Credit Analysis

Submitted By

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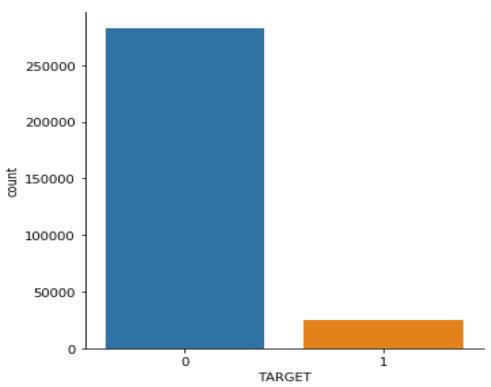
Objective

Analysis of the loan attributes which provides the strong indication to default

Date Cleaning

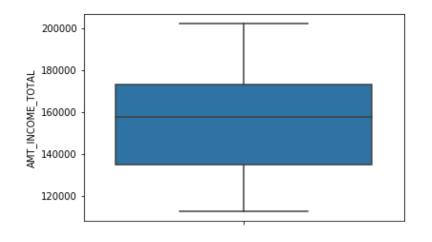
- Remove row/column with all NA
- Removing the columns with lot of Missing values
- Replace XNA/XAP with NA
- Binning of continuous variable
- Derive Columns

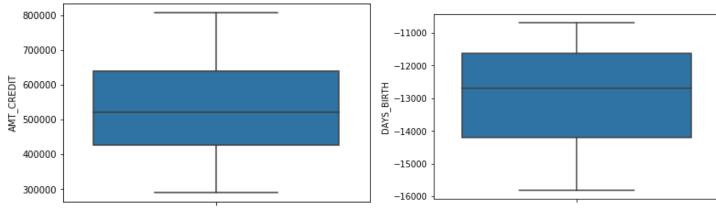
Data Imbalance



• As the data is imbalanced, using the percentage to the target, 8 % before removal, 13% after removal

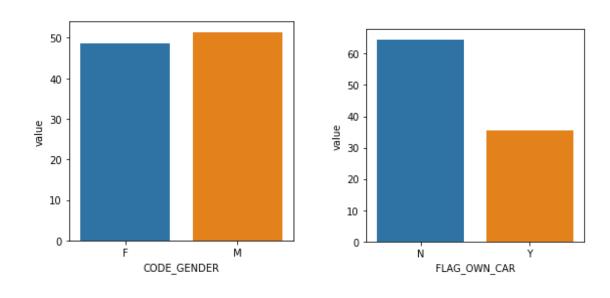
Numeric Attribute Analysis



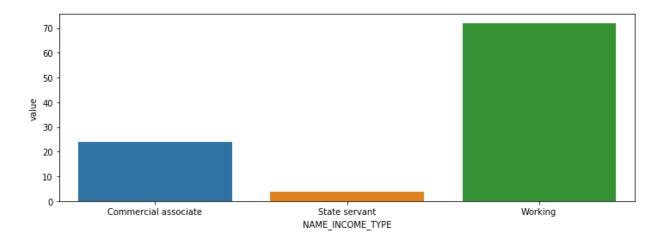


- Outlier are removed by taking the values between 0.25 and 0.75 quantile
- Outliers are removed for AMT_INCOME_TOTAL, AMT_CREDIT, AMT_GOODS_PRICE, DAYS_EMPLOYED, DAYS_REGISTRATION and DAYS_BIRTH

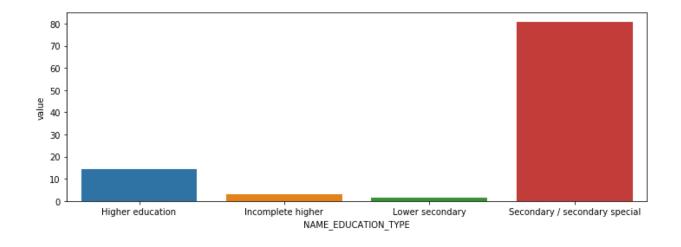
Categorical Attribute Analysis

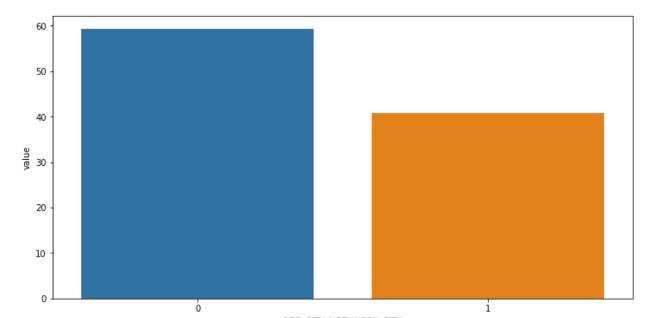


- Males are defaulting more than females
- Clients who don't have cars are more defaulting
- Working clients are defaulting more



