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# ABC Supply Hackathon

An Investigation of Customer Behavior and Branch Performance

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# Questions

1. What are the predicted balances by Agings buckets over the next year for customers? How does this relate to the customer's sales (purchases)?
2. **How can we characterize different kinds of branches in terms of how they use credit? Is there a relationship between branch performance (sales) and Credit?**
3. **How can we characterize different kinds of customers or accounts in terms of how they use credit? Are there customers who have never not paid on time?**
4. Is the current risk score a valid signal for Agings movements?
5. How might we prioritize collections activities?
6. How do credit lines relate to customer spending? If you were going to set a credit limit for an account, what might you consider?
7. **Which customers are most likely to go into bad debt?**

# Background

- Large and diverse customer base
  - [REDACTED] customers under [REDACTED] accounts
  - Over [REDACTED] branches nationwide
- Nature of business necessitates credit
- Huge cash flows = high risk

# Goals

- Understand state of branches
- Identify our most trustworthy customers
- Characterize those customers who default often
- Develop framework for presenting information to business

# Questions - Customers

- How many accounts?
- How long a customer?
- How often do payments go overdue?
- How often are payments delinquent?
- % of Credit Line used?
- Avg ang weighted risk scores?

# Methodology

- Analysis at Customer, Account, and Customer-x-Account level
- Various Clustering/Profiling Methods
  - K-Means
  - GMM
  - DBScan
  - Archetype Analysis
- Random forest to predict likelihood of delinquent payments

# Goals / Methods - Branches

- Revenue vs Credit
  - Affected by Size
  - By Risk-profile of the branch owner
- 
- Missed opportunities?

# Branch Clusters

Dimensions (Averaged across snapshots):

Credit\_Per\_Account, %\_Accounts\_NotBadAccure, RISK\_SCORE, CREDIT\_ACCOUNT\_AGE

0: the Mainstream (191)

1: the Willing (78)

2: the Cautious (97)

3: the Shark (50)

4: the Risky (54)

	perAcctCredit	pct_notBA	RISK_SCORE	CREDIT_ACCOUNT_AGE
clusters				
0				
1				
2				
3				
4				



# Branch Clusters Details

0: the Mainstream (191)

1: the Willing (78)

2: the Cautious (97)

3: the Shark (50)

4: the Risky (54)

	CREDIT_LINE	DelinqToCredit	TOTAL_REVENUE
clusters			
0			
1			
2			
3			
4			

# Customer Clusters

	N_ACCOUNTS	AGE	ROWS	N_OVERDUE	N_DELINQUENT	P_CREDIT	RISK_MEAN	WEIGHTED_RISK_MEAN
Label								
1								

# Trustworthy Customers


N_ACCOUNTS	AGE	ROWS	N_OVERDUE	N_DELINQUENT	P_CREDIT	RISK_MEAN	WEIGHTED_RISK_MEAN
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**Label**

# Next Steps

- Investigate our least trustworthy customers
- Continue temporal analysis
  - DEDICOM model will allow this
- Build frontend for insight consumption

# Sneak Peek at What's in Store!



Branch  - Class

Branch Name 

Search...

Q

Number of Customers

Number of Accounts

Accounts with Overdue Cal

Avg Age

Avg Accounts

Avg RiskScore

Weighted Risk

% Credit Used

Num Overdue

Num Delinquent

Segment 1

[Explore Further](#)

Segment 2

[Explore Further](#)

Segment 3

[Explore Further](#)

Segment 4

[Explore Further](#)