# Measuring Income and Wealth at the Top Using Administrative and Survey Data

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#### **Motivation**

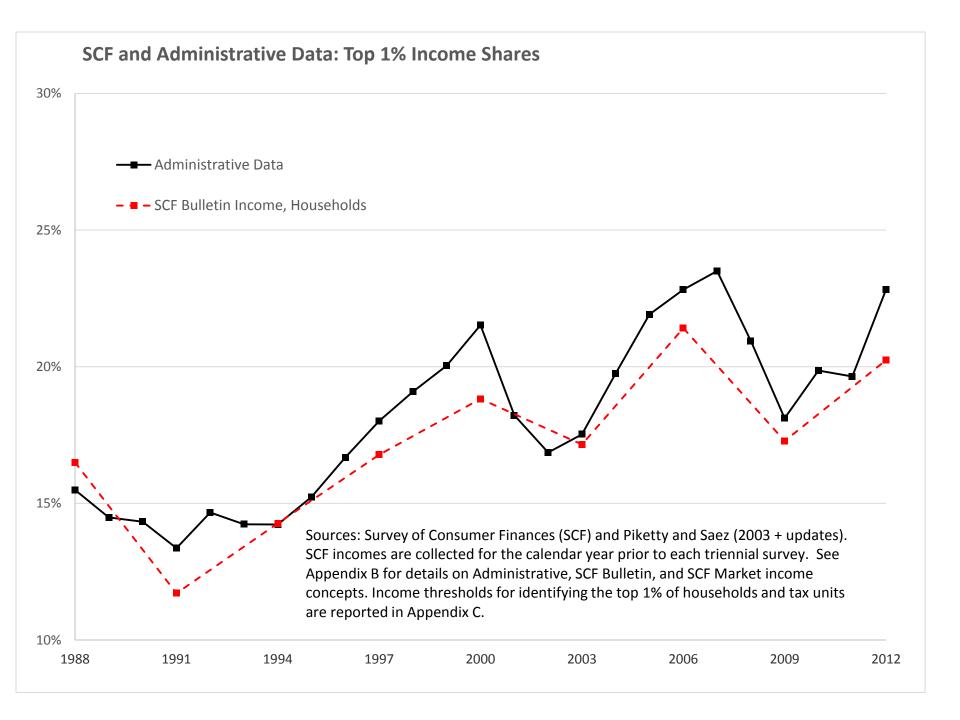
- Top income and wealth shares: high and rising
- Both normative and policy concerns
  - o Is the world becoming less fair? If so, why?
  - Has inequality contributed to macro instability and/or slowed economic growth?
- Not just academic debates, bestsellers!
  - Rajan (2010), Stiglitz (2012), Piketty (2014)

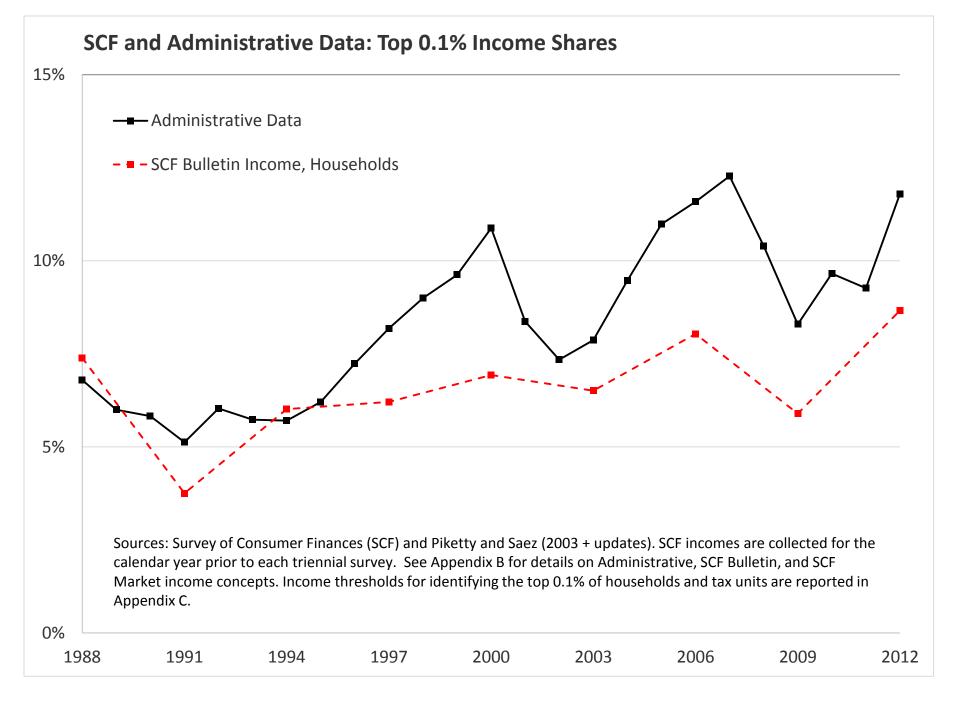
# **Goals for this Paper**

- U.S. top income and wealth shares are high and rising, but how high, and how fast?
- Widely-cited top shares estimates based on administrative income tax data diverge from Survey of Consumer Finances (SCF)
  - Piketty and Saez (2003, updated)
  - Saez and Zucman (2014)
- Primary goal is to understand why the two approaches diverge, and solve for biases

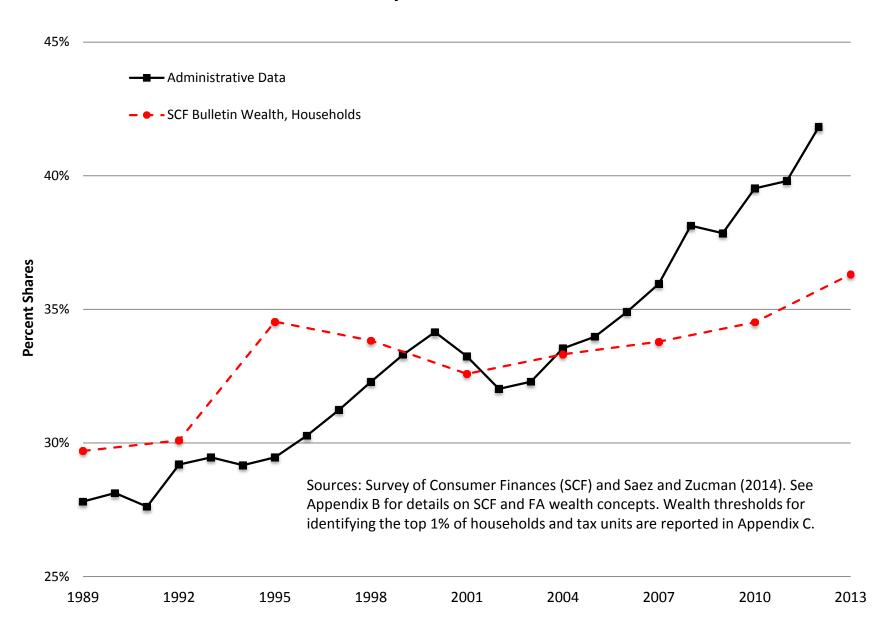
# Why Might Different Approach Matter?

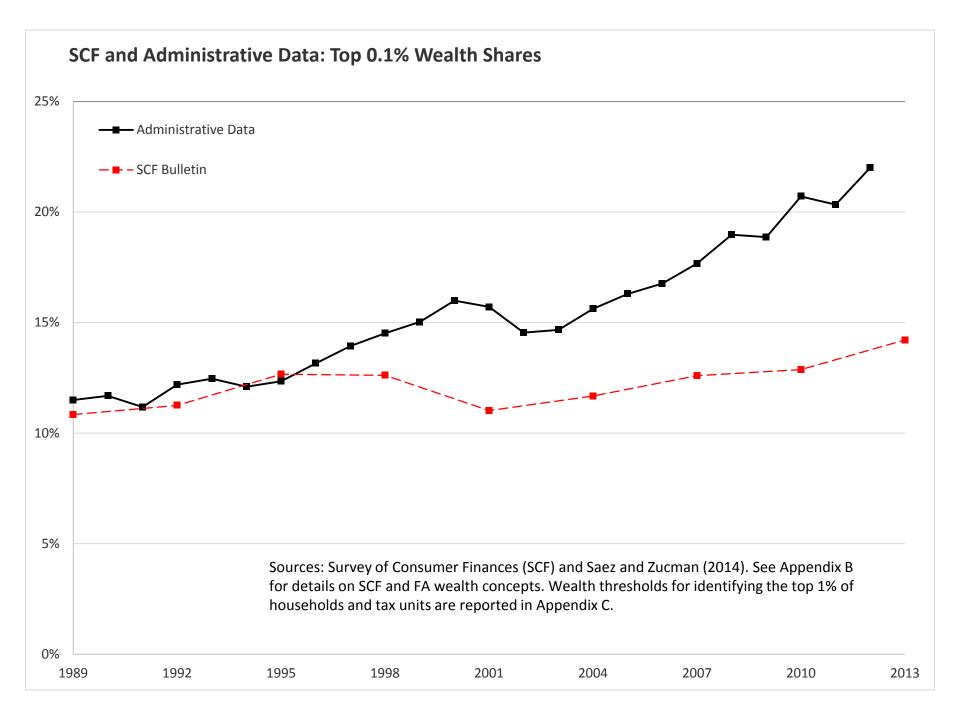
- SCF samples the population and validates representativeness with same administrative data
  - SCF top shares lower/growing more slowly
  - O Does this mean SCF not capturing top shares?
- Could also be problems with tax-based estimates
  - Unit of observation is not households
  - Income and wealth concepts limited by tax system
    - *Unmeasured* incomes and wealth are not distributed the same as measured components
  - Aggregate benchmarks matter, especially for wealth
  - Transitory income fluctuations distort very top shares





#### **SCF and Administrative Data Top 1% Wealth Shares**





### **Road Map for Paper**

- I. Introduce SCF sampling and validation
  - Participants and non-participants look the same
- II. Reconcile top 1%, 0.1% income shares
  - Tax units vs households, income concepts
- III. Reconcile top 1%, 0.1% wealth shares
  - Tax units vs households, wealth concepts
  - Benchmarking to wealth aggregates, Forbes 400
- IV. Show how SCF sampling goal of finding permanently wealthy affects very top shares

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- Introduce SCF sampling and validation
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#### III. Wealth Reconciliation

- Why do SCF and tax-based "Gross Capitalization" top wealth share estimates diverge?
  - Capitalized approach uses taxable SOI incomes for income-generating assets, imputations for rest
  - Capitalized calibrated to Financial Accounts (FA)
  - SCF and FA balance sheet concepts diverge
  - SCF and FA estimated aggregates diverge
  - Some implied capitalization factors problematic
- 160 million tax units versus 120 million families
- SCF (by rule) does not survey Forbes 400

# **Gross Capitalization (GC) Approach**

• Given taxable capital income type k=1,...9 along with estimates of wealth that do not generate taxable income, for family i

