

Revisiting Subprime Lending: Cross-Sectional Patterns of Mortgage Debt During the Housing Boom

Christopher Foote, Lara Loewenstein, and Paul Willen

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Disclaimer: I do not speak for:

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Eric Rosengren, President of Boston Fed

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- **Distorted beliefs/over-optimism:**

higher house prices → higher low-income lending

This paper: Fact #1

- 1 There was no reallocation of mortgage debt to low-income borrowers.

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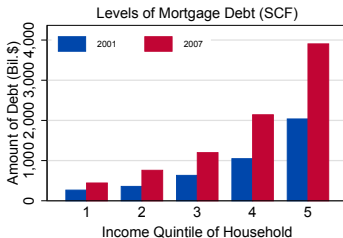
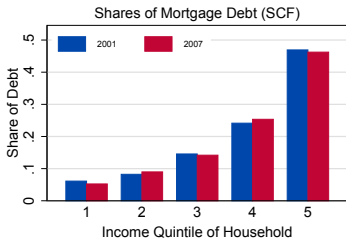
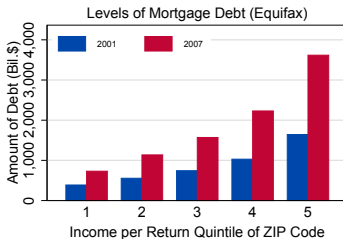
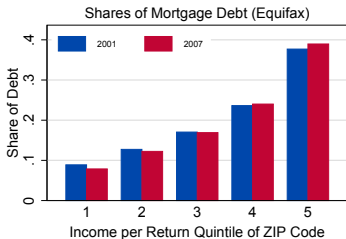
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 - Subprime did not cause a reallocation of debt.
 - It prevented one.

Distribution of Mortgage Debt

Sources: NY Fed Consumer Credit Panel/Equifax, IRS, and SCF.



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 - HMDA data is a good measure of one gross flow (originations)...
 - ... but HMDA measures neither the other gross flow (terminations) nor the stock of debt.
 - Stocks/flows distinction is crucial for understanding the debate between Mian and Sufi (2009) and Adelino, Schoar and Severino (2015).

This paper: Fact #3

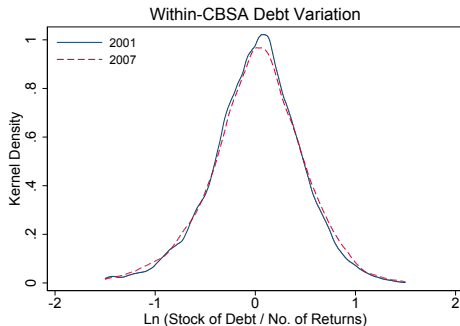
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- Mian and Sufi (2009, p. 1459, emphasis added): “...it is critical to understand the variation [in debt] within counties if we are to understand the causes and consequences of the mortgage default crisis.”
 - Distribution of ZIP-level debt on within-CBSA basis:



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 - Data-suppression rules change over time.
 - Not everyone is required to file a tax return.
 - The number of returns in 2007 increased sharply, especially in low income areas, due to the availability of a stimulus payment.

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 - Relatively small samples (about 3,000 to 6,500 households per survey).

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- Standard errors are clustered by CBSA or county for ZIP-level regs.

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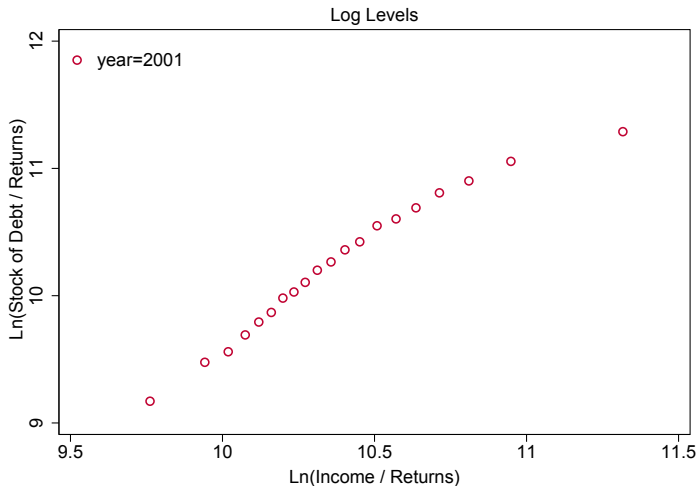
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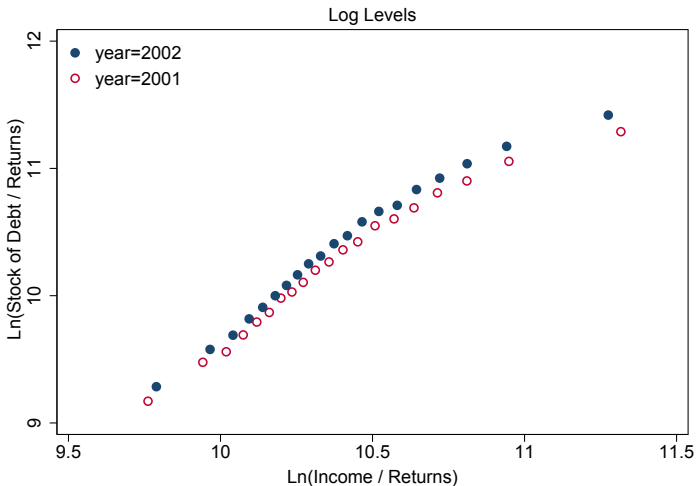
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- We need to estimate β_1 in each year to rule out this possibility.

