## Saranpat Prasertthum (655667271)

IE517 ML in Fin Lab

Module 7 Homework (Random Forest)

## Out[1]:

Click here to toggle on/off the raw code.

#### **Load Data**

### Out[3]:

|                     | LIMIT_BAL | SEX | EDUCATION | MARRIAGE | AGE | PAY_0 | PAY_2 | PAY_3 | PAY_4 | PAY_5 |  |
|---------------------|-----------|-----|-----------|----------|-----|-------|-------|-------|-------|-------|--|
| ID                  |           |     |           |          |     |       |       |       |       |       |  |
| 1                   | 20000     | 2   | 2         | 1        | 24  | 2     | 2     | -1    | -1    | -2    |  |
| 2                   | 120000    | 2   | 2         | 2        | 26  | -1    | 2     | 0     | 0     | 0     |  |
| 3                   | 90000     | 2   | 2         | 2        | 34  | 0     | 0     | 0     | 0     | 0     |  |
| 4                   | 50000     | 2   | 2         | 1        | 37  | 0     | 0     | 0     | 0     | 0     |  |
| 5                   | 50000     | 1   | 2         | 1        | 57  | -1    | 0     | -1    | 0     | 0     |  |
| 5 rows × 24 columns |           |     |           |          |     |       |       |       |       |       |  |

#### Part 1: Random forest estimators

# Out[5]:

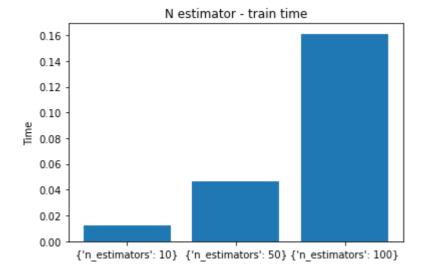
## Out[6]:

|   | Params                | Train Score | Test Score | Time     |
|---|-----------------------|-------------|------------|----------|
| 0 | {'n_estimators': 10}  | 0.979881    | 0.807800   | 0.012062 |
| 1 | {'n_estimators': 50}  | 0.998841    | 0.814600   | 0.046465 |
| 2 | {'n_estimators': 100} | 0.999344    | 0.815633   | 0.160956 |

## Part 2: Random forest feature importance

```
1) PAY_0
                 0.0986534998307748
         0.0675981322288534
2) AGE
3) BILL_AMT1
                 0.060794968463012564
4) LIMIT BAL
                 0.05938585573161959
5) BILL_AMT2
                 0.05451331240022285
6) BILL AMT3
                 0.05185449147076197
7) PAY_AMT1
                 0.05083125934847679
8) BILL_AMT6
                 0.05065825548621419
9) BILL_AMT5
                 0.050542074247345664
10) BILL AMT4
                 0.04989948122514444
11) PAY_AMT2
                 0.04742839312450553
12) PAY_AMT6
                 0.04648475150914067
13) PAY_AMT3
                 0.04623896699943686
14) PAY_AMT5
                 0.04348973831161921
15) PAY_AMT4
                 0.042896326918097034
16) PAY_2
                 0.03944462464112495
17) PAY 3
                 0.029032476832814774
18) PAY_5
                 0.02402127551409982
19) PAY 4
                 0.022996783190280524
20) EDUCATION
                 0.020348760322455986
21) PAY_6
                 0.016465360500640916
22) MARRIAGE
                 0.014096163854901509
23) SEX
                 0.012325047848455981
```

**Part 3: Conclusions** 



a) What is the relationship between n\_estimators, in-sample CV accuracy and computation time?

**ANS** If N estimator increase it will also increase computational time.

#### Out[10]:

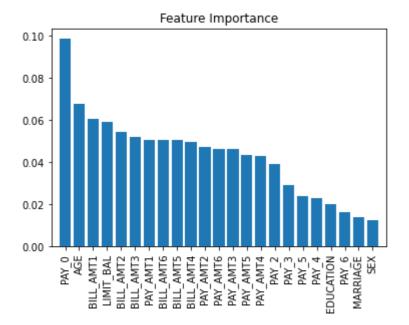
{'n\_estimators': 100}

b) What is the optimal number of estimators for your forest?

ANS The best estimators is 50

#### Out[11]:

(-1.0, 23.0)



### c) Which features contribute the most importance in your model according to scikit-learn function?

ANS The most importance feature in my model is "PAY 0"

# d) What is feature importance and how is it calculated? (If you are not sure, refer to the Scikit-Learn.org documentation.)

**ANS** Feature importance in scikit-learn Random Forest models is calculated based on the decrease in node impurity caused by a feature, and the probability of reaching the node that uses the feature. This probability is determined by dividing the number of samples that reach the node by the total number of samples. The resulting values are then weighted and used to rank the importance of the features. Essentially, the higher the feature importance value, the more significant the feature is in predicting the target variable.

| Impurity   | Task           | Formula   | Description  |
|--|----------------|---|--|
| Gini impurity  | Classification | $\sum\nolimits_{i=1}^{c} f_i (1 - f_i)$           | $f_i$ is the frequency of label $i$ at a node and $C$ is the number of unique labels.  |
| Entropy  | Classification | $\sum\nolimits_{i=1}^{C} -f_i {\rm log}(f_i)$     | $f_i$ is the frequency of label $i$ at a node and $C$ is the number of unique labels.  |
| Variance<br>/ Mean Square Error<br>(MSE)                       | Regression     | $\frac{1}{N}\sum_{i=1}^{N}(y_i-\mu)^2$            | $y_i$ is label for an instance, $N$ is the number of instances and $\mu$ is the mean given by $\frac{1}{N}\sum_{i=1}^{N}y_i$ |
| Variance / Mean Absolute<br>Error (MAE)<br>(Scikit-learn only) | Regression     | $\frac{1}{N} \sum\nolimits_{i=1}^{N}  y_i - \mu $ | $y_i$ is label for an instance, $N$ is the number of instances and $\mu$ is the mean given by $\frac{1}{N}\sum_{i=1}^{N}y_i$ |

Credit: <a href="https://towardsdatascience.com/the-mathematics-of-decision-trees-random-forest-and-feature-importance-in-scikit-learn-and-spark-">https://towardsdatascience.com/the-mathematics-of-decision-trees-random-forest-and-feature-importance-in-scikit-learn-and-spark-</a>

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(https://towardsdatascience.com/the-mathematics-of-decision-trees-random-forest-and-feature-importance-in-scikit-learn-and-spark-

f2861df67e3#:~:text=Feature%20importance%20is%20calculated%20as,the%20more%20important%20the%2

## Part 4: Appendix

4

 $\hbox{My name is Saranpat Prasertthum} \\$ 

My NetID is: 655667271

I hereby certify that I have read the University policy on Academic Integr

ity and that I am not in violation.