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```
In [8]: #to import Libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [9]: #to import dataset
data=pd.read_csv(r"C:\Users\user\Downloads\4_drug200 - 4_drug200.csv")
data
```

```
Out[9]:
```

	Age	Sex	BP	Cholesterol	Na_to_K	Drug
0	23	F	HIGH	HIGH	25.355	drugY
1	47	M	LOW	HIGH	13.093	drugC
2	47	M	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	HIGH	11.567	drugC
196	16	M	LOW	HIGH	12.006	drugC
197	52	M	NORMAL	HIGH	9.894	drugX
198	23	M	NORMAL	NORMAL	14.020	drugX
199	40	F	LOW	NORMAL	11.349	drugX

200 rows × 6 columns

```
In [10]: data.head()
```

```
Out[10]:
```

	Age	Sex	BP	Cholesterol	Na_to_K	Drug
0	23	F	HIGH	HIGH	25.355	drugY
1	47	M	LOW	HIGH	13.093	drugC
2	47	M	LOW	HIGH	10.114	drugC
3	28	F	NORMAL	HIGH	7.798	drugX
4	61	F	LOW	HIGH	18.043	drugY

DATA CLEANING AND PREPROCESSING

```
In [11]: data.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  
dtypes: float64(1), int64(1), object(4)
memory usage: 9.5+ KB
```

```
In [12]: data.describe()
```

```
Out[12]:
```

	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

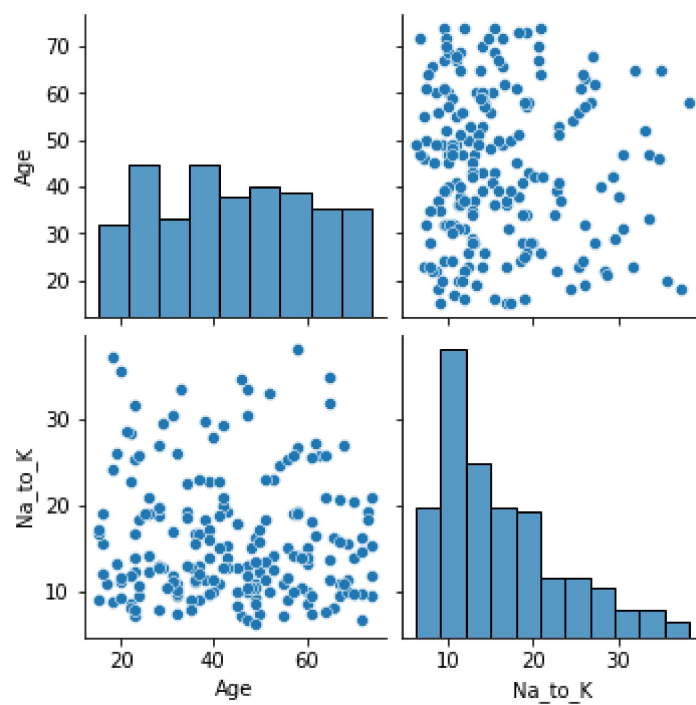
```
In [13]: data.columns
```

```
Out[13]: Index(['Age', 'Sex', 'BP', 'Cholesterol', 'Na_to_K', 'Drug'], dtype='object')
```

EDA and DATA VISUALIZATION

```
In [14]: sns.pairplot(data)
```

```
Out[14]: <seaborn.axisgrid.PairGrid at 0x28d057fa3d0>
```

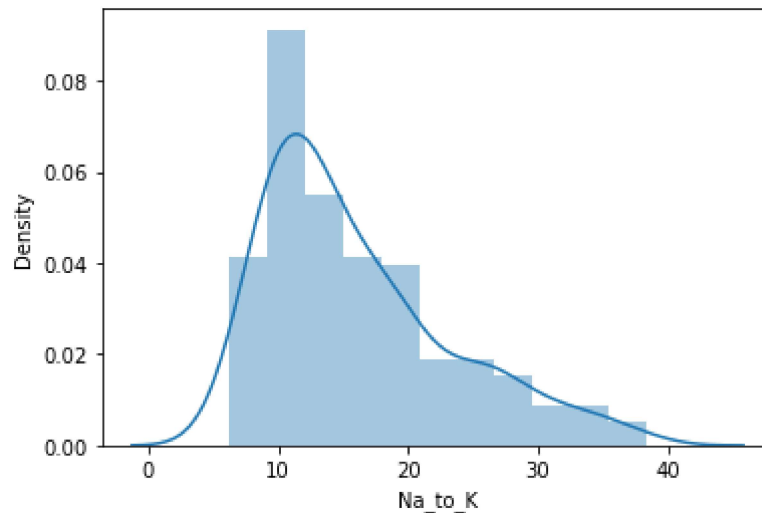


```
In [15]: sns.distplot(data['Na_to_K'])
```

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)
```

