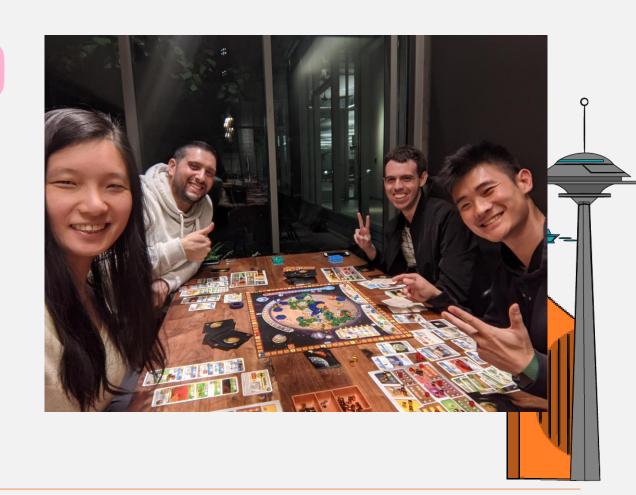






### Meet the author

- Senior Fraud Analyst at Klarna
- Passionate about data quality
- Hobbies
  - organize AI/ML developer meetups
  - boardgames
  - Toastmasters

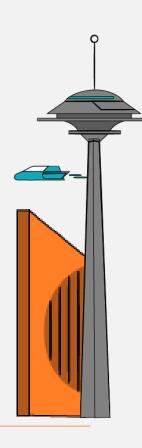






### **Outline**

- Why should we care about data quality in Credit Card Fraud Prevention?
- What data quality problems do we encounter?
- How shall we solve the problems?

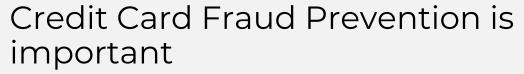






# Why should we care about data quality in Credit Card Fraud Prevention?





 \$10 billion in consumer losses from credit card frauds in the US ~ 150k US households' annual expenses



This domain heavily relies on data

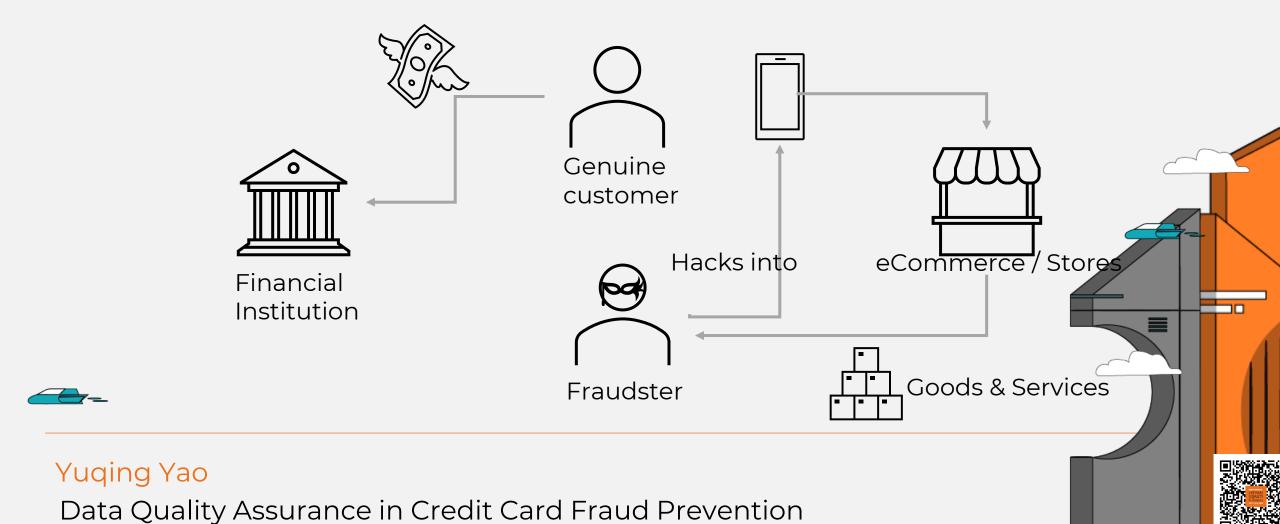
- Behavioral analysis
- Fraud attacks
- Investigations

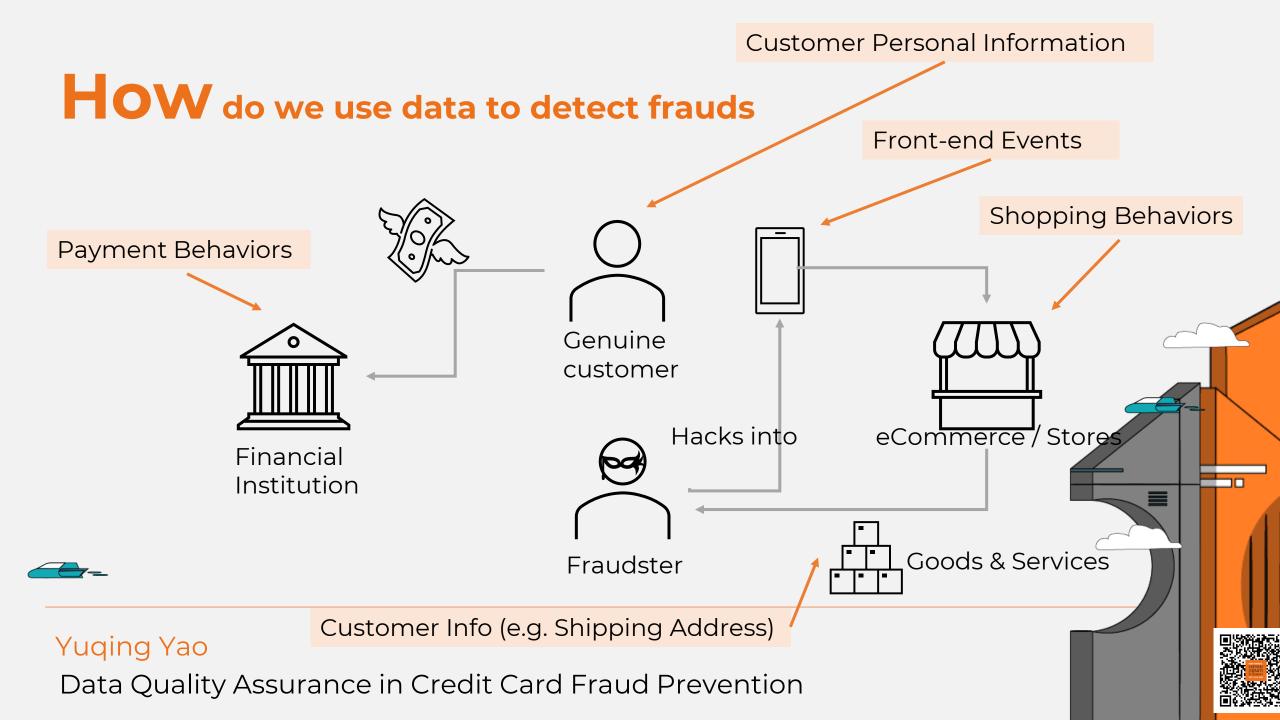






### **HOW** are frauds conducted





# What data quality problems do we encounter?

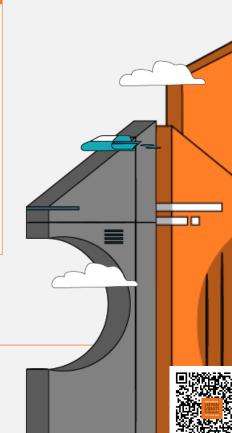
- Types of data used
  - Customer Personal Information
  - Transactions
  - Front-end Events
  - Behaviors
  - External Data Sources
  - Disputes

#### By Nature of Data

- Imbalanced Data
- Wrong Labels
- Incomplete Data

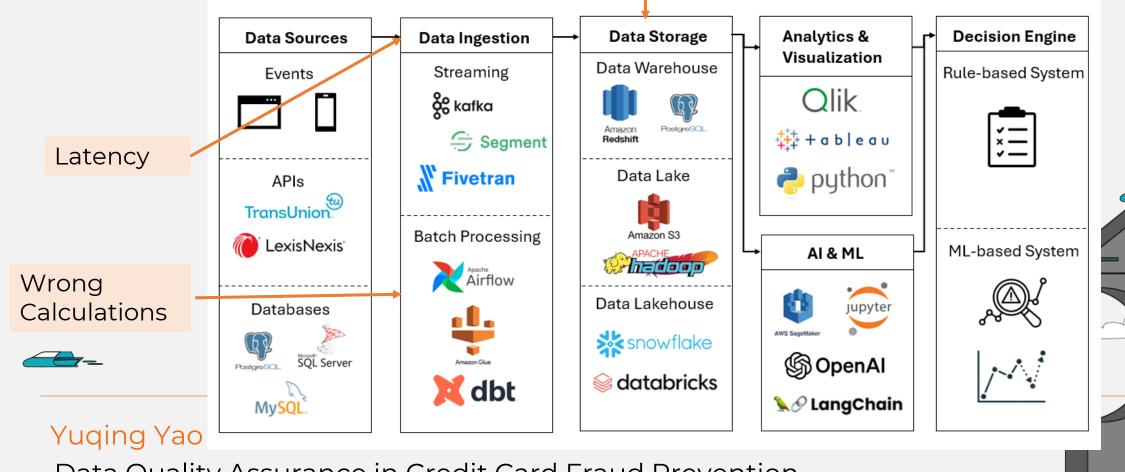






# What data quality problems do we encounter?

• Data Stack – Procedures in Cred t Card Fraud Prevention

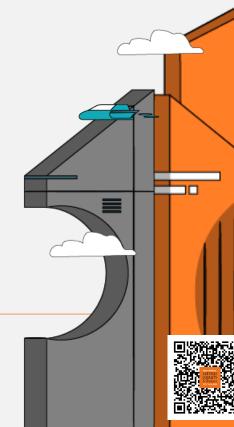


### What data quality problems do we encounter?

- By Nature of the Task
  - Afterthought of the Business
  - Data Quality Assurance could be Tedious







### Recap of data quality problems

#### By Nature of Data

- Imbalanced Data
- Wrong Labels
- Incomplete Data

#### **By Procedure**

- Latency
- Wrong Calculations
- Data Silos

#### By Nature of the Task

- Afterthought of the Business
- Data Quality Assurance could be Tedious

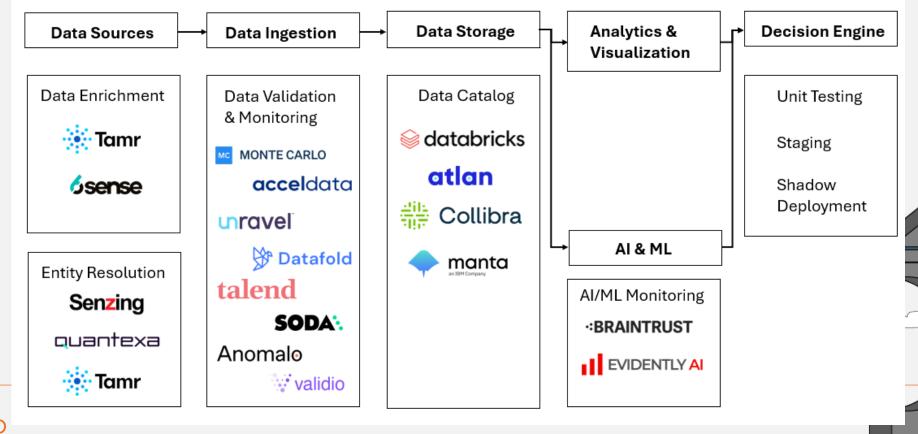






### **HOW** shall we solve the problems?

• Solutions / Tools



Yuqing Yao

• What is the problem?

customer	transaction_date	dispute_date	dispute_reason
John Smith	2024-01-01	2024-01-03	NULL

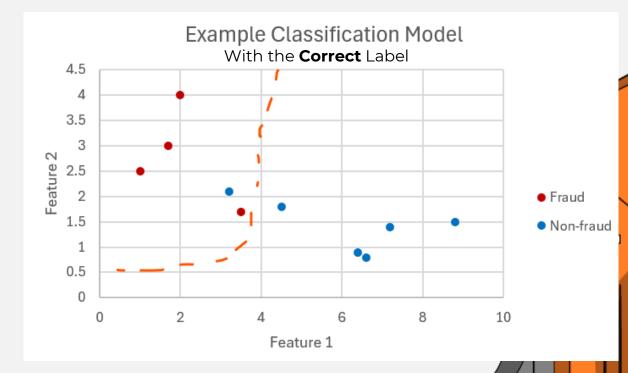




#### Yuqing Yao

- Why is it a problem
  - Misleads supervised learning or statistical analysis





#### Yuqing Yao



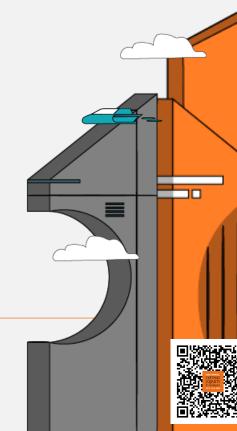
- What caused the problem?
  - No data validation during entry

Dispute Reason:

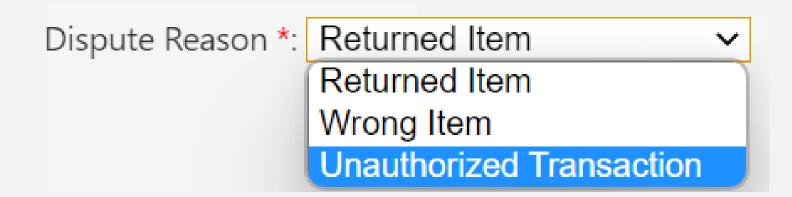






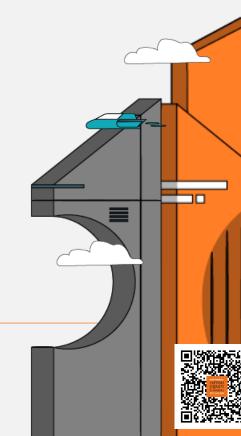


- What's the solution
  - Design the workflow to enforce data entry
  - Data validation during entry





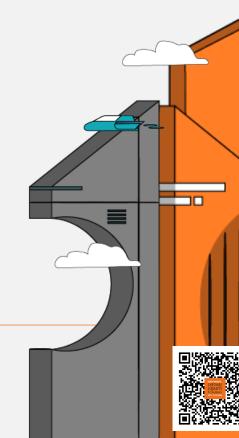
#### Yuqing Yao



- What's result
  - Higher rate of data being populated
  - More standardized values







- What is the problem?
  - Certain attribute suddenly has more occurrences of extreme values (e.g. 9999)

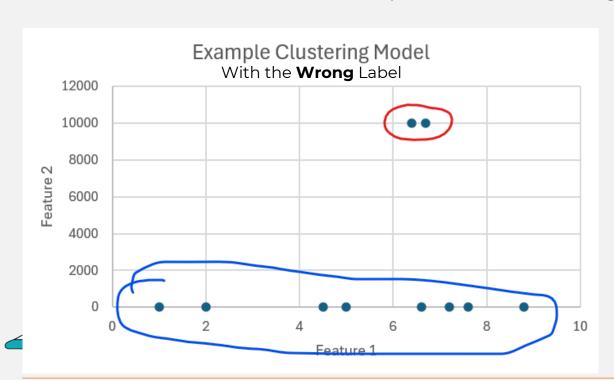
Feature 1	Feature 2	
2	8	
1	9	
5	2.3	
7.6	2.1	
6.7	9999	
4.5	1.8	
6.4	9999	
6.6	8.0	
7.2	1.4	
8.8	1.5	

