```
In [31]:
           #LOADING PYTHON LIBRARIES
           import numpy as np #LIBRARY TO CREATE ARRAYS WHICH IS USEFULL FOR DATA-PROCESSING.ALIAZ
           import pandas as pd #LIBRARY USED TO MAKE DATAFRAMES(STRUCTURED TABLES).
           import matplotlib.pyplot as plt #CREATING PLOTS FOR DATA ANALYSIS(VIZUALISATION LIBRARY
           import seaborn as sns #VIZUALISATION LIBRARY.
           from sklearn.model selection import train test split #SPLIT DATASETS.
           from xgboost import XGBRegressor #FOR THE ALGORITHM.
           from sklearn import metrics #USED TO EVALUATE OUR MODEL.
In [19]:
           #LOADING exercise.csv
          exercise_data = pd.read_csv('exercise.csv')
           exercise data.head() #first five coloumns is displayed.
           exercise data.shape #displays size of the array.
Out[19]: (15000, 8)
In [22]:
           #LOAD calories.csv
           calories = pd.read csv('calories.csv')
           calories.head() #first five coloumns is displayed.
           #alories.shape #displays size of the array.
Out[22]:
              User_ID Calories
            14733363
                        231.0
            14861698
                          66.0
            11179863
                          26.0
            16180408
                          71.0
          4 17771927
                          35.0
In [20]:
           #COMBINE exercise.csv and calories.csv
           calories_data = pd.concat([exercise_data, calories['Calories']], axis=1)# axis=1 repres
In [27]:
           #COMBINED DATASET
           calories data.head()
          #calories data.shape
Out[27]:
              User_ID Gender Age Height Weight Duration Heart_Rate Body_Temp Calories
          0 14733363
                                     190.0
                                             94.0
                                                       29.0
                                                                 105.0
                                                                             40.8
                                                                                     231.0
                        male
                               68
            14861698
                                    166.0
                                             60.0
                                                       14.0
                                                                  94.0
                                                                             40.3
                                                                                      66.0
                       female
                               20
            11179863
                        male
                               69
                                    179.0
                                             79.0
                                                        5.0
                                                                  88.0
                                                                             38.7
                                                                                      26.0
            16180408
                       female
                                    179.0
                                             71.0
                                                       13.0
                                                                 100.0
                                                                             40.5
                                                                                      71.0
                               34
          4 17771927
                       female
                               27
                                    154.0
                                             58.0
                                                       10.0
                                                                  81.0
                                                                             39.8
