Applied Machine Learning Phase 1

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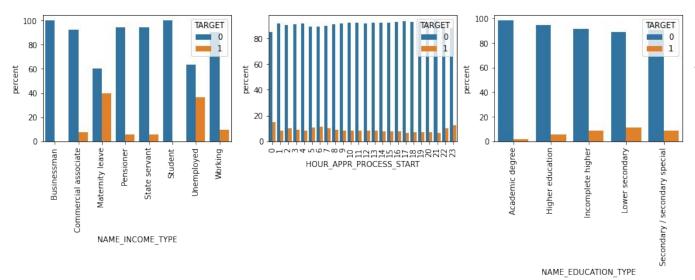






EDA + baselines

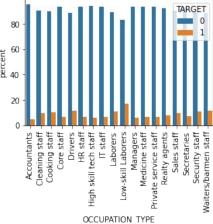
- Found interesting application and secondary features
- 0.74 ROC score using Logistic Regression



Most Positive Correlations:	
FLAG_EMP_PHONE	0.045982
REG_CITY_NOT_WORK_CITY	0.050994
DAYS_ID_PUBLISH	0.051457
DAYS_LAST_PHONE_CHANGE	0.055218
REGION_RATING_CLIENT	0.058899
REGION_RATING_CLIENT_W_CITY	0.060893
(DAYS_CREDIT, min)	0.075248
DAYS_BIRTH	0.078239
(AMT_BALANCE, mean)	0.087177
TARGET	1.000000

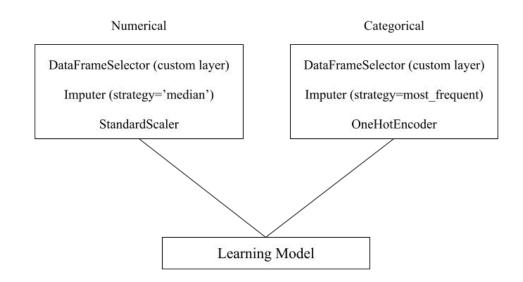
Name: TARGET, dtype: float64

Most Negative Correlations:	
EXT_SOURCE_3	-0.178919
EXT_SOURCE_2	-0.160472
EXT_SOURCE_1	-0.155317
(AMT_CREDIT_LIMIT_ACTUAL, count)	-0.060481
DAYS_EMPLOYED	-0.044932
FLOORSMAX_AVG	-0.044003
FLOORSMAX_MEDI	-0.043768
FLOORSMAX_MODE	-0.043226
AMT_GOODS_PRICE	-0.039645
REGION_POPULATION_RELATIVE	-0.037227
Name: TARGET, dtype: float64	



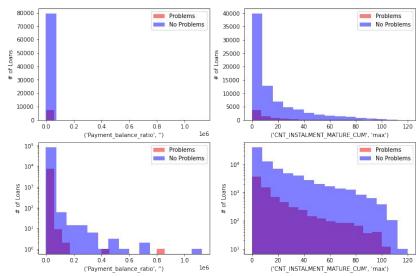
Pipeline

Our Pipeline follows this model



Feature Engineering

- Experimented aggregating and creating new features
 - Looking at aggregations differently helped a lot
 - New features performed terribly
- Moving forward, need to go deeper into our datasets
- Want to add a lot more data



Summary

- Completed baseline models
- Did basic feature engineering
- Want to add more features to model
- Have to figure out the deeper connections between data sources
- Google Colab is finicky with too much data