



Bankruptcy Prediction based on Financial and Economic ratios

David Song

ABOUT DATA

Financial data based on business regulation of the Taiwan Stock Exchange obtained from Taiwan Economic Journal from the years 1999 to 2009. Financial ratios refers to information related to business condition and prospects and are in general utilized to evaluate and analyze company's performance and financial health.

Data Summary

- Total of 95 input variables(features) representing financial ratios
- 1 Target variable named 'Bankrupt': 1 for bankrupt and 0 for not bankrupt
- Features can be categorized into 7 Sub-groups namely: Solvency, Profitability, Capital Structure ratios, Turn over ratios, Cash flow ratios, Growth and others
- Total of 6819 entries of company data
- 93 columns are data type of 'float' : Continuous variables
- 3 columns are data type of 'int' : Categorical variables

OBJECTIVE

The primary goals of this project is to analyze:

- Which financial ratio are most critical in determining the outcome of the bankruptcy?
- Which Supervised Learning models produce the best performance predicting the likelihood of bankruptcy?

DATA IMBALANCE PROBLEM

TOTAL NUMBER OF SAMPLES : 6819

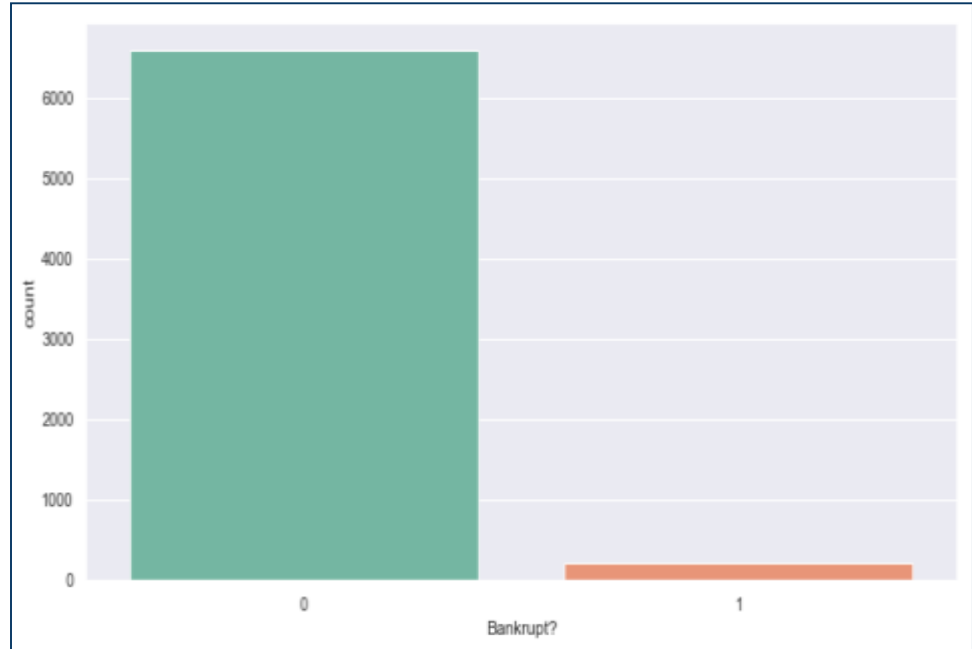
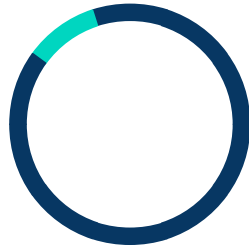
220 
BANKRUPT

6599 
NOT BANKRUPT

PERCENTAGE OF TARGET VARIABLE

3.2%

PERCENTAGE

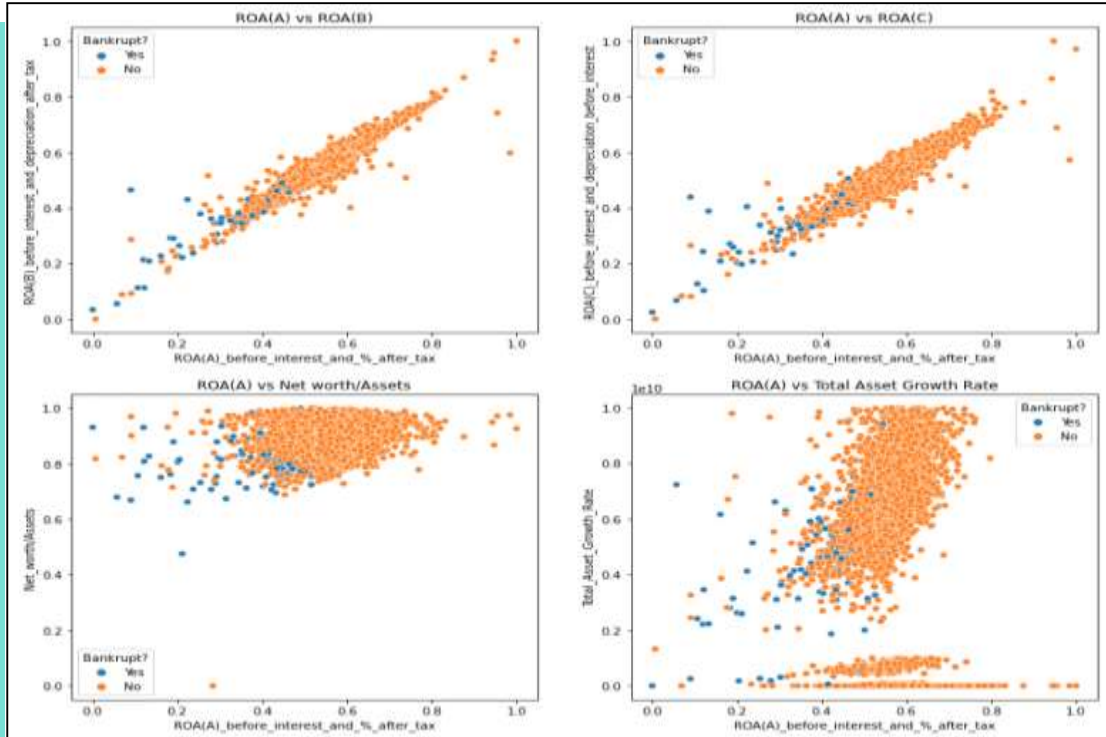


SCATTER PLOTS

Constant variable:	Comparison Variables:
ROA(A) before interest and % after tax return	ROA(B) before interest and depreciation after tax
	ROA(C) before interest and depreciation before interest
	Cash/Total Assets
	Net profit before tax/Paid in capital
	Net worth/Assets
	Total Assets Growth Rate
	Realized Sales Gross Margin
	Debt ratio%

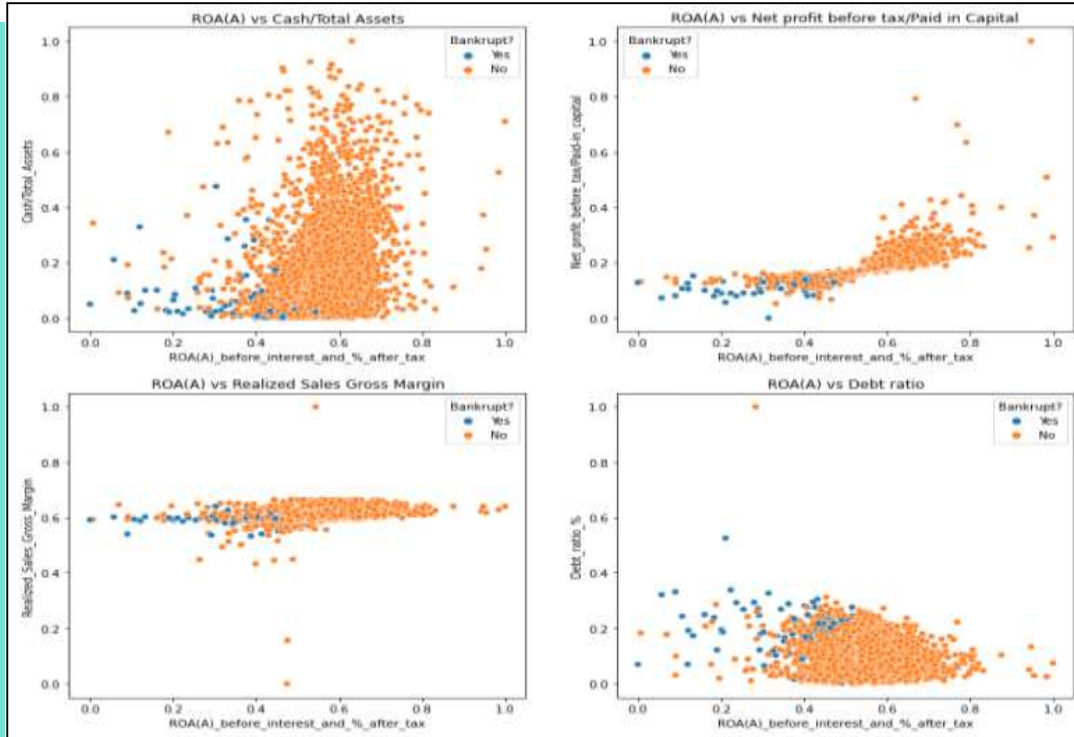
- 9 features have been randomly selected for analysis for any patterns to be observed.
- ROA(A) is selected as independent variable (x-axis)
- All other variables are arranged to dependent variable for comparison

SCATTER PLOTS (CONTINUE)



- Overall scatter plots presents clear difference on Return on Assets (ROA) between target variables. Bankrupt companies show weakness in profitability
- ROA(B) and ROA(C) reveals most significant difference between target variables
- Near perfect linear relationship in ROA(B) and ROA(C) presents they are highly correlated to each other.

SCATTER PLOTS (CONTINUE)



- Cash/Total Assets ratio values are evidently lower for bankrupt companies
- Debt ratio values are moderately higher for bankrupt companies. This signifies that chance of bankruptcy increases as debt and liability increases
- The difference of the target variable on the remaining scatter plots are inconsiderable.

MULTICOLLINEARITY

```
avoid_list = [] # Create empty list for columns to drop
for i in range(len(correlation_matrix.columns)):
    for j in range(i):
        if abs(correlation_matrix.iloc[i,j]) > 0.95: # Check correlation coefficient above 0.95 or below -0.95
            if correlation_matrix.columns[j] not in avoid_list:
                avoid_list.append(correlation_matrix.columns[j])
```

MULTICOLLINEARITY CHECK

- High correlation between the independent variables in the regression analysis could impact overall interpretation of the result.
- The features that have correlation coefficient above 0.95 or below -0.95 are dropped to avoid multicollinearity in the variables.

FEATURE SELECTION

- Backward Elimination method was considered as feature selection technique to reduce down the least significant features.

- The final count of columns filtered through both Multicollinearity check and Feature section is 29.

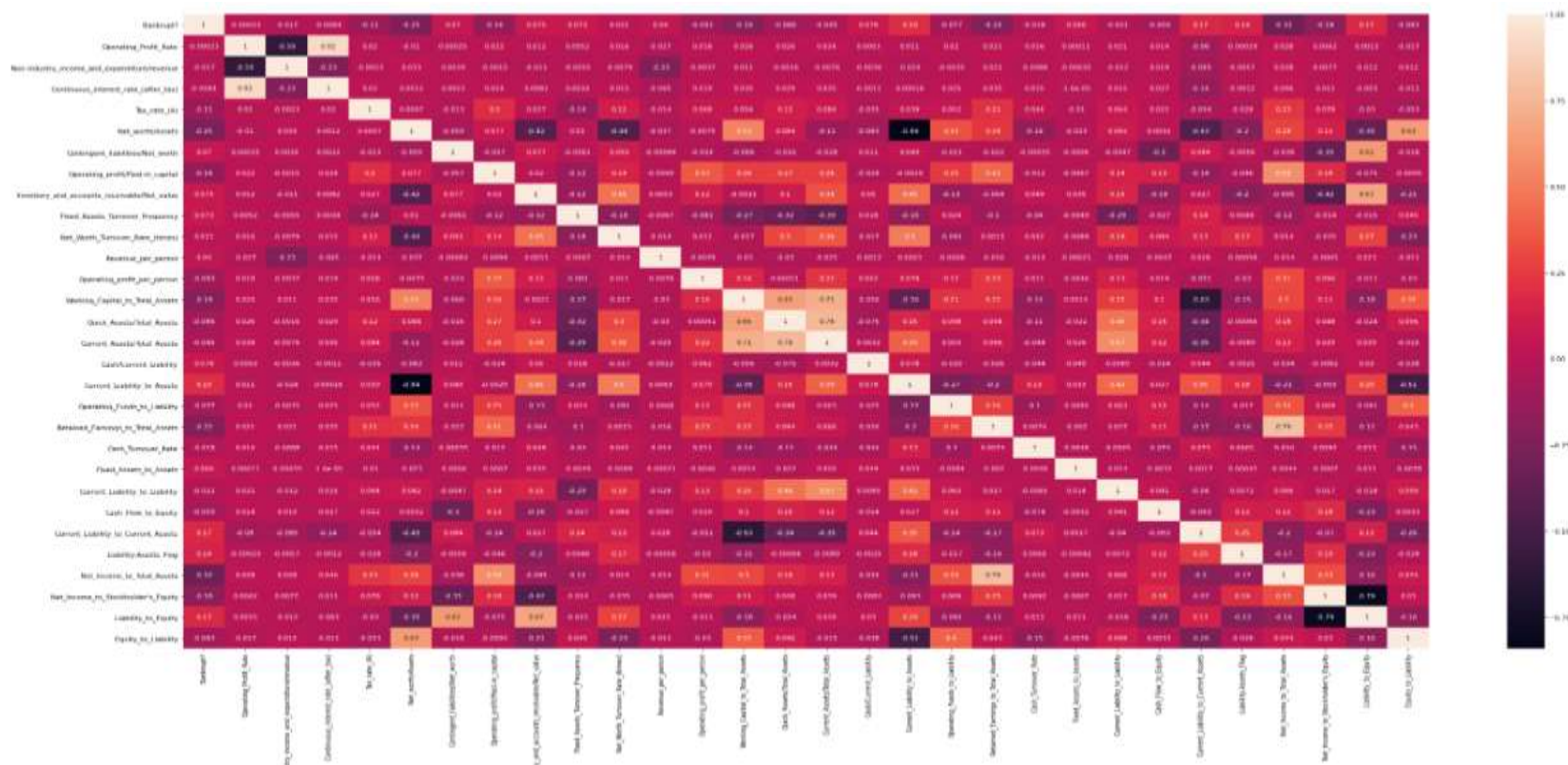
```
cols = list(X.columns)
pmax = 1
while (len(cols)>0):
    p = []
    X_1 = X[cols]
    X_1 = sm.add_constant(X_1)
    model = sm.OLS(y,X_1).fit()
    p = pd.Series(model.pvalues.values[1:],index=cols)
    pmax= max(p)
    feature_with_p_max = p.idxmax()
    if(pmax>0.05):
        cols.remove(feature_with_p_max)
    else:
        break
selected_features = cols
```

