```
In [1]:
                                                                                                H
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings("ignore")
                                                                                                H
In [2]:
social_data = pd.read_csv("Social_Network_Ads.csv")
In [3]:
                                                                                                H
social_data.head()
Out[3]:
        EstimatedSalary Purchased
   Age
                19000
                              0
0
    19
1
    35
                20000
                              0
2
    26
                43000
3
                57000
                              0
    27
    19
                76000
                              0
In [4]:
                                                                                                H
social_data.tail()
Out[4]:
     Age
          EstimatedSalary Purchased
395
                  41000
                                1
      46
396
      51
                  23000
                                1
397
      50
                  20000
398
      36
                  33000
                                0
399
                  36000
      49
                                1
In [5]:
                                                                                                H
social_data.shape
Out[5]:
```

(400, 3)

```
In [6]:
                                                                                         M
social_data.columns
Out[6]:
Index(['Age', 'EstimatedSalary', 'Purchased'], dtype='object')
                                                                                         H
In [7]:
social_data.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 3 columns):
                      Non-Null Count Dtype
 #
     Column
     _____
_ _ _
                       400 non-null
 0
     Age
                                       int64
 1
     EstimatedSalary 400 non-null
                                       int64
     Purchased
                      400 non-null
 2
                                       int64
dtypes: int64(3)
memory usage: 9.5 KB
                                                                                         H
In [8]:
social_data.describe()
Out[8]:
```

| | Age | EstimatedSalary | Purchased |
|-------|------------|-----------------|------------|
| count | 400.000000 | 400.000000 | 400.000000 |
| mean | 37.655000 | 69742.500000 | 0.357500 |
| std | 10.482877 | 34096.960282 | 0.479864 |
| min | 18.000000 | 15000.000000 | 0.000000 |
| 25% | 29.750000 | 43000.000000 | 0.000000 |
| 50% | 37.000000 | 70000.000000 | 0.000000 |
| 75% | 46.000000 | 88000.000000 | 1.000000 |
| max | 60.000000 | 150000.000000 | 1.000000 |

In [9]:

```
social_data.isnull().sum()
```

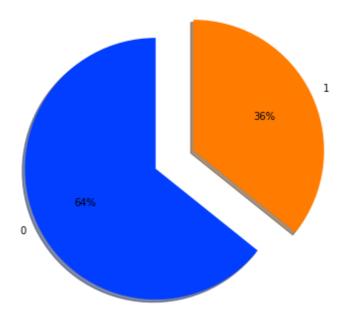
Out[9]:

Age 0
EstimatedSalary 0
Purchased 0

dtype: int64

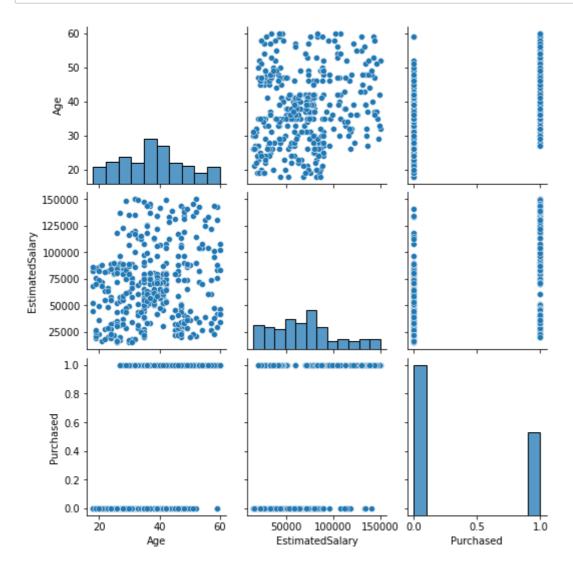
```
In [10]:
                                                                                           M
social_data.nunique()
Out[10]:
                     43
Age
EstimatedSalary
                    117
Purchased
                      2
dtype: int64
In [11]:
                                                                                           H
social_data.Purchased.unique()
Out[11]:
array([0, 1], dtype=int64)
In [12]:
                                                                                           H
social_data.Purchased.value_counts()
Out[12]:
     257
1
     143
Name: Purchased, dtype: int64
In [13]:
                                                                                           H
plt.figure(figsize=(15,6))
sns.countplot('Purchased', data = social_data, palette='hls')
plt.xticks(rotation = 90)
plt.show()
  250
  200
 150
 100
  50
```

In [14]:



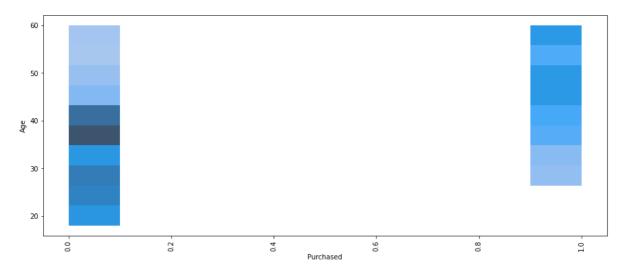
In [15]: ▶

```
sns.pairplot(social_data)
plt.show()
```



In [19]: ▶

```
plt.figure(figsize=(15,6))
sns.histplot(y = 'Age', x = 'Purchased', data = social_data)
plt.xticks(rotation = 90)
plt.show()
```

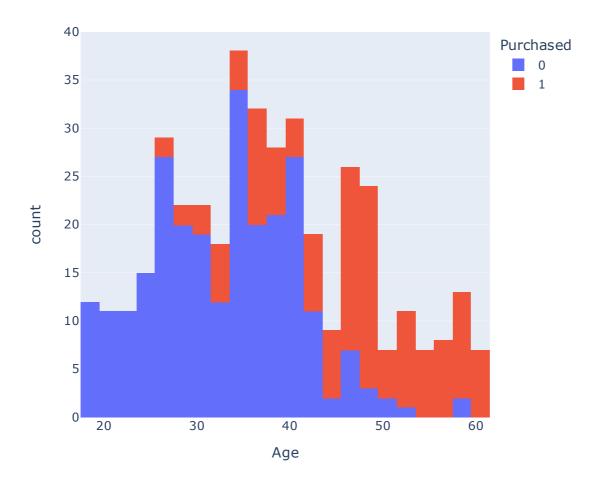


In [20]: ▶

import plotly.express as px

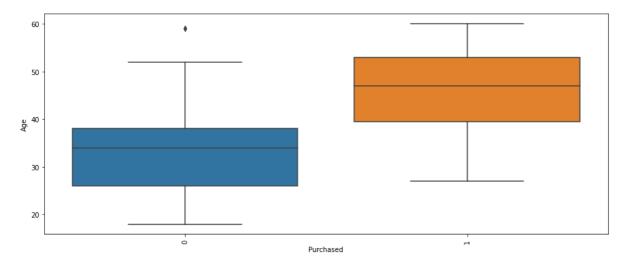
In [22]: ▶

```
fig1 = px.histogram(social_data, x = 'Age', color = 'Purchased')
fig1.show()
```



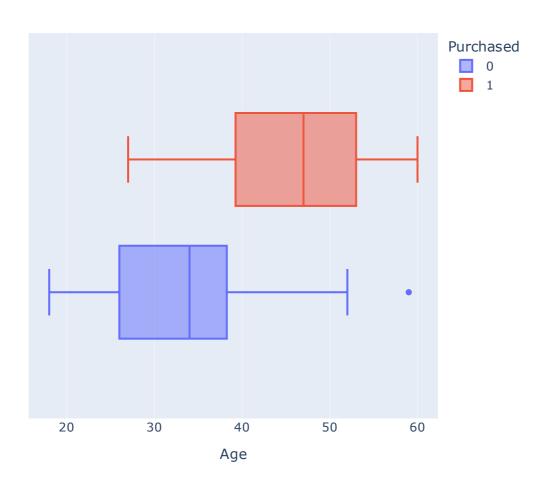
```
In [23]: ▶
```

```
plt.figure(figsize=(15,6))
sns.boxplot(y = 'Age', x = 'Purchased', data = social_data)
plt.xticks(rotation = 90)
plt.show()
```



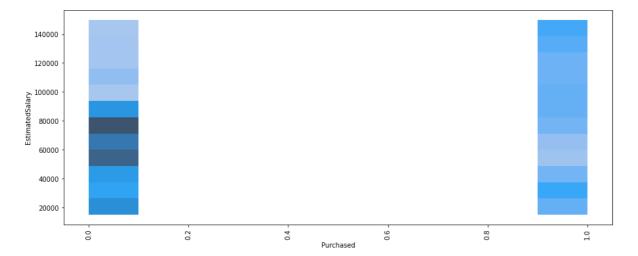
```
In [24]: ▶
```

```
fig2 = px.box(social_data, x = 'Age', color = 'Purchased')
fig2.show()
```



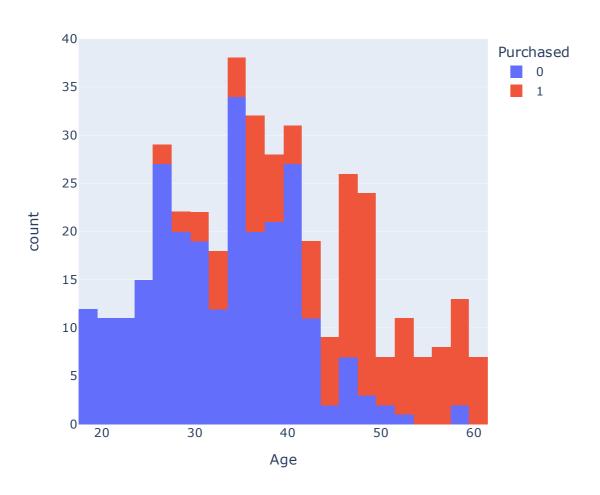
In [25]: ▶

```
plt.figure(figsize=(15,6))
sns.histplot(y = 'EstimatedSalary', x = 'Purchased', data = social_data)
plt.xticks(rotation = 90)
plt.show()
```



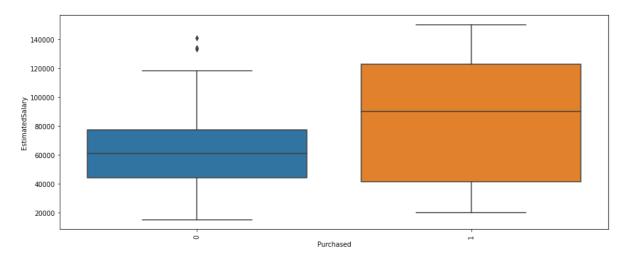
```
In [26]: ▶
```

```
fig3 = px.histogram(social_data, x = 'EstimatedSalary', color = 'Purchased')
fig1.show()
```



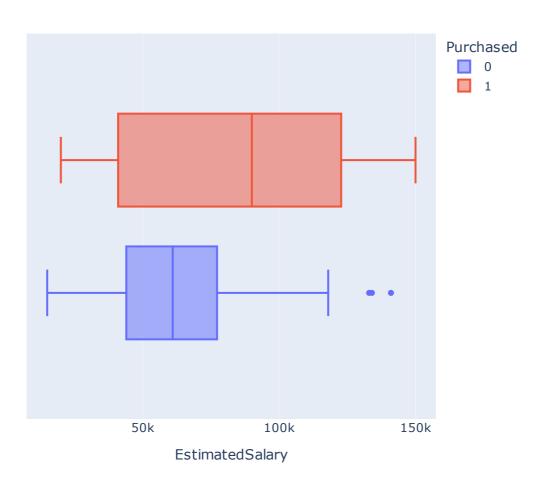
In [27]: ▶

```
plt.figure(figsize=(15,6))
sns.boxplot(y = 'EstimatedSalary', x = 'Purchased', data = social_data)
plt.xticks(rotation = 90)
plt.show()
```



```
In [28]:

fig4 = px.box(social_data, x = 'EstimatedSalary', color = 'Purchased')
fig4.show()
```



```
In [29]:

x = social_data.drop(['Purchased'], axis = 1)
y = social_data.Purchased

In [30]:

x.shape

Out[30]:
(400, 2)

In [31]:

y.shape

Out[31]:
(400,)
```

```
In [32]:
x_{train} = x.iloc[:301]
In [33]:
y_train = y.iloc[:301]
In [34]:
x_{test} = x.iloc[301:401]
In [35]:
y_test = y.iloc[301:401]
In [36]:
from sklearn.preprocessing import StandardScaler
scale = StandardScaler()
x_train = scale.fit_transform(x_train)
x_test = scale.fit_transform(x_test)
In [37]:
                                                                                         M
from sklearn.neighbors import KNeighborsClassifier
In [39]:
                                                                                         H
classifier = KNeighborsClassifier(n_neighbors = 5, metric = 'minkowski',
                                   p = 2
classifier.fit(x_train,y_train)
Out[39]:
KNeighborsClassifier()
In [40]:
                                                                                         M
y_pred = classifier.predict(x_test)
In [41]:
print("Training Accuracy :", classifier.score(x_train, y_train))
print("Testing Accuracy :", classifier.score(x_test, y_test))
```

Training Accuracy: 0.9235880398671097 Testing Accuracy: 0.696969696969697

```
In [42]:
from sklearn.metrics import confusion matrix, accuracy score
cm = confusion_matrix(y_test,y_pred)
print(cm)
accuracy_score(y_test,y_pred)
[[37 1]
 [29 32]]
Out[42]:
0.696969696969697
In [43]:
                                                                                                                                                                                                                  H
from sklearn.model_selection import cross_val_score
from sklearn.linear_model import LogisticRegression, SGDClassifier, Perceptron, RidgeClassifier
from sklearn.discriminant analysis import LinearDiscriminantAnalysis
from sklearn.naive_bayes import GaussianNB, BernoulliNB
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier, AdaBoostingClassifier, AdaBoostin
from sklearn.svm import SVC,NuSVC
from sklearn.linear_model import PassiveAggressiveClassifier
from sklearn.metrics import precision_score
In [45]:
                                                                                                                                                                                                                  H
models =[("LR", LogisticRegression()),("SVC", SVC()),
                      ('KNN', KNeighborsClassifier()),
                      ("DTC", DecisionTreeClassifier()),
                      ("GNB", GaussianNB()),("SGDC", SGDClassifier()),
                      ("Perc", Perceptron()),
                      ("Ridge", RidgeClassifier()),("NuSVC", NuSVC()),
                      ("BNB", BernoulliNB()),('RF',RandomForestClassifier()),
                      ('ADA',AdaBoostClassifier()),('XGB',GradientBoostingClassifier()),('PAC',Passi
pred = []
names = []
modelsprecision = []
In [46]:
                                                                                                                                                                                                                  Ы
for name, model in models:
         model.fit(x_train, y_train)
         prediction = model.predict(x_test)
          score = precision_score(y_test, prediction,average = 'macro')
         pred.append(score)
         names.append(name)
         modelsprecision.append((name,score))
```

modelsprecision.sort(key=lambda k:k[1],reverse=True)

In [47]: ▶

modelsprecision

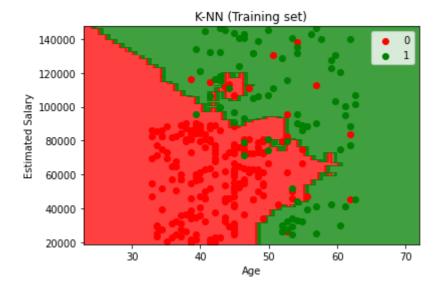
Out[47]:

```
[('SVC', 0.7848735832606801),
('KNN', 0.7651515151515151),
('DTC', 0.7567567567567568),
('NuSVC', 0.7470443349753695),
('GNB', 0.7427062374245472),
('XGB', 0.7384259259259258),
('LR', 0.7341938883034773),
('PAC', 0.7341938883034773),
('SGDC', 0.73),
('RF', 0.73),
('Ridge', 0.7216819221967964),
('ADA', 0.7091772151898734),
('BNB', 0.6723684210526316),
('Perc', 0.6296583850931676)]
```

In [50]: ▶

c argument looks like a single numeric RGB or RGBA sequence, which shoul d be avoided as value-mapping will have precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2D array with a single row if you intend to specify the same RGB or RGBA v alue for all points.

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In [52]: ▶

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