In [1]: import pandas as pd
import pickle
import warnings
warnings.filterwarnings("ignore")

In [2]: a=pd.read\_csv("calories.csv")

In [3]: a

Out[3]:

•		User_ID	Gender	Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Calories
	0	14733363	male	68	190	94	29	105	40.8	231
	1	14861698	female	20	166	60	14	94	40.3	66
	2	11179863	male	69	179	79	5	88	38.7	26
	3	16180408	female	34	179	71	13	100	40.5	71
	4	17771927	female	27	154	58	10	81	39.8	35
	•••									
	14995	15644082	female	20	193	86	11	92	40.4	45
	14996	17212577	female	27	165	65	6	85	39.2	23
	14997	17271188	female	43	159	58	16	90	40.1	75
	14998	18643037	male	78	193	97	2	84	38.3	11
	14999	11751526	male	63	173	79	18	92	40.5	98

15000 rows × 9 columns

In [4]: a.head(10)

231
66
26
71
35
123

**8** 12132339 male 60 186 94 21 97 40.4 134 51 90 40.2 72 **9** 17964668 female 26 146 16

56

85

22

25

95

100

40.5

40.7

112

143

In [5]: a.tail(10)

**6** 19602372

**7** 11117088

female

male

33

41

158

175

```
User_ID Gender Age Height Weight Duration Heart_Rate Body_Temp Calories
Out[5]:
         14990 19715870
                                  22
                                         190
                                                  79
                                                           19
                                                                                40.3
                                                                                          89
                          female
                                                                      96
         14991 10050978
                                  51
                                         181
                                                  87
                                                            9
                                                                      91
                                                                                39.6
                                                                                          44
                           male
         14992 14722670
                                                  70
                                                           13
                           male
                                  27
                                         170
                                                                      92
                                                                                40.1
                                                                                          46
         14993 13584585
                           male
                                  45
                                         179
                                                  78
                                                           11
                                                                      98
                                                                                39.9
                                                                                          60
         14994 18209611
                                  48
                                         159
                                                  57
                                                           10
                                                                      94
                                                                                39.8
                                                                                          52
                          female
                                                                                40.4
         14995 15644082
                          female
                                  20
                                         193
                                                  86
                                                           11
                                                                      92
                                                                                          45
         14996 17212577
                                  27
                                         165
                                                  65
                                                            6
                                                                      85
                                                                                39.2
                                                                                          23
                          female
                                                                                          75
         14997 17271188
                          female
                                  43
                                         159
                                                  58
                                                           16
                                                                      90
                                                                                40.1
         14998 18643037
                           male
                                  78
                                         193
                                                  97
                                                            2
                                                                      84
                                                                                38.3
                                                                                          11
                                                                                40.5
         14999 11751526
                           male
                                  63
                                         173
                                                  79
                                                           18
                                                                      92
                                                                                          98
        a.info()
In [6]:
         <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
          3
              Height
                          15000 non-null int64
          4
              Weight
                          15000 non-null int64
          5
              Duration
                          15000 non-null int64
          6
              Heart_Rate 15000 non-null int64
          7
              Body_Temp
                          15000 non-null float64
                          15000 non-null int64
              Calories
         dtypes: float64(1), int64(7), object(1)
         memory usage: 1.0+ MB
In [7]:
         a.shape
         (15000, 9)
Out[7]:
In [8]:
         a.isnull().sum()
                       0
        User_ID
Out[8]:
                       0
         Gender
                       0
         Age
         Height
                       0
         Weight
                       0
         Duration
                       0
         Heart Rate
                       0
                       0
         Body Temp
         Calories
         dtype: int64
In [ ]:
         b=a.drop('User_ID',axis=1)
In [9]:
```

_			
1	114	1 a i	
J	uч	2	

	Gender	Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Calories
0	male	68	190	94	29	105	40.8	231
1	female	20	166	60	14	94	40.3	66
2	male	69	179	79	5	88	38.7	26
3	female	34	179	71	13	100	40.5	71
4	female	27	154	58	10	81	39.8	35
•••								
14995	female	20	193	86	11	92	40.4	45
14996	female	27	165	65	6	85	39.2	23
14997	female	43	159	58	16	90	40.1	75
14998	male	78	193	97	2	84	38.3	11
14999	male	63	173	79	18	92	40.5	98

15000 rows × 8 columns

In [10]: b=pd.get\_dummies(b,dtype=int)

In [11]: b

Out[11]:

		Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Calories	Gender_female	Gender
	0	68	190	94	29	105	40.8	231	0	
	1	20	166	60	14	94	40.3	66	1	
	2	69	179	79	5	88	38.7	26	0	
	3	34	179	71	13	100	40.5	71	1	
	4	27	154	58	10	81	39.8	35	1	
	•••							•••		
	14995	20	193	86	11	92	40.4	45	1	
	14996	27	165	65	6	85	39.2	23	1	
	14997	43	159	58	16	90	40.1	75	1	
	14998	78	193	97	2	84	38.3	11	0	
	14999	63	173	79	18	92	40.5	98	0	

15000 rows × 9 columns

```
In [12]: y=b['Calories']
x=b.drop('Calories',axis=1)
```

In [13]:

```
231
Out[13]:
                    66
          2
                    26
          3
                    71
                    35
          14995
                    45
          14996
                    23
          14997
                    75
          14998
                    11
          14999
                    98
          Name: Calories, Length: 15000, dtype: int64
```

name. earor les, length. 15000, atype. In

In [14]:

Out[14]:

		Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Gender_female	Gender_male
	0	68	190	94	29	105	40.8	0	1
	1	20	166	60	14	94	40.3	1	0
	2	69	179	79	5	88	38.7	0	1
	3	34	179	71	13	100	40.5	1	0
	4	27	154	58	10	81	39.8	1	0
	•••								
	14995	20	193	86	11	92	40.4	1	0
	14996	27	165	65	6	85	39.2	1	0
	14997	43	159	58	16	90	40.1	1	0
	14998	78	193	97	2	84	38.3	0	1
	14999	63	173	79	18	92	40.5	0	1

15000 rows × 8 columns

```
In [15]: from sklearn.model_selection import train_test_split
In [16]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.22,random_state=49)
In [17]: x_test.head(10)
```

Out[17]:		Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Gender_female	Gender_male
	9944	55	183	83	24	106	40.7	0	1
	12	66	171	79	11	90	40.0	0	1
	5052	60	191	95	2	72	38.1	0	1
	14845	26	186	81	5	72	39.3	0	1
	12637	22	177	78	25	102	40.7	0	1
	9721	24	183	81	12	91	40.1	0	1
	14141	33	168	63	5	81	38.8	1	0
	2372	77	175	79	4	84	38.8	0	1
	431	23	173	72	28	110	40.9	1	0
	6357	59	189	95	22	110	40.5	0	1
In [18]:	x_trai	in.hea	ad(10)						
Out[18]:		Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Gender_female	Gender_male
	11805	28	193	97	13	91	40.0	0	1
	10315	22	145	43	9	77	39.9	1	0
	12700	63	162	58	19	107	40.5	1	0
	1557	78	170	78	14	103	40.1	0	1
	7973	29	160	55	5	85	39.3	1	0
	2827	38	158	63	21	103	40.6	1	0
	1785	39	169	62	29	113	41.0	1	0
					_				
	13914	34	175	65	5	85	38.7	1	0
	13914 11389	34 26	175 187	65 85	25	107	38.7 40.5	0	0
	11389	26	187	85	25	107	40.5	0	1
In [19]:	11389	26 25	187 181	85	25	107	40.5	0	1
In [19]: Out[19]:	y_trai 11805 10315 12700 1557 7973 2827 1785 13914 11389 2342	26 25 in hea 13 16 2 12 26 15	187 181 ad(10) 52 28 31 39 20 25 34 20 51	85	25 21	107	40.5	0	1