Categorical Attribute Analysis

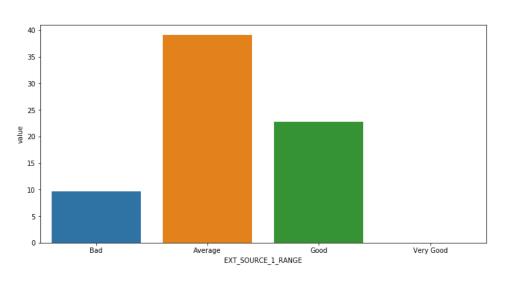


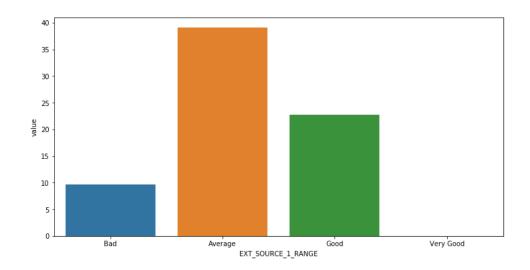


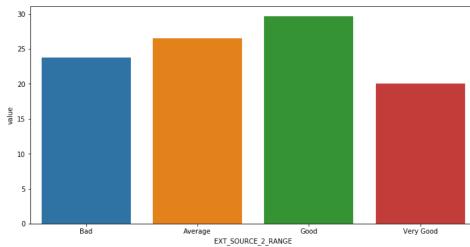
 Clients with Secondary education are defaulting more

