```
1=[15,24,6,37,50]
In [1]:
         import pandas as pd
In [2]:
         l=pd.DataFrame(1)
In [3]:
         print(1.std())
              17,472836
         dtype: float64
In [4]: data=pd.read_csv(r"C:\Users\DELL\Downloads\Feature Scaling Resource16994544271.csv")
In [5]:
         data
Out[5]:
              Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome
           0
                       6
                                                                  0 33.6
                                                                                            0.627
                                                                                                   50
                                            72
                              148
                                                          35
                                                                                                              1
           1
                       1
                              85
                                             66
                                                                  0 26.6
                                                                                            0.351
                                                                                                   31
                                                                                                              0
           2
                       8
                              183
                                            64
                                                           0
                                                                  0 23.3
                                                                                            0.672
                                                                                                    32
                                                                                                              1
           3
                       1
                              89
                                                          23
                                                                  94 28.1
                                                                                            0.167
                                                                                                   21
                                             66
                                                                                                              0
           4
                                            40
                                                          35
                                                                                                    33
                       0
                              137
                                                                 168 43.1
                                                                                            2.288
                                                                                                              1
                             101
                                                                180 32.9
                                                                                                    63
         763
                      10
                                            76
                                                          48
                                                                                            0.171
                                                                                                              0
                                                                                                   27
         764
                       2
                             122
                                            70
                                                          27
                                                                  0 36.8
                                                                                            0.340
                                                                                                              0
         765
                       5
                                            72
                                                                 112 26.2
                                                                                            0.245
                                                                                                   30
                                                                                                              0
                              121
                                                          23
         766
                       1
                             126
                                            60
                                                           0
                                                                  0 30.1
                                                                                            0.349
                                                                                                    47
                                                                                                              1
                                                                  0 30.4
                                                                                            0.315
         767
                       1
                              93
                                            70
                                                          31
                                                                                                    23
                                                                                                              0
        768 rows × 9 columns
```

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data.head()

In [6]:

ıt[6]:	P	regnancies	Glucos	e BloodPres	sure S	kinThickness	Insulin	ВМІ	Diabete	esPedigreeFu	ınction	Age	Outcome		
	0	6	14	8	72	35	0	33.6			0.627	50	1		
	1	1	8	5	66	29	0	26.6			0.351	31	0		
	2	8	18	3	64	0	Ü	23.3			0.672	32	1		
	3	1	8	9	66	23	94	28.1			0.167	21	0		
	4	0	13	7	40	35	168	43.1			2.288	33	1		
[7]:	data	a.isnull()	any .												
7]:	<box< td=""><td></td><td>l NDFrai 1I \</td><td>neadd_num</td><td>eric_o</td><td>perations.<</td><td>locals</td><td>·.any</td><td>of</td><td>Pregnanc</td><td>ies (</td><td>Glucos</td><td>e BloodP</td><td>ressure</td><td>SkinThicknes</td></box<>		l NDFrai 1I \	neadd_num	eric_o	perations.<	locals	·.any	of	Pregnanc	ies (Glucos	e BloodP	ressure	SkinThicknes
	0		ılse	False	F	alse	Fals	se	False	False					
	1		lse	False		alse	Fals			False					
	2		lse	False		alse	Fals			False					
	3	Fa	lse	False	F	alse	Fals	se	False	False					
	4	Fa	lse	False	F	alse	Fals	se	False	False					
										• • •					
	763		lse	False		alse	Fals	se		False					
	764	Fa	ılse	False	F	alse	Fals	se	False	False					
	765	Fa	ılse	False	F	alse	Fals	se		False					
	766	Fa	lse	False	F	alse	Fals	se	False	False					
	767	Fa	ılse	False	F	alse	Fals	se	False	False					
		Diabetes	Pedigr	eeFunction	Age	Outcome									
	0			False	False	e False									
	1			False	False	e False									
	2			False	False	e False									
	3			False	False	e False									
	4			False	False	e False									
	• •														
	763				False										
	764				False										
	765				False										
	766				False										
	767			False	False	e False									
	[768	3 rows x 9	colum	ns]>											
8]:	data	a.dtypes													

