

# NBER 2022 Open banking

## Credit Market Competition When Borrowers Own the Data

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- Many practitioners expect Open Banking (OB) to change the competitive landscape of the financial services industry
- Open banking: Regulatory initiatives that allow third parties access to customer transaction accounts (upon customer approval)

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  - 1) Milgrom (1981) implies that voluntary information sharing (open banking) is effectively not “voluntary” because of adverse inference
  - 2) Hauswald and Marquez (2003) study the effects of changes in lenders’ screening technologies on credit market equilibria

# Model Framework: Adaptation of Hauswald and Marquez

- Two borrower types:
  - ▶ high-types always repay
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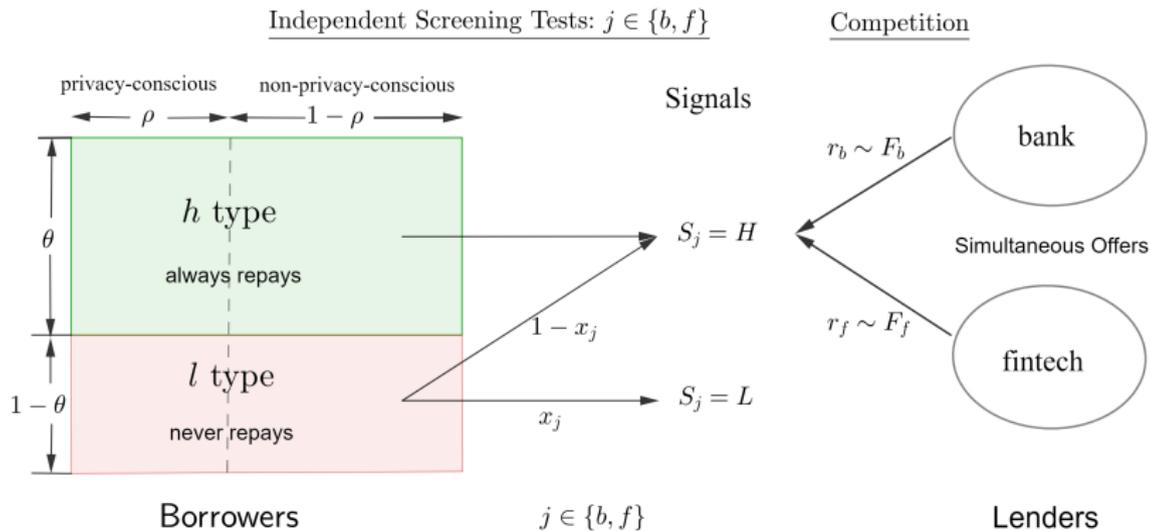
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- Asymmetric lenders
  - ▶ strong lender (Pre OB: Bank, Post OB: Fintech)
  - ▶ weak lender (Pre OB: Fintech, Post OB: Bank)

# Model Overview



**Before open banking:**  $x_f < x_b$

**After open banking, on a borrower who signs up:**  $x'_f > x_b$

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  - ▶ Makes it possible for the weak lender to sometimes undercut the strong lender (and still make a profit) just enough to offset the winner's curse loss when facing a low-type borrower

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- All results of the model can be understood by the interplay of the (1) strategic and 2) information effect

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- Why do customers share data (without privacy consciousness)?  
Because of adverse inference, unraveling à la Milgrom (1981)  
Equilibrium with mandatory sharing = Equil. with voluntary sharing

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  - ▶ Is effect of OB conceptually the same, but simply “large” change? (recall that *voluntary* nature of disclosure is not so important)

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- My preference: Highlight robust mechanism (information & strategic effect) rather than selecting non-obvious results

## Kitchen-sink out-of-the-model comments

- Model dynamics and additional effects:
  - ▶ Fintech: One can only train models with transaction data only **after** attracting a large number of consumers to share data (implications to generate scale as a precondition for future competitiveness)

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  - ▶ banks should be more regulated (see Harris, Opp, Opp, 2030)

# Conclusion

- Well-written paper on an important topic
- Paper highlights relevant trade-offs, maybe adjust overall message
- Integrate novel, specific features of Open Banking  
(because literature on credit market competition is “crowded”)