

# Anomaly detection

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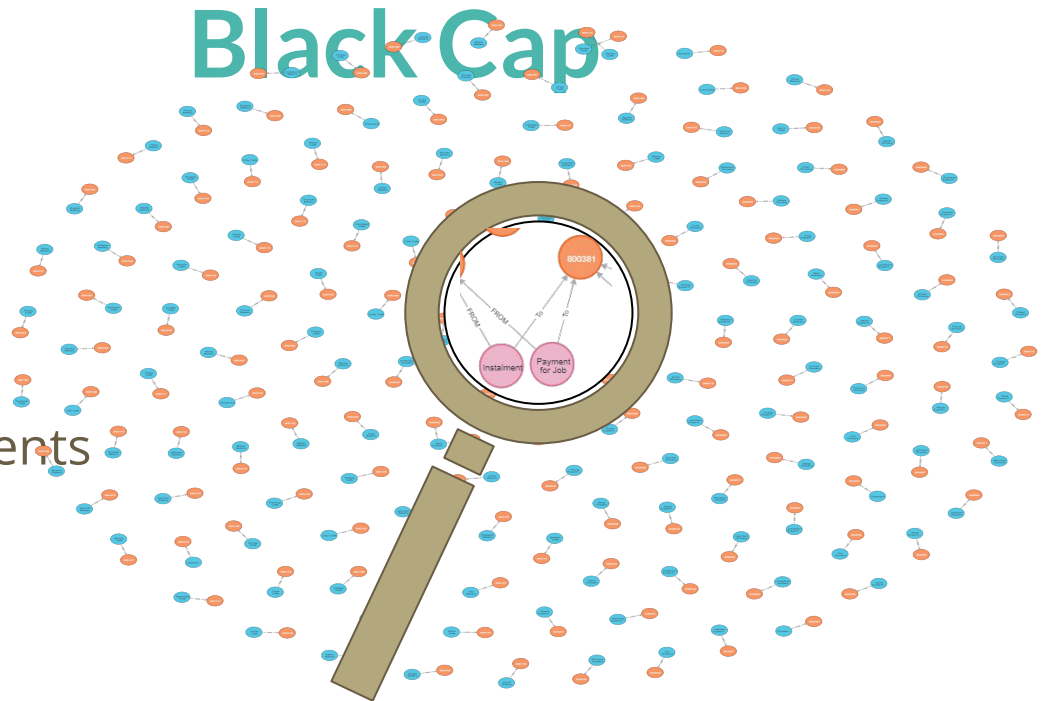
# Overview

# Anomaly detection

By team:

Black Cap

- Data & Problem
- Technologies Used
- Data Model
- Found Anomalies
- Insights
- Solutions
- Future Steps
- References & Acknowledgements



# Data & Problem

- Charlies Crash Repairs (CCR) business
  - Crash and mechanical repairs
  - Their expected turnover is **\$2.1M**, while they operate with a turnover of **\$14.3M**
- Where is the additional **\$12.2M** coming from?
- **Find anomalies in the transaction history and account information**



