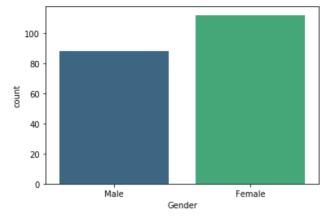
```
In [5]:
           import pandas as pd
           import numpy as np
           import matplotlib.pyplot as plt
           import seaborn as sns
           from plotly.offline import download plotly;s, init notebook mode, plot, iplo
           import cufflinks as cf
           import warnings
           warnings.filterwarnings("ignore")
In [10]: len(df)
Out[10]: 200
In [11]: df.head()
Out[11]:
              CustomerID Gender
                                                        Spending Score (1-100)
                                  Age
                                      Annual Income (k$)
            0
                            Male
                                   19
                                                     15
                       1
                                                                          39
            1
                       2
                                   21
                                                     15
                                                                          81
                            Male
            2
                       3
                          Female
                                   20
                                                     16
                                                                           6
            3
                       4
                          Female
                                   23
                                                     16
                                                                          77
                         Female
                                   31
                                                     17
                                                                          40
In [12]:
           df.isnull()
Out[12]:
                CustomerID Gender
                                         Annual Income (k$) Spending Score (1-100)
                                    Age
              0
                      False
                              False False
                                                     False
                                                                          False
              1
                      False
                              False
                                    False
                                                     False
                                                                          False
              2
                      False
                              False
                                    False
                                                     False
                                                                          False
                                                     False
              3
                      False
                              False
                                    False
                                                                          False
              4
                      False
                              False False
                                                     False
                                                                          False
            195
                      False
                              False
                                    False
                                                     False
                                                                          False
            196
                      False
                              False
                                    False
                                                     False
                                                                          False
            197
                      False
                              False
                                   False
                                                     False
                                                                          False
                                                     False
            198
                      False
                                                                          False
                              False False
            199
                      False
                              False False
                                                     False
                                                                          False
           200 rows × 5 columns
In [13]: | df.isnull().sum()
                                          0
Out[13]: CustomerID
           Gender
                                           0
                                          0
           Age
           Annual Income (k$)
                                          0
           Spending Score (1-100)
                                          0
```

dtype: int64

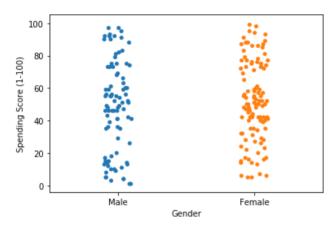
```
In [14]: import missingno as msno
          msno.matrix(df)
Out[14]: <matplotlib.axes._subplots.AxesSubplot at 0x7fd506161668>
In [15]: df['Gender'].unique()
Out[15]: array(['Male', 'Female'], dtype=object)
In [17]: print(sum(df.duplicated()))
          0
In [18]: ig, axes = plt.subplots(1,2, figsize=(21,6))
          sns.distplot(df['Age'], ax=axes[0])
sns.distplot(df['Annual Income (k$)'], ax=axes[1])
Out[18]: <matplotlib.axes._subplots.AxesSubplot at 0x7fd50607bc18>
                                                         0.012
           0.025
                                                         0.010
           0.020
           0.015
                                                         0.006
```

```
In [19]: sns.countplot(x='Gender',data=df,palette='viridis')
Out[19]: <matplotlib.axes._subplots.AxesSubplot at 0x7fd505e7ac50>
```



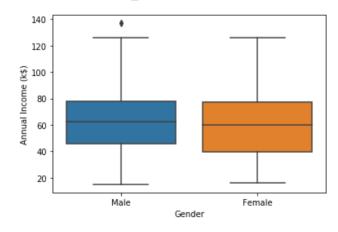
```
In [21]: sns.stripplot(x = 'Gender', y = 'Spending Score (1-100)',data = df)
```

Out[21]: <matplotlib.axes._subplots.AxesSubplot at 0x7fd502e05c50>



In [22]: sns.boxplot(x='Gender',y='Annual Income (k\$)',data=df)

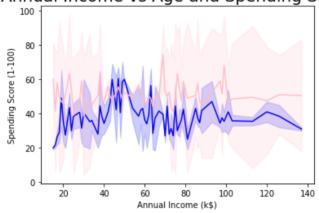
Out[22]: <matplotlib.axes._subplots.AxesSubplot at 0x7fd502c69dd8>



```
In [25]: x = df['Annual Income (k$)']
y = df['Age']
z = df['Spending Score (1-100)']

sns.lineplot(x, y, color = 'blue')
sns.lineplot(x, z, color = 'pink')
plt.title('Annual Income vs Age and Spending Score', fontsize = 20)
plt.show()
```

Annual Income vs Age and Spending Score



```
In [29]: df.head()
```

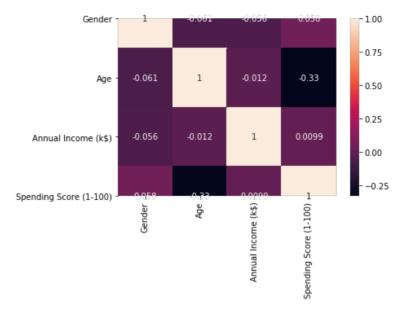
Out[29]:

| | CustomerID | Gender | Age | Annual Income (k\$) | Spending Score (1-100) |
|---|------------|--------|-----|---------------------|------------------------|
| 0 | 1 | 0 | 19 | 15 | 39 |
| 1 | 2 | 0 | 21 | 15 | 81 |
| 2 | 3 | 1 | 20 | 16 | 6 |
| 3 | 4 | 1 | 23 | 16 | 77 |
| 4 | 5 | 1 | 31 | 17 | 40 |

```
In [31]: | df.drop('CustomerID',axis=1,inplace=True)
         ______
         KevFrror
                                                    Traceback (most recent call last)
         <ipython-input-31-57491add0442> in <module>
         ----> 1 df.drop('CustomerID',axis=1,inplace=True)
         /home/tarun/snap/jupyter/common/lib/python3.7/site-packages/pandas/core/fram
         e.py in drop(self, labels, axis, index, columns, level, inplace, errors)
            4115
                              level=level,
            4116
                              inplace=inplace,
          -> 4117
                              errors=errors,
            4118
                          )
            4119
         /home/tarun/snap/jupyter/common/lib/python3.7/site-packages/pandas/core/gener
         ic.py in drop(self, labels, axis, index, columns, level, inplace, errors)
            3912
                         for axis, labels in axes.items():
            3913
                              if labels is not None:
         -> 3914
                                  obj = obj._drop_axis(labels, axis, level=level, error
         s=errors)
            3915
            3916
                         if inplace:
         /home/tarun/snap/jupyter/common/lib/python3.7/site-packages/pandas/core/gener
         ic.py in _drop_axis(self, labels, axis, level, errors)
                                  new_axis = axis.drop(labels, level=level, errors=erro
            3944
         rs
            3945
         -> 3946
                                  new axis = axis.drop(labels, errors=errors)
            3947
                              result = self.reindex(**{axis_name: new_axis})
            3948
         /home/tarun/snap/jupyter/common/lib/python3.7/site-packages/pandas/core/index
         es/base.py in drop(self, labels, errors)
            5338
                         if mask.any():
            5339
                             if errors != "ignore":
                                  raise KeyError("{} not found in axis".format(labels[m
         -> 5340
         ask]))
                              indexer = indexer[~mask]
            5341
            5342
                         return self.delete(indexer)
         KeyError: "['CustomerID'] not found in axis"
In [32]:
         df.head()
Out[32]:
            Gender Age Annual Income (k$) Spending Score (1-100)
          0
                O
                    19
                                   15
                                                    39
                0
                    21
                                   15
                                                    81
          1
          2
                    20
                                   16
                                                     6
                1
          3
                1
                    23
                                   16
                                                    77
          4
                1
                    31
                                   17
                                                    40
```

In [33]: sns.heatmap(df.corr(),annot=True)

Out[33]: <matplotlib.axes._subplots.AxesSubplot at 0x7fd5001bbe10>



```
In [35]: def impute_age(cols):
    spend=cols
    if spend > 55:
        return 1
    else:
        return 0
    df['Spending Score (1-100)'] = df['Spending Score (1-100)'].apply(impute_ag e)
    df.head()
```

Out[35]:

| | Gender | Age | Annual Income (k\$) | Spending Score (1-100) |
|---|--------|-----|---------------------|------------------------|
| 0 | 0 | 19 | 15 | 0 |
| 1 | 0 | 21 | 15 | 1 |
| 2 | 1 | 20 | 16 | 0 |
| 3 | 1 | 23 | 16 | 1 |
| 4 | 1 | 31 | 17 | 0 |

In [36]: df

Out[361:

| | Gender | Age | Annual Income (k\$) | Spending Score (1-100) |
|-----|--------|-----|---------------------|------------------------|
| 0 | 0 | 19 | 15 | 0 |
| 1 | 0 | 21 | 15 | 1 |
| 2 | 1 | 20 | 16 | 0 |
| 3 | 1 | 23 | 16 | 1 |
| 4 | 1 | 31 | 17 | 0 |
| | | | | |
| 195 | 1 | 35 | 120 | 1 |
| 196 | 1 | 45 | 126 | 0 |
| 197 | 0 | 32 | 126 | 1 |
| 198 | 0 | 32 | 137 | 0 |
| 199 | 0 | 30 | 137 | 1 |

