# **Complaints Data as a Tool for Consumer Protection: Lessons from Uganda**

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#### **Outline**

- 1. Potential of complaints data analysis for consumer protection
- 2. Digital Financial Services and the Uganda Communications Commission
- 3. Process and Methods
- 4. Emerging Insights: Market Scope, Segmentation, Challenges
- 5. Next Steps



# Complaints data are a rich source of consumer protection insights

- Complaints data are accessible sources of consumer insights that can be leveraged by Mobile Network Operators (MNOs) and Regulators
- Complaints collected by MNOs offer critical operational data to understand consumer issues with products and customer care
- When consolidated and monitored by regulators, complaints offer market-level observational data serving several key use cases:
  - 1. Monitor real-time issues/events to detect hot spots
  - 2. Track the development of issues in the Digital Financial Services (DFS) ecosystem
  - 3. Test and evaluate policy interventions



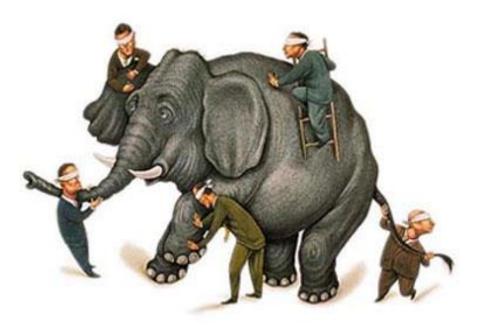
### Uganda, Digital Financial Services, and the UCC

- Uganda Communications Commission (UCC) receives MNO complaints databases to monitor the market and identify areas for improvement in customer products and services
- Uganda has a concentrated DFS market two firms account for 95% of mobile and mobile money subscriptions
- But difficult to scope and diagnose the DFS market: Who makes complaints? What do they complain about? When and how do they complain?
- Diagnostic lays **groundwork** for (a) immediate actions (b) evidence-based policy experiments

## How to use MNO consumer complaints data?

Data Triangulation Strategy

Parable of blind men and the elephant



- Consumer complaints data reveal part of the elephant
  - Analysis of anonymized complaints data sets submitted by MNOs to UCC
  - Variation in types of data MNOs share, as well as way MNOs organize these data
- Conducted a **random digit dial (RDD) phone survey** with 830 DFS subscribers to seek market-level view
  - Triangulated with admin (SIM and Mobile Money subscriptions) and Finscope data
- Related IPA Study on alternative redress mechanism:
  Consumer protection monitoring via social media content of Ugandan financial service providers, including MNOs, from July 2019 June 2020 (https://www.poverty-action.org/event/ipas-consumer-protection-research-initiative-holds-first-practitioner%E2%80%99s-forum-meeting)

# Complaints Data: January 2019 – August 2020

Provider	Months submitted (1/2019-8/2020)	Total # of Customer	Average # of complaints	Complaint	# Complaint categories	
	(1/2013 0/2020)	care logs	per month identified	identified	2019	2020
MNO-1	20	688,976	34,449	4	26	143
MNO-2	16	2,279,691	142,481	36	18,576	71
MNO-3	17	734,124	43,184	9	34	25
MNO-4	3	4,107	1,369	5	20	Not received



# **Consumer Complaints Data - Analytic Approach**

5 Methods, Multi-phase, Iterative process

#### **Data Collection and Processing:**

Collect 2019-20 MNO complaints data, pre-process and seek standardization

#### **Structured - Exploratory Data Analysis (EDA):**

First Contact Resolution (FCR) rates, Service Level Agreements (SLAs), top categories/subcategories, statistics by channels, categories, months and day of week.

#### **Unstructured - Exploratory Data Analysis (EDA):**

Most frequent words overall and by categories, subcategories, and channels.

#### **Topic Modeling:**

Reduce dozens and even thousands of categories to optimal groups

#### **Predictive Modeling:**

Leverage demographic data to understand complaints behavior, based on 2020 data



## **Market Survey Design and Implementation**

#### **MNO Products & Services**

Voice/SMS/data, mobile money, mobile banking, & mobile loans

**Challenges: Products & Services** 

Experiences & risks w/ mobile money, mobile banking, & mobile loans

Demographics, Trust, Financial Self-efficacy, COVID-19 issues

**Consumer redress channels** 

Use of customer care & complaints resolution

Fraud experiences

Experiences with phone-based fraud and responses

- Conducted in August-September 2020, as COVID-19 restrictions eased
- Sought representative sample of active adult DFS users (18+, use in last 90 days, region and education quotas)
- Divided into two rounds within one week, n = 830 (R1), 771 (R2)