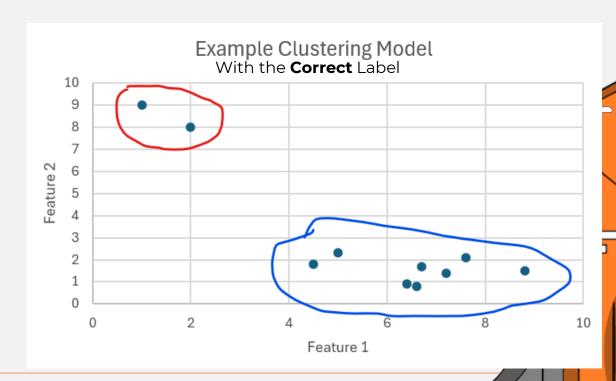




#### Yuqing Yao

- Why is it a problem
  - Misleads the unsupervised learning models





#### Yuqing Yao



- What caused the problem?
  - Vendor changed its missing value definition
  - Upstream team changed how they fill in missing values





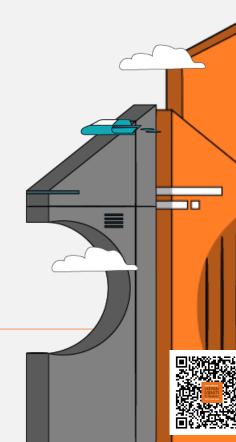
Yuqing Yao

- What's the solution
  - Set up monitoring based on business logic and corresponding alerts

Definition Name	Condition	Action
Column A	X% values over 2 SD (last 1 month)	Alert  Alert Warning Recover



Yuqing Yao



- What's result
  - Discover data errors before they contaminates the model training
  - Fix the problems in time





Yuqing Yao

- What is the problem?
  - Customer created different accounts using different set of contact info



Name John Smith

Email js95@gmail.com

Phone +14159999999

Address 123 ABC St, San Francisco, CA

Name John K.P. Smith

Email johnisawesome@gmail.com

Phone +14159999999

Address 123 ABC St, San Francisco, CA





- Why is it a problem
  - These records are physically the same person, but digitally treated as different people
    - Mistakenly upsample or downsample frauds
    - Miss out fraud rings





#### Yuqing Yao

- What caused the problem?
  - Data collected from different sources (Marketing / Log-in / vendors)
  - How records are created naturally cause this problem

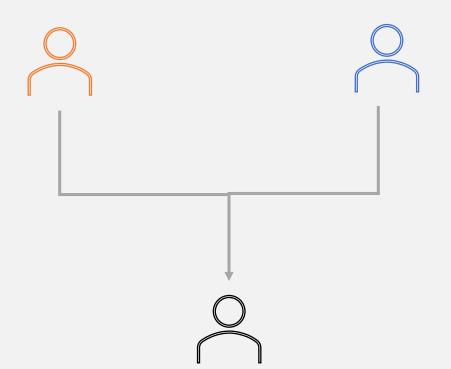




Yuqing Yao

- What's the solution
  - Entity resolution

John Smith
js95@gmail.com
+14159999999
123 ABC St, San Francisco, CA



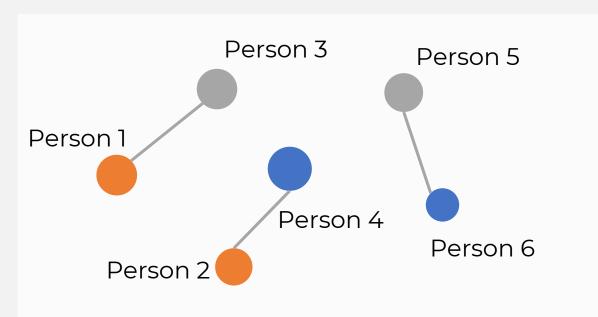
John K.P. Smith
johnisawesome@gmail.com
+14159999999

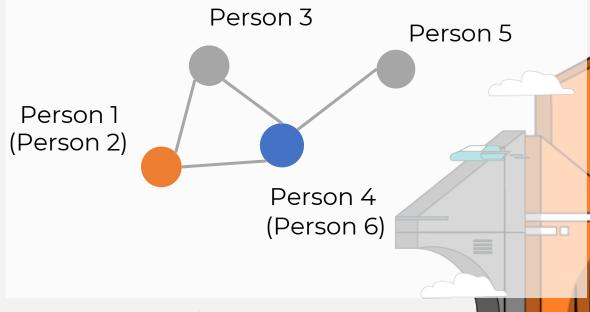
123 ABC St, San Francisco, CA



#### Yuqing Yao

- What's the result
  - Discover new fraud rings and unreported frauds







**Before** Entity Resolution

**After** Entity Resolution

#### Yuqing Yao



# Recap of examples

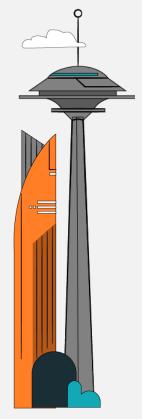
	Wrong labels	Abnormal values	Disconnected data
Impact	Misleads supervised learning or statistical analysis	Misleads the unsupervised learning	Mistakenly upsample or downsample frauds; Miss out fraud rings
Cause	No data validation during entry	Upstream processes changed filler values	Data collected from different sources
Solution	Data validation + Better UI	Set up monitoring based on business logic and corresponding alerts	Entity resolution
Result	More standardized values	Discover and fix data errors	Discover new fraud rings and unreported frauds





### **Conclusion and Future Improvements**

#### Conclusion

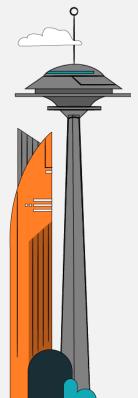


- Data quality is **important** in Credit Card Fraud Prevention because each piece of missing info contributes to solving the case
- Data issues can be caused by the nature of the data, procedures involved, and the nature of the task
- Solutions of the data issues lie in using good tools and controls during each step of conducting Credit Card Fraud Prevention, including data validation, entity resolution, data enrichment, data monitoring, and data catalogs



### **Conclusion and Future Improvements**

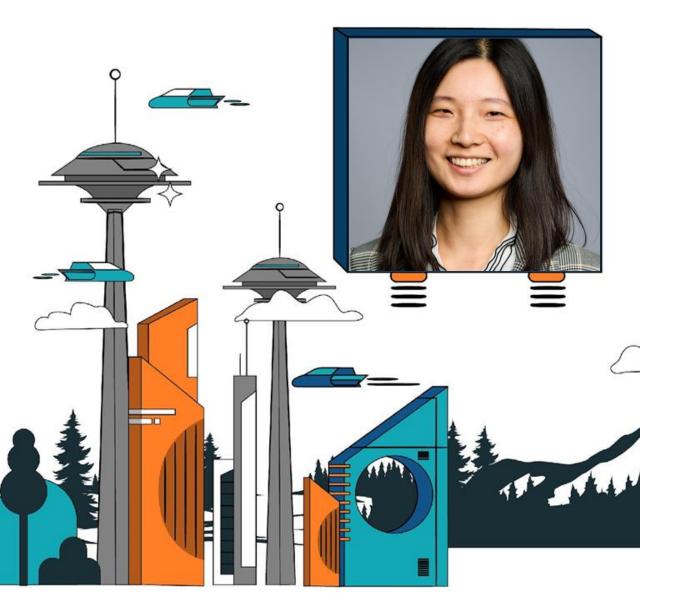
#### Future Improvements



- Use real-world case studies of FinTech companies to show the impact of applying the best practices of data quality assurance, and how that prevents losses
- Discuss how AI is used in Credit Card Fraud Prevention and how AI observability should be conducted in this field









THE FUTURE IS NOW PNSQC.ORG OCTOBER 14-16 2024