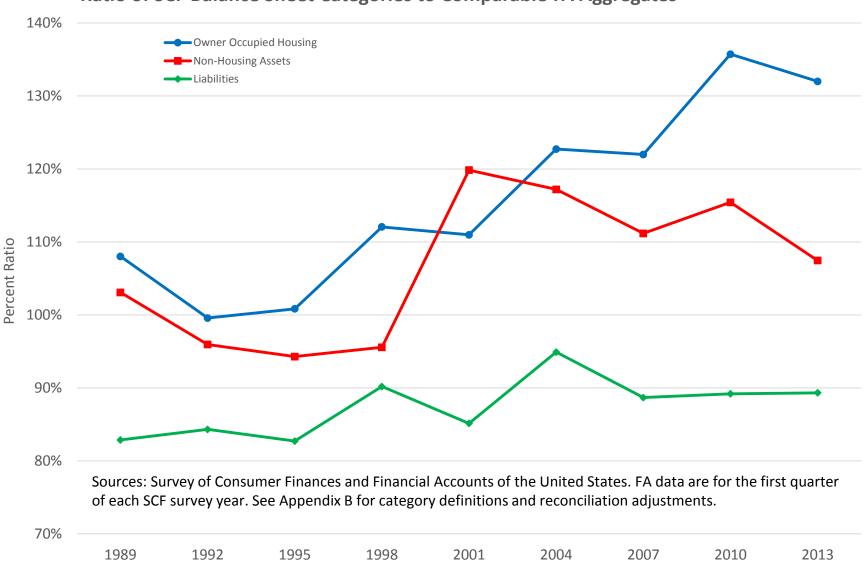
$$\widehat{wealth}_{i}^{GC} = \sum_{\forall k} \frac{SOI\ income_{i}^{k}}{ror^{k}} + nonfinancial_{i}$$

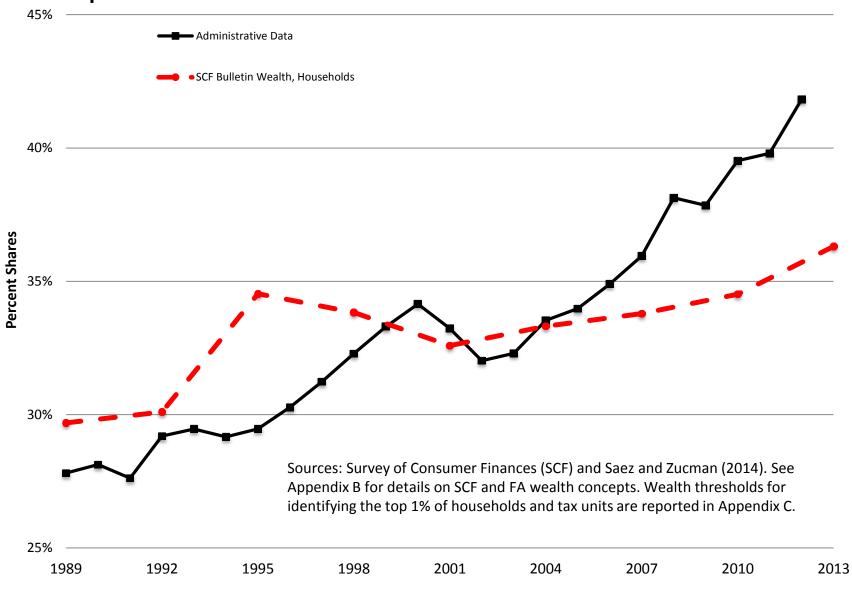
• In practice, Saez and Zucman (2014) compute *ror* for each asset *k* to calibrate to FA aggregates

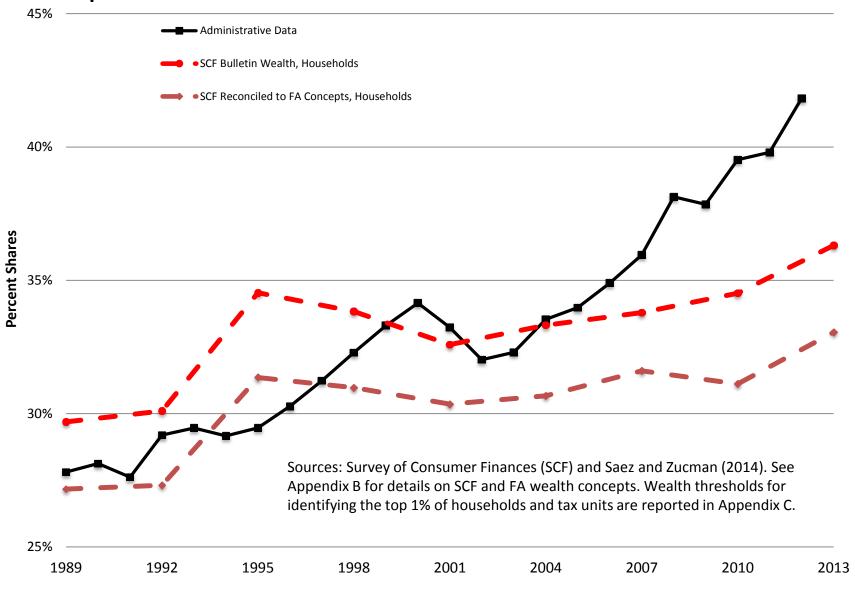
$$ror^{k} = \frac{\sum_{\forall i} SOI \ income_{i}^{k}}{FA \ asset^{k}}$$

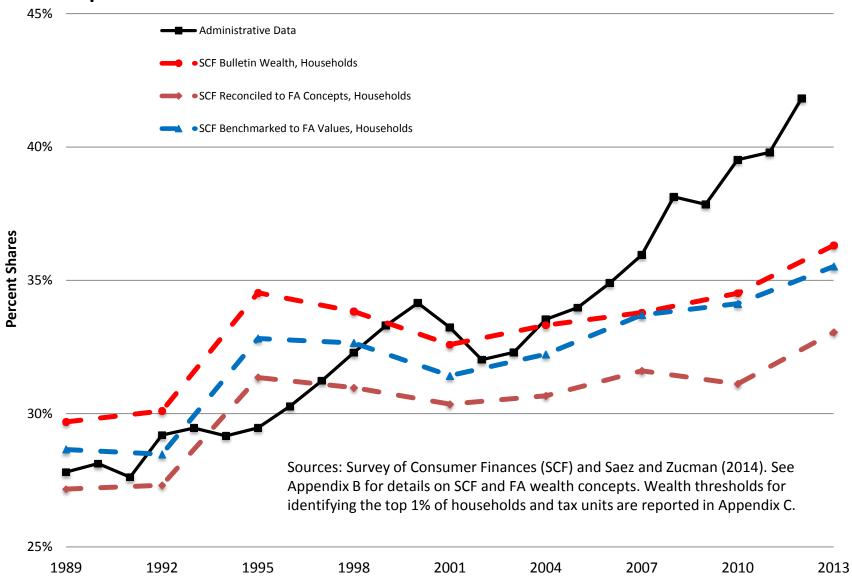
Table B.1. Reconciling SCF and FA Aggregates			
	2013Q1 (\$ Trillions)		
	SCF	FA	Gap
Published Household Net Worth	65.5	72.3	-6.8
- Less Identifiable Nonprofit Net Worth		2.1	
- Less Security Credit, miscellaneous assets and liabilities	1.0	1.1	
- Less Life Insurance	0.8	1.2	
+ Plus DB Pensions	10.9		
- Less Durables	2.4	4.9	
- Less Forbes400 Net Worth		2.0	
= Conceptually Equivalent Net Worth	72.2	61.0	11.2

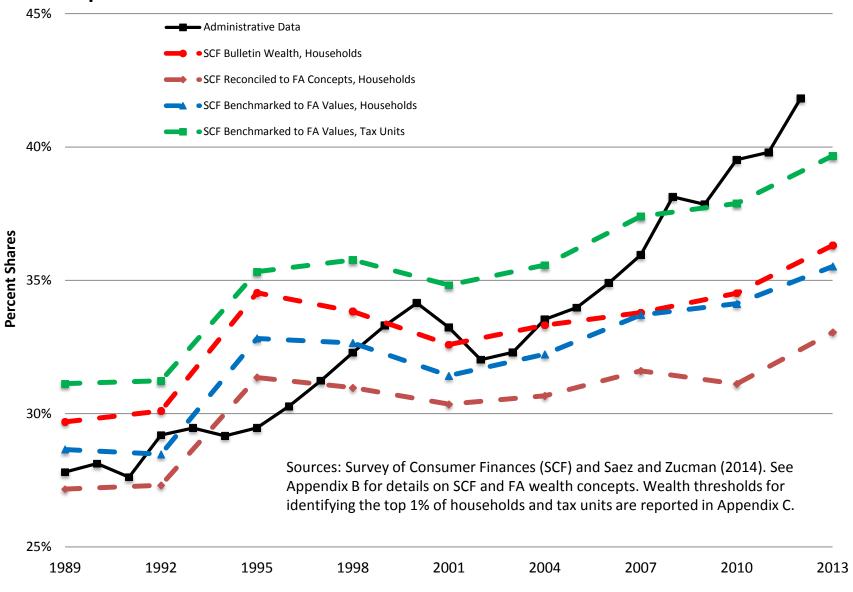
#### **Ratio of SCF Balance Sheet Categories to Comparable FA Aggregates**

