T.C.

MARMARA UNIVERSITY

FACULTY of ENGINEERING

CSE4062 Introduction to Data Science and Analytics

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Group #1

Delivery #4: Predictive Analysis

Title of the Project

Machine Learning Approach to U.S. Stock Investments

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1- Feature Selection Methods

Due to data having large numbers of missing values we drop the attributes that has more than 33% missing values and we replace missing values with mean value of attributes.

Table 1. Mutual Information

Results for Mutual Information		
Attribute	Score	Missing Values %
Total non-current assets	0.022112280957419017	27.27725687366943
EBIT Margin	0.022054091974503587	7.582551977170811
eBITperRevenue	0.01920992042646863	12.963717896453323
Gross Profit	0.019154132780264455	5.467228337183494
payoutRatio	0.018948078374790178	10.82574625175522
Total non-current liabilities	0.018644065311503644	27.340671286859628
Earnings before Tax	0.01823924284012657	6.178375685102142
ebitperRevenue	0.018153038022910906	12.963717896453323
EPS	0.017993695863804504	5.893010825746252
Earnings Before Tax Margin	0.017921437192584877	5.41287312587761
Free Cash Flow margin	0.01782413204082678	10.377315758481679
Profit Margin	0.016968018418720643	11.070344702631699
netProfitMargin	0.01651437783387477	12.963717896453323
EBITDA Margin	0.016334752774368688	11.08393350545817
Total shareholders equity	0.016233635705729332	6.214612492639398
dividendYield	0.016230110792283314	14.938623907233772
Consolidated Income	0.016169867868345955	7.097884676360013
Other Assets	0.016166675280912646	30.601983965212664
Net Profit Margin	0.016102373873498665	7.799972822394347
Net Income	0.01600560547987806	7.088825474475699

Table 2. ANOVA F-Value

Results for ANOVA F-Value		
Attribute	Score	Missing Values %
Dividend payments	40.99933856904691	9.765819631290483
Gross Margin	32.31890334987439	5.4309915296462385
Earnings before Tax	31.54958855350051	6.178375685102142
Free Cash Flow	31.143068580384302	7.424015944195316
Operating Income	27.759629645131373	4.987090637314853
Stock-based compensation	27.303706581127024	7.849798432758074
SG&A Expense	23.464233709464327	7.5599039724600265
Cash and short-term investments	22.72453457323068	11.210762331838565
R&D Expenses	22.336960734419254	9.684286814331657
Issuance (buybacks) of shares	22.1082255055086	8.796485029668887
EBITDA	20.253538575804583	7.944920052543371
Gross Profit	19.585008492335405	5.467228337183494
EBIT	16.072282497737262	6.4546813425737195
Retained earnings (deficit)	15.51009740357719	5.286044299497214
ebitperRevenue	15.122352264938574	12.963717896453323
Goodwill and Intangible Assets	14.643294274095304	6.509036553879604
Total current liabilities	14.41323496095706	11.450831181772886
Short-term investments	14.142473710094146	12.284277755129773
Operating Expenses	13.792811693653622	7.709380803551207
Net Income Com	13.71337691993083	6.300674910540382

To conclude we choose our best 20 attributes by using ANOVA F-value. We also try with Mutual Information but with MI, our 20 attributes with missing value percentages before replacing gets very close to our threshold of 33%. We want to use more raw data; therefore, we choose ANOVA F-value for feature selection.

2- Classification Experiments

We did not use Cross Validation or any other similar methods alike. We split our train and test set according to years. Last year's csv, 2018, is our test set.

The reasoning behind this is we thought ourselves at the end of 2018 and wanted to profit next year by using the algorithm. Thus, we use features selected in ANOVA F-Value.

Methods used for classification in this experiment;

- Decision Tree with Gini Index which is calculated by subtracting the sum of squared probabilities of each class from one.
- Decision Tree with Gain Ratio which determines the information gain of all the attributes, and then computes the average information gain.
- Naïve Bayes which are based on applying Bayes' theorem with strong independence assumptions between the features.
- Artificial Neural Networks which are designed to simulate the way the human brain analyzes and processes information.
- K Nearest Neighbor assumes that similar things exist in close proximity.

Table 3. Table for Evaluation for Classification Experiments

	Experiment	Accuracy	F1-macro	F1-micro	AUC
0	Decision Tree with Gini Index	0.528916	0.510808	0.528916	0.534653
1	Decision Tree with Gain Ratio	0.521403	0.505316	0.521403	0.531224
2	Naive Bayes	0.670993	0.428472	0.670993	0.603608
3	ANN with 1 hidden layer	0.680328	0.629554	0.680328	0.682963
4	ANN with 2 hidden layer	0.561475	0.555338	0.561475	0.631891
5	KNN 3	0.570355	0.550413	0.570355	0.592899
6	KNN 9	0.602687	0.579037	0.602687	0.624357
7	KNN 149	0.665528	0.627639	0.665528	0.670177

The performance evaluation table shows the most accurate method is ANN with 1 hidden layer. Considering Area Under Curve (AUC) which measures performance across all possible classification thresholds it suggests ANN with 1 hidden layer overperforms when compared with Naïve-Bayes which is also less accurate. Also, F1-micro (micro-averages) suggests ANN with 1 hidden layer performs better. Furthermore, F1-macro (macro-averages) indicate ANN with 1 hidden layer performs better.

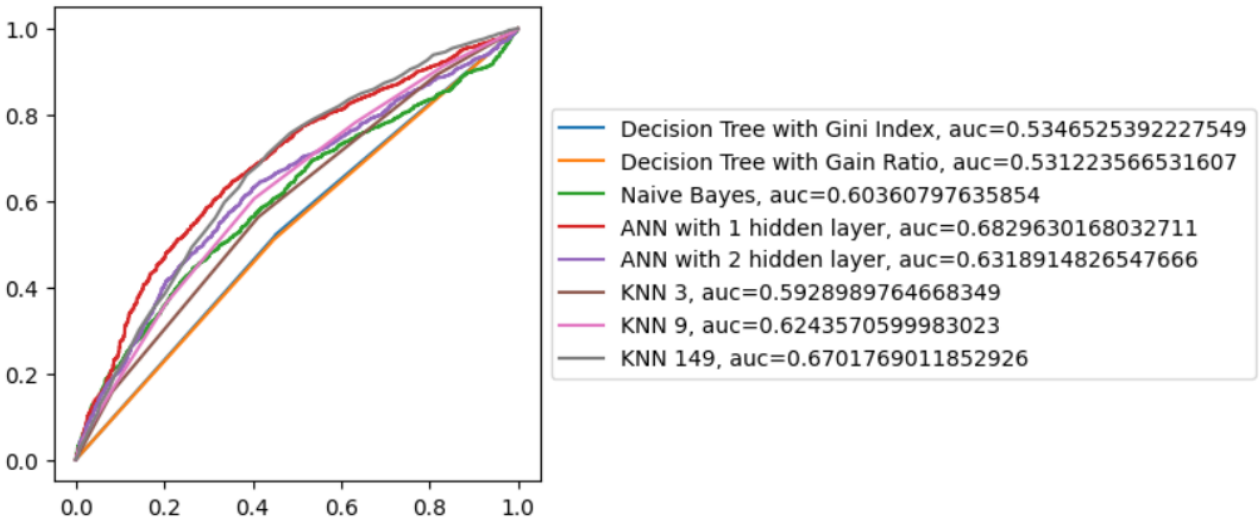

Fig.1 ROC Curve

The ROC curve shows the trade-off between sensitivity and specificity. Classifiers that give curves closer to the top-left corner indicate a better performance. The closer the curve comes to the 45-degree diagonal of the ROC space, the less accurate the test. This also shows ANN with 1 hidden layer is the best performing method followed by Naïve-Bayes.

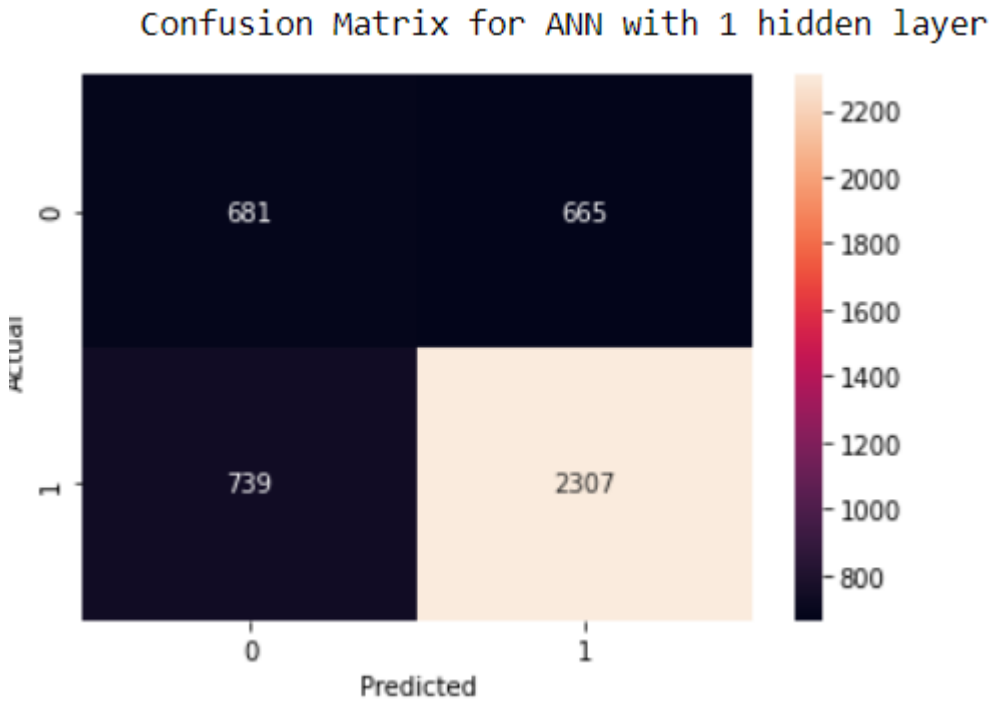

Fig.2 Confusion Matrix for ANN with 1 hidden layer

In Figure.2 it is observed that ANN with 1 hidden layer method predicted true positive cases 2307 times out of 4392 cases and true negative cases 681 times.

This suggests 68% of the algorithm is classifying correctly. Even though theoretically any algorithm can reach 100% accuracy, our achievement of 68% is good enough. Case being that if this algorithm reached maximum accuracy Group-1 would probably be very rich next year because of this achievement.

5- Statistical significance analysis between your best performing model and its closest competitor

Best Model: ANN with 1 hidden layer, Closest Competitor: Naïve-Bayes

Accuracy:

The P-value is = 0.002
The t-statistics is = 5.914
Since $p<0.05$, We can reject the null-hypothesis in terms of accuracy that both models perform equally well on this dataset. We may conclude that the two algorithms are significantly different.

F1-Macro:

The P-value is = 0.005
The t-statistics is = 4.883
Since $p<0.05$, We can reject the null-hypothesis in terms of f1_macro that both models perform equally well on this dataset. We may conclude that the two algorithms are significantly different.

F1-Micro:

The P-value is = 0.059
The t-statistics is = 2.428
Since $p>0.05$, we cannot reject the null hypothesis in terms of f1_micro may conclude that the performance of the two algorithms is not significantly different.

AUC:

The P-value is = 0.398
The t-statistics is = -0.923
Since $p>0.05$, we cannot reject the null hypothesis in terms of AUC and may conclude that the performance of the two algorithms is not significantly different.