BACKWARD ELIMINATION METHOD

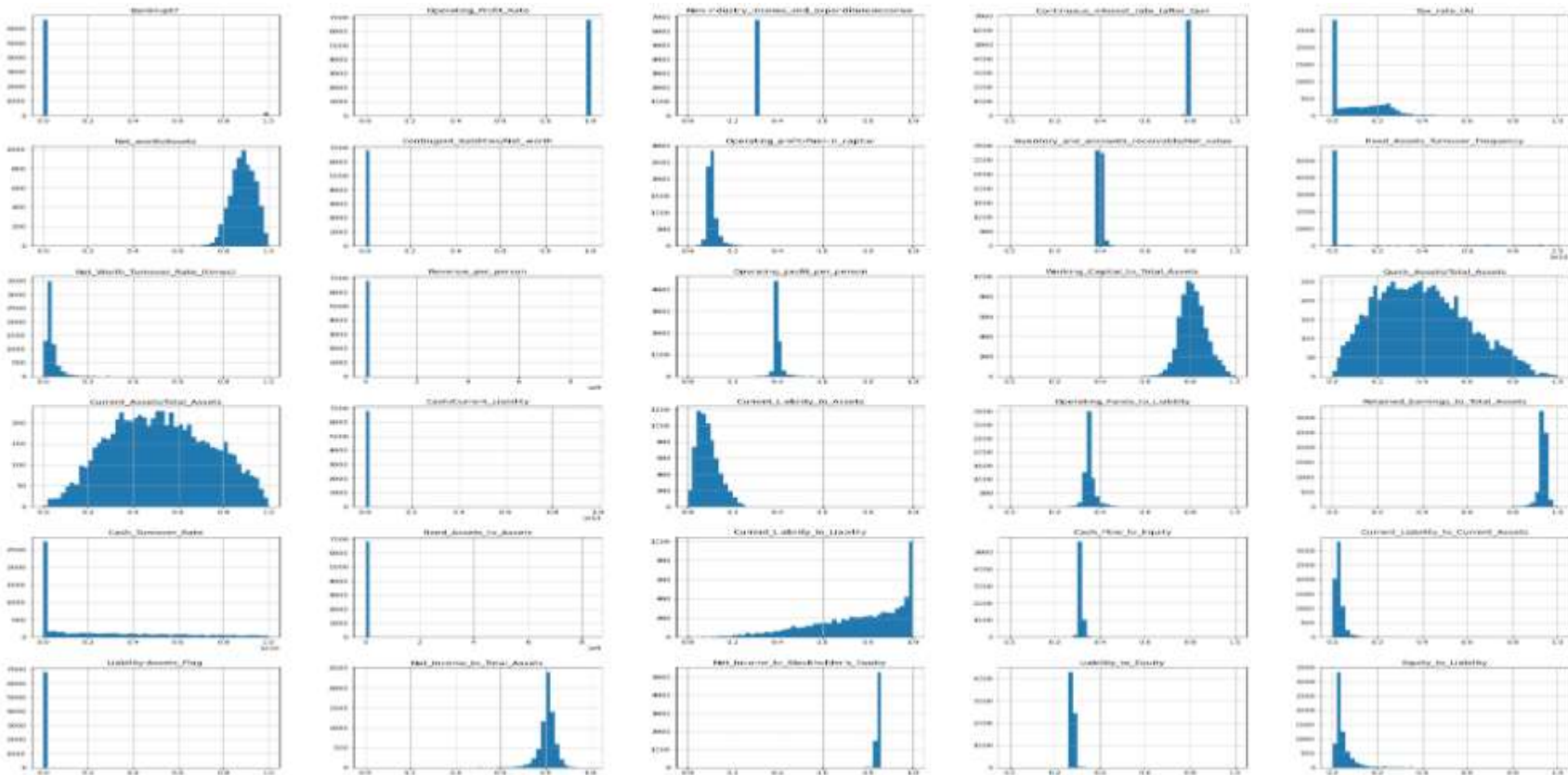
#	Column	Non-Null Count	Dtype
0	Bankrupt?	6819 non-null	int64
1	Operating_Profit_Rate	6819 non-null	float64
2	Non-industry_income_and_expenditure/revenue	6819 non-null	float64
3	Continuous_interest_rate_(after_tax)	6819 non-null	float64
4	Tax_rate_(A)	6819 non-null	float64
5	Net_worth/Assets	6819 non-null	float64
6	Contingent_liabilities/Net_worth	6819 non-null	float64
7	Operating_profit/Paid-in_capital	6819 non-null	float64
8	Inventory_and_accounts_receivable/Net_value	6819 non-null	float64
9	Fixed_Assets_Turnover_Frequency	6819 non-null	float64
10	Net_Worth_Turnover_Rate_(times)	6819 non-null	float64
11	Revenue_per_person	6819 non-null	float64
12	Operating_profit_per_person	6819 non-null	float64
13	Working_Capital_to_Total_Assets	6819 non-null	float64
14	Quick_Assets/Total_Assets	6819 non-null	float64
15	Current_Assets/Total_Assets	6819 non-null	float64
16	Cash/Current_Liability	6819 non-null	float64
17	Current_Liability_to_Assets	6819 non-null	float64
18	Operating_Funds_to_Liability	6819 non-null	float64
19	Retained_Earnings_to_Total_Assets	6819 non-null	float64
20	Cash_Turnover_Rate	6819 non-null	float64
21	Fixed_Assets_to_Assets	6819 non-null	float64
22	Current_Liability_to_Liability	6819 non-null	float64
23	Cash_Flow_to_Equity	6819 non-null	float64
24	Current_Liability_to_Current_Assets	6819 non-null	float64
25	Liability-Assets_Flag	6819 non-null	int64
26	Net_Income_to_Total_Assets	6819 non-null	float64
27	Net_Income_to_Stockholder's_Equity	6819 non-null	float64
28	Liability_to_Equity	6819 non-null	float64
29	Equity_to_Liability	6819 non-null	float64

SELECTED FEATURES

HEATMAP



HISTOGRAM

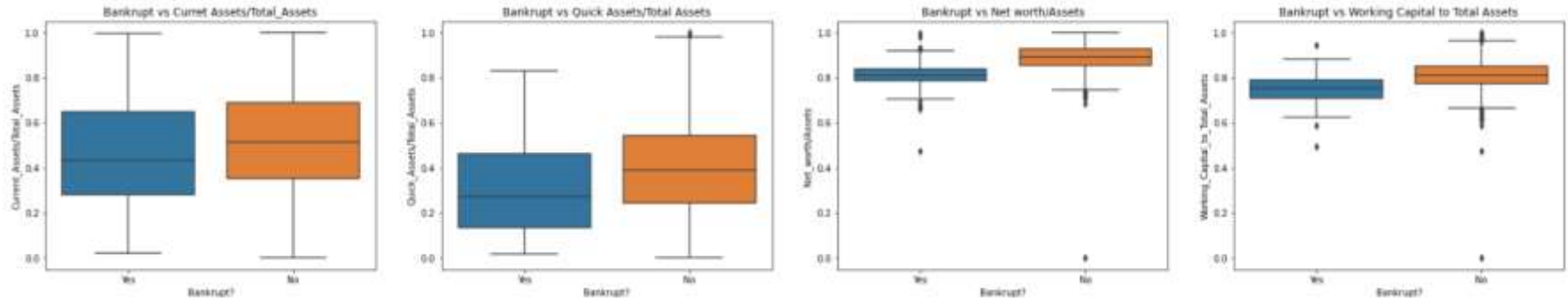


DIVIDING CATEGORIES OF NEW FEATURES

From the new columns, 16 ratios are selected and separated into 4 Major categories:

- Solvency : Asset ratios
 - ❖ Current Assets/Total Assets
 - ❖ Quick Assets/Total Assets
 - ❖ Net Worth/Assets
 - ❖ Working Capital/Total Assets
- Solvency : Liability ratios
 - ❖ Current Liability to Assets
 - ❖ Current Liability to Current Assets
 - ❖ Equity to Liability
 - ❖ Current Liability to Liability
- Turnover and Cash flow ratios
 - ❖ Net Income to Total Assets
 - ❖ Cash turnover rate
 - ❖ Retained Earnings to Total Assets
 - ❖ Cash Flow to Equity
- Operation Measures
 - ❖ Operating Profit rate
 - ❖ Operating Funds to Liability
 - ❖ Operating Profit/Paid in Capital
 - ❖ Fixed Assets Turnover Frequency

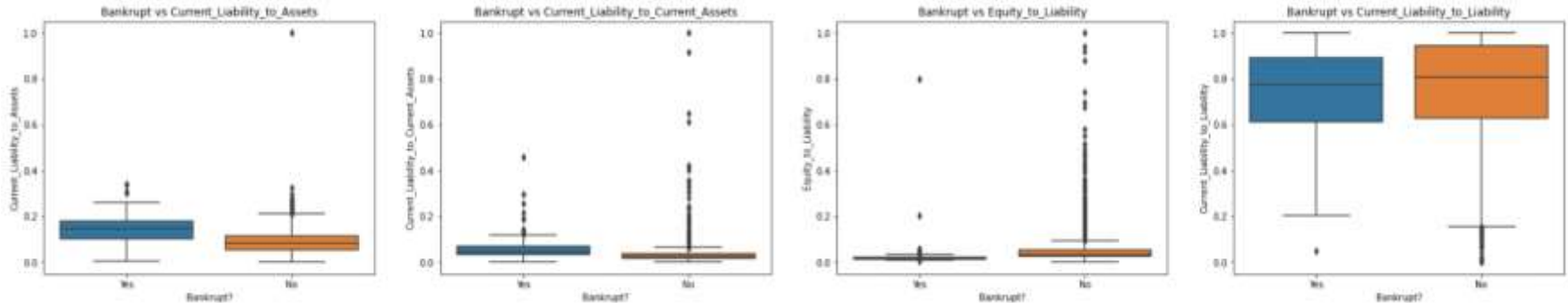
BOX PLOTS (SOLVENCY : ASSETS)



Asset ratios are the features which presents percentage of the company's assets financed by creditors

- Overall, there is noticeable difference between Bankrupt vs. Surviving companies as surviving companies extends higher measurement for all of the Asset ratios.
- The plots indicates that surviving companies are more capable of generating cash to pay off debts and meet its financial obligations thus maintaining the financial strength to avoid bankruptcy

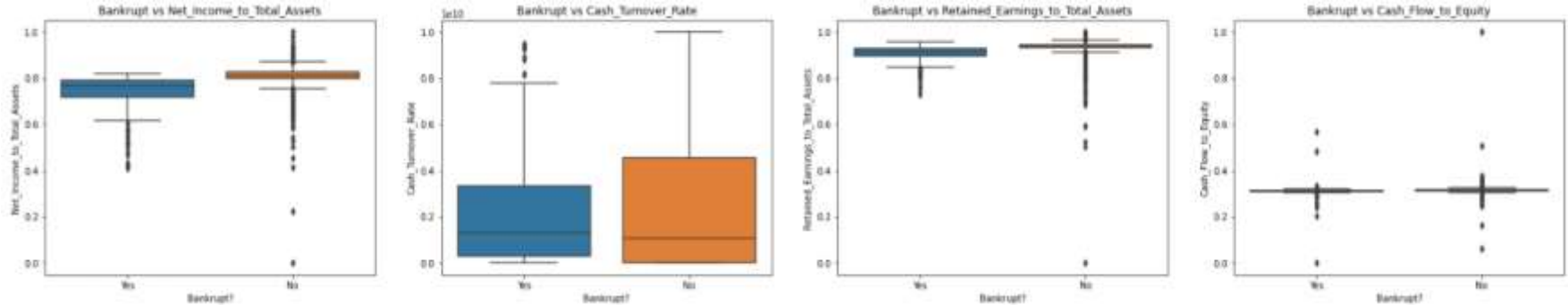
BOX PLOTS (SOLVENCY : LIABILITY)



Liability ratios are the features that define the financial obligation of the company

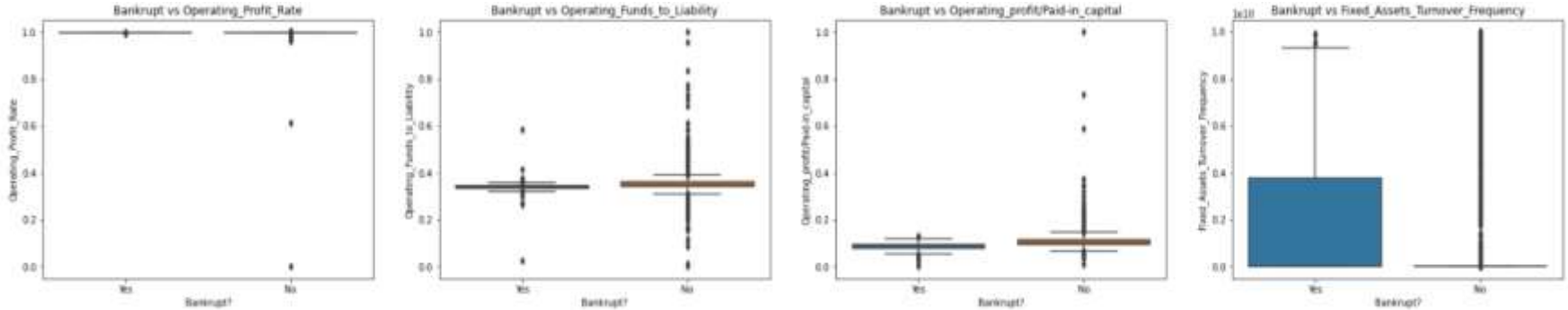
- As expected, bankrupt companies are comprised of higher ratios of liabilities compared to surviving companies
- Bankrupt companies have higher measurement in Current Liability to Assets & Current Liability to Current Assets ratios
- Observation presents that surviving companies exhibit greater strength in financial stability and are able to avoid bankruptcy due to proportionally smaller liabilities in respect to assets

BOX PLOTS (TURNOVER AND CASH FLOW RATIOS)



- Net Income to Total Assets and Retained Earnings to Total Assets display moderately higher values for the surviving companies.
- The pattern signifies that surviving companies are generally more profitable and proficient in retaining its profits to finance assets instead of paying out dividends or converting debt and new capital to fund its operations.

BOX PLOTS (OPERATION MEASURES)



- Operation measure category portrays no difference between the 2 classes. This insignificance in the plots interprets that operation measuring ratios are not a critical contributing factor in determining bankruptcy.

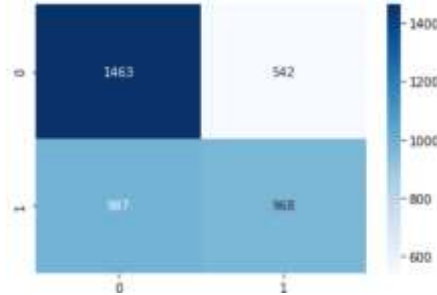
MODELLING

- 3 Supervised Learning Models are considered:
 - 1) Logistic Regression
 - 2) Random Forest Classification
 - 3) Support Vector Machine
- All models are compared between the Base Model vs. Hyperparameter Tuned Model
- Binary Classification problem | “Bankrupt” vs. “Not Bankrupt”
 - Bankrupt companies are labelled as 1
 - Not Bankrupt companies are labelled as 0

LOGISTIC REGRESSION

Accuracy: 0.613888888888889

	precision	recall	f1-score	support
0	0.68	0.73	0.66	2005
1	0.64	0.50	0.56	1955
accuracy			0.61	3960
macro avg	0.62	0.61	0.61	3960
weighted avg	0.62	0.61	0.61	3960



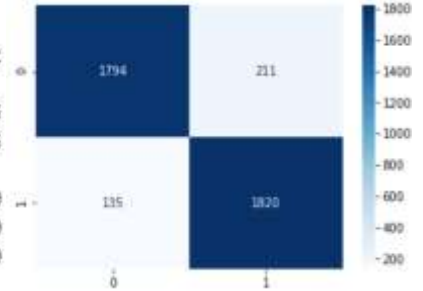
BASE MODEL

OPTIONS:

- ❑ Default parameters
- ❑ Synthetic Minority Oversampling Technique applied to solve imbalance problem

Accuracy: 0.9116262626262627

	precision	recall	f1-score	support
0	0.93	0.89	0.91	2005
1	0.98	0.93	0.91	1955
accuracy			0.91	3960
macro avg	0.91	0.91	0.91	3960
weighted avg	0.91	0.91	0.91	3960

