DATA COLLECTION

In [3]: # import libraries import numpy as np import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns

In [7]: # To Import Dataset

sd=pd.read_csv(r"c:\Users\user\Downloads\16_Sleep_health_and_lifestyle_dataset. sd

Out[7]:

	Person ID	Gender	Age	Occupation	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	BMI Category	Blo Pressi
0	1	Male	27	Software Engineer	6.1	6	42	6	Overweight	126/
1	2	Male	28	Doctor	6.2	6	60	8	Normal	125/
2	3	Male	28	Doctor	6.2	6	60	8	Normal	125/
3	4	Male	28	Sales Representative	5.9	4	30	8	Obese	140/
4	5	Male	28	Sales Representative	5.9	4	30	8	Obese	140/
				•••					•••	
369	370	Female	59	Nurse	8.1	9	75	3	Overweight	140/
370	371	Female	59	Nurse	8.0	9	75	3	Overweight	140/
371	372	Female	59	Nurse	8.1	9	75	3	Overweight	140/
372	373	Female	59	Nurse	8.1	9	75	3	Overweight	140/
373	374	Female	59	Nurse	8.1	9	75	3	Overweight	140/

374 rows × 13 columns

```
In [8]: # to display top 10 rows
sd.head(10)
```

Out[8]:

	Person ID	Gender	Age	Occupation	Sleep Duration	Quality of Sleep	Physical Activity Level	Stress Level	BMI Category	Blood Pressure
0	1	Male	27	Software Engineer	6.1	6	42	6	Overweight	126/83
1	2	Male	28	Doctor	6.2	6	60	8	Normal	125/80
2	3	Male	28	Doctor	6.2	6	60	8	Normal	125/80
3	4	Male	28	Sales Representative	5.9	4	30	8	Obese	140/90
4	5	Male	28	Sales Representative	5.9	4	30	8	Obese	140/90
5	6	Male	28	Software Engineer	5.9	4	30	8	Obese	140/90
6	7	Male	29	Teacher	6.3	6	40	7	Obese	140/90
7	8	Male	29	Doctor	7.8	7	75	6	Normal	120/80
8	9	Male	29	Doctor	7.8	7	75	6	Normal	120/80
9	10	Male	29	Doctor	7.8	7	75	6	Normal	120/80
4 0										•

DATA CLEANING AND PRE_PROCESSING

In [9]: sd.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 374 entries, 0 to 373
Data columns (total 13 columns):

	•	,	
#	Column	Non-Null Count	Dtype
0	Person ID	374 non-null	int64
1	Gender	374 non-null	object
2	Age	374 non-null	int64
3	Occupation	374 non-null	object
4	Sleep Duration	374 non-null	float64
5	Quality of Sleep	374 non-null	int64
6	Physical Activity Level	374 non-null	int64
7	Stress Level	374 non-null	int64
8	BMI Category	374 non-null	object
9	Blood Pressure	374 non-null	object
10	Heart Rate	374 non-null	int64
11	Daily Steps	374 non-null	int64
12	Sleep Disorder	374 non-null	object

dtypes: float64(1), int64(7), object(5)

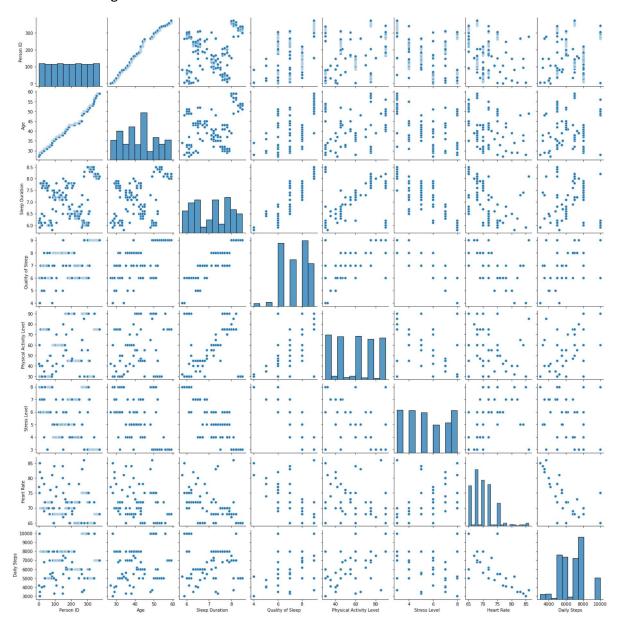
memory usage: 38.1+ KB

```
In [10]: # to display summary of statistics
           sd.describe()
Out[10]:
                                                                  Physical
                                              Sleep
                                                      Quality of
                                                                                Stress
                   Person ID
                                    Age
                                                                   Activity
                                                                                       Heart Rate
                                                                                                    Da
                                           Duration
                                                         Sleep
                                                                                Level
                                                                     Level
            count 374.000000
                              374.000000
                                         374.000000
                                                    374.000000
                                                                374.000000 374.000000
                                                                                       374.000000
                                                                                                    37
            mean 187.500000
                               42.184492
                                           7.132086
                                                       7.312834
                                                                 59.171123
                                                                              5.385027
                                                                                        70.165775
                                                                                                   681
              std 108.108742
                               8.673133
                                           0.795657
                                                       1.196956
                                                                 20.830804
                                                                             1.774526
                                                                                         4.135676
                                                                                                   161
                    1.000000
                               27.000000
                                           5.800000
                                                       4.000000
                                                                 30.000000
                                                                             3.000000
                                                                                        65.000000
                                                                                                   300
             min
             25%
                   94.250000
                               35.250000
                                           6.400000
                                                       6.000000
                                                                 45.000000
                                                                             4.000000
                                                                                        68.000000
                                                                                                   560
                                                                                                   700
             50%
                  187.500000
                                           7.200000
                                                       7.000000
                                                                 60.000000
                                                                              5.000000
                                                                                        70.000000
                               43.000000
             75%
                  280.750000
                               50.000000
                                           7.800000
                                                       8.000000
                                                                 75.000000
                                                                              7.000000
                                                                                        72.000000
                                                                                                   800
             max 374.000000
                               59.000000
                                           8.500000
                                                       9.000000
                                                                 90.000000
                                                                              8.000000
                                                                                                  1000
                                                                                        86.000000
          #to display colums heading
In [11]:
           sd.columns
Out[11]: 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')
```