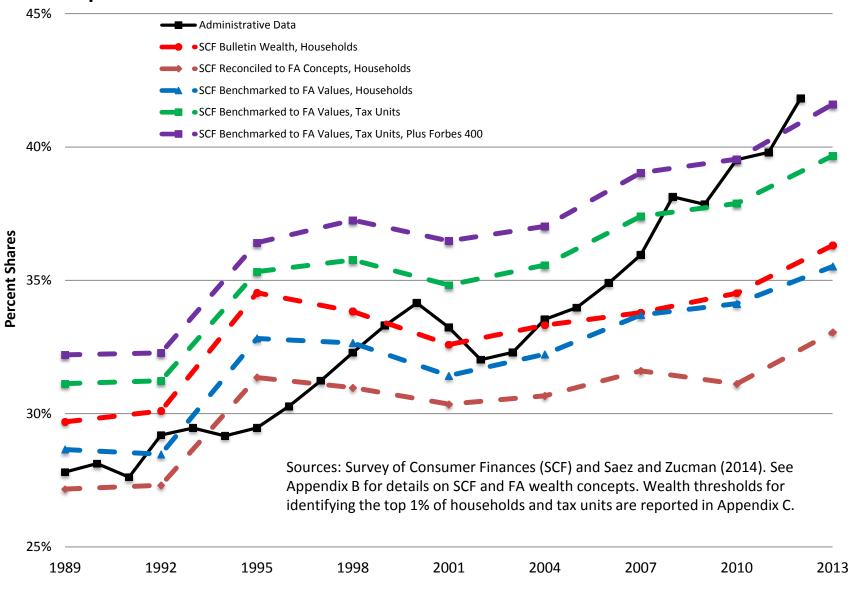


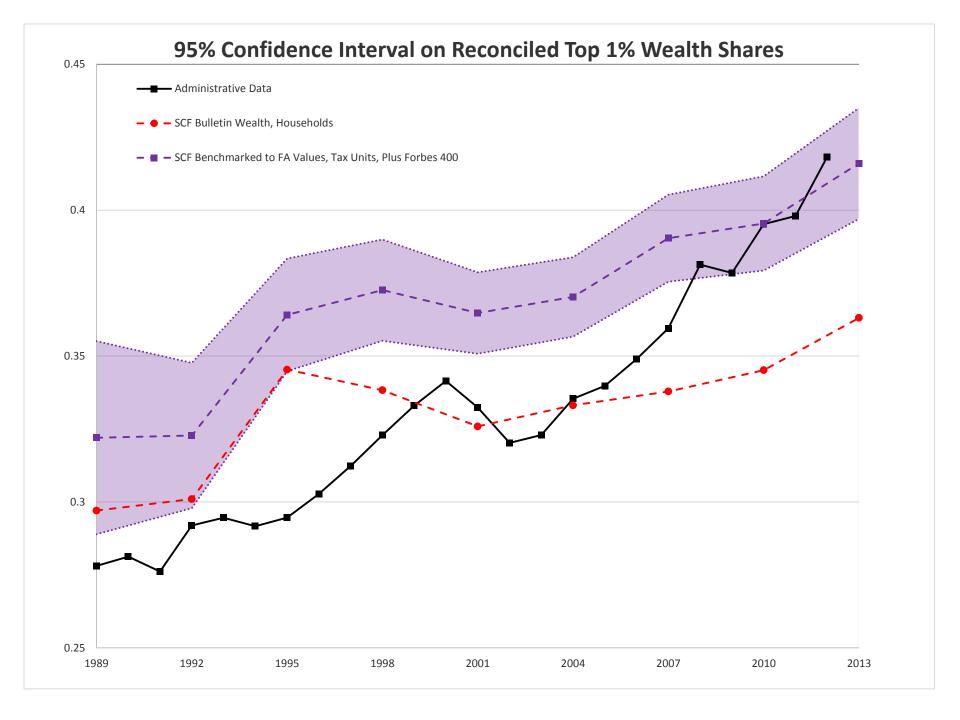


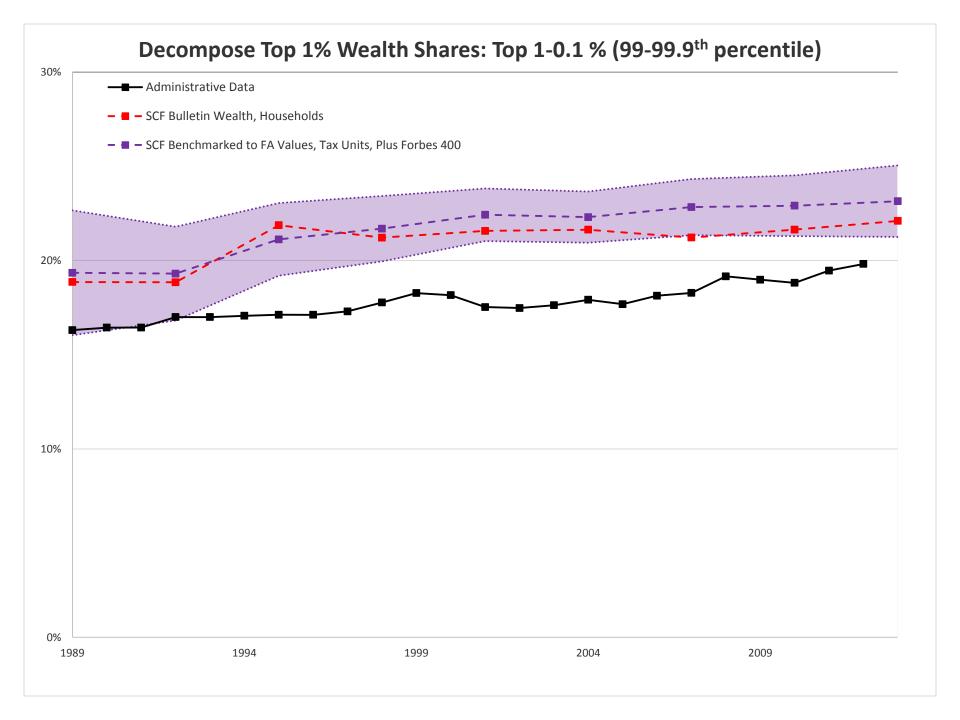


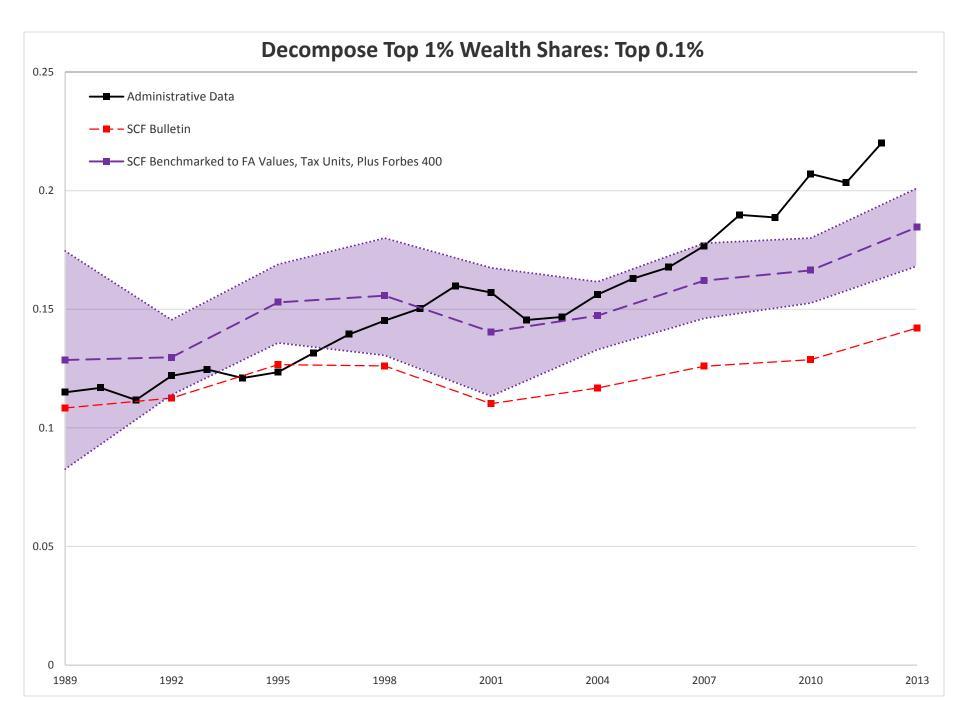






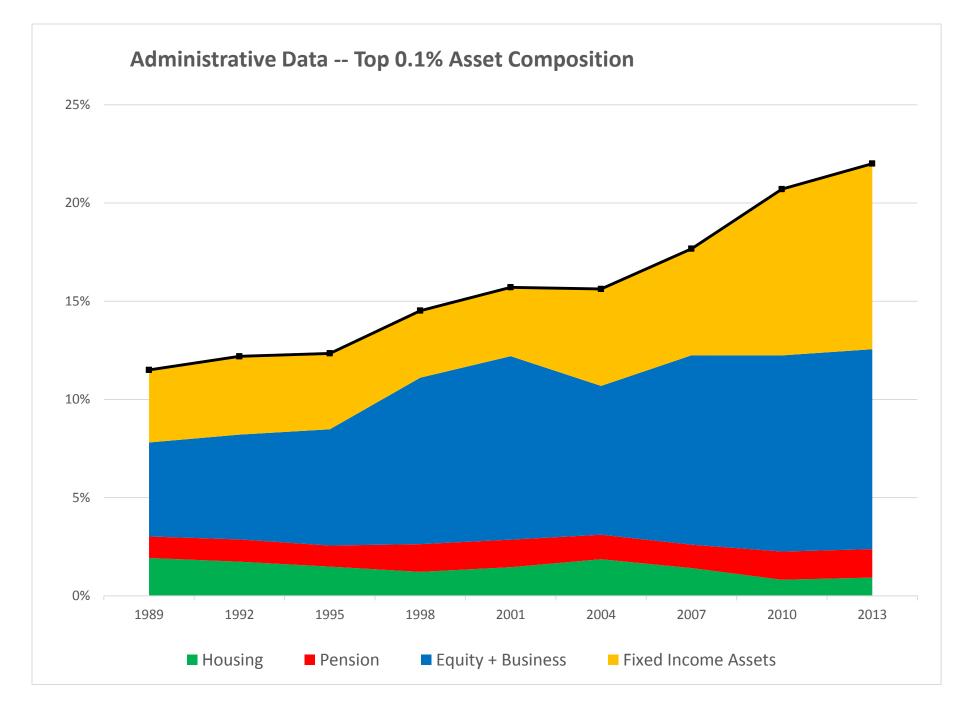


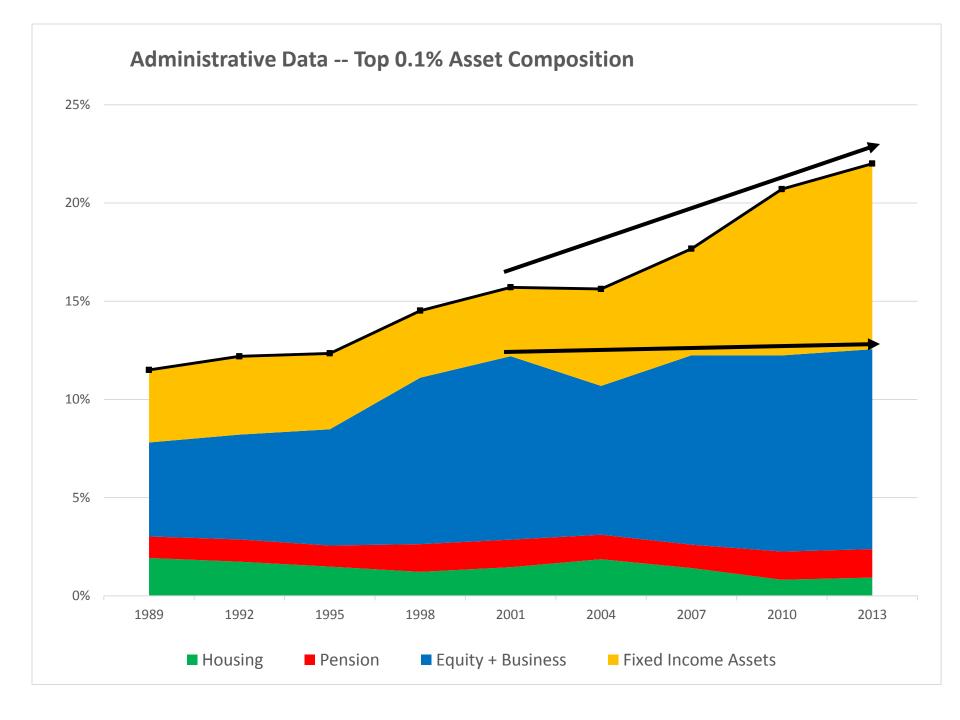




### Wealth Reconciliation at the Very Top

- Still, top 0.1 wealth share is greater in capitalized administrative tax data
- Look closer at asset composition and RoR
  - Fixed-income assets were 25%, now 45% of assets

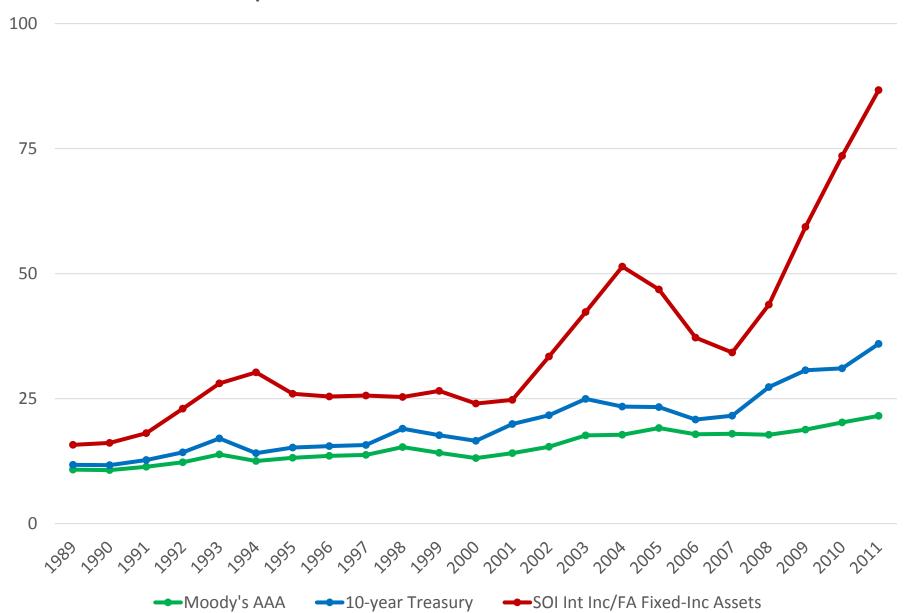




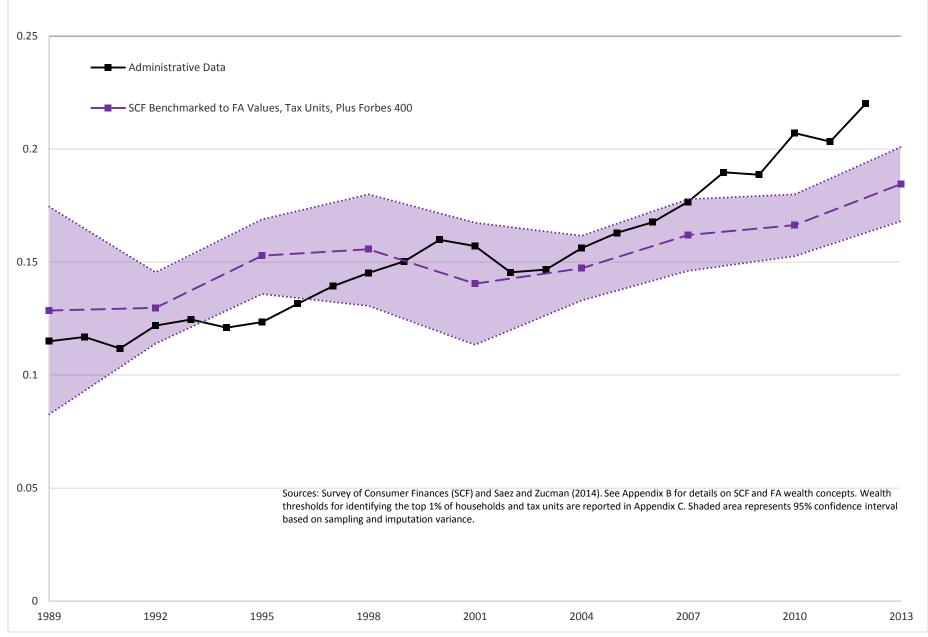
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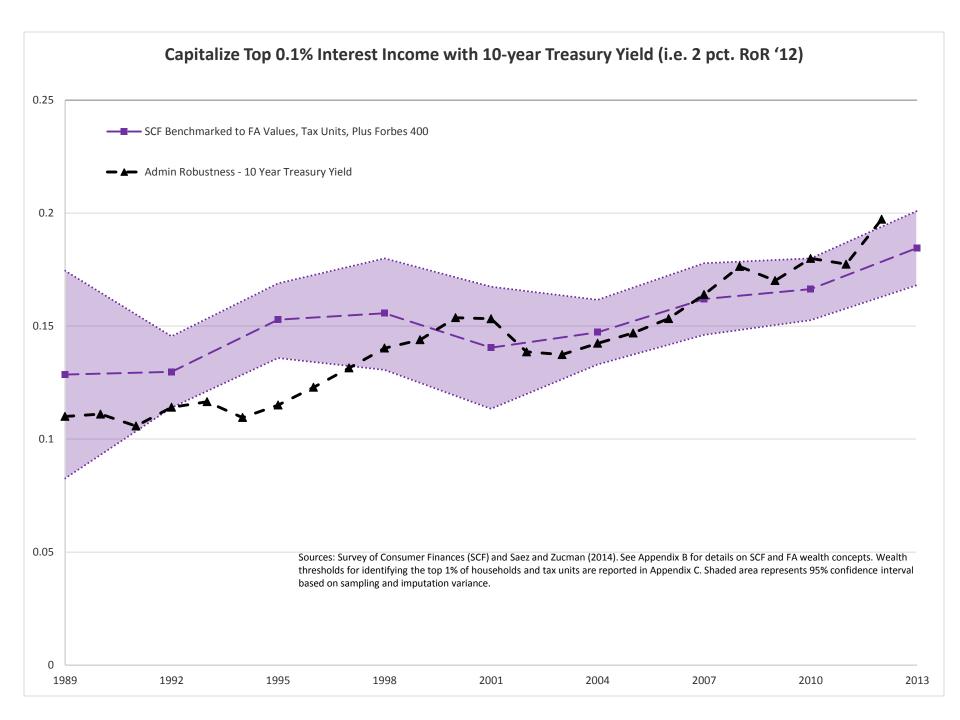
- Still, top 0.1 wealth share is greater in capitalized administrative tax data
- Look closer at asset composition and RoR
  - Fixed-income assets were 25%, now 45% of assets
  - Bonds ≈  $1/3^{rd}$ , deposit accounts are the other  $2/3^{rds}$ .
  - O Do the top 0.1 really hold savings deposit accounts?
- Rate of return on fixed-income = 1 pct. (for all)
  - → capitalization factor of 100x for interest income
  - Compare to market rates of return

#### **Capitalization Factors on Fixed-Income Assets**









### Wealth Reconciliation Takeaways

- As with income, top 1% levels in 2013 similar, but
   SCF trends generally flatter
  - All the difference is due to top 0.1 percent
- The same slower top 0.1 wealth shares, growth
  - When assume reasonable capitalization factor for interest income
  - Why? Allocate FA fixed-income assets to those that realize interest income
  - Small balance deposits growing; 1099-INT if ≥ \$10
  - Further, the SZ bond series is pre-FA revision!

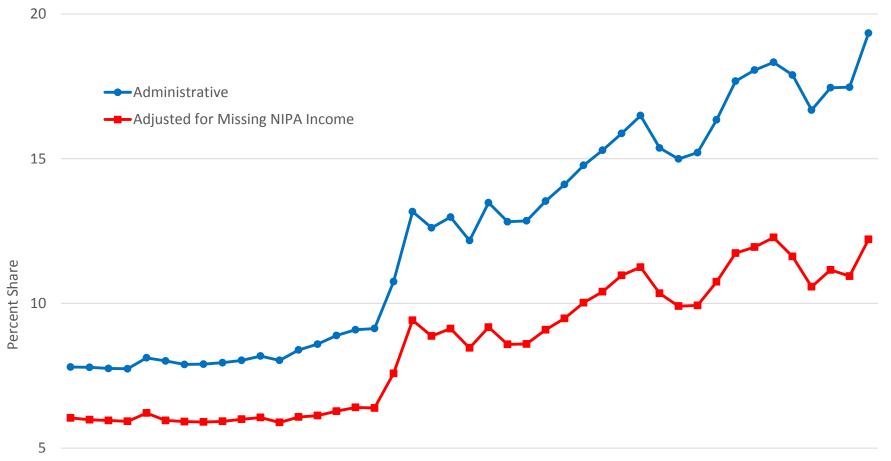
#### **Conclusions**

- Estimates of top income and wealth shares from SCF can be reconciled with estimates derived directly from administrative tax data
- SCF suggests that administrative-based top share estimates too high and rising too fast
- Reconciliations offer direction for future work, as broader income and wealth measures are likely to further reduce estimated top shares

# If Time: Expanded Income Concept

- Cannot distribute all of NIPA personal income, but can at least bracket top income shares
- Assume that missing income in every year, starting in 1970, is allocated per tax unit
  - Top 1 percent gets only 1% of the missing income
- Top 1% income levels and growth much more muted, and tax unit adjustment would add
- Extreme assumption, but brackets truth: missing incomes are transfers, non-wage compensation, retirement saving

Figure 10. Effect of Allocating Missing Personal Income on Top 1% Income Shares



Sources: Bureau of Economic Analysis and Piketty and Saez (2003 + updates). Adjustment assumes all missing NIPA income (government transfers, unreported income, retirement saving, employer-provided health) are allocated to top share group in proportion to numbers of units, not in relation to other incomes. See Appendix B for a discussion of the mismatch between NIPA and administrative data concepts.

1970 1975 1980 1985 1990 1995 2000 2005 2010

# Thanks!

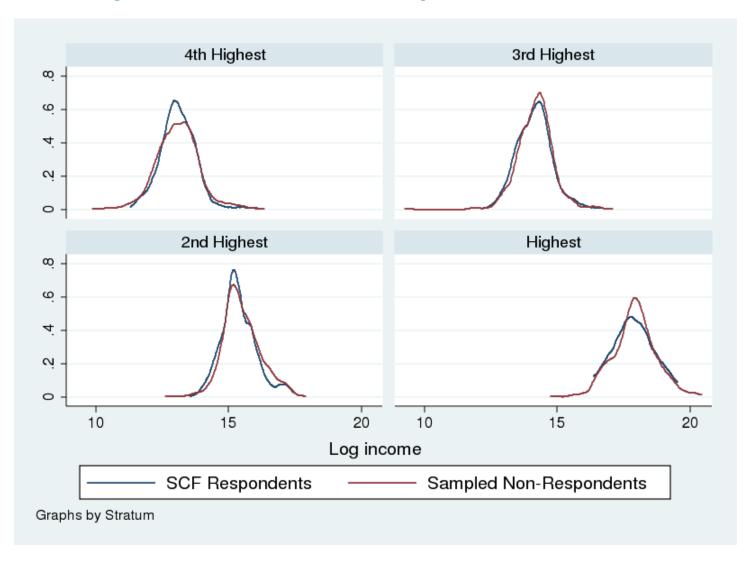
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# **Additional Slides**

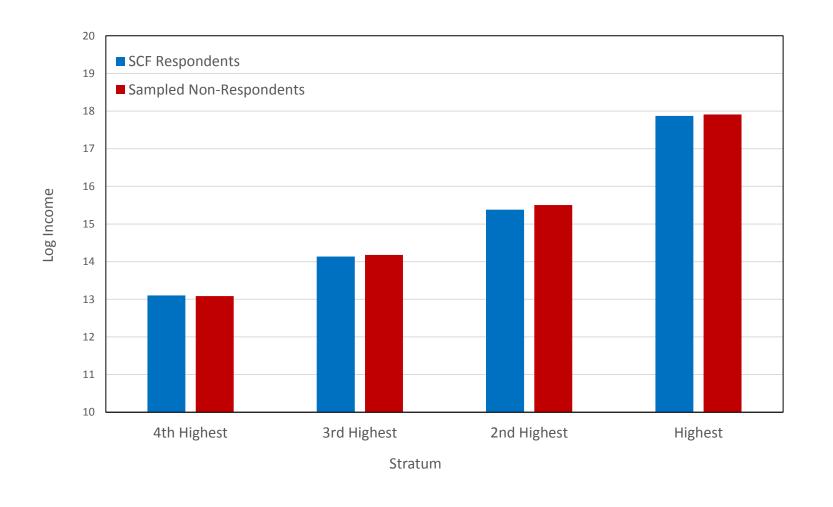
### I. SCF Sampling and Representativeness

- Non-participation among wealthy families in traditional household surveys is a problem
  - Indirect evidence: don't see large incomes, wealth
  - Direct evidence: Sabelhaus et al. (2015)
- SCF solves this by targeting a certain number of cases from very thin strata near the top
  - Accept low response rates, weight accordingly
- Things can go wrong: participants within a given stratum don't look like non-participants

