#### Competition and Selection in Credit Markets

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Fourth Biennial Conference on Auto Lending

July 15, 2021

#### Concentration and Interest Rates



#### Concentration and Interest Rates





Trended Credit and Alternative Data Attributes Improve Dealer and Consumer Experience in Auto Finance Market

TransUnion Client Arivo Acceptance optimizes performance with CreditVision Suite

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# Global Lending Services revamps decisioning software

#### **Automotive News**



March 11, 2019 12:00 AM

#### Alternative credit data gain acceptance

DAVID MULLER 🦄 🖂

#### Spireon Launches GoldStar Wireless GPS Tracking Solution for Buy Here Pay Here Dealerships and Auto Lenders

#### AUTO FINANCE NEWS

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# This Paper

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- Test model predictions using auto loans data
- Rates and concentration pos. correlated for low-risk groups, neg. for high-risk
- Result holds using variation from bank failures, bank mergers
- In more concentrated markets, lenders invest more in observable screening technologies
- Model predictions on default rates, loan quantities also verified

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#### Implications

- New effect of competition in screening markets
- Implications for competition policy/antitrust
- $^{4}$

### Related Literature

- Competition, data acquisition, screening: Broecker (1990), Hauswald & Marquez (2003, 2006), He, Huang, & Zhou (2020)
- Competition in lending markets: Petersen & Rajan (1995), Parlour & Rajan (2001), Sapienza (2002), Agarwal & Hauswald (2010), Giroud & Mueller (2010, 2011), Scharfstein & Sunderam (2016), Drechsler, Savov, & Schnabl (2017, 2018), Egan, Hortacsu, & Matvos (2017), Buchak, Matvos, Piskorski, & Seru (2018, 2020), Robles-Garcia (2020), Buchak & Jorring (2021)
- Competition and selection: Agarwal, Chomsisengphet, Mahoney & Stroebel (2015), Mahoney & Weyl (2017), Crawford, Pavanini & Schivardi (2018), Lester, Shourideh, Venkateswaran, & Zetlin-Jones (2019), DeFusco, Tang, & Yannelis (2021)
- Auto lending: Einav, Jenkins, & Levin (2012, 2013), Benmelech, Meisenzahl, & Ramcharan (2017), Argyle, Nadauld, & Palmer (2020a, 2020b), Grunewald, Lanning, Low, & Salz (2020)

### Model Overview

Lenders compete to make loans to consumers

- Market power: Lenders set prices above marginal costs
- Screening: Lenders make fixed-cost investments to identify and screen out high-risk borrowers

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In high-risk populations, second effect can dominate, so more competition leads to higher interest rates

- $\bullet~N$  identical lenders, indexed by j
- Unit mass of consumers, located on a Salop circle
- $\bullet$  Unit mass of type-G borrowers never default
- $\bullet\,$  Measure q of type-B borrowers always default
- Lenders compete in two stages:
  - 1. Pay fixed cost  $c_{q}\left(\delta_{j}\right)$  to screen out type B's, decrease default rate to  $\delta_{j}$
  - 2. Set interest rate  $r_j$

#### Price-setting

 $\bullet\,$  Suppose default rate is  $\delta_j.$  Lenders' profits:

$$\Pi_{j} = s_{j} \left( r_{j} - \frac{\delta_{j}}{1 - \delta_{j}} \right)$$

• Optimal markups:

$$s_{j} - \frac{\partial s_{j}}{\partial r_{j}} \left( r_{j} - \frac{\delta_{j}}{1 - \delta_{j}} \right) = 0$$
$$\implies r_{j} - \frac{\delta_{j}}{1 - \delta_{j}} = \frac{\theta}{N}$$

• Higher  $N \implies$  lower markups

### **Optimal Screening**

- By paying fixed cost  $c_q(\delta_j)$ , lender removes some type-B consumers from population, decreasing default rate to  $\delta_j$ 
  - $\bullet\,$  Higher cost when population fraction of type-B's, q, is higher
- In baseline model, assume screening outcomes perfectly correlated among firms
  - Hence, no "winner's curse"
- With screening cost  $c_q(\delta_j)$ , lender solves:

$$\max_{\delta_{j}} \max_{r_{j}} \left( s_{j} \left( r_{j} \right) \right) \left( r_{j} - \frac{\delta_{j}}{1 - \delta_{j}} \right) - c_{q} \left( \delta_{j} \right)$$

$$\Longrightarrow \frac{s_{j}}{\left(1-\delta_{j}\right)^{2}} = -c_{q}'\left(\delta_{j}\right)$$

 $\bullet \ {\rm Higher} \ N \ \Longrightarrow \ {\rm lower} \ s_j \ \Longrightarrow \ {\rm lower} \ {\rm screening \ incentives}$ 

# Equilibrium

In symmetric equilibrium, lenders' market shares are:

$$s_j = \frac{1}{N}$$

Market shares determine markups:



Market shares also determine screening incentives:



#### Equilibrium Outcomes



#### Model Predictions

- 1. Effect of concentration on interest rates depends on population riskiness:
  - Low risk: Concentration  $\uparrow \implies$  interest rates  $\uparrow$
  - High risk: Concentration  $\uparrow \implies$  interest rates  $\downarrow$

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- 2. Concentration  $\uparrow \implies$  default rates  $\downarrow$  for all groups
- 3. Concentration  $\uparrow$  can cause loan quantity  $\downarrow,$  even if prices  $\downarrow$ 
  - Demand curves slope down, so this can't happen without some kind of screening/credit rationing

#### Data

#### Empirical setting

- Auto loans market
  - Third largest source of household debt in US
    - \$1.4 trillion outstanding in 2020
  - Segmented by borrower risk
  - Lenders bear most losses
    - Loan generally not guaranteed
    - Securitization rate low ( $\approx 20\%$ ) (SPG Global 2020)
  - Screening investments
    - Predictive analytics, ML/AI, GPS

#### Data

#### Transunion Consumer Credit Panel

- 10% sample of TransUnion credit records, 2009-2020
- Observe loan balance, payments, maturity: back out interest rates
- Loan volumes comparable to other datasets
- Observe lender ID, allowing us to compute HHIs

#### Call Reports

• Bank merger + market share data

#### Concentration and Interest Rates



# **Regression Specification**

- Begin with correlation between rates and HHI
- Estimate variants of

 $\label{eq:rct} \mathsf{ln}\left(r_{ct}\right) = \alpha_{c} + \alpha_{t} + \beta \, \mathsf{ln}\left(\mathsf{HHI}_{ct}\right) + \epsilon_{ct}$ 

HHI<sub>ct</sub> is Σ<sup>N</sup><sub>i</sub> s<sup>2</sup><sub>i</sub> (within credit score group)
α<sub>c</sub>, α<sub>t</sub> are county and year fixed effects

### Correlation between Concentration and Interest Rates

Interest Rates and and Market Competition									
	(1)	(2)	(3)	(4)	(5)	(6)			
			Ln(Interest F	late)					
	Cre	edit Score 300-	600	Cred	it Score 60	0-850			
Ln(HHI)	-0.0632***	-0.0723***	-0.0641***	0.269**	0.207*	0.134**			
	(0.0189)	(0.0200)	(0.0127)	(0.109)	(0.111)	(0.0628)			
Year FE	No	Yes	Yes	No	Yes	Yes			
County FE	No	No	Yes	No	No	Yes			
Obs	27,887	27,887	27,826	31,773	31,773	31,733			
R <sup>2</sup>	0.013	0.045	0.571	0.013	0.043	0.845			

### Finer Credit Score Buckets

#### Interest Rates and and Market Competition

	(1)	(2)	(3)	(4)	(5)	(6)			
	Ln(Interest Rate)								
	Credit Score	Credit Score	Credit Score	Credit Score	Credit Score	Credit Score			
	300-550	550-600	600-650	650-700	700-750	750 +			
Ln(HHI)	$-0.0715^{***}$	-0.0267***	$0.0467^{*}$	$0.106^{*}$	0.0471	0.103			
	(0.0129)	(0.00956)	(0.0242)	(0.0541)	(0.0315)	(0.0731)			
Year FE	Yes	Yes	Yes	Yes	Yes	Yes			
County FE	Yes	Yes	Yes	Yes	Yes	Yes			
Obs	25,985	27,019	29,162	30,093	29,794	29,989			
R <sup>2</sup>	0.392	0.565	0.619	0.665	0.760	0.844			

# Do Lenders Screen More in Concentrated Markets?



- Lenders can purchase proprietary product from TU
- Additional information on consumer behavior and histories
- Includes predictive modeling
- Purpose-built scores, propensity models, attributes, algorithms, estimators, etc.

