

# Company Bankruptcy Prediction with Naive Bayes Algorithm

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# Situation & Project Goals

- Situation

- Due to economical difficulties, companies are facing bankruptcy
- Over the 5 years period, **2091** Polish companies have bankrupted among **43405** observations.

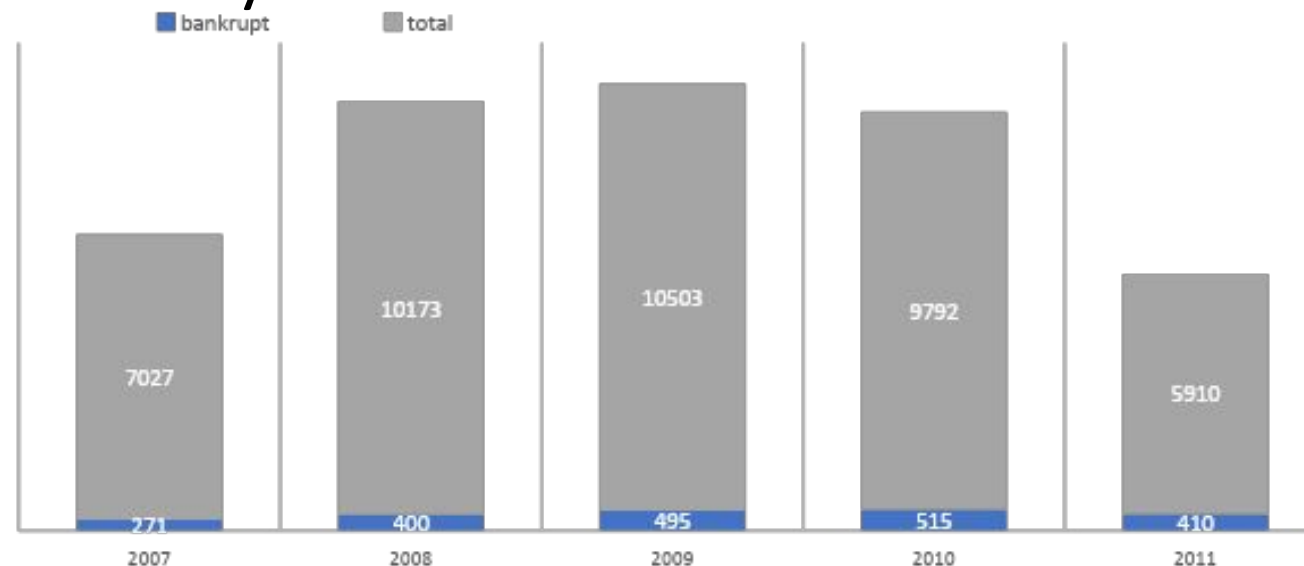
- Goals

- Develop a prediction model to decide whether a company will bankrupt in the following year or not
- Model's bankruptcy prediction power should be as good as Random Forest classifier
- Model should be scalable and fast to generate results

# Main Findings

- Predicting whether a company will bankrupt in the following year or not will give the investors and banks a better insight, and reduce possible waste of time and money.

- Proposed model has an accuracy score of **92.70%**



- Proposed model can run within 2 seconds.

# Approach

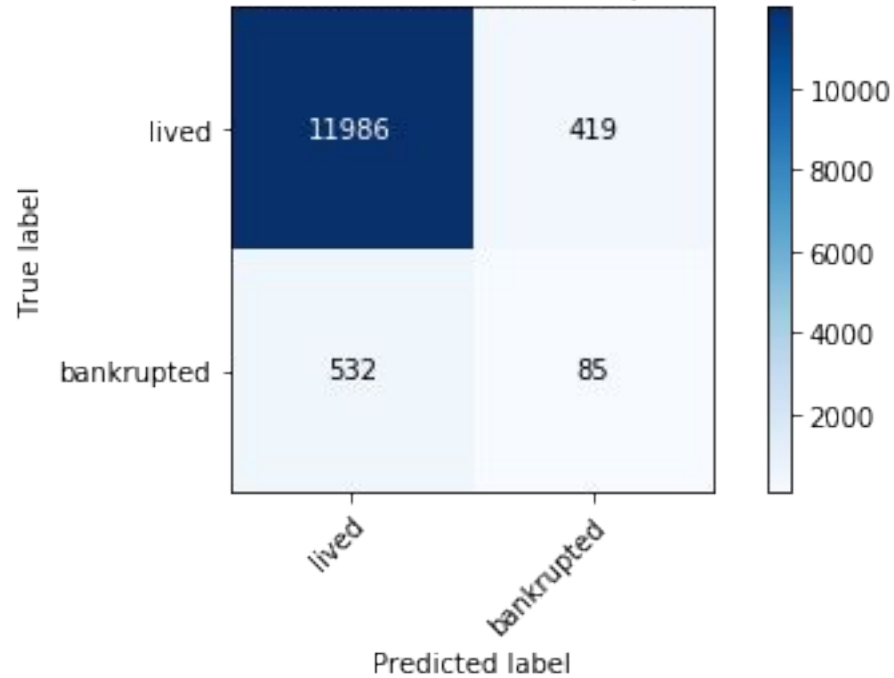
- Developed a bankruptcy model in Python using Naive Bayes based techniques
  - Easy to build, powerful on even large datasets
  - Classification based on conditional probability
  - Can handle multiple features
- Impact of using different subsets of features is investigated
- Model can be easily deployed and run

# Model Description

- **Overview of the Basic Methodology:** Predict whether the given company will bankrupt in the following year or not
- **Model:** Naive Bayesian Classifier
- **Dependent Variable:** Binary variable, of bankrupt/no bankrupt
- **Scope:**
  - 43405 instances of 5 year's company financial data
  - 64 features and the outcome (bankrupt/no bankrupt)
  - Different feature subsets are selected
- **Sampling:**
  - 70 / 30 stratified train test split (StratifiedShuffleSplit) (30383 – 13022)
- Model's bankruptcy prediction power should be as good as Random Forest
  - Created a baseline model with Random Forest Classifier

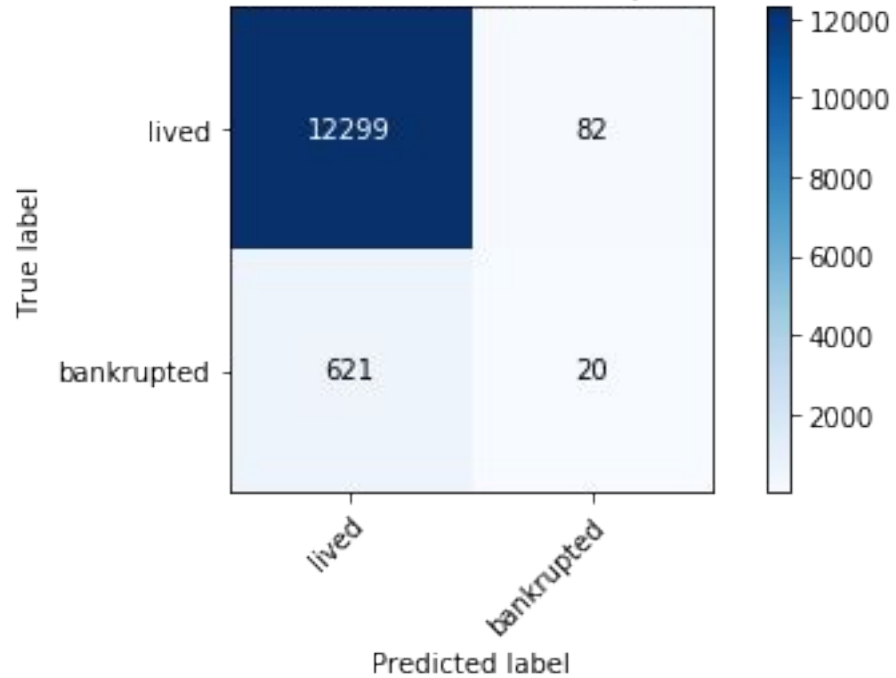
# Key Points

Confusion matrix, Bernoulli NB 9 important features



Accuracy: 92.70%  
Precision: 95.75%  
Recall: 96.62%

Confusion matrix, Random Forest Important Features



Accuracy: 94.60%  
Precision: 95.19%  
Recall: 99.34%

# Model Details

- Candidate Variables: 64 financial indicators
- Used variables: 9 features according to feature ranking
- Naive Bayes Model built in Python
  - Stratified sampling as the dataset is imbalanced
  - Accuracy, Precision and Recall calculated to measure success

$$P(x_i | y) = P(i | y)x_i + (1 - P(i | y))(1 - x_i)$$

```
from sklearn.naive_bayes import BernoulliNB
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```
bnb = BernoulliNB()
```

```
y_pred = bnb.fit(X_train, Y_train).predict(X_test)
```

# Recommendations

- Train the model again every year, as more recent data becomes available
  - Otherwise the model can be outdated
- Run the model to predict a company when an indicator is updated
  - A trigger can be set on the DB
- Re-evaluate the feature importances again when a new model is built
  - Using only 9 most important features gave the best result
  - Try different subsets of features



# References

Tomczak, S., 2016. **Polish Companies Bankruptcy Dataset**. [Online]

Available at: <https://archive.ics.uci.edu/>

Zieba, M., Tomczak, S. K. & Tomczak, J. M., 2016. **Ensemble Boosted Trees with Synthetic Features Generation in Application to Bankruptcy Prediction**. Expert Systems with Applications, Volume 58, pp. 93-101.

