

D3

```
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(r"C:\Users\user\Downloads\4_drug200.csv")
df
```

Out[2]:

| | 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 [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
dtypes: float64(1), int64(1), object(4)
memory usage: 9.5+ KB
```

```
In [4]: df.describe()
```

```
Out[4]:
```

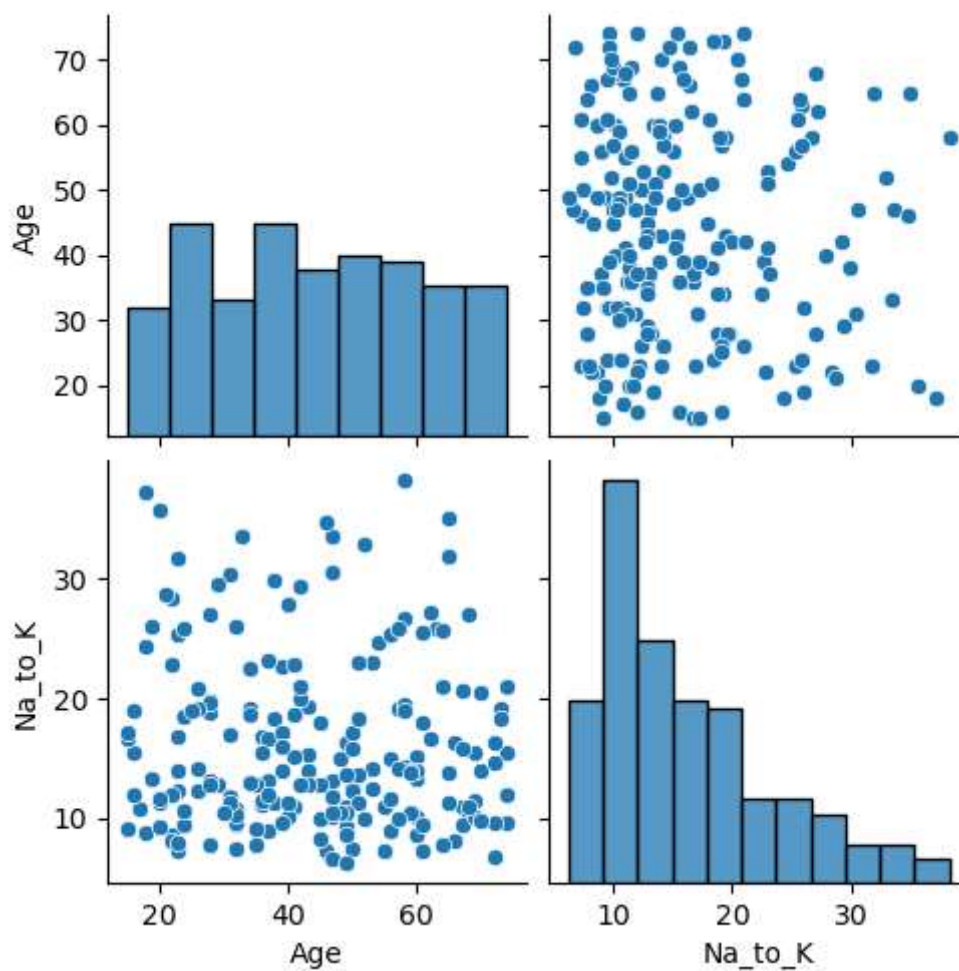
| | Age | Na_to_K |
|-------|------------|------------|
| count | 200.000000 | 200.000000 |
| mean | 44.315000 | 16.084485 |
| std | 16.544315 | 7.223956 |
| min | 15.000000 | 6.269000 |
| 25% | 31.000000 | 10.445500 |
| 50% | 45.000000 | 13.936500 |
| 75% | 58.000000 | 19.380000 |
| max | 74.000000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

```
In [5]: df.columns
```

```
Out[5]: Index(['Age', 'Sex', 'BP', 'Cholesterol', 'Na_to_K', 'Drug'], dtype='object')
```

