What's in a Picture? Evidence of Discrimination from Prosper.com

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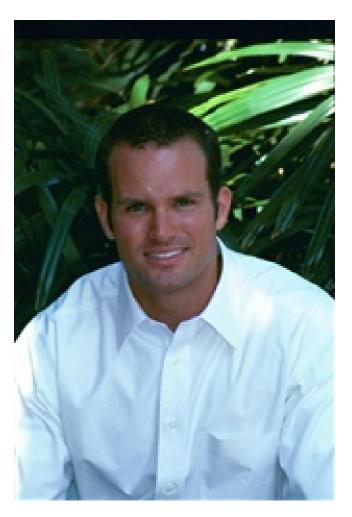
Justin R. Sydnor Economics Department Case Western Reserve University Imagine you are a loan officer...



"Need a car"



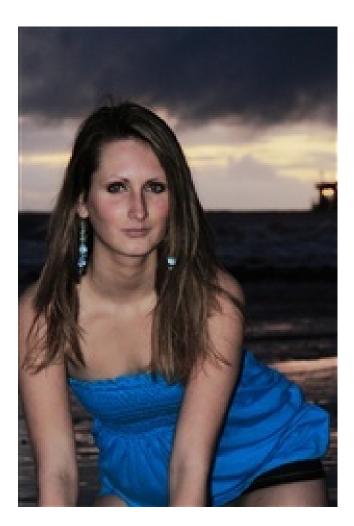
"Medical expenses"



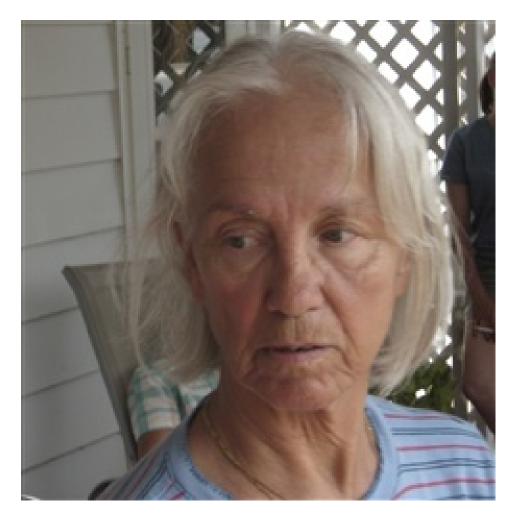
"Becoming debt free"



"Take a trip"



"Need school supplies"



"Fix my roof"

We study discrimination in the Prosper.com marketplace.

PRIMARY RESEARCH QUESTIONS

- How do lenders respond to information found in the pictures and text of online credit listings?
- Is the lender response to this information efficient?
 (Becker (1957) and Arrow (1973))

KEY ADVANTAGES TO USING PROSPER.COM DATA

- All information about borrowers that is seen by lenders is available to the econometrician.
- Loan terms and relevant ex-post performance of borrowers is available.

Outline

- Introduction
- Background of Prosper.com
- Data
- Determinants of access to credit (disparate treatment?)
- Efficiency of discrimination (taste-based discrimination?)
- Discussion and conclusion

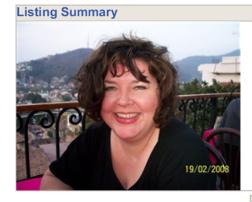
Background of Prosper.com

- Borrowers create loan listing
- All loans have 3-year terms and are uncollateralized
- Individual lenders bid to fund a portion of a loan (min. \$50)
- Loan funds only if total bids fully cover the requested loan size
- Once fully funded, auction process determines final interest rate
- Prosper charges closing costs of 1.5 3% of loan amount
- Defaulted loans reported to credit agencies and collection agency

An example of a Prosper.com loan listing

Paying off my high interest debt

Personal loan (Listing #351871)



	Help						
\$5,500.00 @ 14.85%							
	Bid down from 17.50%						
	Bid Now						
	Bid \$50 or more						
Funding:							
	100% funded						
Bids:	<u>113 bids</u>						
	1 day 19 hours 25 minutes left						
Borrower APR:	17.02%						
Mo. payment:	\$190.26 (3y loan)						

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/	e	u	IL.		ofi	le

creat Frome						A Lieth
C credit grade	🟠 Homeo	wner		15% debt to inco		
Now delinquent:	0	First credit line:	Jul-1987	Employment status:	Full-time	
Amount delinguent:	\$0	Current / open credit lines:	16 / 13	Length of status:	0y <mark>10</mark> m	
Delinquencies in last 7y:	0	Total credit lines:	31	Income range:	\$75,000-\$99,999	
Public records last 12m / 10y:	0/0	Revolving credit balance:	\$23,317	Occupation:	Professional	
Inquiries last 6m:	1	Bankcard utilization:	86%	8		

