UNDP SHOWCASE PRESENTATION

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Data

Microbusiness Landscape

Methodology

Results

06 Policy Recommendation





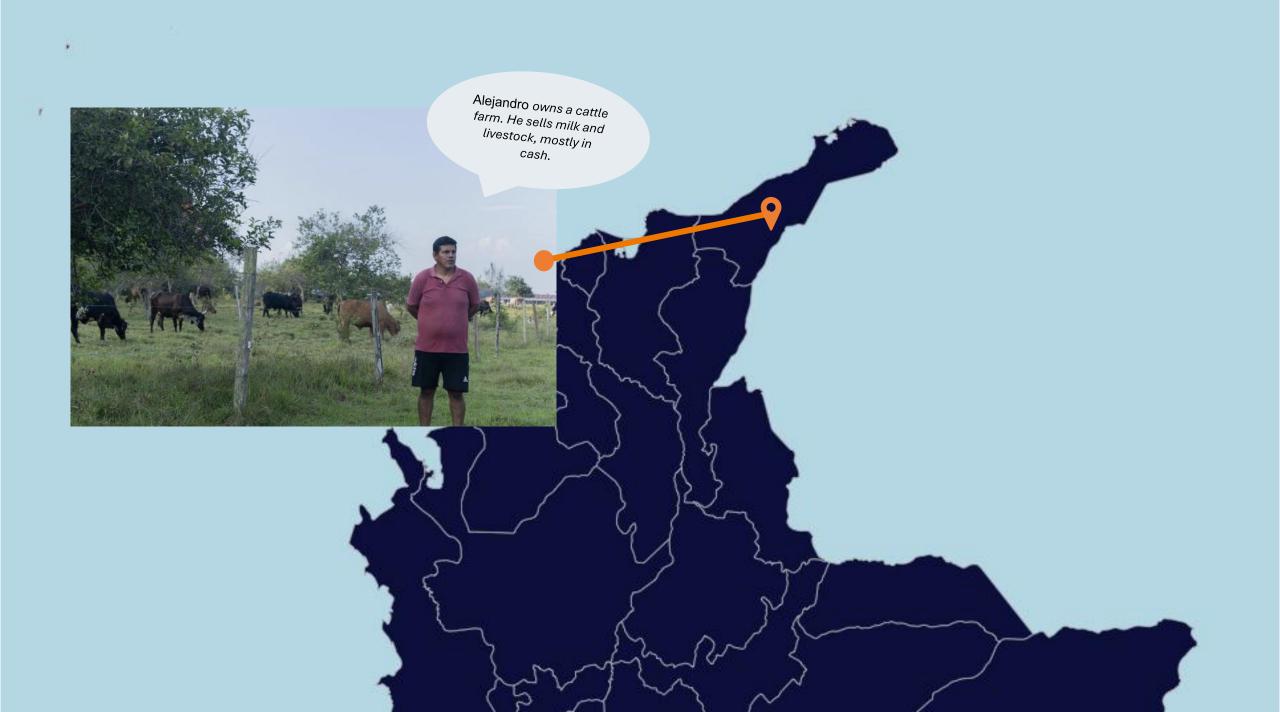






INTRODUCTION

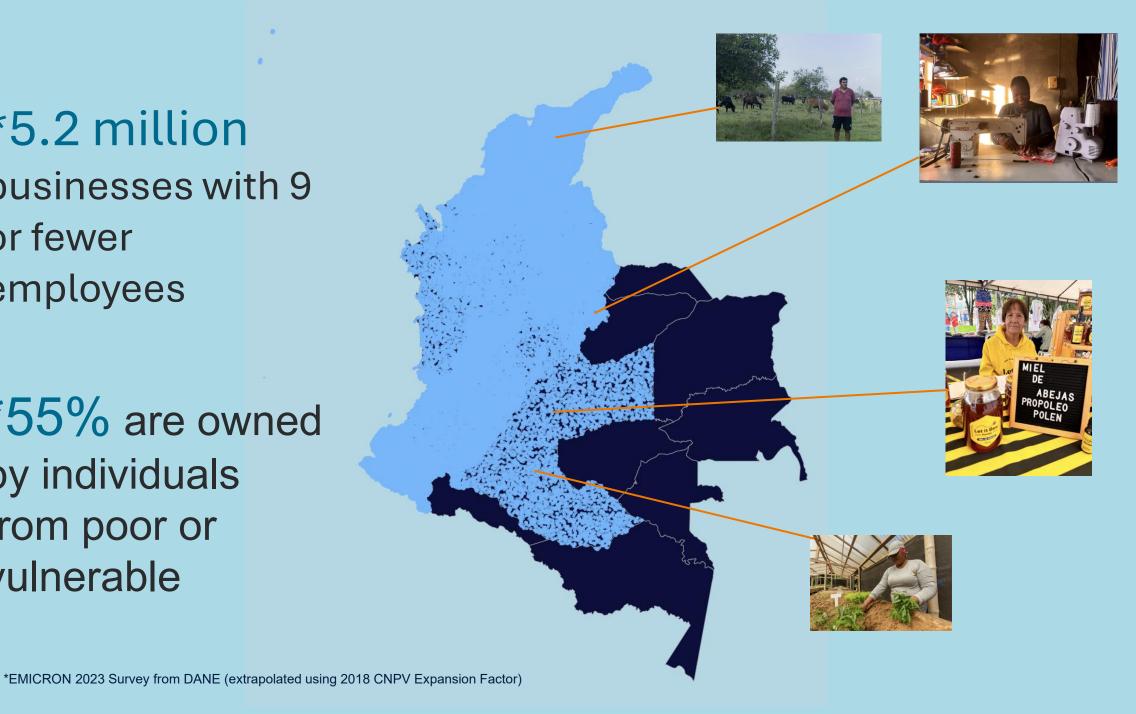




*5.2 million

businesses with 9 or fewer employees

*55% are owned by individuals from poor or vulnerable



BROAD PROBLEM

Informality

- 87% of microbusinesses operate entirely informally
- Neighborhood stores rely on informal financial networks
- These networks limit growth and increase risk
- Some turn to illicit funding in absence of support

Financial Exclusion

- 70% of small shop owners are unable to obtain formal credit due to inadequate financial records
- UNDP is working to "score the invisible" and expand access
- Alternative credit scoring
- Financially inclusive policies

OBJECTIVES

Understand Colombia's Microbusiness Landscape

- Understand regional trends and the underlying causes of microbusiness informality and financial exclusion
- Characterize businesses that remain informal or excluded
- Identify traits of businesses that are highly formal, financially included, or successful

Develop Quantifiable Indices to Summarize Complex Business Traits

- Indices should help guide financial inclusion policies for poor and vulnerable microbusinesses
- Can serve an alternative credit scoring model
- Produce concrete, data-driven recommendations that are transferrable to other countries







DATA





GEIH

(Gran Encuesta Integrada de Hogares)





Demographics

Household Labor & Connectivity

General Household Population Survey

EMICRON

(Microbusiness Survey)



