Random Forest and Linear Regression Assignment

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RANDOM FOREST

CODE

```
import <u>numpy</u> as <u>nm</u>
import <u>matplotlib.pyplot</u> as <u>mtp</u>
import <u>pandas</u> as <u>pd</u>
data set= <u>pd</u>.read csv('areas.csv')
#Extracting Independent and dependent Variable
x= data set.iloc[:, [2,3]].values
y= data set.iloc[:, 4].values
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test= train_test_split(x, y, test_size=
0.25, random state=0)
#feature Scaling
from sklearn.preprocessing import StandardScaler
st x= StandardScaler()
x train= st x.fit transform(x train)
x test= st x.transform(x test)
#Fitting Decision Tree classifier to the training set
from <u>sklearn</u>.<u>ensemble</u> import <u>RandomForestClassifier</u>
classifier= RandomForestClassifier(n estimators= 10,
criterion="entropy")
classifier.fit(x train, y train)
y pred= classifier.predict(x test)
#Creating the Confusion matrix
from <u>sklearn.metrics</u> import confusion matrix
cm= confusion_matrix(y_test, y_pred)
from matplotlib.colors import ListedColormap
x_set, y_set = x_train, y_train
```

```
x1, x2 = nm.meshgrid(nm.arange(start = x_set[:, 0].min() - 1, <math>stop = nm
x set[:, 0].max() + 1, step = 0.01),
\underline{nm}.arange(start = x_set[:, 1].min() - 1, <math>stop = x_set[:, 1].max() + 1,
step = 0.01)
mtp.contourf(x1, x2, classifier.predict(nm.array([x1.ravel(),
x2.ravel()]).T).reshape(x1.shape),
alpha = 0.75, cmap = ListedColormap(('purple','green')))
mtp.xlim(x1.min(), x1.max())
mtp.ylim(x2.min(), x2.max())
 for i, j in enumerate(nm.unique(y_set)):
         \underline{mtp}.scatter(x set[y set == j, 0], x set[y set == j, 1],
                   c = ListedColormap(('purple', 'green'))(i), label = j)
mtp.title('Random Forest Algorithm (Training set)')
mtp.xlabel('Age')
mtp.ylabel('Estimated Salary')
mtp.legend()
mtp.show()
from <u>matplotlib</u>.<u>colors</u> import <u>ListedColormap</u>
x set, y set = x test, y test
x1, x2 = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, stop = nm.meshgrid(nm.arange(start = x set[:, 0].min() - 1, st
x set[:, 0].max() + 1, step = 0.01),
mtp.contourf(x1, x2, classifier.predict(nm.array([x1.ravel(),
x2.ravel()]).T).reshape(x1.shape),
alpha = 0.75, cmap = ListedColormap(('purple','green')))
\underline{\mathsf{mtp}}.\mathtt{xlim}(\mathtt{x1.min}(), \mathtt{x1.max}())
mtp.ylim(x2.min(), x2.max())
for i, j in enumerate(nm.unique(y set)):
         mtp.scatter(x_set[y_set == j, 0], x_set[y_set == j, 1],
                   c = ListedColormap(('purple', 'green'))(i), label = j)
mtp.title('Random Forest Algorithm(Test set)')
mtp.xlabel('Age')
mtp.ylabel('Estimated Salary')
mtp.legend()
mtp.show()
```

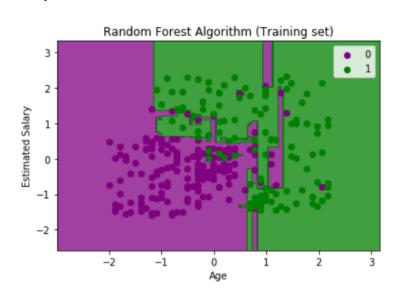
DATA FROM CSV FILE

Index	User ID	Gender	Age	EstimatedSalary	Purchased
0	15624510	Male	19	19000	0
1	15810944	Male	35	20000	0
2	15668575	Female	26	43000	0
3	15603246	Female	27	57000	0
4	15804002	Male	19	76000	0
5	15728773	Male	27	58000	0
6	15598044	Female	27	84000	0
7	15694829	Female	32	150000	1
8	15600575	Male	25	33000	0
9	15727311	Female	35	65000	0
10	15570769	Female	26	80000	0
11	15606274	Female	26	52000	0
12	15746139	Male	20	86000	0
13	15704987	Male	32	18000	0

OUTPUT

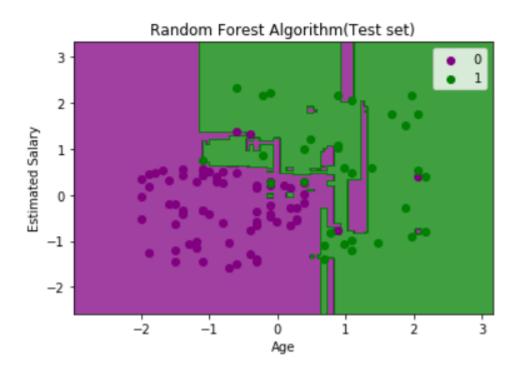
FOR TRAINING

Output:



FOR TEST

Output:



LINEAR REGRESSION

CODE

Code source: Jaques Grobler

License: BSD 3 clause

import matplotlib.pyplot as plt import numpy as np from sklearn import datasets, linear_model from sklearn.metrics import mean_squared_error, r2_score

Load the diabetes dataset diabetes_X, diabetes_y = datasets.load_diabetes(return_X_y=True)

Use only one feature diabetes_X = diabetes_X[:, np.newaxis, 2]

Split the data into training/testing sets diabetes_X_train = diabetes_X[:-20]

```
diabetes_X_test = diabetes_X[-20:]
# Split the targets into training/testing sets
diabetes_y_train = diabetes_y[:-20]
diabetes y test = diabetes y[-20:]
# Create linear regression object
regr = linear model.LinearRegression()
# Train the model using the training sets
regr.fit(diabetes_X_train, diabetes_y_train)
# Make predictions using the testing set
diabetes_y_pred = regr.predict(diabetes_X_test)
# The coefficients
print("Coefficients: \n", regr.coef_)
# The mean squared error
print("Mean squared error: %.2f" % mean_squared_error(diabetes_y_test,
diabetes_y_pred))
# The coefficient of determination: 1 is perfect prediction
print("Coefficient of determination: %.2f" % r2_score(diabetes_y_test, diabetes_y_pred))
# Plot outputs
plt.scatter(diabetes_X_test, diabetes_y_test, color="black")
plt.plot(diabetes_X_test, diabetes_y_pred, color="blue", linewidth=3)
plt.xticks(())
plt.yticks(())
plt.show()
```

OUTPUT

Coefficients: [938.23786125] Mean squared error: 2548.07 Coefficient of determination: 0.47