A Hole

Description

Purpose of loan:

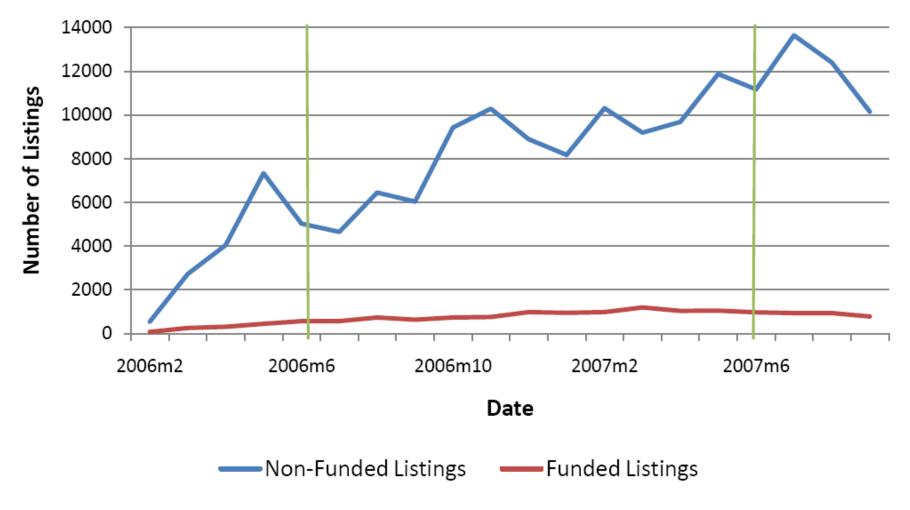
This loan will be used to... pay off a high interest credit card. I am making good, solid payments but almost half of my installment is going to interest and it seems that I am just spinning my wheels.

My financial situation:

I am a good candidate for this loan because... I have had some challenges in the past but I have overcome them all and have not taken the easy way out. I accept my obligations and am serious about becoming debt free within the next five years. I have successfully paid off a number of my accounts and am proud of that accomplishment.

Figure 1. Total Listings Across Time

By Funded Status



Funded over 24,000 loans worth \$154 million to date

Data collection process

DATA

- Our analysis focuses on all loan listings between June 2006 and May 2007.
- 110,333 distinct loan listings (10,207 funded loans)
- All credit and listing information that was provided to lenders was made available by Prosper.com
- Undergraduates systematically coded information from borrower's picture and one-line description without seeing other parameters of the listing.

Information from Listings

Information from Pictures (for those with a picture)

Information from Listings							
	All	Funded					
Variables	Listings	Listings	Variat				
Credit Grade			Main				
AA (760-800)	0.03	0.10	Adı				
A (720-760)	0.03	0.09	Jus				
B (680-720)	0.04	0.11	Bui				
C (640-680)	0.07	0.16	Ani				
D (600-640)	0.11	0.17	Aut				
E (560-600)	0.18	0.17					
HR (520-560)	0.54	0.20	For P				
NC	0.01	0.01	Gend				
			Sin				
Loan Information	7 4 5 4	5 000	Sin				
\$ Requested	7,154	5,930	Cou				
Borrower's Max Rate	0.17	0.20	Gro				
Final Rate	NA 0.12	0.18					
Fraction Funded Closed Auction Loans	0.12 0.43	1.00 0.37	Race				
Closed Auction Loans	0.43	0.37	Wh				
Other Information Provided			Bla				
Debt to Income Ratio	0.63	0.39	Asi				
Group Member	0.51	0.69	His				
Owns a Home	0.27	0.39	nis • • • •				
Provided a Picture	0.46	0.64	Age				
	0.40	0.04	Le				
Information Coded From			35				
Descriptions:			Μ				
Purpose of Loan							
Consolidate or Pay Debt	0.30	0.33	Нар				
Business/Entrepreneurship	0.10	0.10	Ha				
Pay Bills	0.04	0.02	Ne				
Education Expenses	0.03	0.03	Ui				
Medical/Funeral Expenses	0.03	0.02					
Home Repairs	0.02	0.03	Weig				
Auto Purchase	0.02	0.02	N				
Home/Land Purchase	0.02	0.02	So				
Auto Repairs	0.01	0.01	Ve				
Luxury Item Purchase	0.01	0.01					
Wedding	0.01	0.01	Attra				
Reinvest in Prosper	0.01	0.02	Ve				
Taxes	0.01	0.01	A				
Vacation or Trip	0.01	0.01	Ve				
Multiplie of Above Reasons	0.06	0.05					
Unclear/Other	0.34	0.33	Othe				
			Pr				
			CI				
			Si				
Observations	110,333	10,207					