```
In [6]: sns.pairplot(df)
```

```
Out[6]: <seaborn.axisgrid.PairGrid at 0x1eb144434d0>
```



```
In [7]: sns.distplot(df["Age"])
```

C:\Users\user\AppData\Local\Temp\ipykernel_7792\2732350774.py:1: UserWarning:

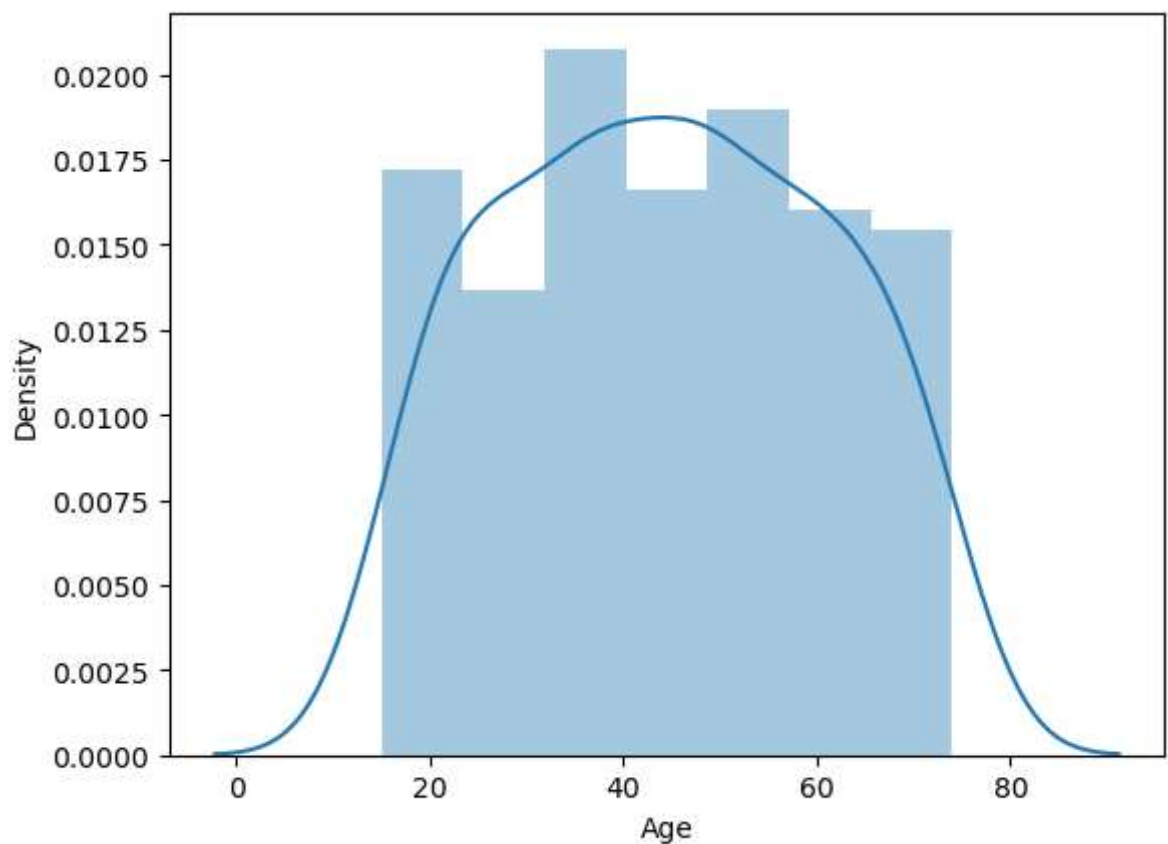
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(df["Age"])
```

Out[7]: <Axes: xlabel='Age', ylabel='Density'>

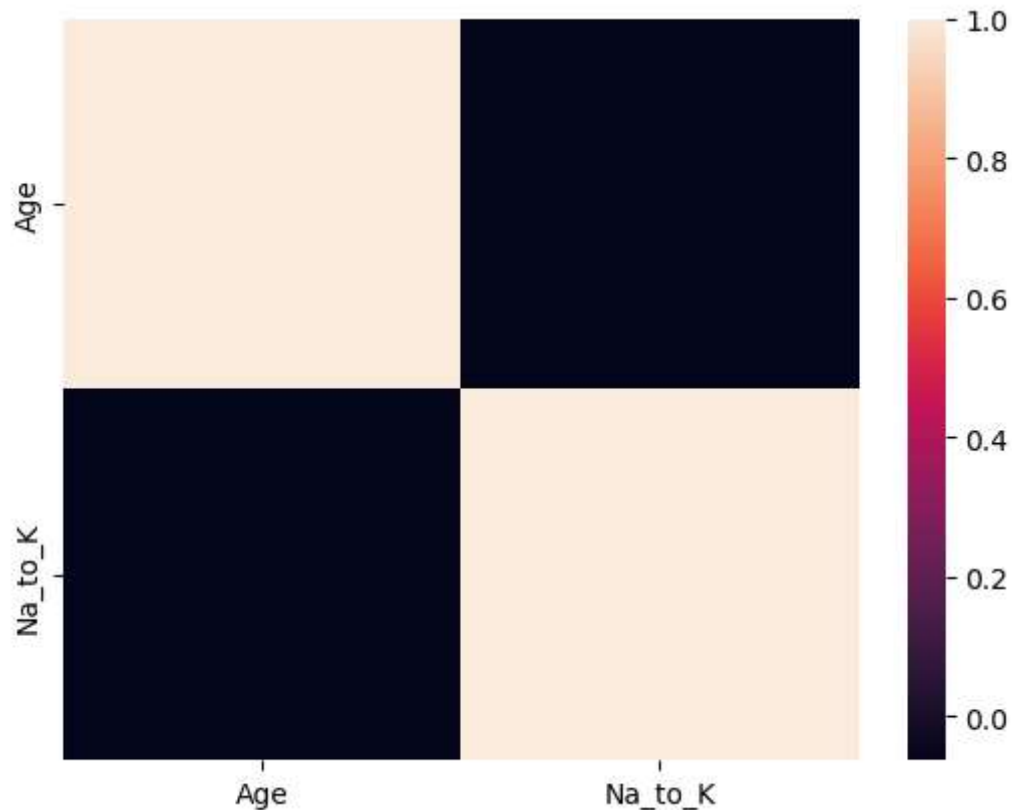


```
In [8]: df1=df[['Age', 'Sex', 'BP', 'Cholesterol', 'Na_to_K', 'Drug']]
```

In [9]: `sns.heatmap(df1.corr())`

C:\Users\user\AppData\Local\Temp\ipykernel_7792\781785195.py:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.
`sns.heatmap(df1.corr())`

Out[9]: <Axes: >



In [10]: `x=df1[['Age']]`
`y=df1['Na_to_K']`

In [11]: `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 [12]: `from sklearn.linear_model import LinearRegression`
`lr=LinearRegression()`
`lr.fit(x_train,y_train)`

Out[12]: `LinearRegression`
`LinearRegression()`

In [13]: `print(lr.intercept_)`
15.435320212781702

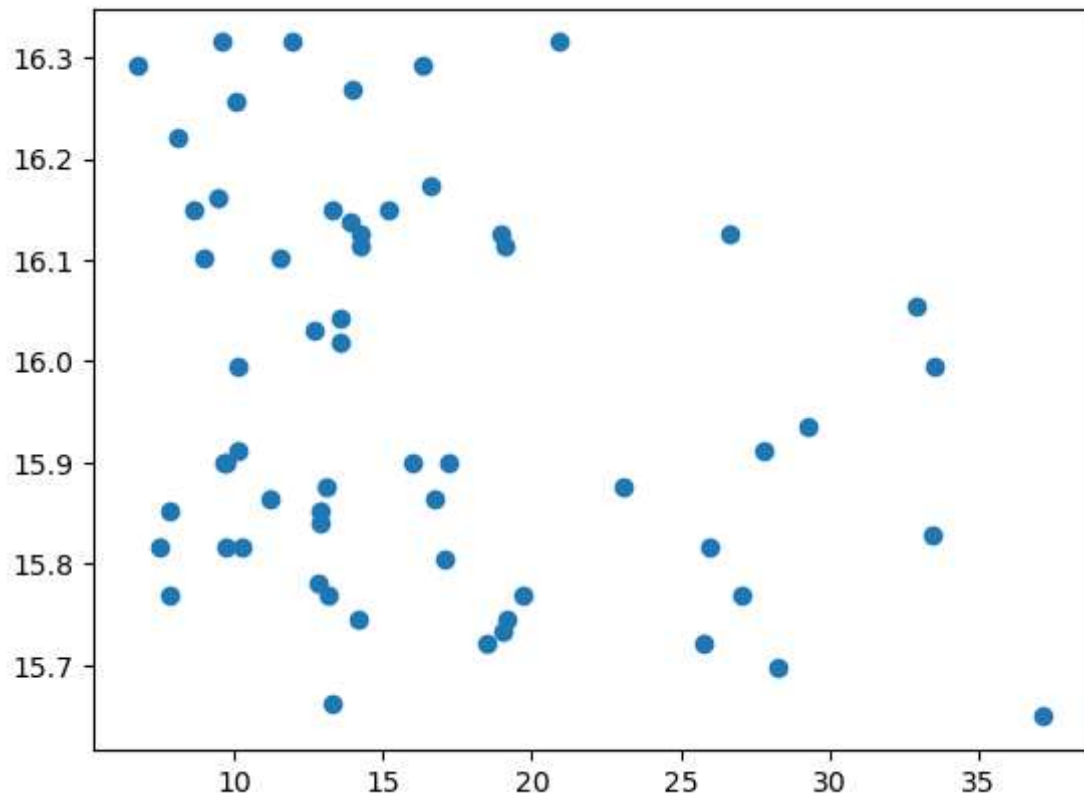
```
In [14]: coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
coeff
```

Out[14]:

| | Co-efficient |
|-----|--------------|
| Age | 0.011889 |

```
In [15]: prediction=lr.predict(x_test)
plt.scatter(y_test,prediction)
```

Out[15]: <matplotlib.collections.PathCollection at 0x1eb0f056250>



```
In [16]: print(lr.score(x_test,y_test))
```

-0.017709603394363116

```
In [17]: from sklearn.linear_model import Ridge,Lasso
```

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

Out[18]:

```

Ridge
Ridge(alpha=10)
```

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

```
Out[19]: -0.017705688018912147
```

```
In [20]: la=Lasso(alpha=10)  
la.fit(x_train,y_train)
```

```
Out[20]: 

▼



Lasso



Lasso(alpha=10)


```

```
In [21]: la.score(x_test,y_test)
```

```
Out[21]: -0.0030546432738367546
```

```
In [ ]:
```