Comparative Model Analysis

Jiacheng Yao

Process



1. Exploratory Data Analysis

Exploratory Data Analysis

Application

Dimension: 276686, 122

One row = One loan application

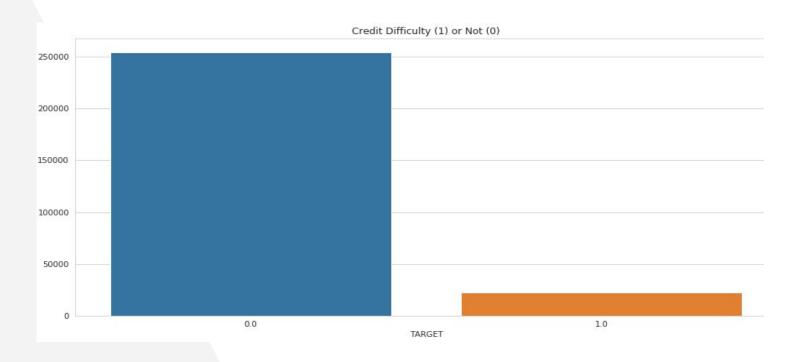
Bureau

Dimension: 1716428, 17

One row = One previous

Loan

Exploratory Data Analysis



2. Preprocess

Preprocess - Merge Bureau Data

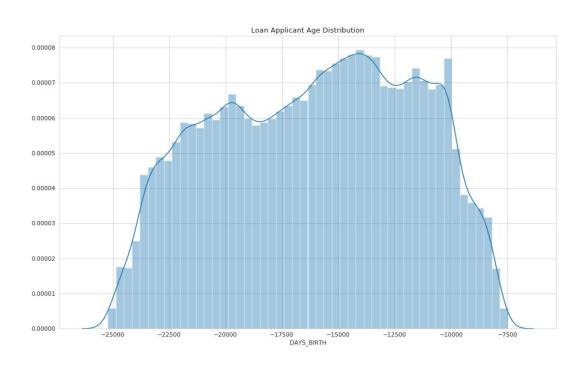
Categorical

Aggregate to Most Common per SK_ID_CURR

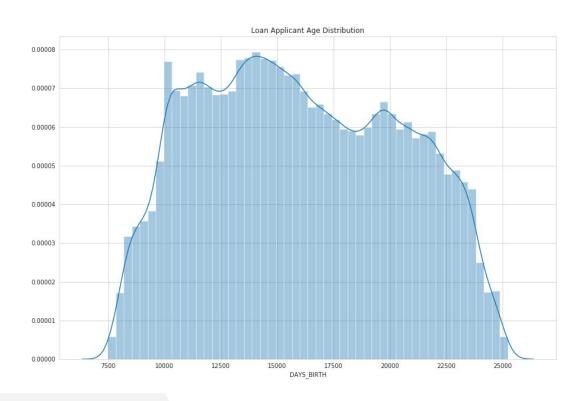
Numeric

Aggregate to Median per SK_ID_CURR

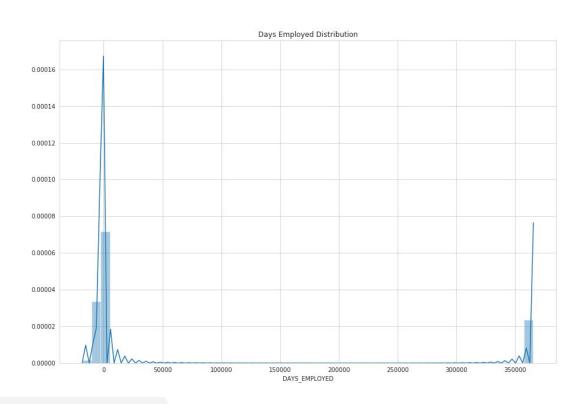
Preprocess - Loan Applicant Age



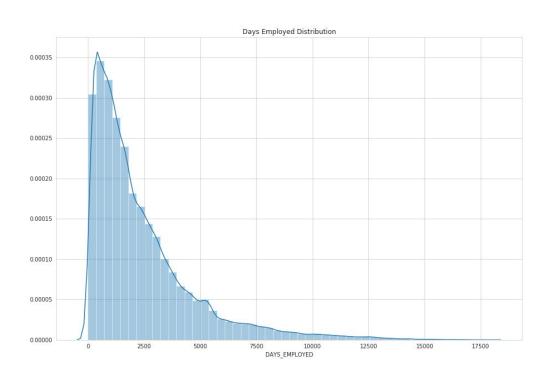
Preprocess - Loan Applicant Age (clean)



