

# Consumer Durables and the Distributional Effects of Credit Supply Shocks

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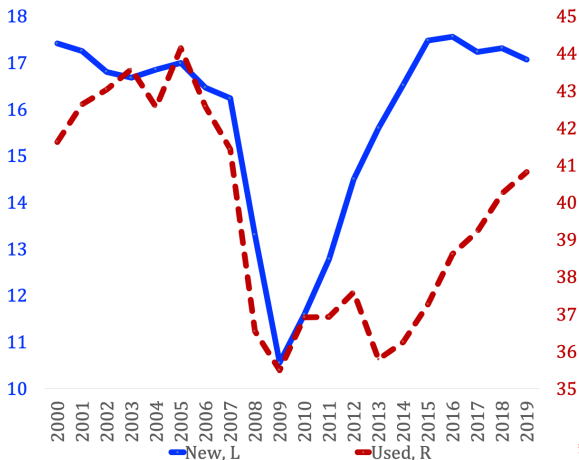
October 29, 2020

## Motivation

The Great Recession featured

1. severely worsened liquidity in various financial markets
2. a large decline in the expenditure on consumer durable goods

U.S. Auto Sales (Millions of Vehicles)

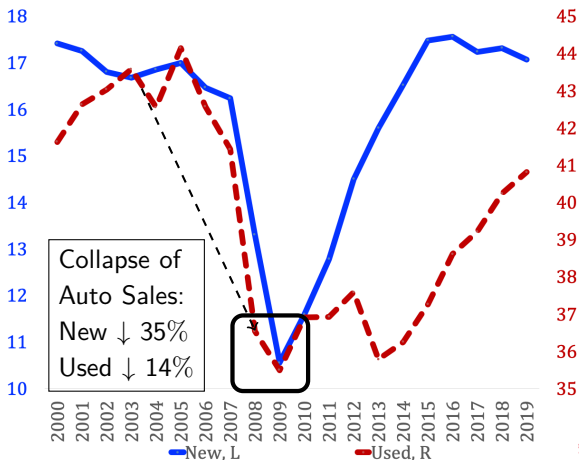


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### Why Cars? Why the Great Recession? Why auto credit?

- ❏ CD: a large, highly volatile, and procyclical component of GDP:  
 $\downarrow \Delta(\text{CD}) = 24\% \downarrow \Delta(rGDP)$  the GR
- ❏ auto: **biggest, most volatile** component of CD
- ❏ not the cause of the credit crunch during the GR
- ❏ relatively higher frequency of trading
- ❏ straightforward to measure quality

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Why Cars? [Why the Great Recession?](#) Why auto credit?

- ❖ Narrowly defined durables consumption ↓ 14.2% 2007Q4 - 2009Q2, >> 9.7% average across all US post-war recessions
- ❖ sharper decline in the availability of credit compared to other recessions

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Why Cars? Why the Great Recession? [Why auto credit?](#)

- ❑ most auto sales are financed (>50% used, >80% new)
- ❑ heterogeneity among auto lenders :  
**banks v.s. nonbank financial institutions**



## Nonbanks. v.s. Banks as Auto Lender

### Nonbanks

✎ Market Share 2006	44.3%
✎ Primary Fund Source	Asset Backed Securities (ABS)
✎ Repossession Costs	Lower
✎ Example	Carmax
✎ Median Fico Score	655 Fair
✎ Median Loan Rate	10%

### Banks

55.7%
Deposits
Higher
Wells Fargo
703 Good
8.5%

## Nonbanks. v.s. Banks as Auto Lender

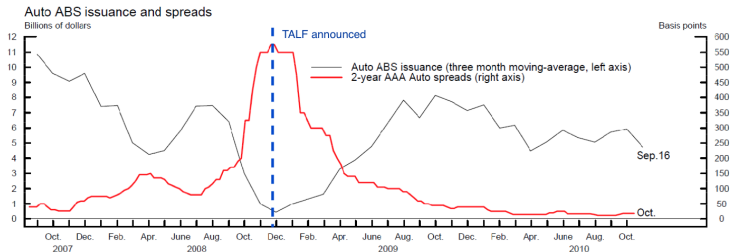
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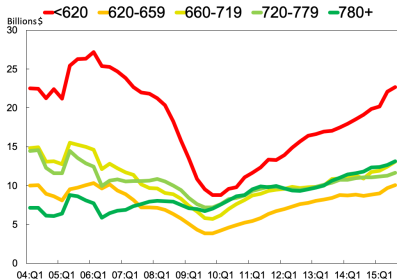
**collapse of the ABS market** → **dramatic increase** in the nonbank fund cost.



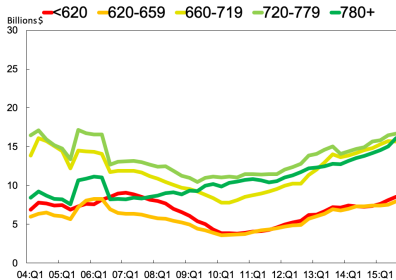
Source. JPMorgan Chase for spreads and Bloomberg for issuance.

# Suggestive Evidence of Distributional effects

## Figure 1: Auto Loan Origination by Riskscore



(a) Auto Finance Company



(b) Banks & Credit Unions

Source: New York Fed Consumer Credit Panel/Equifax. Billions



subprime lending concentrated on nonbanks

Delinquency



more severe shrinkage of credit provided to subprime from nonbanks.

# Applicants

## What I do in this paper: Facts

### 1. Empirical: document novel facts

So far, in the auto loan market

- ❏ subprime lending is concentrated on nonbanks
- ❏ nonbank lending to the subprime shrank dramatically v.s. banks

Later:

- ❏ auto purchase behavior: liquidation ↑ Retention ↑ Replacement ↓
- ❏ auto loan market: individual auto loan characteristics by lender type, pre and during GR

## What I do in this paper: Theory

**2. Theoretical** develop a dynamic equilibrium model with heterogeneous households and lenders

- ❖ **Lenders** differ in fund costs and repossession/foreclosure costs
- ❖ **Households** face uninsurable income and car quality shocks
  - ❖ choice of car qualities
  - ❖ saving borrowing decision **with the choice of lender** **new**
  - ❖ default option → endogenous auto loan rate schedules based on individual default risk
- ❖ Car markets clear

## Main Mechanism: **Asymmetric** Ability to Borrow

When **nonbank** credit supply shock occurs,

### Safe Household

easily switch to bank loans  
→ **limited increase** in loan rate  
→ **little impact** on car purchase decision

### Risky Household

nowhere else to borrow from  
→ **big increase** in loan rate if borrowing from bank  
→ big impact: buy **no** car, or buy a **cheaper** car

