



# Home Credit Data Set

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# Predicting Loan Default Risk



## The Company

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Home Credit:  
lender who  
provides loans to  
populations  
unable to use  
traditional credit  
services.

## The Problem

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how to lower loan  
default risk by  
identifying  
patterns from  
within historical  
data.

## The Method

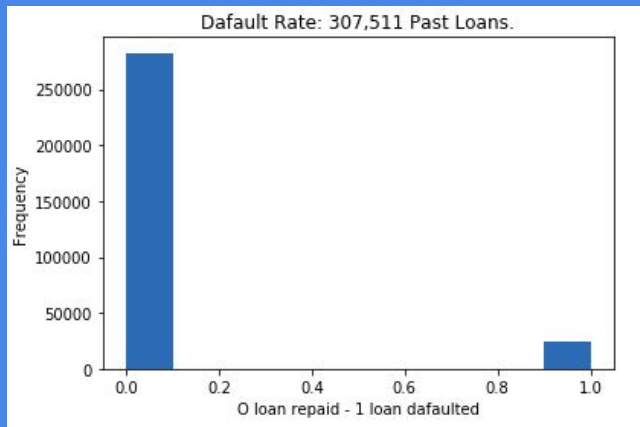
-

a combination of  
Exploratory Data  
Analysis, Principal  
Component  
Analysis, and  
various Machine  
learning  
Algorithms to  
identify the most  
predictive data  
points.





The Data  
307,511 loans  
122 features -  
including age,  
employment, past  
credit, region...



Default Rate

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8.8%

## The Potential

Even a small decrease in the default rate would translate to a significant reduction in losses and an increase in capital available for further loans.



# TOP THREE DATA FEATURES

## Outside Credit Agency Scores

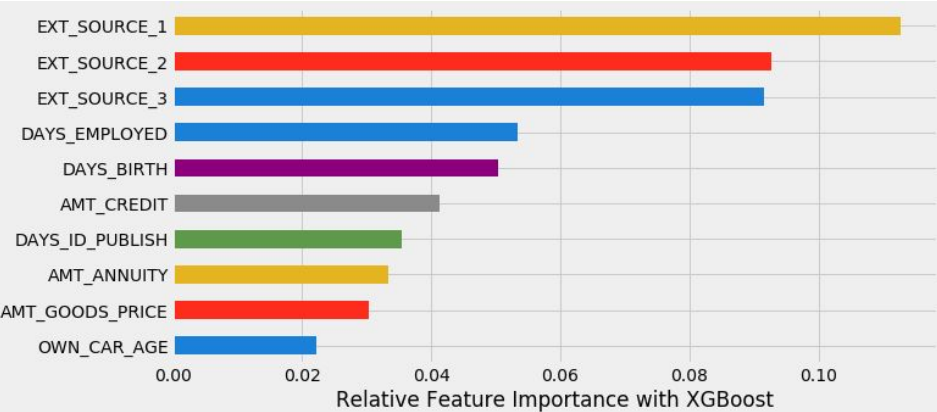
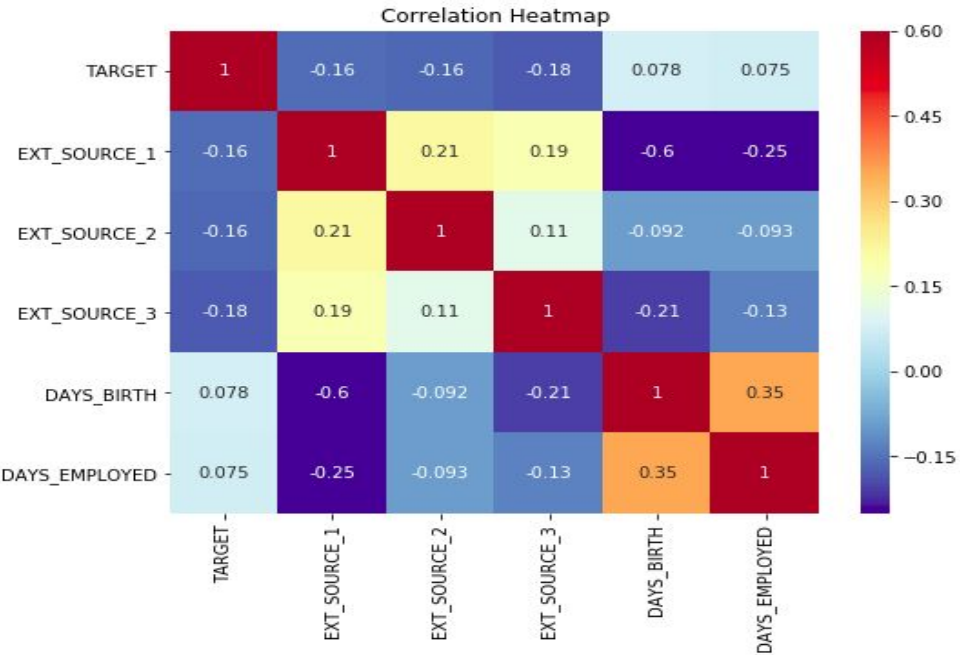
(EXT\_SOURCE\_[1-3])

