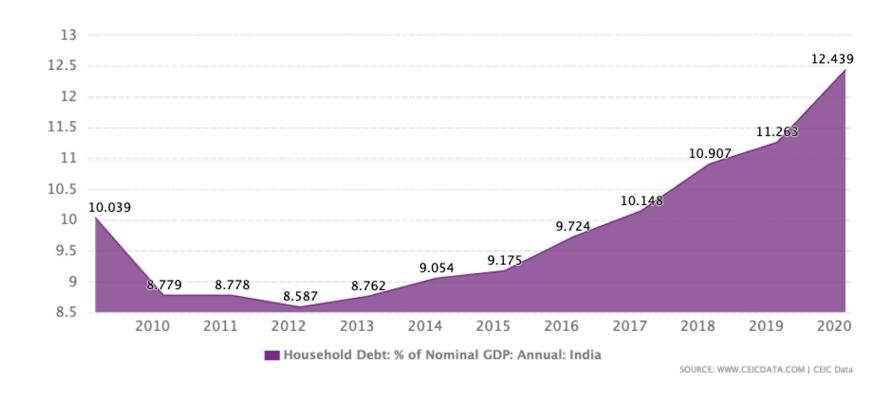


Extending a lending hand

Analyzing consumer credit data to provide financial help to consumers in need

Sam Reiff Sept 2021

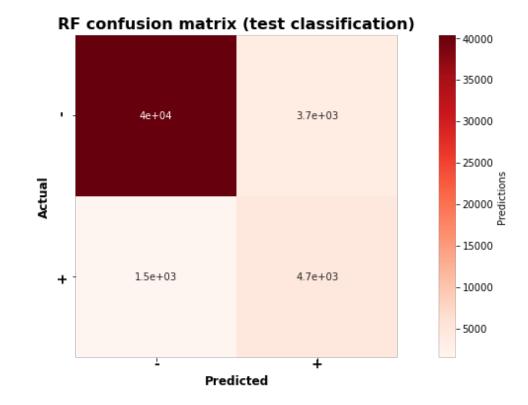
Background



- Growing household debt in India
- Consumer debt relief programs could stabilize economic growth
- Indian government seeks to use loan data to identify citizens who may benefit greatly from relief programs
- Solution should be scalable to thousands/millions more observations

Executive summary

- Motivation: Identify default-prone consumers for debt relief outreach
- Objective: Establish an effective classification algorithm with acceptable recall/F-beta
- Conclusion: Random Forest produces results superior to LR, in-line with KNN with better scalability
 - Target: default / no-default
 - Recall on test: 76%
 - F-beta on test: 71%
 - ROC on test: 94%



Random Forest has reasonable performance metrics and can scale with the business need

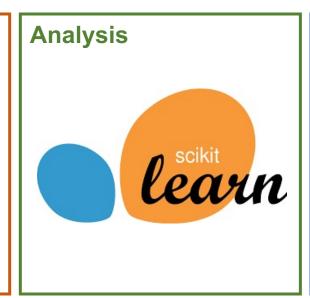
Methodology

Project workflow

Data acquisition









252k observations

— Target class: loan default

— Target distr.: ~10% default rate on sample

11 features

— Target: default/no default

— Numerical: income, age, work experience, employment, time at residence

— Categorical: marriage status, house ownership, car ownership, profession

Methodology

Project workflow



Partitioning	Training	Diagnostics	Model selection	Testing
— 60 / 20 / 20	RegularizationResampling	RecallFβROC-AUC	— Best Fβ— Interpretability— Computational cost	

Results

Validation

• KNN:

— Recall: 0.81

— Fβ: 0.72

— ROC: 0.88

Logistic Regression:

- Recall: 0.03

— Fβ: 0.04

— ROC: 0.58

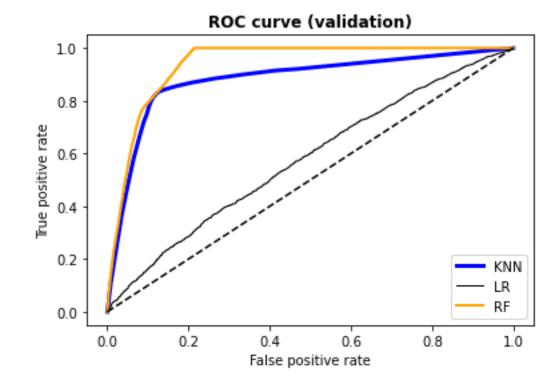
Random Forest:

- Recall: 0.76

— Fβ: 0.71

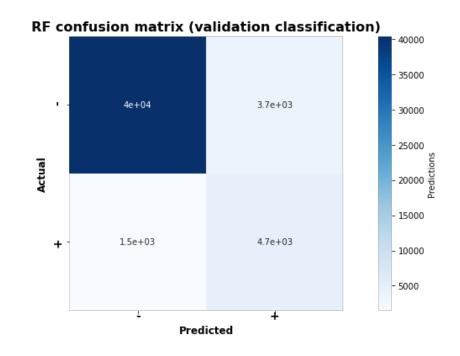
- ROC: 0.94

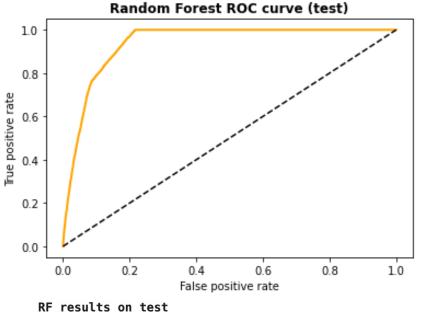
Marginally worse recall/f-beta than KNN but acceptable, and will scale better with Ks/Ms more observations



		model	recall	precision	f1	f-beta	roc
	0	KNN	0.812893	0.501441	0.620266	0.723071	0.880015
	1	Logistic Regression	0.032393	0.276860	0.058000	0.039341	0.577027
	2	Random Forest	0.760516	0.557867	0.643617	0.709006	0.938439

Conclusions





Accuracy: 0.8956150793650793 Recall: 0.7580361426515273 Precision: 0.558435438265787

F1: 0.6431042670103793 F-beta: 0.7074626865671643 ROC: 0.9378033040255929

The model holds reasonably stable on test, and could be a viable tool for credit relief outreach

Further due diligence/future work



Parameter tuning

- Decision thresholds
- KNN: neighbors, distance metric
- RF: depth, more estimators



More models to employ

- Boosted trees
- Naïve Bayes



Incorporate more data

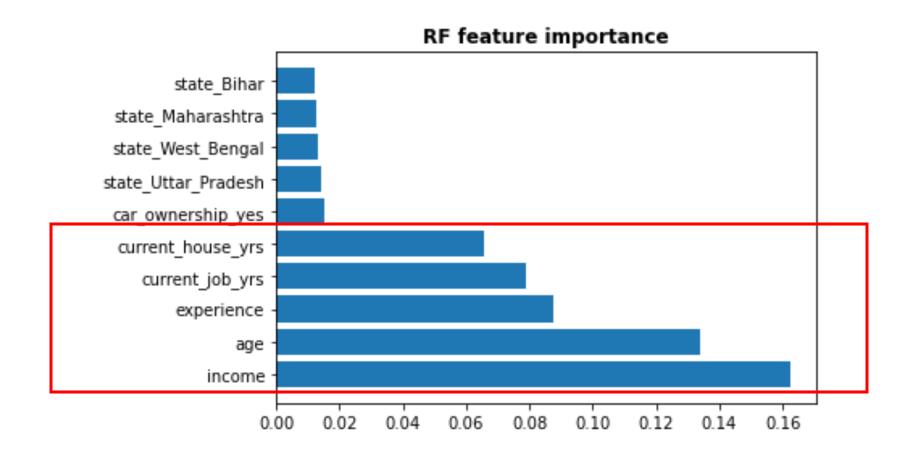
- Monthly payment data
- More demographics

Sam Reiff

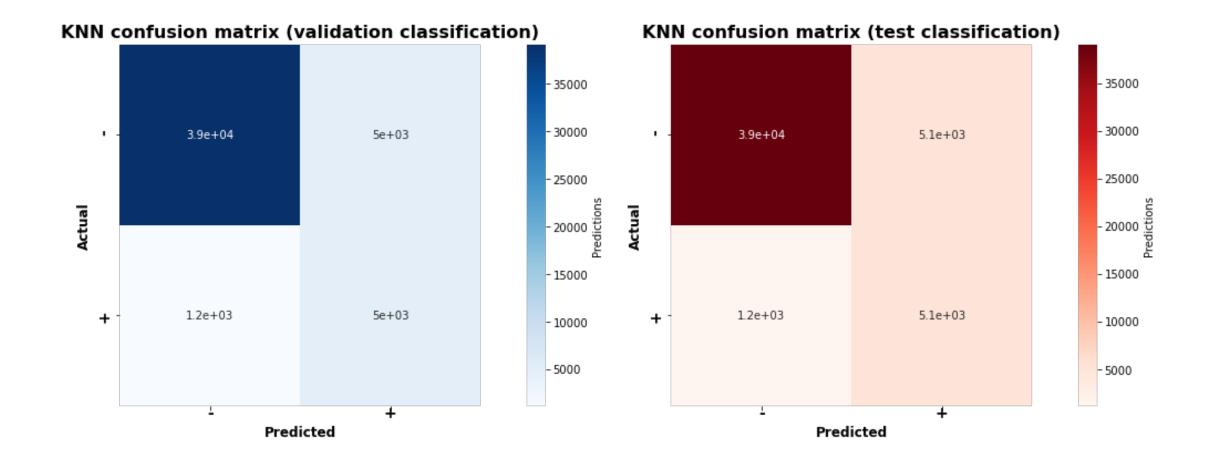
reiff.sam@gmail.com 479.426.3700

Appendix

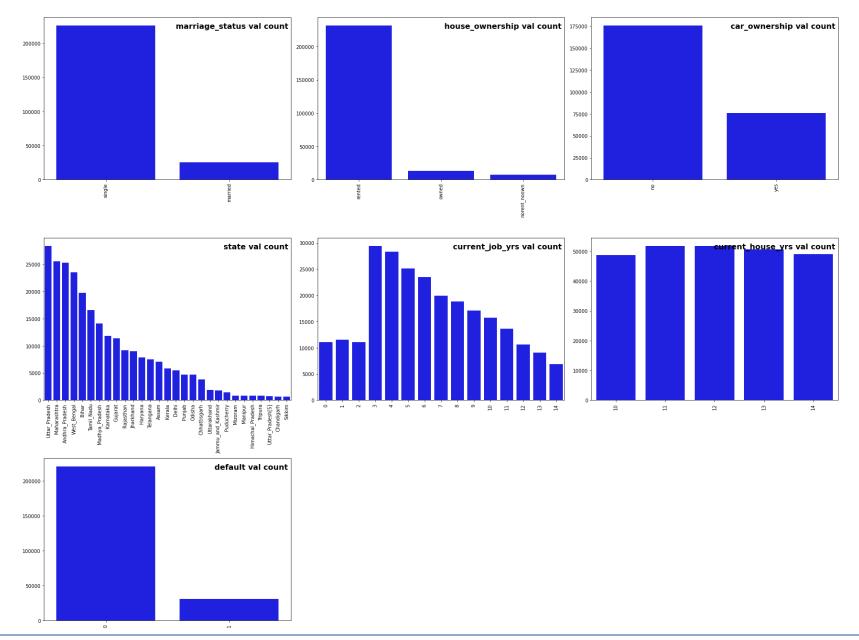
Random Forest feature importance, top ten influential feats



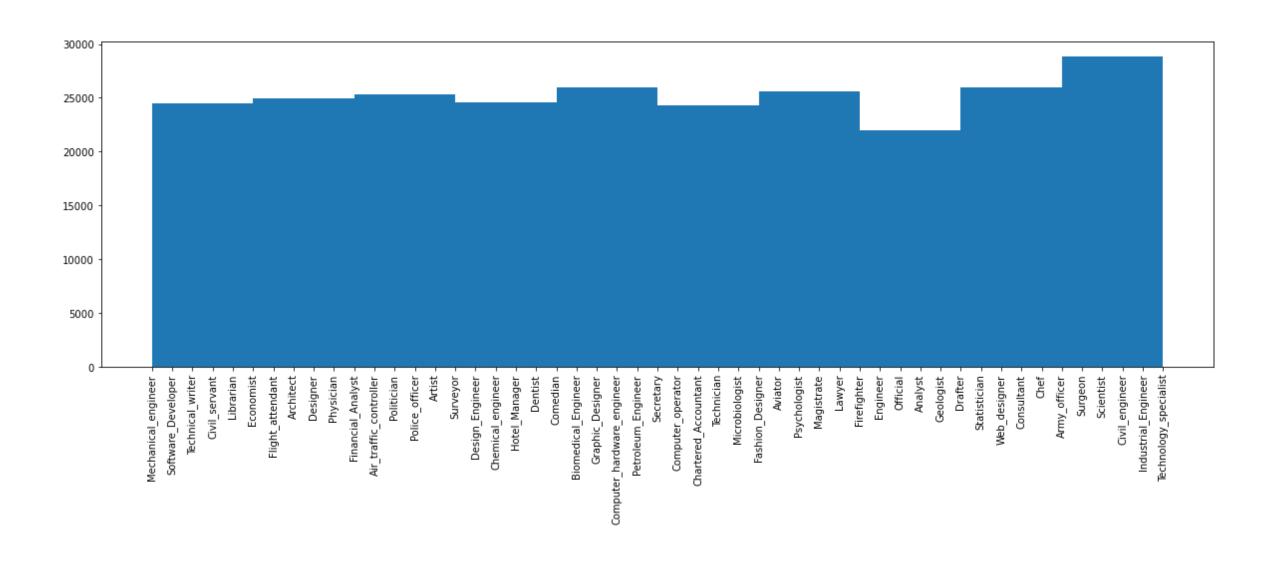
KNN confusion matrix on validation and test



Feature value counts



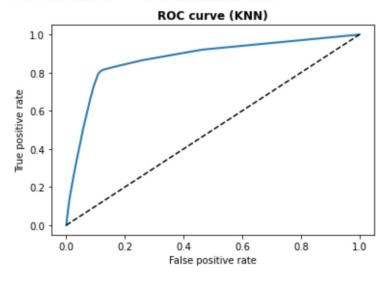
Profession feature distribution



KNN validation results

KNN classification metrics with no resampling

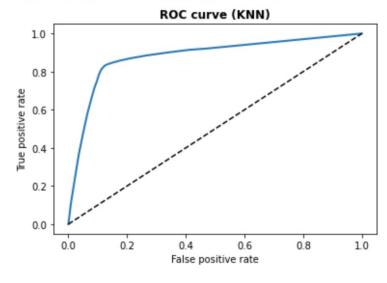
KNN classification accuracy of: 0.8882539682539683
KNN classification recall of: 0.41418211120064463
KNN classification precision of: 0.5627326472520254
KNN classification F1 of: 0.47716301522465654
KNN classification F-beta (2) of: 0.4372681798073978



KNN validation results

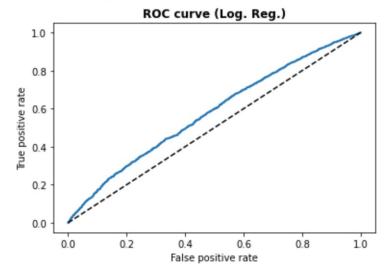
KNN classification metrics with resampling

KNN classification accuracy of: 0.8774603174603175
KNN classification recall of: 0.8128928283642224
KNN classification precision of: 0.5014414951784472
KNN classification F1 of: 0.6202656173143137
KNN classification F-beta (2) of: 0.7230711889675736



Logistic regression validation results

Logistic Regression classification metrics with no resampling

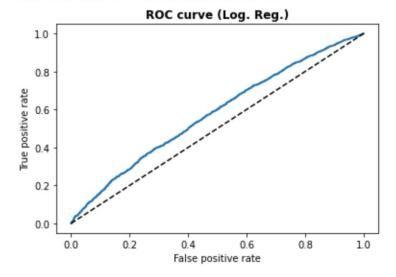


Logistic regression validation results

Logistic Regression classification metrics with resampling

```
Logistic regression (hard preds) accuracy of: 0.8704563492063492
Logistic regression (hard preds) recall of: 0.032393231265108784
Logistic regression (hard preds) precision of: 0.2768595041322314
Logistic regression (hard preds) F1 of: 0.05800028855864955
Logistic regression (hard preds) F-beta (2) of: 0.0393407969936585

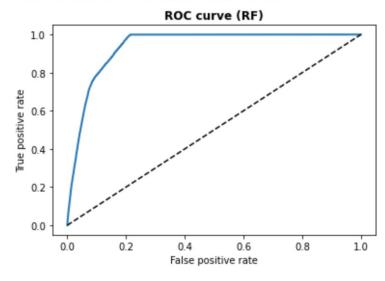
Logistic regression (soft preds with 0.17 decision threshold) accuracy of: 0.1252579365079365
Logistic regression (soft preds with 0.17 decision threshold) recall of: 0.9988718775181306
Logistic regression (soft preds with 0.17 decision threshold) precision of: 0.12327459326146625
Logistic regression (soft preds with 0.17 decision threshold) F1 of: 0.21946426358373317
Logistic regression (soft preds with 0.17 decision threshold) F-beta (2) of: 0.4126607899011958
Logistic regression (soft preds with 0.17 decision threshold) log loss of: 0.5096175315499527
```



Random forest validation results

Random Forest classification metrics with no resampling

Random forest classification accuracy of: 0.8998015873015873
Random forest classification recall of: 0.5381144238517325
Random forest classification precision of: 0.6045627376425855
Random forest classification F1 of: 0.569406548431105
Random forest classification (hard preds) F-beta (2) of: 0.5502092739676366



Random forest validation results

Random Forest classification metrics with resampling

Random forest classification accuracy of: 0.8963095238095238
Random forest classification recall of: 0.7605157131345689
Random forest classification precision of: 0.5578673602080624
Random forest classification F1 of: 0.6436170212765957
Random forest classification F-beta (2) of: 0.7090056792571892