When **bank** credit supply shock occurs,

Being a bank borrower means **safe** → **little impact** on car purchase decision

## What I do in this paper: Quantitative Analysis

**3. Estimation** of the structural model by Simulated Method of Moments

**4. Counterfactual:** quantify the contribution of credit supply shocks by comparing Scenario 1, 2, 3

S1 only income shocks

S2 income shocks + nonbank credit supply shocks

S3 income shocks + bank credit supply shocks

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$S2 - S1 \rightarrow$  contribution of **Nonbank** credit supply shocks



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**S1 only income shocks**

S2 income shocks + nonbank credit supply shocks

**S3 income shocks + bank credit supply shocks**

S3 - S1 → contribution of **Bank** credit supply shocks

## Preview of Findings

- the estimated model generates **21%** decline in auto sales
  - very close to 22 % actual decline documented in Johnson et al. (2014).
- contribution of **nonbank** shock: **37%**
  - close to 33% in Benmelech et al. (2017)
  - Policy: Term Asset-backed securities Loan Facility (TALF)
- contribution of **bank** shock: **merely 0.28%**
- bank v.s. nonbank shocks: different distributional implications

## Relation to the Literature

- Dynamics of Durable Expenditures  
e.g. Mankiw (1982) Bernanke (1985) Eberly (1994) Leahy and Zeira (2005) Berger and Vavra(2015) Guerrieri and Lorenzoni (2017)
- Role of Secondary Markets of Durable Goods  
e.g. Chen et al (2013), Gavazza et al (2014), Oh (2019)
- Aggregate Dynamics of Automobile Sales  
e.g. Attanasio (2000), Adda and Cooper (2006), Dupor et al. (2018)

Despite the richness of auto financing, not enough attention paid to

- Consumer credit and Auto Purchase

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Despite the richness of auto financing, not enough attention paid to

- ❖ Consumer credit and Auto Purchase
  - ❖ Benmelech et al. (2017) :the illiquidity of nonbank lenders contributed to  $\frac{1}{3}$  of the decline in auto sales: **no structural model, missing distributional implications**
  - ❖ Gavazza and Lanteri (2020) provide a model to study how consumers respond to credit tightening shocks: **the distinction between banks and nonbanks is missing**

# Facts

## Facts



auto purchasing



auto loan

## Auto Purchasing and the Loan Market during GR

### ❏ substitution from new to used cars

$$❏ \frac{\text{new car sales}}{\text{total sales}} 32\% \xrightarrow{\downarrow 25\%} 24\%$$

### ❏ less replacement with new cars

$$❏ \% \text{ hh replacing used car with new } 7\% \xrightarrow{\downarrow 43\%} 4\%$$

### ❏ more liquidation and retention

$$❏ \% \text{ hh liquidation } 7\% \xrightarrow{\uparrow 21\%} 8.5\%$$

$$❏ \% \text{ hh retention 4-15 yrs old car } 40\% \xrightarrow{\uparrow 6pp} 46\%$$

Calculated from CEX

### ❏ nonbank market share **declined** during GR

### ❏ nonbank loans to the subprime group **dropped** dramatically during the Great Recession: shift to safer borrowers

Calculated from Equifax Data [Table](#)

## Stylized Facts in the Auto Loan Market

November 2006

Rating	Nonbank					Bank				
	w	R%	FICO	y	b	w	R%	FICO	y	b
Exceptional	8%	6.0	813	55	19.6	12%	6.9	813	54	17.5
Very Good	15%	6.1	771	57	23.9	24%	7.1	771	55	20.1
Good	21%	7.6	702	46	26.1	31%	7.9	704	45	19.8
Fair	34%	11.2	626	37	26.6	25%	10.1	633	36	18.4
Poor	22%	14.2	526	30	22.8	7%	12.9	537	31	16.0
Average		10.0	656	41.6	24.6		8.5	703	45.1	19.0

y:annual income, b:loan amount in 10k dollars, R:loan rate, w:fraction of each group within bank/nonbank borrower

Source: [Equifax](#) Archive.



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1. Nonbank borrowers **riskier** than banks: average nonbank loan rate  $>$  bank loan rate

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y:annual income, b:loan amount in 10k dollars, R:loan rate, w:fraction of each group within bank/nonbank borrower

- within group, nonbank and bank borrowers similar income level
- $\text{corr}(y, \text{FICO}) = 0.4712$

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y:annual income, b:loan amount in 10k dollars, R:loan rate, w:fraction of each group within bank/nonbank borrower

4. Nonbank borrowers: **higher** loan amount (higher  $\frac{b}{y}$ )

## How do Loan Rates Depend on Ind. Characteristics

$$R^B = x\beta^B + u^B \quad (1)$$

$$R^N = x\beta^N + u^N \quad (2)$$

$$y_1 = \mathbb{I}(R^B < R^N) \quad (3)$$

Table 1: the Poor: Selected Results

	(1) Probit	(2) Bank	(3) Nonbank
y	0.009*** (0.002)	-0.515* (0.216)	-0.195*** (0.017)
b	-0.009*** (0.001)	0.375 (0.229)	-0.009 (0.006)
FICO	0.004*** (0.000)	-0.228* (0.099)	-0.074*** (0.007)
N	6810	1874	4756

Standard errors in parentheses \* ( $p < .10$ ), \*\* ( $p < .05$ ), \*\*\* ( $p < .01$ )

In addition to  $y$ ,  $b$  and FICO,  $X$  includes: length of loan contract, bank pct (county), cash pct (county), constant

# Model

## Environment

Time is discrete and infinite.

Agents in the model:

1. Households
2. Auto lender: banks and nonbanks
3. New car producer

Clearing markets:

1. auto loan market: perfect competition
2. car market:

Car Age	Quality	Supply	Demand
$\leq 4$ yrs old	high(H)	hh + producer	hh
4 – 14 yrs old	middle(M)	hh	hh
$> 15$ yrs old	low(L)	hh	hh

## Environment: Households

1. Observe state  $s_t$ ,  $h_t$  and  $\epsilon_t$

$$s_t \equiv (e_t, d_t, l_t)$$

$e_t$  idio. earning shock,

$d_t \in \{0, H, M, L\}$  car ownership ,

$l_t$  level of net wealth

$h_t \in \{0, 1\}$  default record

$\epsilon_t$  EV1 shock

## Environment: Households

1. Observe state  $s_t$ ,  $h_t$  and  $\epsilon_t$
2. For each car choice  $\hat{d}$   
→ If can borrow ( $h_t = 0$ , no default record):  
Default or Repay

- ❑ **default:** current debt clears, car taken away, default record  $h_{t+1} = 1$
- ❑ **Repay:** choose  $l_{t+1}$  and financial institution

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→ If cannot borrow ( $h_t = 1$ , with default record): choose saving amount  $l_{t+1} > 0$

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3. car choice prob formed
4. (flag goes away next period w.p.  $\lambda$  for  $h = 1$ )

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## Environment: Financing Choices

### Borrowing

Competitive Financial market: bank (B), nonbank (NB)

- Date  $t$ , Lend  $q^I(l', s)l'$ ,  $I \in \{B, NB\}$  at cost  $r^I$   $r^B < r^{NB}$
- Date  $t+1$ , receive  $l'$  if repay,  $\theta^I P_d l'$  if default  $\theta^B < \theta^{NB}$

### Saving

through a safe bond at risk free  $r_f$ .

