



# Monitoring Consumer Protection Risks in the Digital Age

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March 30, 2022

*Innovations for Poverty Action*

*Consumer Protection Practitioner's Forum*



# Why Market Monitoring?

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Background and Motivation



# Market Monitoring

## Goals:

1. Identify potential consumer protection concerns
2. Inform development of policies and interventions

## Examples:

1. Kenya: IPA with the Competition Authority of Kenya<sup>1</sup>
2. Sierra Leone: IPA with UNCDF & the Bank of Sierra Leone<sup>2</sup>
3. Tanzania: CGAP with Central Bank of Tanzania<sup>3</sup>



<sup>1</sup> Putman, D., Mazer, R., & Blackmon, W. (2021). Report on the Competition Authority of Kenya Digital Credit Market Inquiry Competition Authority of Kenya and Innovations for Poverty Action.

<sup>2</sup> Blackmon, W., Cuccaro, F., Holzinger, A., Mazer, R., Ngwabe, W., & Putman, D. (2021). From the Field to Policy Formulation—How Research is Informing Consumer Protection in Sierra Leone. Innovations for Poverty Action. <https://www.poverty-action.org/blog/field-policy-formulation—how-research-informing-consumer-protection-sierra-leone>

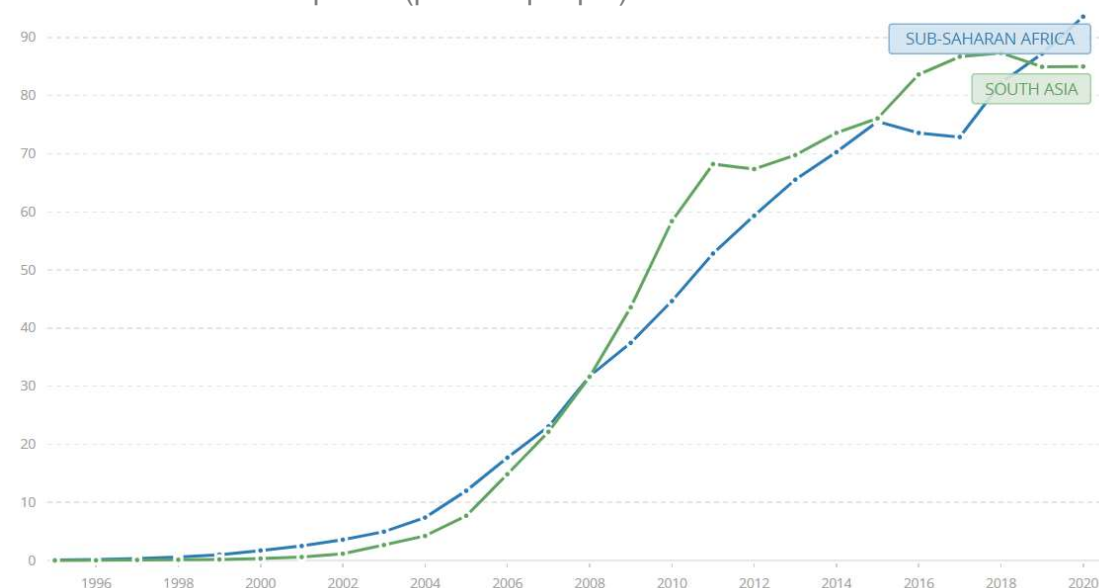
<sup>3</sup> Izaguirre, Juan Carlos, Rafe Mazer, and Louis Graham. "Digital Credit Market Monitoring in Tanzania," no. September (2018).



# Growth in Digital Financial Services

- Large growth in Mobile Money in South Asia and Sub-Saharan Africa<sup>4,5</sup>
- Digital credit has been built on this foundation – strong growth in markets like Kenya where MM adoption was high<sup>6,7</sup>

Mobile cellular subscriptions (per 100 people)<sup>8</sup>



4 Andersson, Simon, and Nika Naghavi. "State of the Industry Report on Mobile Money," GSMA 2021.

5 Demircig-Kunt, Asli, Leora Klapper, Dorothe Singer, Saniya Ansar, and Jake Hess. "The Global Findex Database 2017: Measuring Financial Inclusion and the Fintech Revolution," 2018. <https://doi.org/10.1596/978-1-4648-1259-0>.

6 Anderson, C. Leigh, Travis Reynolds, Marieka Klawitter, Pierre Biscaye, Kirby Callaway, Melissa Greenaway, Daniel Lunchick-Seymour, and Max McDonald. "Review of Digital Credit Products in India, Kenya, Nigeria, Tanzania, and Uganda." EPAR Technical Report, 2017.