Levels Binscatters: Year-by-Year



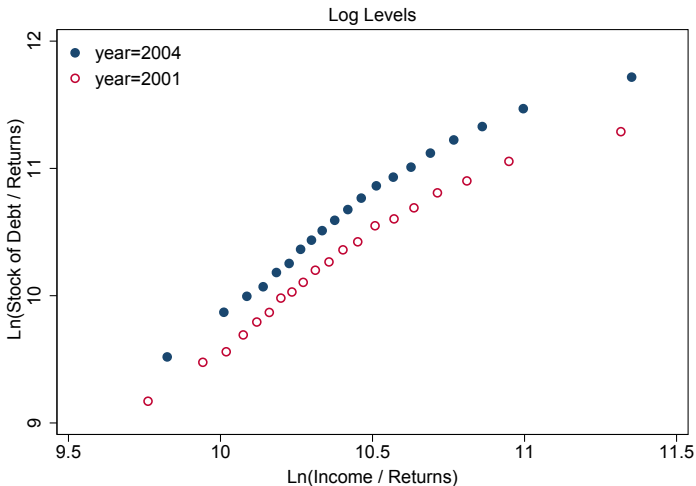
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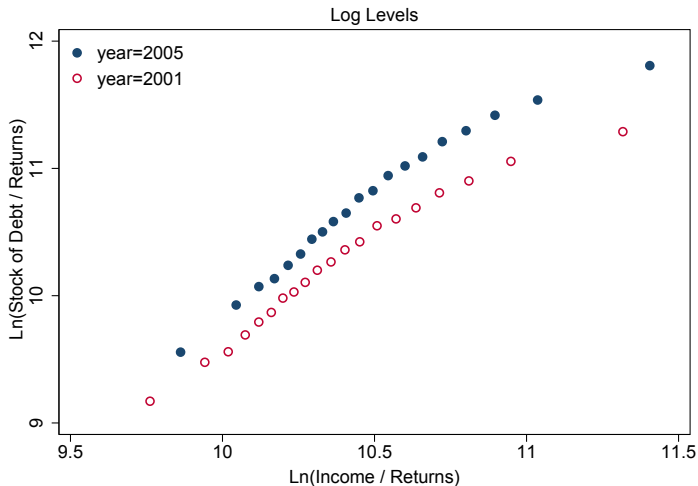
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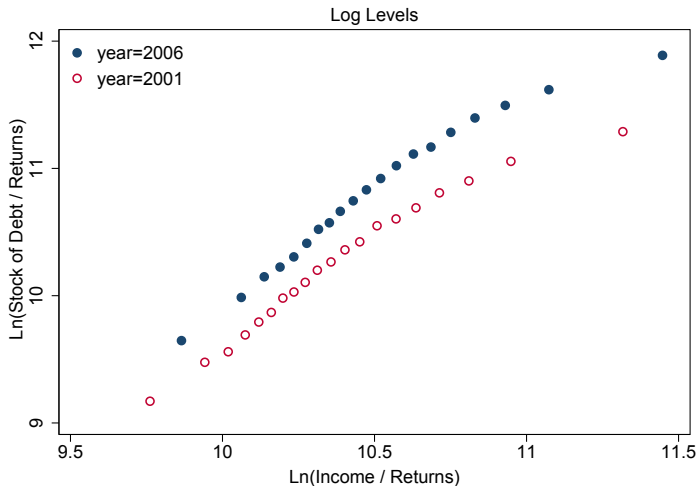
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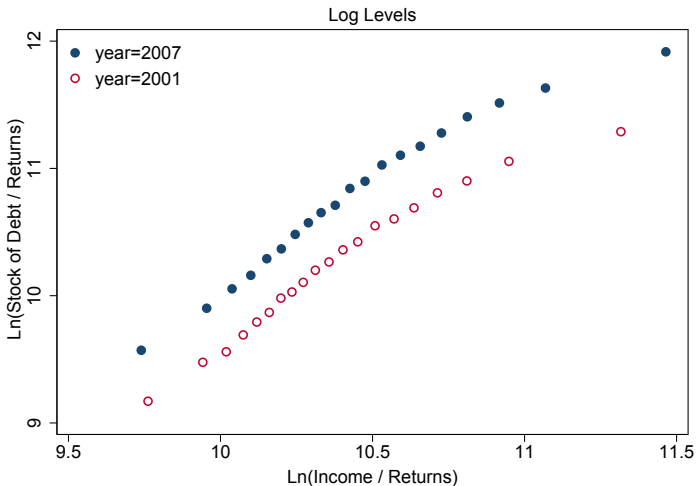
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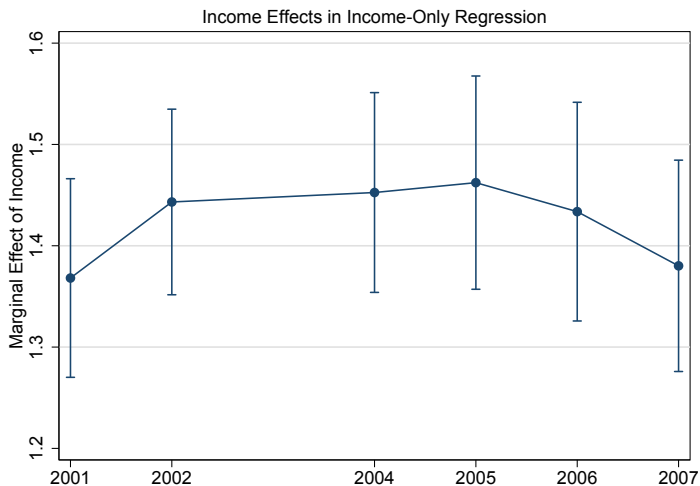
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Levels Binscatters: Year-by-Year



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Levels Regressions: Income Effects