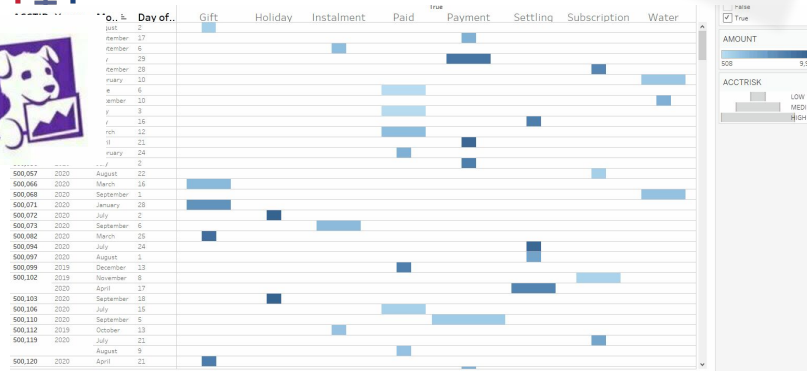
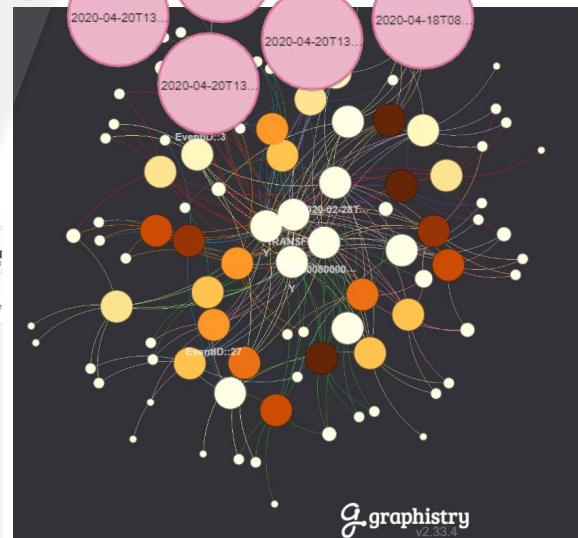
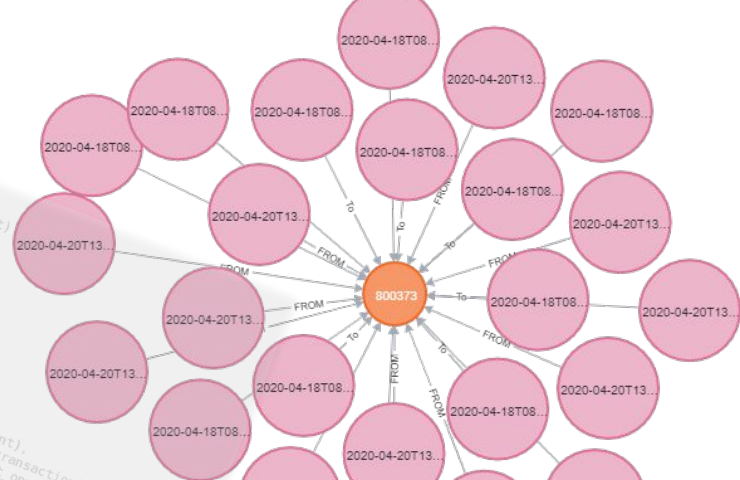
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# Tools Used

- Python
- NEO4J
- Graphistry
- Tableau
- DataDog

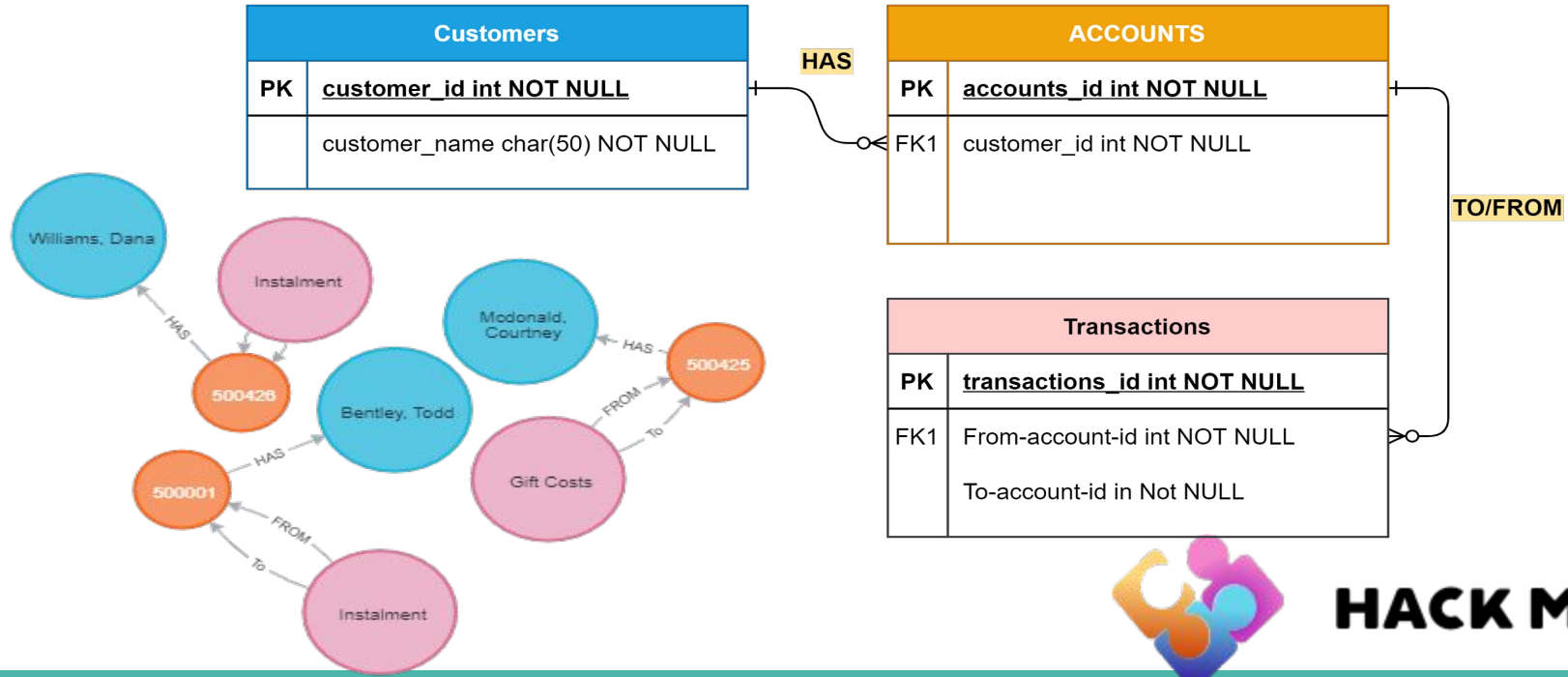


```
def read_in_account(account):  
    df_faccount = pd.read_csv(account)  
    return df_faccount  
  
def read_in_transactions(transaction):  
    df_ftxn = pd.read_csv(transaction)  
    return df_ftxn  
  
def merge_datasets():  
    df_merged = pd.merge(read_in_account(account),  
                          read_in_transactions(transaction),  
                          left_on='ACCTID', right_on='TOACCTID')  
    return df_merged  
  
def transaction_onto_self(df_transactions):  
    self_transactions = df_transactions[ df_transactions['FROMACCTID'] ==  
                                         df_transactions['TOACCTID'] ]
```



# DataModel

Field in	ACCTID	ACCTTYPE	ACCTNAM	FIRSTNAM	SURNAME	DOB	ACCTCRE	ACCTRISK	ADDRESS	ADDRESS	ADDRESS	CATEGORY
Data	TXN_ID	TXTYPE	AMOUNT	FROMACCT	TOACCTID	TXDA	REFERENC	ISFRAUD	ISFLAGGED			



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# Scenarios

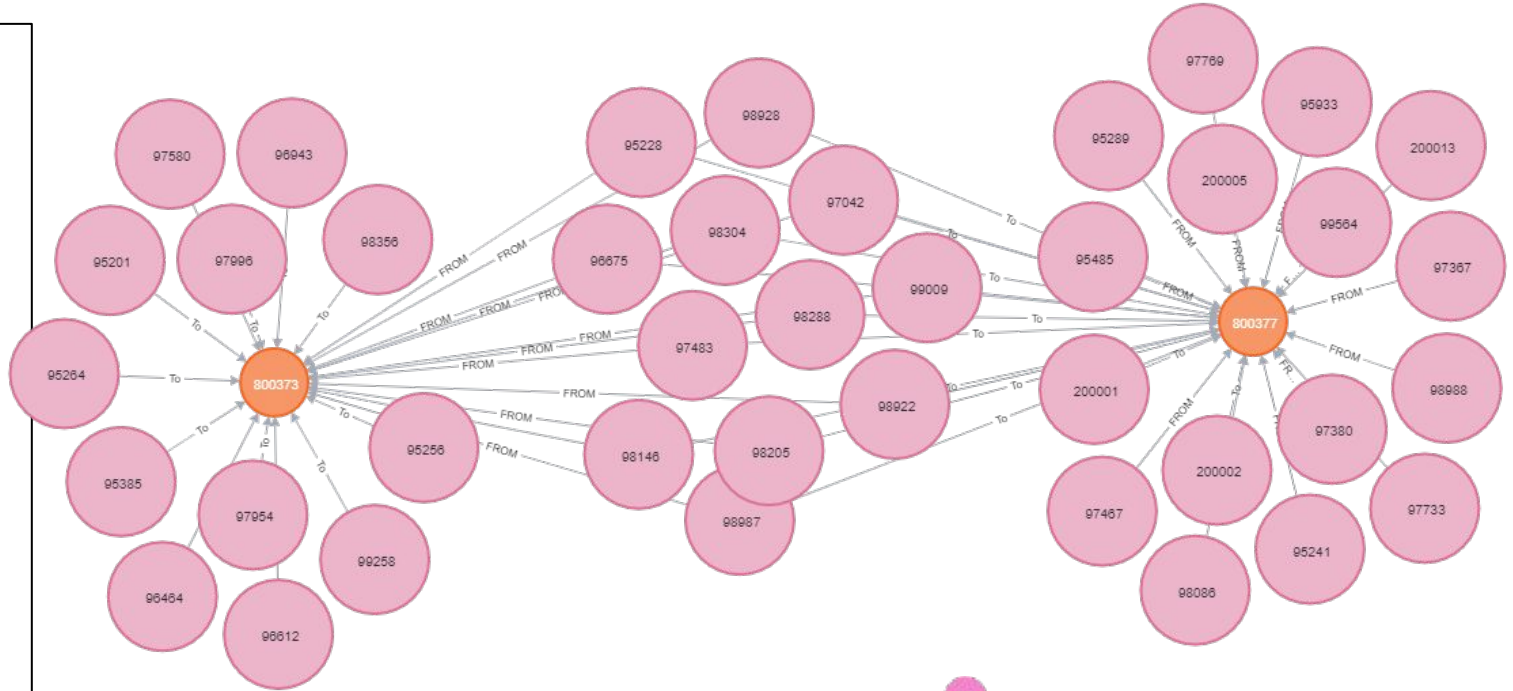
- Large transactions with > 90K
- Self-Self transactions
- Circular transactions
- Unusual amount spent on Meals
- Two accounts per person



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# Found Anomaly: Large transactions