Information from Pictures (for those with a picture)				
	All	Funded		
Variables	Listings	Listings		
Main Content of Picture				
Adult/Adults	0.65	0.67		
Just Children	0.10	0.07		
Buildings	0.04	0.05		
Animals	0.04	0.04		
Automobiles	0.02	0.02		
For Pictures with Adults:				
Gender				
Single Male	0.38	0.40		
Single Female	0.35	0.31		
Couple	0.20	0.22		
Group	0.07	0.07		
Race				
White	0.67	0.83		
Black	0.20	0.11		
Asian	0.03	0.03		
Hispanic	0.03	0.02		
Age				
Less than 35 yrs	0.53	0.54		
35-60 yrs	0.41	0.41		
More than 60 yrs	0.02	0.02		
Happiness				
Нарру	0.74	0.77		
Neutral	0.23	0.21		
Unhappy	0.01	0.01		
Weight				
Not overweight	0.73	0.75		
Somewhat overweight	0.20	0.18		
Very overweight	0.03	0.02		
Attractiveness				
Very attractive	0.05	0.06		
Average	0.91	0.91		
Very unattractive	0.03	0.02		
Other				
Profesionally Dressed	0.13	0.14		
Child Also in Picture	0.21	0.21		
Signs of Military Involvement	0.02	0.02		
	50,820	6,571		

....

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Related research on Prosper.com

FREEDMAN & JIN (2008)

- Study evolution of market
- Analyze Prosper.com's business model and profits to lenders

RAVINA (2008)

- 1-month of Prosper data
- Primarily focuses on the effect of beauty

Determinants of Access to Credit

Approach to identifying discrimination (disparate treatment)

ESTIMATION

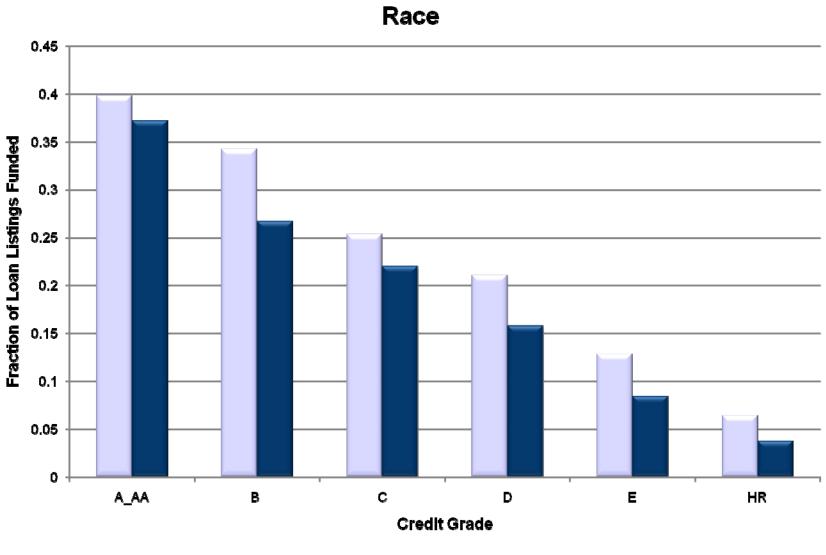
- Probability of receiving loan funding and interest rate conditional on funding as functions of listing variables
- Focusing on effect of characteristics such as race, age, and gender

OBSERVATIONAL DATA

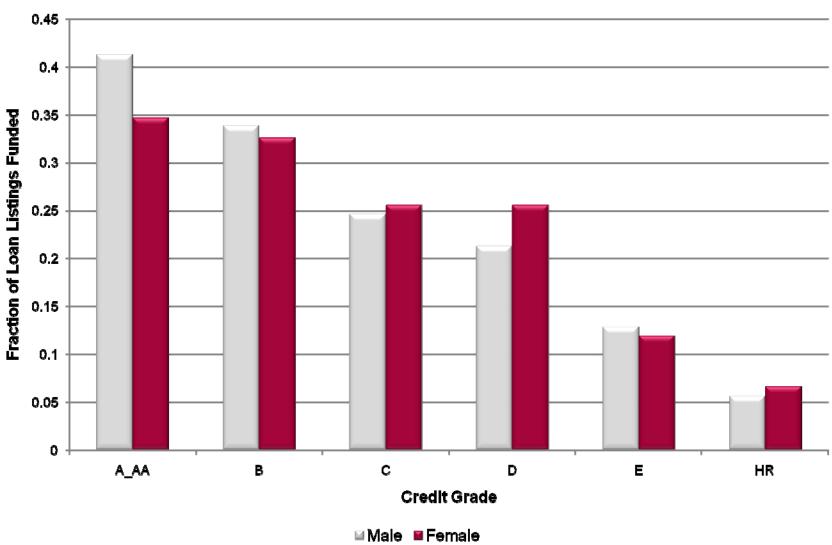
- E.g., Boston Fed studies (1996)
- Potential omitted variable bias & model specification questions