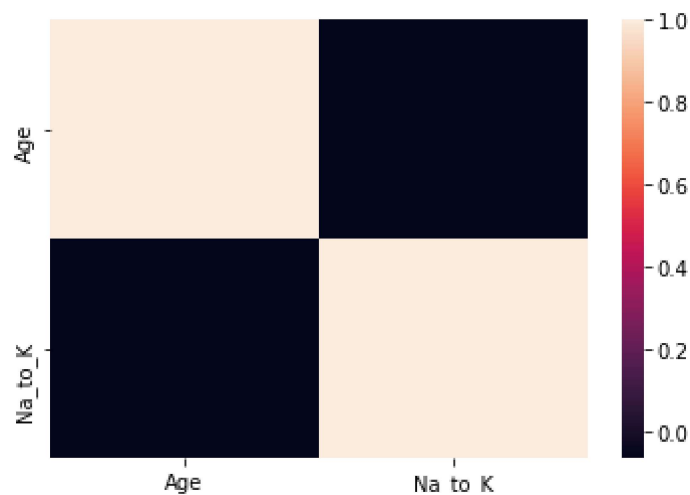
```
Out[15]: <AxesSubplot:xlabel='Na_to_K', ylabel='Density'>
```



```
In [16]: df=data[['Age', 'Sex', 'BP', 'Cholesterol', 'Na_to_K', 'Drug']]
```

```
In [17]: sns.heatmap(df.corr())
```

```
Out[17]: <AxesSubplot:>
```



TRAINING MODEL

```
In [18]: x=df[['Age']]  
y=df[['Na_to_K']]
```

```
In [19]: #to split my dataset into training and test  
  
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)
```

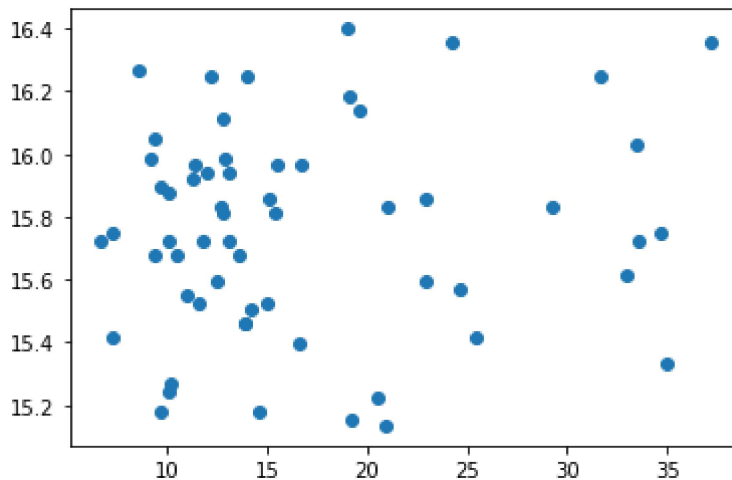
Out[20]: LinearRegression()

```
In [21]: #to find intercept  
print(lr.intercept_)
```

[16.74885176]

```
In [22]: prediction = lr.predict(x_test)  
plt.scatter(y_test,prediction)
```

Out[22]: <matplotlib.collections.PathCollection at 0x28d05c025b0>



```
In [23]: print(lr.score(x_test,y_test))
```

-0.009955722953167712

RIDGE AND LASSO REGRESSION

```
In [24]: from sklearn.linear_model import Ridge,Lasso
```

```
In [25]: rr=Ridge(alpha=10)
rr.fit(x_train,y_train)
```

```
Out[25]: Ridge(alpha=10)
```

```
In [26]: rr.score(x_test,y_test)
```

```
Out[26]: -0.009956517886995808
```

```
In [27]: la=Lasso(alpha=10)
la.fit(x_train,y_train)
```

```
Out[27]: Lasso(alpha=10)
```

```
In [28]: la.score(x_test,y_test)
```

```
Out[28]: -0.01491531607825447
```

```
In [29]: from sklearn.linear_model import ElasticNet
en=ElasticNet()
en.fit(x_train,y_train)
```

```
Out[29]: ElasticNet()
```

```
In [30]: print(en.coef_)
```

```
[-0.02006718]
```

```
In [31]: print(en.predict(x_test))
```

```
[15.80946899 16.21081255 15.72920028 15.72920028 16.09040948 15.82953617
15.68906592 15.56866285 16.31114844 16.15061101 16.23087973 15.60879721
15.84960334 15.86967052 15.72920028 15.54859567 16.03020795 15.48839414
15.60879721 15.44825978 15.92987206 15.30778954 15.72920028 16.01014077
15.28772236 15.36799107 15.5285285 15.90980488 15.22752082 15.58873003
15.74926745 15.62886439 15.44825978 15.20745365 15.94993923 15.82953617
15.82953617 16.21081255 15.68906592 16.11047666 15.92987206 15.8897377
16.21081255 15.54859567 16.31114844 15.68906592 15.94993923 15.18738647
15.84960334 15.26765518 15.80946899 15.42819261 15.94993923 15.22752082
15.74926745 15.72920028 15.48839414 15.97000641 15.97000641 16.35128279]
```

```
In [32]: print(en.score(x_test,y_test))
```

```
-0.010226915185498786
```

Evaluation metrics

```
In [33]: from sklearn import metrics
```

```
In [34]: print("Mean Absolute error",metrics.mean_absolute_error(y_test,prediction))
```

Mean Absolute error 6.1202398832739195

```
In [35]: print("Mean Squared error",metrics.mean_squared_error(y_test,prediction))
```

Mean Squared error 63.86700838972528

```
In [36]: print("Root Mean Absolute error",np.sqrt(metrics.mean_squared_error(y_test,prediction)))
```

Root Mean Absolute error 7.991683701806853