                                                                                      35.0
```

calories\_data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15000 entries, 0 to 14999
Data columns (total 9 columns):
```

Column Non-Null Count Dtype User ID 0 15000 non-null int64 1 Gender 15000 non-null object 2 Age 15000 non-null int64 15000 non-null float64 3 Height 4 15000 non-null float64 Weight 5 15000 non-null float64 Duration 6 Heart Rate 15000 non-null float64 7 15000 non-null float64 Body\_Temp 8 Calories 15000 non-null float64 dtypes: float64(6), int64(2), object(1) memory usage: 1.0+ MB

In [26]:

# CHECKING FOR NULL VALUES IN THE DATA TO IMPROVE THE QUALITY OF DATA
calories\_data.isnull().sum()

Out[26]: User\_ID Gender 0 0 Age 0 Height Weight 0 0 Duration Heart\_Rate 0 Body\_Temp 0 Calories

dtype: int64

In [29]:

#STATISTICAL MEASURES ABOUT THE DATA
calories\_data.describe()

Out[29]:

|       | User_ID      | Age          | Height       | Weight       | Duration     | Heart_Rate   | Body_Tem    |
|-------|--------------|--------------|--------------|--------------|--------------|--------------|-------------|
| count | 1.500000e+04 | 15000.000000 | 15000.000000 | 15000.000000 | 15000.000000 | 15000.000000 | 15000.00000 |
| mean  | 1.497736e+07 | 42.789800    | 174.465133   | 74.966867    | 15.530600    | 95.518533    | 40.02545    |
| std   | 2.872851e+06 | 16.980264    | 14.258114    | 15.035657    | 8.319203     | 9.583328     | 0.77923     |
| min   | 1.000116e+07 | 20.000000    | 123.000000   | 36.000000    | 1.000000     | 67.000000    | 37.10000    |
| 25%   | 1.247419e+07 | 28.000000    | 164.000000   | 63.000000    | 8.000000     | 88.000000    | 39.60000    |
| 50%   | 1.499728e+07 | 39.000000    | 175.000000   | 74.000000    | 16.000000    | 96.000000    | 40.20000    |
| 75%   | 1.744928e+07 | 56.000000    | 185.000000   | 87.000000    | 23.000000    | 103.000000   | 40.60000    |
| max   | 1.999965e+07 | 79.000000    | 222.000000   | 132.000000   | 30.000000    | 128.000000   | 41.50000    |

In [32]:

#DATA VIZUALIZATION
sns.set() #THEMES FOR PLOTTING

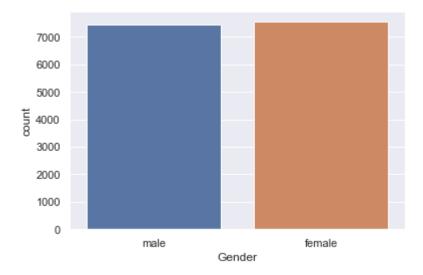
In [34]:

#DATA ANALYSIS OF GENDER USING COUNT PLOT

sns.countplot(calories\_data['Gender'])#SHOWS EQUAL DISTRIBUTION OF MALE AND FEMALE AS T

c:\users\hp-kle\appdata\local\programs\python\python39\lib\site-packages\seaborn\\_decora
tors.py:36: FutureWarning: 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(

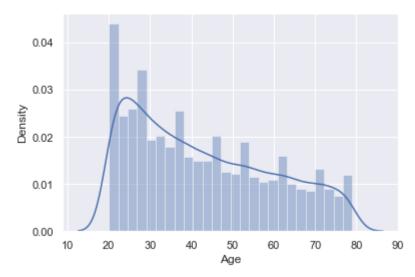
Out[34]: <AxesSubplot:xlabel='Gender', ylabel='count'>



In [35]: #DATA ANALYSIS OF AGE USING DISTRIBUTION PLOT
sns.distplot(calories\_data['Age'])#BASED ON THE DISTRIBUTION PLOT MORE VALUES IN THE RA

c:\users\hp-kle\appdata\local\programs\python\python39\lib\site-packages\seaborn\distrib
utions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed i
n a future version. Please adapt your code to use either `displot` (a figure-level funct
ion with similar flexibility) or `histplot` (an axes-level function for histograms).
 warnings.warn(msg, FutureWarning)

Out[35]: <AxesSubplot:xlabel='Age', ylabel='Density'>



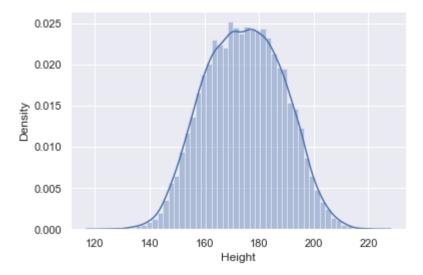
In [36]: #DATA ANALYSIS OF HEIGHT USING DISTRIBUTION PLOT
 sns.distplot(calories\_data['Height'])#NORMAL DISTRIBUTION

c:\users\hp-kle\appdata\local\programs\python\python39\lib\site-packages\seaborn\distrib
utions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed i

n a future version. Please adapt your code to use either `displot` (a figure-level funct ion with similar flexibility) or `histplot` (an axes-level function for histograms).

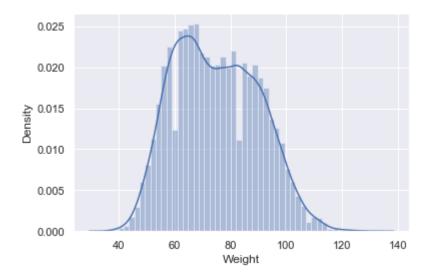
warnings.warn(msg, FutureWarning)

Out[36]: <AxesSubplot:xlabel='Height', ylabel='Density'>



c:\users\hp-kle\appdata\local\programs\python\python39\lib\site-packages\seaborn\distrib
utions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed i
n a future version. Please adapt your code to use either `displot` (a figure-level funct
ion with similar flexibility) or `histplot` (an axes-level function for histograms).