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Feature Scaling Resource

```
Pregnancies
                                      int64
Out[8]:
        Glucose
                                      int64
        BloodPressure
                                      int64
        SkinThickness
                                      int64
        Insulin
                                      int64
        BMI
                                    float64
        DiabetesPedigreeFunction
                                    float64
        Age
                                      int64
        Outcome
                                      int64
        dtype: object
        data.info()
In [9]:
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 768 entries, 0 to 767
        Data columns (total 9 columns):
                                       Non-Null Count Dtype
             Column
                                        _____
             _____
                                        768 non-null
             Pregnancies
                                                        int64
         1
             Glucose
                                       768 non-null
                                                        int64
             BloodPressure
                                       768 non-null
                                                       int64
             SkinThickness
                                       768 non-null
                                                       int64
             Insulin
                                       768 non-null
                                                       int64
         5
             BMI
                                       768 non-null
                                                       float64
             DiabetesPedigreeFunction 768 non-null
                                                       float64
                                        768 non-null
         7
                                                       int64
             Age
         8
             Outcome
                                        768 non-null
                                                       int64
        dtypes: float64(2), int64(7)
        memory usage: 54.1 KB
```

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In [10]:

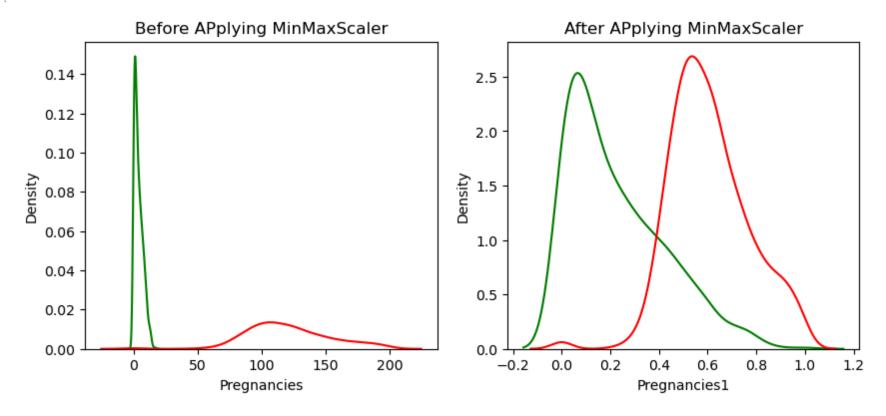
data.describe()

Out[10]:		Pregnancies Glucose		BloodPressure SkinThickness		Insulin	ВМІ	DiabetesPedigreeFunction	Age	Outcome			
	count	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000			
	mean	3.845052	120.894531	69.105469	20.536458	79.799479	31.992578	0.471876	33.240885	0.348958			
	std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	0.331329	11.760232	0.476951			
	min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.078000	21.000000	0.000000			
	25%	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000	0.243750	24.000000	0.000000			
	50%	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000	0.372500	29.000000	0.000000			
	75%	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000	0.626250	41.000000	1.000000			
	max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	2.420000	81.000000	1.000000			
n [11]:	from	sklearn.prep	orocessing	<pre>import MinMax</pre>	xScaler								
n [49]:	<pre>scaler1=MinMaxScaler()</pre>												
n [50]:	<pre>data[["Pregnancies1","Glucose1"]]=scaler1.fit_transform(data[["Pregnancies","Glucose"]])</pre>												
n [13]:	<pre>scaler1.fit_transform(data[["Pregnancies","Glucose"]])</pre>												
ut[13]:	array([[0.35294118, 0.74371859],												
n [14]:	data[ˈ	"Pregnancies	s"].min()										
ut[14]:	0												
n [15]:	data['	"Pregnancies	s"].max()										
ut[15]:	17												
n [16]:	data[d	data["Pregna	ancies"]>15	5]									

