CAPSTONE 2: LENDING CLUB LOAN DEFAULT PREDICTIONS

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LENDING CLUB'S BUSINESS PROBLEM

5% of borrowers defaulted on loans between 2007 and 2017 equating to

\$8,695,172,153 in lost principal, interest, and fees

WHO CARES ABOUT DEFAULTS?

















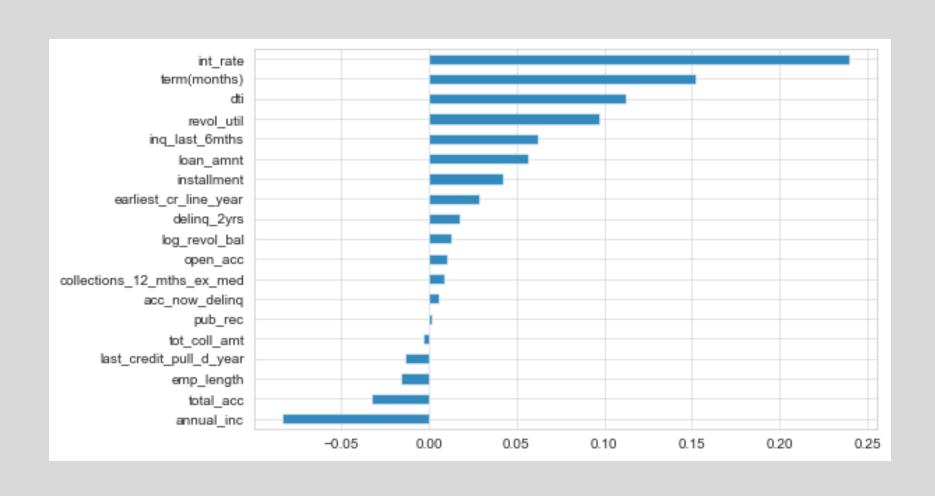
Lenders

- Banks
- Credit Unions

Investors



WHAT FEATURES INFLUENCE DEFAULTS?



BEST MODEL PARAMETERS

RandomForestClassifier(

```
bootstrap = True,

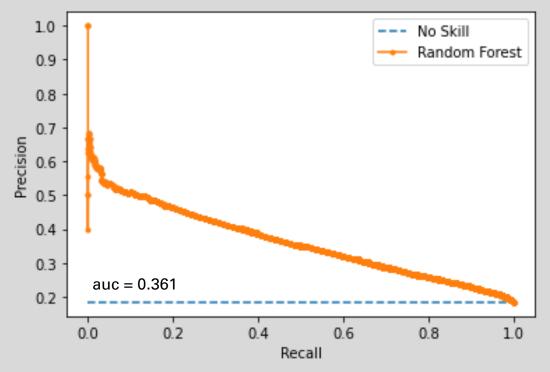
max_depth = 40,

max_features = 'auto',

min_samples_split = 2,

n_estimators = 200)
```

Random Forest Precision-Recall Curve on Test Set



MODEL TRAINING RESULTS

Train Result:

CLASSIFICATION REPORT:

	0	1	accuracy	macro avg	weighted avg
precision	1.0	1.0	1.0	1.0	1.0
recall	1.0	1.0	1.0	1.0	1.0
f1-score	1.0	1.0	1.0	1.0	1.0
support	122183	36655	5 1.0	158838	158838

Confusion Matrix:

[[122183 0] [0 36655]]

Test Result:

CLASSIFICATION REPORT:

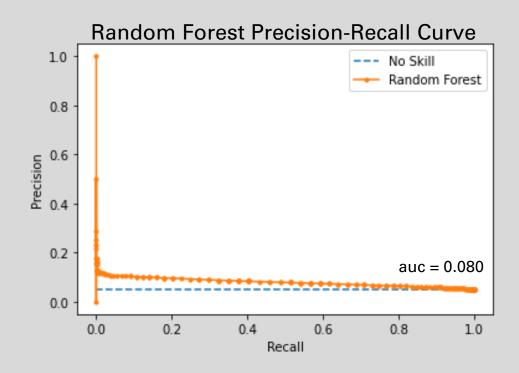
	0	1	accuracy	macro avg	weighted avg
precision	0.829281	0.506489	0.816682	0.667885	0.769944
recall	0.976400	0.107539	0.816682	0.541970	0.816682
f1-score	0.896847	0.177410	0.816682	0.537129	0.764597
support	62839.00	14153.00	0.816682	76992.00	76992.00

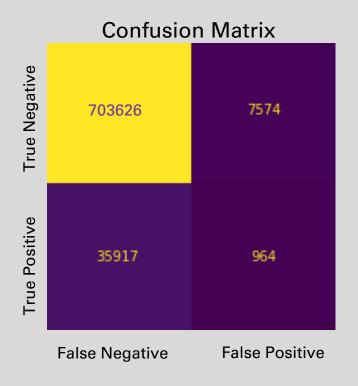
Confusion Matrix:

[[61356 1483]

[12631 1522]]

MODELING RESULTS ON HOLDOUT DATA





Model is only predicting 2.6% of true defaults on holdout data

MODEL REPORT ON HOLDOUT DATA

CLASSIFICATION REPORT								
	0	1	accuracy	macro avg	weighted avg			
precision	0.951434	0.112907	0.941863	0.532170	0.910093			
recall	0.989350	0.026138	0.941863	0.507744	0.941863			
f1-score	0.970022	0.042449	0.941863	0.506235	0.924291			
support	711200.0	36881.00	0.941863	748081.00	748081.00			

SUMMARY AND CONCLUSION

Model is not ready for business use

 Predicting more false defaults than true defaults leads to missed revenue by rejecting loan applications for loans that would be paid back

FUTURE WORK TO IMPROVE MODEL PERFORMANCE

- Collect additional data e.g., FICO scores, or data Lending Club already uses before issuing loans
- > Engineer better features e.g., applicant's debt to income
- Identify cause of the number of outlier and missing values in the data in order to start with better data