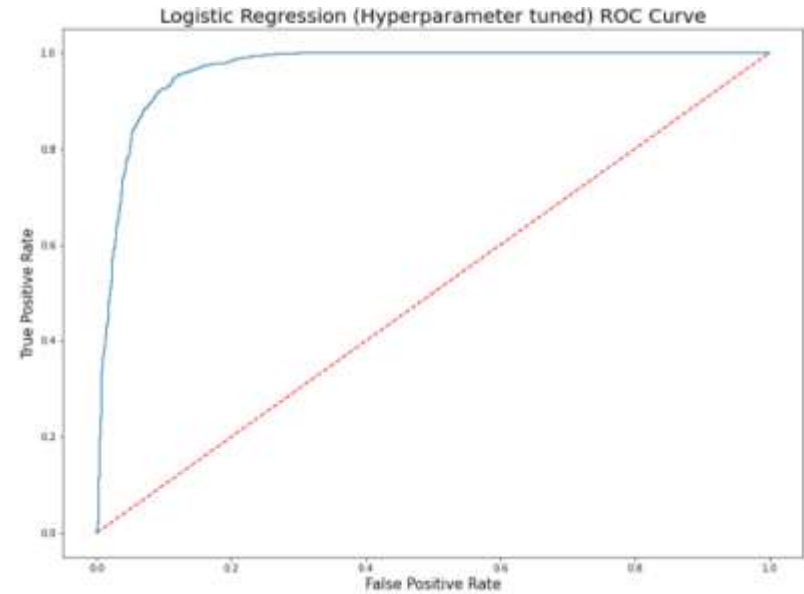
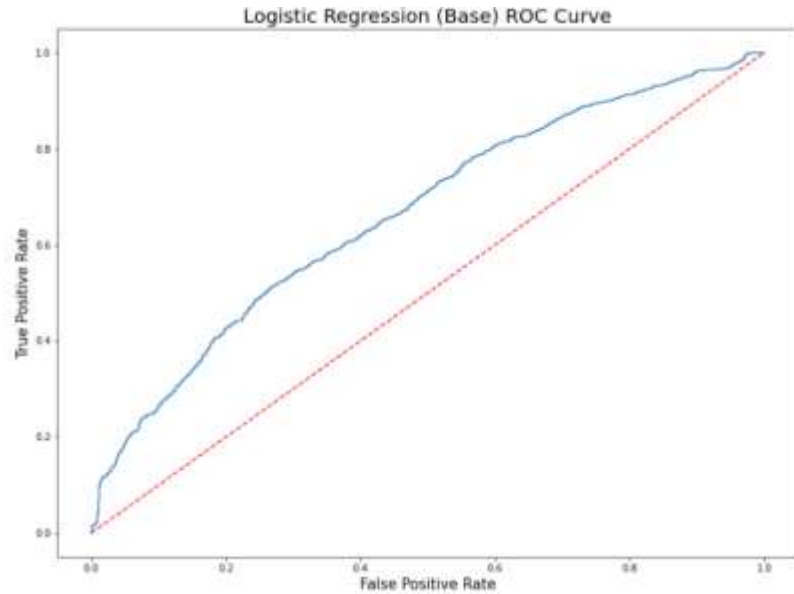


HYPERPARAMETER TUNED

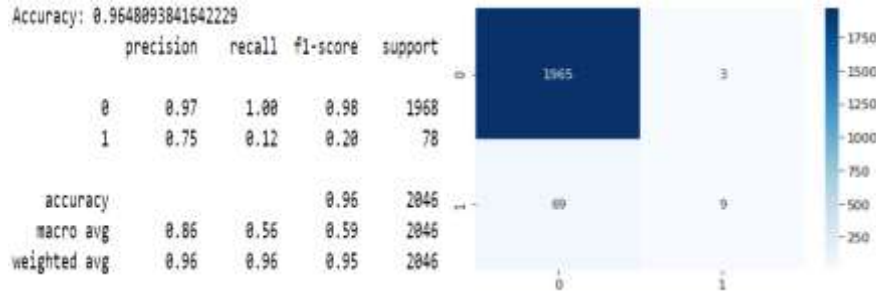
OPTIONS:

- ❑ Inverse of regularization strength(C) = 1.0
- ❑ Regularization = Ridge (L2)
- ❑ Solver option = lbfgs
- ❑ Synthetic Minority Oversampling Technique applied
- ❑ Scaling of Data-set

LOGISTIC REGRESSION (CONTINUED)



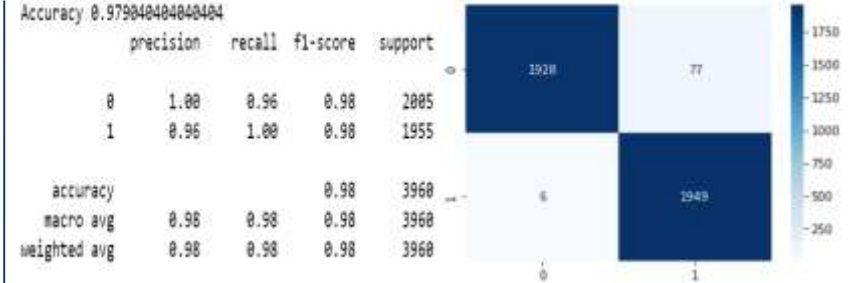
RANDOM FOREST CLASSIFIER



BASE MODEL

OPTIONS:

- ☐ Default parameters
- ☐ Original Data-set

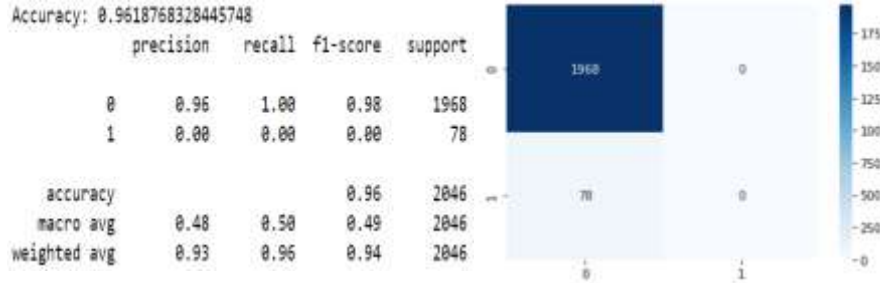


HYPERPARAMETER TUNED

OPTIONS:

- ☐ Estimators = 500
- ☐ Max tree depth = 50
- ☐ Max feature = Auto
- ☐ Criterion = Gini Impurity
- ☐ Synthetic Minority Oversampling Technique applied
- ☐ Scaling of Data-set

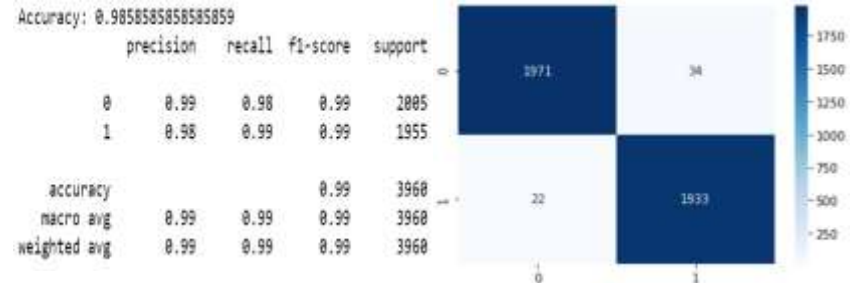
SUPPORT VECTOR MACHINE



BASE MODEL

OPTIONS:

- ☐ Default parameters
- ☐ Original Data-set



HYPERPARAMETER TUNED

OPTIONS:

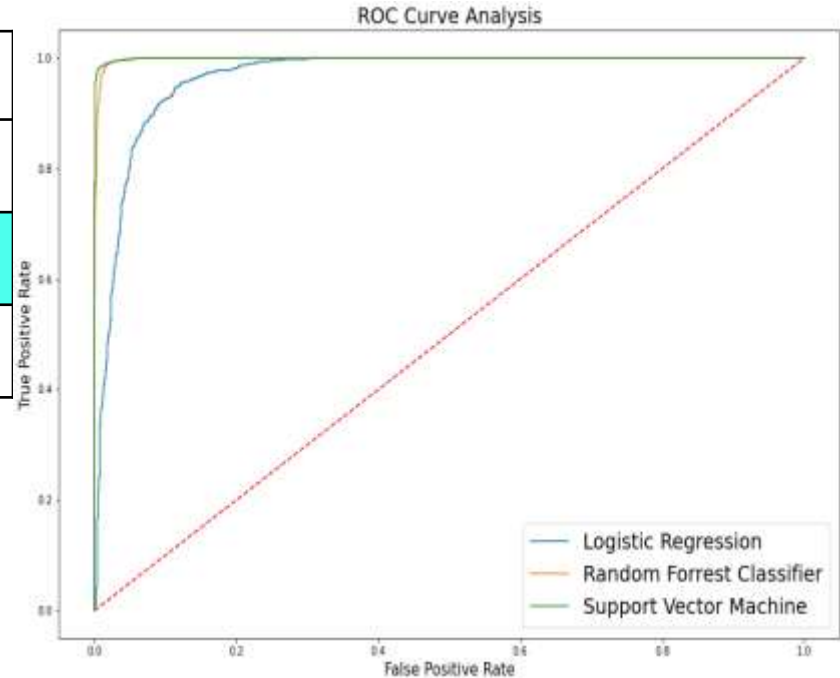
- ☐ Inverse of regularization strength (C) = 10
- ☐ Gamma = 0.1
- ☐ Kernel= Radian Basis Function (rbf)
- ☐ Synthetic Minority Oversampling Technique applied
- ☐ Scaling of Data-set

MODEL COMPARISON

Models	Accuracy	Precision	Recall	Recall Prediction
Logistic Regression	91.1%	90%	93%	1820/1955
Random Forest Classifier	97.9%	96%	100%	1949/1955
Support Vector Machine	98.6%	98%	99%	1933/1955

HYPERPARAMETER TUNED MODEL RESULTS (TARGET VARIABLE OF 1 ONLY)

- ❑ Support Vector Machine produces the highest accuracy score however,
- ❑ Random Forest Classifier has the highest Recall score predicting 1949 samples out of 1955 bankruptcy counts from the test data-set



CONCLUSION

Which financial ratio are most critical in determining the outcome of the bankruptcy?

- ❖ Profitability and Solvency are most critical features affecting company's financial stability
- ❖ Asset and Liability ratios has the largest divergences between Bankrupt vs. Not Bankrupt companies

Which Supervised Learning models produce the best performance predicting the likelihood of bankruptcy?

- ❖ Random Forest Classifier demonstrates the best performance of Bankruptcy prediction
- ❖ Recall score of 100% | Prediction: 1949 out of 1955