EDA and visualization

In [12]: sns.pairplot(sd)

Out[12]: <seaborn.axisgrid.PairGrid at 0x22087112f70>

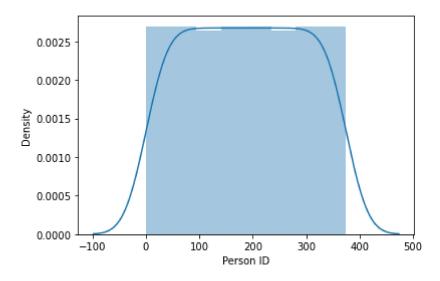


In [14]: sns.distplot(sd['Person ID'])

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: Fut ureWarning: `distplot` is a deprecated function and will be removed in a futu re version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

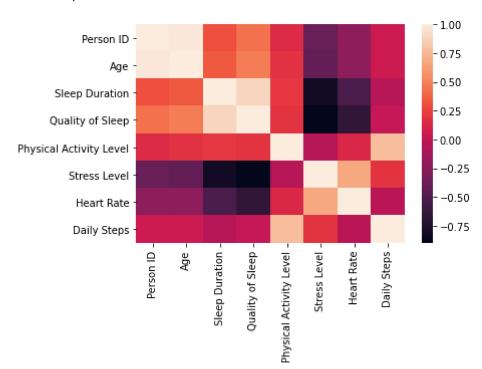
warnings.warn(msg, FutureWarning)

Out[14]: <AxesSubplot:xlabel='Person ID', ylabel='Density'>



In [15]: | sns.heatmap(sd.corr())

Out[15]: <AxesSubplot:>



TO TRAIN THE MODEL _MODEL BUILDING

we are goint train Liner Regression model; we need to split out the data into two varibles x and y where x is independent on x (output) and y is dependent on x(output) adress coloumn as it is not required our model

```
In [18]: x= sd1[['Person ID', 'Age', 'Sleep Duration',
                  'Quality of Sleep', 'Physical Activity Level']]
         y=sd1[ 'Stress Level']
In [19]: # To split my dataset into training data and test data
         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 [20]: | from sklearn.linear_model import LinearRegression
         lr=LinearRegression()
         lr.fit(x train,y train)
Out[20]: LinearRegression()
In [21]: from sklearn.linear model import LinearRegression
         lr=LinearRegression()
         lr.fit(x_train,y_train)
Out[21]: LinearRegression()
In [22]: |print(lr.intercept_)
         13.367928432947835
         coeff= pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
In [23]:
         coeff
Out[23]:
                              Co-efficient
                    Person ID
                               -0.006654
                                0.073910
                         Age
                Sleep Duration
                               -0.212024
                Quality of Sleep
                               -1.229610
          Physical Activity Level
                                0.010932
```

```
In [24]: | prediction = lr.predict(x_test)
         plt.scatter(y_test,prediction)
Out[24]: <matplotlib.collections.PathCollection at 0x2208c8bd700>
          9
          8
          7
          6
          5
          4
          3
                              5
                                      6
In [25]: print(lr.score(x_test,y_test))
         0.8363845510278012
In [26]: |lr.score(x_train,y_train)
Out[26]: 0.8299925338176495
In [27]: from sklearn.linear_model import Ridge,Lasso
In [28]: dr=Ridge(alpha=10)
         dr.fit(x_train,y_train)
Out[28]: Ridge(alpha=10)
In [29]: |dr.score(x_test,y_test)
Out[29]: 0.8341848701704608
In [30]: | dr.score(x_train,y_train)
Out[30]: 0.8288171569244359
In [31]: la=Lasso(alpha=10)
         la.fit(x_train,y_train)
Out[31]: Lasso(alpha=10)
In [32]: la.score(x_test,y_test)
Out[32]: 0.05571687468560049
```

In [33]:	<pre>la.score(x_train,y_train)</pre>
Out[33]:	0.19568057920856774
In []:	
In []:	