

```
In [35]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [36]: calories_data=pd.read_csv('C:/Users/ASUS/Desktop/calories burnt predictor/calories.csv')
calories_data.head(3)
```

```
Out[36]:
```

	User_ID	Calories
0	14733363	231.0
1	14861698	66.0
2	11179863	26.0
3	16180408	71.0
4	17771927	35.0

```
In [37]: exercise_data=pd.read_csv('C:/Users/ASUS/Desktop/calories burnt predictor/exercise.csv')
exercise_data.head()
```

```
Out[37]:
```

	User_ID	Gender	Age	Height	Weight	Duration	Heart_Rate	Body_Temp
0	14733363	male	68	190.0	94.0	29.0	105.0	40.8
1	14861698	female	20	166.0	60.0	14.0	94.0	40.3
2	11179863	male	69	179.0	79.0	5.0	88.0	38.7
3	16180408	female	34	179.0	71.0	13.0	100.0	40.5
4	17771927	female	27	154.0	58.0	10.0	81.0	39.8

```
In [38]: new_df=exercise_data.merge(calories_data,on='User_ID')
new_df.head(3)
```

```
Out[38]:
```

	User_ID	Gender	Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Calories
0	14733363	male	68	190.0	94.0	29.0	105.0	40.8	231.0
1	14861698	female	20	166.0	60.0	14.0	94.0	40.3	66.0
2	11179863	male	69	179.0	79.0	5.0	88.0	38.7	26.0

```
In [39]: new_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 15000 entries, 0 to 14999
Data columns (total 9 columns):
 #   Column      Non-Null Count  Dtype
---  --
 0   User_ID    15000 non-null   int64
 1   Gender     15000 non-null   object
 2   Age        15000 non-null   int64
 3   Height     15000 non-null   float64
 4   Weight     15000 non-null   float64
 5   Duration   15000 non-null   float64
 6   Heart_Rate 15000 non-null   float64
 7   Body_Temp  15000 non-null   float64
 8   Calories   15000 non-null   float64
dtypes: float64(8), int64(2), object(1)
memory usage: 1.1+ MB
```

```
In [40]: new_df.describe()
```

```
Out[40]:
```

	User_ID	Gender	Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Calories
count	1.500000e+04	15000.000000	15000.000000	15000.000000	15000.000000	15000.000000	15000.000000	15000.000000	15000.000000
mean	1.497736e+07	42.788900	174.465133	74.966867	15.530600	95.518533	40.025453	89.539533	
std	2.872891e+06	16.860264	14.256114	15.030667	8.319203	9.583328	0.779230	62.456978	
min	1.000116e+07	20.000000	123.000000	36.000000	1.000000	67.000000	37.100000	1.000000	
25%	1.247413e+07	28.000000	164.000000	63.000000	8.000000	88.000000	39.600000	35.000000	
50%	1.497736e+07	38.000000	175.000000	74.000000	16.000000	96.000000	40.000000	79.000000	
75%	1.744059e+07	55.000000	185.000000	97.000000	23.000000	103.000000	40.600000	139.000000	
max	1.999955e+07	79.000000	222.000000	132.000000	30.000000	128.000000	41.500000	314.000000	

```
In [40]: new_df.drop('User_ID',axis=1,inplace=True)
```

```
In [41]: new_df.head(3)
```

```
Out[41]:
```

	Gender	Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Calories
0	male	68	190.0	94.0	29.0	105.0	40.8	231.0
1	female	20	166.0	60.0	14.0	94.0	40.3	66.0
2	male	69	179.0	79.0	5.0	88.0	38.7	26.0

```
In [42]: #Checking how many males and females are there
sns.countplot(new_df.Gender)
```

```
Out[42]:
```

```
In [43]: #Its giving equal distribution for both
```

```
In [45]: #finding the distribution of "Age" column and
sns.distplot(new_df.Age)
```

```
Out[45]:
```

```
In [46]: sns.set()
sns.distplot(new_df['Height'])
```

```
Out[46]:
```

```
In [47]: sns.distplot(new_df.Weight)
```

```
Out[47]:
```

```
In [48]: sns.distplot(new_df.Duration)
```

```
Out[48]:
```

```
In [49]: sns.distplot(new_df.Calories)
```

```
Out[49]:
```

```
In [50]: sns.distplot(new_df.Heart_Rate)
```

```
Out[50]:
```

```
In [51]: #finding correlation to check how strongly data is related to each other
cor_realion=new_df.corr()
```

```
Out[52]:
```

	Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Calories
Age	1.000000	0.009584	0.006004	0.013247	0.010482	0.013175	0.154395
Height	0.009584	1.000000	0.094851	-0.004825	0.009528	0.001200	0.017537
Weight	0.006004	0.094851	1.000000	-0.001894	0.004311	0.004909	0.030461
Duration	0.013247	-0.004825	-0.001894	1.000000	0.026369	0.003157	0.955421
Heart_Rate	0.010482	0.009528	0.004311	0.026369	1.000000	0.771529	0.897882
Body_Temp	0.013175	0.001200	0.004909	0.003157	0.771529	1.000000	0.824558
Calories	0.154395	0.017537	0.030461	0.955421	0.897882	0.824558	1.000000

```
In [53]: #Building heatmap for correlation
plt.figure(figsize=(10,10))
sns.heatmap(data=cor_realion,annot=True)
#we infer that if values are high means positively correlated or else negatively
```

```
Out[53]:
```

```
In [54]: new_df.replace({"gender":{"male":0,"female":1}},inplace=True)
```

```
In [55]: new_df.head()
```

```
Out[55]:
```