# **Regional Distribution of DFS Users**

Geography



Census: 25.4%

Finscope (MM): 12.3%

Survey: 11.4%

#### Western

Census: 19.8%

Finscope (MM): 27.0%

Survey: 20.8%



Census: 19.8%

Finscope (MM): 20.7%

Survey: 17.3%

#### Central

Census: 30.2%

Finscope (MM): 40.0%

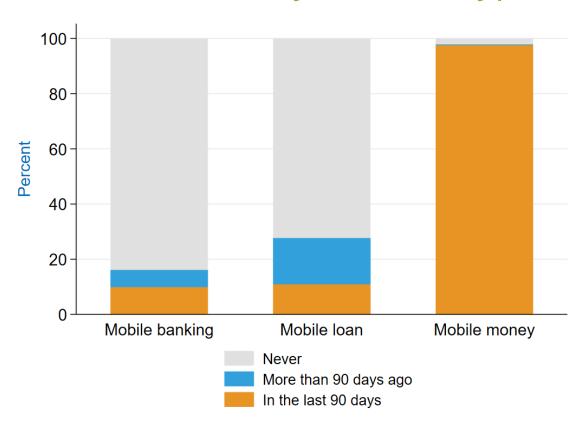
Survey: 50.5%



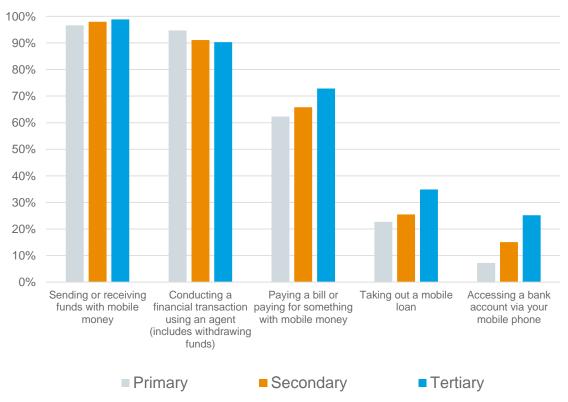
### High levels of access, varied penetration of DFS products

Reported by Survey Respondents

#### DFS Use by Product Type



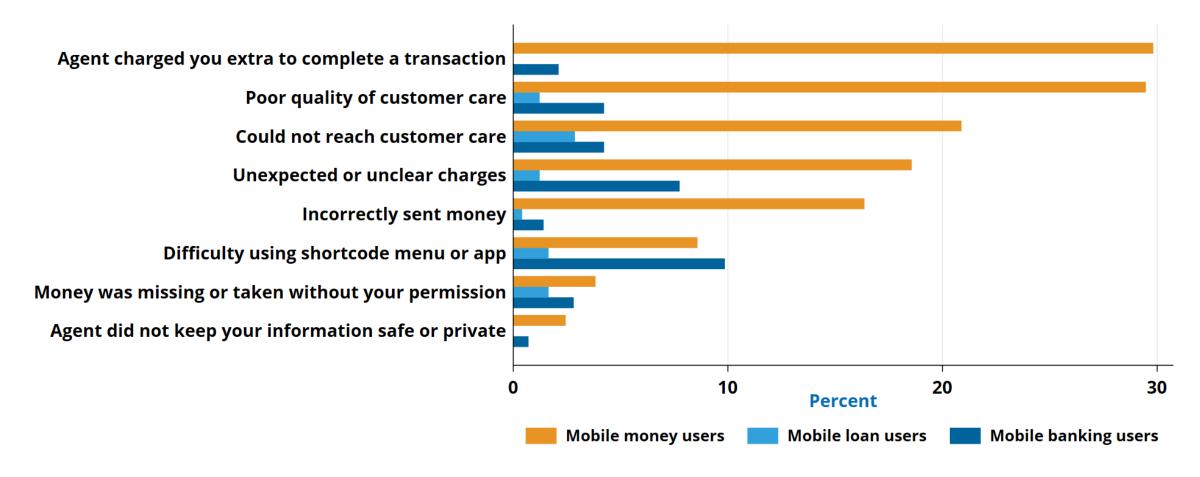
#### DFS Usage by Education



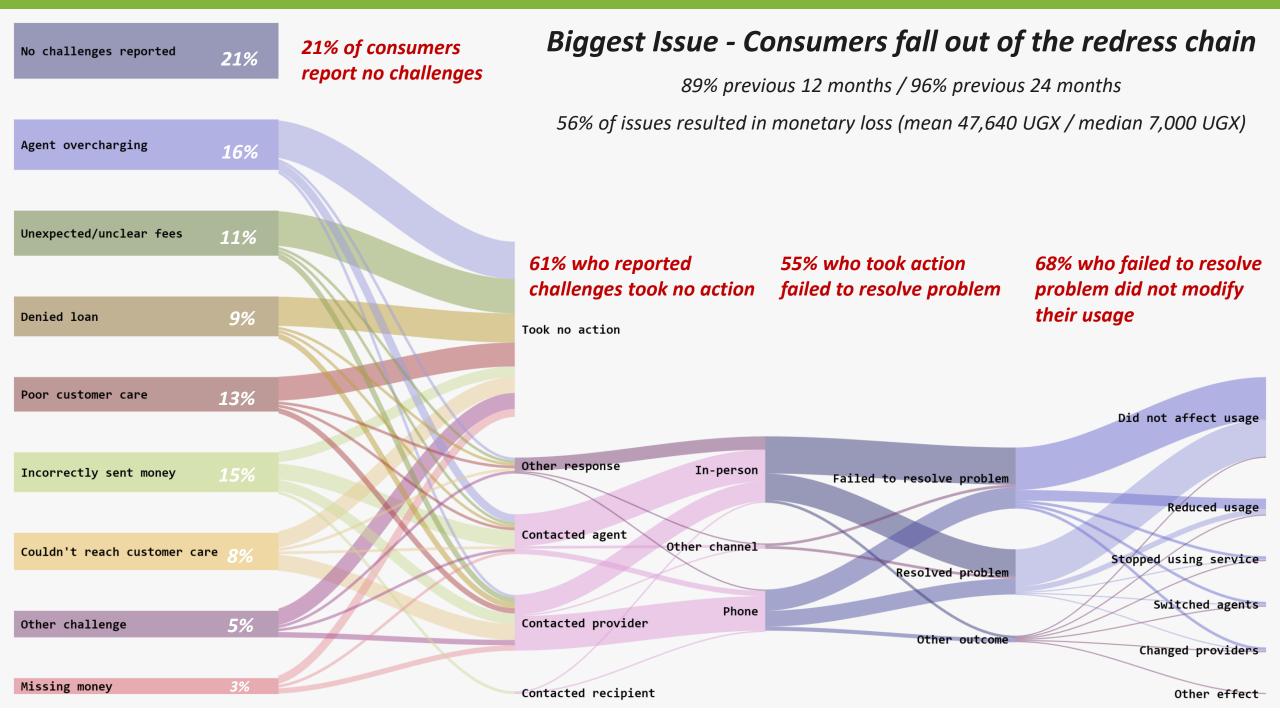


# Fees and customer care are biggest challenges

DFS Challenges reported by Survey Respondents

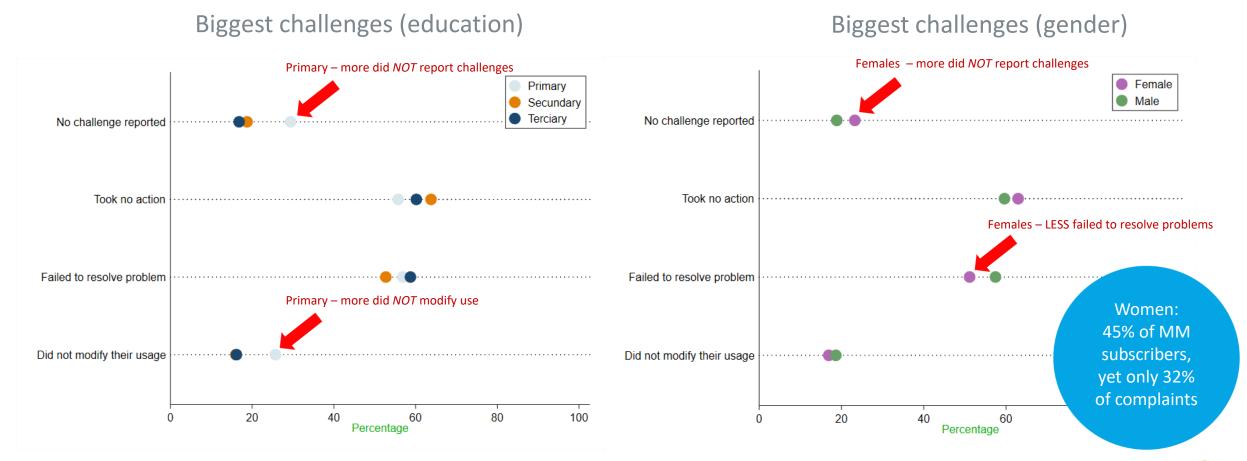






#### Primary educated and women appear to have less redress issues

Differences in usage-based risk or less aware?

