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

In [2]: df=pd.read_csv("4_drug.csv")
df

Out[2]:

	Age	Sex	BP	Cholesterol	Na_to_K	Drug
0	23	F	HIGH	HIGH	25.355	drugY
1	47	М	LOW	HIGH	13.093	drugC
2	47	М	LOW	HIGH	10.114	drugC
3	28	F	NORMAL	HIGH	7.798	drugX
4	61	F	LOW	HIGH	18.043	drugY
195	56	F	LOW	H I GH	11.567	drugC
196	16	М	LOW	HIGH	12.006	drugC
197	52	М	NORMAL	HIGH	9.894	drugX
198	23	М	NORMAL	NORMAL	14.020	drugX
199	40	F	LOW	NORMAL	11.349	drugX

200 rows × 6 columns

In [3]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 6 columns):

#	Column	Non-Null Count	Dtype			
0	Age	200 non-null	int64			
1	Sex	200 non-null	object			
2	BP	200 non-null	object			
3	Cholesterol	200 non-null	object			
4	Na_to_K	200 non-null	float64			
5	Drug	200 non-null	object			
<pre>dtypes: float64(1), int64(1), object(4)</pre>						
memory usage: 9.5+ KB						

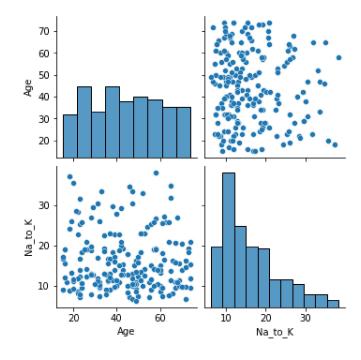
In [4]: df.describe()

Out[4]:

	Age	Na_to_K
count	200.000000	200.000000
mean	44.315000	16.084485
std	16.544315	7.223956
min	15.000000	6.269000
25%	31.000000	10.445500
50%	45.000000	13.936500
75%	58.000000	19.380000
max	74.000000	38.247000

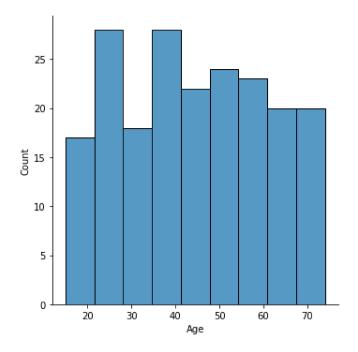
In [5]: sns.pairplot(df)

Out[5]: <seaborn.axisgrid.PairGrid at 0x1bed1a34af0>



In [6]: sns.displot(df['Age'])

Out[6]: <seaborn.axisgrid.FacetGrid at 0x1bed1c53bb0>



In [7]: df1=df.drop(['Sex'],axis=1)
 df1

Out[7]:

	Age	ВР	Cholesterol	Na_to_K	Drug
0	23	HIGH	HIGH	25.355	drugY
1	47	LOW	HIGH	13.093	drugC
2	47	LOW	H I GH	10.114	drugC
3	28	NORMAL	HIGH	7.798	drugX
4	61	LOW	H I GH	18.043	drugY
195	56	LOW	HIGH	11.567	drugC
196	16	LOW	HIGH	12.006	drugC
197	52	NORMAL	HIGH	9.894	drugX
198	23	NORMAL	NORMAL	14.020	drugX
199	40	LOW	NORMAL	11.349	drugX

200 rows × 5 columns

```
Untitled7 - Jupyter Notebook
 In [8]: sns.heatmap(df1.corr())
 Out[8]: <AxesSubplot:>
                                                       -1.0
                                                       - 0.8
           Age
                                                       -0.6
                                                       - 0.4
                                                       - 0.2
           Na to K
                                                        0.0
                     Age
                                       Na_to_K
 In [9]: from sklearn.model_selection import train_test_split
          from sklearn.linear_model import LinearRegression
In [10]: | y=df['Age']
          x=df1.drop(['Age','BP','Cholesterol','Drug'],axis=1)
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
          print(x_train)
               Na_to_K
          62
                20.693
                11.227
          156
          187
                10.403
                14.239
          41
          118
                10.292
          193
                 6.769
          71
                19.675
          93
                29.271
          57
                27.826
          195
                11.567
          [140 rows x 1 columns]
In [11]:
          model=LinearRegression()
          model.fit(x_train,y_train)
          model.intercept_
Out[11]: 47.90162037012079
In [12]: | coeff=pd.DataFrame(model.coef_,x.columns,columns=["Coefficient"])
          coeff
```

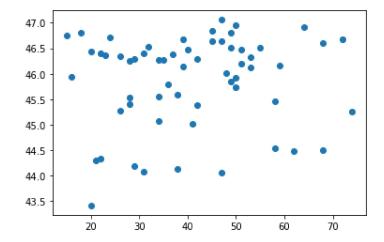
```
Coefficient
```

Out[12]:

-0.125986 Na_to_K

```
In [13]: prediction=model.predict(x_test)
plt.scatter(y_test,prediction)
```

Out[13]: <matplotlib.collections.PathCollection at 0x1bed360dee0>



```
In [14]: model.score(x_test,y_test)
```

Out[14]: -0.11864238220653989

```
In [15]: from sklearn.linear_model import Ridge,Lasso
```

```
In [16]: rr = Ridge(alpha=10)
    rr.fit(x_train,y_train)
```

Out[16]: Ridge(alpha=10)

```
In [17]: rr.score(x_test,y_test)
```

Out[17]: -0.11864770796837432

```
In [18]: la = Lasso(alpha=10)
la.fit(x_train,y_train)
```

Out[18]: Lasso(alpha=10)

```
In [19]: la.score(x_test,y_test)
```

Out[19]: -0.12627217665641743

```
In [20]: from sklearn.linear_model import ElasticNet
         en=ElasticNet()
         en.fit(x_train,y_train)
         print(en.coef_)
         print(en.intercept )
         print(en.predict(x_test))
         print(en.score(x test,y test))
         from sklearn import metrics
         print("Mean Absolute Error:",metrics.mean absolute error(y test,prediction))
         print("Mean Squared Error:",metrics.mean squared error(y test,prediction))
         print("Root Mean Squared Error:",np.sqrt(metrics.mean squared error(y test,prediction))
         [-0.1153991]
         47.73161475153885
         [46.31682177 46.46487882 44.22740564 44.61006906 45.08839834 46.57566195
          46.12318208 46.36171202 46.76572427 45.99647387 45.90946294 46.86727548
          46.5440426 46.42264275 46.10067925 46.61489765 44.6568057 46.31139801
          45.4910258 46.16253317 46.68332932 45.85349438 45.56107306 45.9410823
          46.28970298 45.57330536 46.24365874 43.61890618 46.63820827 44.59472098
          46.35224929 46.72187262 46.96040256 44.2840666 45.74548082 46.83600233
          45.44717414 46.12941363 46.21677075 44.20409503 46.46060905 46.569892
          46.42195035 46.24804391 46.61120488 46.48068849 45.31873494 45.6203882
          44.42750769 44.33311122 45.14021254 45.79833361 46.25842983 46.38306085
          46.72764257 46.24031217 45.42213254 44.46651258 46.34613314 45.31492677]
         -0.11899860556912256
         Mean Absolute Error: 12.810936451640638
         Mean Squared Error: 240.64483513223132
         Root Mean Squared Error: 15.512731388515412
```