## Flag 1 households: with default record

$$V_1(e, l, d, \epsilon) = \max_{\hat{d} \in \{0, H, M, L\}} \{v_1(e, l, d, \hat{d}) + \sigma_\epsilon \epsilon(\hat{d})\} \quad (4)$$

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where

$$v_1(e, l, d, \hat{d}) = \max_{l' \geq 0} u(c^1, \hat{d} + \xi) + \beta \mathbb{E}_{e', d' | e, \hat{d}} \left\{ \lambda \underbrace{EV^1(e', l', d')}_{\text{Value if flag remains}} \right. \\ \left. + (1 - \lambda) \underbrace{EV^0(e', l', d')}_{\text{value if flag disappears}} \right\} \quad (5)$$

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$$c^1 = l + (1 - \Gamma)e - \underbrace{\frac{1}{r_f} l'}_{\text{saving}} + P_d - P_{\hat{d}} - \underbrace{\kappa(d, \hat{d})}_{\text{trans. cost}} \quad (6)$$

$\Gamma$  : loss of income due to credit flag

$EV^i(e', l', d') \equiv E_{e'} V_i(e', l', d', \epsilon')$

$\kappa(d, \hat{d}) = \begin{cases} \lambda_1 P_d + \lambda_0, & \text{if } d \neq 0 \text{ and } \hat{d} \neq d \\ 0, & \text{Otherwise} \end{cases}$

## Flag 0 households: no default record

$$V_0(e, l, d, \epsilon) = \max_{\hat{d} \in \{0, H, M, L\}} \{v_0(e, l, d, \hat{d}) + \sigma_\epsilon \epsilon(\hat{d})\} \quad (7)$$

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$$v_0(e, l, d, \hat{d}) = \max\{v_0^B(e, l, d, \hat{d}), v_0^{NB}(e, l, d, \hat{d}), v_0^S(e, l, d, \hat{d}), v_0^{def}(e, l, d, \hat{d})\}$$



## Flag 0 households: no default record

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$$\forall I \in \{B, NB\}$$

$$v_0^I(e, l, d, \hat{d}) = \max_{l' < 0} u(c^I, \hat{d} + \xi) + \beta \mathbb{E}_{e', d' | e, \hat{d}} EV^0(e', l', d') \quad (8)$$

$$c^I = l + e - q^I(l', s) l' + P_d - P_{\hat{d}} - \underbrace{\kappa(d, \hat{d})}_{\text{trans. cost}}$$

$$EV^i(e', l', d') \equiv E_{\epsilon'} V_i(e', l', d', \epsilon')$$

## Flag 0 households: no default record

$$V_0(e, l, d, \epsilon) = \max_{\hat{d} \in \{0, H, M, L\}} \{v_0(e, l, d, \hat{d}) + \sigma_\epsilon \epsilon(\hat{d})\} \quad (7)$$

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$$v_0^S(e, l, d, \hat{d}) = \max_{l' \geq 0} u(c^S, \hat{d} + \xi) + \beta \mathbb{E}_{e', d' | e, \hat{d}} EV^0(e', l', d') \quad (8)$$

$$c^S = l + e - \frac{1}{r_f} l' + P_d - P_{\hat{d}} - \underbrace{\kappa(d, \hat{d})}_{\text{trans. cost}}$$

$$EV^i(e', l', d') \equiv E_{\epsilon'} V_i(e', l', d', \epsilon')$$

## Flag 0 households: no default record

$$V_0(e, l, d, \epsilon) = \max_{\hat{d} \in \{0, H, M, L\}} \{v_0(e, l, d, \hat{d}) + \sigma_\epsilon \epsilon(\hat{d})\} \quad (7)$$

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$$v_0^{def}(e, l, d, \hat{d}) = u(c^{def}, 0 + \xi) + \beta \mathbb{E}_{e', d' | e, 0} EV^1(e', 0, d') \quad (8)$$

$$c^{def} = (1 - \Gamma)e - P_{\hat{d}}$$

$\Gamma$  : loss of income due to default      $EV^i(e', l', d') \equiv E_{\epsilon'} V_i(e', l', d', \epsilon')$

## Financial Institutions

Competitive Financial market:  $I \in \{ \text{bank (B), nonbank (NB)} \}$

❖ Date  $t$ , Lend  $q^I(l', s)l'$  at cost  $r^I$   $r^B < r^{NB}$

❖ Date  $t+1$ , receive  $l'$  if repay,  $\theta^I P_{d'}$  if default  $\theta^B < \theta^{NB}$

**Loan Contract, for  $l' < 0$ ,**

$$\underbrace{q^I(l', s)l' r^I}_{\text{total fund cost}} = \underbrace{\mathbb{E}_{s'|s}(1 - \mathbb{D}(s')) l'}_{\text{repay}} + \underbrace{\mathbb{E}_{s'|s}\{\mathbb{D}(s') \theta^I P_{d'}\}}_{\text{value covered from repossessed car}}, \quad (9)$$

$\mathbb{D}(s') \equiv \mathbb{E}_\epsilon \mathbb{I}(s' \in \Psi)$ ,  $\Psi$  the default set

## Auto Market Clearing

- + New Car Producers: perfect competition, linear technology,  $P_H = mc$ . New production  $x$

$$\forall \tilde{d} \in \{H, M, L\},$$

$$\underbrace{\int \mathbb{I}(g_d(s) = \tilde{d}, d \neq \tilde{d}) \mu(s) ds}_{\text{demand of } \tilde{d} \text{ cars}} = \underbrace{\int \mathbb{I}(g_d(s) \neq \tilde{d}, d = \tilde{d}) \mu(s) ds}_{\text{supply of } \tilde{d} \text{ cars}} + x_{\hat{d}} \quad (10)$$

$$x_{\hat{d}} = \begin{cases} x, & \text{if } \hat{d} = H \\ 0, & \text{Otherwise} \end{cases}$$

$g_d(s)$  policy function of car choice for  $s$  household

$\mu(s)$  measure of  $s$  households

## A Recursive Stationary Competitive Equilibrium

is (i) a value function  $V(\mathbf{s}, h) \equiv EV_h(\mathbf{s})$  and associated policy functions  $\hat{d} = g_d(\mathbf{s}, \epsilon)$ ,  $l' = g_l(\mathbf{s}, \epsilon)$ ,  $I_{def} = g_{def}(\mathbf{s}, \epsilon)$ ,  $I_{fi} = g_{fi}(\mathbf{s}, \epsilon)$  (ii) a stationary distribution  $\mu^*(e, l, d, h)$ , (iii) a vector of prices  $\mathbf{P}^* \equiv (P_M^*, P_L^*)$ , and (iv) loan rate schedules  $q^I(l', \mathbf{s})$ ,  $I \in \{N, NB\}$  such that

- Individual Optimization:**  $V(\mathbf{s}, h)$  satisfies (7) and (4) with policy functions  $\hat{d} = g_d(\mathbf{s}, \epsilon)$ ,  $l' = g_l(\mathbf{s}, \epsilon)$ ,  $I_{def} = g_{def}(\mathbf{s}, \epsilon)$ ,  $I_{fi} = g_{fi}(\mathbf{s}, \epsilon)$
- Consistency of Loan Rates:** the loan rate schedules  $q^I(l', \mathbf{s})$  satisfy (9), where the lenders' perceived default set is consistent with households policy function:  $\tilde{g}_{def}(s, \epsilon) = g_{def}(s, \epsilon)$
- Stationarity and Consistency of Beliefs**  $\mu^*(e, l, d, h)$  is consistent with exogenous processes and policy functions  $l' = g_l(s, \epsilon)$ ,  $I_{def} = g_{def}(s, \epsilon)$ ,  $I_{fi} = g_{fi}(s, \epsilon)$ :  $\mu^*(e, l, d, h) = H(\mu^*(e, l, d, h))$
- Car markets clear:** (10) determines the flow  $x$  of production of new cars and holds for  $\hat{d} \in \{M, L\}$ .

# Estimation

## Taking the Model to the Data

- Income Processes: high v.s. low education groups estimated from PSID *a la* Guvenen (2007) [detail](#)
- parameters calibrated outside of the model [detail](#)
- parameters estimated in the structural model

Utility Function:

$$u(c, d) = \begin{cases} \log(c^\alpha d^{1-\alpha}) & \text{if } \gamma = 1 \\ \frac{\{c^\alpha d^{1-\alpha}\}^{1-\gamma} - 1}{1-\gamma} & \text{if } \gamma \geq 0, \gamma \neq 1 \end{cases} \quad (11)$$



## Calibrated Parameters

$\sigma_\epsilon$	0.0500	scale of taste shock $\epsilon$	literature
$\lambda$	0.9000	Prob. default record re- remaining	Chp 7
$p_H$	$\frac{1}{3}$	prob. $H \rightarrow M$	Gavazza and Lanteri (2020)
$p_M$	0.1000	prob. $M \rightarrow L$	Gavazza and Lanteri (2020)
$p_L$	0.5000	prob. $L \rightarrow 0$	Gavazza and Lanteri (2020)
$\delta$	0.0300	prob. car loss	CEX
$P_H$	\$21675	price of H car	VA regist. record
$r^B$	1.0600	bank, fund cost	equifax data
$r^{NB}$	1.0624	nonbank, fund cost	equifax data
$r_f$	1.0200	risk free rate	literature
$\bar{\theta}$	0.98	nonbank recovery	repo cost \$200

$$\Pi_{i,j} = \begin{pmatrix} 1 - p_H - \delta & p_H & 0 & \delta \\ 0 & 1 - p_M - \delta & p_M & \delta \\ 0 & 0 & 1 - p_L & p_L \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

## Calibrated Parameters

$\beta_h$	0.9450	discount factor, hi edu	Gavazza and Lanteri (2020)
$\gamma$	1	risk aversion	literature
$d_H$	1	util from H car	normalization

- ❏ moments less responsive to these parms compared to estimated ones
- ❏ later robustness check [table](#)

[back](#)

## Estimation

$\Theta \equiv (\beta_l, \alpha, \Gamma, d_M, d_L, \xi, \underline{\theta}, P_H, P_L)$  solves:

$$\Upsilon = \min_{\Theta} (M^s(\Theta) - M^d)'W(M^s(\Theta) - M^d) \quad (12)$$

Moments (match pre-crisis):

- Financial: wealth-to-income, debt-to-income Ratio
- Car stock: fraction of  $\tilde{d}$  car owners, car ownership rates
- loan rate: dependence of loan rate on  $y$  by lender type
- delinquency: fraction of loans flowing into delinquency
- nonbank market share

Full Table of Moments

## Selected Moments

	Data	Model
nonbank share	0.4433	0.4206
Med wi ratio	0.0045	0
% hh indebt	0.3479	0.3740
% default	0.0196	0.0221
Med b/y ratio, all	0.2870	0.2894
Med b/y ratio, B	0.2053	0.1549
Med b/y ratio, N	0.2517	0.5061
B, coef y	-0.7366	-0.1202
N, coef y	-0.0941	-0.0369

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## Estimation Results

Table 2: Estimated Parameters

Estimated Parameters		
$\beta_l$	0.8965	discount factor, low education group
$\alpha$	0.9520	weight of non-durable consumption in the utility function
$\Gamma$	0.0007	fraction of income loss due to default or default record
$d_M$	0.6869	utility flow from middle quality car
$d_L$	0.3060	utility flow from low quality car
$\xi$	0.0139	utility flow from alternative ways of transportation
$P_M$	0.2652	the price of a middle quality car
$P_L$	0.0976	the price a of low quality car
$\theta$	0.5387	bank fraction of recovery from foreclosure

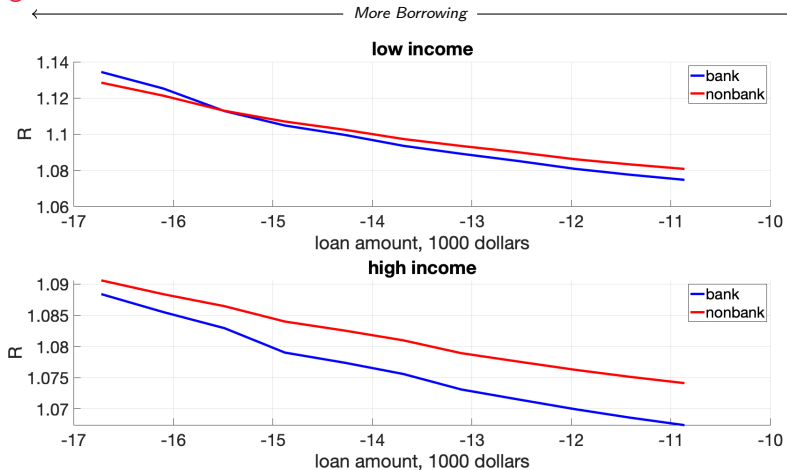
Identification

# Model Implications

## Model Implications

- Loanrate Schedules
- Lender Choices
- Evaluation of credit supply shocks

## Endogenous Loan Rates

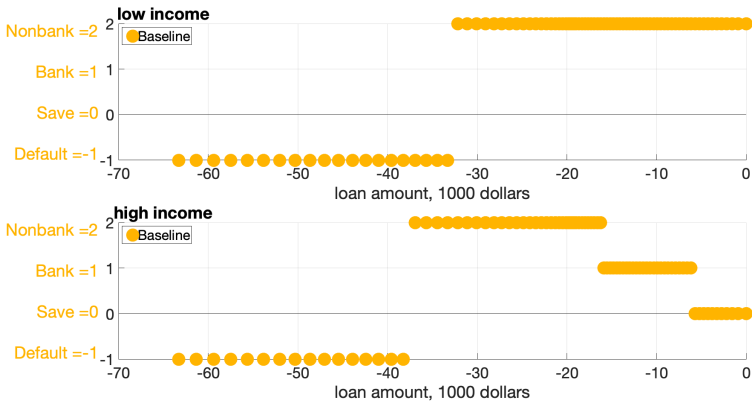


both B and NB lower loan rate for higher income today

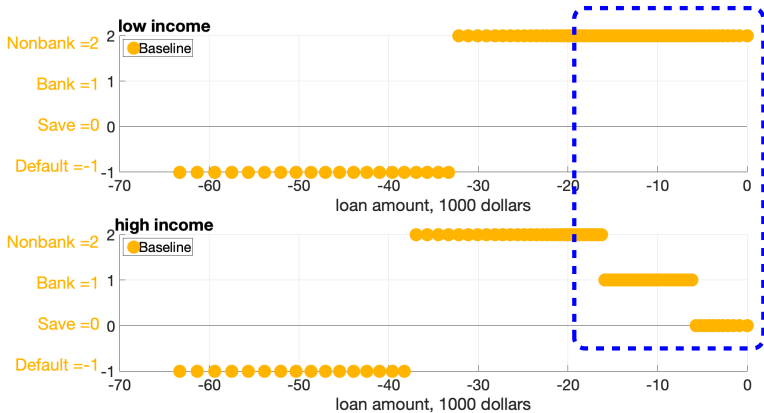
bank needs more compensation for risk

## Choice of Lenders

HH with same  $d$ ,  $\hat{d} = M$  in ss. eqm:



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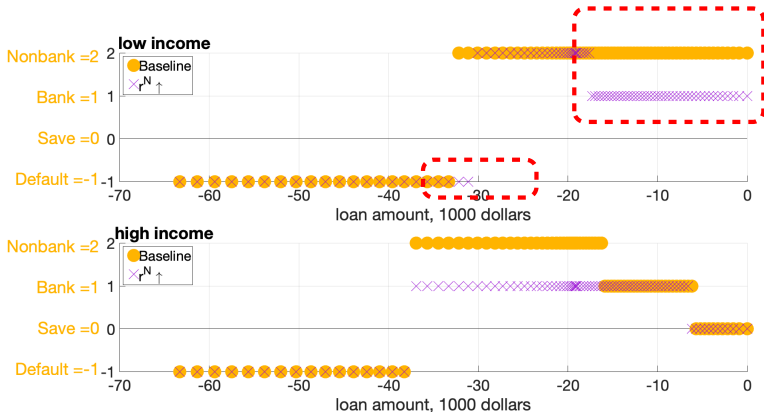
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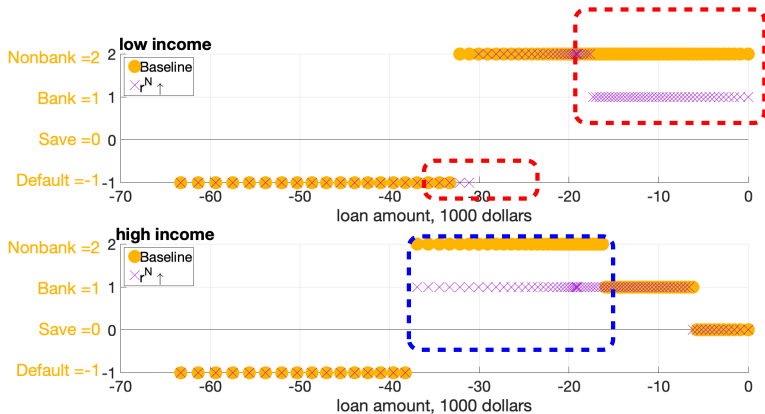
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## Credit Supply Shocks

Table 3: Timeline, Shocks and  $\mu$

t	0	1	2	3	...
shocks	no	yes	no	no	...
hh exp. $P'$	$P^*$	$P^*$	$P^*$	$P^*$	..
actual P	$P(\mu^*)$	$P(\mu_1)$	$P(\mu_2)$	$P(\mu_3)$	..

S1 income shocks

S2 income shocks + nonbank credit supply shocks

S3 income shocks + bank credit supply shocks

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**S1 income shocks**

**S2 income shocks + nonbank credit supply shocks**

S3 income shocks + bank credit supply shocks

$S2 - S1 \rightarrow$  contribution of **Nonbank** credit supply shocks

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### S1 income shocks

S2 income shocks + nonbank credit supply shocks

S3 income shocks + **bank** credit supply shocks

$S3 - S1 \rightarrow$  contribution of **Bank** credit supply shocks

## Contribution of Credit Supply Shocks

Table 4: Percentage Changes in Auto Sales and Prices

	S1	S2	S3
	income only	income & nonbank	income & bank
Sales (H)	-13.21% (0.336%)	<b>-20.88%</b> (0.297%)	-13.25% (0.353%)
Price M	-1.10 % (0.085%)	-2.97% (0.084%)	-1.08 % (0.085%)
Price L	-3.25 % (0.254%)	-4.08% (0.262%)	-3.24 % (0.249%)
Contribution		<b>nonbank</b> <b>36.74%</b>	<b>bank</b> <b>0.28%</b>

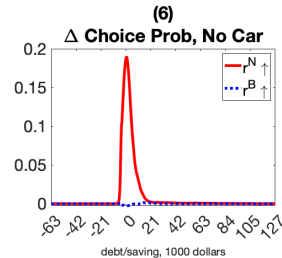
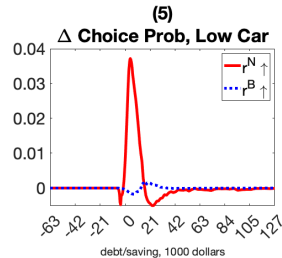
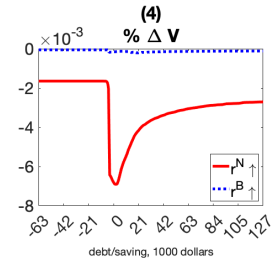
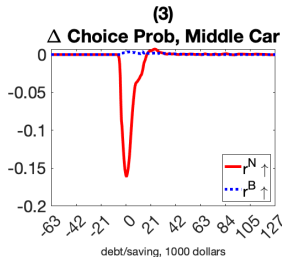
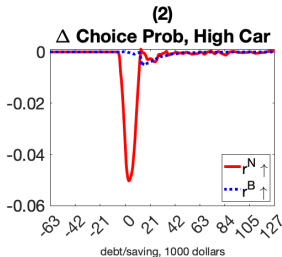
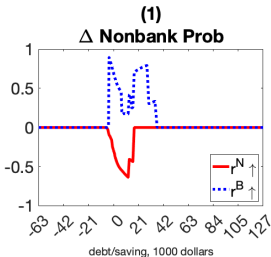
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- inaction↑: 69.2% s.s. → 73.5% S2
- substitution↓: replacement with new purchase 8.3% s.s. → 6.9% S2
- liquidation↑: 15.4% more hh disposed M cars

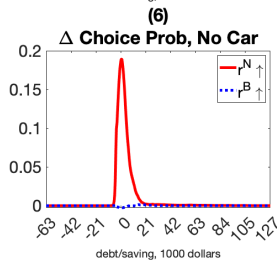
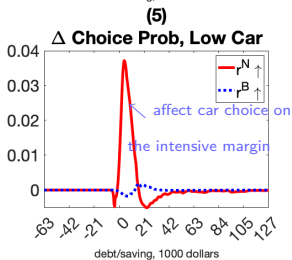
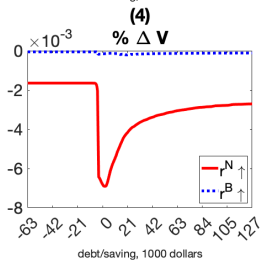
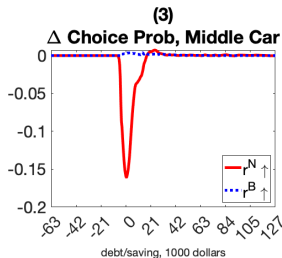
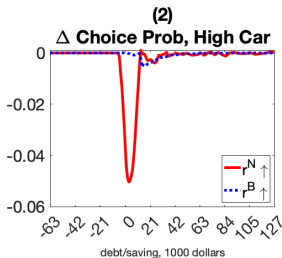
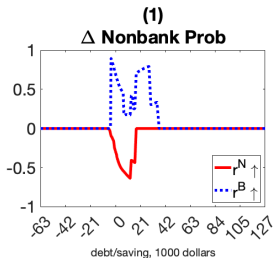
## Distributional Effects: bank v.s. nonbank shock only



low income hh with no car today

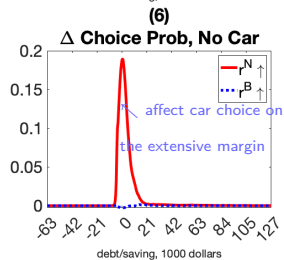
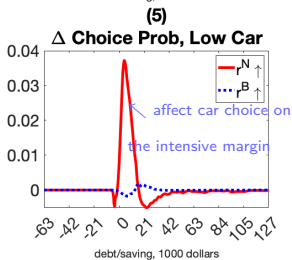
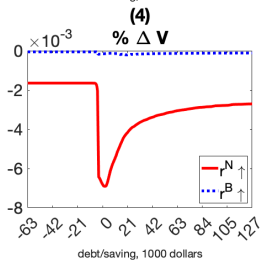
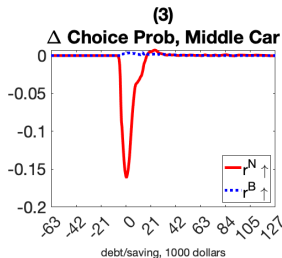
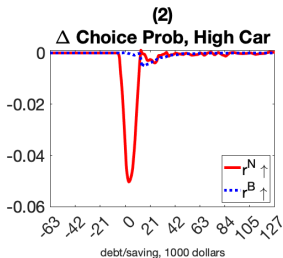
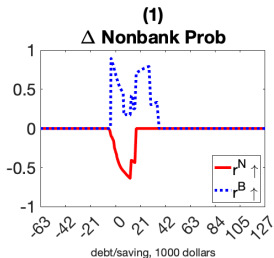


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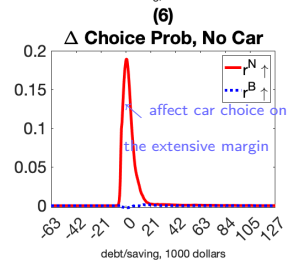
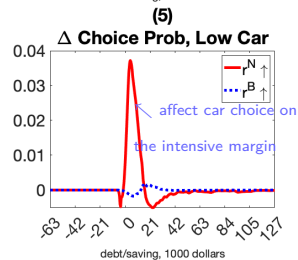
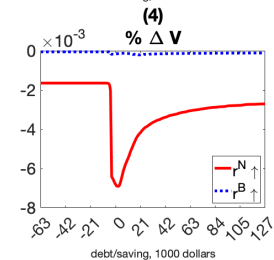
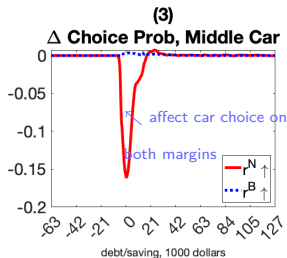
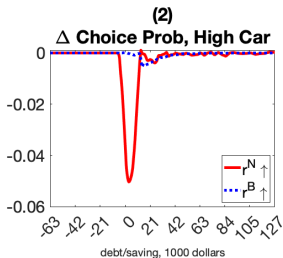
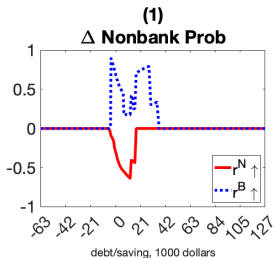
low income hh with no car today

