

Bivariate Tests of Credit Market Discrimination with an Application to Intersectionality

Anastasia Cozarenco

Montpellier Business School, France
& Centre for European Research in Microfinance (CERMi)

Ariane Szafarz

Université Libre de Bruxelles (ULB), SBS-EM, Centre Emile Bernheim, Belgium
CERMi & New York University

ASSA Annual Meetings, January 8, 2022

Motivation (1)

- **Discrimination in credit market**

- **Definition:** loan application rejected due to *personal* rather than *economic* characteristics of the borrower (Black et al., 1978)

⇒ **Economic AND ethical topic**

- **Social lending is not immune to discriminatory biases**

- Female borrowers receive smaller loans than men in [Brazil](#) (Agier & Szafarz, 2013); [France](#) (Cozarenco & Szafarz, 2018; Brana, 2013); [Italy](#) (Alesina et al., 2013); [Peru](#) (Buvinic & Berger, 1990); [Paraguay](#) (Fletschner, 2009).
- Disabled people are discriminated against in [Africa & Asia](#) (Cramm & Finkenflugel, 2008); [Uganda](#) (Mersland et al., 2009; Labie et al., 2010); [Zambia & Zimbabwe](#) (Lewis, 2004).

Motivation (2)

Why do we need a new approach to detect biases in social lending?

- **Methodological issues**

- Data limitations (Delis et al., 2020): Often no data on rejected applicants, amounts requested, default records
- Tests on approval rates assume equal creditworthiness across groups
- Coexistence of positive & negative biases depending on the reference group (Ruffle & Shtudiner, 2014)
- New stream of papers using "outcome tests" (=profit for the lender) (Dobbie et al., 2020)
 - Major achievement in literature using lenders' profitability of marginal applicants

⇒ Not suitable for non-profit lenders

Objective of this paper

- **New approach to detect biases in social lending**
 - Outcome = recovery rate $\neq$ profit
 - Applicable to non-profit social lenders
- **Application to intersectional biases (women/migrants) in microcredit in France**

Test for discrimination (based on Ferguson & Peters, 1995) Heckman (1979) approach to address selection bias:

$$\text{Recovery rate}_i = \alpha_{\mathbf{R}} \mathbf{F_i} + \beta'_R X_i + \epsilon_i$$

$$E[\text{Approval}_i | \mathbf{F_i}, Z_i] = \Phi(\alpha_{\mathbf{A}} \mathbf{F_i} + \beta'_A Z_i)$$

$\mathbf{F_i}$ = Female

	Higher approval rate for women: $\alpha_{\mathbf{A}} > 0$	Nonsignificant difference between approval rates: $\alpha_{\mathbf{A}} = 0$	Lower approval rate for women: $\alpha_{\mathbf{A}} < 0$
Higher recovery rate for women: $\alpha_{\mathbf{R}} > 0$	No bias detected	Weak negative bias	Strong negative bias
Insignificant difference between recovery rates: $\alpha_{\mathbf{R}} = 0$	Weak positive bias	No bias detected	Weak negative bias
Lower recovery rate for women: $\alpha_{\mathbf{R}} < 0$	Strong positive bias	Weak positive bias	No bias detected

Extension to intersectional discrimination

$$\text{Recovery rate}_i = \alpha_R \mathbf{F}_i + \theta_R \mathbf{D}_i + \mu_R \mathbf{F}_i \cdot \mathbf{D}_i + \beta'_R X_i + \epsilon_i$$

$$E[\text{Approval}_i | \mathbf{F}_i, \mathbf{D}_i, Z_i] = \Phi(\alpha_A \mathbf{F}_i + \theta_A \mathbf{D}_i + \mu_A \mathbf{F}_i \cdot \mathbf{D}_i + \beta'_A Z_i)$$