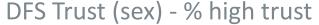


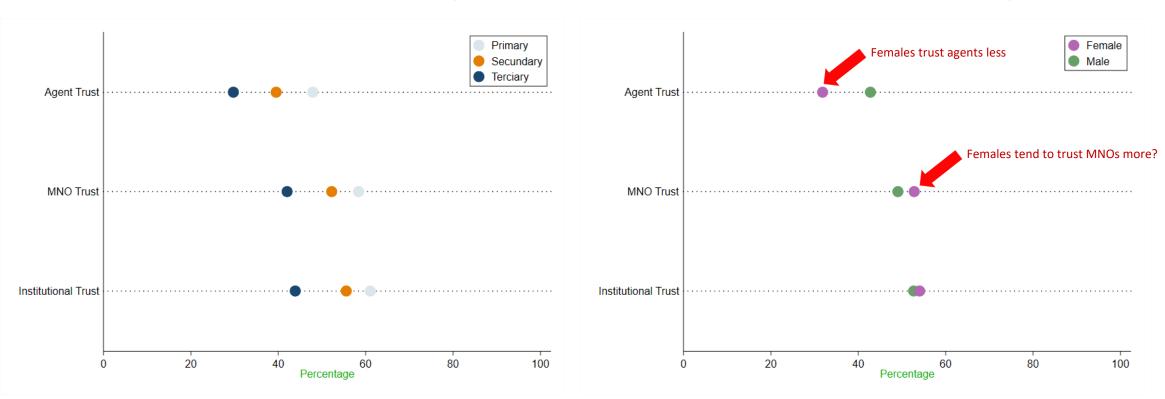


#### Less education, more trust - while women trust less agents

Differences in usage-based risk or less aware?



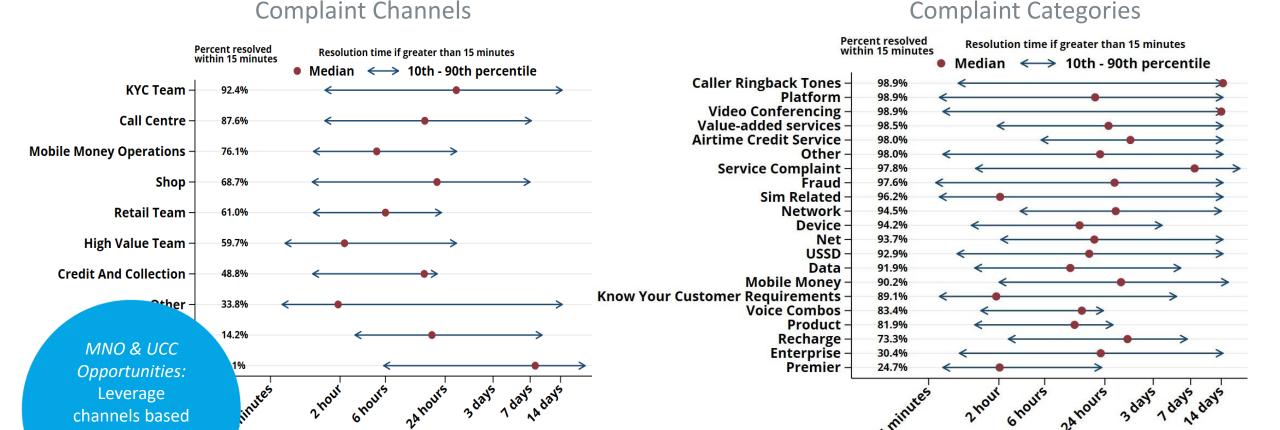






#### Variation in resolution times by complaint channel and MNO

Structured EDA of Complaints Data



channels based on complaint and

> consumer segments

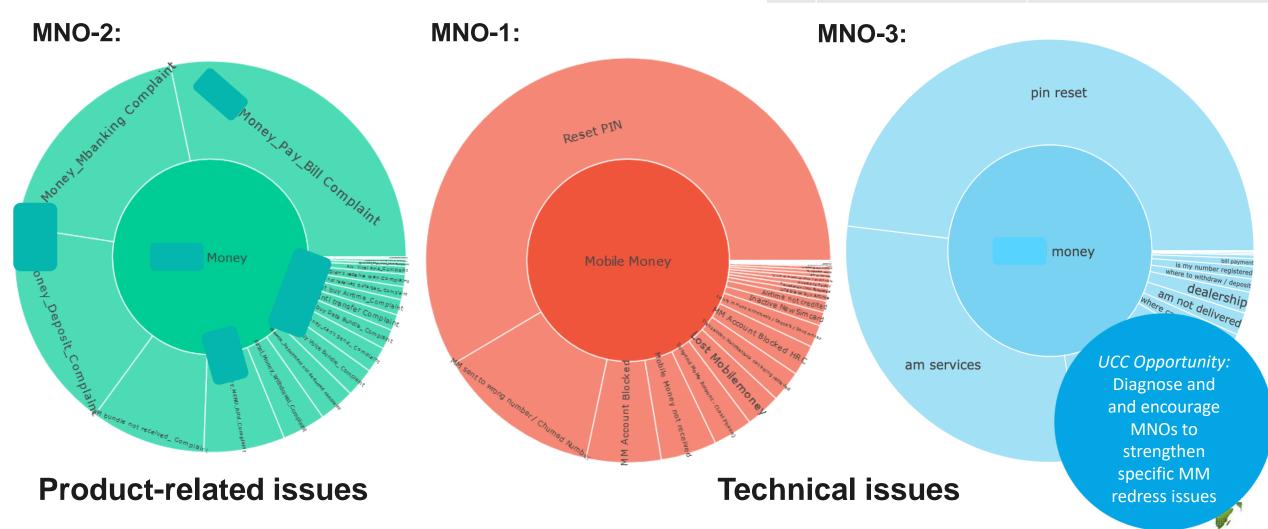
Resolution times (log scale)