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Out[16]:	Pregnancies		Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction	Age	Outcome	e	
	159	17	163	72	41	114	40.9	0.817	47		1	
[17]:	_	matplot1 seaborn		ot as plt								
51]:	<pre>fig,(ax1,ax2)=plt.subplots(ncols=2,figsize=(10,4))</pre>											
	<pre>ax1.set_title("Before APplying MinMaxScaler") sns.kdeplot(data["Pregnancies"],ax=ax1,color="green") sns.kdeplot(data["Glucose"],ax=ax1,color="red")</pre>											
	<pre>ax2.set_title("After APplying MinMaxScaler") sns.kdeplot(data["Pregnancies1"],ax=ax2,color="green") sns.kdeplot(data["Glucose1"],ax=ax2,color="red")</pre>											
	<axes:< td=""><td>title={'</td><td>center':</td><td>'After APply</td><td>ving MinMaxSca</td><td>aler'}.</td><td>xlab</td><td>el='Pregnancies1'. vlab</td><td>el='[</td><td>Density'</td><td>></td></axes:<>	title={'	center':	'After APply	ving MinMaxSca	aler'}.	xlab	el='Pregnancies1'. vlab	el='[Density'	>	

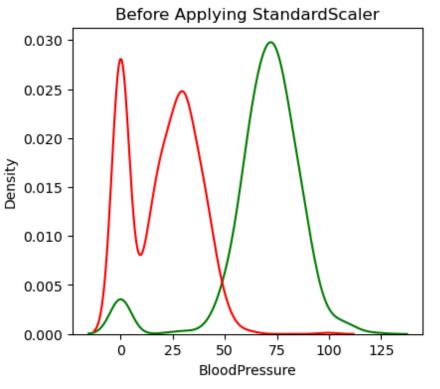


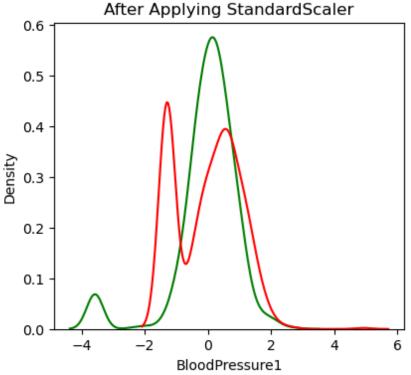


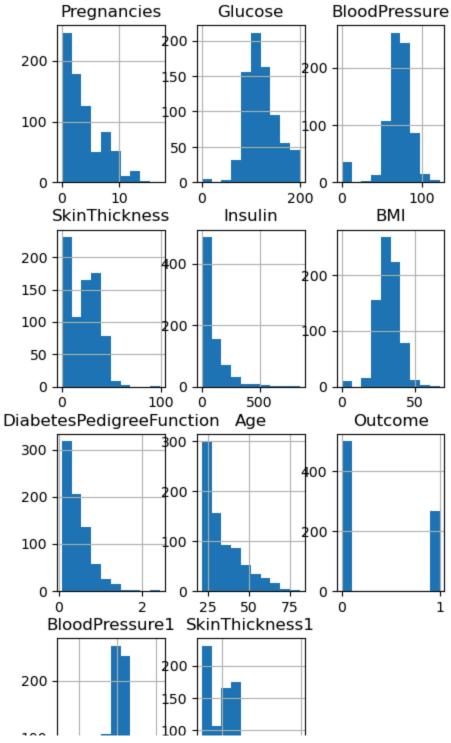
```
from sklearn.preprocessing import StandardScaler
In [21]:
          scaler2=StandardScaler()
In [22]:
          data[["BloodPressure1", "SkinThickness1"]]=scaler2.fit_transform(data[["BloodPressure", "SkinThickness"]])
In [26]:
          data[["BloodPressure", "BloodPressure1"]]
In [27]:
               BloodPressure BloodPressure1
Out[27]:
            0
                         72
                                  0.149641
            1
                         66
                                  -0.160546
            2
                         64
                                  -0.263941
            3
                         66
                                  -0.160546
            4
                         40
                                  -1.504687
          763
                         76
                                  0.356432
                                  0.046245
          764
                         70
          765
                         72
                                  0.149641
          766
                         60
                                  -0.470732
          767
                         70
                                  0.046245
         768 rows × 2 columns
          data["BloodPressure"].mean()
In [28]:
          69.10546875
Out[28]:
In [32]: fig,(ax1,ax2)=plt.subplots(ncols=2,figsize=(10,4))
          ax1.set_title("Before Applying StandardScaler")
          sns.kdeplot(data["BloodPressure"],ax=ax1,color="green")
          sns.kdeplot(data["SkinThickness"],ax=ax1,color="red")
```

```
ax2.set_title("After Applying StandardScaler")
sns.kdeplot(data["BloodPressure1"],ax=ax2,color="green")
sns.kdeplot(data["SkinThickness1"],ax=ax2,color="red")
```

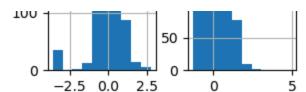
Out[32]: <Axes: title={'center': 'After Applying StandardScaler'}, xlabel='BloodPressure1', ylabel='Density'>







7/26/24, 11:09 PM



```
from sklearn.preprocessing import RobustScaler
In [37]:
          scaler3=RobustScaler()
In [38]:
          data[["Insulin1","BMI1"]]=scaler3.fit_transform(data[["Insulin","BMI"]])
          data[["Insulin1","BMI1"]]
In [44]:
Out[44]:
                Insulin1
                           BMI1
            0 -0.239686 0.172043
            1 -0.239686 -0.580645
            2 -0.239686 -0.935484
            3 0.499018 -0.419355
            4 1.080550 1.193548
               1.174853 0.096774
          764 -0.239686 0.516129
               0.640472 -0.623656
          766 -0.239686 -0.204301
          767 -0.239686 -0.172043
         768 rows × 2 columns
         data["Insulin"].mean()
In [46]:
         79.79947916666667
Out[46]:
```

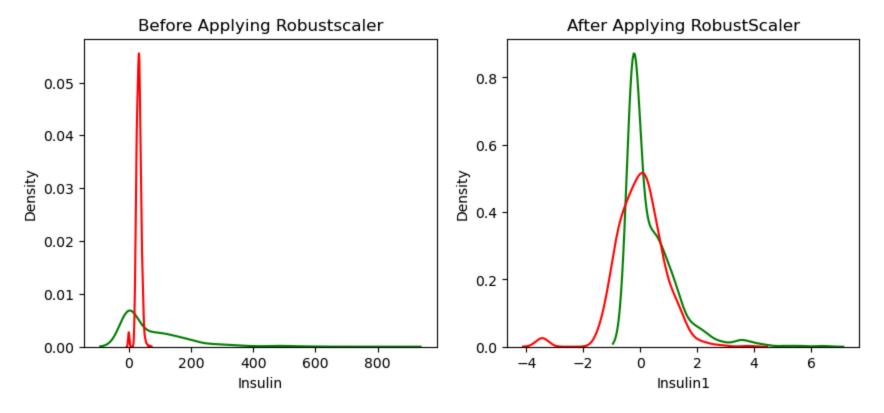
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```
In [47]: fig,(ax1,ax2)=plt.subplots(ncols=2,figsize=(10,4))

ax1.set_title("Before Applying Robustscaler")
    sns.kdeplot(data["Insulin"],ax=ax1,color="green")
    sns.kdeplot(data["BMI"],ax=ax1,color="red")

ax2.set_title("After Applying RobustScaler")
    sns.kdeplot(data["Insulin1"],ax=ax2,color="green")
    sns.kdeplot(data["BMI1"],ax=ax2,color="red")
```

Out[47]: <Axes: title={'center': 'After Applying RobustScaler'}, xlabel='Insulin1', ylabel='Density'>



In []: