Modeling the Credit Card Revolution: The Role of IT Reconsidered

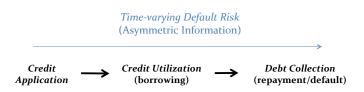
Lukasz A. Drozd¹ Ricardo Serrano-Padial²

¹Wharton School of the University of Pennsylvania

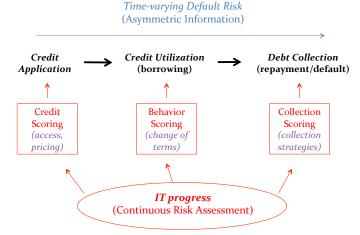
²University of Wisconsin-Madison

April, 2014

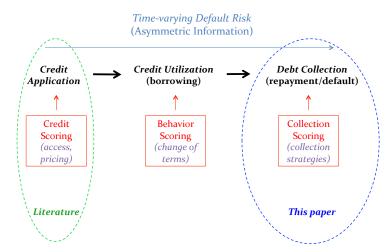
The Role of IT in Credit Markets



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 - o court-based process, truthful revelation of state
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 - o court-based process, truthful revelation of state
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- Conventional approach at odds with data
 - o [1.] most debt discharged informally
 - Dawsey & Ausubel (2004): >50% of \$ defaulted on
 - o [2.] vast resources involved in collection of unpaid debt
 - employment: 350k+ ($\approx 30\%$ share of cc-receivables)

Basic Idea of the Paper

- In the model:
 - Enforcement by the lending industry with access to IT
 - enforcement = Ex post 'State verification' (solvency status)
 - $\mathsf{IT} = \mathsf{signal} \ \mathsf{extraction} \ \mathsf{technology}$

Basic Idea of the Paper

- In the model:
 - o Enforcement by the lending industry with access to IT
 - enforcement = Ex post 'State verification' (solvency status)
 - IT = signal extraction technology

- Comparative Statics Exercise: IT progress
 - o Increase in signal precision (main channel)
 - Reduction in transaction costs

Basic Idea of the Paper



Daily Scoring Directs Core Asset Collections



- Account characteristics
- Demographic characteristics
- ← Supplemental data from external sources
- ← Reponses to collection actions

signal of solvency

Hundreds of thousands of data points are distilled to produce a likelihood of ability to contact and collect

An ROI threshold is established

Actions are taken to the extent warranted



- → Pursue legal collections
- → Send a letter
- → Make a phone call
- → Do nothing

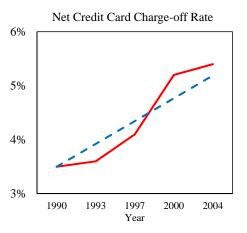
action or no action

PRA, Investor Presentation, 2011 Q3

Portfolio Recovery Associates, Inc.

Preview of Results

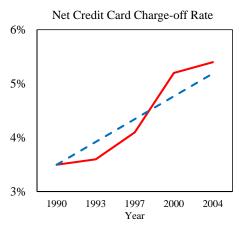
Better enforcement technology implies



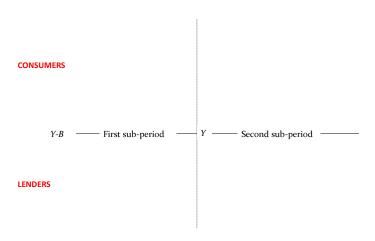
⇒ accounts for most puzzling development in cc-market

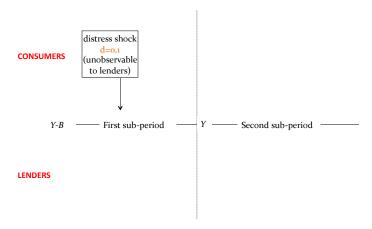
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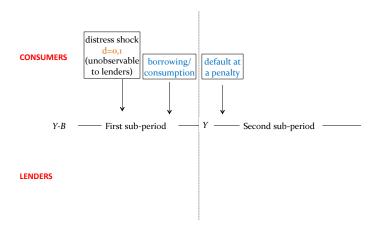
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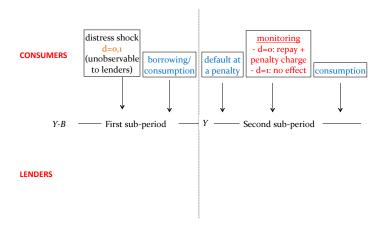


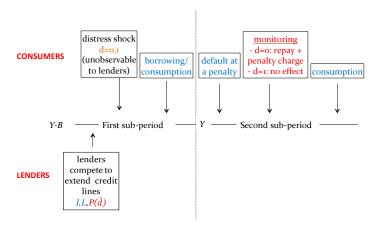
charge-off rate = (net) debt discharged / total debt

