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IE517 MLF F21

Module 6 Homework (Cross validation)

Using the ccdefault dataset, with 90% for training and 10% for test (stratified sampling) and the decision tree model that you did in Module 2:

Part 1: Random test train splits

Run in-sample and out-of-sample accuracy scores for 10 different samples by changing random_state from 1 to 10 in sequence.

Display the individual scores, then calculate the mean and standard deviation on the set of scores. Report in a table format.

Part 2: Cross validation

Now rerun your model using cross_val_scores with stratified k-fold CV (k=10).

Report the individual fold accuracy scores, the mean CV score and the standard deviation of the fold scores. Now run the out-of-sample accuracy score. Report in a table format.

Part 3: Conclusions

Write a short paragraph summarizing your findings. Which method of measuring accuracy provides the best estimate of how a model will do against unseen data? Which one is more efficient to run?

The out-of-sample accuracy score provides the best estimate of how a model will do against unseen data. Since the data used to find this accuracy score is unseen by the model since it is not in the training sample, it provides the best estimate. Method-wise, the K-fold method does the best by making sure sampling is done without-replacement. K-fold is also the most efficient method of doing this, as opposed to writing and running multiple for-loops, the K-fold does this more effectively under the hood.

Part 4: Appendix

Link to github repo: https://github.com/eemayes2/IE517 F21 HW6

IE517_HWK6

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```
[1]: #Import libraries needed
    import numpy as np
    import pandas as pd
    import seaborn as sns
    import matplotlib.pyplot as plt
    import warnings
    warnings.filterwarnings('ignore')
    #Check if any null values we need to change
    def num_missing(x):
        return sum(x.isnull())
[5]: from google.colab import drive
    drive.mount('/content/gdrive')
```

Mounted at /content/gdrive

```
[6]: #Read in Data
    df = pd.read_csv('gdrive/MyDrive/Colab Notebooks/ccdefault.csv', header=0)
    df.head()
```

[6]:	ID	LIMIT_BAL	SEX	EDUCATION	 PAY_AMT4	PAY_AMT5	PAY_AMT6	DEFAULT
0	1	20000	2	2	 0	0	0	1
1	2	120000	2	2	 1000	0	2000	1
2	3	90000	2	2	 1000	1000	5000	0
3	4	50000	2	2	 1100	1069	1000	0
4	5	50000	1	2	 9000	689	679	0

[5 rows x 25 columns]

[7]: df.describe()

[7]:		ID	LIMIT_BAL	 PAY_AMT6	DEFAULT
	count	30000.000000	30000.000000	 30000.000000	30000.000000
	mean	15000.500000	167484.322667	 5215.502567	0.221200
	std	8660.398374	129747.661567	 17777.465775	0.415062
	min	1.000000	10000.000000	 0.000000	0.000000
	25%	7500.750000	50000.000000	 117.750000	0.000000
	50%	15000.500000	140000.000000	 1500.000000	0.000000

```
75% 22500.250000 240000.000000 ... 4000.000000 0.0000000 max 30000.000000 1000000.000000 ... 528666.000000 1.000000
```

0.0.1 Try different random states in test-train-split

```
[13]: #Split into training-test sets
     from sklearn.tree import DecisionTreeClassifier
     from sklearn.model_selection import train_test_split
     x = df.drop(columns = ['DEFAULT'])
     y = df['DEFAULT']
     out_of = []
     in_of = []
     for i in range(1,11):
       X_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.1,_
      →random_state=i, stratify = y)
       tree = DecisionTreeClassifier(criterion = 'gini', max depth = 4, random state_
      \rightarrow = 1)
       tree.fit(X train, y train)
       y_pred = tree.predict(X_test)
       acc_out = accuracy_score(y_test, y_pred)
       out_of.append(acc_out)
       y_pred_train = tree.predict(X_train)
       acc_in = accuracy_score(y_train, y_pred_train)
       in_of.append(acc_in)
       print("Seed Number:" + str(i) + "\n\tOut of Sample Accuracy: " + str(acc_out))
       print("\tIn Sample Accuracy: " + str(acc_in))
    Seed Number: 1
            Out of Sample Accuracy: 0.82833333333333334
            In Sample Accuracy: 0.82255555555556
    Seed Number: 2
            Out of Sample Accuracy: 0.824
            In Sample Accuracy: 0.8225925925925925
    Seed Number: 3
            Out of Sample Accuracy: 0.81733333333333334
            In Sample Accuracy: 0.8241111111111111
    Seed Number: 4
            Out of Sample Accuracy: 0.819666666666667
            In Sample Accuracy: 0.8237037037037037
    Seed Number:5
            Out of Sample Accuracy: 0.818
            In Sample Accuracy: 0.8231481481481482
    Seed Number:6
            Out of Sample Accuracy: 0.819
            In Sample Accuracy: 0.823925925925926
```

```
Seed Number:7
            Out of Sample Accuracy: 0.82533333333333334
            In Sample Accuracy: 0.8232592592592592
    Seed Number:8
            Out of Sample Accuracy: 0.816666666666667
            In Sample Accuracy: 0.8238148148148148
    Seed Number:9
            Out of Sample Accuracy: 0.81733333333333334
            In Sample Accuracy: 0.823925925925926
    Seed Number: 10
            Out of Sample Accuracy: 0.82
            In Sample Accuracy: 0.8236296296296
[20]: print("Out of Sample: \n\tMean: " + str(np.mean(out_of)) + "\n\tStand. Dev: " + u

→str(np.std(out_of)))
     print("In Sample: \n\tMean: " + str(np.mean(in_of)) + "\n\tStand. Dev: " +__

→str(np.std(in_of)))
    Out of Sample:
            Mean: 0.820566666666666
            Stand. Dev: 0.0037566237797019685
    In Sample:
            Mean: 0.8234666666666666
            Stand. Dev: 0.0005272284340525735
    0.0.2 Cross Validation with K-fold CV
[40]: from sklearn.model_selection import StratifiedKFold, cross_val_score
     x = df.drop(columns = ['DEFAULT'])
     y = df['DEFAULT']
     cv = StratifiedKFold(n_splits=10, random_state=1, shuffle=True)
     scores = cross_val_score(tree, x, y, scoring='accuracy', cv=cv)
     scores
                                  , 0.82366667, 0.82466667, 0.82033333,
[40]: array([0.819
                      , 0.827
            0.817
                      , 0.81233333, 0.81833333, 0.81366667, 0.82566667])
[35]: print("Cross Val Scores: \n\tMean: " + str(np.mean(scores)) + "\n\tStand. Dev:
      →" + str(np.std(scores)))
    Cross Val Scores:
            Mean: 0.82073333333333333
            Stand. Dev: 0.006848844184726854
[41]: #Out of sample accuracy
     kf = StratifiedKFold(n_splits=10, random_state = 1, shuffle = True)
```

```
kf.get_n_splits(x,y)
     i = 1
     out of = []
     for train_index, test_index in kf.split(x):
      X_train, X_test = x.iloc[list(train_index)], x.iloc[list(test_index)]
      y_train, y_test = y.iloc[list(train_index)], y.iloc[list(test_index)]
      tree.fit(X_train, y_train)
      y_pred = tree.predict(X_test)
       acc_out = accuracy_score(y_test, y_pred)
       out_of.append(acc_out)
      y_pred_train = tree.predict(X_train)
       print("Fold Index:" + str(i) + "\n\t0ut of Sample Accuracy: " + str(acc_out))
      print("\tIn Sample Accuracy: " + str(acc_in))
       i += 1
            TypeError
                                                       Traceback (most recent call_
     →last)
            <ipython-input-41-63251d908dc7> in <module>()
              5 \text{ out of } = []
        ---> 7 for train_index, test_index in kf.split(x):
                  X_train, X_test = x.iloc[list(train_index)], x.
     →iloc[list(test_index)]
                  y_train, y_test = y.iloc[list(train_index)], y.
     →iloc[list(test_index)]
            TypeError: split() missing 1 required positional argument: 'y'
[38]: print("Out of Sample: \n\tMean: " + str(np.mean(out_of)) + "\n\tStand. Dev: " +__
      →str(np.std(out_of)))
    Out of Sample:
            Mean: 0.8207333333333333
            Stand. Dev: 0.006848844184726854
[39]: | wget -nc https://raw.githubusercontent.com/brpy/colab-pdf/master/colab_pdf.py
     from colab_pdf import colab_pdf
     colab_pdf('IE517_HWK6.ipynb')
```

```
--2021-10-01 16:48:11-- https://raw.githubusercontent.com/brpy/colab-
pdf/master/colab_pdf.py
Resolving raw.githubusercontent.com (raw.githubusercontent.com)...
185.199.110.133, 185.199.109.133, 185.199.108.133, ...
Connecting to raw.githubusercontent.com
(raw.githubusercontent.com) | 185.199.110.133 | :443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 1864 (1.8K) [text/plain]
Saving to: colab_pdf.py
                  colab_pdf.py
                                                                  in Os
2021-10-01 16:48:11 (27.8 MB/s) - colab_pdf.py saved [1864/1864]
Mounted at /content/drive/
WARNING: apt does not have a stable CLI interface. Use with caution in scripts.
WARNING: apt does not have a stable CLI interface. Use with caution in scripts.
Extracting templates from packages: 100%
[NbConvertApp] Converting notebook /content/drive/MyDrive/Colab
Notebooks/IE517_HWK6.ipynb to pdf
[NbConvertApp] Writing 38290 bytes to ./notebook.tex
[NbConvertApp] Building PDF
[NbConvertApp] Running xelatex 3 times: [u'xelatex', u'./notebook.tex',
'-quiet']
[NbConvertApp] Running bibtex 1 time: [u'bibtex', u'./notebook']
[NbConvertApp] WARNING | bibtex had problems, most likely because there were no
citations
[NbConvertApp] PDF successfully created
[NbConvertApp] Writing 35185 bytes to /content/drive/My Drive/IE517_HWK6.pdf
<IPython.core.display.Javascript object>
<IPython.core.display.Javascript object>
```

[39]: 'File ready to be Downloaded and Saved to Drive'