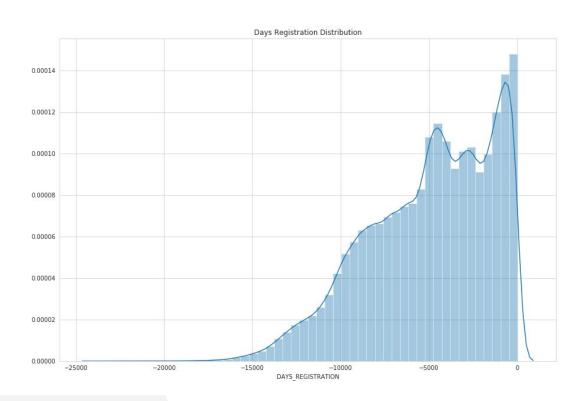
Preprocess - Days Employed



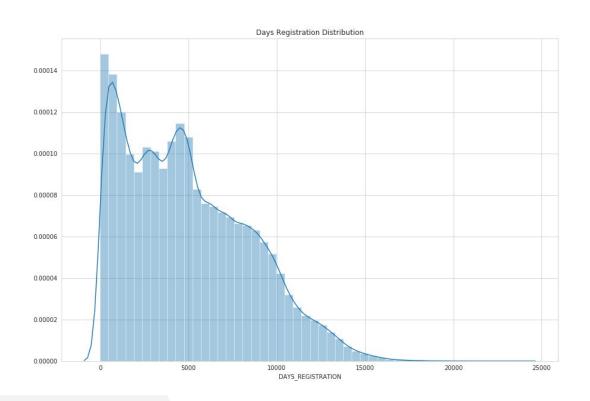
Preprocess - Days Employed (clean)



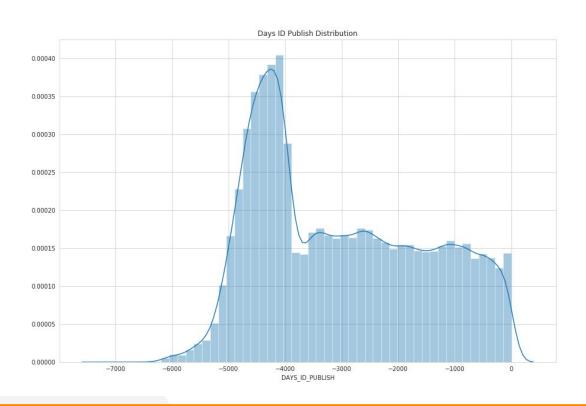
Preprocess - Days Registration



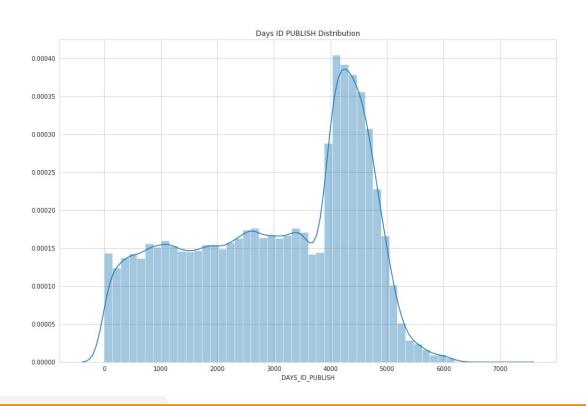
Preprocess - Days Registration (clean)



Preprocess - Days ID Public



Preprocess - Days ID Public (clean)



Preprocess - Last Steps

Categorical

- 1. Null -> "Missing"
- 2. One Hot Encoding

Numeric

- 1. Drop features with high correlation
- 2. Impute with MICE
- 3. Standardize

3. Model Comparison

Model Comparison

	Fast and Frugal	Random Forest	XgBoost
BACC (10Fold AVG)	0.6805	0.6808	0.8129
Time	12'20"	14'26"	71'18"

4. Discussion

Discussion

Simplicity and Interpretability

FFT: more time efficient, interpretable, human readable and usable, potentially more robust against change of circumstances (regime changes).

Others: more complicated, difficult to interpret, prone to overfit, less robust, more time-consuming

When to use FFT

- 1. If used by human without aid,
- 2. If reducing time and cost is crucial,
- If understanding and transparency is important,
- If performance similar to more complicated models,
- 5. If robustness is priority.

Discussion

Further Steps:

- Closer inspection of individual features,
- Integration of alternative data, e.g. credit card data, loan installment payment data, etc,
- 3. Acquire cost matrix for each prediction scenario (false negative, true negative, false positive, true positive) from experts and optimize models on it to tackle imbalancedness of the dataset.
- 4. etc.

Behavioral Interventions with MI

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Approaches - Unsupervised

Unsupervised

- Treat as clustering problem, group debtors into clusters and adopt different interaction strategies for different clusters.
- To find the optimal strategy for each individual cluster, conduct A/B test on several strategies and choose the one with the best KPI (debt repayment percentage, click through rate, customer retention rate, etc.)

Approaches - Supervised

Classification

- Treat as classification problem, engineer a target signal to indicate if debtors will continue to repay, or will have difficulty with payments.
- For debtors with and without difficulty paying back debts, adopt different interaction strategies for the two classes.
- 3. To find the optimal strategy for each individual class, conduct A/B test on several strategies and choose the one with the best KPI (debt repayment percentage, click through rate, customer retention rate, etc.)

Regression

- Treat as regression problem, engineer a target signal to forecast future lifetime value(LV) of debts.
- Divide debtors into different tranches based on future LV, adopt different interaction strategies for different tranches.
- To find the optimal strategy for each individual tranch, conduct A/B test on several strategies and choose the one with the best KPI (debt repayment percentage, click through rate, customer retention rate, etc.)

Recommendations

- Monitor the KPIs vigilantly with incoming data to validate the performance of the models and do optimization based on results.
- If possible, conduct surveys to better gauge satisfaction levels of different groups of debtors and how to improve interaction methods.
- Integrate findings from experts in the field of behavioral science, especially in the use case of debt collection.
- 4. etc.

THANKS!

Any questions?

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Appendix - Feature Importance