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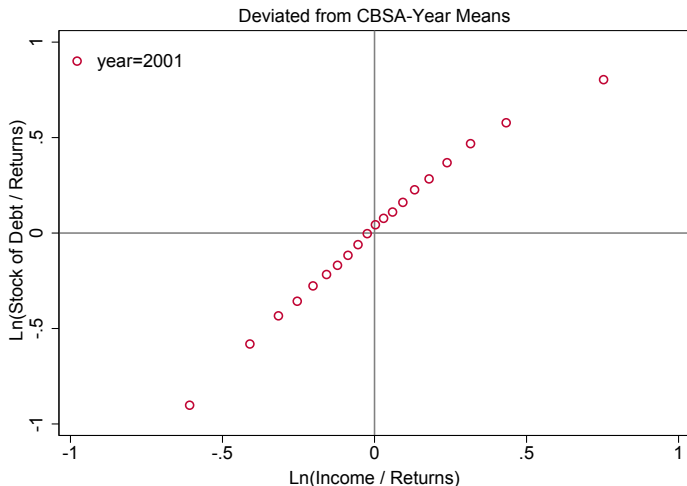
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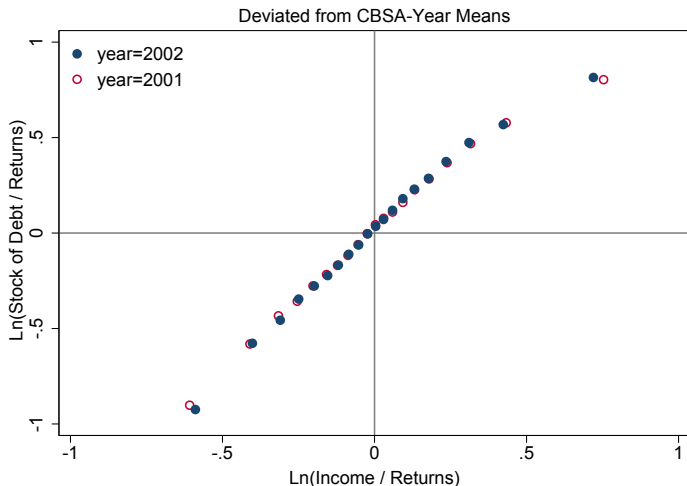
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- Additionally, the intercepts (β_c) can be analyzed in their own right.
 - Why did debt rise more in Phoenix than in Wichita?

CBSA-Deviated Levels Binscatters: Yr-by-Yr



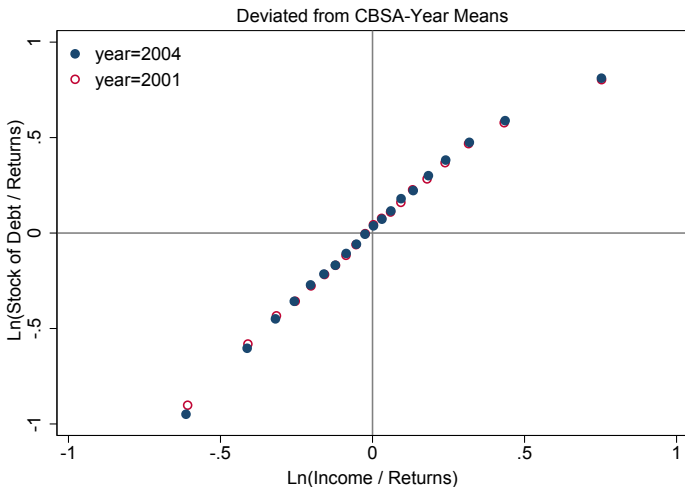
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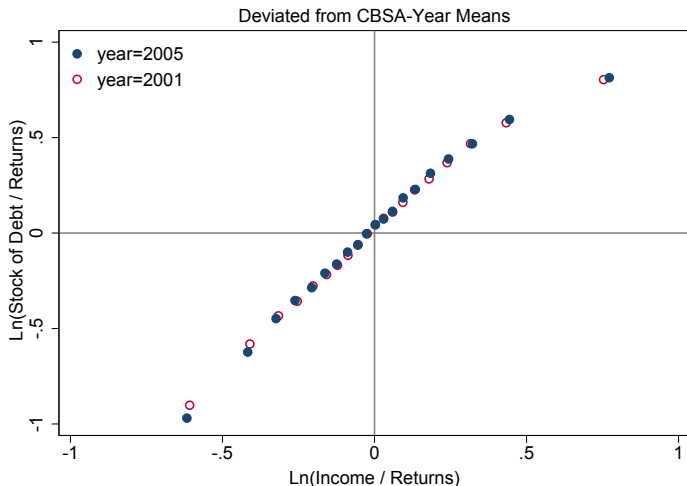
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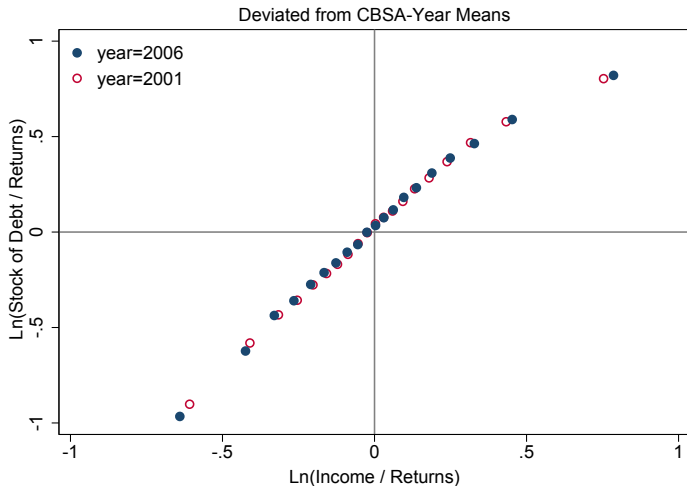
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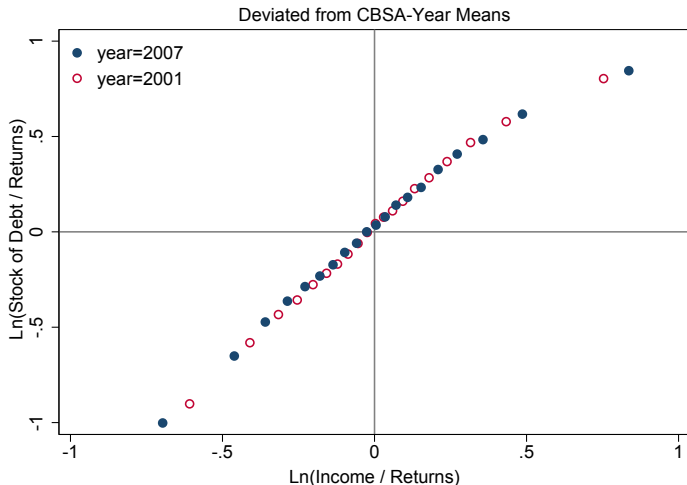
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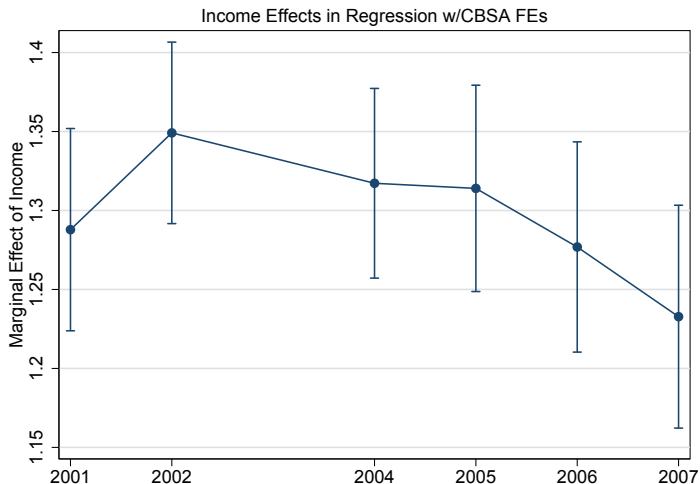
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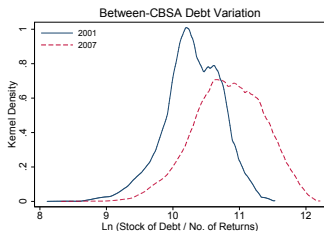
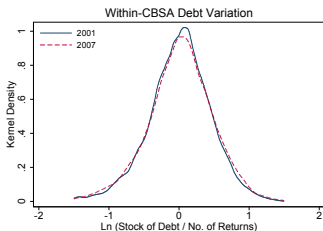
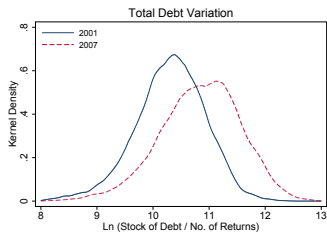
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Regressions w/CBSA FEs: Income Effects



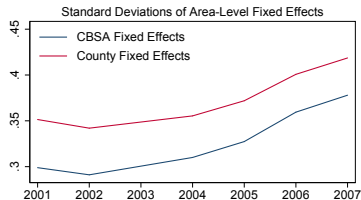
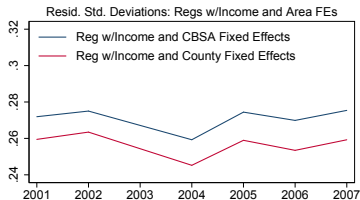
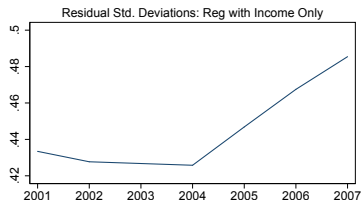
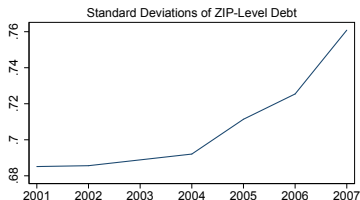
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Total, Within- and Between-CBSA Variation



Source: NY Fed Consumer Credit Panel/Equifax and IRS.

Between and Within Variation in Regressions



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Credit Allocation Function in Long-Difference Form

$$Debt_{ic,2007} = \beta_c,2007 + \beta_1 Income_{ic,2007} + \epsilon_{ic,2007}$$

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- How does debt for a given ZIP code change over time?

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- If β_1 s do not change over time, then estimating a long-difference regression is easy:

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- If β_1 s do not change over time, then estimating a long-difference regression is easy:

$$\Delta Debt_{ic} = \Delta \beta_c + \beta_1 \Delta Income_{ic} + \Delta \epsilon_{ic}$$

- If β_1 s do change, then we need to put an income level in the regression as well.

Long-Difference Specification Issues

$$y_2 = \beta_2 x_2$$

$$y_1 = \beta_1 x_1$$

$$y_2 - y_1 = \beta_2 x_2 - \beta_1 x_1$$

$$y_2 - y_1 = \beta_2(x_2 - x_1) + x_1(\beta_2 - \beta_1)$$

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- Note that coefficient on level (x_1 or x_2) is always the same: $\beta_2 - \beta_1$.

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 - ...depends on which level is included...
 - ... but always reflects a level effect (β_1 or β_2).

Long-Difference Regression Results

Dependent Variable: 2001-07 ZIP-Level Change in Ln Mortgage Debt per Return

Sample Restriction	(1) None	(2) None	(3) 1% Trim	(4) 5% Trim
Panel A: All ZIP Codes				
2001-07 Change in Ln Income per Return	1.071*** (0.040)	1.050*** (0.043)	1.170*** (0.054)	1.031*** (0.061)
2001 Ln Income per Return Level		0.019 (0.012)	0.010 (0.014)	0.031 (0.016)
Constant	0.527*** (0.008)	0.527*** (0.008)	0.528*** (0.008)	0.528*** (0.008)
R-sq.	0.122	0.122	0.150	0.117
Observations (No. of ZIP Codes)	35,595	35,595	27,337	18,313
Expected Diff. in Debt Growth: 90th 2001 Income Pctile vs. 10th 2001 Income Pctile		0.017	0.009	0.027

Standard errors in parentheses

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

Source: NY Fed Consumer Credit Panel/Equifax and IRS.

Long-Difference Regression Results

Dependent Variable: 2001-07 ZIP-Level Change in Ln Mortgage Debt per Return

Sample Restriction	(1) None	(2) None	(3) 1% Trim	(4) 5% Trim
Panel B: CBSA ZIP Codes without Fixed Effects				
2001-07 Change in Ln Income per Return	1.088*** (0.043)	1.059*** (0.046)	1.192*** (0.057)	1.028*** (0.064)
2001 Ln Income per Return Level		0.027* (0.014)	0.010 (0.016)	0.032 (0.018)
Constant	0.527*** (0.009)	0.527*** (0.009)	0.528*** (0.009)	0.529*** (0.008)
R-sq.	0.141	0.142	0.164	0.120
Observations (No. of ZIP Codes)	27,567	27,567	21,634	15,165
Expected Diff. in Debt Growth: 90th 2001 Income Pctile vs. 10th 2001 Income Pctile		0.023	0.009	0.028

Standard errors in parentheses

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Source: NY Fed Consumer Credit Panel/Equifax and IRS.

Long-Difference Regression Results

Dependent Variable: 2001-07 ZIP-Level Change in Ln Mortgage Debt per Return

Sample Restriction	(1) None	(2) None	(3) 1% Trim	(4) 5% Trim
Panel C: CBSA ZIP Codes with CBSA Fixed Effects				
2001-07 Change in Ln Income per Return	0.827*** (0.064)	0.858*** (0.060)	0.990*** (0.062)	0.925*** (0.066)
2001 Ln Income per Return Level		-0.027 (0.015)	-0.057*** (0.016)	-0.052*** (0.015)
Constant	0.527*** (0.000)	0.527*** (0.000)	0.528*** (0.000)	0.529*** (0.000)
R-sq.	0.429	0.429	0.553	0.580
Observations (No. of ZIP Codes)	27,567	27,567	21,634	15,165
Expected Diff. in Debt Growth: 90th 2001 Income Pctile vs. 10th 2001 Income Pctile		-0.023	-0.049	-0.045

Standard errors in parentheses

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Source: NY Fed Consumer Credit Panel/Equifax and IRS.

Determinants of CBSA-Level Debt Growth

Dependent Variable: CBSA-Level Fixed Effects from ZIP-Level Long-Diff Regressions

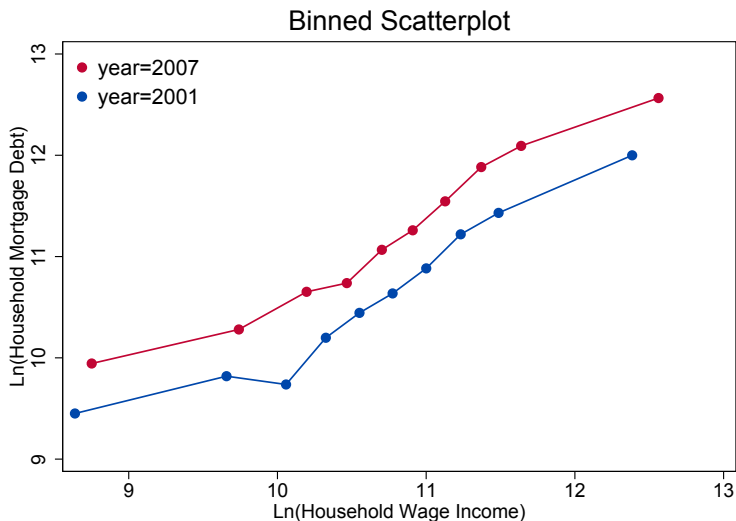
	(1)	(2)	(3)	(4)	(5)	(6)
2001-07 Change in Ln CBSA Income		0.68*** (0.16)	0.55** (0.17)		-0.28 (0.17)	-0.45* (0.18)
2001 Ln CBSA Income Level			0.22*** (0.05)			0.24*** (0.04)
2001-07 Change in Ln CBSA House Price				0.40*** (0.04)	0.44*** (0.04)	0.45*** (0.04)
Constant	0.53*** (0.01)	0.53*** (0.01)	0.53*** (0.01)	0.36*** (0.02)	0.35*** (0.02)	0.34*** (0.02)
Observations (No. of CBSAa)	934	934	934	934	934	934
R-sq.	0.00	0.05	0.13	0.31	0.32	0.42

Standard errors in parentheses

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Source: NY Fed Consumer Credit Panel/Equifax and IRS.

Household-Level Data from SCF



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 - Number of children.

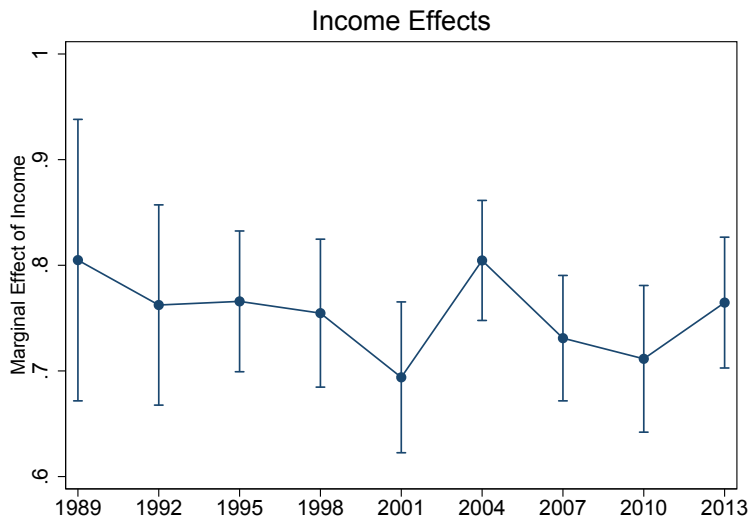
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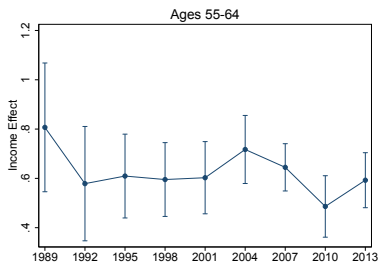
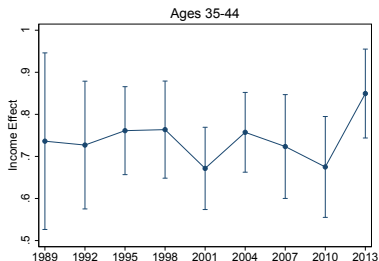
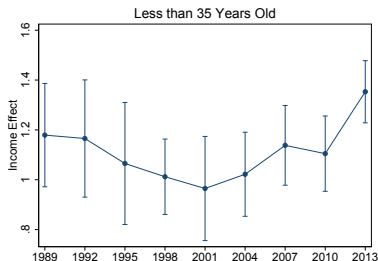
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- Other regressors:
 - Indicators for age group of household head (<35,35-44,45-54,55-64), nonwhite, and marital status.
 - Number of children.
- Households headed by persons 65 or older are excluded, as are people with no wage income.

SCF Results

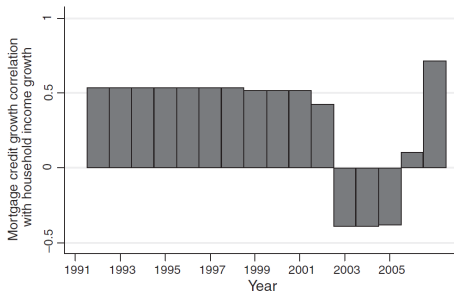


SCF Results: With Age \times Income Interactions



The Negative Correlation in Mian and Sufi (2009)

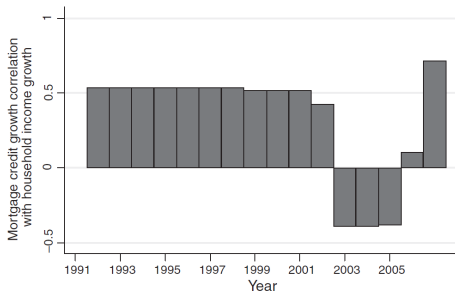
Note: All correlations calculated on a within-county basis.



- Potential specifications for regressions of Δ debt on Δ income:

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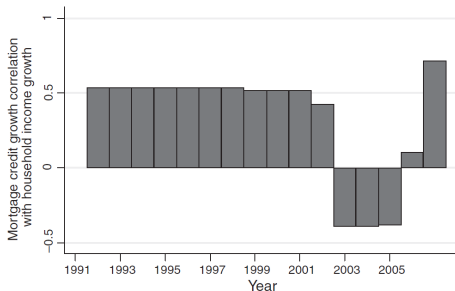
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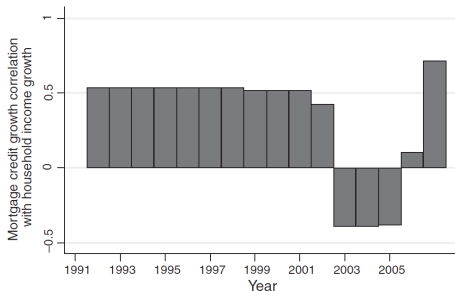
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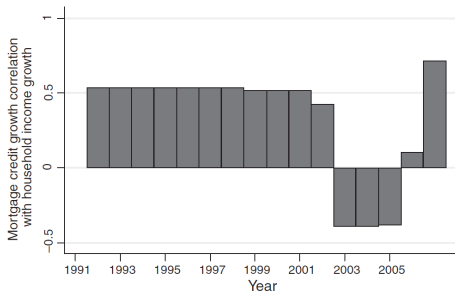
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 - 4 Adelino et al. (2015): Value of loans versus number of loans?

2002-06 Debt-Growth Regressions

All regressions include county FEs and use AGI as income measure

	Stock				Flow			
Control for 2002 level?	Yes		No		Yes		No	
Value or Number of Loans?	\$	#	\$	#	\$	#	\$	#
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
$\Delta \ln(\text{AGI}/\text{Returns})$	0.42*** (0.05)	0.09** (0.04)	0.29*** (0.05)	0.14*** (0.03)	0.29*** (0.04)	0.60*** (0.07)	0.17*** (0.03)	-0.34*** (0.07)
$\ln(\text{AGI}/\text{Returns in 2006})$	-0.05*** (0.01)	0.02 (0.01)			-0.04*** (0.01)	-0.34*** (0.02)		
Constant	0.55*** (0.04)	0.07 (0.05)	0.37*** (0.01)	0.14*** (0.01)	0.36*** (0.04)	1.52*** (0.07)	0.21*** (0.01)	0.33*** (0.01)
Observations	35,683	35,864	35,683	35,864	29,008	29,008	29,008	29,008
R-sq.	0.40	0.28	0.39	0.28	0.67	0.63	0.67	0.57
R-sq. w/o County FE	0.04	0.02	0.04	0.02	0.08	0.15	0.07	0.00

Standard errors in parentheses

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2002-06 Debt-Growth Regressions