### Market Concentration and Increased Screening



### Market Concentration and Credit Scores



# Bank Market Structure Variation

- Market HHI may be correlated with many factors associated with price, even within county
- Want source of variation in market structure uncorrelated with other determinants of prices
- Exploit variation from bank market structure shocks:
  - 2008 large bank failures (Buchak & Jorring, 2021)
  - Bank mergers

#### Auto Loan HHI and Bank Deposit Market HHI



#### **Bank Failures**

- Three large banks failed during 2008 crisis: Wachovia, Washington Mutual, Countrywide (Buchak & Jorring, 2021)
- Counties where these banks had high pre-crisis market share have lower post-crisis concentration
- First stage:

$$ln(HHI_{ct}) = \psi Share_c + X_{ct} + e_{ct}$$

• Second stage:

$$ln(r_{ct}) = \zeta ln(\hat{HHI}_{ct}) + X_{ct} + v_{ct}$$

• Identifying assumption: pre-crisis market share of failed banks not correlated with auto loan rates, except through concentration effects

### Bank Failures: IV Estimates



# Bank Mergers

• Use bank mergers as a shock to market concentration

• Estimate:

$$\ln(\text{HHI}_{cst}) = \alpha_{cs} + \alpha_t + \alpha_y + \sum_{i=-5}^{5} \zeta_i \mathbb{1}[i=t] + \xi_{cst}$$
$$\ln(r_{cst}) = \alpha_{cs} + \alpha_t + \alpha_y + \sum_{i=-5}^{5} \zeta_i \mathbb{1}[i=t] + \xi_{cst}$$

# Bank Mergers: Effect on Concentration



# Bank Mergers: Effect on Rates

Credit Score 300-600

Credit Score 600-850



#### Bank Mergers: Reduced Form Estimates



# **Delinquency** Rates

#### **Delinquency Rates and and Market Competition**

Panel A: Delinquency and HHI									
	(1)	(2)	(3)	(4)	(5)	(6)			
			Ln(Delin	quency)					
	Cr	edit Score 300-6	300	Cr	edit Score 600-8	350			
Ln(HHI)	-0.0974**	-0.0534	-0.0290**	-0.00550	-0.0225	-0.0272**			
	(0.0381)	(0.0421)	(0.0143)	(0.0202)	(0.0183)	(0.0125)			
Year FE	No	Yes	Yes	No	Yes	Yes			
County FE	No	No	Yes	No	No	Yes			
Obs	27,887	27,887	27,826	31,773	31,773	31,733			
R <sup>2</sup>	0.027	0.710	0.827	0.089	0.617	0.825			

#### Panel B: Delinquency and HHI by Credit Score

	(1)	(2)	(3)	(4)	(5)	(6)
			Ln(Delir	quency)		
	Credit Score					
	300-550	550-600	600-650	650-700	700-750	750 +
Ln(HHI)	0.00641	-0.0651***	-0.0528***	-0.0543***	-0.0301***	-0.0147**
	(0.0131)	(0.0145)	(0.0151)	(0.0139)	(0.0114)	(0.00746)
Year FE	Yes	Yes	Yes	Yes	Yes	Yes
County FE	Yes	Yes	Yes	Yes	Yes	Yes
Obs	25,985	27,019	29,162	30,093	29,794	29,989
R <sup>2</sup>	0.762	0.724	0.694	0.621	0.514	0.503

# Loan Quantities

#### Loans and and Market Competition

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
				Ln(L	oans)			
	Credit Score							
	300-600	600-850	300-550	550-600	600-650	650-700	700-750	750 +
Ln(HHI)	-0.105***	-0.0480***	-0.0663***	-0.0669***	-0.0412***	-0.0350***	-0.0336***	-0.0286***
	(0.0112)	(0.0129)	(0.00873)	(0.00794)	(0.00731)	(0.00740)	(0.00689)	(0.00942)
Year FE	Yes							
County FE	Yes							
Obs	27,826	31,733	25,985	27,019	29,162	30,093	29,794	29,989
R <sup>2</sup>	0.974	0.983	0.977	0.979	0.982	0.985	0.986	0.983

• Results from panel regression of loan quantities:

$$Q_{\texttt{ct}} = \alpha_{\texttt{c}} + \alpha_{\texttt{t}} + \beta \ln{(\texttt{HHI}_{\texttt{ct}})} + \epsilon_{\texttt{ct}}$$

- In high-score buckets, concentration  $\uparrow \implies$  interest rate  $\uparrow$ , so quantity decrease is intuitive
- In low-score buckets, concentration  $\uparrow \implies$  interest rate  $\downarrow$ , so quantity decrease shouldn't happen, with downward-sloping demand
  - Suggests some sort of screening/credit rationing

#### Specialized lenders

# Specialized and Non-specialized Lenders

#### **Concentration and Interest Rates by Lender**

-	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
				Ln(	Interest Rate)			
		Full Sa	mple		Auto I	Lenders	All L	enders
	Credi	t Score	Credit	t Score	Credit Score	Credit Score	Credit Score	Credit Score
	300	-600	600	-850	300-600	600-850	300-600	600-850
Ln(HHI)	-0.0737***	-0.0727***	0.110**	$0.105^{**}$	-0.0689***	0.173**	-0.0785***	0.0282
	(0.0103)	(0.00996)	(0.0517)	(0.0502)	(0.0114)	(0.0764)	(0.00921)	(0.0245)
County X Score FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Lender FE	No	Yes	No	Yes	Yes	Yes	Yes	Yes
Observations	1,866,741	1,865,935	5,213,296	5,212,710	1,107,222	2,401,999	758,663	2,810,678
R <sup>2</sup>	0.678	0.689	0.802	0.805	0.728	0.813	0.622	0.760

### Lender size

#### Lender size

-				
	(1)	(2)	(3)	(4)
	(p 50) interest_rate	(p 50) interest_rate	(p 50) interest_rate	(p 50) interest_rate
HHI	-0.0806***	0.0998**	-0.0909***	0.0881*
	(0.0106)	(0.0485)	(0.0108)	(0.0456)
X 7 1	0.0000118**	0.00000=**		
Volume	0.000244	0.000367		
	(0.0000254)	(0.000146)		
HHI Y Volumo	0.0000800***	0.000112**		
IIIII A volume	0.0000800	0.000113		
	(0.0000833)	(0.0000453)		
# Counties			0.0000327***	0.0000369***
<i>// 0000000000</i>			(0.0000021	(0.000000000000000000000000000000000000
			(0.00000368)	(0.0000117)
HHI X # Counties			0.0000105***	0.0000132***
			(0.00000121)	(0.00000415)
Observations	1866741	5913996	1866741	5213296
- 2	1000/41	0210290	1000/41	0210290
R <sup>2</sup>	0.678	0.802	0.678	0.802

• For large lenders, screening decisions may be made at aggregated level: local HHIs should matter less

• However, large and small lenders' markups should be similarly sensitive to HHI

•  $\implies$  Rates should be more positively correlated with HHI for larger lenders

• However, subprime coef is negative even for large lenders, suggesting there's some local component to screening costs

# Alternative Explanations

Concentration  $\uparrow$  leads to:

- 1. Interest rates  $\uparrow$  in low-risk groups,  $\downarrow$  in high-risk groups
- 2. Default rate  $\downarrow$  in both groups
- 3. Loan quantity  $\downarrow$  in both groups

Other explanations:

- Adverse selection: concentration should always increase interest rates (Mahoney & Weyl 2017 ReStat)
- Competition and loan standards: concentration should increase interest rates (Mian + Sufi 2009, Favara & Imbs 2015)
- Moral hazard: Doesn't explain asymmetry between groups.
   Doesn't explain rates ↑ in high group, but default rate ↓
- Dealer markups, het. funding costs: Doesn't explain asymmetry between groups. Also, our results hold for pure auto lenders
- Improved collections technology: similar effect to our channel

# Robustness to Alternative Specifications

Results Robust to Varying Main Specification:

- Diff-in-Diff Specification
- Alternative Weighting
- Restricting to Large Counties
- Including All Counties
- Winsorizing
- Loan Based HHI
- Using Number of Lenders

# Concluding Remarks

- Simple model of screening and price-setting
- When population risk is high, higher concentration can decrease interest rates
- In auto loans data, interest rates + correlated with concentration for low-risk groups, for high-risk, as model predicts
- Suggestive evidence that lenders screen more in more concentrated markets
- Result holds using variation from failures and bank mergers
- Other model predictions (default rates, quantities) also hold in data

 $\Rightarrow$  May need to rethink effects of competition in credit markets, and incorporate more heterogeneity between different types of borrowers