In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

In [2]: s=pd.read\_csv(r"C:\Users\user\Downloads\16\_Sleep\_health\_and\_lifestyle\_dataset - 16\_Sleep\_health\_and\_lifestyle\_dataset.cs\
s

Out[2]:

	Person ID	Gender	Age	Occupation	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	BMI Category	Blood Pressure	Heart Rate	Daily Steps	Sleep Disorder
0	1	Male	27	Software Engineer	6.1	6	42	6	Overweight	126/83	77	4200	None
1	2	Male	28	Doctor	6.2	6	60	8	Normal	125/80	75	10000	None
2	3	Male	28	Doctor	6.2	6	60	8	Normal	125/80	75	10000	None
3	4	Male	28	Sales Representative	5.9	4	30	8	Obese	140/90	85	3000	Sleep Apnea
4	5	Male	28	Sales Representative	5.9	4	30	8	Obese	140/90	85	3000	Sleep Apnea
369	370	Female	59	Nurse	8.1	9	75	3	Overweight	140/95	68	7000	Sleep Apnea
370	371	Female	59	Nurse	8.0	9	75	3	Overweight	140/95	68	7000	Sleep Apnea
371	372	Female	59	Nurse	8.1	9	75	3	Overweight	140/95	68	7000	Sleep Apnea
372	373	Female	59	Nurse	8.1	9	75	3	Overweight	140/95	68	7000	Sleep Apnea
373	374	Female	59	Nurse	8.1	9	75	3	Overweight	140/95	68	7000	Sleep Apnea

374 rows × 13 columns

In [3]: s=s.head(100)
s

Out[3]:

	Person ID	Gender	Age	Occupation	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	BMI Category	Blood Pressure	Heart Rate	Daily Steps	Sleep Disorder
0	1	Male	27	Software Engineer	6.1	6	42	6	Overweight	126/83	77	4200	None
1	2	Male	28	Doctor	6.2	6	60	8	Normal	125/80	75	10000	None
2	3	Male	28	Doctor	6.2	6	60	8	Normal	125/80	75	10000	None
3	4	Male	28	Sales Representative	5,9	4	30	8	Obese	140/90	85	3000	Sleep Apnea
4	5	Male	28	Sales Representative	5.9	4	30	8	Obese	140/90	85	3000	Sleep Apnea
95	96	Female	36	Accountant	7.1	8	60	4	Normal	115/75	68	7000	None
96	97	Female	36	Accountant	7.2	8	60	4	Normal	115/75	68	7000	None
97	98	Female	36	Accountant	7.1	8	60	4	Normal	115/75	68	7000	None
98	99	Female	36	Teacher	7.1	8	60	4	Normal	115/75	68	7000	None
99	100	Female	36	Teacher	7.1	8	60	4	Normal	115/75	68	7000	None

100 rows × 13 columns

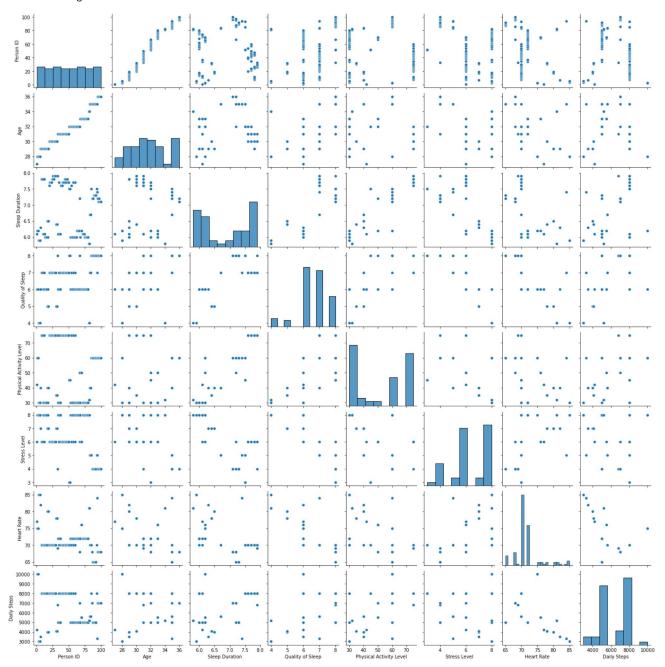
```
In [4]: s.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 100 entries, 0 to 99
         Data columns (total 13 columns):
          #
              Column
                                          Non-Null Count Dtype
          0
              Person ID
                                          100 non-null
                                                            int64
          1
              Gender
                                          100 non-null
                                                            object
          2
                                          100 non-null
                                                            int64
              Age
          3
              Occupation
                                          100 non-null
                                                            object
              Sleep Duration
          4
                                          100 non-null
                                                            float64
          5
              Quality of Sleep
                                          100 non-null
                                                            int64
              Physical Activity Level 100 non-null
                                                            int64
                                          100 non-null
              Stress Level
          7
                                                            int64
          8
              BMI Category
                                          100 non-null
                                                            object
          9
              Blood Pressure
                                          100 non-null
                                                            object
          10
              Heart Rate
                                          100 non-null
                                                            int64
              Daily Steps
                                          100 non-null
          11
                                                            int64
          12 Sleep Disorder
                                          100 non-null
                                                            object
         dtypes: float64(1), int64(7), object(5)
         memory usage: 10.3+ KB
In [5]: s.describe()
Out[5]:
                 Person ID
                                Age Sleep Duration Quality of Sleep Physical Activity Level Stress Level
                                                                                                  Heart Rate
                                                                                                               Daily Steps
         count 100.000000 100.00000
                                        100.000000
                                                       100.000000
                                                                           100.000000
                                                                                                  100.000000
                                                                                                               100.000000
                                                                                       100.000000
          mean
                 50.500000
                            31.69000
                                          6.871000
                                                         6.590000
                                                                            51.910000
                                                                                         6.420000
                                                                                                   71.610000
                                                                                                              6426.000000
            std
                 29.011492
                             2.26388
                                          0.766903
                                                         1.005992
                                                                            19.429279
                                                                                         1.485145
                                                                                                    4.240009
                                                                                                              1689.517294
                  1.000000
                            27.00000
                                          5.800000
                                                         4.000000
                                                                            30.000000
                                                                                         3.000000
                                                                                                   65.000000
                                                                                                              3000.000000
           min
           25%
                 25.750000
                            30.00000
                                          6.100000
                                                         6.000000
                                                                            30.000000
                                                                                         6.000000
                                                                                                   70.000000
                                                                                                              5000.000000
                                                                                                              7000.000000
                 50.500000
                            31.50000
                                          7.100000
                                                         7.000000
                                                                            60.000000
                                                                                         6.000000
                                                                                                   70.000000
           50%
```

75% 75.250000 33.00000 7.700000 7.000000 75.000000 8.000000 72.000000 8000.00000 max 100.000000 36.00000 7.900000 8.000000 75.000000 000000.8 85.000000 10000.000000 In [6]: s.columns

```
Out[6]: Index(['Person ID', 'Gender', 'Age', 'Occupation', 'Sleep Duration',
                     'Quality of Sleep', 'Physical Activity Level', 'Stress Level', 'BMI Category', 'Blood Pressure', 'Heart Rate', 'Daily Steps',
                      'Sleep Disorder'],
                    dtype='object')
```

In [7]: sns.pairplot(s)

Out[7]: <seaborn.axisgrid.PairGrid at 0x1bbe85fb850>

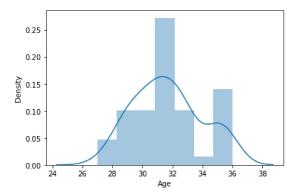


## In [8]: sns.distplot(s['Age'])

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: `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[8]: <AxesSubplot:xlabel='Age', ylabel='Density'>



In [9]: s1=s[['Person ID','Age','Sleep Duration','Quality of Sleep','Stress Level']]
s

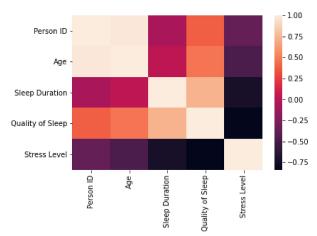
## Out[9]:

	Person ID	Gender	Age	Occupation	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	BMI Category	Blood Pressure	Heart Rate	Daily Steps	Sleep Disorder
0	1	Male	27	Software Engineer	6.1	6	42	6	Overweight	126/83	77	4200	None
1	2	Male	28	Doctor	6.2	6	60	8	Normal	125/80	75	10000	None
2	3	Male	28	Doctor	6.2	6	60	8	Normal	125/80	75	10000	None
3	4	Male	28	Sales Representative	5.9	4	30	8	Obese	140/90	85	3000	Sleep Apnea
4	5	Male	28	Sales Representative	5.9	4	30	8	Obese	140/90	85	3000	Sleep Apnea
					•••								
95	96	Female	36	Accountant	7.1	8	60	4	Normal	115/75	68	7000	None
96	97	Female	36	Accountant	7.2	8	60	4	Normal	115/75	68	7000	None
97	98	Female	36	Accountant	7.1	8	60	4	Normal	115/75	68	7000	None
98	99	Female	36	Teacher	7.1	8	60	4	Normal	115/75	68	7000	None
99	100	Female	36	Teacher	7.1	8	60	4	Normal	115/75	68	7000	None

100 rows × 13 columns

In [10]: sns.heatmap(s1.corr())

## Out[10]: <AxesSubplot:>



```
In [11]: x=s1[['Person ID','Age','Sleep Duration','Quality of Sleep','Stress Level']]
         y=s1['Stress Level']
In [12]: from sklearn.model_selection import train_test_split
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
In [13]: from sklearn.linear_model import LinearRegression
         lr=LinearRegression()
         lr.fit(x_train,y_train)
Out[13]: LinearRegression()
In [14]: | lr.intercept_
Out[14]: -5.5067062021407764e-14
In [15]: coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
         coeff
Out[15]:
                          Co-efficient
               Person ID -2.266715e-16
                   Age 2.789289e-15
           Sleep Duration -1.649334e-15
          Quality of Sleep -8.884300e-16
             Stress Level 1.000000e+00
In [16]: prediction=lr.predict(x_test)
         plt.scatter(y_test,prediction)
Out[16]: <matplotlib.collections.PathCollection at 0x1bbf1a08670>
                                      6
In [17]: print(lr.score(x_test,y_test))
         1.0
In [18]: from sklearn.linear_model import Ridge,Lasso
In [19]: from sklearn.linear_model import Ridge,Lasso
In [20]: rr=Ridge(alpha=10)
         rr.fit(x_train,y_train)
         rr.score(x_test,y_test)
Out[20]: 0.9781426348036008
In [21]: |la=Lasso(alpha=10)
         la.fit(x_train,y_train)
         la.score(x_test,y_test)
Out[21]: -0.03476371537208034
In [22]: from sklearn.linear_model import ElasticNet
         en=ElasticNet()
         en.fit(x_train,y_train)
Out[22]: ElasticNet()
```

```
In [23]: print(en.coef_)
         [-0.00428925 -0.
                                   -0.
                                               -0.
                                                            0.61743786]
In [24]: print(en.intercept_)
         2.7101852037236402
In [25]: print(en.predict(x_test))
         [7.62395257 4.79390399 6.18319277 6.17032501 6.32473807 5.43278811
          7.38375447 6.95933294 6.25182079 6.20034977 6.89928341 5.39847409
          7.63253107 6.38049834 6.11456474 7.34086196 6.41052311 4.33945769
          7.60250631\ 5.44136661\ 7.40091148\ 7.58106005\ 6.31187032\ 7.64110957
          6.11885399 5.43707736 4.75958998 6.21321753 7.63682032 4.76816848]
In [26]: print(en.score(x_test,y_test))
         0.8837501365595812
In [27]: from sklearn import metrics
In [28]: print("Mean Absolute Error", metrics.mean_absolute_error(y_test, prediction))
         Mean Absolute Error 1.1694349192718315e-15
In [29]: print("Mean squared Error", metrics.mean_squared_error(y_test, prediction))
         Mean squared Error 2.4520426470619784e-30
In [30]: print("Root Mean squared Error",np.sqrt(metrics.mean_squared_error(y_test,prediction)))
         Root Mean squared Error 1.5658999479730429e-15
In [ ]:
```