## Distributional Effects: bank v.s. nonbank shock only



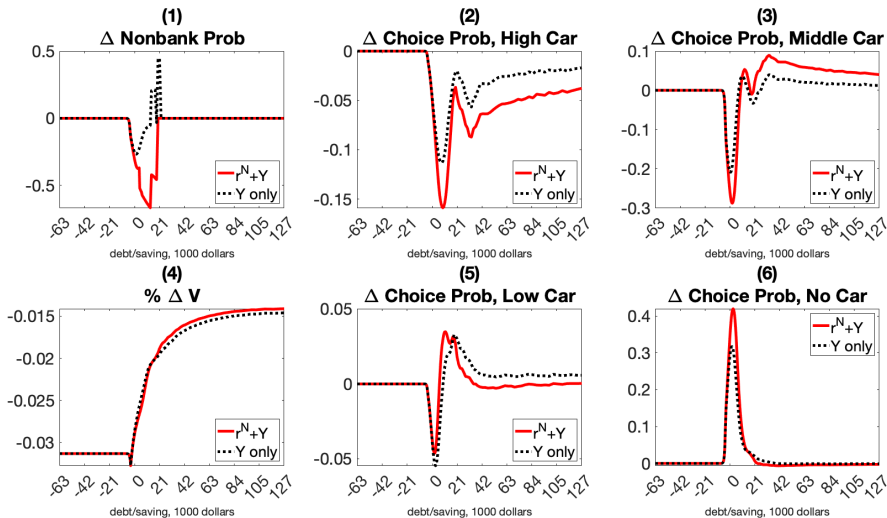
low income hh with no car today

## Distributional Effects: bank v.s. nonbank shock only



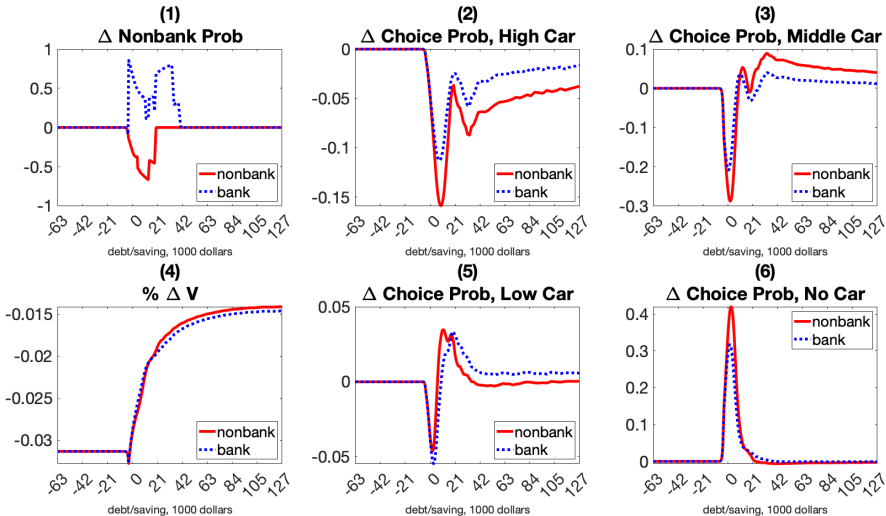
low income hh with no car today

## Distributional Effects: bigger real effects with nonbank shock



low income hh with no car

# Distributional Effects: trivial real effects with bank shock



low income hh with no car

## Conclusions

**Facts:** rich heterogeneity between bank v.s. nonbank borrowers

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**Equilibrium model** with hetero. hh and lenders

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**Equilibrium model** with hetero. hh and lenders

- asymmetric ability to borrow
- the estimated model generates **21%** decline in auto sales
  - very close to 22 % actual decline documented in Johnson et al. (2014).



## Conclusions

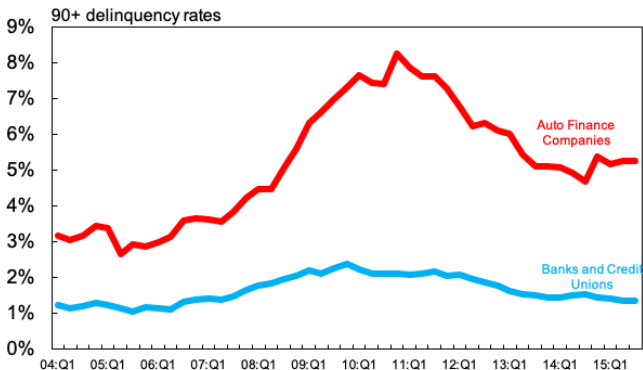
**Facts:** rich heterogeneity between bank v.s. nonbank borrowers  
**Equilibrium model** with hetero. hh and lenders

- ❖ asymmetric ability to borrow
- ❖ the estimated model generates **21%** decline in auto sales
  - ❖ very close to 22 % actual decline documented in Johnson et al. (2014).
- ❖ contribution of **nonbank** shock: **37%**
  - ❖ close to 33% in Benmelech et al. (2017)
  - ❖ Policy: Term Asset-backed securities Loan Facility (TALF)
- ❖ contribution of **bank** shock: **merely 0.28%**
- ❖ bank v.s. nonbank shocks: different distributional implications

# Appendix

# Auto Finance Co. v.s. Banks

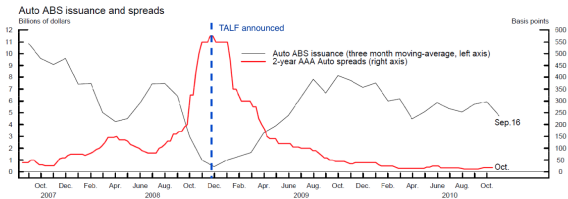
Figure 3: Flow into 90+ delin



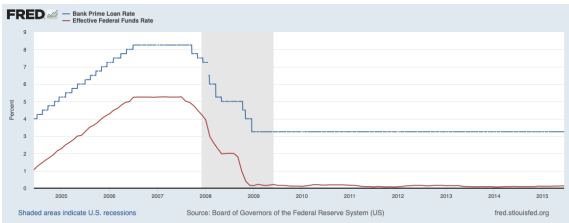
Source: New York Fed Consumer Credit Panel / Equifax

## Costs Bank v.s. Nonbank

**Figure 4:** Auto ABS spreads; Bank Prime Loan Rates and Federal Funds Rates



Source: JPMorgan Chase for spreads and Bloomberg for issuance.



