# FAIRNESS IN CONSUMER CREDIT MARKETS

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Think Forward Initiative

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Strong growth in consumer credit markets

- → How is credit allocated? based on perceived default risk
- $\rightarrow\,$  How is this credit risk calculated?

US: credit scores

Europe: government credit registries, internal models

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#### → Questions:

Do better data and technologies improve the allocation of credit?
Who is left out of credit markets, why and what are the consequences?
Can we reduce default risk while increasing access to credit?

## FAIRNESS IN CONSUMER CREDIT MARKETS

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### Algorithmic notions:

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- 3 Calibration  $\rightarrow$  outcomes independent of protected attributes

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#### Economic notions:

- 4 Access to credit
- 5 Price of credit
- 6 Access to information, transparency

Credit allocation mainly based on credit scores

- 1 Anti-classification: mandated by law (Fair Credit Reporting Act 1970, Equal Credit Opportunity Act 1974)
- 2 Classification parity: very little public information
- 3 Calibration: large body of evidence that outcomes are very unequal across protected groups
- ightarrow access and price of credit, financial literacy
  - 4 Low perceived transparency
- → despite legally mandated disclosure requirements

Unequal outcomes justified based on credit scores

#### Concerns:

- 1 Historically biased features lead to low credit scores for some groups
- 2 Potential for self confirming cycle of low credit scores, costly credit and high default rates
- 3 Large number of unscored consumers

Who are the unscored?

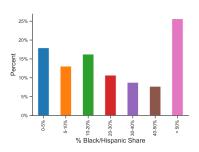
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Fraction of unscored by race. Source: Albanesi and Vamossy (2019).

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#### Lessons

- Credit scores key in enhancing fairness
- Similar concerns apply to Europe as consumer credit expands and societies become more diverse