AUDIT/FIELD EXPERIMENTS

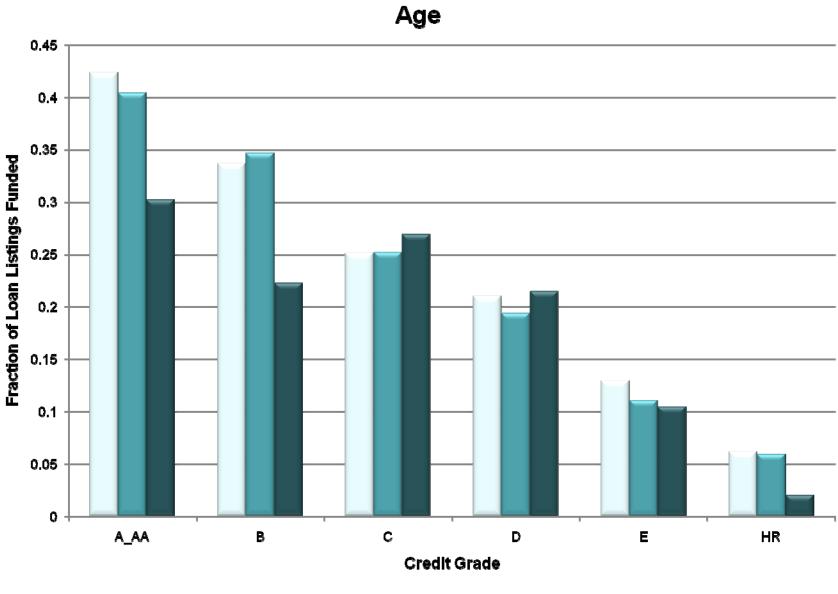
- E.g., Bertrand & Mullainathan (2004)
- Random assignment
- Potential to track ex-post performance
 - Typically difficult to track ex-post



🖬 White 🔳 Black



Gender



≌Young ≌Middle ∎Old

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Main OLS specification

$$Y_i = \alpha + X_i \beta + Z_i \theta + \varepsilon_i$$

- Y_i is an indicator for whether the listing got funded.
- Xi is a matrix of variables coded from pictures and text
- Zi is a matrix of other listing variables (e.g., credit controls)

ESTIMATION TECHNIQUES

- OLS
- Logit (marginal effects)
- Propensity score

	Dependent	Variable: Indicator = 1 if the Loan wa	as Funded
	Logit		Logit
Mean of Dependent Variable	0.093	Loan Purpose (BG: Unclear)	
Gender (BG: Single Male)		Consolidate or Pay Debt	0.004
Single Female	0.011		(0.002)*
	(0.003)**:	Business/Entrepreneurship	-0.006
Couple	-0.001		(0.003)**
·	(0.003)	Pay Bills	-0.010
Group	-0.004		(0.006)
Crosp	(0.004)	Education Expenses	0.001
Race (BG: White)	(0.004)		(0.005)
Black	-0.024	Medical/Funeral Expenses	-0.014
DIACK	(0.003)**		(0.006)**
A = i = -	. ,	Home Repairs	0.005
Asian	0.004		(0.006)
	(0.006)	Auto Purchase	-0.005
Hispanic	-0.006		(0.006)
Age (BG: 35-60 yrs)	(0 005)	Home/Land Purchase	-0.015
Less than 35 yrs	0.004		(0.006)***
Less than 55 yrs	(0.002)*	Auto Repairs	-0.015
More than 60 yrs	-0.011		(0.007)**
More than oo yrs	(0.007)*	Luxury Item Purchase	-0.011
Happiness (BG: Neutral)	(0.007)		(0.008)
Happy	0.002	Wedding	-0.006
Парру	(0.002)	Wedding	(0.007)
Unhappy	-0.016	Reinvest in Prosper	-0.010
Отпарру	(0.009)*	Reinvest in Fresper	(0.006)*
Weight (BG: Not Overweight)	(0.005)	Taxes	0.008
Somewhat overweight	0.003	Taxes	(0.011)
Come what over weight	(0.003)	Vacation or Trip	0.006
Very overweight	-0.008	vacation of Thp	-(0.011)
very overweight	(0.006)	Multiplie of Above Reasons	-0.003
Attractiveness (BG: Average)	(0.000)	Multiple of Above Reasons	
Very attractive	0.004		(0.003)
	(0.005)	Picture Characteristics	Х
Very unattractive	-0.005	Month Fixed Effects	Х
	(0.007)	Credit Controls	Х
Misc. Adult Information	(0.007)	Observations	110,332
Profesionally Dressed	0.002		
	(0.003)		
Child With Adult in Picture	0.001		
	(0.002)		
Signs of Military Involvement	0.025		
eight of Millary involvement	(0.009)***		

