

# **The Payday Loan Puzzle: A Credit Scoring Explanation**

Tsung-Hsien Li  
University of Mannheim

Jan Sun  
University of Mannheim

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# Introduction

# Motivation

- Payday loans: Unsecured, small amount (\$300), short-term (2 weeks), and high-cost (400%)
- 12 million user and \$50 billion (Stagman, 2007; PEW, 2016)
- Hotly debated regulatory topic
- **Payday loan puzzle** in U.S. (Agarwal et al, 2009):
  - **2/3** of payday loan borrowers have liquidity left on credit cards ( $< 20\%$ )
  - Significant extra monetary costs about **\$200** over a year

# Credit Scoring Explanation

*“Why are people taking out [payday] loans instead of using their cards?” Ranney told me, “This guy was implying that these people weren’t smart enough to make the ‘right’ decision. I laughed in his face. ‘They’re protecting the card!’ I told him. [...]” Whereas failure to repay a payday loan won’t affect a consumer’s credit score, failure to repay a credit card will.*

— Lisa Servon (2017): *The Unbanking of America*

# What We Do

- Reputation protection: **“Using payday loans to protect credit scores!”**
  - Credit scores very important in U.S. [▶ Credit Scores](#)
  - Payday lenders do not report to credit bureaus in U.S. (CFPB, 2017)
- Build a Huggett-type model of two assets, two default options, both hidden information (type scores) and hidden actions
  - Rationalize the puzzling behavior via type score protection
- Use calibrated model to understand payday loan puzzle
- Policy experiments: Quantity caps, full ban

# Intuition — Type Score Protection via Payday Loans

- HHs' types (discount factors)  $\beta$  are unobservable
- Types score: Probability of being patient (good type)
- Types score updated with observable bank loan choice and default
- Credit terms thus condition on type scores
- $\text{Income} \downarrow \implies \text{Borrowing bank loans} \implies \text{Type score} \downarrow (\text{today})$   
 $\implies \text{Fail to repay bank loans} \implies \text{Type score} \downarrow (\text{tomorrow})$
- Payday loans and payday default are unobservable “to banks”
- HHs might use more expensive payday loans to protect type scores

# Key Findings

- Endogenously generate the payday loan puzzle
  - Account for 40% of the puzzle occurrence
  - Match the magnitude of monetary losses
- Restricting the size of or banning payday loans are welfare-reducing
  - Heterogeneity across types

# Literature / Contribution

- **Consumer finance and default:** Chatterjee et al. (2007), Livshits et al. (2007), Chatterjee et al. (2020), Exler (2020), Saldain (2021)  
First to model defaultable bank and payday loans with hidden information and actions
- **Pecuniary mistakes:** Agarwal et al. (2009), Cartel et al. (2011)  
First to endogenously generate and rationalize the payday loan puzzle
- **Payday loan policy debate:** Zinman (2010), Morgan et al. (2012), Skiba and Tobacman (2019), Melzer (2011)  
First to analyze welfare implications of policies in a richer framework



**Model**

# Model Environment

- Time is discrete
- Endowment economy with idiosyncratic shocks
- Incomplete market: Bank assets, payday loans
- Banks, payday lenders, and households
- Households' hidden types (discount factors)  $\implies$  Type scores
- Banks cannot see payday loans

# Households

- Infinitely-lived with survival rate  $\rho$
- Risk-averse, derive utility from consumption  $c$
- Two types of HHs:  $\beta_L$  and  $\beta_H$  (stochastic persistent)
- Receive stochastic earnings  $\mathbf{z}$  (transitory) and  $\mathbf{e}$  (persistent)
- Have bank assets  $\mathbf{b}$ , payday debts  $\mathbf{p}$ , type score  $\mathbf{s}$  (Prob. of  $\beta_H$ )
- Repay or default  $\mathbf{d}$ 
  - Formal default (both), payday default (payday loan only)
  - Filing costs, stigma costs, exclusion in the filing period
- Can borrow/save  $\mathbf{b}'$  in banking sector
- Can borrow  $\mathbf{p}'$  in payday lending sector (if  $\mathbf{b}' \leq 0$ )
- Subject to action-specific utility shocks  $\epsilon \implies \sigma^{(\mathbf{d}, \mathbf{b}', \mathbf{p}')}(\beta, \mathbf{z}, \mathbf{e}, \mathbf{b}, \mathbf{s}, \mathbf{p})$

# Banks and Payday Lenders

- Risk-neutral
- Different information set
  - Banks cannot observe payday variables ( $p$ ,  $p'$ , and  $PD$ )
- Different operating costs:  $r_p \gg r_f$
- Different default probabilities
  - Banks: Formal default
  - Payday lenders: Formal default, payday default
- Both can't see  $z$  (i.i.d.) and  $\beta$  (persistent)  $\rightarrow$  Type score  $s$
- Perfect competition: Risk-based discount loan prices  $q_b$  and  $q_p$

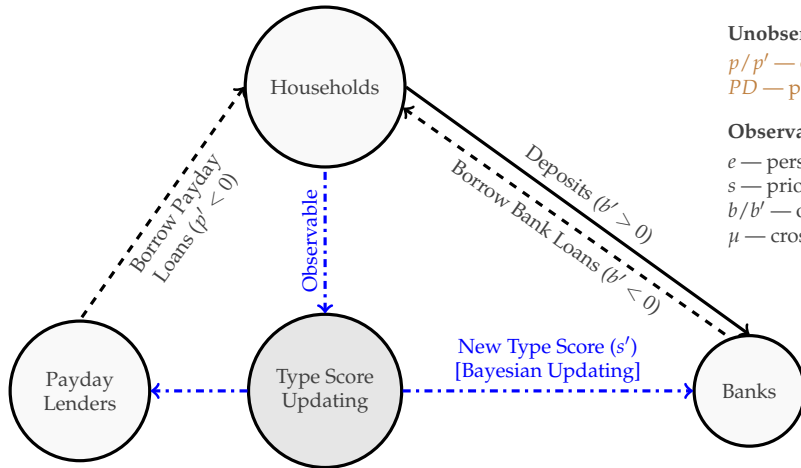
► Bank Problem

► Bank Loan Price Schedule

► Payday Problem

► Payday Loan Price Schedule

# Information Structure



**Unobservable to Both**

$\beta$  — type (discount factor)

$z$  — transitory earnings

**Unobservable to Banks only**

$p/p'$  — old/new payday loans

$PD$  — payday default

**Observable to Both**

$e$  — persistent earnings

$s$  — prior type score

$b/b'$  — old/new bank assets

$\mu$  — cross-sectional distribution

# Type Score Updating

- Bank-observable choice probabilities  $\omega_b \equiv (e, b, s)$ :

$$\sigma^{(d,b',p')}(\beta, z, \omega_b, \mathbf{p}) \xrightarrow{\tilde{\mu}(\mathbf{p})} \sigma^{(d,b',p')}(\beta, z, \omega_b) \xrightarrow{p', \tilde{d}=RVPD} \tilde{\sigma}_b^{(\tilde{d},b')}(\beta, z, \omega_b)$$

- Type score  $s$  (Prob. of  $\beta_H$ ) updated via Bayes rule:

$$\underbrace{s'(\beta_H)}_{\text{posterior}} = \underbrace{Q_{H \rightarrow H}^\beta}_{\text{transition}} \cdot \frac{\overbrace{\tilde{\sigma}_b^{(\tilde{d},b')}(\beta_H)}^{\text{updating}} \cdot \overbrace{s(\beta_H)}^{\text{prior}}}{\sum_{\hat{\beta}} \tilde{\sigma}_b^{(\tilde{d},b')}(\hat{\beta}) \cdot s(\hat{\beta})} + Q_{L \rightarrow H}^\beta \cdot \frac{\tilde{\sigma}_b^{(\tilde{d},b')}(\beta_L) \cdot (1 - s(\beta_H))}{\sum_{\hat{\beta}} \tilde{\sigma}_b^{(\tilde{d},b')}(\hat{\beta}) \cdot s(\hat{\beta})}$$

► Rigorous  $\psi$

► Rigorous  $\tilde{\sigma}_b$

► Likelihood Ratio and Type Score

► Stationary Equilibrium

► Grid Specification

# Calibration

# Strategy

- Model period is a year
- Whole population in 2004
- Two sets of parameters
  - Exogenously calibrated
    - ▶ Discount factors from Chatterjee et al. (2020)
    - ▶ Earnings processes from Floden and Linde (2001)
    - ▶ Standard values or direct empirical evidence
  - Internally calibrated to match formal and payday default rates



# Exogenous Calibration

Parameter		Value	Source
Low discount factor	$\beta_L$	0.886	Chatterjee et al. (2020)
High discount factor	$\beta_H$	0.915	Chatterjee et al. (2020)
Transition from low to high	$Q^\beta(\beta_L \beta_H)$	0.013	Chatterjee et al. (2020)
Transition from high to low	$Q^\beta(\beta_H \beta_L)$	0.011	Chatterjee et al. (2020)
Discount factor at birth	$G_\beta$	(0.72,0.28)	Chatterjee et al. (2020)
AR(1) of persistent earnings	$\rho_e$	0.9136	Floden and Linde (2001)
S.D. of persistent earnings	$\sigma_e^2$	0.0426	Floden and Linde (2001)
S.D. of transitory earnings	$\sigma_z^2$	0.0421	Floden and Linde (2001)
Persistent earnings at birth	$G_e$	(1,0,0)	Upward earnings profile
Transitory earnings at birth	$G_z$	(1/3,1/3,1/3)	Upward earnings profile

# Exogenous Calibration (cont.)

Parameter		Value	Source
CRRA	$\gamma$	2	Standard
Survival probability	$\rho$	0.975	40 years
Risk-free rate	$r_f$	0.014	Effective interest rate = 4%
Formal default cost	$\kappa_{FD}$	0.02	Albanesi and Nosal (2020)
Payday default cost	$\kappa_{PD}$	0.002	Montezemolo and Wolff (2015)
Operating cost for payday lenders	$r_p$	1.925	Flannery and Samolyk (2005)
Dispersion of extreme value shocks	$\alpha$	0.005	

# Internal Calibration

Parameter		Value	Target	Data	Model
Formal stigma cost	$\zeta_{FD}$	0.02235	Formal default rate	0.99%	0.99%
Payday stigma cost	$\zeta_{PD}$	0.00704	Payday default rate (cond.)	29.7%	29.7%

Source: ABI, CPS, Skiba and Tobacman (2018)

# Untargeted Moments Aligned with Data

Moment (in %)	Data	Model
<b>Households in Debt</b>		
Fraction of bank loan users	20.9	24.26
Fraction of payday loan users	5.61	9.46
Bank debt-to-earnings (cond. on borr.)	11.75	6.48
<b>Interest Rate</b>		
Avg. interest rate for bank loans	9.26	8.56
Avg. interest rate for payday loans	447.88	410.85

# Pooling and Cross-Subsidization

# Pooling

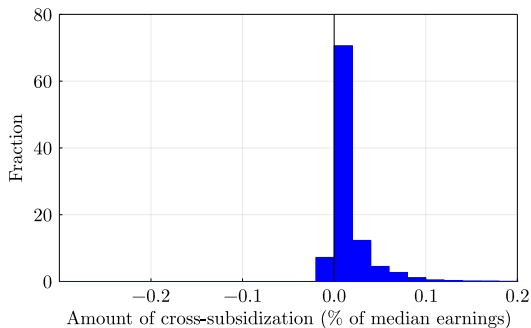
- Banks cannot see types and payday loan choices
- Conditional on the same borrowing of bank loans
  - Impatient  $\rightarrow$  Default  $\uparrow \rightarrow$  **Riskier**
  - Payday loan users  $\rightarrow$  Total debt burden  $\uparrow \rightarrow$  Default  $\uparrow \rightarrow$  **Riskier**
- **Riskier** faces lower bank loan rates than “actuarially fair rates” (FI)
- Cross-subsidization in “bank loan market”

► Formal Default Prob. by Types and Payday Users

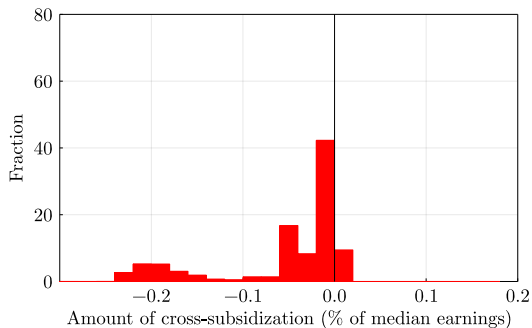
► Equilibrium Outcomes by Types

# Cross-Subs. of Bank Loans across Types

(a) Impatient Households

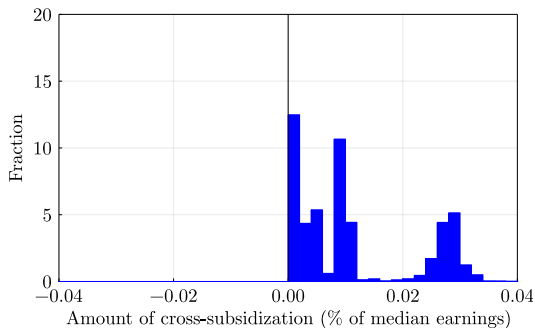


(b) Patient Households

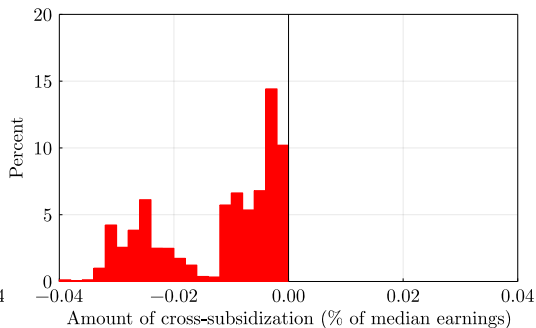


# Cross-Subs. of Bank Loans across Payday Loan Users

(a) Payday Loan Users



(b) Non Payday Loan Users





# Payday Loan Puzzle

# Account for 40% of Puzzle Occurrence

- Payday loan puzzle: Using payday loans before maxing out credit cards
- In data  $\approx 66\%$  (Agarwal et al., 2009)
- Define “**Rate of Puzzle Occurrence**” as:

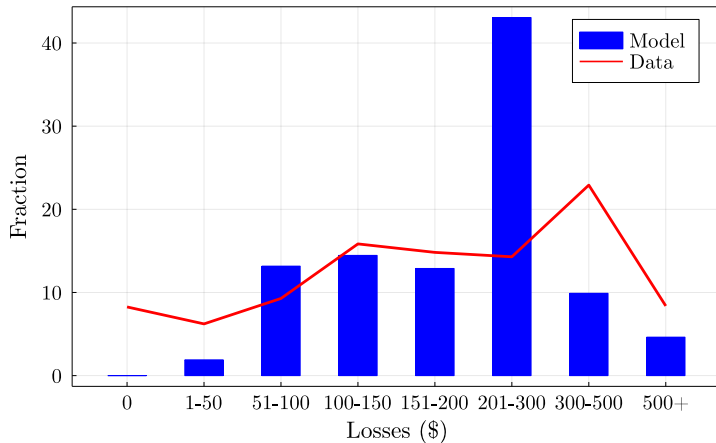
$$\left( \frac{\text{Both loan users making “Seeming Pecuniary Mistake”}}{\text{Both loan users}} \right) \times 100$$

- In model = 26.44%  $\implies$  40% of puzzle occurrence
- Puzzle users: HHs of this puzzling behavior

► Rigorous Puzzle Definition

► Type Score Protection

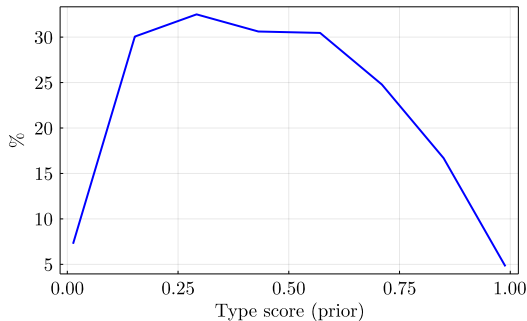
# Match Magnitude of Monetary Losses



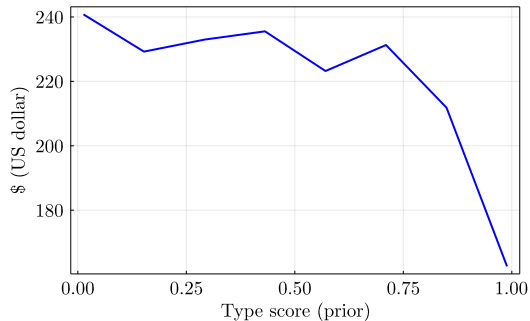
Source: Agarwal et al. (2009)

# Reputation Gain vs. Interest Loss

(a) Reputation Gain



(b) Interest Loss



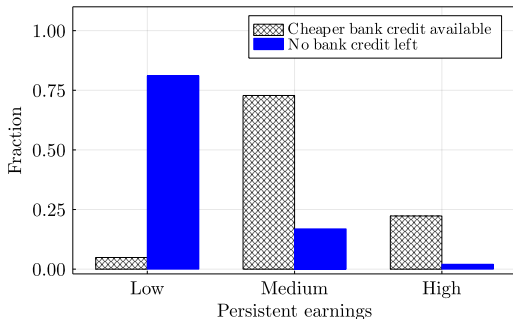
► Type Score Dynamics

► Interest Loss Dynamics

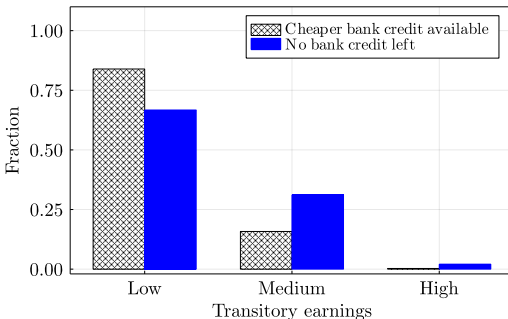
Note: Avg. Losses  $\approx$  \$200 (Agarwal et al., 2009)

# Why? Smooth Out Temporary Shortfall in Earnings

(a) Persistent Earnings (Observable)



(b) Transitory Earnings (Unobservable)



► Payday Loan Users Across Earnings

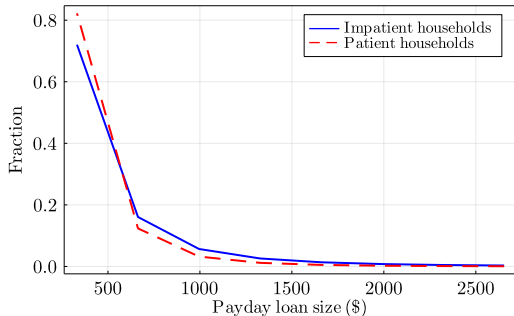
► Type Score Updating Across  $e$

► Fraction of Payday Users Across Income in SCF

# Quantity Caps

# Policy Debate about Payday Lending

- Hotly debated policy topic
- Benchmark:  $p \in [0, \$6000]$
- Two counterfactuals
  - Quantity Cap:  $p \in [0, \$300]$
  - Full Ban:  $p = 0$



# Pooling vs. Insurance of Payday Loans

Variables	Benchmark	Quantity Cap	Full Ban
Welfare (CEV)	–	–0.0012%	–0.0291%
Welfare (CEV) – Impatient	–	–0.0029%	–0.0331%
Welfare (CEV) – Patient	–	0.0013%	–0.0233%
Avg. Cross-Sub. of Bank Loans ( $\beta_L$ )	\$4.57	\$4.09	\$3.88
Avg. Cross-Sub. of Bank Loans ( $p' < 0$ )	\$30.33	\$23.95	–

- Quantity cap → Less pooling → Good: Patient / Bad: Impatient
- Insurance of payday loans → Smoothing bad shocks (e.g., puzzle users)



# Conclusion

# Conclusion

- Payday loan puzzle can be rationalized by “credit scoring protection”
  - Account for 40% of puzzle occurrence
  - Match magnitude of monetary losses
- Puzzle users are not “stupid”—They want to maintain access to credit market in the future via type score protection!
- Restricting the size of payday loans affects (im)patient HHs differently: Impatient, worse off while patient, better off
  - Less cross-subsidization in bank loan market (less pooling)
- Eliminating payday loans is overall welfare-reducing
  - Both types do use payday loans to smooth out bad shocks (insurance)