200 rows × 4 columns

```
In [37]: from sklearn.model_selection import train_test_split
    X_train, X_test, y_train, y_test= train_test_split(df.drop('Spending Score
    (1-100)',axis=1), df['Spending Score (1-100)'], test_size=0.30, random_stat
    e=101)
    from sklearn.linear_model import LogisticRegression
    log=LogisticRegression()
    log.fit(X_train,y_train)
    pred=log.predict(X_test)
    from sklearn.metrics import classification_report
    print(classification_report(y_test, pred))
```

```
recall f1-score
              precision
                                                 support
           0
                    0.63
                               0.86
                                         0.73
                                                      36
           1
                    0.55
                               0.25
                                         0.34
                                                      24
                                         0.62
                                                      60
    accuracy
   macro avg
                    0.59
                               0.56
                                         0.54
                                                      60
weighted avg
                    0.60
                                         0.57
                                                      60
                               0.62
```

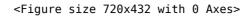
```
In [38]: from sklearn.preprocessing import StandardScaler
    scaler = StandardScaler()
    scaler.fit(df.drop('Spending Score (1-100)',axis=1))
    scaled_features = scaler.transform(df.drop('Spending Score (1-100)',axis=1))
    df_feat = pd.DataFrame(scaled_features,columns=df.columns[:-1])
    df_feat.head()
```

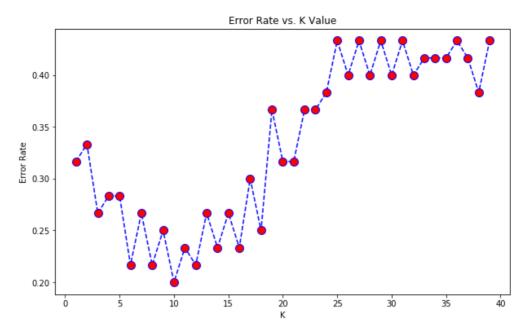
Out[38]:

| | Gender | Age | Annual Income (k\$) |
|---|-----------|-----------|---------------------|
| 0 | -1.128152 | -1.424569 | -1.738999 |
| 1 | -1.128152 | -1.281035 | -1.738999 |
| 2 | 0.886405 | -1.352802 | -1.700830 |
| 3 | 0.886405 | -1.137502 | -1.700830 |
| 4 | 0.886405 | -0.563369 | -1.662660 |

```
In [39]:
          from sklearn.model_selection import train_test_split
          X_train, X_test, y_train, y_test = train_test_split(scaled_features,df['Spen
ding Score (1-100)'],
                                                                    test size=0.30)
          from sklearn.neighbors import KNeighborsClassifier
          knn = KNeighborsClassifier(n_neighbors=1)
          knn.fit(X_train,y_train)
          pred = knn.predict(X_test)
from sklearn.metrics import classification_report,confusion_matrix
          print(confusion_matrix(y_test,pred))
          print(classification report(y test,pred))
          [[28 13]
           [ 6 13]]
                         precision
                                        recall f1-score
                                                             support
                                                     0.75
                      0
                               0.82
                                          0.68
                                                                  41
                               0.50
                                          0.68
                                                     0.58
                                                                  19
                                                     0.68
                                                                  60
              accuracy
                               0.66
                                          0.68
                                                     0.66
                                                                  60
             macro avg
                                                     0.69
          weighted avg
                               0.72
                                          0.68
                                                                  60
```

```
Out[40]: Text(0, 0.5, 'Error Rate')
         <Figure size 720x432 with 0 Axes>
         <Figure size 720x432 with 0 Axes>
```





```
In [41]: knn = KNeighborsClassifier(n_neighbors=3)
    knn.fit(X_train,y_train)
    pred = knn.predict(X_test)
    print('WITH K=3')
    print('\n')
    print(confusion_matrix(y_test,pred))
    print('\n')
    print(classification_report(y_test,pred))
```

WITH K=3

[[31 10] [6 13]]

| | precision | recall | f1-score | support |
|---------------------------------------|--------------|--------------|----------------------|----------------|
| 0 1 | 0.84 0.57 | 0.76 0.68 | 0.79 0.62 | 41 19 |
| accuracy macro avg weighted avg | 0.70 0.75 | 0.72 0.73 | 0.73 0.71 0.74 | 60 60 60 |

```
In [ ]:
```