## Length of Employment

(DAYS\_EMPLOYED)

## Age of Borrower

(DAYS\_BIRTH)



Feature Selection via XGboost

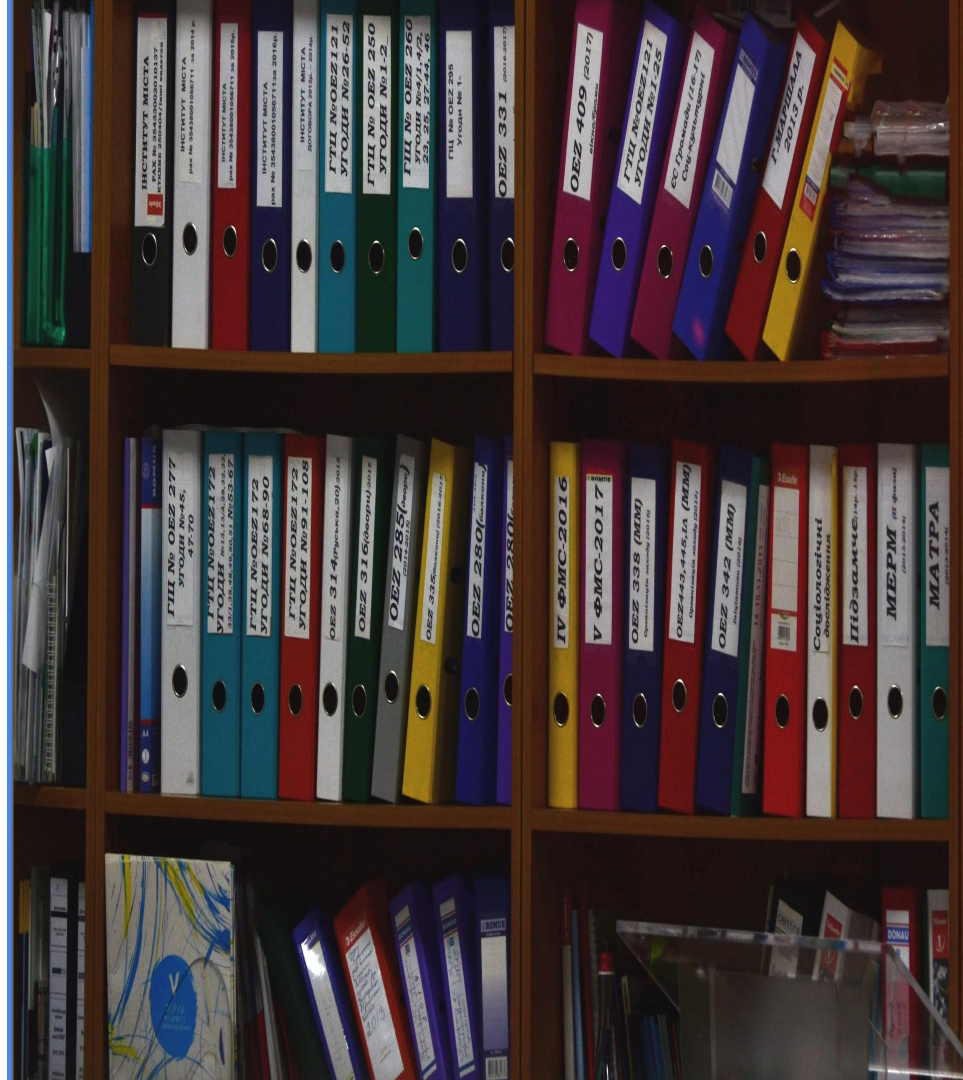
Training Accuracy: 92.01%

Testing Accuracy: 91.99%

# Recommendation - 1

## Trust and Emulate the Credit Agencies

The top three indicators of repayment correlate to credit scores assigned by outside agencies.





## Recommendation - 2 Investigate Employment Anomaly

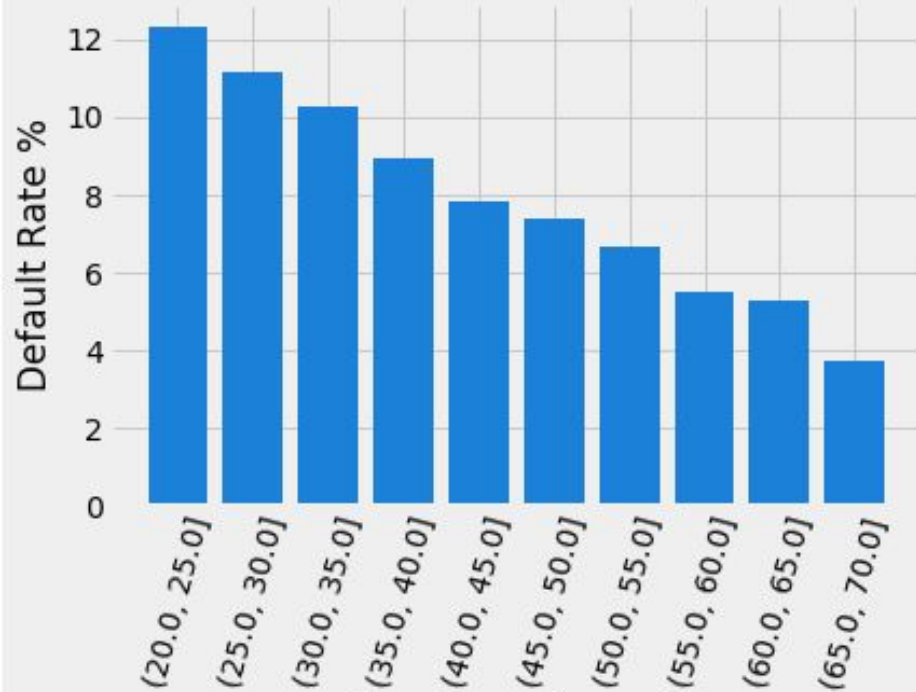
Anomaly: Max Days Employed 365243

Anomaly Default Rate: 5.4%

Potential Gain  
3.26%

Non-Anomaly Default Rate: 8.66%

Default Rate by Age



## Recommendation - 3

Aggressive debt counseling for younger segments .



# Future Work



- ⚙️ fully Explore the Employment Anomaly for Insights
- ⚙️ engineer Domain Specific Data Features
- ⚙️ research the Best Practices credit agencies are currently employing and apply them to the data set





THANK YOU

Full credit and appreciation to Will Koehrsen for his Kaggle guides to this data set. All hail Will!