7 Totolo, Edoardo. "The Digital Credit Revolution in Kenya: An Assessment of Market Demand, 5 Years On," no. March (2018): 28. <https://www.microfinancegateway.org/library/digital-credit-revolution-kenya-assessment-market-demand-5-years>.

8 World Bank Data: <https://data.worldbank.org/indicator/IT.CEL.SETS.P2?end=2020&locations=ZG-8S&start=1995>



# Benefits have accompanied digitization of financial services

In tandem to increased adoption of financial services:

- Reduced transaction costs<sup>9</sup>
- Resiliency to risk<sup>10,11</sup>
- Poverty reduction<sup>12</sup>

9 Chen, Greg, and Rafe Mazer. "Instant, Automated, Remote: The Key Attributes of Digital Credit." *CGAP Blog*, 2016. <https://www.cgap.org/blog/instant-automated-remote-key-attributes-digital-credit>.

10 Jack, William, and Tavneet Suri. "Risk Sharing and Transactions Costs: Evidence from Kenya's Mobile Money Revolution." *American Economic Review* 104, no. 1 (2014): 183–223. <https://doi.org/10.1257/aer.104.1.183>.

11 Suri, Tavneet, Prashant Bharadwaj, and William Jack. "Fintech and Household Resilience to Shocks: Evidence from Digital Loans in Kenya." *Journal of Development Economics* 153, no. April 2020 (2021): 102697. <https://doi.org/10.1016/j.jdeveco.2021.102697>.

12 Suri, Tavneet, and William Jack. "The Long-Run Poverty and Gender Impacts of Mobile Money." *Science* 354, no. 6317 (2016): 1288–92. <https://doi.org/10.1126/science.aah5309>.



# Digitization is a double-edged sword

- Opportunities for misconduct and fraud<sup>13</sup>
- Products are often not fully understood by consumers<sup>14</sup>
- Aspects of digital credit can heighten risks: speed of disbursement,<sup>15</sup> increases in credit limits<sup>16</sup>
- Product information and experience may help alleviate some risks<sup>17,18</sup>

13 Garz, Seth, Xavier Giné, Dean Karlan, Rafe Mazer, Caitlin Sanford, and Jonathan Zinman. "Consumer Protection for Financial Inclusion in Low- and Middle-Income Countries: Bridging Regulator and Academic Perspectives." *Annual Review of Financial Economics* 13, no. 17 (2021): 1–28. <https://doi.org/10.1146/annurev-fe-2-110110-100001>.

14 Brailovskaya, Valentina, Pascaline Dupas, and Jonathan Robinson. "Is Digital Credit Filling a Hole or Digging a Hole? Evidence from Malawi," 2021.

15 Burlando, Alfredo, Michael A. Kuhn, and Silvia Prina. "Too Fast, Too Furious? Digital Credit Delivery Speed and Repayment Rates." *CEGA Working Paper*, 2021. <https://doi.org/10.11436/mssj.15.250>.

16 Shema, Alain. "Effects of Increasing Credit Limit in Digital Microlending: A Study of Airtime Lending in East Africa." *Electronic Journal of Information Systems in Developing Countries*, no. July 2020 (2021): 1–14. <https://doi.org/10.1002/isd2.12199>.

