

1 - 2D Matrix

```
In [ ]:

import numpy
import pandas

myarray = numpy.array([[1,2,3],[4,5,6]]) # Creating Array
rownames = ['a','b'] # Naming Row
colnames=['f1','f2','f3'] # Naming Column

mydataframe = pandas.DataFrame(myarray, index = rownames, columns=colnames)
print(mydataframe)
```

	f1	f2	f3
a	1	2	3
b	4	5	6

```
In [ ]:

import numpy
import pandas

myarray = numpy.array([[ 'a', 'sandhya', 9.6],[4, 'shreya', 6.5]])
rownames = ['r1','r2']
colnames=['f1','f2','f3']

mydataframe = pandas.DataFrame(myarray, index = rownames, columns=colnames)
print(mydataframe)
```

	f1	f2	f3
r1	a	sandhya	9.6
r2	4	shreya	6.5

2 - Import CSV

```
In [ ]:

# Load csv file using pandas from a specific path or url

from pandas import read_csv

data=read_csv('/content/Diabetes.csv')
colnames=['Pregnancies','Glucose','BloodPressure','SkinThickness','Insulin','BMI','DiabetesPedigreeFunction','Age','Outcome']

print(data.shape)
```

(768, 9)

3 - Statistical Summary

```
In [ ]:

# This will give statistics of each column in the dataset.

description = data.describe()

print(description)
```

	Pregnancies	Glucose	...	Age	Outcome
count	768.000000	768.000000	...	768.000000	768.000000
mean	3.845052	120.894531	...	33.240885	0.348958
std	3.369578	31.972618	...	11.760232	0.476951
min	0.000000	0.000000	...	21.000000	0.000000
25%	1.000000	99.000000	...	24.000000	0.000000
50%	3.000000	117.000000	...	29.000000	0.000000
75%	6.000000	140.250000	...	41.000000	1.000000
max	17.000000	199.000000	...	81.000000	1.000000

[8 rows x 9 columns]

```
In [ ]:

# Size of matrix

print(data.shape)
```

(768, 9)

```
In [ ]:

# Peek at data

print(data.head(4))
```

	Pregnancies	Glucose	BloodPressure	...	DiabetesPedigreeFunction	Age	Outcome
0	6	148	72	...	0.627	50	1
1	1	85	66	...	0.351	31	0
2	8	183	64	...	0.672	32	1
3	1	89	66	...	0.167	21	0

[4 rows x 9 columns]

In []:

```
# Group on the basis of a particular attribute

print(data.groupby("Outcome").size())
```

Outcome
0 500
1 268
dtype: int64

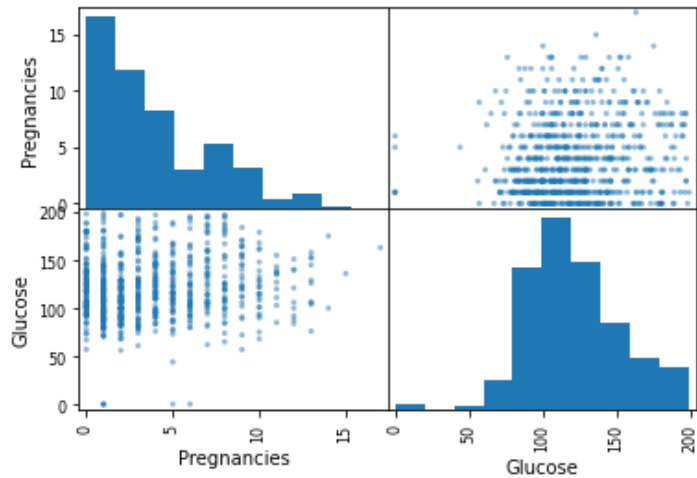
4 - Data Visualization

In []:

```
# Plotting pairs of attributes as scattered plot, specify the attributes to be plotted explicitly

import matplotlib.pyplot as plt
import pandas
from pandas.plotting import scatter_matrix

scatter_matrix(data[['Pregnancies', 'Glucose']])
plt.show()
```



In [133]:

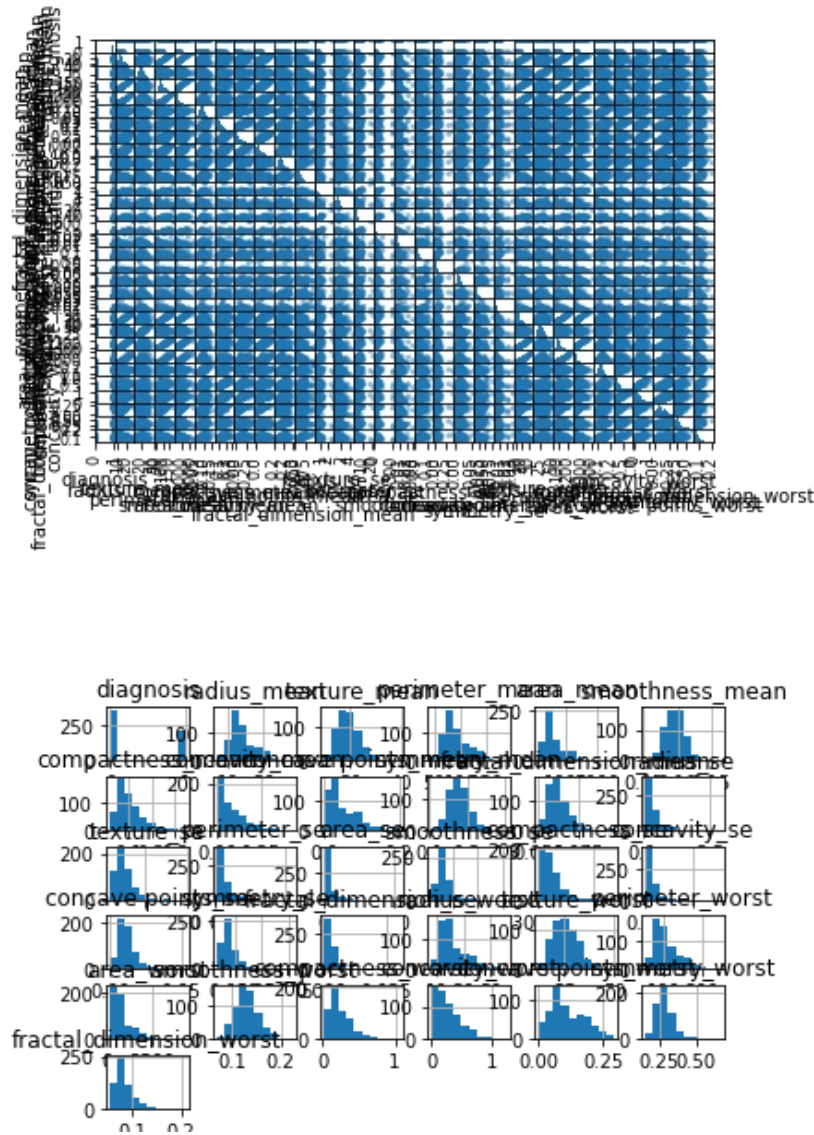
```
# Plot all pairs of attributes in data

import matplotlib.pyplot as plt
import pandas
from pandas.plotting import scatter_matrix

scatter_matrix(data) # Scatter plot

print()
plt.show()
print("\n\n")

data.hist() # Histogram
plt.show()
```



5 - Standardization of dataset

In [132]:

```
from sklearn.preprocessing import StandardScaler
import pandas
import numpy

arr=data.values  # Convert data frame to array

X=arr[:,0:8] # Split columns
Y=arr[:,8] # Only 7th Column since index starts from 0

scaler=StandardScaler().fit(X) # Fit data for standardization
rescaledX=scaler.transform(X) # Convert the data as per (x-μ)/σ
numpy.set_printoptions(precision=3)

print()
print(rescaledX[0:2,:])
print()

print(X[0:2,:])
```

```
[[ 1.298  1.097 -2.073  1.27  0.984  1.568  3.284  2.653]
 [ 1.298  1.83  -0.354  1.686  1.909 -0.827 -0.487 -0.024]]

[[1.000e+00 1.799e+01 1.038e+01 1.228e+02 1.001e+03 1.184e-01 2.776e-01
 3.001e-01]
 [1.000e+00 2.057e+01 1.777e+01 1.329e+02 1.326e+03 8.474e-02 7.864e-02
 8.690e-02]]
```

6 - Normalizing a Column

In [131]:

```
# Create a dataframe for a set of values in an array

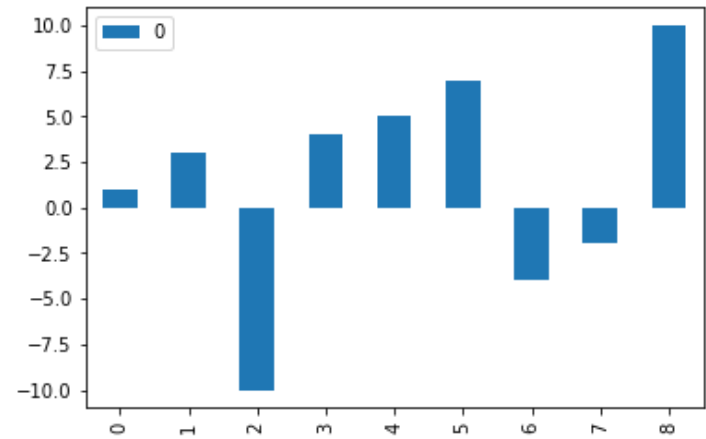
myarray=numpy.array([1,3,-10,4,5,7,-4,-2,10])
mydataframe = pandas.DataFrame(myarray)

print()
print(mydataframe)
print()

# Plot the data

mydataframe.plot(kind='bar')
plt.show()
```

```
0
0  1
1  3
2 -10
3  4
4  5
5  7
6 -4
7 -2
8 10
```



In [134]:

```
# Plot normalized data

from sklearn import preprocessing
fl_x = mydataframe.values.astype(float)

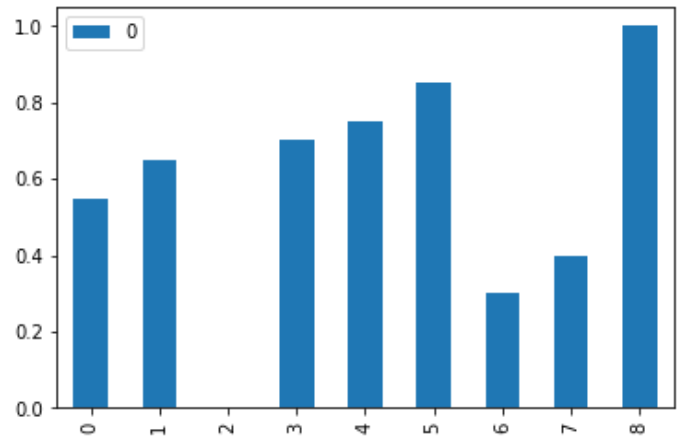
# fl_x=mydataframe[['fl']].values.astype(float) # If specific feature name is to be converted

min_max_scaler = preprocessing.MinMaxScaler()
X_scaled=min_max_scaler.fit_transform(fl_x)
df_normalized = pandas.DataFrame(X_scaled)

print()
print(df_normalized)
print("\n\n")
```

```
df_normalized.plot(kind='bar')
plt.show()
```

```
0
0  0.55
1  0.65
2  0.00
3  0.70
4  0.75
5  0.85
6  0.30
7  0.40
8  1.00
```



Normalization and Standardization

Difference between Normalization and Standardization

- The terms normalization and standardization are sometimes used interchangeably, but they usually refer to different things.
- Normalization usually means to scale a variable to have a values between a desired range (like [-1,1] or [0,1]).
- Standardization transforms data to have a mean of zero and a standard deviation of 1.
- The result of standardization (or Z-score normalization) is that the features will be rescaled to ensure the mean and the standard deviation to be 0 and 1, respectively.
- Advantage of Normalization over Standardization is that we are not bound to any specific distribution.
- In addition to that Normalization also suppresses the effect of outliers to some extent.

Breast Cancer

In []:

```
# Load csv file using pandas from a specific path or url

from pandas import read_csv

data=read_csv('/content/Breast Cancer.csv')

print(data.shape)
```

(569, 33)

In []:

```
# This will give statistics of each column in the dataset.

description = data.describe()

print(description)
```

	id	radius_mean	...	fractal_dimension_worst	Unnamed: 32
count	5.690000e+02	569.000000	...	569.000000	0.0
mean	3.037183e+07	14.127292	...	0.083946	NaN
std	1.250206e+08	3.524049	...	0.018061	NaN
min	8.670000e+03	6.981000	...	0.055040	NaN
25%	8.692180e+05	11.700000	...	0.071460	NaN
50%	9.060240e+05	13.370000	...	0.080040	NaN
75%	8.813129e+06	15.780000	...	0.092080	NaN
max	9.113205e+08	28.110000	...	0.207500	NaN

[8 rows x 32 columns]

In []:

```
# Size of matrix

print(data.shape)
```

(569, 33)

In []:

```
# Peek at data

print(data.head(4))
```

	id	diagnosis	...	fractal_dimension_worst	Unnamed: 32
0	000000	M	...	0.110000	0.000000

0	842302	M	...	0.11890	NaN
1	842517	M	...	0.08902	NaN
2	84300903	M	...	0.08758	NaN
3	84348301	M	...	0.17300	NaN

[4 rows x 33 columns]

In []:

Group on the basis of a particular attribute

```
print(data.groupby("perimeter_se").size())
```

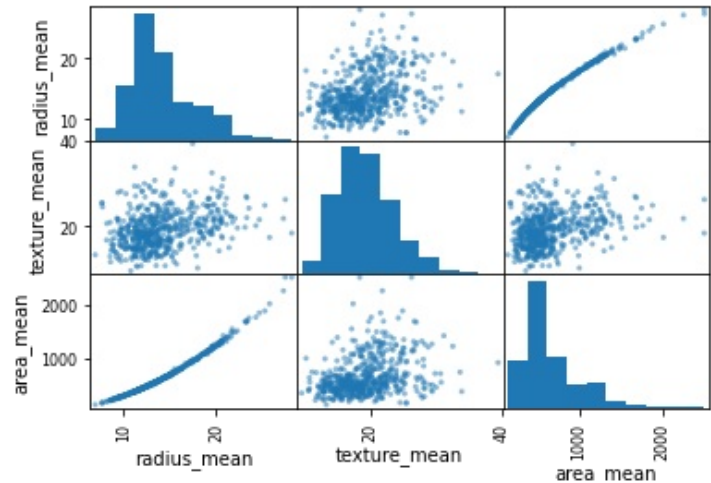
```
perimeter_se
0.7570      1
0.7714      1
0.8439      1
0.8484      1
0.8730      1
..
10.0500     1
10.1200     1
11.0700     1
18.6500     1
21.9800     1
Length: 533, dtype: int64
```

In []:

Plotting pairs of attributes as scattered plot, specify the attributes to be plotted explicitly

```
import matplotlib.pyplot as plt
import pandas
from pandas.plotting import scatter_matrix

scatter_matrix(data[['radius_mean', 'texture_mean', 'area_mean']])
plt.show()
```



In []:

Plot all pairs of attributes in data

```
import matplotlib.pyplot as plt
import pandas
from pandas.plotting import scatter_matrix

# Cleaning and modifying the data
data = data.drop('id',axis=1)
data = data.drop('Unnamed: 32',axis=1)

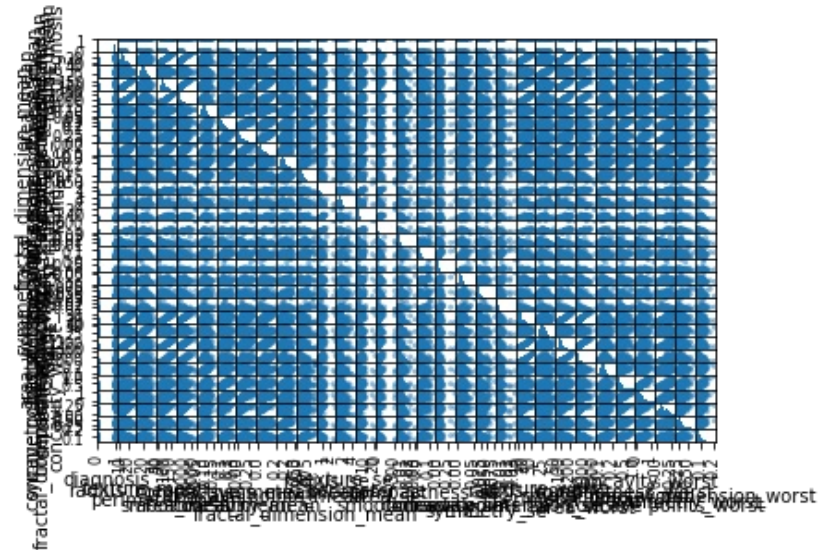
# Mapping Benign to 0 and Malignant to 1
data['diagnosis'] = data['diagnosis'].map({'M':1, 'B':0})

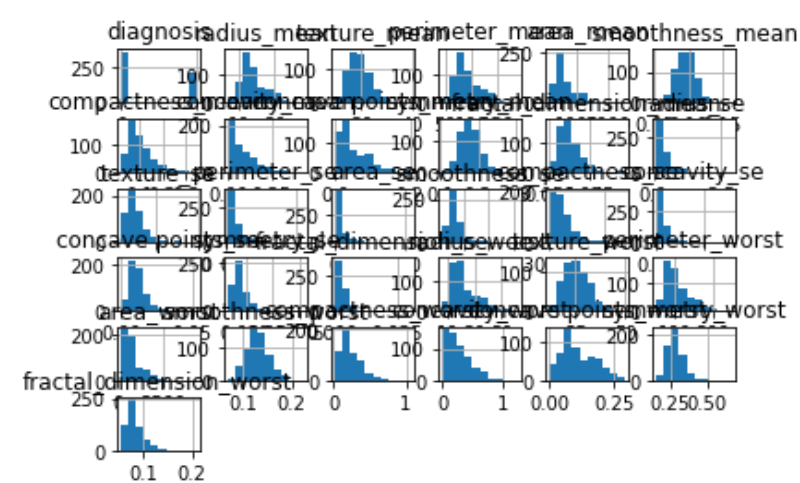
#Check the data stats
data.describe()

scatter_matrix(data) # Scatter plot
plt.show()

print("\n\n")

data.hist() # Histogram
plt.show()
```





```
In [ ]:
data.shape
```

