CS598 - Project 3

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Computer System

Hardware

- Dell Precision Tower 5810
- CPU: Intel Xeon E5-1607 @ 3.10GHz
- Memory: 32GB
- GPU: Nvidia GeForce GTX 1080 (2 cards)

Software

- OS: Windows 10 Professional 64bit
- R: 3.5.1
- R Packages:
 - catboost 0.11.1
 - xgboost_0.71.2
 - randomForest 4.6-14
 - glmnet 2.0-16
 - kernlab 0.9-26

Preprocessing and Feature Engineering

Several approaches are taken to pre-process the data.

- Response label: merge Charged Off with Default and convert label values to 0 or 1.
 - Build new predictors to help training/prediction:
 - earliest_cr_line_mon: derived from earliest_cr_line that indicates how many months has elapsed till 2019-1-1 when the borrower's earliest reported credit line was opened.
 - fico_score: consolidate fico_range_high and fico_range_low using formula: (fico_range_high + fico_range_low) / 2.
 - Level grouping:
 - zip_code: it has more than 900 levels, I group these values to 10 new levels to reduce memory usage and improve performance.
 - $\bullet\,$ Remove predictors: remove less useful and redundant predictors.
 - emp_title (too many levels), title (redundant with purpose), grade (redundant with sub_grade)
 ,earliest_cr_line, fico_range_high, fico_range_low.

Models

For testing purpose, I build 7 models,

- Dumb model: just predict 0.2 for every sample.
- Logistic Regression
- Boosting (XGBoost, CatBoost)
- RandomForest
- Lasso

$\bullet \;\; liner \; SVM$

Suprisingly, Dumb model can achieve 0.504 logloss score. kernlab ksvm() fails to build the model (hang forever). lasso and random forest don't give me significant improvement than logistic regression and they take much longer time to build. Thus, I will pick dumb, Logistic Regression and Boosting as my final models.

Note: My testing shows CatBoost performs at least 10x faster than XGBoost (even without GPU). In case catboost library is not installed, xgboost will be used.

Evaluation

I tested all 3 test datasets against these models with the parameters,

- Dumb: None
- Logistic Regression: liner with all predictors.
- Boosting: learning_rate = 0.09, iterations = 1200.

The LogLoss scores are,

	Dumb	LogisticRegression	Boosting
Test1	0.5038353	0.4539925	0.4445856
Test2	0.5038353	0.4548663	0.4462146
Test3	0.5038353	0.4539786	0.4448837
Average	0.5038353	0.4542791	0.4452280

Computation time: 1244 seconds