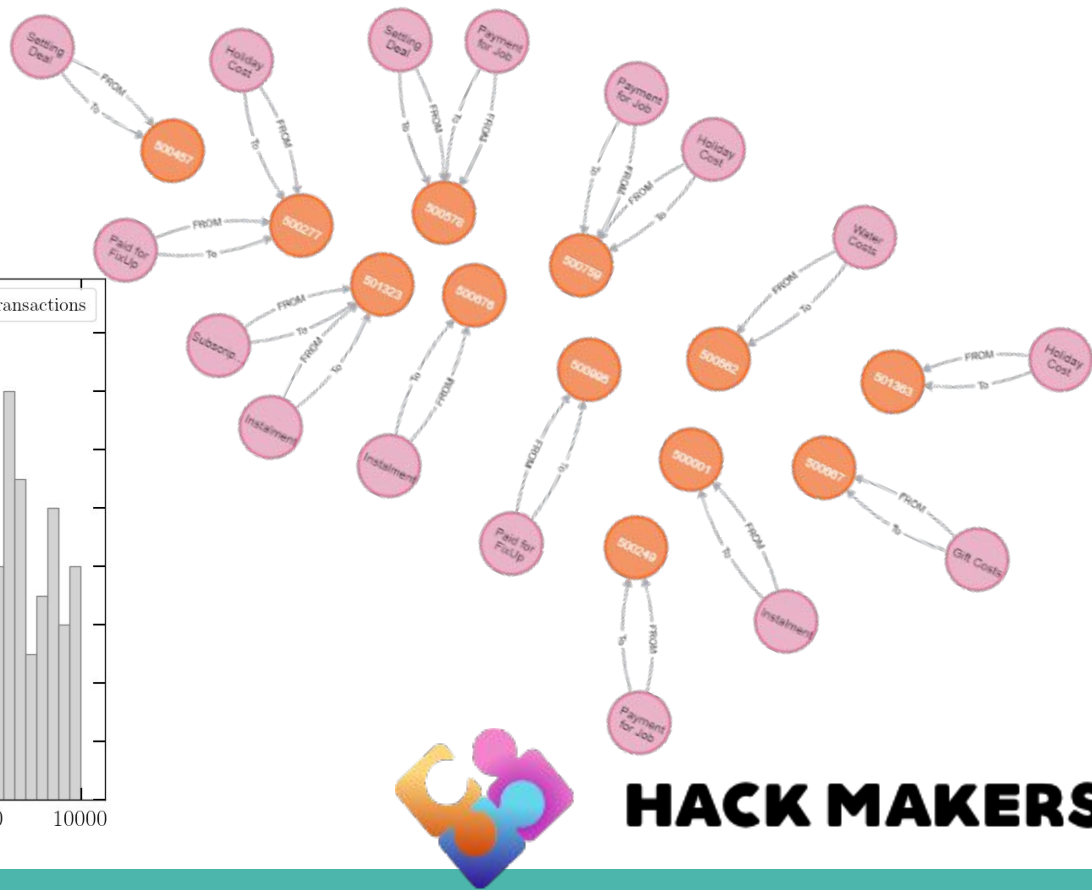
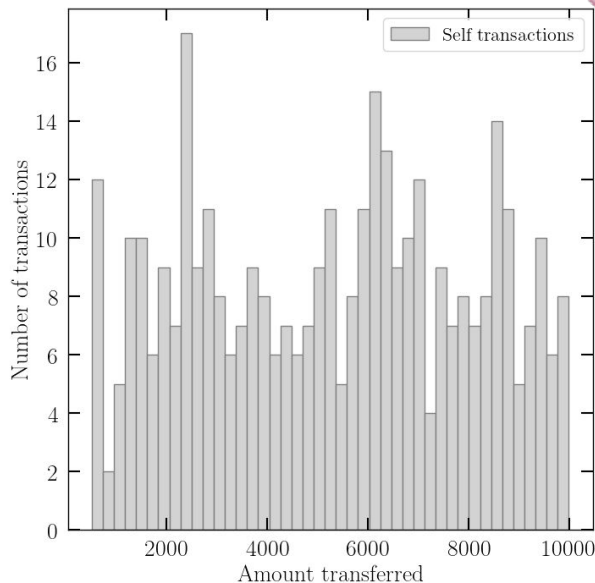
> 90K  
Transactions  
FROM/TO  
suspicious  
accounts  
with  
transfer  
amounts



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# Found Anomaly: Self-Self transactions

- With **maximum up to ~ \$10K**  
(Possibly to avoid Threshold  
transaction reports)

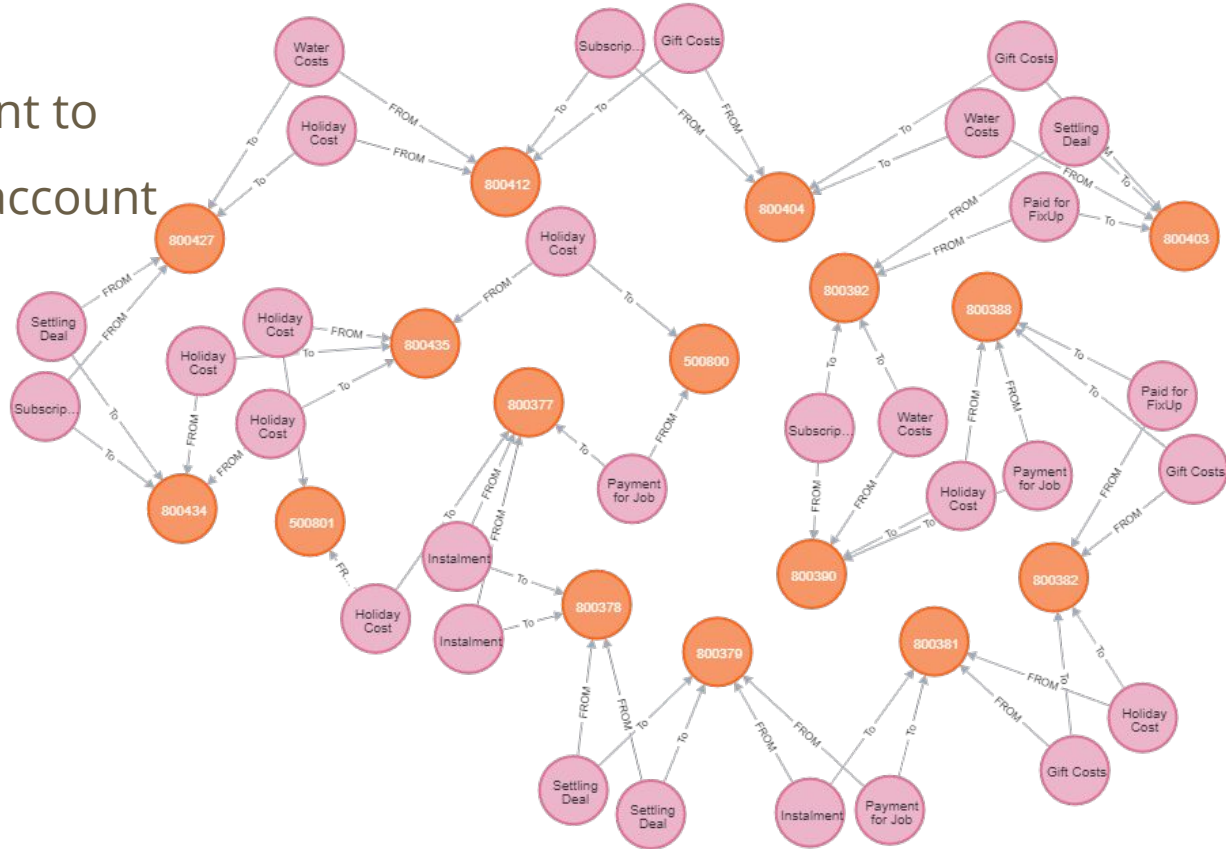


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# Found Anomaly: Circular transactions

- From account to account to account to account to account to account to account
- Within the same day!



# Solutions

- **We detect and interactively visualise Anomalies**
- Tools built:
  - **NEO4J & Graphistry** - To visualise Networks between accounts
  - **Tableau** - To interactively explore nuances of each account
  - **Coding scripts (Jupyter Notebook)** - To explore the datasets