Sci-kit Learn Developers, 2019. Naive Bayes. [Online]

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# Model Details: Variables

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1	X16 (gross profit + depreciation) / total liabilities
2	X52 (short-term liabilities * 365) / cost of products sold
3	X32 (current liabilities * 365) / cost of products sold
4	X28 working capital / fixed assets
5	X5 [(cash + short-term securities + receivables - short-term liabilities) / (operating expenses - depreciation)] * 365
6	X40 (current assets - inventory - receivables) / short-term liabilities
7	X9 sales / total assets
8	X11 (gross profit + extraordinary items + financial expenses) / total assets
9	X59 long-term liabilities / equity
10	X23 net profit / sales
11	X25 (equity - share capital) / total assets
12	X55 working capital
13	X17 total assets / total liabilities
14	X14 (gross profit + interest) / total assets
15	X29 logarithm of total assets
16	X13 (gross profit + depreciation) / sales
17	X58 total costs / total sales
18	X30 (total liabilities - cash) / sales
19	X57 (current assets - inventory - short-term liabilities) / (sales - gross profit - depreciation)
20	X56 (sales - cost of products sold) / sales

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# All Features

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X1 net profit / total assets	X33 operating expenses / short-term liabilities
X2 total liabilities / total assets	X34 operating expenses / total liabilities
X3 working capital / total assets	X35 profit on sales / total assets
X4 current assets / short-term liabilities	X36 total sales / total assets
X5 [(cash + short-term securities + receivables - short-term liabilities) / (operating expenses - depreciation)] * 365	X37 (current assets - inventories) / long-term liabilities
X6 retained earnings / total assets	X38 constant capital / total assets
X7 EBIT / total assets	X39 profit on sales / sales
X8 book value of equity / total liabilities	X40 (current assets - inventory - receivables) / short-term liabilities
X9 sales / total assets	X41 total liabilities / ((profit on operating activities + depreciation) * (12/365))
X10 equity / total assets	X42 profit on operating activities / sales
X11 (gross profit + extraordinary items + financial expenses) / total assets	X43 rotation receivables + inventory turnover in days
X12 gross profit / short-term liabilities	X44 (receivables * 365) / sales
X13 (gross profit + depreciation) / sales	X45 net profit / inventory
X14 (gross profit + interest) / total assets	X46 (current assets - inventory) / short-term liabilities
X15 (total liabilities * 365) / (gross profit + depreciation)	X47 (inventory * 365) / cost of products sold
X16 (gross profit + depreciation) / total liabilities	X48 EBITDA (profit on operating activities - depreciation) / total assets
X17 total assets / total liabilities	X49 EBITDA (profit on operating activities - depreciation) / sales
X18 gross profit / total assets	X50 current assets / total liabilities
X19 gross profit / sales	X51 short-term liabilities / total assets
X20 (inventory * 365) / sales	X52 (short-term liabilities * 365) / cost of products sold
X21 sales (n) / sales (n-1)	X53 equity / fixed assets
X22 profit on operating activities / total assets	X54 constant capital / fixed assets
X23 net profit / sales	X55 working capital
X24 gross profit (in 3 years) / total assets	X56 (sales - cost of products sold) / sales
X25 (equity - share capital) / total assets	X57 (current assets - inventory - short-term liabilities) / (sales - gross profit - depreciation)
X26 (net profit + depreciation) / total liabilities	X58 total costs / total sales
X27 profit on operating activities / financial expenses	X59 long-term liabilities / equity
X28 working capital / fixed assets	X60 sales / inventory
X29 logarithm of total assets	X61 sales / receivables
X30 (total liabilities - cash) / sales	X62 (short-term liabilities * 365) / sales
X31 (gross profit + interest) / sales	X63 sales / short-term liabilities
X32 (current liabilities * 365) / cost of products sold	X64 sales / fixed assets

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