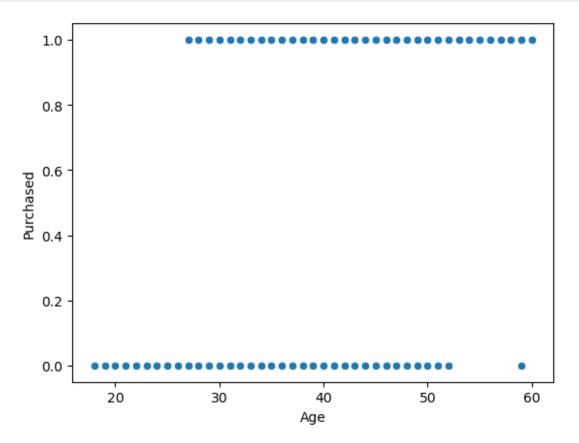
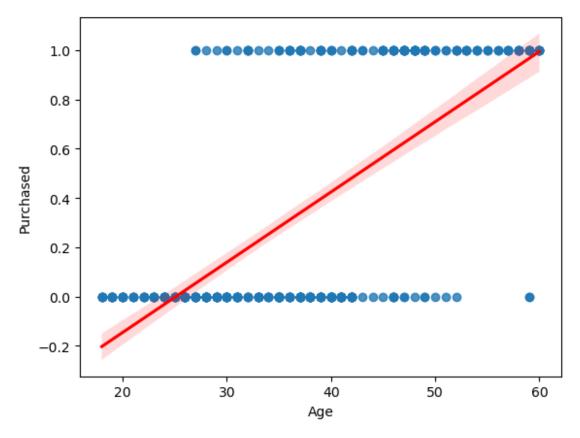
```
import pandas as pd
import seaborn as sns
from sklearn.model selection import train test split
from sklearn.linear model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier as dtc
from sklearn.ensemble import RandomForestClassifier as rfc
from sklearn.metrics import r2 score, classification report as cr,
confusion matrix as cm
import warnings
warnings.filterwarnings("ignore")
import matplotlib.pyplot as plt
import numpy as np
z = pd.read csv(r"C:\Users\skj h\OneDrive\Desktop\dataset\
Social Network Ads.csv")
     Age EstimatedSalary
                            Purchased
      19
0
                    19000
1
      35
                                    0
                    20000
2
      26
                    43000
                                    0
3
      27
                                    0
                    57000
4
      19
                    76000
                                    0
     . . .
395
      46
                    41000
                                    1
396
      51
                                    1
                    23000
397
      50
                                    1
                    20000
                    33000
398
      36
                                    0
399
      49
                    36000
                                    1
[400 rows x 3 columns]
z.isnull().sum()
Age
                   0
EstimatedSalary
                   0
Purchased
                   0
dtype: int64
z.shape
(400, 3)
z.size
1200
z.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 3 columns):
```

```
#
     Column
                       Non-Null Count
                                       Dtype
- - -
 0
     Age
                       400 non-null
                                       int64
     EstimatedSalary
                       400 non-null
 1
                                       int64
                       400 non-null
 2
     Purchased
                                       int64
dtypes: int64(3)
memory usage: 9.5 KB
z.dtypes
Age
                    int64
EstimatedSalary
                   int64
Purchased
                   int64
dtype: object
z.corr()
                            EstimatedSalary
                                             Purchased
                       Age
                 1.000000
                                   0.155238
                                              0.622454
Age
EstimatedSalary
                 0.155238
                                   1.000000
                                              0.362083
Purchased
                 0.622454
                                   0.362083
                                              1.000000
sns.scatterplot(x = z["Age"], y = z["Purchased"], data = z)
<Axes: xlabel='Age', ylabel='Purchased'>
```



```
sns.regplot(x = z["Age"], y = z["Purchased"], data = z, line_kws =
{"color": "red"})
<Axes: xlabel='Age', ylabel='Purchased'>
```



```
x = z[["Age", "Purchased"]]
X = x
Y = x["Purchased"]
x_train, x_test, y_train, y_test = train_test_split(X, Y, train_size = 0.7, test_size = 0.3, random_state = 100)
x_train = x_train.drop(["Purchased"], axis = 1)
x_test = x_test.drop(["Purchased"], axis = 1)
y_train = np.array(y_train).reshape(-1, 1)
y_test = np.array(y_test).reshape(-1, 1)
n = LogisticRegression()
n.fit(x_train, y_train)
LogisticRegression()
y_predict_train = n.predict(x_train)
```

```
y_predict_test = n.predict(x_test)
cr train logistic = cr(y true = y train, y pred = y predict train)
cr_test_logistic = cr(y_true = y_test, y_pred = y_predict_test)
a = dtc()
a.fit(x train, y train)
DecisionTreeClassifier()
v predict train1 = a.predict(x train)
cr train decision = cr(y true = y train, y pred = y predict train)
y predict test1 = a.predict(x test)
cr test decision = cr(y true = y test, y pred = y predict test1)
v5 = rfc()
y5.fit(x train, y train)
RandomForestClassifier()
y_predict_train2 = y5.predict(x_train)
cr train random = cr(y true = y train, y pred = y predict train2)
y predict test2 = y5.predict(x test)
cr_test_random = cr(y_true = y_test, y_pred = y_predict_test2)
def upper name(s):
    return(s.upper())
print(upper name("Logistic Regression training"),cr train logistic)
print(upper name("Decision Tree classifier
training"),cr train decision)
print(upper name("Random forest classifier training"),cr train random)
print(upper name("Logistic Regression testing"),cr test logistic)
print(upper name("Decision Tree classifier testing"),cr test decision)
print(upper name("Random forest classifier testing"),cr test random)
LOGISTIC REGRESSION TRAINING
                                           precision recall f1-
score
        support
                   0.84
                             0.92
                                       0.88
                                                  182
           0
                   0.81
                             0.67
                                       0.74
                                                   98
           1
                                       0.83
                                                  280
    accuracy
                   0.83
                             0.80
                                       0.81
                                                  280
   macro avg
weighted avg
                   0.83
                             0.83
                                       0.83
                                                  280
DECISION TREE CLASSIFIER TRAINING
                                                              recall
                                                 precision
f1-score
         support
```

0 0.84 0.92 0.88 182 1 0.81 0.67 0.74 98	
accuracy 0.83 280 macro avg 0.83 0.80 0.81 280 weighted avg 0.83 0.83 0.83 280	
RANDOM FOREST CLASSIFIER TRAINING precision refl-score support	recall
0       0.88       0.88       0.88       182         1       0.78       0.77       0.77       98	
accuracy 0.84 280 macro avg 0.83 0.82 0.83 280 weighted avg 0.84 0.84 0.84 280	
LOGISTIC REGRESSION TESTING precision recall score support	f1-
0       0.84       0.96       0.89       75         1       0.91       0.69       0.78       45	
accuracy 0.86 120 macro avg 0.87 0.82 0.84 120 weighted avg 0.87 0.86 0.85 120	
DECISION TREE CLASSIFIER TESTING precision ref fl-score support	ecall
0       0.79       0.87       0.83       75         1       0.74       0.62       0.67       45	
accuracy 0.78 120 macro avg 0.76 0.74 0.75 120 weighted avg 0.77 0.78 0.77 120	
RANDOM FOREST CLASSIFIER TESTING precision ref f1-score support	ecall
0       0.83       0.84       0.83       75         1       0.73       0.71       0.72       45	
accuracy 0.79 120 macro avg 0.78 0.78 0.78 120 weighted avg 0.79 0.79 0.79 120	