Out[]:

(569, 31)

In [128]:

```
from sklearn.preprocessing import StandardScaler
import pandas
import numpy

arr=data.values # Convert data frame to array

X=arr[:,0:31] # Split columns
Y=arr[:,30]

scaler=StandardScaler().fit(X) # Fit data for standardization
rescaledX=scaler.transform(X) # Convert the data as per (x-μ)/σ
numpy.set_printoptions(precision=3)

print(rescaledX[0:2,:])

print()

print(X[0:2,:])
```

```
[[ 1.298e+00  1.097e+00 -2.073e+00  1.270e+00  9.844e-01  1.568e+00
  3.284e+00  2.653e+00  2.532e+00  2.218e+00  2.256e+00  2.490e+00
 -5.653e-01  2.833e+00  2.488e+00 -2.140e-01  1.317e+00  7.240e-01
  6.608e-01  1.149e+00  9.071e-01  1.887e+00 -1.359e+00  2.304e+00
  2.001e+00  1.308e+00  2.617e+00  2.110e+00  2.296e+00  2.751e+00
  1.937e+00]
[ 1.298e+00  1.830e+00 -3.536e-01  1.686e+00  1.909e+00 -8.270e-01
 -4.871e-01 -2.385e-02  5.481e-01  1.392e-03 -8.687e-01  4.993e-01
 -8.762e-01  2.633e-01  7.424e-01 -6.054e-01 -6.929e-01 -4.408e-01
  2.602e-01 -8.055e-01 -9.944e-02  1.806e+00 -3.692e-01  1.535e+00
  1.890e+00 -3.756e-01 -4.304e-01 -1.467e-01  1.087e+00 -2.439e-01
  2.812e-01]]

[[1.000e+00 1.799e+01 1.038e+01 1.228e+02 1.001e+03 1.184e-01 2.776e-01
 3.001e-01 1.471e-01 2.419e-01 7.871e-02 1.095e+00 9.053e-01 8.589e+00
 1.534e+02 6.399e-03 4.904e-02 5.373e-02 1.587e-02 3.003e-02 6.193e-03
 2.538e+01 1.733e+01 1.846e+02 2.019e+03 1.622e-01 6.656e-01 7.119e-01
 2.654e-01 4.601e-01 1.189e-01]
[1.000e+00 2.057e+01 1.777e+01 1.329e+02 1.326e+03 8.474e-02 7.864e-02
 8.690e-02 7.017e-02 1.812e-01 5.667e-02 5.435e-01 7.339e-01 3.398e+00
 7.408e+01 5.225e-03 1.308e-02 1.860e-02 1.340e-02 1.389e-02 3.532e-03
 2.499e+01 2.341e+01 1.588e+02 1.956e+03 1.238e-01 1.866e-01 2.416e-01
 1.860e-01 2.750e-01 8.902e-02]]
```