Real

Estate





Labor

Finances

Sub-module Focused on Micro-Businesses



MICROBUSINESS LANDSCAPE





BUSINESSES APPLYING FOR LOANS

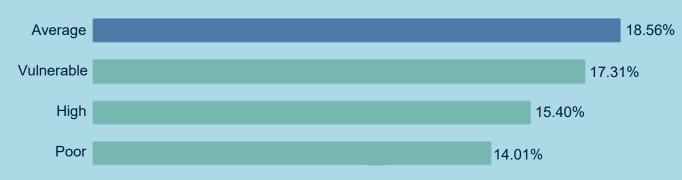
18%

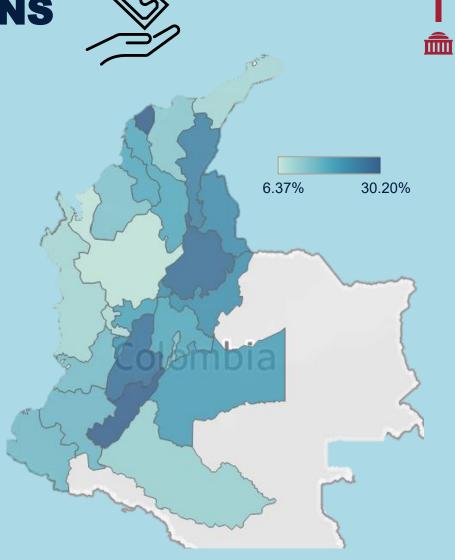
Applied for a loan in the last year

43%

Did not apply due to fear of debt 94%

Of loan applications are approved







BUSINESSES SAVING MONEY IN FINANCIAL INSTITUTIONS

19%

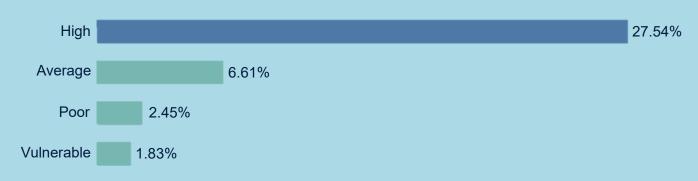
Saved money in the last year

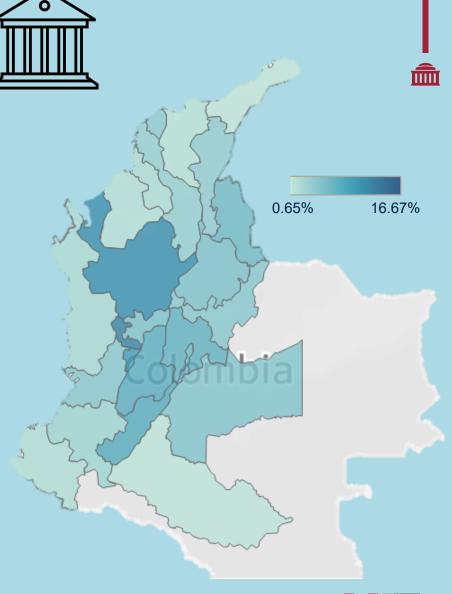
24%

Of savers do so in financial institutions

67%

Of savers do so at home









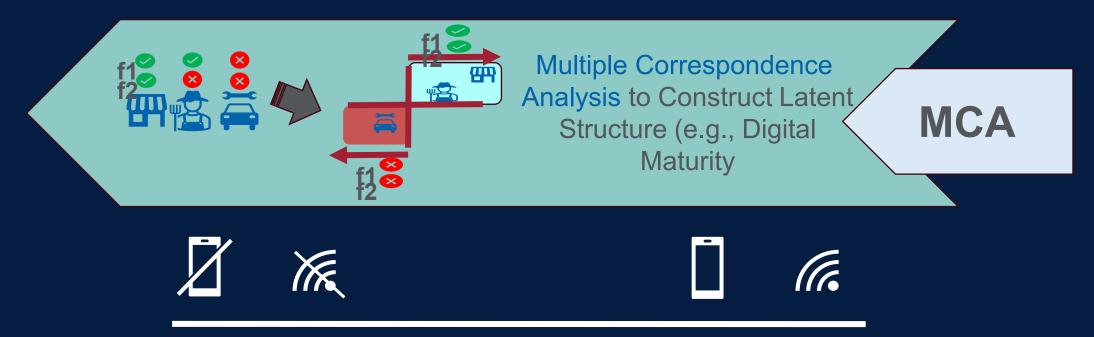




How naturally do businesses group together — and what features explain these groupings best?



How well does the data reveal underlying structures across categorical features?





How well does our theory explain the relationships in the data? If we can predict whether a business will invest its loan productively—
then the most important features tell us what sets them apart!

Modeling

Structural Equation Modeling (SEM) - interrelationship between latent structures

Assessment: Comparative Fit Index

Random Forest (RF) identify features that
predict productive loan use

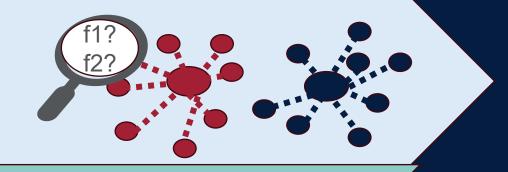
Assessment: Out-of-Bag Error Rate

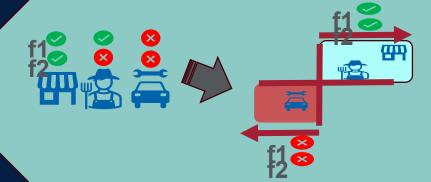


Clustering

KMeans Clustering

Identify Differentiating Features





Multiple Correspondence
Analysis to Construct Latent
Structure (e.g., Digital
Maturity

MCA

Modeling

Structural Equation Modeling (SEM) - interrelationship between latent structures

Assessment: Comparative Fit Index

Random Forest (RF) identify features that
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Assessment: Out-of-Bag Error Rate







RESULTS

RESULTING INDICES

- Registered with the Chamber of Commerce?
- Has a single tax registry?
- Paid Occupational Risk insurance?
- Paid compensation fund or SENA/ICBF (social and labor welfare)?
- Paid for health or pension in the last month?

- Saves money in a financial institution?
- Accepts online payments?
- Accepts credit cards as payment?
- Applied for a credit or loan in the last year?

- Has an email?
- Uses a cellphone?
- Uses a laptop or tablet?
- Has internet service?

- Has a trade name?
- Has workers who receive payments?
- Has a website?
- Does accounting or daily bookkeeping?

FORMALITY



FINANCIAL FORMALITY '



DIGITAL READINESS

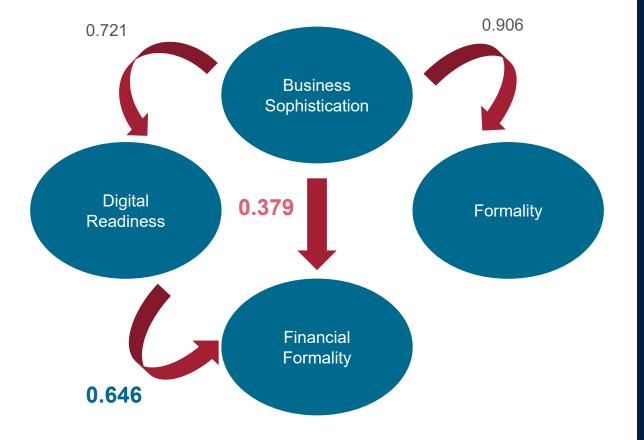


BUSINESS SOPHISTICATION





RELATIONSHIP BETWEEN THE INDICES



Structural Equation Modeling (SEM)



 Multiple regressions between hidden traits like formalization, digital readiness, or financial inclusion — using the data we can observe.

0.972 Comparative Fit Index (CFI)



INDICATIVE POWER AS AN ALTERNATIVE SOURCE

23.73% Outof-Bag Error

Random Forest

Can we predict if a business will invest its loan in business development?

