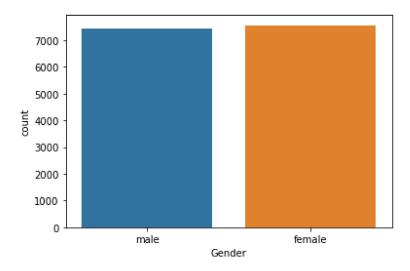
## Importing the Dependencies

```
In [2]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.model_selection import train_test_split
        from sklearn import metrics
        from sklearn.linear_model import LinearRegression
        from sklearn.ensemble import RandomForestRegressor
        Data Collection and Processing
In [3]:
        calories data=pd.read csv('calories.csv')
        exercise_data=pd.read_csv('exercise.csv')
In [4]:
        calories=pd.concat([exercise data,calories data['Calories']],axis=1)
In [5]:
        calories.head()
In [6]:
            User_ID Gender Age Height Weight Duration Heart_Rate Body_Temp Calories
Out[6]:
        0 14733363
                             68
                                  190.0
                                          94.0
                                                   29.0
                                                             105.0
                                                                         40.8
                                                                                231.0
                      male
        1 14861698
                                  166.0
                                          60.0
                                                   14.0
                                                              94.0
                                                                         40.3
                                                                                 66.0
                     female
                             20
        2 11179863
                                  179.0
                                          79.0
                                                    5.0
                                                              88.0
                                                                         38.7
                                                                                 26.0
                      male
                             69
        3 16180408
                     female
                                  179.0
                                          71.0
                                                   13.0
                                                             100.0
                                                                         40.5
                                                                                 71.0
        4 17771927
                     female
                             27
                                  154.0
                                          58.0
                                                   10.0
                                                              81.0
                                                                         39.8
                                                                                 35.0
        calories.shape
In [7]:
         (15000, 9)
Out[7]:
        calories.isnull().sum()
In [8]:
        User ID
Out[8]:
        Gender
                        0
        Age
                        0
        Height
                        0
        Weight
                       0
        Duration
        Heart Rate
                       0
                        0
        Body Temp
        Calories
        dtype: int64
        Data Visualization
        sns.countplot(calories['Gender'])
In [9]:
```

C:\Users\DELL\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWar ning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments wit hout an explicit keyword will result in an error or misinterpretation.

warnings.warn(

Out[9]: <AxesSubplot:xlabel='Gender', ylabel='count'>

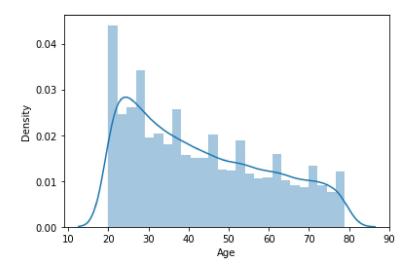


In [10]: sns.distplot(calories['Age'])

C:\Users\DELL\anaconda3\lib\site-packages\seaborn\distributions.py:2619: Futur eWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[10]: <AxesSubplot:xlabel='Age', ylabel='Density'>



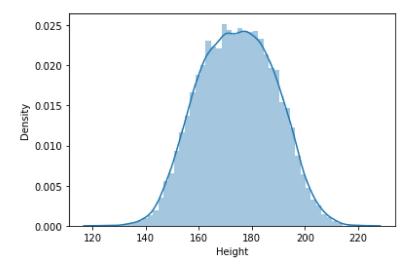
In [11]: sns.distplot(calories['Height'])

C:\Users\DELL\anaconda3\lib\site-packages\seaborn\distributions.py:2619: Futur eWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings, FutureWarning)

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Out[11]: <AxesSubplot:xlabel='Height', ylabel='Density'>

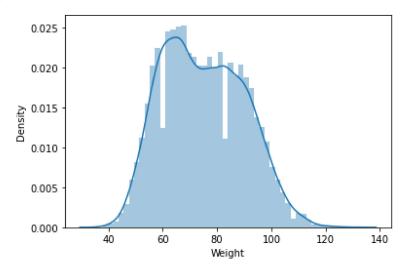


In [12]: sns.distplot(calories['Weight'])

C:\Users\DELL\anaconda3\lib\site-packages\seaborn\distributions.py:2619: Futur eWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

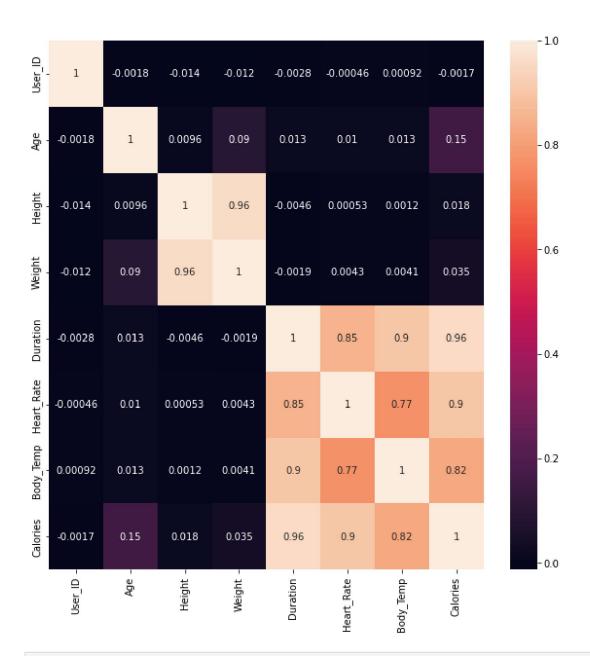
warnings.warn(msg, FutureWarning)

Out[12]: <a href="AxesSubplot:xlabel='Weight'"> <a href="AxesSubplot:xlabel='Weight'>



```
In [13]: correlation=calories.corr()
  plt.figure(figsize=(10,10))
  sns.heatmap(correlation,annot=True)
```

Out[13]: <AxesSubplot:>



In [14]: | calories.replace({"Gender":{'male':0,'female':1}},inplace=True)

In [15]: calories.head()

Out[15]:		User_ID	Gender	Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Calories
	0	14733363	0	68	190.0	94.0	29.0	105.0	40.8	231.0
	1	14861698	1	20	166.0	60.0	14.0	94.0	40.3	66.0
	2	11179863	0	69	179.0	79.0	5.0	88.0	38.7	26.0
	3	16180408	1	34	179.0	71.0	13.0	100.0	40.5	71.0
	4	17771927	1	27	154.0	58.0	10.0	81.0	39.8	35.0

## Separating features and target

```
In [17]: print(X)
               Gender Age Height Weight Duration Heart Rate Body Temp
        0
                      68
                                   94.0
                                           29.0
                                                     105.0
                 0
                           190.0
                                                                      40.8
        1
                   1
                       20
                            166.0
                                     60.0
                                               14.0
                                                          94.0
                                                                      40.3
        2
                   0
                       69
                            179.0
                                     79.0
                                               5.0
                                                           88.0
                                                                      38.7
                                     71.0
        3
                   1
                       34
                            179.0
                                               13.0
                                                          100.0
                                                                      40.5
        4
                   1
                       27
                            154.0
                                    58.0
                                              10.0
                                                          81.0
                                                                      39.8
                  ... ...
                             . . .
                                     . . .
                                               . . .
                                                           . . .
                                                                      . . .
        14995
                  1 20
                           193.0
                                   86.0
                                             11.0
                                                          92.0
                                                                     40.4
                   1 27
        14996
                            165.0
                                    65.0
                                               6.0
                                                          85.0
                                                                     39.2
                   1 43
                                    58.0
                                              16.0
                                                          90.0
                                                                     40.1
        14997
                           159.0
                       78
                   0
                                     97.0
                                               2.0
        14998
                            193.0
                                                           84.0
                                                                      38.3
        14999
                   0
                       63
                             173.0
                                     79.0
                                               18.0
                                                           92.0
                                                                      40.5
        [15000 rows x 7 columns]
In [18]: print(Y)
                 231.0
        1
                  66.0
        2
                  26.0
        3
                 71.0
                 35.0
        14995
                45.0
        14996
                 23.0
        14997
                 75.0
        14998
                 11.0
                  98.0
        14999
        Name: Calories, Length: 15000, dtype: float64
        Splitting the data into training and testing data
In [19]: X_train, X_test, Y_train, Y_test=train_test_split(X,Y,test_size=0.2)
        Model Training
        Linear Regressor
In [20]: | model1=LinearRegression()
        model1.fit(X train, Y train)
        prediction1=model1.predict(X test)
        print(prediction1)
        [ 18.90073679 112.37963061 2.12558084 ... 30.60482603 176.6454258
         182.98259929]
        Mean Absolute Error for Linear Regressor
In [21]: | mae=metrics.mean_absolute_error(Y_test,prediction1)
        print("Mean Absolute Error = ", mae)
        Mean Absolute Error = 8.379646102595844
```

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Random Forest Regressor

```
In [22]: model2=RandomForestRegressor()
         model2.fit(X_train,Y_train)
         prediction2=model2.predict(X_test)
         print(prediction2)
         [ 33.31 101.05 11.47 ... 37.42 184.96 172.61]
         Mean Absolute Error for Random Forest Regressor
In [23]: | mae=metrics.mean_absolute_error(Y_test,prediction2)
         print("Mean Absolute Error= ",mae)
        Mean Absolute Error = 1.84171
In [28]: pred = model2.predict([[1, 68, 50.0, 94.0, 90.0, 105.0, 10.8]])
         pred
        C:\Users\DELL\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X
         does not have valid feature names, but RandomForestRegressor was fitted with f
         eature names
          warnings.warn(
        array([204.28])
Out[28]:
In [ ]:
 In [ ]:
```