$F_i = \text{Female}$, $D_i = \text{Migrant (non-EU)}$

Panel I: Definition of coefficients			
Comparing	with EU men	with EU women	with migrant men
EU women	$\delta_1 = \Phi(\alpha_A + \beta'_A \mathbf{Z}_i) - \Phi(\beta'_A \mathbf{Z}_i)$ $\gamma_1 = \alpha_R$		
migrant men	$\delta_2 = \Phi(\theta_A + \beta'_A \mathbf{Z}_i) - \Phi(\beta'_A \mathbf{Z}_i)$ $\gamma_2 = \theta_R$		
migrant women	$\delta_3 = \Phi(\alpha_A + \theta_A + \mu_A + \beta'_A \mathbf{Z}_i) - \Phi(\beta'_A \mathbf{Z}_i)$ $\gamma_3 = \alpha_R + \theta_R + \mu_R$	$\delta_4 = \Phi(\alpha_A + \theta_A + \mu_A + \beta'_A \mathbf{Z}_i) - \Phi(\alpha_A + \beta'_A \mathbf{Z}_i)$ $\gamma_4 = \theta_R + \mu_R$	$\delta_5 = \Phi(\alpha_A + \theta_A + \mu_A + \beta'_A \mathbf{Z}_i) - \Phi(\theta_A + \beta'_A \mathbf{Z}_i)$ $\gamma_5 = \alpha_R + \mu_R$
Panel II: Test results and interpretation			
$\forall j = 1, \dots, 5$	$\delta_j > 0$	$\delta_j = 0$	$\delta_j < 0$
$\gamma_j > 0$	No bias detected	Weak negative bias	Strong negative bias
$\gamma_j = 0$	Weak positive bias	No bias detected	Weak negative bias
$\gamma_j < 0$	Strong positive bias	Weak positive bias	No bias detected

An application to microcredit in Europe

Context

- **Microcredit in Europe** is a nascent, yet growing market
- 1.26 million active borrowers in 2019, +14% since 2018
- 3.70 billion gross microloan portfolio outstanding
- Highly standardized loans
- **Main objective:** financial inclusion & job creation
- Strong dependence on subsidies

Hand-collected dataset from a French NGO

- Microcredit to **unbanked** individuals
- Gender & citizenship (EU vs. migrants)
- Average loan size: EUR 2,231
- Average duration: 33 months
- Fixed interest rate: 1% - 4%

An application to microcredit in Europe

Descriptive statistics

Full sample		Panel I: Applicants		Panel II: Borrowers	
		Rejected	Accepted	Repaying	Defaulting
Variables	(1)	(2)	(3)	(4)	(5)
Percentage		0.437	0.563	0.871	0.129
Recovery rate				1.000	0.415
Female	0.485	0.450	0.511***	0.517	0.473*
Migrant	0.086	0.089	0.083	0.082	0.090
Observations	5,789	2,528	3,261	2,840	421

T-tests for equal means. Significance: *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Results (1) – Simple discrimination

	Recovery rate		Approval	
	Coef	SE	Marginal effect	SE
Female ($\hat{\alpha}$)	0.016**	(0.007)	0.028**	(0.014)
Migrant ($\hat{\theta}$)	-0.016	(0.014)	-0.009	(0.022)
Instruments:				
Unemployment rate in town of residence			0.016***	(0.004)
Unemployment rate at national level			-0.131***	(0.048)
Mills ratio	0.018***	(0.008)		
Number of observations	3,262		5,792	

*** p<0.01, ** p<0.05, * p<0.1

- **No bias:** women are more likely to receive a loan because they are more creditworthy than men
- Significant Mills ratio: Heckman approach is relevant

Results (2) – Intersectional discrimination

	Recovery rate		Approval	
	Coef	SE	Marginal effect	SE
Panel I: Marginal effects				
Female ($\hat{\alpha}$)	0.021***	(0.007)	0.020	(0.015)
Migrant ($\hat{\theta}$)	0.009	(0.016)	-0.050*	(0.027)
Female*Migrant ($\hat{\mu}$)	-0.055**	(0.027)	0.114***	(0.038)
Panel II: Total marginal effects				
Migrant women vs. EU women	-0.046**	(0.023)	0.065**	(0.032)
Migrant women vs. EU men	-0.025	(0.025)	0.085**	(0.033)
Migrant women vs. migrant men	-0.034	(0.028)	0.134***	(0.038)
EU women vs. EU men	0.021***	(0.007)	0.020	(0.015)
Migrant men vs. EU men	0.009	(0.016)	-0.050*	(0.027)

*** p<0.01, ** p<0.05, * p<0.1

- **Strong positive bias** in favor of migrant women vs. EU women
- **Weak positive bias** in favor of migrant women vs. EU/migrant men
- **Weak negative bias** against EU women vs. EU men
- **Weak negative bias** against migrant men vs. EU men

Summary of results

- Expectation : Social mission

	Men	Women
EU	—	+
Migrant	+	++

- Reality: Social mission & Stereotypes

	Men	Women
EU	+	—
Migrant	—	++

Conclusion

● Case Study:

- Positive and socially consistent bias in favor of migrant women in pro-social lending can hide entrenched stereotypes

⇒ Pro-social lenders are not immune to discriminatory attitude

● Method:

- Our test detects biases otherwise invisible: Intersectional discrimination can occur even without any simple discrimination

- Intersectional discrimination in lending deserves attention

⇒ Our test design can be extended to any type of lender (for-profit, non-profit, hybrid)

Thank you for your attention.