Resolution times (log scale)

### Mobile Money Example: Variation in Key Issues across MNOs

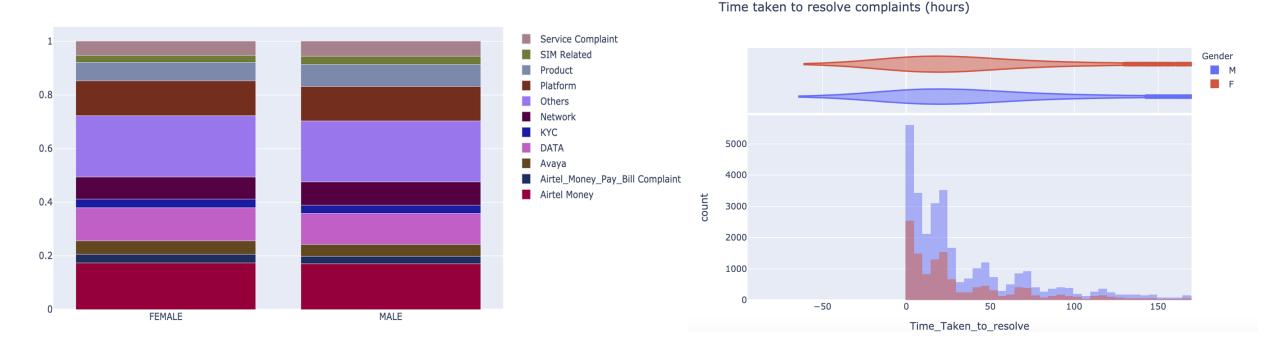
Most common customer issues in 2019					
	MNO 1	MNO 2			
1.	Mobile money	Mobile money			
2.	Data	Prepaid queries/problems			
3.	Network issues	Lost airtime			



### Women complain about same things but complain less

Examples from Structured EDA of Complaints Data

Complaints by category and gender



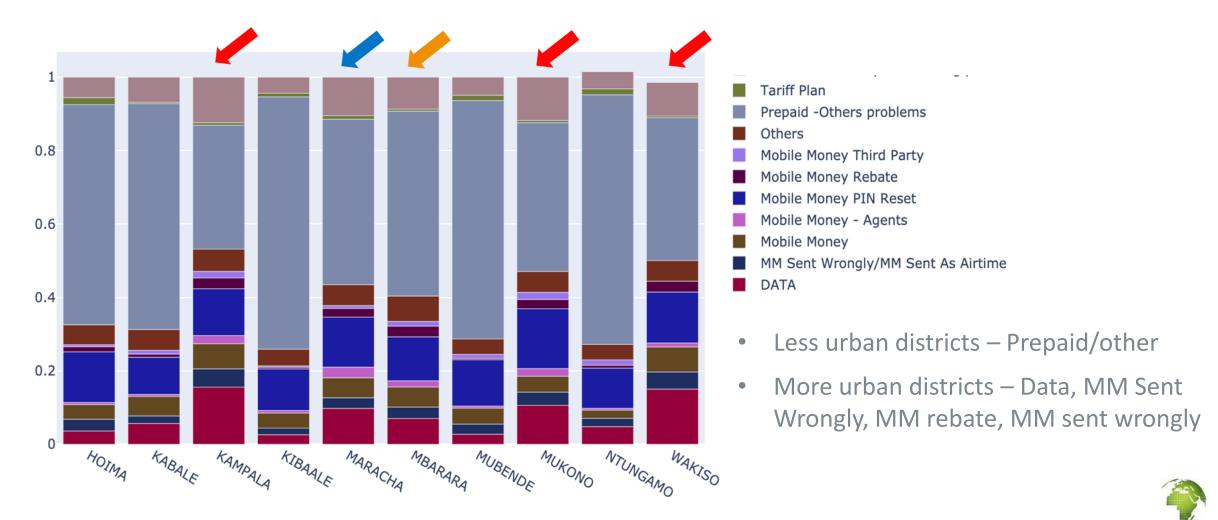
- No *visible* differences in main categories of complaints
- Women are 45% of subscribers yet account for only 32% of complaints
- Women have faster first contact resolution times (FCR) across MNOs Why?



### Differences in distribution of complaint types across geographies

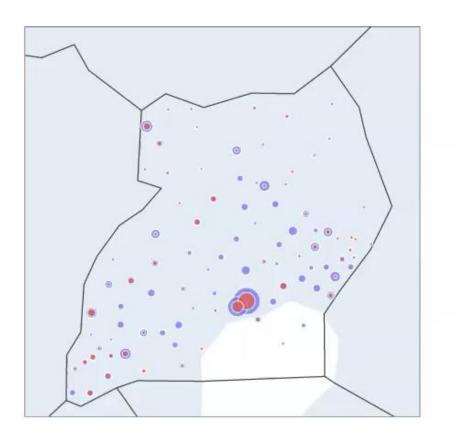
Example from Structured EDA of Complaints Data

Complaints by category and location



## **Complaints across Time and Geography**

Video Demonstration



Gender



D F

UCC and MNO
Opportunity:
Develop an MNO or
market-wide hotspot and alert
system, which
automates responses

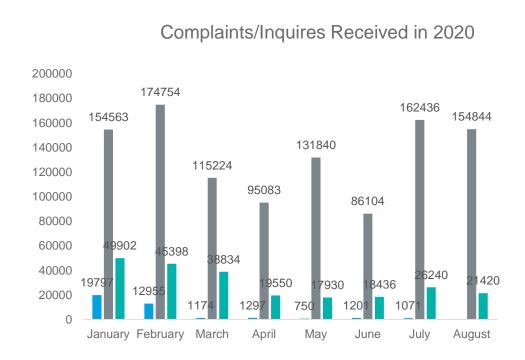
Date=2020-01-01



### Changes in complaints could be capacity and channel issue

#### Structured EDA of Complaints Data

Fewer complaints March-June 2020 (coinciding with most restrictive COVID-19 lockdown period





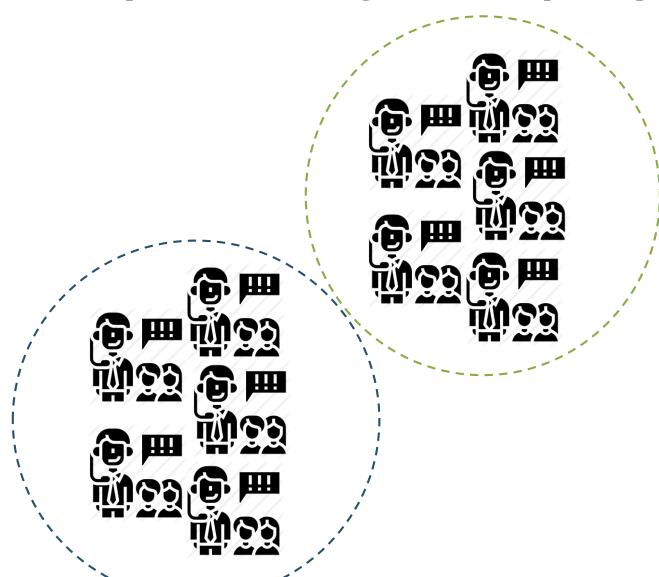
https://www.poverty-action.org/event/ipas-consumer-protection-research-initiative-holds-first-practitioner%E2%80%99s-forum-meeting

However, the reduced volumes may be misleading ...

- Social media data shows rise in complaints to MNOs—especially for customer care matters
- Likely drop off is due to reduced call center staff not fewer consumer issues



## **Topic Modeling Can Help Improve Comparative Analysis**



- Objective is to group similar complaints under the same topic, given their description
- Train topic modeling algorithm to find the topics given complaints description.
- Find the best number of topics
- Understand what the topics represent
- Basis for standardization for MNOs and UCC

MNOs	Category Reduction
MNO-1	35 - > 7
MNO-2	3,797 -> 13
MNO-3	35 -> 7
MNO-4	20 - 4

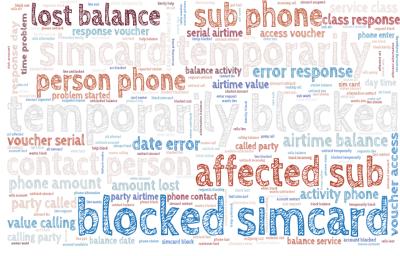


# **Topic Modeling: 3 Examples from an MNO**

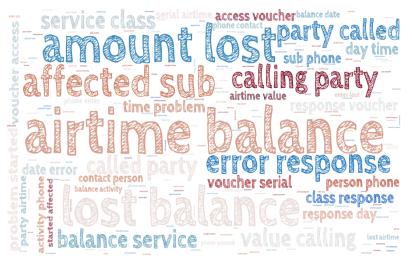
#### **Errors in sending MM**



## Blocked sim card



# Lost airtime balance



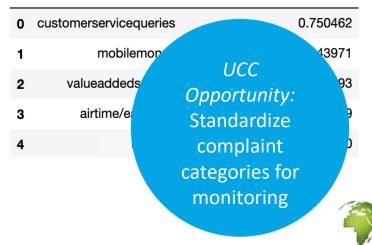
#### index CATEGORIZATION\_TIER\_1

0 customerservicequeries	0.473322
1	0.330612
MNO	0.105554
Opportunity: Create most	0.055734
appropriate	0.031017
categories for	
CRM, voice	
transcription opp	

#### index CATEGORIZATION\_TIER\_1

0	customerservicequeries	0.372885
1	prepaidqueries	0.328923
2	mobilemoney	0.182030
3	voicecalls/sms	0.093154
4	internet	0.017241

#### index CATEGORIZATION\_TIER\_1



# What if we wanted to predict a fraud complaint?

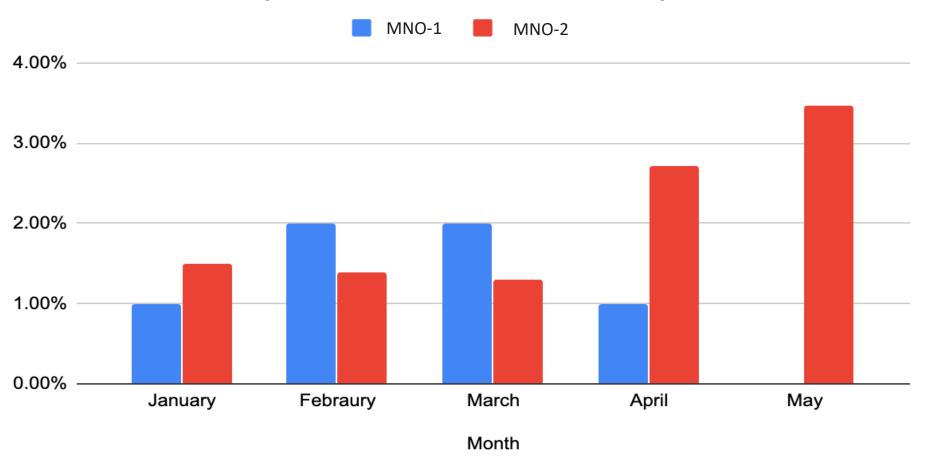
#### Predictive Model Example

- Survey: Nearly 90% had experienced a scam/fraud attempt in last 12 months
- Survey: Consumers susceptible during the pandemic, 75% report making less income during the pandemic, 30% reported loan repayment issues
- Survey: 46% reported receiving a COVID-19 related fraud/scam call
- Survey: 49% of scammers claimed to be the MNO, 24% didn't id themselves
- Complaints data: Better prediction would enable:
  - Preventive efforts to mitigate or eliminate the fraud threat
  - Customized response to the complaint



# Complaints data tell different stories about fraudrelated issues during COVID-19

% of total complaints as fraud-related complaints in 2020

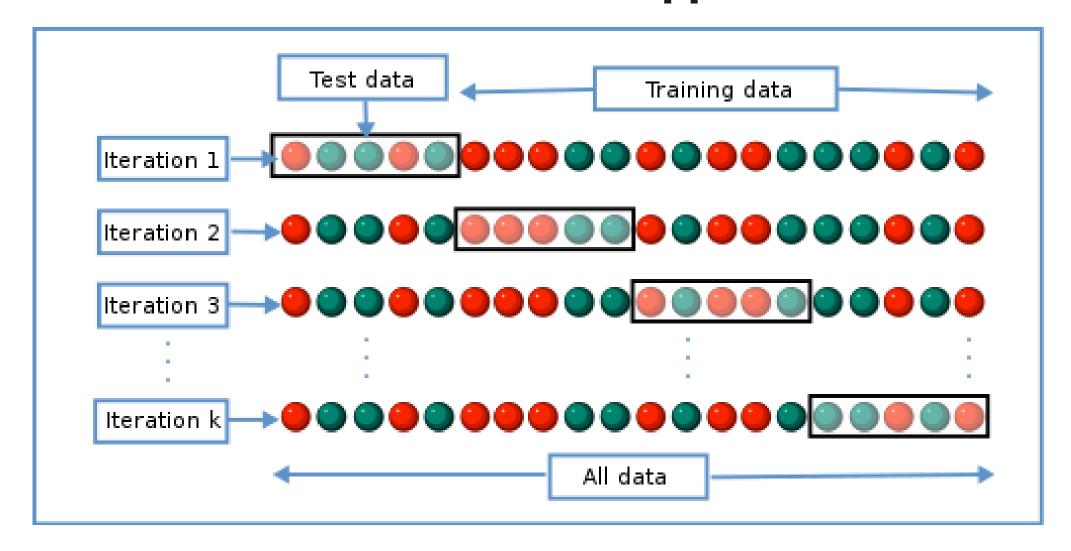


Is this a difference in providers - or a difference in reporting and classification?