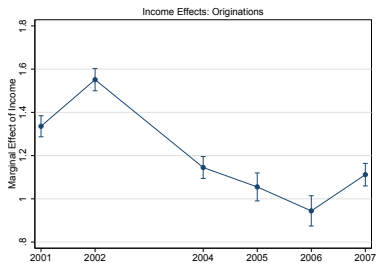
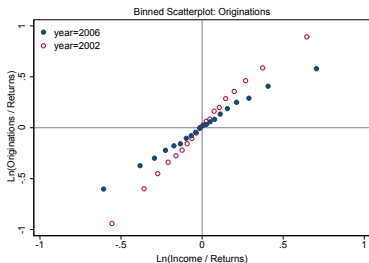
All regressions include county FEs and use salary and wages as income measure

	Stock				Flow			
	Yes		No		Yes		No	
	\$	#	\$	#	\$	#	\$	#
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
$\Delta \ln(\text{Salary/Returns})$	0.74*** (0.05)	0.38*** (0.06)	0.70*** (0.05)	0.47*** (0.07)	0.29*** (0.06)	0.68*** (0.08)	0.22*** (0.05)	0.02 (0.09)
$\ln(\text{Salary/Returns in 2006})$	-0.02 (0.01)	0.06*** (0.02)			-0.04*** (0.01)	-0.37*** (0.02)		
Constant	0.41*** (0.05)	-0.07 (0.06)	0.34*** (0.01)	0.12*** (0.01)	0.36*** (0.04)	1.52*** (0.08)	0.22*** (0.01)	0.27*** (0.01)
Observations	35,611	35,788	35,611	35,788	28,967	28,967	28,967	28,967
R-sq.	0.41	0.30	0.41	0.30	0.67	0.62	0.66	0.57
R-sq. w/o County FE	0.07	0.04	0.07	0.03	0.10	0.13	0.09	0.00

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Source: NY Fed Consumer Credit Panel/Equifax and IRS.

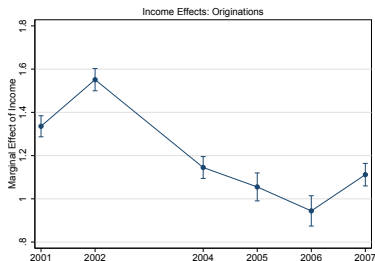
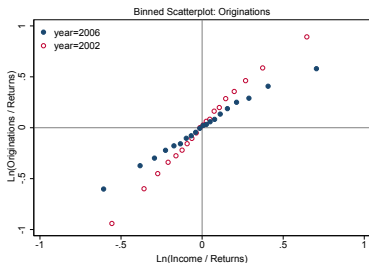
Stocks vs. Flows: Total Value of Originations



Source: NY Fed Consumer Credit Panel/Equifax and IRS.

- Using the Equifax data, we also find decline in the positive relationship between income and originations at the ZIP code level.

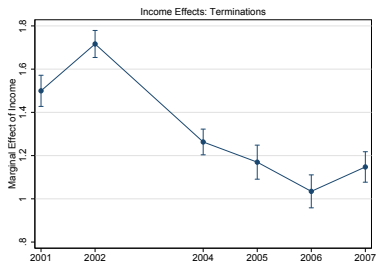
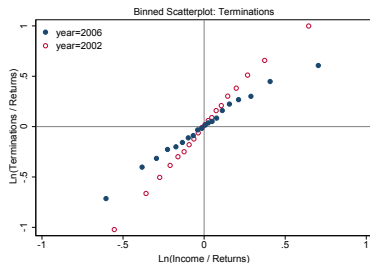
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- Over the course of the housing boom, originations rose more in ZIP codes that were relatively poor compared to others in their county.

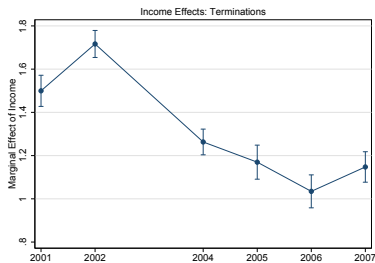
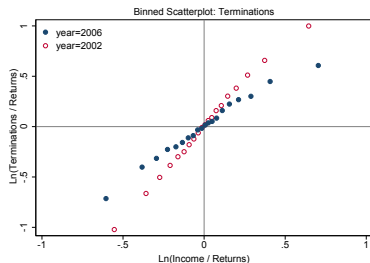
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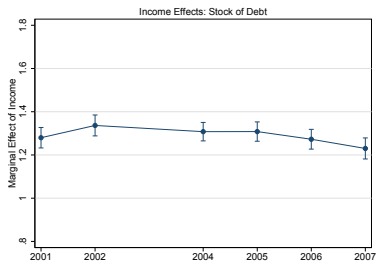
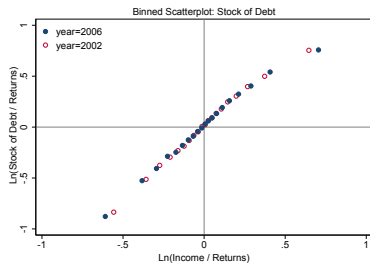
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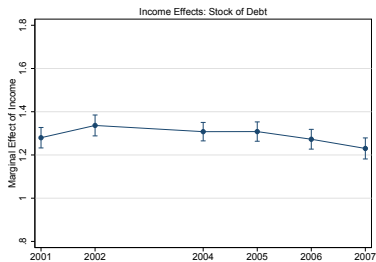
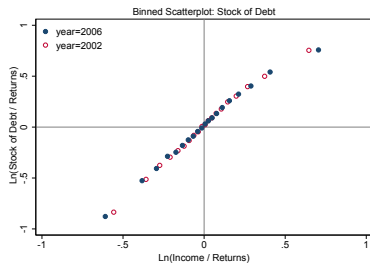
Stocks vs. Flows: Stock of Debt



Source: NY Fed Consumer Credit Panel/Equifax and IRS.

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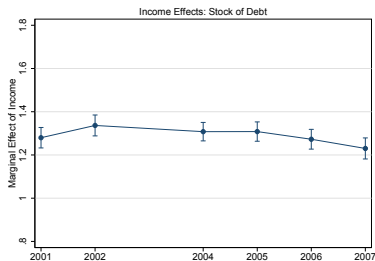
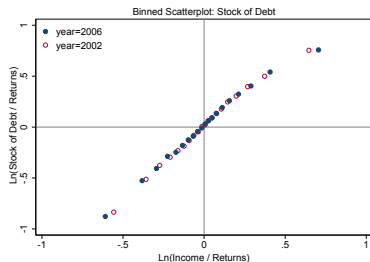
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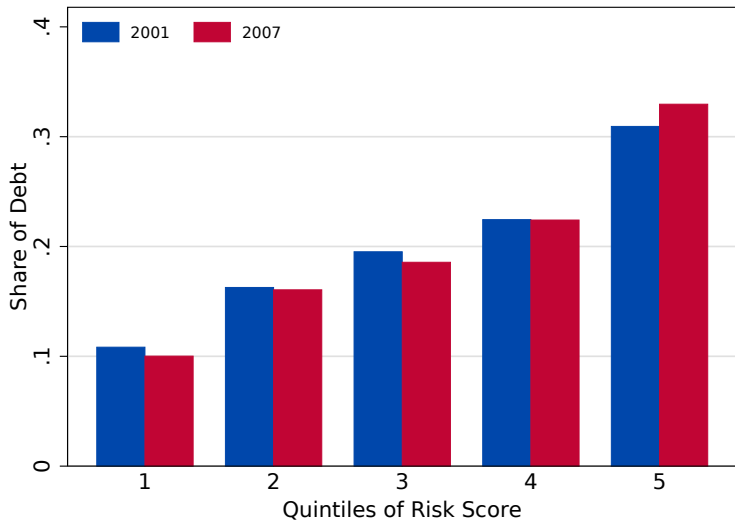
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- Only difference was time period (2001-07) and use of CBSA (not county) fixed effects.

What About Credit Scores?



Source: NY Fed Consumer Credit Panel/Equifax.

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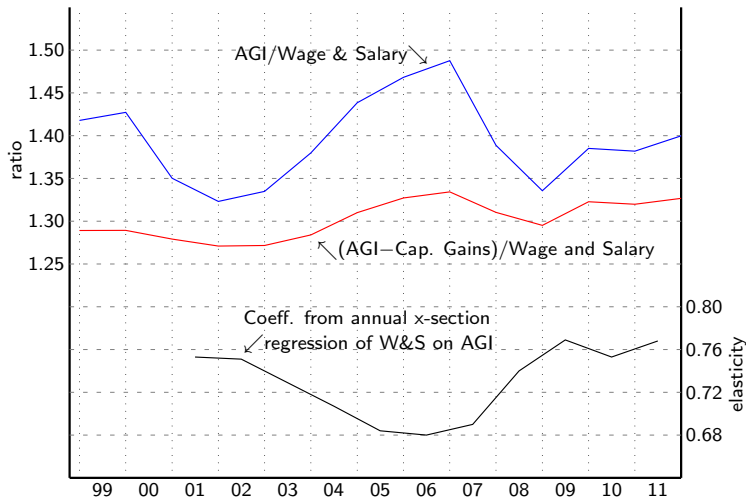
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 - **Distorted beliefs/over-optimism:**
 - higher house prices → higher ~~low-income~~ lending

Supplementary Slides

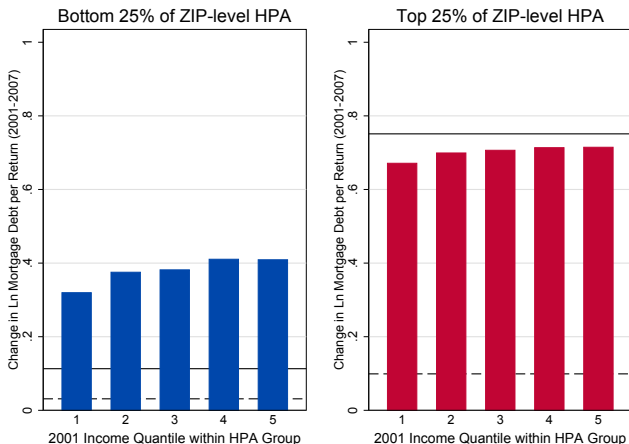
Adjusted Gross Income (AGI) vs. Salary and Wages



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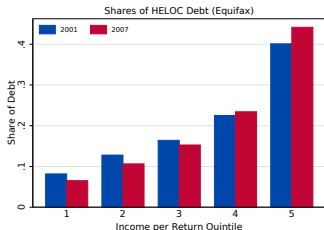
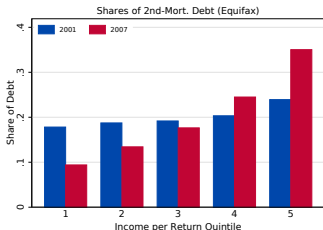
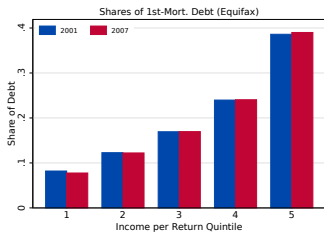
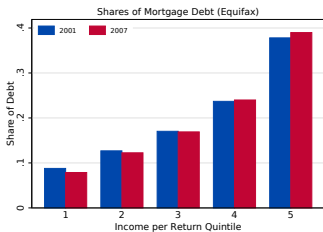
Growth in ZIP-level Debt, House Prices, and Income

Solid line: House price appreciation (HPA); dashed line: income growth



Source: NY Fed Consumer Credit Panel/Equifax and IRS.

Distribution of Mortgage Debt by Type



Source: NY Fed Consumer Credit Panel/Equifax and IRS.

Comparing Debt Aggregates