<u>Table 2.</u>	The Effect of Borrower Characteristics and Purpose on Loans Being Funded	
		=

		Dependent Variable: Indicator = 1 if the Loan was Funded							
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
Mean of Dependent Variable	0.093	0.093	0.093	0.093	0.093	0.093	0.093	0.093	
Race (BG: White)									
Black	-0.051	-0.028	-0.027	-0.022	-0.021	-0.021	-0.024	-0.024	
	(0.002)***	(0.003)***	(0.002)***	(0.003)***	(0.003)***	(0.003)***	(0.003)***	(0.003)***	
Asian	0.010	0.000	0.006	0.006	0.007	0.006	0.005	0.004	
	(0.008)	(0.006)	(0.006)	(0.006)	(0.006)	(0.006)	(0.006)	(0.006)	
Hispanic	-0.028	-0.013	-0.012	-0.007	-0.006	-0.005	-0.006	-0.006	
	(0.006)***	(0.006)**	(0.006)**	(0.006)	(0.006)	(0.005)	(0.005)	(0.005)	
Credit Grade (BG: HR & NC)									
AA		0.745							
		(0.004)***							
A		0.704							
		(0.004)***							
В		0.624							
		(0.004)***							
С		0.477							
-		(0.004)***							
D		0.315							
2		(0.004)***							
E		0.106							
-		(0.003)***							
Other Key Credit Variables		(0.000)							
Maximum Borrower's Rate		1.756							
Maximum Borrower's reate		(0.022)***							
Debt to Income Ratio		-0.014							
Debt to income Ratio		(0.001)***							
\$ Requested (thousands)		-0.000							
\$ Requested (indusands)		(.000)***							
Outline of Democrate Mars		(.000)							
Cubic of Borrower's Max								X	
Rate x Credit Grades			Х	Х	Х	Х	Х	х	
Cubic of Debt to Income									
Ratio and Amount			Х	Х	Х	Х	Х	Х	
All other Credit Controls				Х	Х	Х	Х	Х	
Long Description Text Controls					Х	Х	Х	Х	
Month Fixed Effects						Х	Х	Х	
Other Picture Characteristics							Х	Х	
Loan Purpose Fixed Effects								Х	
Observations	110,333	110,333	110,333	110,333	110,333	110,333	110,333	110,333	

Table 3. The Effect of Race on Loans Being Funded - Specification Robustness

		Depende	int variabler indio			
Panel A		Sample Cut by	Credit Grades		Sample C	ut by Time
	AA & A	B&C	D & E	HR & NC	First 6 Months	Last 6 Months
	(1)	(2)	(3)	(4)	(5)	(6)
Mean of Dependent Variable	0.335	0.225	0.108	0.035	0.087	0.096
Race (BG: White)						
Black	-0.041	-0.054	-0.033	-0.013	-0.020	-0.026
	(0.031)	(0.014)***	(0.005)***	(0.002)***	(0.004)***	(0.003)***
Asian	0.011	0.000	-0.004	0.006	0.002	0.005
	(0.027)	(0.019)	(0.011)	(0.006)	(0.011)	(0.006)
Hispanic	0.072	0.005	-0.011	-0.006	-0.022	0.006
	(0.054)	(0.026)	(0.010)	(0.004)	(0.008)***	(0.007)
Other Picture Characteristics	Х	Х	Х	Х	X	Х
Loan Purpose Fixed Effects	Х	Х	Х	Х	Х	Х
Month Fixed Effects	Х	Х	Х	Х	Х	Х
Credit Controls	Х	Х	Х	Х	Х	Х
Observations	5,587	12,123	32,154	60,391	45,941	64,386
			·			