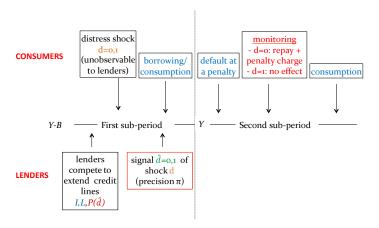


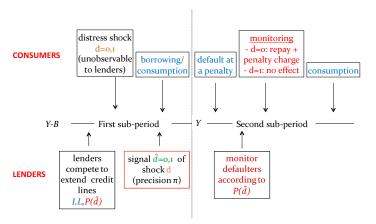












Equilibrium Contracts

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$$\circ~L>L_{min}(d=1)$$
: Risky Contracts (positive probability of default)

- $L \in (L_{min}(d=1), L_{min}(d=0)]$: Non-monitored contracts (default if d=1 for all P(s))
- $L > L_{min}(d=0)$: Monitored contracts (default if d=1, or if d=0 and $P(s) < \bar{P}$)

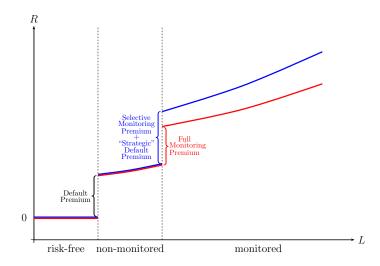
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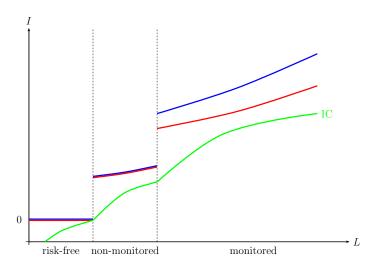
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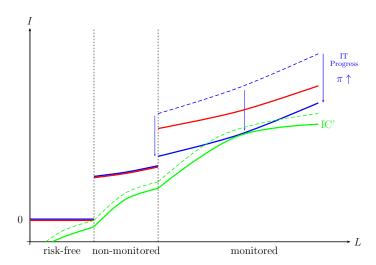
How Do Lenders Price Defaultable Debt



How Does π Impact Pricing?



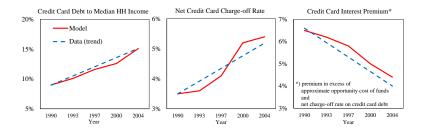
How Does π Impact Pricing?



Quantitative Extension

- Life-cycle environment (27 periods)
- Analytic model embedded within each period
 - \circ baseline period length (1 sub-period) = 1 year
- ullet B endogenous
- Y stochastic
- $\mathsf{E} = (Y < .25\bar{Y}) + \mathsf{medical\ bills} + \mathsf{divorce} + \mathsf{unwanted\ pregnancy}$
- ullet Only medical shock assumed directly defaultable o low ϕ

Model Accounts for Both Trends and Levels



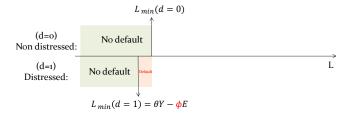
- information precision x3 over the 90s
- transaction cost declines by 20% (Berger, 2003)

Why Model Matches Trends?

	Benchmark Model		Decomposition		
	90s	00s	$ au_{90s}$	π_{90s}	
(in % unless otherwise noted)			π_{00s}	τ_{00s}	$ au_{fit}$
CC Debt to Med. Income	9.0	15.1	11.2	13.9	15.1
CC Charge-off Rate	3.5	5.4	5.5	4.1	4.1
Defaults (per 1000) - fraction monitored - fraction strategic	4.5 30 0.0	10.8 18 19	9.0 17 19	7.5 31 0	7.9 32 0
Frequency of Risky Cont fraction fully monitored - fraction sel. monitored	21.4 100 0	36.6 1 99	35.7 0 100	31.3 100 0	31.1 100 0
Discharge to Income	74	89	82	80	82
CC Interest Premium	6.5	4.4	6.1	5.3	4.6

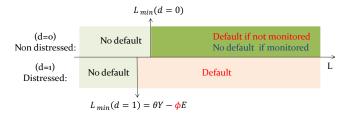
Why Model Matches Levels?

Standard model:



Why Model Matches Levels?

Our model:



Conclusions

- Complementary mechanism of IT-driven expansion of credit card lending
 - o departure motivated by:
 - prevalence of informal bankruptcy
 - involvement of lenders in debt collection
- Addresses Achilles' heel of existing models

THE END



Literature: Unsecured Credit and IT

Adverse Selection and Ex-ante Role of IT

Narajabad (2012), Athreya, Tam and Young (2008), Sanchez (2012) Livshits, MacGee and Tertilt (2011)

• Informal Bankruptcy

Benjamin and Mateos-Planas (2011), Athreya, Sanchez, Tam and Young (2012), Chatterjee (2010)

• Standard Modeling Frameworks

Livshits & MacGee and Tertilt (2006, 2010), Chatterjee, Corbae, Nakajima and Rios-Rull (2007), Athreya (2003) etc...

⇒ define modeling issues / challenges

Lenders: Contract Assignment

• Choose K = (R, L) & P(s) to maximize

$$\max_{K,P} V(K,P)$$

subject to

$$\mathbb{E}\Pi(I,K,P) - \lambda \sum_{I=(d,s)} \delta(I,K,P) P(s) Prob(I) \geq 0,$$

where $I \equiv (d,s)$ and ex-post profit function $\Pi(I,K,P)$ given by

$$\Pi(I,K,P) = \begin{cases} R \max\{b(I,K,P),0\} & \text{if } \delta(I,K,P) = 0 \\ -L + L(1+\bar{R})(1-d)P(s) & \text{if } \delta(I,K,P) = 1 \end{cases}$$

Consumers: Decision to Default

 \bullet Choose $\delta \in \{0,1\}$ to maximize

$$V(K, P) \equiv \mathbb{E} \max_{\delta \in \{0, 1\}} [(1 - \delta)N(I, K, P) + \delta D(I, K, P)]$$

where I = (d, s) and

- $N(\cdot)$ is indirect utility fcn. associated with repayment
- $D(\cdot)$ is indirect utility fcn. associated with default

Consumers: Indirect Utility from Repayment

• Under repayment, choose b, c, c' to maximize

$$N(I, K) \equiv \max_{b \le L} U(c, c')$$

subject to

$$\begin{cases} c = Y - B + b - \rho(K, b) \\ c' = Y - b - dE - \rho(K, b) \end{cases}$$

where I = (d, s) and

$$\rho(K, b) = R \max\{b(I, K, P), 0\}/2$$

Consumers: Indirect Utility from Default

• Under default, choose b, c, c' to maximize

$$D(I, K, P) \equiv \max_{-L \le b \le 0} \mathbb{E}_{I} U(c, c')$$

subject to

$$\begin{cases} c = Y - B + L + b \\ c' = (1 - \theta)Y - (1 - \phi)dE - b - mX(d) \end{cases}$$

where I = (d, s) and

$$X(d) = (1 - d)((\underline{\theta} - \theta)Y + L(1 + \bar{R}))$$

 $\theta Y + \bar{R}L$ s.t. d=0-consumer does not default if P=1

Definition of Equilibrium

• Equilibrium is: indirect utility functions

$$V(\cdot), N(\cdot), D(\cdot)$$

and decision functions

$$\delta(\cdot), b(\cdot), K(\cdot), P(\cdot)$$

s.t. consistent with problems defined above.

Parameterization

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- Choose β , $\bar{\theta}$, θ , π , λ
 - o indebtedness for 2004: 15%
 - o charge-off rate for 2004: 5%
 - o discharge to income of bankruptcy filer in the 90s
 - \circ 3 fold increase in π centered around .5
 - $\lambda = .3$ to get regime switch around $\pi = .5$

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- Decline of transaction cost by 20% (consistent with Berger, 2003)

Direct Impact of IT-Based Solution

- In early 90s, GE capital developed PAYMENT; first comprehensive solution (Markuch et al., 1992) to direct collection resources:
 - Markov model of evolution of delinquent debt as a function of possible actions taken by collectors
 - o systematic comparison of accounts treated vs non-treated
 - report 7-9% gain in overall effectiveness and improved borrower goodwill
 - explicit mention that most gains due to more frequent selection of no action
 - as for first implementation of this sort of system this is big number

Direct Impact of IT-Based Solution

- Banerjee (2001) directly looks at yield from litigation on cc-receivables:
 - o yield from litigation boosted from 24% to 40% by IT!

Direct Impact of IT-Based Solution

- Other industry studies report even higher numbers:
 - PRA, major debt collection agency, reports 120% gain in debt recovered per dollar spent on collection over the years 1997-2004 (Annual Report, 2011)
 - Trustmark National Bank, discussed adoption of Fair ISAAK debt collection system in late 90s: 35-58% gain on consumer receivables with same staff

Other Important Evidence

- In 90s all 3 major credit bureaus started offering collection scores, marketed to debt collection industry; this accounts for 7% of their revenue, which suggests:
 - o 1. these scores aid collection by segmenting/prioritizing debtors
 - o 2. segmentation and prioritization is of first order importance

Comparison to the Model

- IT progress rate in the ballpark of assumed numbers:
 - $\circ\,$ in model 33% gain in efficiency, industry data report vary between 9%-120%
- Cost of monitoring on the high side, but not unreasonable:
 - o pre-PAYMENT GE spent \$150 million on final write-offs \$400 million
 - suggests 150/(400/.74)=.28 as upper bound on monitoring cost (we use .3)
 - aggregate costs also consistent with the model's implication: data: 350k*\$50k*30% -2% x \$800 billion on 5%x800 billion aggregate charge-offs