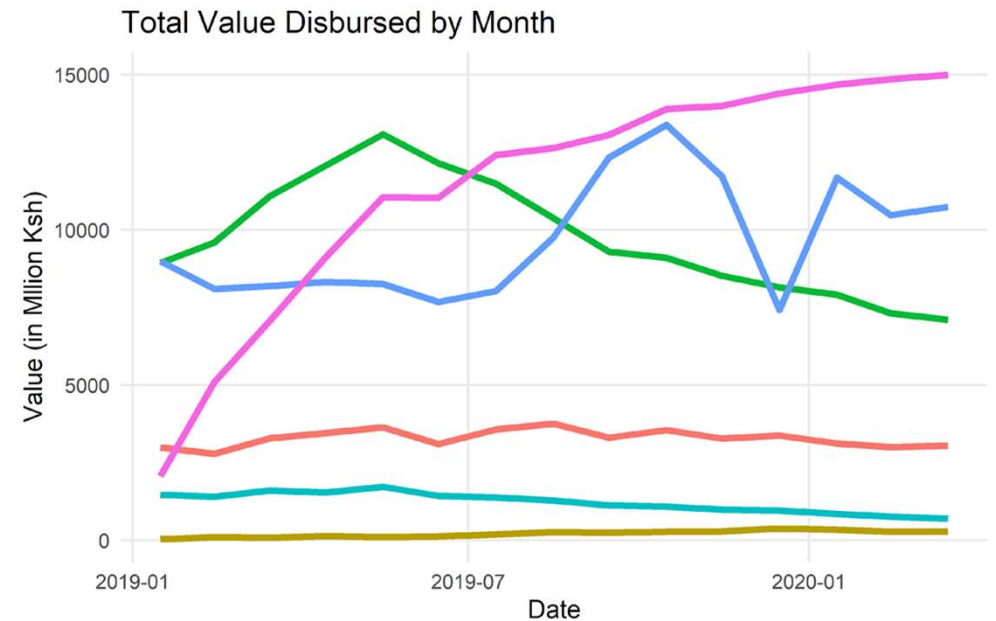
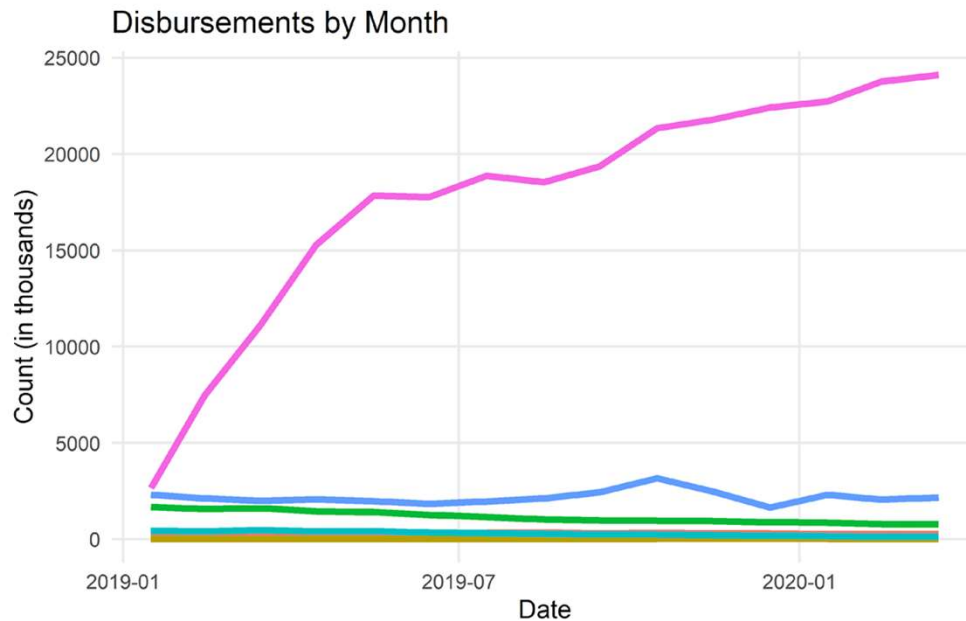
17 Breza, Emily, Martin Kanz, and Leora Klapper. "Learning to Navigate a New Financial Technology: Evidence from Payroll Accounts," 2020.

18 Annan, Francis. "Misconduct and Reputation Under Imperfect Information," 2021.



# Digital Financial Services are still evolving

Market dynamics in Kenya shifted significantly in 2019 with a new overdraft product



Provider: A F H1  
B G H2



# Why use administrative data?

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## ***What is administrative data?***

*Any data that is collected and stored by organizations for operational as opposed to research purposes*

As digital financial services have grown, automation has meant digitized record keeping, which provides an opportunity:

1. Digital credit administrative data has become larger scale and better kept
2. Includes records like account information, disbursements, fees, repayments, rollovers, etc.



## *Advantages of administrative transaction data*

1. Lower **cost** of data collection both participants and researchers
2. Detailed data is generated **quickly** and is more up to date
3. Can also trace out the **evolution** of outcomes over time
4. Better **measurement** of outcomes<sup>19</sup>
  - *Accuracy and precision when studying difficult to recall outcomes*
  - *Avoids social desirability biases*

<sup>19</sup> See also: Feeney, Laura, Jason Bauman, Julia Chabrier, Geeti Mehra, Michelle Woodford, and J-PAL North America. "Using Administrative Data for Randomized Evaluations," 2018.

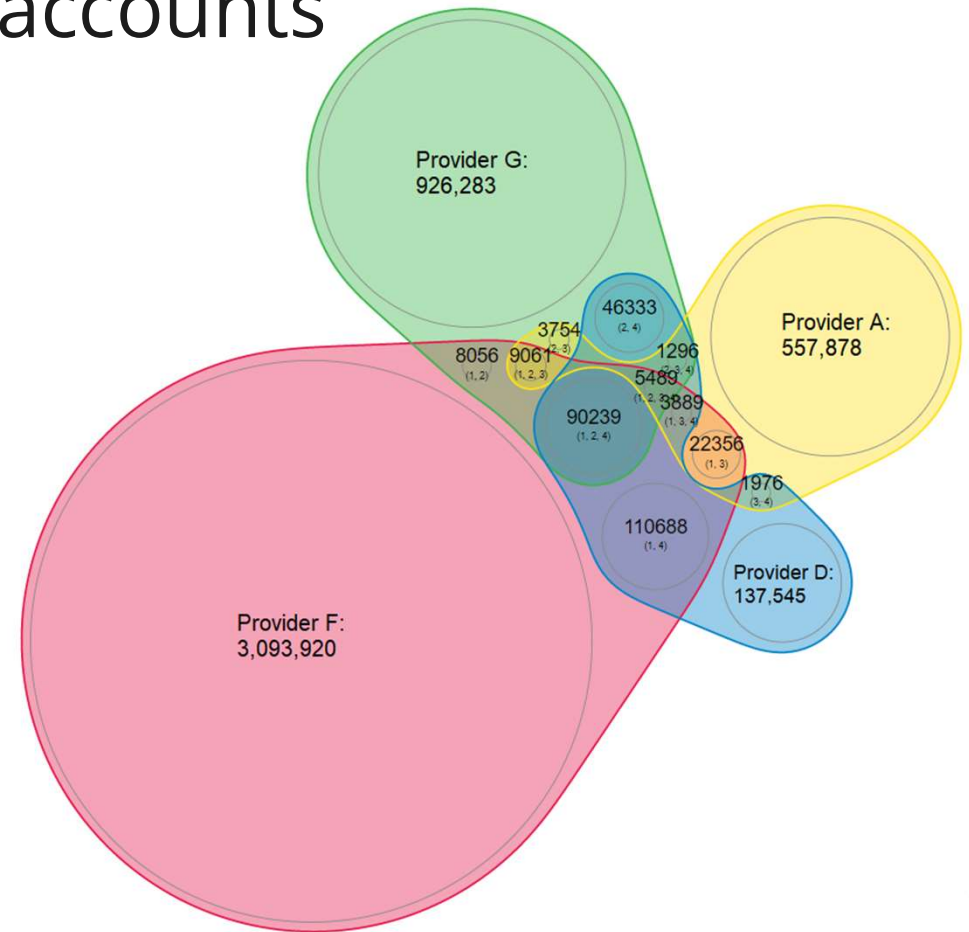


# ***Digital Credit Market Inquiry:***

## Consumers with multiple accounts

Multiple Account Holding in the Digital Credit Market Inquiry:<sup>20</sup>

1. De-identified phone numbers through a common process which allowed us to match them later
2. Found 6% of our sample had multiple accounts



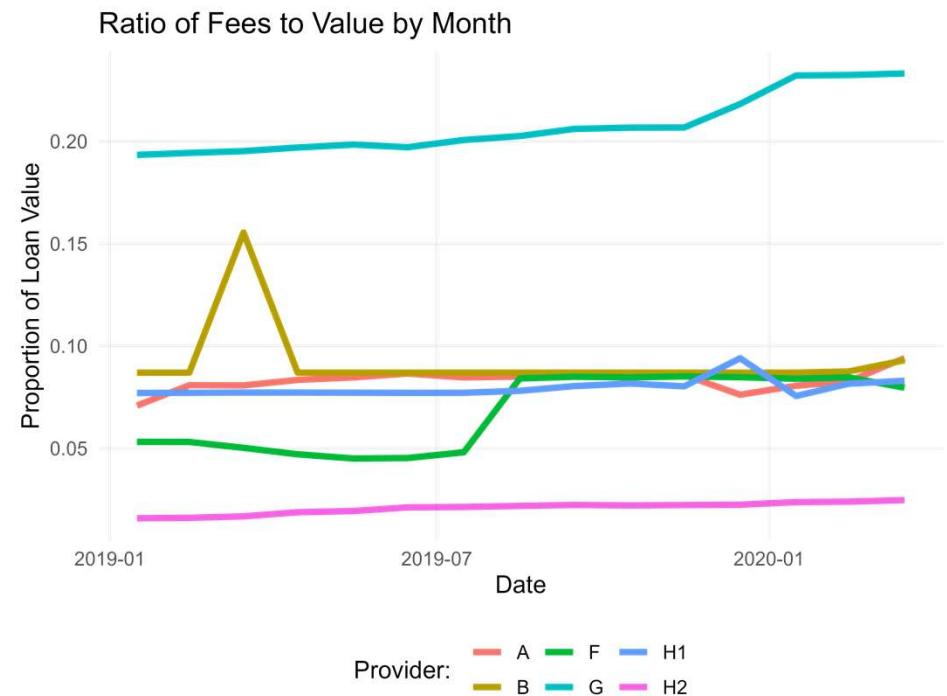
<sup>20</sup> Putman, Daniel, Rafe Mazer, and William Blackmon. "Report on the Competition Authority of Kenya Digital Credit Market Inquiry," 2021.