Remember:
Evidence complaints
went up in social
media



# Predictive Modeling Machine-learning Algorithm with a Cross-validation Approach





### Different variables predicted fraud-related issues for different MNOs

#### **MNO-1 Fraud**

#### 96% precision

Most important variables to predict a fraud call, ranked:

- 1. *Hour*: Complaints **LATER** *in the day*, more likely to be fraud\*
- 2. *Month*: Complaints in **EARLIER** *months* more likely fraud\*
- 3. Time-as-client: LESS time as MNO client, more likely fraud
- 4. Day: Days earlier in the month, more likely fraud
- 5. *Gender*: *Women more likely* to file a fraud-related complaint
- 6. Location: More likely fraud from Kampala and Wakiso

If female who is a new MNO-1 client from Kampala calls late in day in early January ... probably calling about fraud

#### **MNO-2 Fraud**

#### 84% precision

Most important variables to predict a fraud call are, ranked:

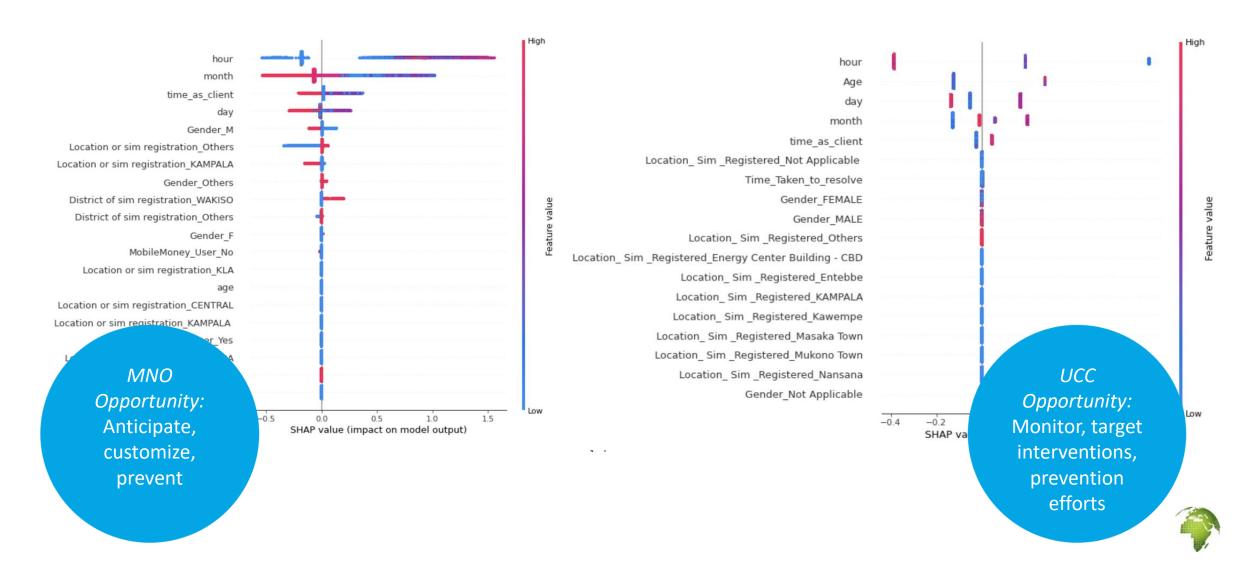
- 1. *Hour*: Calls made **EARLIER** *in the day*, more likely to be fraud\*
- 2. Age: **Older people** more likely to final fraud complaint\*
- 3. Day: Days **earlier in the month**, evidence at the end\*
- 4. *Month*: **MIDDLE** *months in timeframe* more likely fraud\*
- 5. *Time-as-client*: **MORE** *time as MNO client*, more likely
- 6. Location: More likely fraud **from Kampala**

If older, established MNO-2 client from Kampala calls earlier in day in early April ... probably calling about fraud



#### **MNO-1 Fraud**

#### **MNO-2 Fraud**



## **Takeaways and Next Steps**

- Complaints data are accessible and rich source of insights for MNOs and regulators
- Seek to triangulate the complaints data
- MNOs can use complaints to (a) anticipate, (b) customize, and (c) prevent
- Regulators can use complaints as a market monitoring tool and as basis for evidence-based policy interventions
- Next steps:
  - Develop a comprehensive and standardized template for use after January 2021
  - Train UCC staff on data analysis to generate monthly statistics
  - Consider experimentation to address consumer challenges



# Thank you



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