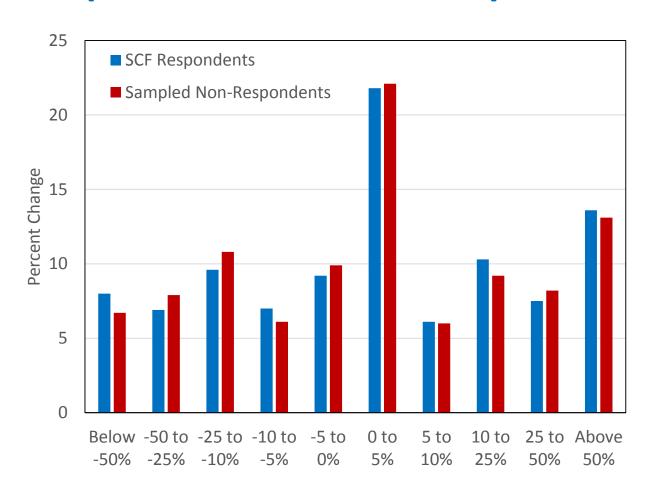
# SCF Respondents are Similar to Non-Respondents by Income Within Top Share Strata



# Mean Incomes within Top Share Strata are Similar for Respondents and Non-Respondents



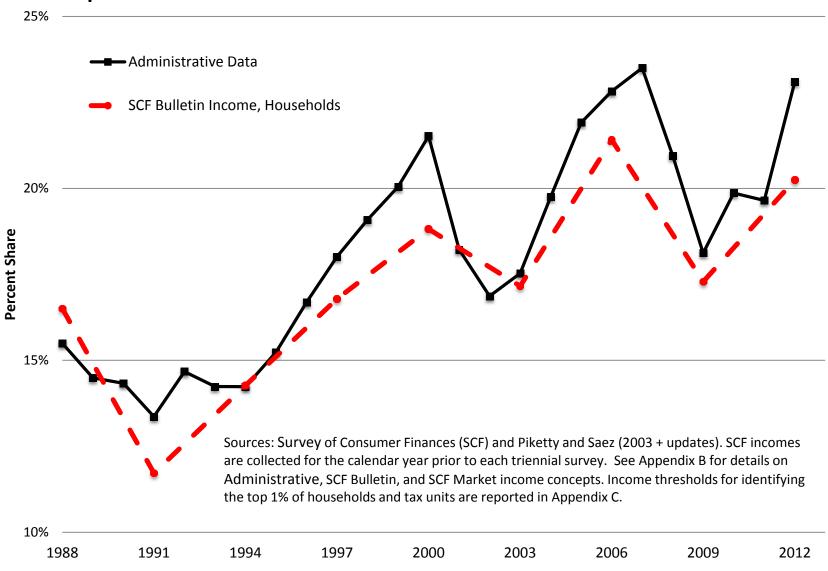
# Pre-Survey Income Volatility is Similar for Respondents and Non-Respondents



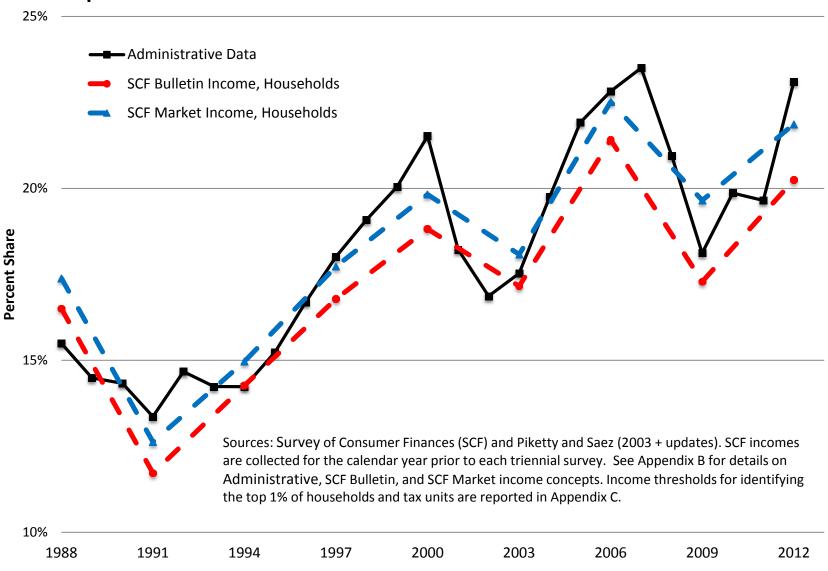
#### II. Income Reconciliation

- Comparisons between Piketty-Saez and SCF income shares is "apples-to-oranges"
- First, adjust income concept
  - Strip out "non-market" incomes from SCF
- Second, adjust for tax units vs households
  - In 2013, 160 mill tax units, 120 mill households
  - In adjusted 1% series, just count 1.6 mill from top

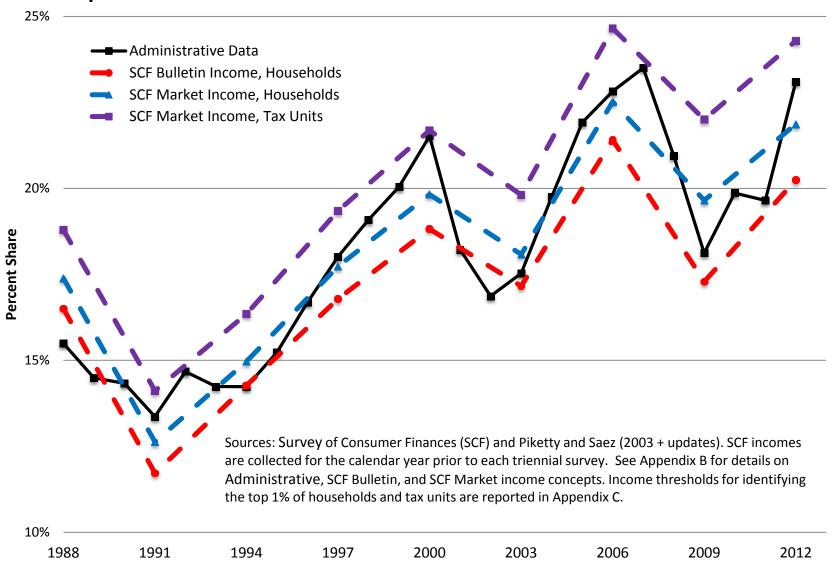
# Reconciling Survey of Consumer Finances (SCF) and Administrative Data Top 1% Total Income Shares

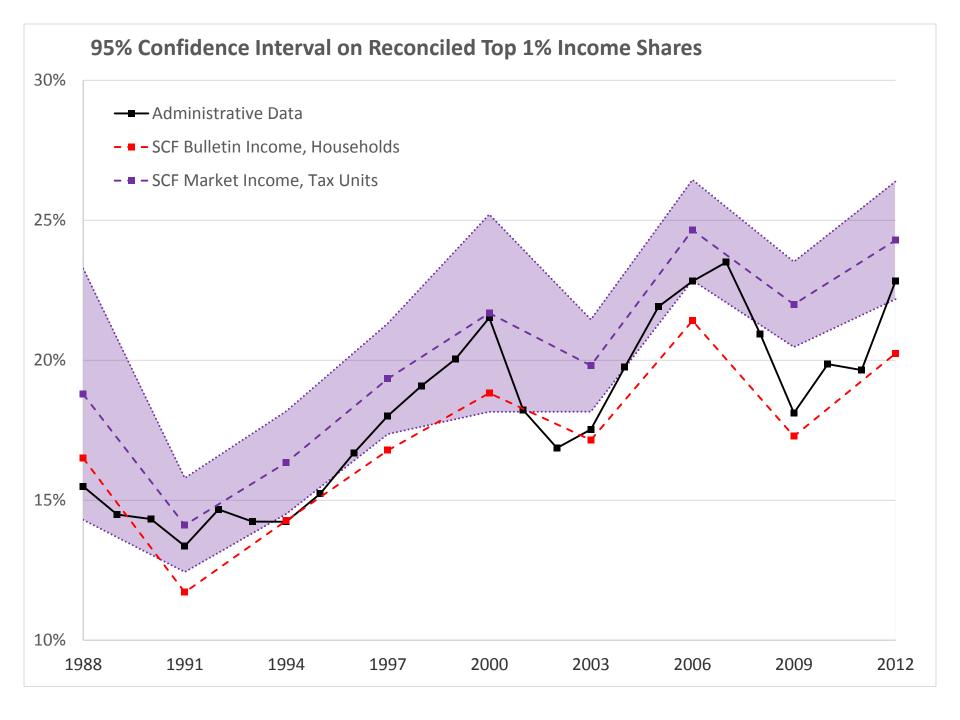


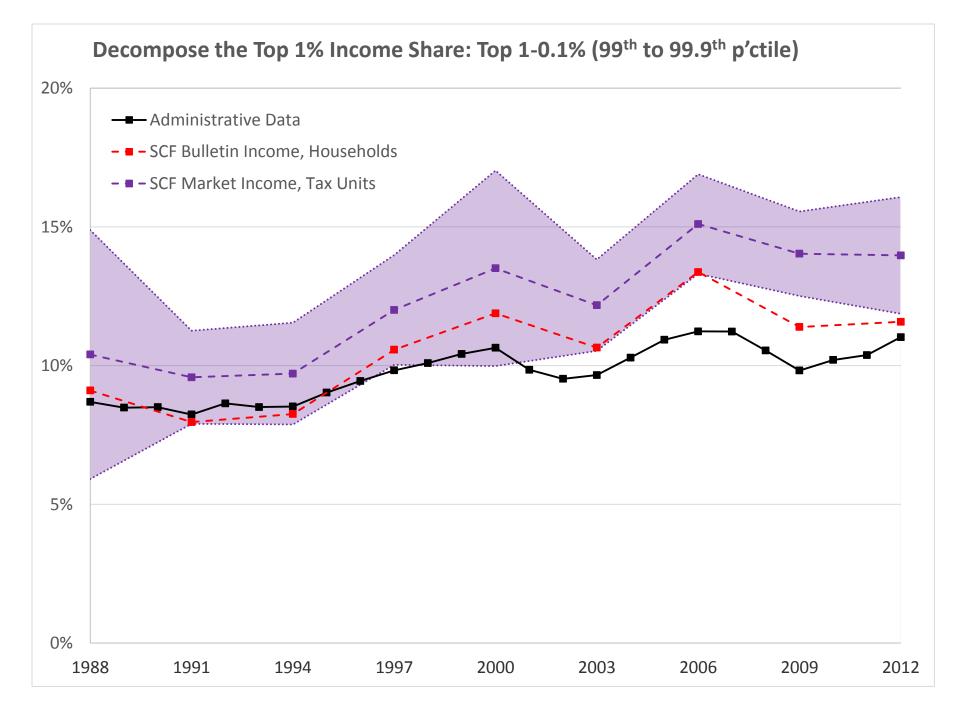
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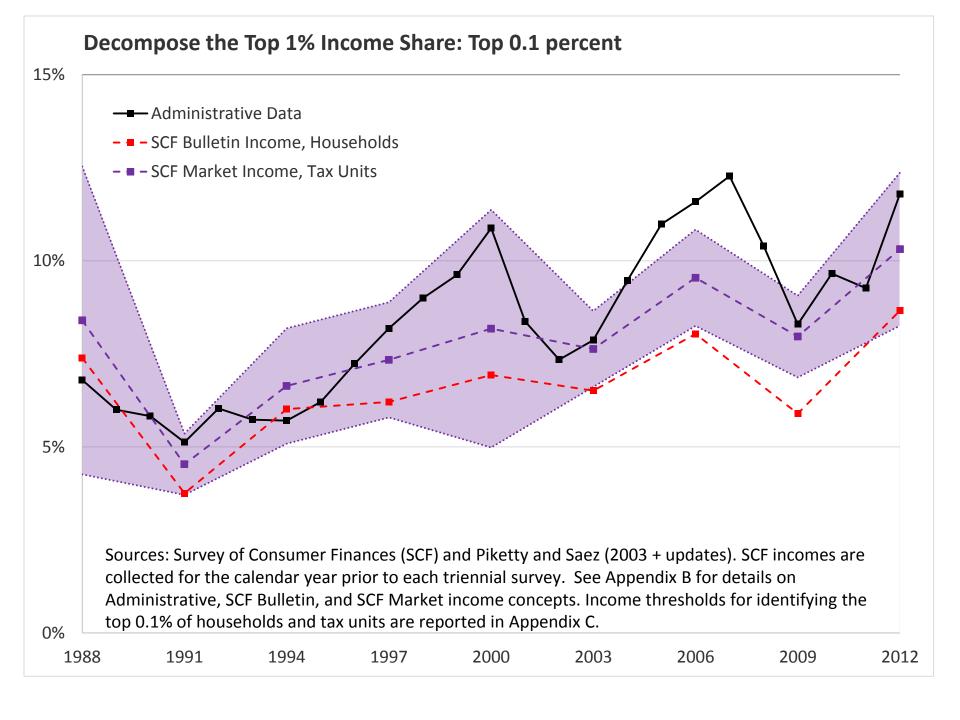