 if clients permanent address does not match contact address defaults more

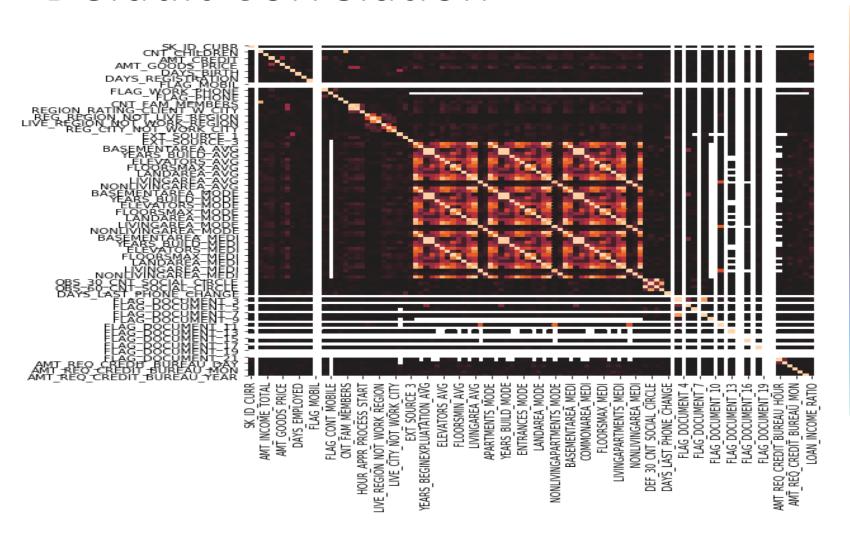
External Source Analysis after binning







Default Correlation



- 0.8

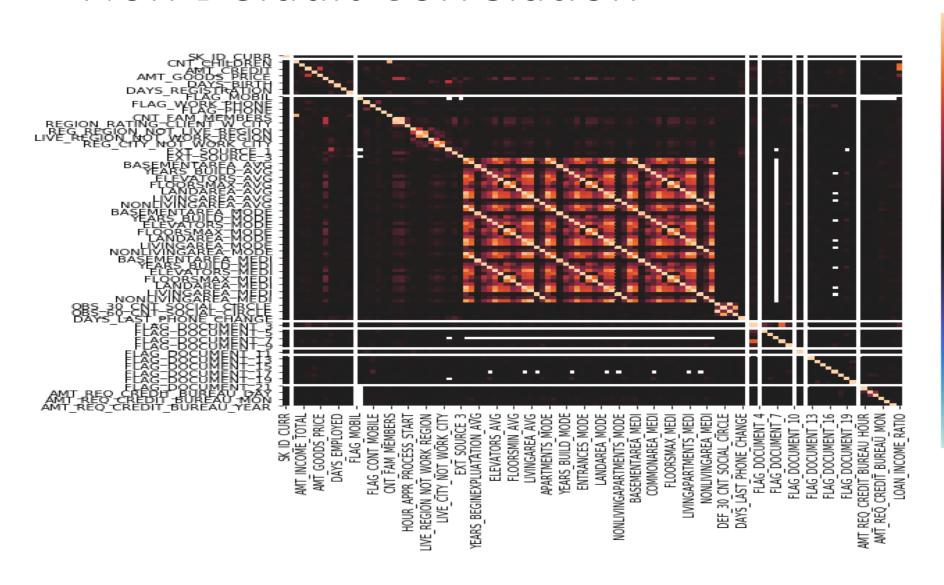
- 0.4

- 0.0

- -0.4

- -0.8

Non Default Correlation



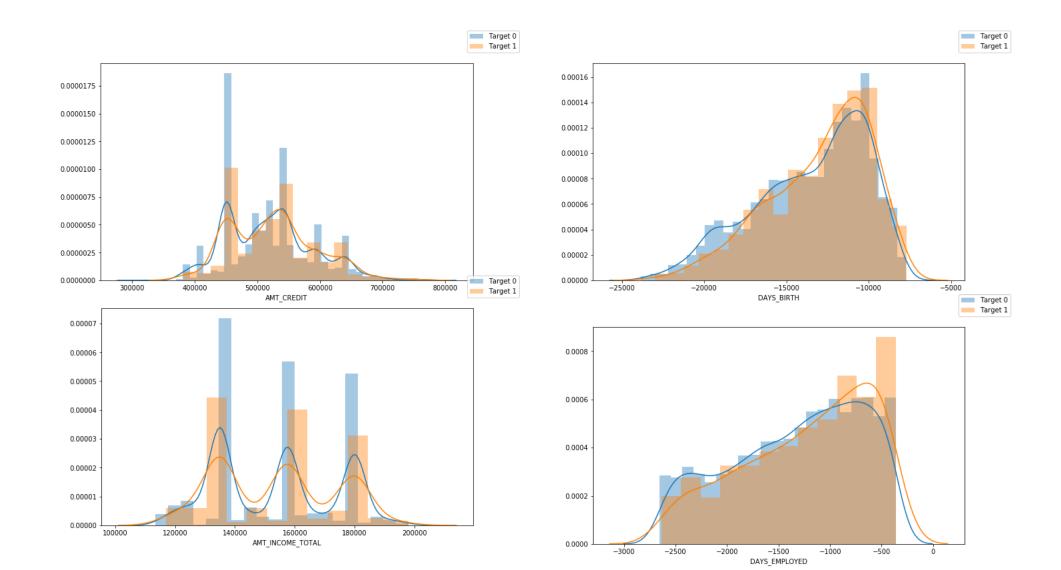
- 0.8

- 0.4

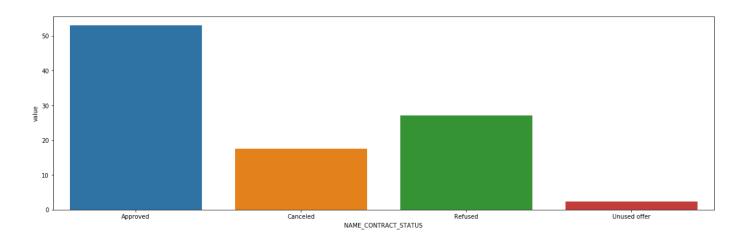
-0.4

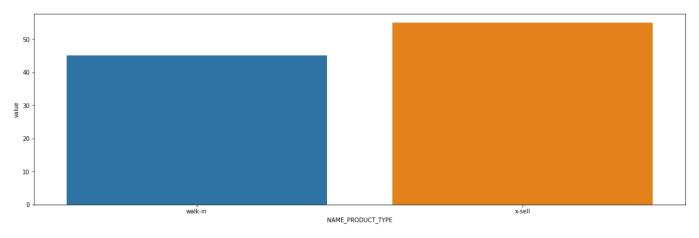
-0.8

Numeric Attribute Distribution



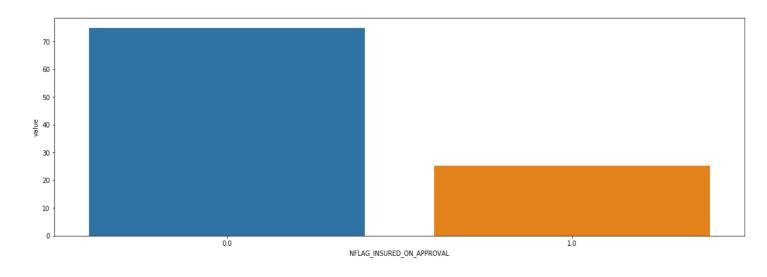
Category Analysis





- Walk in clients are more likely to default
- Previous Refused clients are likely to default

Insured or Not



 If client not requested insurance, will default

Summary

- Working Clients with Secondary Education, whose previous application is rejected is more likely to default
- Client who didn't request insurance, and walked in clients are more like to default