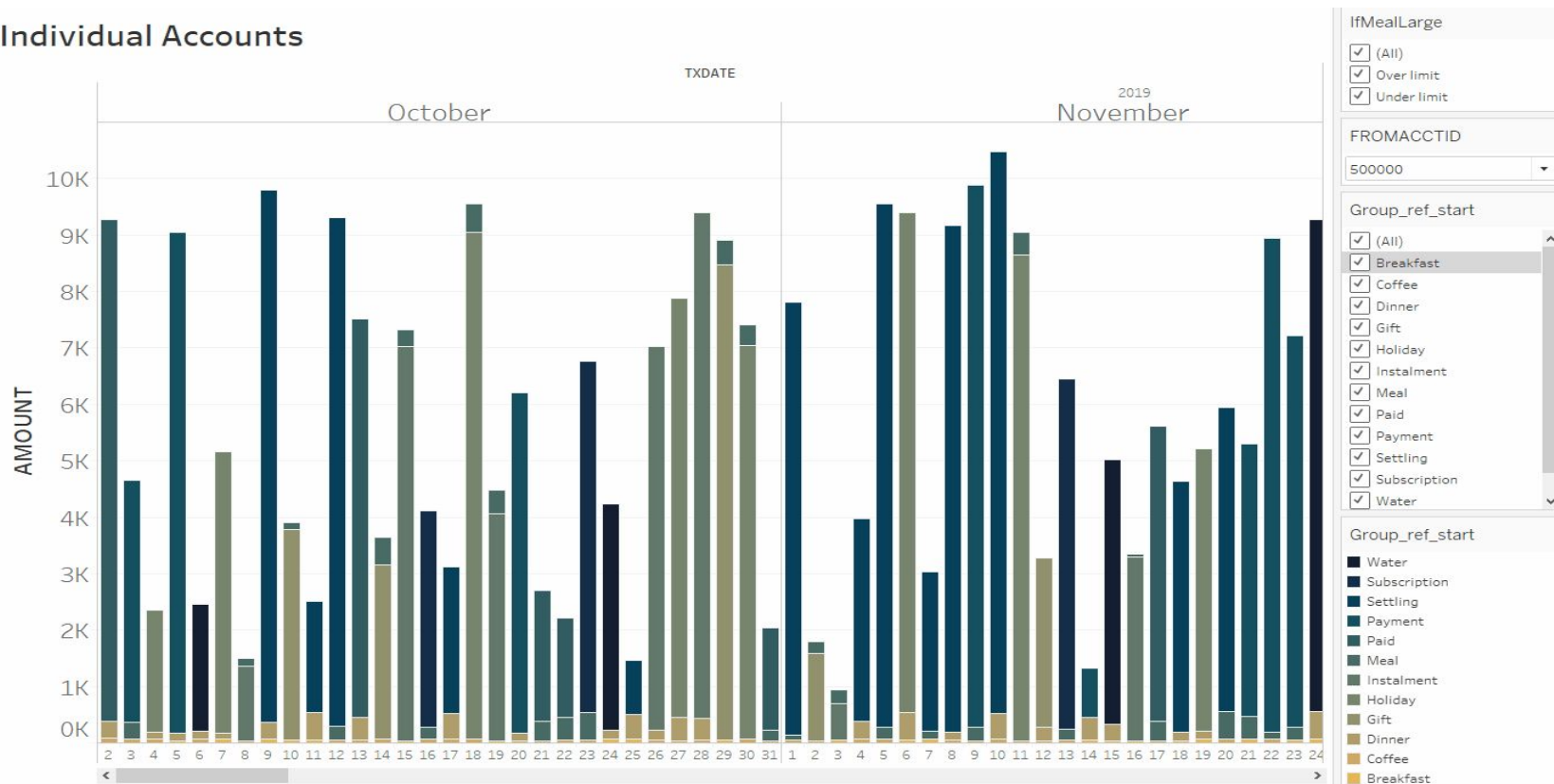
**Our developed tools can be used to inspect accounts**



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# Example: Our developed tools can be used to inspect accounts

## Individual Accounts



# Future Steps

- Create a web-app that will connect our
  - **Networks**
  - **Jupyter Notes**
  - **Tableau Dashboards**
- Function of the Web-app is to
  - **Automatically flag accounts**
  - **Provide Stakeholders a tool to inspect accounts timely & interactively**
- Using machine learning approach to detect anomalies
  - **This requires the appropriate labelled dataset**
  - **Dataset can be developed by crowdsourcing current anomalies**



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# References & Acknowledgements

- We used datasets that were provided by Organisers:
  - **faccount.txt** - Contains Account information  
(<https://objectstorage.ap-sydney-1.oraclecloud.com/n/sdc90vkxb5rj/b/data/o/anomaly-detection%2Ffaccount.txt>)
  - **ftxn2.txt** - Contains transaction information  
(<https://objectstorage.ap-sydney-1.oraclecloud.com/n/sdc90vkxb5rj/b/data/o/anomaly-detection%2Fftxn2.txt>)
- We are acknowledging support from the Mentors
  - Riyaz Ahamed
  - Jigna Thacker

Thank you #DigitalDefence Hack organisers!



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