 warnings.warn(msg, FutureWarning)

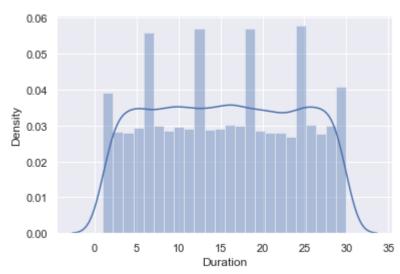
Out[37]: <AxesSubplot:xlabel='Weight', ylabel='Density'>



In [38]: #DATA ANALYSIS OF DURATION USING DISTRIBUTION PLOT
sns.distplot(calories\_data['Duration'])

c:\users\hp-kle\appdata\local\programs\python\python39\lib\site-packages\seaborn\distrib
utions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed i
n a future version. Please adapt your code to use either `displot` (a figure-level funct
ion with similar flexibility) or `histplot` (an axes-level function for histograms).
 warnings.warn(msg, FutureWarning)

Out[38]: <AxesSubplot:xlabel='Duration', ylabel='Density'>

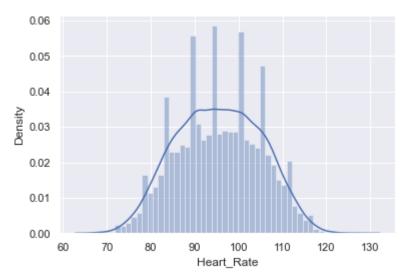


In [39]:

#DATA ANALYSIS OF HEART RATE USING DISTRIBUTION PLOT
sns.distplot(calories\_data['Heart\_Rate'])

c:\users\hp-kle\appdata\local\programs\python\python39\lib\site-packages\seaborn\distrib
utions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed i
n a future version. Please adapt your code to use either `displot` (a figure-level funct
ion with similar flexibility) or `histplot` (an axes-level function for histograms).
 warnings.warn(msg, FutureWarning)

Out[39]: <AxesSubplot:xlabel='Heart\_Rate', ylabel='Density'>

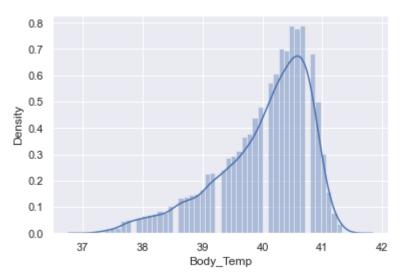


In [40]:

#DATA ANALYSIS OF BODY TEMPERATURE USING DISTRIBUTION PLOT
sns.distplot(calories data['Body Temp'])

c:\users\hp-kle\appdata\local\programs\python\python39\lib\site-packages\seaborn\distrib
utions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed i
n a future version. Please adapt your code to use either `displot` (a figure-level funct
ion with similar flexibility) or `histplot` (an axes-level function for histograms).
 warnings.warn(msg, FutureWarning)

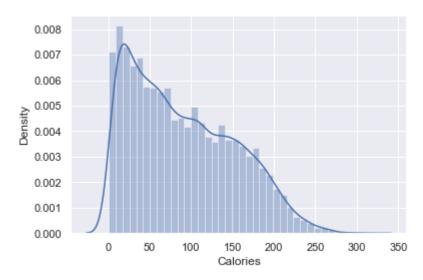
Out[40]: <AxesSubplot:xlabel='Body\_Temp', ylabel='Density'>



In [41]: #DATA ANALYSIS OF CALORIES USING DISTRIBUTION PLOT
sns.distplot(calories data['Calories'])

c:\users\hp-kle\appdata\local\programs\python\python39\lib\site-packages\seaborn\distrib
utions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed i
n a future version. Please adapt your code to use either `displot` (a figure-level funct
ion with similar flexibility) or `histplot` (an axes-level function for histograms).
 warnings.warn(msg, FutureWarning)

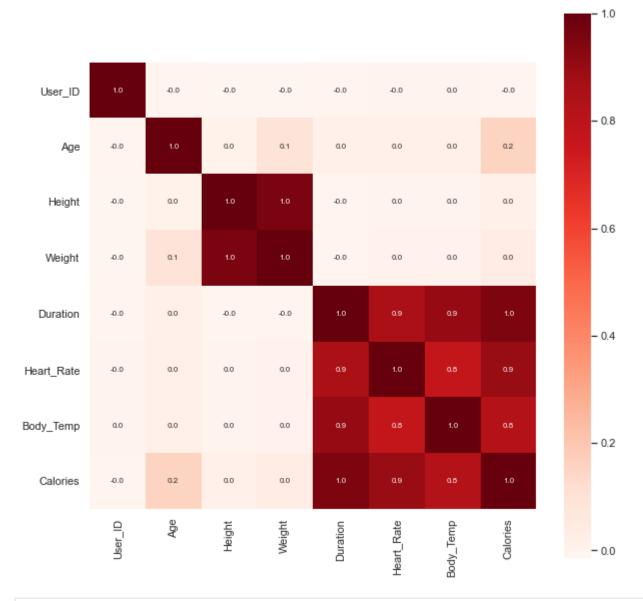
Out[41]: <AxesSubplot:xlabel='Calories', ylabel='Density'>



In [47]: #FINDING THE CORRELATION BETWEEN THE DATA
correlation = calories\_data.corr()

In [55]: #CONSTRUCTING A HEATMAP TO UNDERSTAND CORRELATION
plt.figure(figsize=(10,10))
sns.heatmap(correlation, cbar=True, square=True, fmt='.1f', annot=True, annot\_kws={'siz}
# ACCORDING TO THE HEAT MAP THE FIELDS Duration Heart\_Rate AND Body\_Temp ARE POSITIVELY
#THE Height AND Weight ARE POSITIVELY CORRELATED.
#MEANING THE DATA ARE DIRECTLY PROPOTIONAL TO EACH OTHER.

Out[55]: <AxesSubplot:>



In [59]:
#CONVERTING THE GENDER FEILD IN THE DATASET FROM TEXT TO NUMERICAL VALUES.
calories\_data.replace({"Gender":{'male':0,'female':1}}, inplace=True)
calories\_data
#GENDER IS CONVERTED TO NUMERICAL VALUES

| Out[59]: |       | 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     |
|          | •••   |          |        |     |        |        |          |            |           |          |
|          | 14995 | 15644082 | 1      | 20  | 193.0  | 86.0   | 11.0     | 92.0       | 40.4      | 45.0     |
|          | 14996 | 17212577 | 1      | 27  | 165.0  | 65.0   | 6.0      | 85.0       | 39.2      | 23.0     |
|          | 14997 | 17271188 | 1      | 43  | 159.0  | 58.0   | 16.0     | 90.0       | 40.1      | 75.0     |

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|  |   |                      |                                      |                                      |               | ML             |                                      |       |                                      |      |          |
|--|---|----------------------|--------------------------------------|--------------------------------------|---------------|----------------|--------------------------------------|-------|--------------------------------------|------|----------|
|  | User_ID   | Gender               | Age                                  | Height                               | Weight        | Duration       | Heart_                               | Rate  | Body_Te                              | mp   | Calories |
| 14998  | 18643037  | 0                    | 78                                   | 193.0                                | 97.0          | 2.0            | )                                    | 84.0  | 3                                    | 38.3 | 11.0     |
| 14999  | 11751526  | 0                    | 63                                   | 173.0                                | 79.0          | 18.0           | )                                    | 92.0  | 2                                    | 40.5 | 98.0     |
| 15000 r  | ows × 9 co  | olumns               |                                      |                                      |               |                |                                      |       |                                      |      |          |
| <pre>#ALL THE COLOUMNS ARE KNOW AS FEAUTURES. #THE CALORIES FEILD IS CALLED TARGET. #SEPERATING FEATURES AND TARGETS BEFORE TRAINING X = calories_data.drop(columns=['User_ID', 'Calories']) Y = calories_data['Calories']</pre> |   |                      |                                      |                                      |               |                |                                      |       |                                      |      |          |
| print  | (X)#User_   | .ID and              | Calor                                | ies Drop                             | oped.         |                |                                      |       |                                      |      |          |
| 0<br>1<br>2  | 0<br>1<br>0                                       | 68 1<br>20 1<br>69 1 | .90.0<br>.66.0<br>.79.0              | 94.0<br>60.0<br>79.0                 | 5             | .0<br>.0<br>.0 | rt_Rate<br>105.0<br>94.0<br>88.0     | Bod   | y_Temp<br>40.8<br>40.3<br>38.7       |      |          |
| 3<br>4   | 1<br>1  | 27 1                 | .79.0<br>.54.0                       | 71.0<br>58.0                         | 10            | .0             | 100.0<br>81.0                        |       | 40.5<br>39.8                         |      |          |
