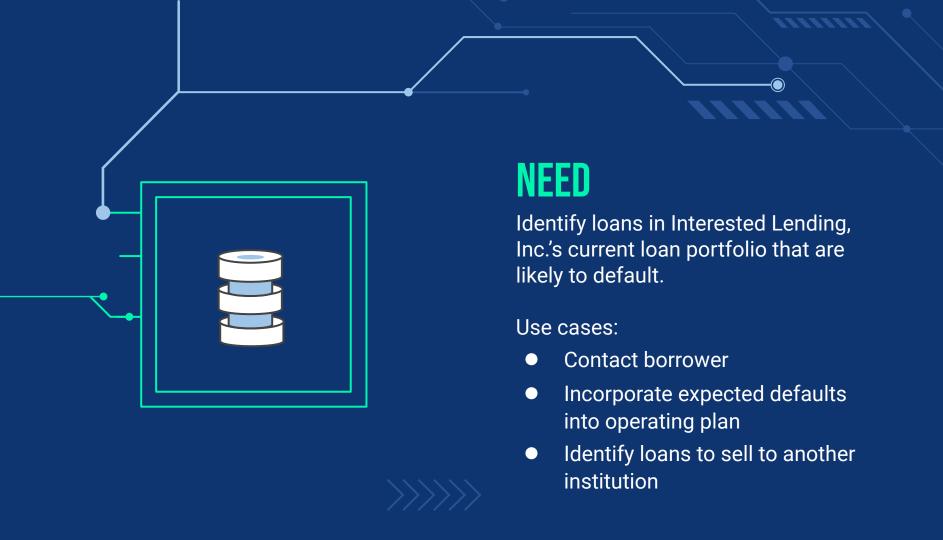
LOAN DEFAULT CLASSIFICATION

Manveer Sadhal Oct 29, 2021



DATA AND SCOPE

DATA

- Snapshot of 887,000+ individual loans.
- Filtered to approximately 250,000 loans.
- 73 columns reduced to 19 features for final model (e.g. annual income, interest rate)

SCOPE

- Current and closed loans
- Model developed with closed loans (fully paid, charged off, or defaulted).

METHODOLOGY PANDAS NUMPY MATPLOTLIB **SCIKIT-LEARN SEABORN** DATA **EVALUATE FINAL EXPLORATION & BASELINE ALTERNATE MODEL** MODEL **FEATURE MODELS SELECTION SELECTION**

XGBOOST

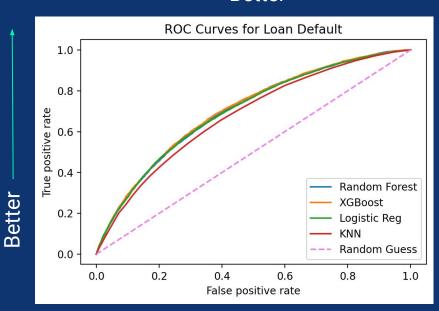
TUNE,

CROSS-VAL

& TEST

RESULTS

Better



- Logistic regression model, random forest, and XGB had nearly identical performance.
- Logistic regression selected simpler, more interpretable.

RESULTS (LOGISTIC REGRESSION)

	Precision	Recall	F1	
Paid	0.89	0.63	0.74	
Default or Charge Off	0.29	0.66	0.40	
Accuracy	0.63			
Macro Average	0.59	0.64	0.57	
Weighted Average	0.78	0.63	0.68	

RESULTS

Feature	Coefficient
Interest Rate	0.469
Annual Income	-0.284
Total Number of Accounts	-0.187
Term of 60 months (vs. 36)	0.176
Debt to Income Ratio	0.174

INTERACTIVE STREAMLIT APP

Loan Default Classification		
Enter information below. Default prediction will be displayed at the bottom of the scr	een.	
Annual income (USD)		
0		+
Number of accounts in collections within past 12 months (excluding medical)		
5	-	+
Number of delinquencies over 30 days in the last two years		
29	-	
Debt to income ratio		
29	-	+
Credit inquiries in the last 6 months		
5	-	+
Interest rate (%)		
29.00	-	

Number of open credit lines in borrower's file		
46	-	+
Number of derogatory public records		
5	-	+
Revolving debt utilization (%)		
62	-	+
Total number of credit lines in borrower's file (open or closed)		
16	-	+
Total number of accounts currently delinquent		
5	-	
Loan Term (months)		
60		-
Loan Grade		
G		*
Installment (USD)		
1215	-	+
Loan is expected to default!		

CONCLUSIONS

- Model offers predictive capability for approved loans that may go into default
- Most impactful features can inform loan screening process
- Possible actions for loans likely to default:
 - Borrower outreach
 - Identify loans to sell to other institutions
 - Account for expected loan defaults in operating plan

FUTURE WORK

- Additional feature engineering
- Model stacking
- Develop separate model to evaluate whether high risk loans should be sold
- Add functionality to app to allow user to upload CSV with data on multiple loans rather than inputting one at a time.

THANKS!









Do you have any questions?

addyouremail@freepik. com +91 620 421 838 yourcompany.com

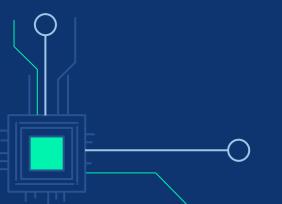
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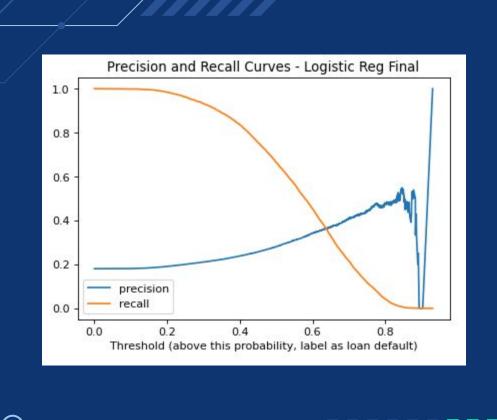




APPENDIX









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