## Reconciling Survey of Consumer Finances (SCF) and Administrative Data Top 1% Total Income Shares



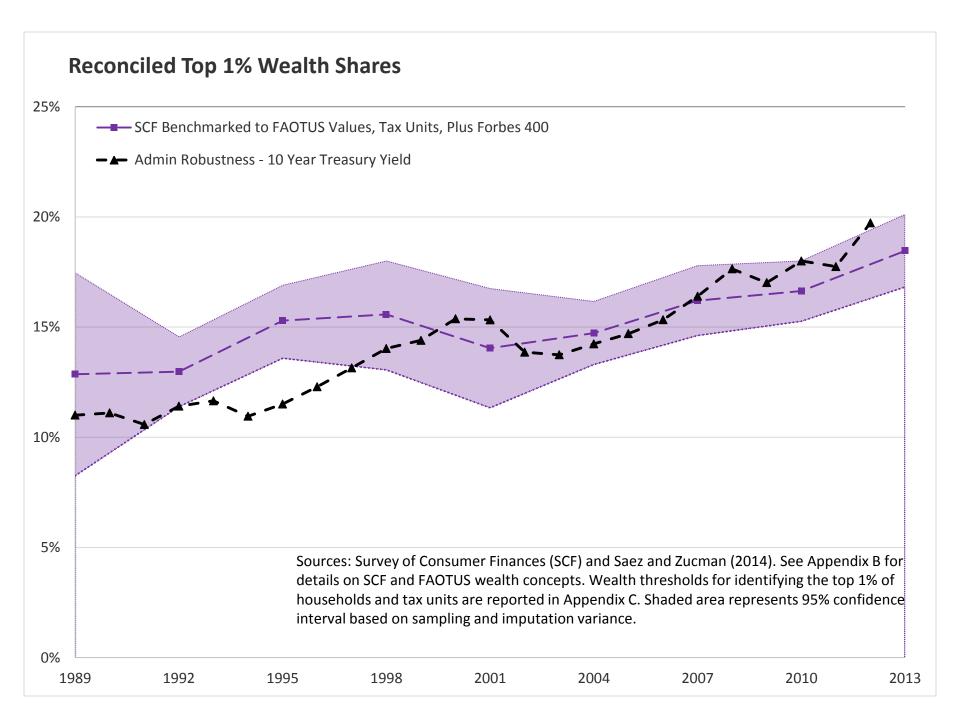






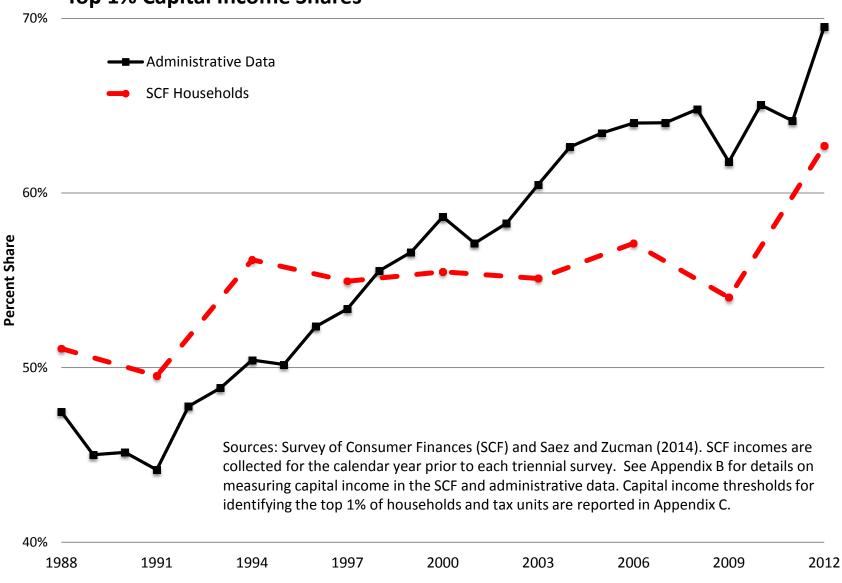
#### **Income Reconciliation Takeaways**

- Income concepts and tax unit adjustments reconcile top 1% share, levels and trends
- Top 0.1%, Top 1-0.1%: growth is the same in recent years
  - SCF does not "spike" as much for top 0.1%
- Levels: Top 0.1% slightly higher, Top 1-0.1% slightly lower in level in recent years

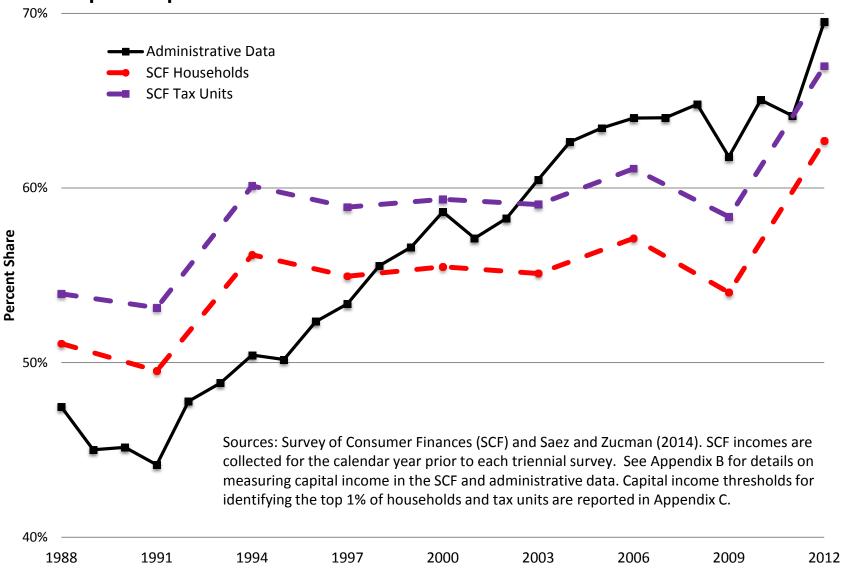


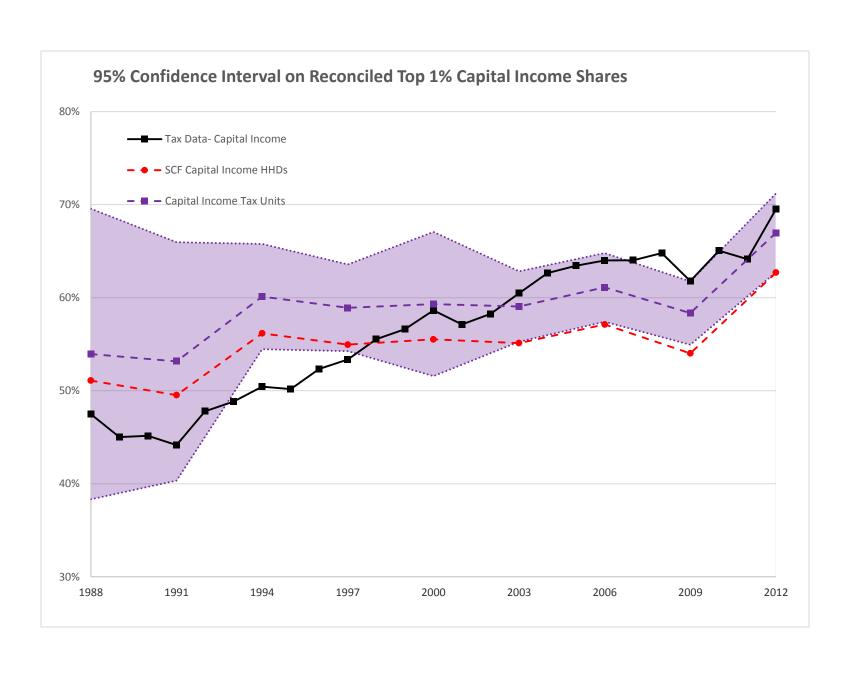


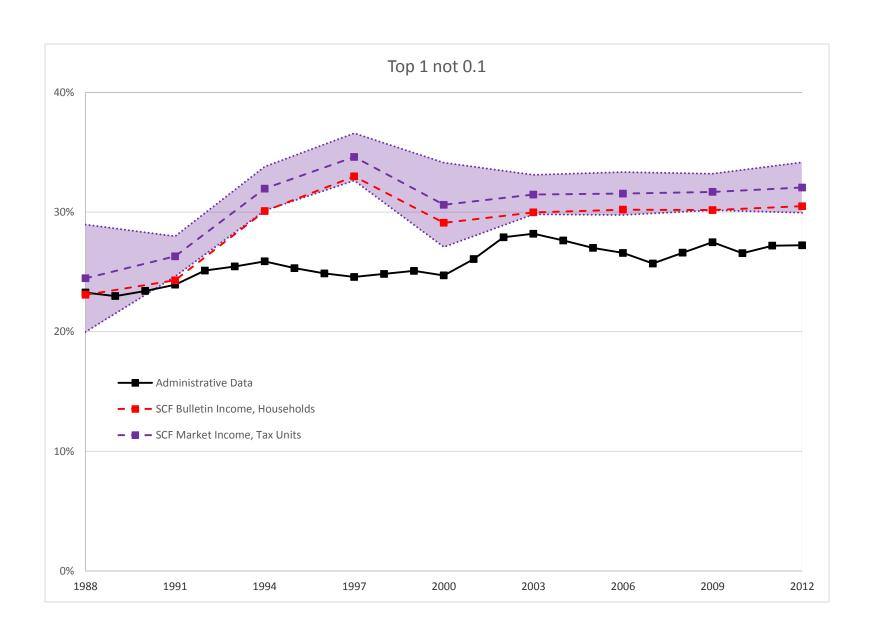
# Reconciling Survey of Consumer Finances (SCF) and Administrative Data Top 1% Capital Income Shares

