

```
import numpy as np
import pandas as pd
from google.colab import files
fupload = files.upload()
```

No file chosen

Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

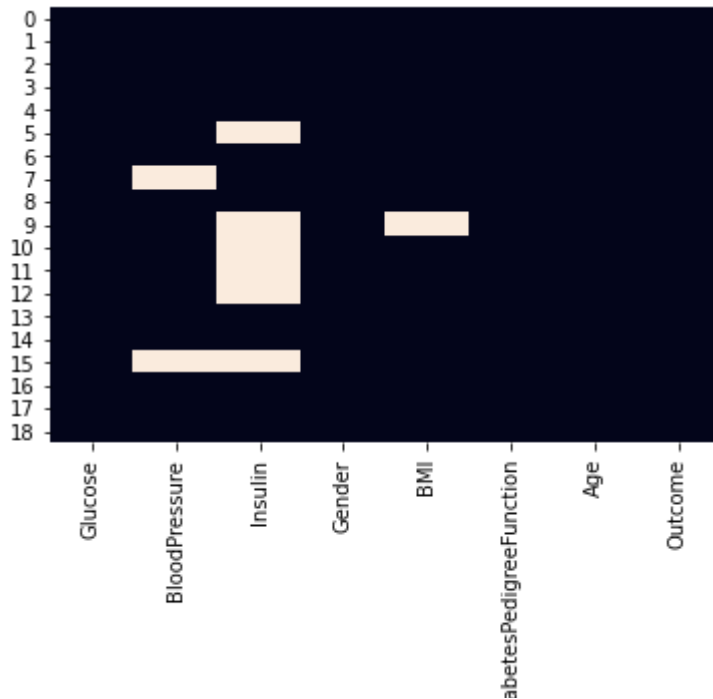
Saving diabetes2.csv to diabetes2.csv

```
data=pd.read_csv("diabetes2.csv")
data
```

	Glucose	BloodPressure	Insulin	Gender	BMI	DiabetesPedigreeFunction	Age	Out
0	148	72.0	88.0	M	33.6	0.627	50	
1	85	66.0	90.0	F	26.6	0.351	31	
2	183	64.0	75.0	M	23.3	0.672	32	
3	89	66.0	94.0	M	28.1	0.167	21	
4	137	40.0	168.0	M	43.1	2.288	33	
5	116	74.0	NaN	F	25.6	0.201	30	
6	78	50.0	88.0	M	31.0	0.248	26	
7	115	NaN	78.0	F	35.3	0.134	29	
8	197	70.0	543.0	F	30.5	0.158	53	
9	125	96.0	NaN	M	NaN	0.232	54	
10	110	92.0	NaN	F	37.6	0.191	30	
11	168	74.0	NaN	M	38.0	0.537	34	
12	139	80.0	NaN	F	27.1	1.441	57	
13	189	60.0	846.0	M	30.1	0.398	59	
14	166	72.0	175.0	M	25.8	0.587	51	
15	100	NaN	NaN	M	30.0	0.484	32	
16	118	84.0	230.0	M	45.8	0.551	31	
17	107	74.0	98.0	M	29.6	0.254	31	
18	103	30.0	83.0	F	43.3	0.183	33	

```
#checking NULL values is present or not in CSV file using heatmap
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
sns.heatmap(data.isnull(), cbar=False)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fa4cd0c85c0>
```



```
print(data.describe())
```

	Glucose	BloodPressure	...	Age	Outcome
count	19.000000	17.000000	...	19.000000	19.000000
mean	130.157895	68.470588	...	37.736842	0.631579
std	36.155933	16.845317	...	11.836113	0.495595
min	78.000000	30.000000	...	21.000000	0.000000
25%	105.000000	64.000000	...	30.500000	0.000000
50%	118.000000	72.000000	...	32.000000	1.000000
75%	157.000000	74.000000	...	50.500000	1.000000
max	197.000000	96.000000	...	59.000000	1.000000

```
[8 rows x 7 columns]
```

```
print(data.shape)
```

```
(19, 8)
```

```
print(data.head(5))
```

	Glucose	BloodPressure	Insulin	...	DiabetesPedigreeFunction	Age	Outcome
0	148	72.0	88.0	...	0.627	50	1
1	85	66.0	90.0	...	0.351	31	0
2	183	64.0	75.0	...	0.672	32	1
3	89	66.0	94.0	...	0.167	21	0
4	137	40.0	168.0	...	2.288	33	1

```
[5 rows x 8 columns]
```

```
x=data[['Glucose','BloodPressure','Insulin','Gender','DiabetesPedigreeFunction','Age']].va
print(x[0:8])
#print(x)
```

```
[[148 72.0 88.0 'M' 0.627 50]
 [85 66.0 90.0 'F' 0.35100000000000003 31]
```

```
[183 64.0 75.0 'M' 0.672 32]
[89 66.0 94.0 'M' 0.16699999999999998 21]
[137 40.0 168.0 'M' 2.2880000000000003 33]
[116 74.0 nan 'F' 0.201 30]
[78 50.0 88.0 'M' 0.248 26]
[115 nan 78.0 'F' 0.134 29]]
```

```
from sklearn import preprocessing
le = preprocessing.LabelEncoder()
le.fit(['M','F'])
x[:,3]=le.transform(x[:,3])
print(x)
```

```
[[148 72.0 88.0 1 0.627 50]
 [85 66.0 90.0 0 0.35100000000000003 31]
 [183 64.0 75.0 1 0.672 32]
 [89 66.0 94.0 1 0.16699999999999998 21]
 [137 40.0 168.0 1 2.2880000000000003 33]
 [116 74.0 nan 0 0.201 30]
 [78 50.0 88.0 1 0.248 26]
 [115 nan 78.0 0 0.134 29]
 [197 70.0 543.0 0 0.158 53]
 [125 96.0 nan 1 0.23199999999999998 54]
 [110 92.0 nan 0 0.191 30]
 [168 74.0 nan 1 0.537 34]
 [139 80.0 nan 0 1.4409999999999998 57]
 [189 60.0 846.0 1 0.39799999999999996 59]
 [166 72.0 175.0 1 0.5870000000000001 51]
 [100 nan nan 1 0.484 32]
 [118 84.0 230.0 1 0.551 31]
 [107 74.0 98.0 1 0.254 31]
 [103 30.0 83.0 0 0.183 33]]
```

```
y=data['Outcome'].values
print(y[0:5])
print(y)
```

```
[1 0 1 0 1]
[1 0 1 0 1 0 1 0 1 1 0 1 0 1 1 1 1 0]
```

```
from sklearn.impute import SimpleImputer
imputer=SimpleImputer(missing_values=np.nan,strategy="mean")
x=imputer.fit_transform(x)
print(x)
```

```
[[1.48000000e+02 7.20000000e+01 8.80000000e+01 1.00000000e+00
 6.27000000e-01 5.00000000e+01]
 [8.50000000e+01 6.60000000e+01 9.00000000e+01 0.00000000e+00
 3.51000000e-01 3.10000000e+01]
 [1.83000000e+02 6.40000000e+01 7.50000000e+01 1.00000000e+00
 6.72000000e-01 3.20000000e+01]
 [8.90000000e+01 6.60000000e+01 9.40000000e+01 1.00000000e+00
 1.67000000e-01 2.10000000e+01]
 [1.37000000e+02 4.00000000e+01 1.68000000e+02 1.00000000e+00
 2.28800000e+00 3.30000000e+01]
 [1.16000000e+02 7.40000000e+01 2.04307692e+02 0.00000000e+00
 0.00000000e+00 0.00000000e+00]]
```

```

2.01000000e-01 3.00000000e+01]
[7.80000000e+01 5.00000000e+01 8.80000000e+01 1.00000000e+00
 2.48000000e-01 2.60000000e+01]
[1.15000000e+02 6.84705882e+01 7.80000000e+01 0.00000000e+00
 1.34000000e-01 2.90000000e+01]
[1.97000000e+02 7.00000000e+01 5.43000000e+02 0.00000000e+00
 1.58000000e-01 5.30000000e+01]
[1.25000000e+02 9.60000000e+01 2.04307692e+02 1.00000000e+00
 2.32000000e-01 5.40000000e+01]
[1.10000000e+02 9.20000000e+01 2.04307692e+02 0.00000000e+00
 1.91000000e-01 3.00000000e+01]
[1.68000000e+02 7.40000000e+01 2.04307692e+02 1.00000000e+00
 5.37000000e-01 3.40000000e+01]
[1.39000000e+02 8.00000000e+01 2.04307692e+02 0.00000000e+00
 1.44100000e+00 5.70000000e+01]
[1.89000000e+02 6.00000000e+01 8.46000000e+02 1.00000000e+00
 3.98000000e-01 5.90000000e+01]
[1.66000000e+02 7.20000000e+01 1.75000000e+02 1.00000000e+00
 5.87000000e-01 5.10000000e+01]
[1.00000000e+02 6.84705882e+01 2.04307692e+02 1.00000000e+00
 4.84000000e-01 3.20000000e+01]
[1.18000000e+02 8.40000000e+01 2.30000000e+02 1.00000000e+00
 5.51000000e-01 3.10000000e+01]
[1.07000000e+02 7.40000000e+01 9.80000000e+01 1.00000000e+00
 2.54000000e-01 3.10000000e+01]
[1.03000000e+02 3.00000000e+01 8.30000000e+01 0.00000000e+00
 1.83000000e-01 3.30000000e+01]]

