Bankruptcy Prediction

Problem Statement

To predict bankruptcy and the main features involved, which could help a company detect early signs and can help reduce economic losses.

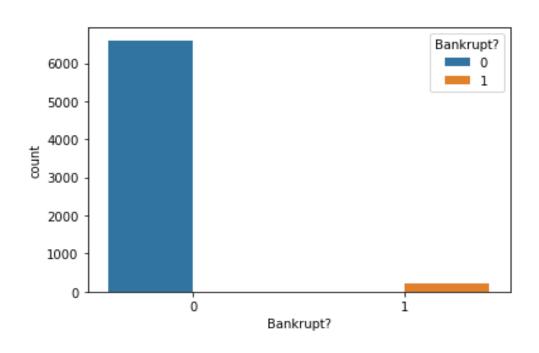
Bankruptcy Prediction

- Classification model
- ~ 7000 rows and 95 columns
- Yes/ No class label

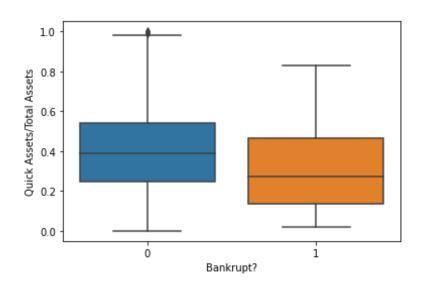
Data Wrangling

- No missing values
- Explored Outliers

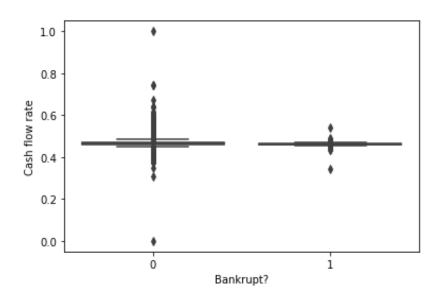
Exploratory Data Analysis



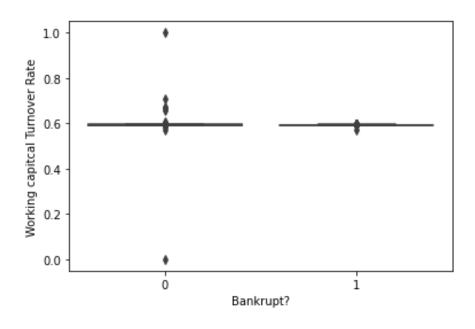
- 6599 No class label
- 220 Yes class label
- ~ 3% Yes class



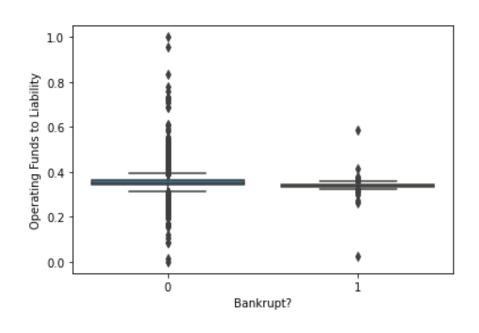
- Compared features with the class labels
- The average of Total Assets of the companies that went bankrupt were lower



Cash flow rate was ranged higher for the No class label



Working capital turnover rate ranged higher for the No class label



Operating Funds to Liability ranged higher for the No class label

Preprocessing and Training Data

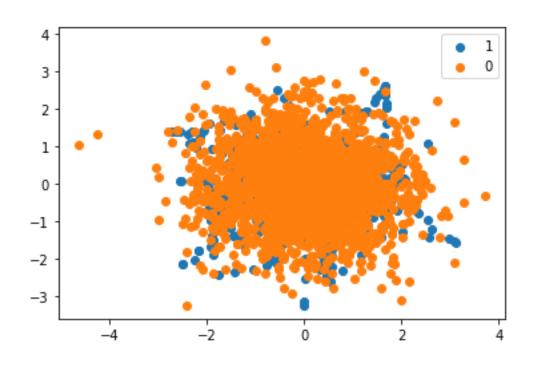
- 80-20 split
- Initial models
- Logistic Regression
- Decision trees

Preprocessing and Training Data

	precision	recall	f1-score	support
0	0.97	0.99	0.98	1313
1	0.61	0.22	0.32	51
accuracy	/		0.97	1364
macro a	0.70	0.61	0.65	1364
Wt. avg	0.96	0.97	0.96	1364

- F1-score
- Between .3-.4 for No class label for all models

Preprocessing and Training Data



- Class imbalance
- SMOTE (Synthetic Minority Oversampling Technique)

Modelling – Logistic Regression

	precision	recall	f1-score	support
0	0.98	0.97	0.97	387
1	0.94	0.96	0.95	207
accuracy			0.96	594
macro avg	0.96	0.96	0.96	594
weighted avg	0.96	0.96	0.96	594

Modelling – Decision Tree

	precision	recall	f1-score	support
0	0.90	0.93	0.91	387
1		0.94	0.95	207
accuracy	0 03	0 03	0.93	594
macro avg	0.93	0.93	0.93	594
weighted avg	0.93	0.93	0.93	594

Decision Tree – Important Features

- Liability to Equity
- Working Capital Turnover Rate
- Revenue per Share
- ROA before share % and after tax
- Operating Profit Rate
- Quick Assets/ Total Assets
- Net worth
- Average collection days
- Cash flow to sales
- Debt Ratio %

Modelling – Gradient Boosting

	precision	recall	f1-score	support
0	0.98	0.97	0.98	387
1	0.95	0.96	0.95	207
accuracy			0.97	594
macro avg	0.96	0.97	0.96	594
weighted avg	0.97	0.97	0.97	594

Modelling – XGBoost

	precision	recall	f1-score	support
0 1	0.99	0.98	0.99 0.97	387 207
accuracy			0.98	594
macro avg	0.98	0.98	0.98	594
weighted avg	0.98	0.98	0.98	594

Summary – Important steps

- Looked at the data/ outliers.
- Compared features with class labels and with each other.
- Applied initial models and got low F1-score for No class label.
- Oversampled No class label to solve class imbalance using SMOTE.
- Applied models again and saw a significant increase in accuracy.
- Important features through decision trees.
- Best prediction with XGBoost.