## Not just a Demand Side Story

Despite the relatively **stable demand** for auto loans:

- ❏ Survey of Consumer Finance 07-09 panel
- ❏ "What type of credit did you apply for?"
- ❏ 2007: all 19.67% vehicle loan 4.36% (in the past 5 years)
- ❏ 2009: all 14.73% vehicle loan 3.48% (in the past 2 years)

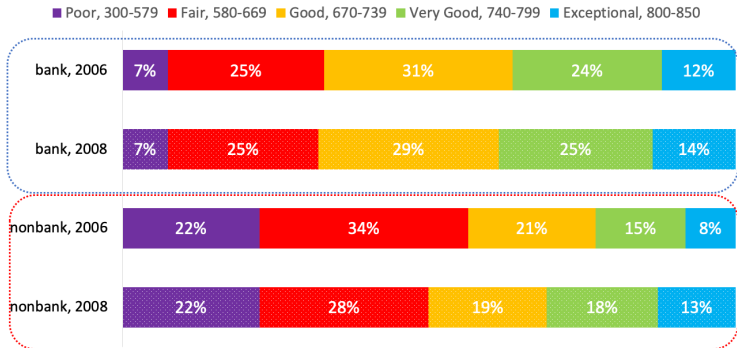
Nonbank auto loan originated to the subprime **drop dramatically**

**Table 5:** % Change in Total Amount of Loan Originated

	Nonbank	Bank
Exceptional	35%	8%
Very Good	7%	-1%
Good	-28%	-11%
Fair	-47%	-15%
Poor	-45%	-15%

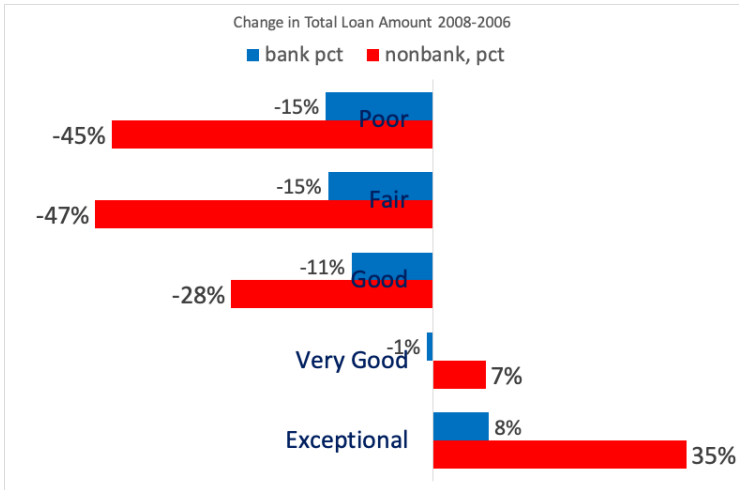
## Nonbank Credit Shrinkage to the Subprime

Figure 5: Composition of Borrower by Lender and Time



1. Cross Sectional: more Fair and Poor in Nonbanks
2. Over Time: obvious shift from riskier to safer for Nonbank
3. Market Share of Nonbank :

# consumers: 44.3% → 41% ;  
 \$ loan origination: 51% → 45%

[Back](#)

## Data Description

- ❖ 2 periods from Equifax Archive: 2006 November and 2008 November
- ❖ Each period, 50k individuals are randomly drawn from records if she opened an auto loan within 30 Days
- ❖ City, State, Zip code, Lender industry code, Loan amount, Loan terms, Loan rate, Income and Fico

Back



## Income Process

$$y_t^i = \bar{y}_t^i + \tilde{y}_t^i$$

$$\bar{y}_t^i = \gamma_0 + \gamma_{11}a_t^i + \gamma_{12}a_t^{i,2} + \gamma_2t + \gamma_3X_i$$

$$\tilde{y}_t^i = z_t^i + \eta_t^i \quad (13)$$

$$z_t^i = \rho_z z_{t-1}^i + \epsilon_{z,t}^i \quad (14)$$

$\delta_{eu}$	0.1453	probability of being separated from current job
$\delta_{ue}$	0.9683	probability of finding a job computed
$\rho_z^h$	0.8865	persistence parameter of the permanent shock, high education group
$\rho_z^l$	0.8681	persistence parameter of the permanent shock, low education group
$\sigma_{\tilde{z}}^h$	0.1784	std parameter of the innovation to the permanent shock, high education group
$\sigma_{\tilde{z}}^l$	0.1662	std parameter of the innovation to the permanent shock, low education group
$\sigma_{\eta}^h$	0.0615	std parameter of the transitory shock, high education group
$\sigma_{\eta}^l$	0.0590	std parameter of the transitory shock, low education group

## Model Fit

Table 6: Moments: Model v.s. Data

	All		High Edu		Low Edu	
	data	model	data	model	data	model
$w_i$	0.0045	0	0.0650	0.1475	0	0
$f_H$	0.3201	0.1704	0.3964	0.2739	0.2849	0.1211
$f_M$	0.6069	0.6916	0.5641	0.6058	0.6267	0.7325
$f_L$	0.0730	0.1380	0.0395	0.1203	0.0884	0.1464
$f_0$	0.1091	0.1714	0.0649	0.1097	0.0649	0.1982
$wi_{p10}$	-0.6569	-0.4884	-0.6784	-0.2197	-0.6424	-0.5420
$wi_{p25}$	-0.1777	-0.1557	-0.1894	-0.0447	-0.1727	-0.2487
$wi_{p75}$	0.1812	0.3334	0.5593	0.5109	0.0685	0.2271
$f_{l<0}$	0.3479	0.3740	0.3501	0.2907	0.3465	0.4097
$\bar{d}_i^m$	0.2870	0.2894	0.2570	0.2074	0.3326	0.3338



## Percentage Changes in Auto Sales and Contribution of Nonbank Shocks

	S2 % H Sales Benchmark	Contribution	
		-20.88 %	36.74%
<b>Calibrated Parameters</b>			
$\beta_h = 0.898$	0.945	-22.51%	32.67%
$\gamma = 1.500$	1	-19.24%	37.37%
$d_H = 0.900$	1	-22.13%	31.49%
$\bar{\theta} = 0.882$	0.98	-20.86%	21.97%
<b>Estimated Parameters</b>			
$\underline{\theta} = 0.5925$	0.5383	-21.54%	33.19%
$\beta_l = 0.9427$	0.8965	-17.22%	26.18%
$\alpha = 0.9055$	0.9520	-10.96%	30.40%
$d_M = 0.6176$	0.6869	-21.63%	32.90%
$d_L = 0.2742$	0.3060	-21.39%	33.33%
$\xi = 0.0151$	0.0139	-21.73%	32.90%
$\Gamma = 7.4e(-3)$	$6.73e(-3)$	-21.65%	34.20%

## The Auto Loan Market pre v.s. during GR

- nonbank market share **declined** during GR
- nonbank loans to the subprime group **dropped** dramatically during the Great Recession: shift to safer borrowers

% total loan amount change

Category (FICO)	Nonbank Share		loan amount $\Delta_{08-06}$	
	2006	2008	Nonbank	Bank
Exceptional (800-850)	35.2%	39.4%	+35%	+8%
Very Good (740-799)	32.3%	33.4%	+7%	-1%
Good (670-739)	35.1%	31.3%	-28%	-11%
Fair (580-669)	51.5%	43.9%	<b>-47%</b>	-15%
Poor (300-579)	70.6%	67.7%	<b>-45%</b>	-15%

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