Top 10 Predictors of Productive (Business Investment) Loan Usage





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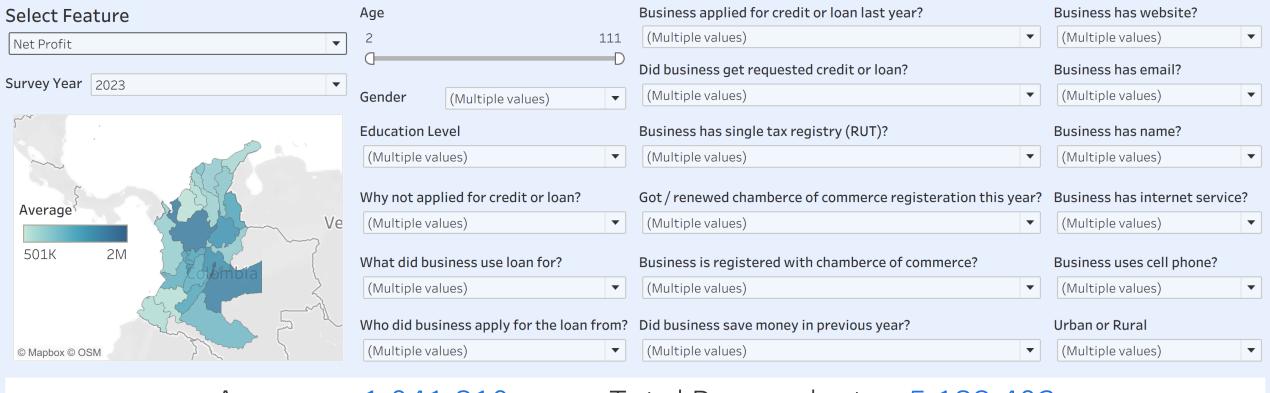
DASHBOARD





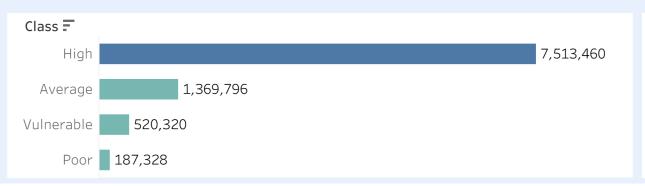
Colombia Microbusiness Credit Scoring Index

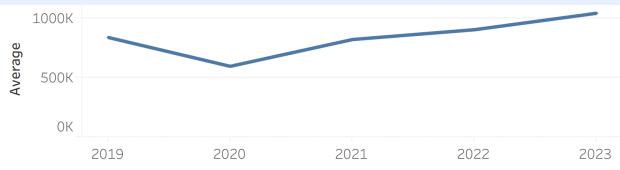
Select the feature you'd like to assess and apply filters as needed



Average 1,041,810

Total Respondents 5,188,402





Formality

0.51/5

Financial Formality

0.50/4

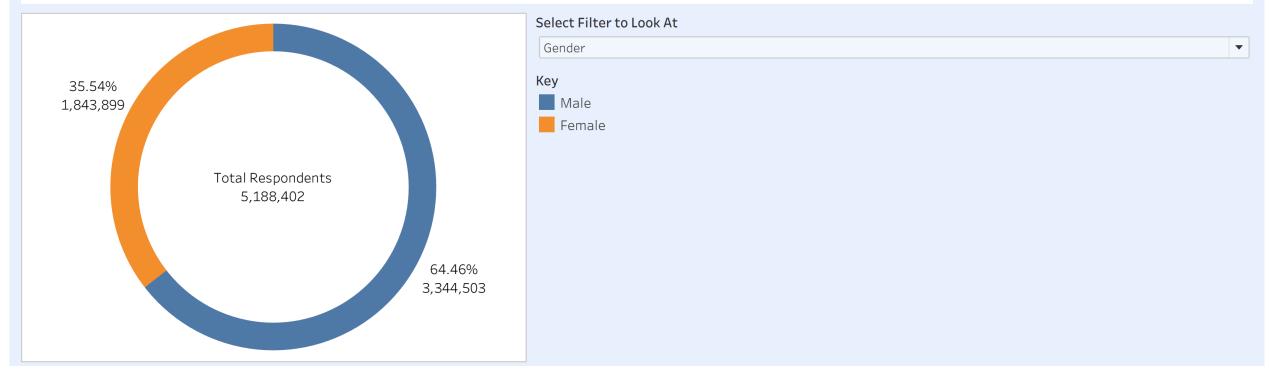
Digital Readiness

1.32/4

Business Sophistication

0.22/4

Optional: Use the chart below to explore the different filters and more









4 POLICY RECOMMENDATIONS







Encourage Digital Transformation



Survey Business Progress, Not Only Status

Promote Formality via
Business
Sophistication



Boost Loan Uptake and Savings Accounts





THANK YOU!

MIT, UNDP, AND OUR MENTOR PROF. ZHENG