Table 4. The Effect of Race on Loans Being Funded - Sample Cuts

Dependent Variable: Indicator = 1 if the Loan was Funded

	Dependent V	ariable: The Final Interest Rate for Fun	ded Loans
	OLS		OLS
Mean of Dependent Variable	0.182	Loan Purpose (BG: Unclear)	
Gender (BG: Single Male)		Consolidate or Pay Debt	-0.002
Single Female	-0.004		(0.001)*
	(0.001)***	Business/Entrepreneurship	0.002
Couple	-0.001		(0.001)*
	(0.001)	Pay Bills	0.007
Group	0.001	-	(0.004)*
	(0.002)	Education Expenses	0.003
Race (BG: White)			(0.002)
Black	0.006	Medical/Funeral Expenses	0.005
	(0.002)***		(0.003)*
Asian	0.002	Home Repairs	-0.001
	(0.002)		(0.002)
Hispanic	0.002	Auto Purchase	0.001
	(0.003)		(0.003)
Age (BG: 35-60 yrs)	()	Home/Land Purchase	0.000
Less than 35 yrs	-0.001		(0.003)
	(0.001)	Auto Repairs	0.004
More than 60 yrs	0.000		(0.004)
	(0.003)	Luxury Item Purchase	-0.001
lappiness (BG: Neutral)	(0.000)		(0.003)
Нарру	-0.001	Wedding	0.010
	(0.001)		(0.004)*
Unhappy	0.002	Reinvest in Prosper	0.004
•····	(0.004)		(0.002)*
Neight (BG: Not Overweight)		Taxes	-0.007
Somewhat overweight	0.002	i diveo	(0.004)*
Somewhat ever weight	(0.001)*	Vacation or Trip	0.005
Very overweight	0.003	Vacation of Thp	(0.004)
tory overweight	(0.002)	Multiplie of Above Reasons	0.003
Attractiveness (BG: Average)	(0.002)	Multiple of Above Redsolls	(0.002)
Very attractive	0.003	Open Funding Option Only	(0.002)
	(0.002)*	Picture Characteristics	х
Very unattractive	0.006	Month Fixed Effects	X
	(0.003)**	Credit Controls	X
Misc. Adult Information	(0.003)	R-Squared	^ 0.79
Profesionally Dressed	0.003	R-Squared Observations	0.79 10,207
Thesionally Diesseu	(0.001)**		10,207
Child With Adult in Picture	0.003		
Child With Adult III Ficture	(0.001)***		
Signo of Militory Involvement	· · ·		
Signs of Military Involvement	-0.002		
	(0.003)		

Table 5.	The Effect of Borrower	Characteristics and	Purpose on the Final Interest Rate
		for Funded Loans	

Efficiency of Discrimination

Disentangling motives and understanding the efficiency of discrimination

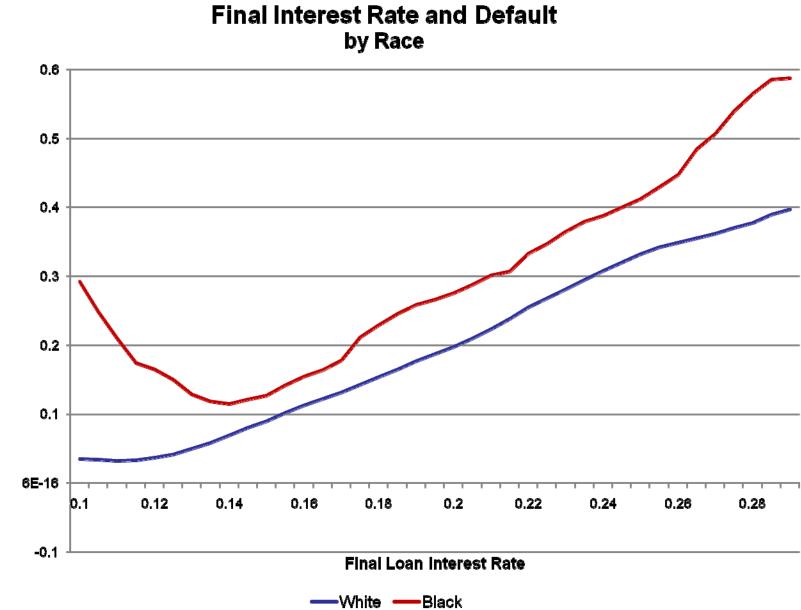
IMPLICATIONS OF DIFFERENT THEORIES OF DISCRIMINATION

Assuming lenders interested in expected net returns...

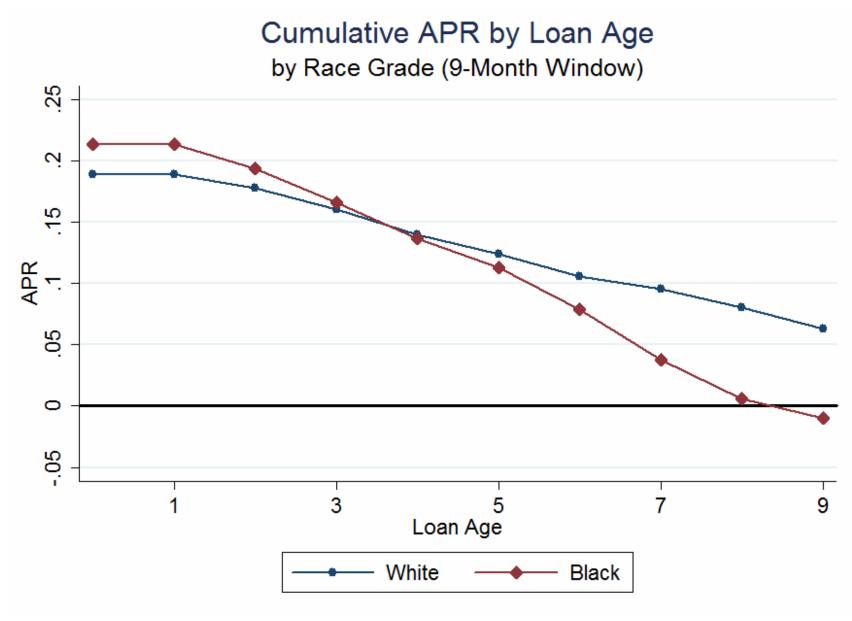
- Accurate statistical discrimination \rightarrow Identical net returns
- Taste-based discrimination \rightarrow Different net returns
- Inaccurate beliefs

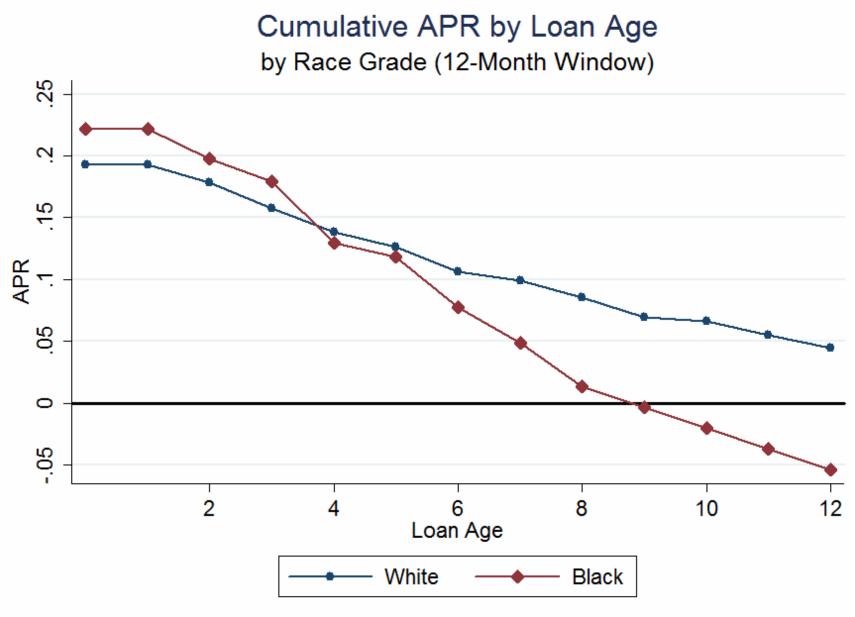
	Current Status of All Funded Loans (Fractions Reported)						
	Current or	1 Month	2 Months	3 Months	4+ Months	Total # of	
	Paid Off	Late	Late	Late	Late	Loans	
All Loans	0.78	0.02	0.02	0.02	0.17	10,118	
Age of the Loan							
(Months)							
7	0.91	0.02	0.04	0.01	0.02	230	
8	0.86	0.03	0.02	0.02	0.07	1,073	
9	0.84	0.03	0.02	0.01	0.10	1,095	
10	0.84	0.02	0.01	0.02	0.11	1,153	
11	0.77	0.02	0.02	0.03	0.16	912	
12	0.76	0.02	0.02	0.02	0.19	982	
13	0.74	0.02	0.02	0.02	0.20	950	
14	0.73	0.01	0.02	0.02	0.22	701	
15	0.72	0.01	0.02	0.02	0.24	704	
16	0.76	0.02	0.01	0.01	0.20	613	
17	0.71	0.01	0.01	0.02	0.24	753	
18	0.74	0.01	0.01	0.01	0.24	560	
19	0.71	0.02	0.01	0.01	0.25	392	
Race							
White	0.79	0.02	0.02	0.02	0.15	3,756	
Black	0.63	0.03	0.02	0.03	0.29	533	
Asian	0.80	0.02	0.02	0.02	0.14	163	
Hispanic	0.69	0.02	0.04	0.04	0.21	103	