```

#Handling High Variance

```

from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
x=scaler.fit_transform(x)
print(x)

```

```

[[ 0.50699897  0.22831789 -0.63356898  0.76376262  0.22633617  1.06447085]
 [-1.28320093 -0.15982253 -0.62267428 -1.30930734 -0.31096889 -0.58477368]
 [ 1.50155447 -0.28920267 -0.70438456  0.76376262  0.31394026 -0.49797134]
 [-1.16953745 -0.15982253 -0.60088487  0.76376262 -0.66917226 -1.45279712]
 [ 0.19442438 -1.84176434 -0.19778079  0.76376262  3.45990032 -0.411169 ]
 [-0.40230892  0.35769803  0.          -1.30930734 -0.60298251 -0.67157603]
 [-1.48211203 -1.19486364 -0.63356898  0.76376262 -0.51148491 -1.0187854 ]
 [-0.43072479  0.          -0.6880425  -1.30930734 -0.73341526 -0.75837837]
 [ 1.89937667  0.09893775  1.84497634 -1.30930734 -0.68669308  1.32487788]
 [-0.14656607  1.78087957  0.          0.76376262 -0.54263303  1.41168022]
 [-0.57280415  1.52211929  0.          -1.30930734 -0.62245008 -0.67157603]
 [ 1.0753164   0.35769803  0.          0.76376262  0.051128   -0.32436665]
 [ 0.25125613  0.74583845  0.          -1.30930734  1.81099675  1.67208725]
 [ 1.6720497  -0.54796294  3.4955241   0.76376262 -0.21947129  1.84569194]
 [ 1.01848466  0.22831789 -0.15964933  0.76376262  0.14846587  1.15127319]
 [-0.85696286  0.          0.          0.76376262 -0.05205015 -0.49797134]
 [-0.34547717  1.00459873  0.13995505  0.76376262  0.0783826  -0.58477368]
 [-0.65805176  0.35769803 -0.57909546  0.76376262 -0.49980436 -0.58477368]
 [-0.77171525 -2.48866504 -0.66080574 -1.30930734 -0.63802414 -0.411169  ]]

```

#min-max normalization

```

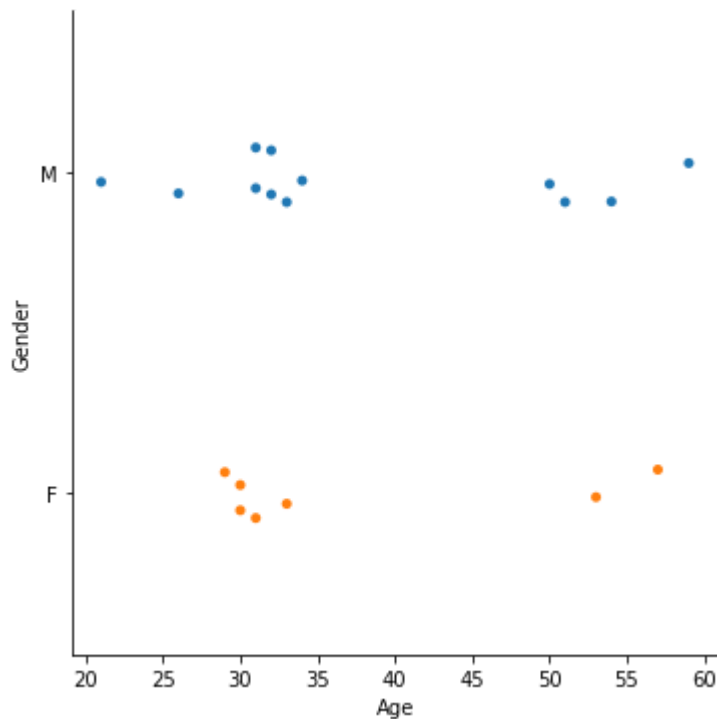
from sklearn.preprocessing import MinMaxScaler
s=MinMaxScaler(feature_range=(0,1))
rescaledx=s.fit_transform(x)
nn.set_printoptions(precision=3)

```

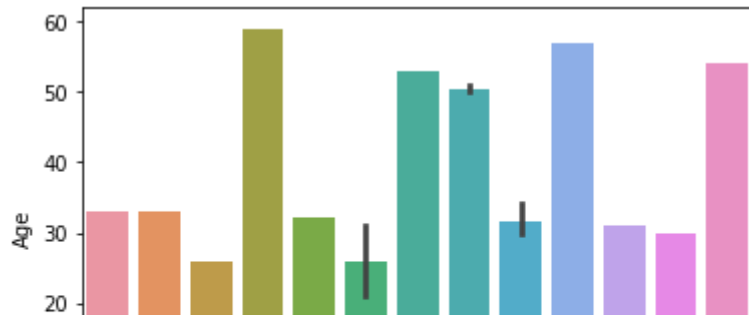
```
np.set_printoptions(precision=3)
print(x)
```

```
[[ 0.507  0.228 -0.634  0.764  0.226  1.064]
 [-1.283 -0.16  -0.623 -1.309 -0.311 -0.585]
 [ 1.502 -0.289 -0.704  0.764  0.314 -0.498]
 [-1.17  -0.16  -0.601  0.764 -0.669 -1.453]
 [ 0.194 -1.842 -0.198  0.764  3.46  -0.411]
 [-0.402  0.358  0.    -1.309 -0.603 -0.672]
 [-1.482 -1.195 -0.634  0.764 -0.511 -1.019]
 [-0.431  0.    -0.688 -1.309 -0.733 -0.758]
 [ 1.899  0.099  1.845 -1.309 -0.687  1.325]
 [-0.147  1.781  0.    0.764 -0.543  1.412]
 [-0.573  1.522  0.    -1.309 -0.622 -0.672]
 [ 1.075  0.358  0.    0.764  0.051 -0.324]
 [ 0.251  0.746  0.    -1.309  1.811  1.672]
 [ 1.672 -0.548  3.496  0.764 -0.219  1.846]
 [ 1.018  0.228 -0.16  0.764  0.148  1.151]
 [-0.857  0.    0.    0.764 -0.052 -0.498]
 [-0.345  1.005  0.14  0.764  0.078 -0.585]
 [-0.658  0.358 -0.579  0.764 -0.5   -0.585]
 [-0.772 -2.489 -0.661 -1.309 -0.638 -0.411]]
```

```
import seaborn as sns
res = sns.catplot(x="Age", y="Gender", data=data)
plt.show()
```

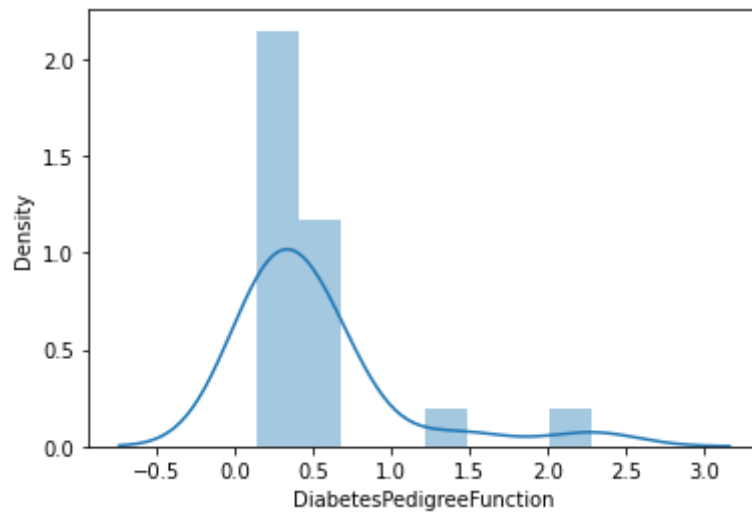


```
res = sns.barplot(x=data['BloodPressure'], y=data['Age'])
plt.show()
```



```
sns.distplot(data["DiabetesPedigreeFunction"], hist=True)
plt.show()
```

```
/usr/local/lib/python3.6/dist-packages/seaborn/distributions.py:2551: FutureWarning:
warnings.warn(msg, FutureWarning)
```



```
sns.histplot(data["Age"], binwidth=3)
plt.show()
```