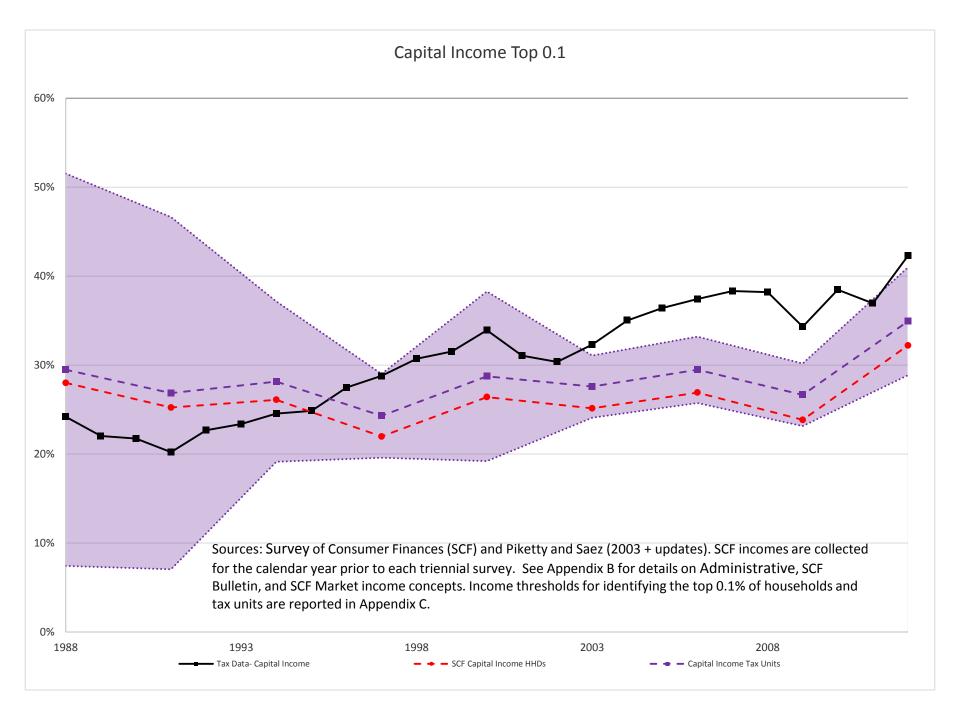


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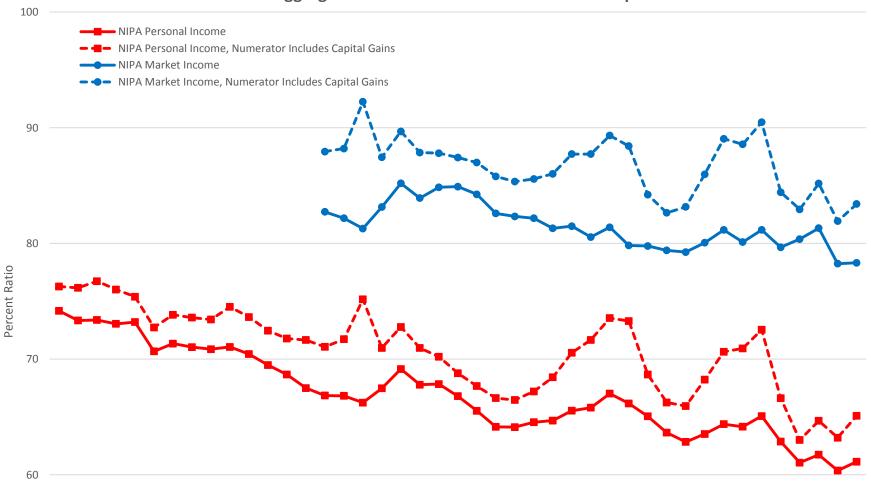




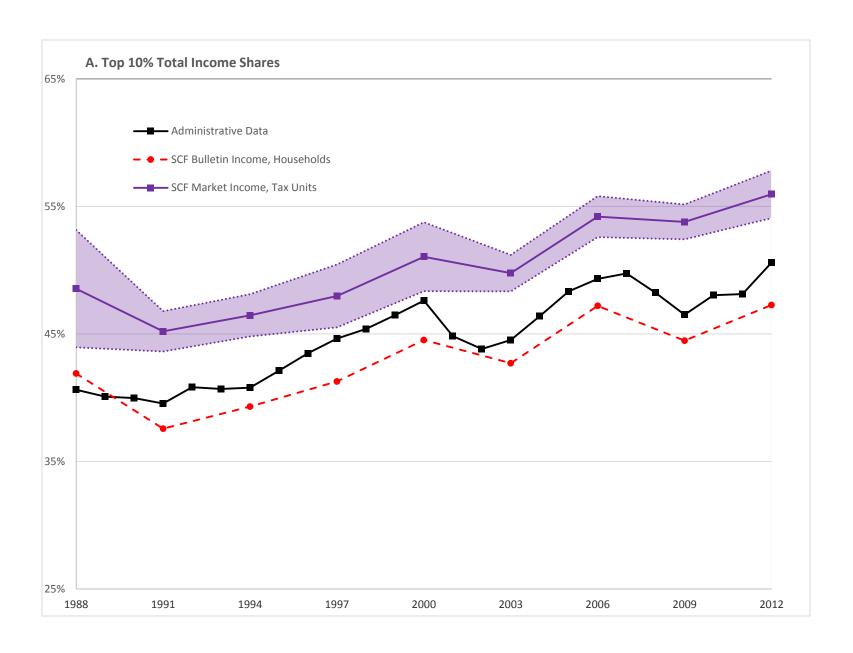


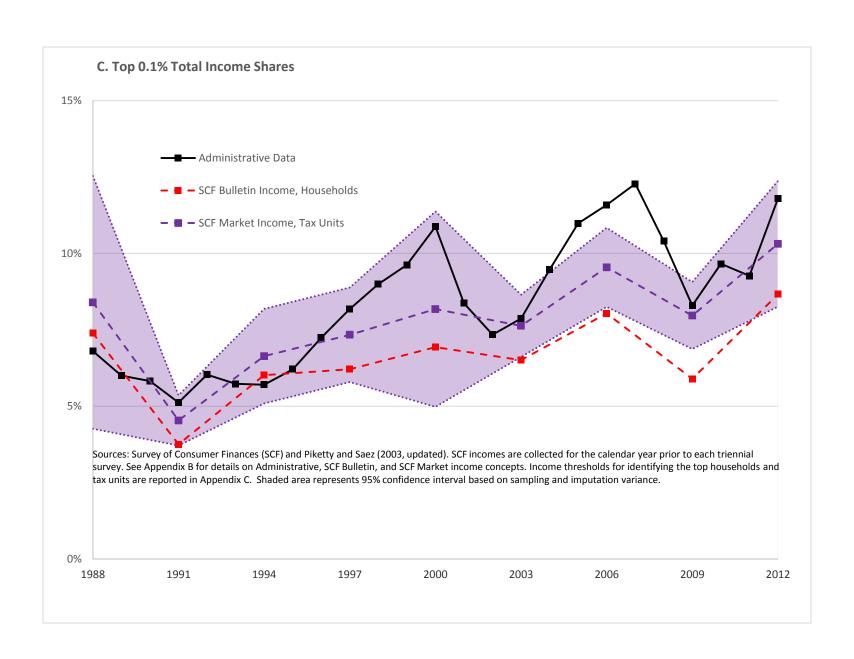


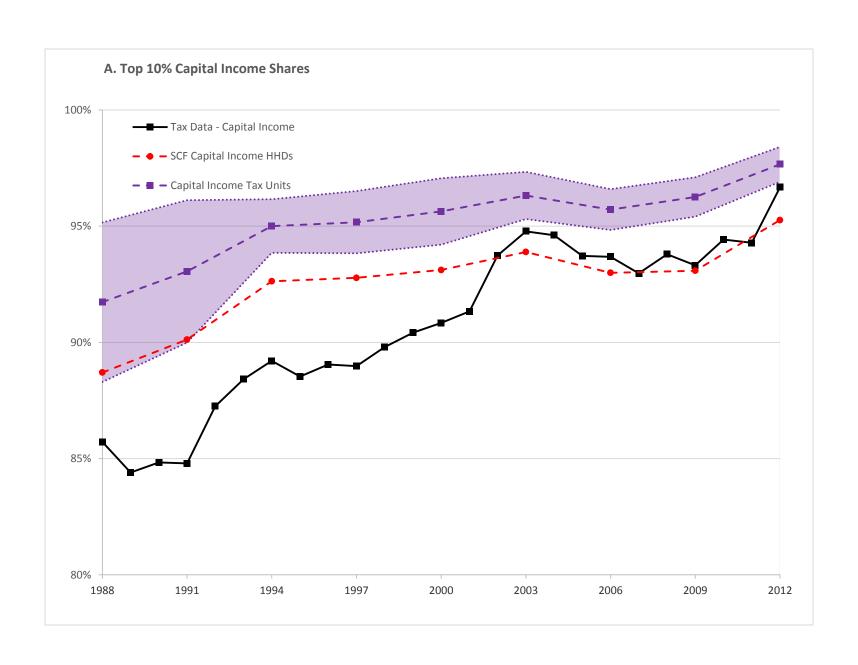
#### Ratio of Administrative Data Aggregate Incomes to Alternative NIPA Concepts

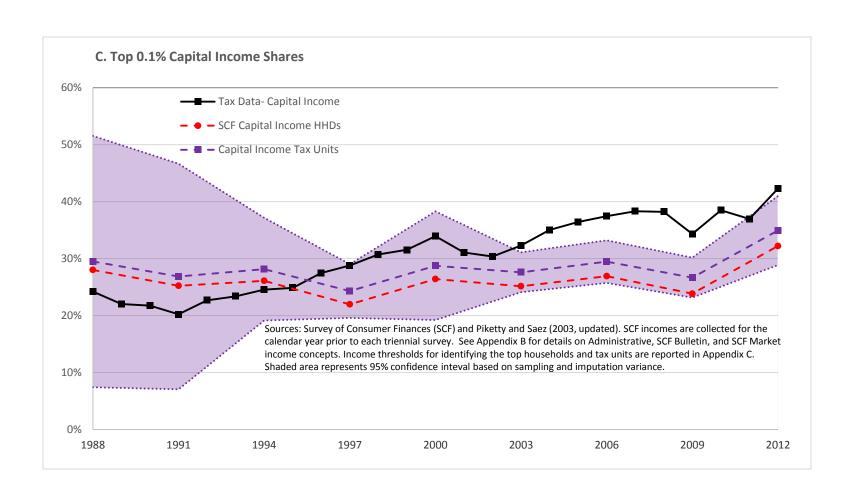


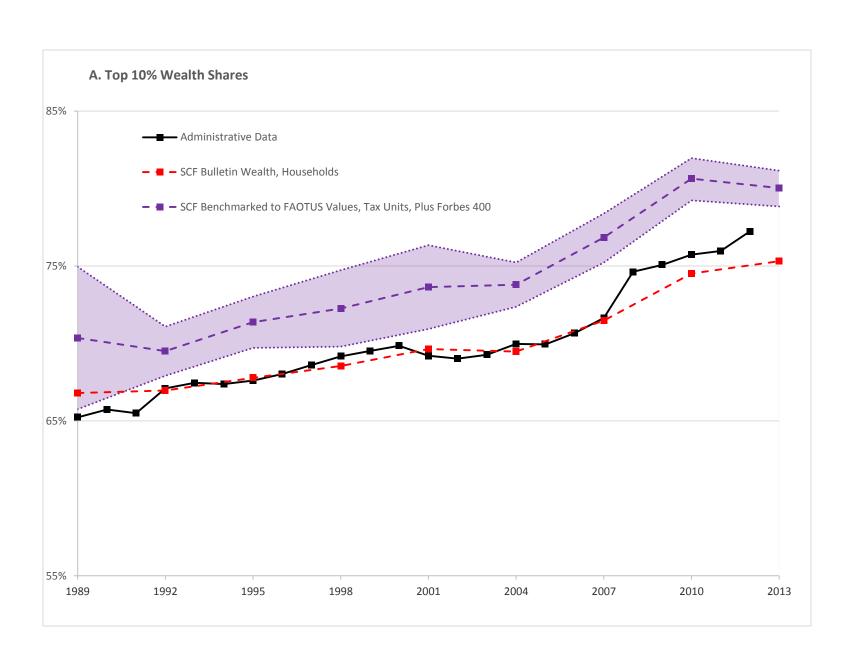
Sources: Bureau of Economic Analysis and Piketty and Saez (2003 + updates). NIPA Market Income is Personal Income less government transfers to persons, employer contributions for pension and insurance funds, and interest and dividends earned on retirement funds. Retirement benefits received are then added back in. NIPA data for retirement funds is available beginning in 1984. See Appendix B for details.