# ***Digital Credit Market Inquiry:***

## Three Pricing Insights from Fee Data

1. Prices converge among similar lending by deposit-taking institutions (A, B, F, H1)
2. Higher fees at non-deposit taking institution (G)
3. Lower fees (and shorter tenure) for overdraft product (H2)



# Previewing the Toolkit

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A detailed and practical market monitoring toolkit, focused on digital credit



# There are many\* admin data toolkits, how is this one different?

*The toolkit is aimed at addressing the unique opportunities and challenges of using digital credit transaction data for consumer protection market monitoring*

## We view this toolkit as a complement, not a substitute

\*A few handbooks and toolkits on the subject:

1. Cole, Shawn A, Iqbal Dhaliwal, Anja Sautmann, and Lars Vilhuber. Handbook on Using Administrative Data for Research and Evidence-Based Policy, 2020.
2. Feeney, Laura, Jason Bauman, Julia Chabrier, Geeti Mehra, Michelle Woodford, and J-PAL North America. "Using Administrative Data for Randomized Evaluations," 2018.
3. Burke, Laura, Jennifer Cowman, and Heidi Mcannally-linz. "Using Administrative Data for Monitoring and Evaluation," no. February (2016).
4. Brown, Julia, Lucia Goin, Nora Gregory, Katherine Hoffmann, and Kim Smith. "Evaluating Financial Products and Services in the US: A Toolkit for Running Randomized Controlled Trials," 2015.



# ***Digital Credit Outcomes in the Toolkit***

Definitions and interpretations in seven areas:

1. Market size
2. Pricing and fees
3. Loan repayment
4. Competition and concentration
5. Multiple borrowing and switching
6. Over-indebtedness
7. Credit applications



# Level of Aggregation: Digital Credit Data

Topic	Provider/Product	Account	Loan	Transaction
Loan contracts	Average loan size, contracted tenure (when fixed)	(Consumer weighted) average loan size, Distribution of number of loans and loan size	Distribution of loan sizes, tenure, contracted APR	Effective tenure
Pricing and fees	Total cost and per loan cost		APR, Distribution of APR	Effective APR
Repayment behavior	Total value defaulted and outstanding loans		Late repayment, default, rollovers	Detailed repayment behavior: early repayment
Multiple borrowing		Multiple account holding	Multiple borrowing	Loan repayment as a function of taking second loan
Statistics:	Means and totals	Means, SDs, Distributions, Regressions	Means, SDs, Distributions, Regressions	





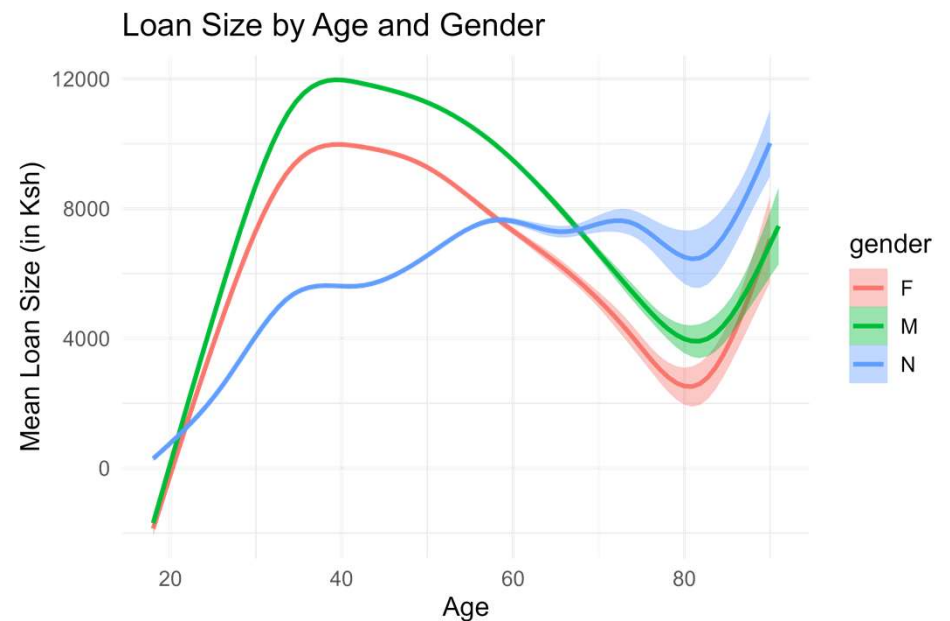
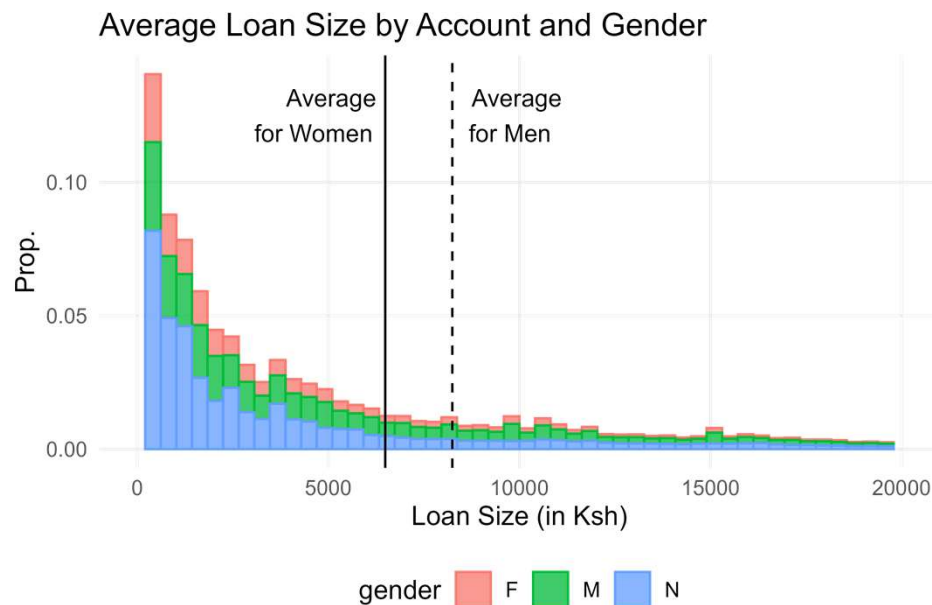
# Consumer Segmentation

Goal	Relevant characteristics	Example method
Explore differences in consumer outcomes by <b>consumer</b> and <b>provider characteristics</b>	<ul style="list-style-type: none"><li>• Gender</li><li>• Age</li><li>• Delivery channel (e.g., USSD/SIM Toolkit v. App)</li></ul>	<p>Plot distributions of outcomes</p> <p>Plot means over continuous variables (e.g., average loan size over age)</p>
Explore <b>seasonality</b> and <b>market evolution</b>	<ul style="list-style-type: none"><li>• Date and time</li><li>• Years, Quarters or Months</li></ul>	Visualize trends over time (e.g., number of disbursements by month by provider)
To segment borrowers into different <b>behavioral groups</b> with different policy needs	<ul style="list-style-type: none"><li>• Consumer protection outcomes, e.g., APR, late repayment, multiple borrowing</li></ul>	Use cluster analysis on borrower outcomes to identify similar groups of consumers



# ***Digital Credit Market Inquiry:***

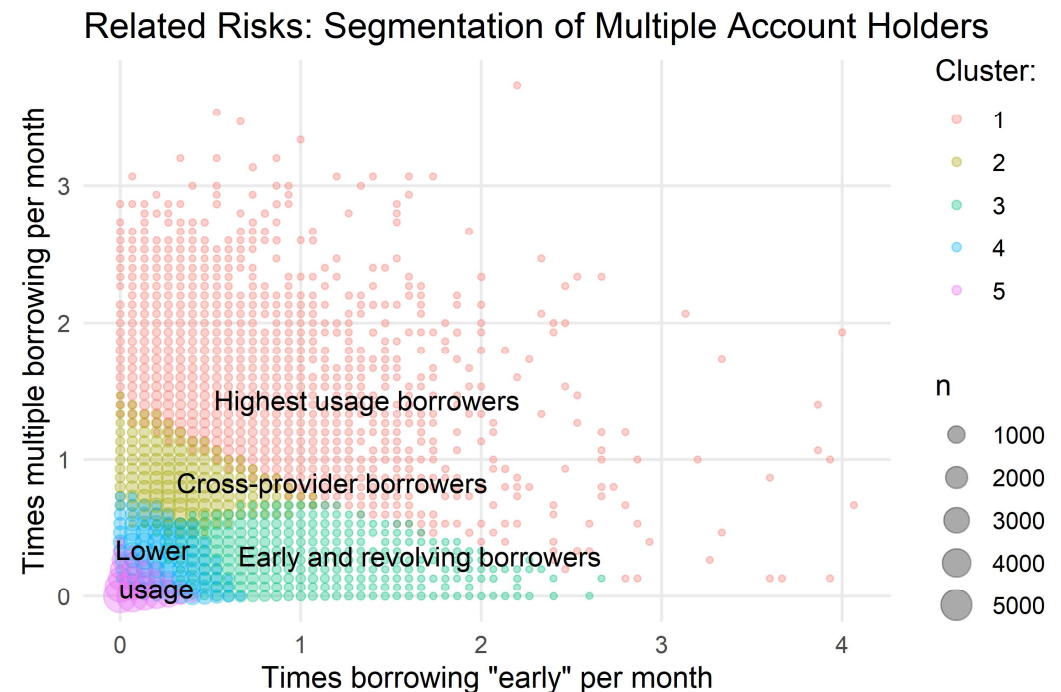
## Average Loan Size by Age and Gender



# ***Digital Credit Market Inquiry:***

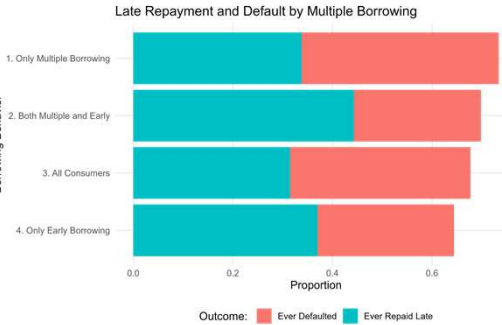
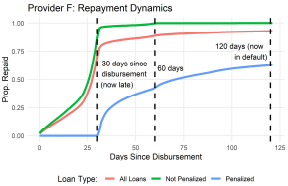
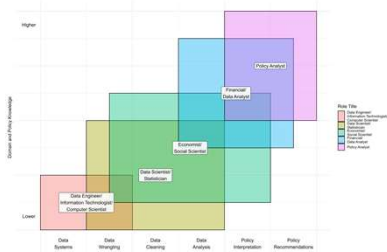
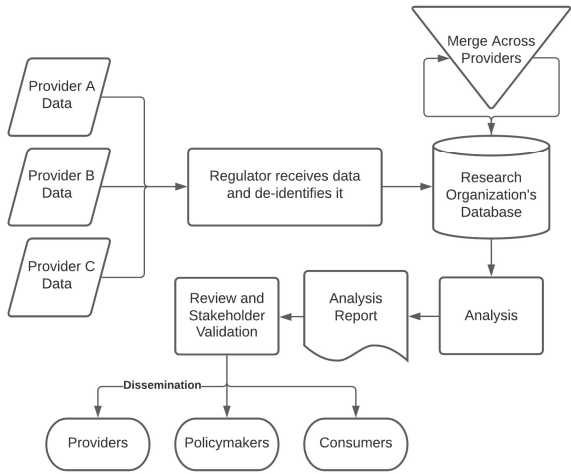
## Outcome Segmentation

- Used borrowing outcomes: number of loans, number of times multiple borrowing across providers, returning early to the same provider, average loan size, etc.
- K-means clustering gives an alternative segmentation approach



# Other Elements of the Toolkit

- 1. Resources for more in-depth analysis of transaction data
- 2. Resources for data security and processing
- 3. Request and planning documents



```
## 3.1.1 Input Format Read
# Specify file format as xls, xls, csv, or txt
file_format = "csv"