| 14995<br>14996<br>14997<br>14998<br>14999  | 1<br>1<br>1<br>0<br>0                             | 27 1<br>43 1<br>78 1 | <br>.93.0<br>.65.0<br>.59.0<br>.93.0 | 86.0<br>65.0<br>58.0<br>97.0<br>79.0 | 11<br>6<br>16 | .0<br>.0<br>.0 | 92.0<br>85.0<br>90.0<br>84.0<br>92.0 |       | 40.4<br>39.2<br>40.1<br>38.3<br>40.5 |      |          |
| [15000   | rows x 7  | olumn column         | ıs]                                  |                                      |               |                |                                      |       |                                      |      |          |
| print  | (Y)   |                      |                                      |                                      |               |                |                                      |       |                                      |      |          |
| 0<br>1<br>2<br>3<br>4  | 231.0<br>66.0<br>26.0<br>71.0<br>35.0             |                      |                                      |                                      |               |                |                                      |       |                                      |      |          |
| 14995<br>14996<br>14997<br>14998<br>14999<br>Name:   | 45.0<br>23.0<br>75.0<br>11.0<br>98.0<br>Calories, | Length               | ı: 150                               | 00, dtyμ                             | oe: floa      | t64            |                                      |       |                                      |      |          |
|  | TTING DAT   |                      |                                      |                                      |               |                | lit(X, Y                             | ′, te | st_size=                             | =0.2 | , random |

#TRAINING DATA IS 12000 AND TESTING DATA IS 3000, THERFORE TOTAL DATA IS 15000.

model = XGBRegressor() #LOADING THE MODEL local host: 8888/nbconvert/html/Desktop/ML.ipynb?download=false

print(X.shape,X\_train.shape,X\_test.shape)

#MODEL TRAINING USING XGBoost Regressor

(15000, 7) (12000, 7) (3000, 7)

In [68]:

In [69]:

```
In [70]:
          model.fit(X train, Y train)# MODEL TRAINING WHEN THIS DATA IS GIVEN TO XGBoost Regresso
Out[70]: XGBRegressor(base score=0.5, booster='gbtree', colsample bylevel=1,
                       colsample bynode=1, colsample bytree=1, gamma=0, gpu id=-1,
                       importance_type='gain', interaction_constraints='',
                      learning rate=0.300000012, max delta step=0, max depth=6,
                      min child weight=1, missing=nan, monotone constraints='()',
                      n estimators=100, n jobs=4, num parallel tree=1, random state=0,
                      reg alpha=0, reg lambda=1, scale pos weight=1, subsample=1,
                      tree method='exact', validate parameters=1, verbosity=None)
In [71]:
          #MODEL IS TRAINED NOW EVALUATION OF MODEL BASED ON TEST DATA
          #PREDICTION ON TEST DATA
          test data prediction = model.predict(X test)
In [72]:
          print(test data prediction)
          #VALUES PREDICTED BY OUR MODEL
          [127.823784 226.00154
                                  38.66253 ... 144.3636
                                                            22.767195 89.87375 ]
In [73]:
          #COMPARING VALUES PREDICTED BY OUR MODEL VS ORIGINAL VALUES(Y test)
          #MEAN ABSOLUTE ERROR METRICS(MAGNITUDE OF ERROR OUR MODEL IS MAKING COMPARED TO ORIGINA
          mae = metrics.mean absolute error(Y test, test data prediction)
In [74]:
          print("MEAN ABSOLUTE ERROR=", mae)#GOOD SCORE AS THE MAGNITUDE OF ERROR IS LESS
         MEAN ABSOLUTE ERROR= 1.4807048829992613
In [75]:
          print(Y test)
         7592
                  127.0
         3551
                  224.0
         9698
                   38.0
         3759
                    6.0
         2353
                  137.0
                  177.0
         8859
         2886
                   49.0
         14357
                  145.0
         9430
                   24.0
                   90.0
         11870
         Name: Calories, Length: 3000, dtype: float64
 In [ ]:
```