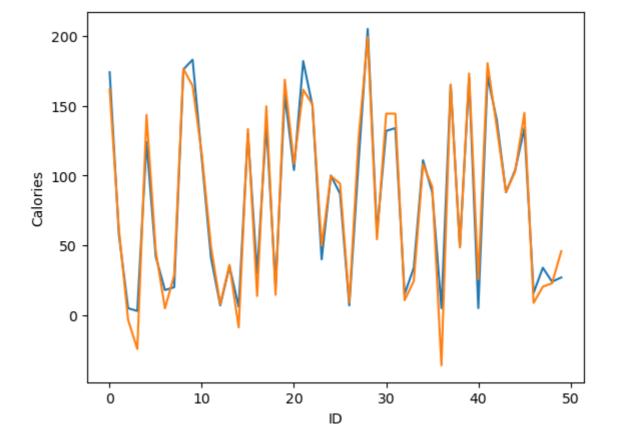
```
9944
                   174
Out[20]:
         12
                   58
         5052
                    5
         14845
                    3
         12637
                  124
         9721
                   42
         14141
                    18
         2372
                   20
         431
                   176
         6357
                   183
         Name: Calories, dtype: int64
In [21]:
         from sklearn.linear_model import LinearRegression
In [22]:
          reg=LinearRegression()
          reg.fit(x_train,y_train)
         LinearRegression()
Out[22]:
          ypred=reg.predict(x_test)
In [23]:
         ypred
In [24]:
         array([161.64451313, 61.70210153,
                                              -3.32685259, ...,
                                                                 64.76631124,
Out[24]:
                  88.71631251, 157.51493911])
          from sklearn.metrics import r2_score
In [25]:
          r2_score(y_test,ypred)
In [26]:
         0.966580087122947
Out[26]:
In [27]:
          from sklearn.metrics import mean_squared_error
In [28]:
         mean_squared_error(ypred,y_test)
         128.97933287284866
Out[28]:
In [29]:
         res=pd.DataFrame(columns=['Calories','predicted'])
          res['Calories']=y_test
          res['predicted']=ypred
          res=res.reset_index()
          res['ID']=res.index
In [30]:
         res.head(10)
```

Out[30]:		index	Calories	predicted	ID
	0	9944	174	161.644513	0
	1	12	58	61.702102	1
	2	5052	5	-3.326853	2
	3	14845	3	-24.238001	3
	4	12637	124	143.377735	4
	5	9721	42	46.007323	5
	6	14141	18	4.872376	6
	7	2372	20	28.560636	7
	8	431	176	176.460214	8
	9	6357	183	164.176525	9

```
In [31]: import seaborn as sns
import matplotlib.pyplot as plt

In [32]: sns.lineplot(x='ID',y='Calories',data=res.head(50))
sns.lineplot(x='ID',y='predicted',data=res.head(50))
plt.plot()
```

Out[32]: []



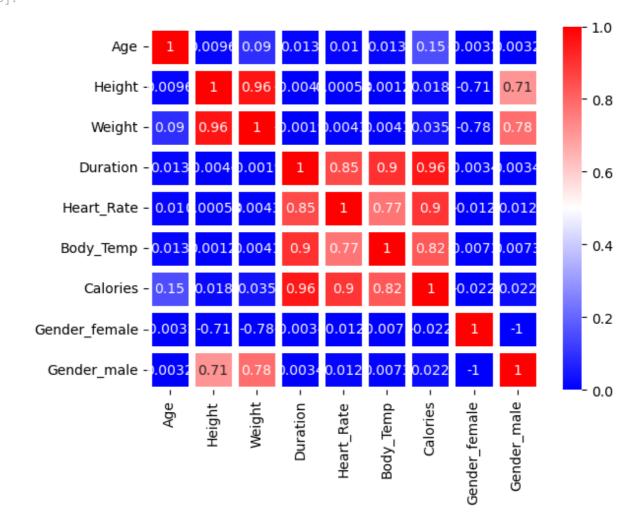
```
In [33]: new=[[22,175,80,25,75,43.5,1,0]]
In [34]: real=reg.predict(new)
    real
```

```
Out[34]: array([43.57603218])
```

In [35]: cor\_mat=b.corr()
 cor\_mat

Out[35]:		Age	Height	Weight	Duration	Heart_Rate	Body_Temp	Calories	Gen
	Age	1.000000	0.009554	0.090094	0.013247	0.010482	0.013175	0.154395	
	Height	0.009554	1.000000	0.958451	-0.004625	0.000528	0.001200	0.017537	
	Weight	0.090094	0.958451	1.000000	-0.001884	0.004311	0.004095	0.035481	
	Duration	0.013247	-0.004625	-0.001884	1.000000	0.852869	0.903167	0.955421	
	Heart_Rate	0.010482	0.000528	0.004311	0.852869	1.000000	0.771529	0.897882	
	Body_Temp	0.013175	0.001200	0.004095	0.903167	0.771529	1.000000	0.824558	
	Calories	0.154395	0.017537	0.035481	0.955421	0.897882	0.824558	1.000000	
	Gender_female	-0.003222	-0.710534	-0.783186	-0.003440	-0.011555	-0.007264	-0.022357	
	Gender_male	0.003222	0.710534	0.783186	0.003440	0.011555	0.007264	0.022357	

In [36]: sns.heatmap(cor\_mat,vmax=1,vmin=0,annot=True,linewidth=5,cmap='bwr')
Out[36]: <AxesSubplot:>



In [ ]: