Quiz: Decision Trees with Python

Due Feb 17 at 3:30pm Points 11 Questions 4 Time Limit None

Instructions

This exercise is a continuation of the quiz "OLS with Python". Make sure that you complete that quiz correctly first. Then follow the instructions in the document

(https://app.box.com/embed_widget/s/vk8159bxwdlflc30sr5kxyumw9uwvogi?

view=list&sort=name&direction=ASC&theme=dark)

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(https://app.box.com/embed_widget/s/if89p4wxoa1okap00l9cial8n60wtkcb?

<u>view=list&sort=name&direction=ASC&theme=dark</u>), and finally answer the questions below.

This is an individual exercise, and has no time limit.

Attempt History

	tempt	Time	Score
LATEST Att	empt 1	4 minutes	9 out of 11 *

^{*} Some questions not yet graded

Score for this quiz: **9** out of 11 * Submitted Feb 8 at 11:11am This attempt took 4 minutes.

	Question 1	3 / 3 pts
	What is the mean squared error of the decision tree model?	
	0.7718440195526606	
	1.0718440195526606	
Correct!	0.9718440195526606	

0.8718440195526606

	Question 2 3 / 3 pts)		
	Take a look at the graph you plotted using the tree.plot_tree function. How many levels are there (including the first main level)?			
Correct!	22	*		
	O 26			
	O 20			
	O 24			

	Question 3	3 / 3 pts	
	What is the mean squared error of the random forest algorithm?		
	0.9514334662236628		
	0.9614334662236628		
	0.9714334662236628		
Correct!	0.9814334662236628		

Question 4 Not yet graded / 2 pts

```
Post the code below.
Your Answer:
# # OLS
#
# In[102]:
import pandas as pd
import numpy as np
from sklearn.datasets import make classification
from sklearn.linear_model import LinearRegression
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import roc_curve
from sklearn.metrics import roc_auc_score
from sklearn.metrics import mean_squared_error
import math
from matplotlib import pyplot
import os
import statsmodels.api as sm
# In[103]:
df=pd.read csv('Stock Return Data.csv')
df.columns
# In[104]:
df.head()
# In[105]:
df['Mkt rf lead'] = df['Mkt rf'].shift(-1)
df['Mkt_rf_lead']
```

```
# In[106]:
#df['Mkt_rf_lead'].dropna(how='any', inplace=True)
df.dropna(how='any', inplace=True)
df
# In[107]:
X = np.asarray(df.drop(['Date','Mkt_rf','Mkt_rf_lead'],axis=1))
y = np.asarray(df['Mkt_rf_lead'])
# Create linear regression object
reg = linear_model.LinearRegression()
# Train the model using the training sets
reg.fit(X, y)
# Make predictions using the testing set
pred x = reg.predict(X)
# The mean absolute error
MSE = mean squared error(y, pred x, squared=True)
RMSE = mean_squared_error(y, pred_x, squared=False)
print(MSE)
print(RMSE)
# In[108]:
dif=y-y_pred
dif_sq=dif*dif
n=len(y)
mse=np.sum(dif sq)/n
mse2=mean_squared_error(y, y_pred)
rmse=math.sqrt(mse)
print("mse:", mse)
print("mse2:", mse2)
print("rmse:", rmse)
```

```
# In[109]:
df.describe()
# In[110]:
X = sm.add\_constant(X)
fit = sm.OLS(y, X).fit()
y_pred=np.asarray(fit.predict())
# Print out the statistics
fit.summary()
# In[111]:
dif=y-y_pred
dif_sq=dif*dif
n=len(y)
mse=np.sum(dif_sq)/n
mse2=mean_squared_error(y, y_pred)
rmse=math.sqrt(mse)
print(mse)
print(mse2)
print(rmse)
## Decision Trees
#
# In[137]:
from sklearn import tree
from sklearn.datasets import load_iris
from sklearn.tree import DecisionTreeRegressor
```

```
from matplotlib import pyplot
from sklearn.ensemble import RandomForestRegressor
# In[138]:
Tree = DecisionTreeRegressor(random_state=42,
min_samples_leaf=100)
Tree.fit(X,y)
pred x = Tree.predict(X)
MSE = mean_squared_error(y, pred_x, squared=True)
RMSE = mean_squared_error(y, pred_x, squared=False)
print(MSE)
print(RMSE)
# In[139]:
tree.plot_tree(Tree)
# In[142]:
from sklearn.ensemble import RandomForestRegressor
# In[145]:
forest = RandomForestRegressor(random_state=42, n_estimators=100,
min_samples_leaf=100, max_features='sqrt')
forest.fit(X,y.ravel())
pred x = forest.predict(X)
MSE = mean squared error(y, pred x, squared=True)
RMSE = mean_squared_error(y, pred_x, squared=False)
print(MSE)
print(RMSE)
```

Quiz Score: 9 out of 11