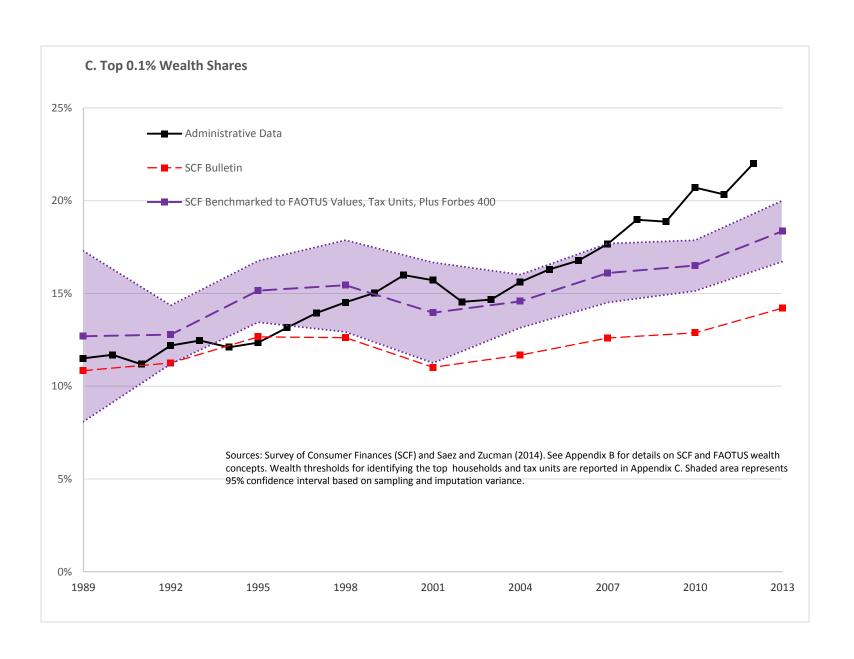












**Table 2. Correlation Between SCF Wealth and Predicted Gross-Capitalization and Empirical Correlation Wealth** 

	(1)	(2)	(3)			
ln(GC model wealth)	0.85		0.26			
	(0.02)	•••	(0.02)			
ln(Corr. model wealth)		1.02	0.76			
m(corr. moder weartii)	•••	(0.01)	(0.03)			
		(010-)	(0.00)			
Constant	1.57	-0.46	-0.73			
	(0.25)	(0.23)	(0.22)			
$\mathbb{R}^2$	0.69	0.78	0.80			
Obs.	1,450	1,450	1,450			
D. P. (117 - 14)						
Predicted ln(wealth) at mean:						
	15.42	15.43	15.35			

Note: Regression of log of SCF family net worth in 2013 on log of predicted wealth of gross capitalization model (col. 1), correlation model (col. 2), and both (col. 3). Data from first implicate of SCF survey data matched to the wealth predictions that were used to stratify the list sample. Standard error in ().

Table 4. Impact of using multiple years of data to classify families

Bottom 90

 Bottom 90
 90-99
 99-99.9
 99-99.9
 99.9-99.9
 99.9-99.9
 99.99+

 0.98
 0.02
 0.00
 0.00
 0.00

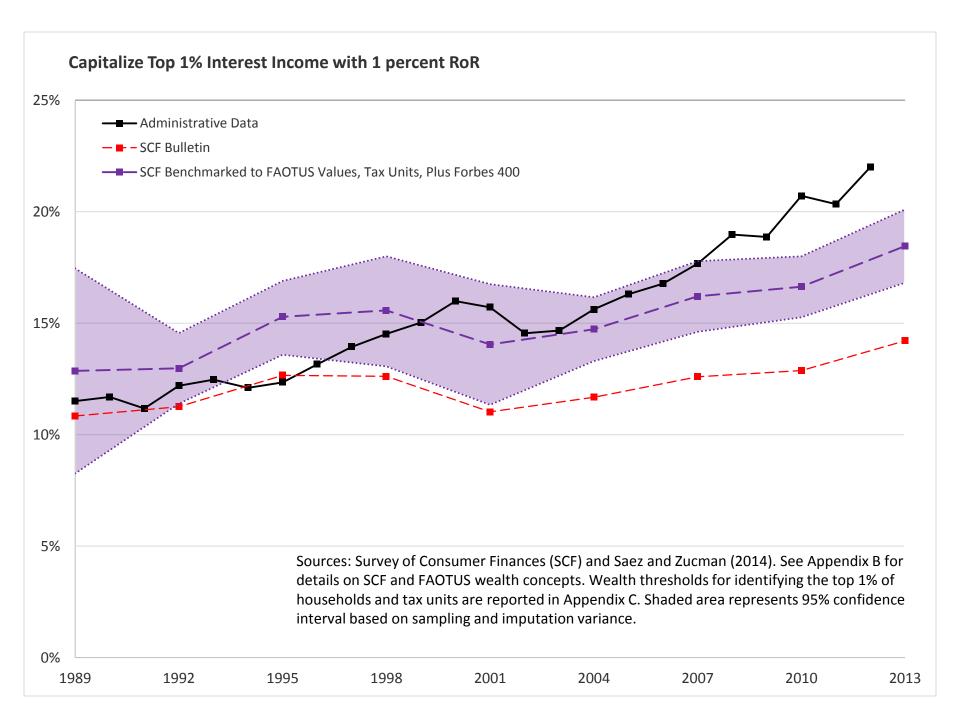
 0.04
 0.93
 0.03
 0.00
 0.00

2011-only gross capitalization model

2011 2000	Dolloin 90	0.96	0.02	0.00	0.00	0.00
2011-2009	90-99	0.04	0.93	0.03	0.00	0.00
gross- capitalization	(Top 1) 99-99.9	0.00	0.06	0.89	0.05	0.00
model	(Top 0.1) 99.9-99.99	0.00	0.00	0.06	0.90	0.04
	(Top 0.01) 99.99+	0.00	0.00	0.01	0.05	0.94
			2011-or	nly correla	tion model	
				(TC 1)	(TE 0.1)	(TE 0.01)

				(Top 1)	(Top 0.1)	(Top 0.01)
		Bottom 90	90-99	99-99.9	99.9-99.99	99.99+
	Bottom 90	0.97	0.03	0.00	0.00	0.00
2011-2009 90-99	0.07	0.87	0.06	0.00	0.00	
correlation	(Top 1) 99-99.9	0.00	0.08	0.86	0.06	0.00
model (T	(Top 0.1) 99.9-99.99	0.00	0.00	0.11	0.84	0.05
	(Top 0.01) 99.99+	0.00	0.00	0.00	0.14	0.86

Note: Rows sum to 1. Tables show the impact of using 3 years of administrative data (2011, 2010, and 2009) versus 1 year of data (2011) to organize top-end families and are organized similarly to table 1. Source: 2011 INSOLE data (supplemented with two years of INSOLE or CDW panel data) compared to 2011 INSOLE data only.



#### IV. Targeting Permanently Wealthy

- Why does SCF find lower and flatter trajectory top 0.1% shares for both income or wealth?
- SCF sampling strategy finds permanently
  wealthy by grossing up capital income and
  empirical correlation between wealth/income
- Predicted wealth based on just gross capitalization too high relative to empirical

Figure 13. Predicted Top 0.1 Percent Wealth Share from Gross-Capitalization and Empirical Correlation Model

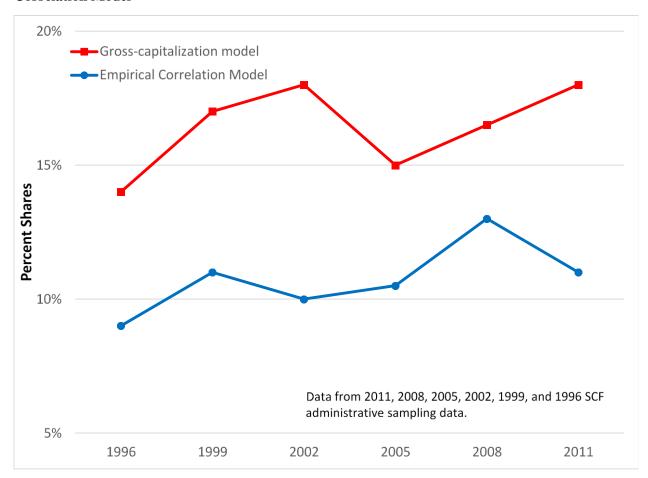


Table 1. Impact of ranking top-end families by an alternate model

Correlation Model Percentile

				(Top 1)	(Top 0.1)	(Top 0.01)
		Bottom 90	90-99	99-99.9	99.9-99.99	99.99+
	Bottom 90	0.89	0.10	0.01	0.00	0.00
Gross-	90-99	0.20	0.48	0.28	0.04	0.00
capitalization	(Top 1) 99-99.9	0.05	0.22	0.48	0.23	0.02
Percentile	(Top 0.1) 99.9-99.99	0.03	0.10	0.31	0.46	0.10
	(Top 0.01) 99.99+	0.01	0.03	0.11	0.39	0.47

Note: Rows sum to 1. Table describes where a family ranked in gross capitalization model would be ranked in the empirical correlation model. For example, in the last row, of families ranked in top 0.01 percentile in the gross capitalizations model, 1 percent of families are ranked in the bottom 90 percentiles by the correlation model, 3 percent are ranked between the 90-99<sup>th</sup> percentiles by the correlation model, 11 percent are ranked between the 99<sup>th</sup>-99.9<sup>th</sup> percentile by the correlation model, 39 percent are ranked between the 99.9<sup>th</sup> and 99.99<sup>th</sup> percentile by the correlation model, and 47 percent are ranked in the top 0.01 percent by the correlation model. Source: 2011 INSOLE data, supplemented with two years of INSOLE or CDW panel data.

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				(Top 1)	(Top 0.1)	(Top 0.01)
		Bottom 90	90-99	99-99.9	99.9-99.99	99.99+
	Bottom 90	0.89	0.10	0.01	0.00	0.00
Gross-	90-99	0.20	0.48	0.28	0.04	0.00
capitalization	(Top 1) 99-99.9	0.05	0.22	0.48	0.23	0.02
Percentile	(Top 0.1) 99.9-99.99	0.03	0.10	0.31	0.46	0.10
	(Top 0.01) 99.99+	0.01	0.03	0.11	0.39	0.47

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Table 3. Pearson and Spearman Correlations: SCF Wealth and Predicted Gross-Capitalization and Empirical Correlation Wealth

	Spearman correlations					
	2013	2010	2007	2004	2001	
Gross-capitalization model	0.83	0.82	0.83	0.82	0.78	
Empirical correlation model	0.90	0.91	0.90	0.91	0.87	

Note: Data from first implicate of SCF survey data matched to wealth indices used to stratify the list sample.

#### Why Does Sampling Matter?

- Samples at very top diverge because of nonincome generating wealth and volatility
- Imagine two very wealthy families
  - 50% chance of 2\*average income, 50% zero
  - Remember high end volatility figure!
- Gross capitalization model only counts the one who realizes the income, SCF gets both
- Is classification bias increasing over time?

### Why Do We Care?

- Are top shares just high and rising, or really high and rising really fast?
- Useful to (1) confirm SCF is capturing high end and (2) understand why administrative and survey-based top share estimates diverge
- Reasons for divergence vis a vis NIPA speak to shortcomings of SCF concepts as well
  - SCF income and wealth also not comprehensive
  - SCF top shares predictably biased up!