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

```
In [56]: from sklearn.model_selection import train_test_split
```

```
In [57]: X=new_df.drop('Calories',axis=1)
y=new_df.Calories
```

```
In [58]: X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.2,random_state=0)
```

```
In [59]: new_df.shape
```

```
Out[59]: (15000, 8)
```

MODEL 1:LINEAR REGRESSION

```
In [61]: from sklearn.linear_model import LinearRegression
```

```
In [62]: reg=LinearRegression()
```

```
In [63]: reg.fit(X_train,y_train)
```

```
Out[63]: LinearRegression()
```

```
In [64]: reg.score(X_test,y_test)
```

```
Out[64]: 0.969214323020104
```

```
In [66]: y_predicted=reg.predict(X_test)
```

```
In [67]: y_predicted
```

```
Out[67]: array([ 37.87781302,  4.05170739, 119.59156716, ...,  28.18588645,
        156.51274893, 146.45426893])
```

```
In [68]: from sklearn.metrics import mean_absolute_error,mean_squared_error
```

```
In [69]: mae = mean_absolute_error(y_test,y_predicted)
```

```
Out[69]: 8.099679630313151
```

```
In [70]: mse=mean_squared_error(y_test,y_predicted)
```

```
Out[70]: 118.79074698385787
```

```
In [71]: new_df.columns
```

```
Out[71]: Index(['Gender', 'Age', 'Height', 'Weight', 'Duration', 'Heart_Rate',
        'Body_Temp', 'Calories'],
        dtype='object')
```

PREDICTION

```
In [72]: Gender=int(input('Enter Gender: ')) # 0 for male and 1 for female
Age=int(input('Enter Age: '))
Height=float(input('Enter Height: '))
Weight=float(input('Enter Weight: '))
Duration=float(input('Enter Duration: '))
Heart_Rate=float(input('Enter Heart_Rate: '))
Body_Temp=float(input('Enter Body_Temp: '))
X_array = [(Gender,Age,Height,Weight,Duration,Heart_Rate,Body_Temp)]
y_pred = reg.predict(X_array)
print(y_pred)
```

```
Enter Gender: 0
Enter Age: 20
Enter Height: 164
Enter Weight: 65
Enter Duration: 30
Enter Heart_Rate: 128
Enter Body_Temp: 41.2
[201.2904411]
```

MODEL-2 : XGBOOST

```
In [73]: !pip install xgboost
from xgboost import XGBRegressor
```

```
Collecting xgboost
  Downloading xgboost-1.6.1-py3-none-win_amd64.whl (125.4 MB)
Requirement already satisfied: numpy in c:\users\asus\anaconda3\lib\site-packages (from xgboost) (1.20.3)
Requirement already satisfied: scipy in c:\users\asus\anaconda3\lib\site-packages (from xgboost) (1.7.1)
Installing collected packages: xgboost
Successfully installed xgboost-1.6.1
```

XGBoost is a powerful approach for building supervised regression models. The validity of this statement can be inferred by knowing about its (XGBoost) objective function and base learners. The objective function contains loss function and a regularization term. It tells about the difference between actual values and predicted values, i.e how far the model predicted are from the real values. The most common loss functions in XGBoost for regression problems is reg.linear, and that for binary classification is reg.logistics. XGBoost is one of the ensemble learning methods.

```
In [74]: model = XGBRegressor()
```

```
In [75]: model.fit(X_train,y_train)
```

```
Out[75]: XGBRegressor(base_score=0.5, booster='gbtree', callbacks=None,
        colsample_bylevel=1, colsample_bynode=1, colsample_bynode=1,
        early_stopping_rounds=None, enable_categorical=False,
        eval_metric=None, gamma=0, gpu_id=-1, grow_policy='depthwise',
        importance_type=None, interaction_constraints='',
        learning_rate=0.300000012, max_bin=256, max_cat_to_onehot=4,
        max_delta_step=0, max_depth=6, max_leaves=0, min_child_weight=1,
        missing=nan, monotone_constraints='', n_estimators=100, n_jobs=0,
        num_parallel_trees=1, predictor='auto', random_state=0, reg_alpha=0,
        reg_lambda=1, ...)
```

```
Out[77]: 0.998801753229742
```

```
In [78]: y_predicted=model.predict(X_test)
```

```
In [79]: mae = mean_absolute_error(y_test,y_predicted)
```

```
Out[79]: 1.511337571414184
```

```
In [80]: mse=mean_squared_error(y_test,y_predicted)
```

```
Out[80]: 4.623595982666398
```

PREDICTION

```
In [81]: Gender=int(input('Enter Gender: ')) # 0 for male and 1 for female
Age=int(input('Enter Age: '))
Height=float(input('Enter Height: '))
Weight=float(input('Enter Weight: '))
Duration=float(input('Enter Duration: '))
Heart_Rate=float(input('Enter Heart_Rate: '))
Body_Temp=float(input('Enter Body_Temp: '))
X_array = np.array([(Gender,Age,Height,Weight,Duration,Heart_Rate,Body_Temp)])
y_pred = model.predict(X_array)
```

```
Enter Gender: 0
Enter Age: 20
Enter Height: 164
Enter Weight: 65
Enter Duration: 30
Enter Heart_Rate: 128
Enter Body_Temp: 41.2
array([219.4555], dtype=float32)
```

```
In [ ]:
```

```
In [ ]:
```