Table 6. Loan Performance Summary Statistics



Fraction of Loans Defaulted





	Hazard Model - Dep Var: Default	OLS - Dep Var: 3-Year Return on Each Dollar Invested by Return Type		
	Default	Return Type I	Return Type II	Return Type III
Mean of Dependent Variable		1.047	1.066	1.084
Gender (BG: Single Male)				
Single Female	0.139	-0.023	-0.016	-0.016
2	(0.094)	(0.016)	(0.016)	(0.016)
Couple	-0.064	0.001	0.002	0.003
	(0.110)	(0.016)	(0.016)	(0.016)
Group	0.077	0.001	-0.001	0.005
	(0.161)	(0.026)	(0.026)	(0.025)
Race (BG: White)	(01202)		, , , , , , , , , , , , , , , , , , ,	, , ,
Black	0.346	-0.086	-0.082	-0.084
Diack	(0.100)***	((0.023)***	(0.023)***	(0.023)***
Asian	-0.230	0.017	0.017	0.015
, totali	(0.210)	(0.034)	(0.033)	(0.032)
Hispanic	0.050	-0.050	-0.051	-0.031
Thopanio	(0.231)	(0.047)	(0.047)	(0.046)
Age (BG: 35-60 yrs)	(0.201)	(0.0.17)	(0.0.17)	(01010)
Less than 35 yrs	-0.008	0.000	-0.003	-0.003
	(0.082)	(0.014)	(0.013)	(0.013)
More than 60 yrs	-0.042	-0.037	-0.039	-0.012
More than 00 yrs	(0.362)	(0.044)	(0.043)	(0.041)
Happinger (BC: Noutral)	(0.302)	(0.044)	(0.043)	(0.041)
Happiness (BG: Neutral)	-0.072	0.022	0.017	0.012
Нарру	(0.083)	(0.015)	(0.014)	(0.012)
Unhappy	0.443	-0.016	-0.039	-0.070
Оппарру	(0.266)*	(0.060)	(0.060)	(0.060)
	(0.266)	(0.060)	(0.060)	(0.060)
Weight (BG: Not Overweight)	0 116	-0.021	-0.017	-0.008
Somewhat overweight	0.116			
N/	(0.095)	-0.017	(0.017)	(0.017)
Very overweight	-0.164	0.072	0.052	0.052
	(0.207)	(0.039)*	(0.039)	(0.040)
Attractiveness (BG: Average)	0.007	0.010	0.044	0.010
Very attractive	-0.097	-0.012	-0.014	0.012
	(0.181)	(0.030)	(0.030)	(0.028)
Very unattractive	0.309	-0.049	-0.046	-0.056
	(0.234)	(0.048)	(0.046)	(0.045)
Misc. Adult Information				
Profesionally Dressed	0.129	-0.025	-0.013	-0.012
	(0.117)	(0.018)	(0.018)	(0.017)
Child With Adult in Picture	0.130	-0.035	-0.029	-0.022
	(0.080)	(0.014)**	(0.014)**	(0.013)*
Signs of Military Involvement	0.458	-0.034	-0.029	-0.052
	(0.260)*	(0.048)	(0.047)	(0.046)
Loan Purpose Fixed Effects	X	X	X	X
Month Fixed Effects	Х	х	Х	х
Credit Controls	Х	х	Х	х
R-Squared		0.26	0.26	0.26
Observations	9,963	10,113	10,113	10,113

Table 7.	The Effect of Borrower	<u>Characteristics on Ne</u>	<u>t Return on Investment</u>

Interpreting the results

HIGHER DEFAULT RATE AMONG BLACKS

- Skin color is clearly not a causal factor of default
- Must be factors that lenders cannot perfectly observe that predict default and are correlated with race (e.g., education, neighborhood effects, support networks, and labor-market discrimination)

LOWER NET RETURNS AMONG BLACKS

Maintaining the assumption of accurate beliefs...

– Statistical discrimination against blacks

&

Taste-based discrimination against whites

Interpreting the results cont.

AN ALTERNATIVE EXPLANATION

- Lenders understand that due to certain unobservables blacks are more likely to default than whites.
- Lenders under appreciate the impact that unobservables have on the likelihood of default or the correlation between race and unobservables.

LONG RUN EQUILIBRIUM IN MARKET

- Difference in net returns may disappear in the long run.
- Disappearance in the long run could be interpreted as evidence of inaccurate beliefs.

Conclusions

KEY RESULTS

- Evidence of racial discrimination in a new credit market.
- Differences in net returns across races suggests taste-based discrimination in favor of blacks or innaccurate beliefs.

IMPLICATIONS FOR DISCRIMINATION LITERATURE

- Results highlight potential importance of inaccurate beliefs when examining theories of discrimination.
- Simple evidence of disparate treatment may be misleading when thinking about the sources of discrimination.

Thank You!!