## 3.1.2 Output Format Read
# Specify output file format as xls or csv
output_file_format = "csv"

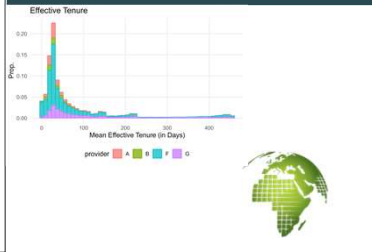
## 3.1.3 Filenames Read
# We add the file names() to quotes below, separated by commas, without the
# file extension. If there is only one dataset, then only include one entry in
# the list.
filenames = c("")

# We also have written a function read_filenames() that automatically
# collects all files with the format specified above. To use, uncomment the
# function below and comment the line above. Please situate the files in a
# directory (input_path) with no other data in it.
filenames = read_filenames(input_path = input_path, file_format = file_format)

## 3.2 Variables Read
# Please add desired variable names in quotes, separated by commas in the
# code line above. In the data, there should be one name for each column in
# the data. Please be sure to title variable containing date of birth "dob" and
# variable containing phone number "phone".
vars_names = c("row_number", "account_id", "msidn", "dob", "gender",
              "loan_id", "type", "interest_rate", "credit", "debit", "date")

## 3.2.2 Variables to Drop Read
# These are variables that contain personally identifiable information. This
# should include "dob" and "phone" if present as well as other sensitive data
# (e.g., account number).
dropvars = c("msidn", "dob", "account_id")

## 3.2.3 Date of Birth Read
# We please specify the date format for date of birth by writing a three letter
# string where 'y' is year, 'm' is month and 'd' is day. It should look like one
# of the following: ymd, ym, ymd, ymd, ymd, or dmy. Note: if the month is written
# in text format please replace "w" with "b" for abbreviated months ("Jan")
# and "B" for full month names, e.g., "ymd" becomes "ydb" or "ydb".
date_format = "ymd"
```



