000 Conclusion

References

Concluding remarks

- First systematic estimates of capital and labor income Pareto exponents in U.S., 1916–2019
- Post-1985, exponents stable at capital lpha pprox 1.2, labor lpha pprox 2.0
- α 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

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Statistics of Income 2019

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Table 1.4. All Returns: Sources of Income, Adjustments, and Tax Items, by Size of Adjusted Gross Income, Tax Year 2019 (Filing Year 2020)					
(All figures are estimates based on samples-money amounts are in thousands of dollars)					
 adjusted gross	Total inco	ime	Salaries and wages		
income	Number of returns	Amount	Number of returns	Amount	
	(3)	(4)	(5)	(6)	
All returns, total	157,197,473	12,111,799,488	129,775,754	8,273,071,046	
No adjusted gross income	1,528,166	-234,790,580	569,047	23,421,857	
\$1 under \$5,000	9,866,880	25,527,613	6,672,531	23,927,191	
\$5,000 under \$10,000	9,925,940	76,354,338	7,622,306	58,927,624	
\$10,000 under \$15,000	11,087,737	141,617,199	8,277,447	100,631,654	
\$15,000 under \$20,000	10,039,446	178,282,897	7,931,946	134,897,400	
\$20,000 under \$25,000	9,493,968	216,874,301	7,855,283	173,142,941	
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\$50,000 under \$75,000	22,238,948	1,382,388,717	18,976,338	1,071,062,478	
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\$100,000 under \$200,000	21,997,582	3,036,219,498	19,028,674	2,209,484,837	
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\$500,000 under \$1,000,000	1,162,371	795,166,354	1,010,488	449,489,139	
\$1,000,000 under \$1,500,000	254,197	310,038,042	214,955	141,101,999	
\$1,500,000 under \$2,000,000	103,075	179,090,416	85,285	72,754,006	

Statistics of Income 1919

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	TABLE 2.—PERSONA class	L RETU:	RNS—DIST the number of [1	RIBUTION 2 returns, net inc	BY INCOM ome, persona r the calendar y	E CLASS l exemption,	ES, FOR dividends, t	THE UNII ax paid, and	TED STATE percentages.	SS; showing	for each
	Income class,	Number of returns.	Net income.	Exemption.	ons from norms Dividends,	Interest on Government obligations. ¹	Normal tax.	Surtax.	Total tax.	Average amount of tax per individual.	Average rate of tax per cent.
	1,001 to 2,000 to 2,0	$\begin{array}{c} 536, 046\\ 1, 888, 826\\ 259, 219\\ 259, 219\\ 1, 040, 472\\ 277, 505\\ 277, 505\\ 277, 505\\ 492$	$\begin{array}{c} \textbf{4} \\ \textbf{4} \\ \textbf{5} \\ $	$\begin{array}{c} \mathbf{R}, 907, 574, 900\\ \mathbf{I}, 382, 868, 700\\ \mathbf{I}, 382, 868, 700\\ \mathbf{I}, 382, 868, 700\\ \mathbf{I}, 900, 575, 560\\ \mathbf{I}, 165, 774, 100\\ \mathbf{I}, 052, 77, 007\\ \mathbf{I}, 052, 700\\ \mathbf$		\$60,402 3,129 3,129 450,403 3,005,007 450,403 450,4	23, 557, 581 23, 557, 581 24, 557, 581 25, 555, 507 25, 508 25, 508 26, 508 2	\$ \$ \$ \$ \$ \$ \$ \$	\$24, 666, 200 26, 575, 685, 789 37, 685, 789 38, 585, 789 39, 587, 585, 789 30, 587, 585, 789 30, 587, 585, 789 30, 587, 585, 789 30, 587, 587, 587 30, 587, 597 30, 5	\$77.78 27.16 39.13 121.77 39.13 39.13 30.57 30.53 30.57 40.09 50.53 50.55 40.09 50.55 40.09 50.55 50.55 40.09 50.55 50.5	1.23 1.60 1.33 1.99 2.28 2.73 3.94 4.53 4.53 4.53 4.53 4.53 1.99 1.55 5.55 5.55 2.55 2.165

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Top relative income shares

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



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Size distribution of COVID cases across U.S. counties

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