	A	B	C	D	E	F	G	H	I	J
	gender	dob	id	loan_id	transaction_type	transaction_date	debit	credit	fees	msidn_prefix
1			efa908ba3f2834e7157ead16f9891f814a7cb87592c30d7efa409fe24cc59ff	4a4042cb3c7	disbursement	4/1/2018	974.78		236.27	915
2	Female	1949	17d3c81c925667a61d03f636bb8780c72607117ab2a38cd049898993c75d4	e6068a08031	repayment	4/1/2018		1681.87	333.99	382
4	Female		e8b2b080d4a724edc7898e6510ed0426a02c13ba309206df5c0ab3311a3e	52e76782786	repayment	4/1/2018		1831.92		699
5	Female		8697777141ec570969be9a4a012ab8c02a68509b2290Aee48a7a1ecf1034c3	66e9bdf530v	disbursement	4/1/2018	1196.18		452.56	634
6	Female		d7023372525bb770e9f62a76582f366cba563cb3f9bb76c2109ce2873ddc0b7d	b3f6bad9a6b	disbursement	4/1/2018	1295.60		538.65	205
7	Female		1970_d452eaf80cca98acbe5a8aea075dc84fd245f8304d4dec5b7f279887d33067	c45a0a2290e	disbursement	4/1/2018	1081.09		408.61	794
8	Female		1962_dff3e5b2ce3c7b4b0809c92f852bb4d60bfc51e5842733d5a94077e00b461	178db1ab31c	disbursement	4/1/2018	900.98		343.13	443
9	Female		c96b520fde7a19d5ac987c38ea0205e7735b02e5f0933ab36486038684ce8	ad5a038050e	repayment	4/1/2018		1549.33	310.11	974
10	Female		1948_cca01ab4c551c817b0586c83c75b13414d8fe3d44e28d745f989b495912222	2607b17feca0	disbursement	4/1/2018	1221.80		476.36	714
11	Male		43b5d766dc3c621f8a0f8b0d82b5a9670376803299cc31824e98697cd1952397	119ca1612a	disbursement	4/1/2018		282.57		590
12	Male		1992_27c23438d6a6f8e4bc2b55cd02ac73b9758b1284a49b81b98113cde3c91063	6c8cf7bdd141	repayment	4/1/2018		1233.04	346.42	682
13			1945_18f2a5605c12fcd8fd5e689c342b0888e8ba4c1cf11969026eb84605b19a1	6827a5a3c39	disbursement	4/1/2018	1218.73		432.60	858

[DATE]  
[PROVIDER CONTACT  
PROVIDER NAME  
PROVIDER ADDRESS]

[AGENCY] is undertaking an analysis of [COUNTRIES] digital lending market. Digital credit is an important service for millions of consumers, but also raise consumer protection risks which must be mitigated through industry and policy measures. [AGENCY] is developing new approaches to data analysis and reporting to help the industry better understand and develop consumer protection solutions for this growing sector.

This analysis is being undertaken in collaboration with [PARTNER]. [description of PARTNER]. The analysis includes measuring the advancement of the market and recent trends in its evolution. To this end we are requesting the provision of transaction-level data on borrowing and

## ***When to consult the toolkit:***

1. When determining if administrative transaction data is the right tool for your work
2. When planning to obtain and use transaction data for market monitoring



# IPA's Consumer Protection Research Initiative

Innovations for Poverty Action (IPA) is a research and policy nonprofit that creates and shares evidence, while equipping decision-makers to use evidence to reduce poverty.

## Two research methods:

- Data collection, analysis and market monitoring
- Impact evaluation testing of new consumer protection solutions

## Four key risks:

- Fraud in digital channels
- Consumer redress and complaints handling
- Product information and consumer choice
- Over-indebtedness



**Bangladesh**



**Kenya**



**Nigeria**



**Uganda**



## *Is administrative data the right tool for me?*

1. Does my organization have the staffing and technical capacity to support obtaining and analyzing the data?
  - Data validation, cleaning, wrangling, analysis, computing, etc. - talk to your technical staff
2. Can I measure the relevant outcome with administrative data?
  - Transaction data is useful for risks like high prices, defaults, debt, multiple borrowing
  - The evolution of these outcomes in addition to segmentation
  - Not the tool for every risk (e.g., explicit misconduct)

