# **Linear Regression on Diabetes dataset**

#### In [9]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

#### In [10]:

```
pdata= pd.read_csv('C:\\Users\\Jai mata di\\Downloads\\diabetes.csv')
pdata.head()
```

#### Out[10]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction
0	6	148	72	35	0	33.6	0.62
1	1	85	66	29	0	26.6	0.35
2	8	183	64	0	0	23.3	0.67:
3	1	89	66	23	94	28.1	0.16
4	0	137	40	35	168	43.1	2.28
4							<b>)</b>

#### In [11]:

```
pdata.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
```

#	Column	Non-Null Count	Dtype
0	Pregnancies	768 non-null	int64
1	Glucose	768 non-null	int64
2	BloodPressure	768 non-null	int64
3	SkinThickness	768 non-null	int64
4	Insulin	768 non-null	int64
5	BMI	768 non-null	float64
6	DiabetesPedigreeFunction	768 non-null	float64
7	Age	768 non-null	int64
8	Outcome	768 non-null	int64

dtypes: float64(2), int64(7)
memory usage: 54.1 KB

```
In [30]:
```

```
class_counts = pdata.groupby('Outcome').size()
print("Class breakdown of the data:\n")
print(class_counts)
```

Class breakdown of the data:

Outcome 0 500 1 268 dtype: int64

#### In [12]:

pdata.corr()

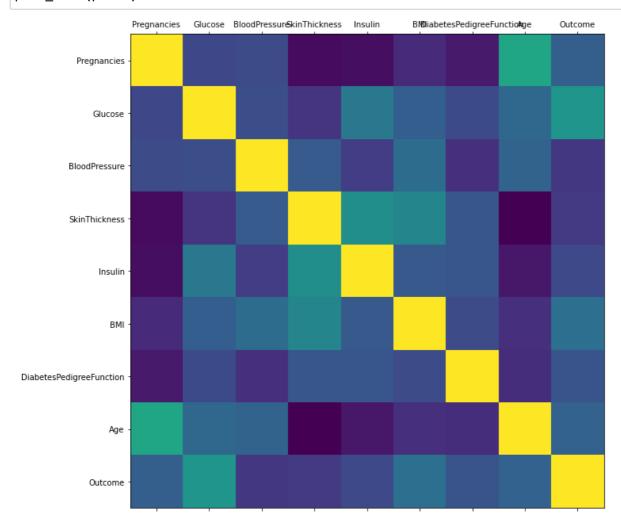
#### Out[12]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	
Pregnancies	1.000000	0.129459	0.141282	-0.081672	-0.073535	0.0
Glucose	0.129459	1.000000	0.152590	0.057328	0.331357	0.2
BloodPressure	0.141282	0.152590	1.000000	0.207371	0.088933	0.2
SkinThickness	-0.081672	0.057328	0.207371	1.000000	0.436783	0.3
Insulin	-0.073535	0.331357	0.088933	0.436783	1.000000	0.1
ВМІ	0.017683	0.221071	0.281805	0.392573	0.197859	1.0
DiabetesPedigreeFunction	-0.033523	0.137337	0.041265	0.183928	0.185071	0.1
Age	0.544341	0.263514	0.239528	-0.113970	-0.042163	0.0
Outcome	0.221898	0.466581	0.065068	0.074752	0.130548	0.2
4						•

#### In [17]:

### In [18]:

## plot\_corr(pdata)



```
In [22]:
```

```
# from sklearn.model_selection import train_test_split
X = pdata.drop('Outcome',axis=1) # Predictor feature columns (8 X m)
Y = pdata['Outcome'] # Predicted class (1=True, 0=False) (1 X m)
x_train, x_test, y_train, y_test = train_test_split(X, Y, test_size=0.3, random_state=1)
# 1 is just any random seed number
x_train.head()
```

#### Out[22]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunct
88	15	136	70	32	110	37.1	0.
467	0	97	64	36	100	36.8	0.
550	1	116	70	28	0	27.4	0
147	2	106	64	35	119	30.5	1.
481	0	123	88	37	0	35.2	0.

```
→
```

#### In [23]:

```
print(x_train.shape)
```

(537, 8)

#### In [24]:

```
print(y_train.shape)
```

(537,)

#### In [31]:

```
skew = pdata.skew()
print("Skew of attribute distributions in the data:\n")
skew
```

Skew of attribute distributions in the data:

#### Out[31]:

Pregnancies	0.901674
Glucose	0.173754
BloodPressure	-1.843608
SkinThickness	0.109372
Insulin	2.272251
BMI	-0.428982
DiabetesPedigreeFunction	1.919911
Age	1.129597
Outcome	0.635017

dtype: float64

## Visualizing data using Pandas

Now, we visualize data using Python's Pandas library. We discuss both univariate plots and multivariate plots using Pandas.

Univariate plots:

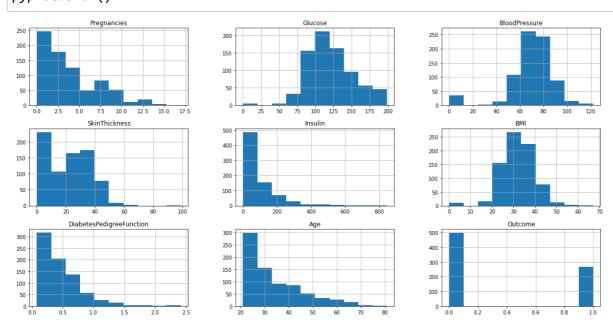
Histograms. Density Plots. Box and Whisker Plots.

#### In [32]:

```
# Import required package
from matplotlib import pyplot
pyplot.rcParams['figure.figsize'] = [20, 10] # set the figure size
```

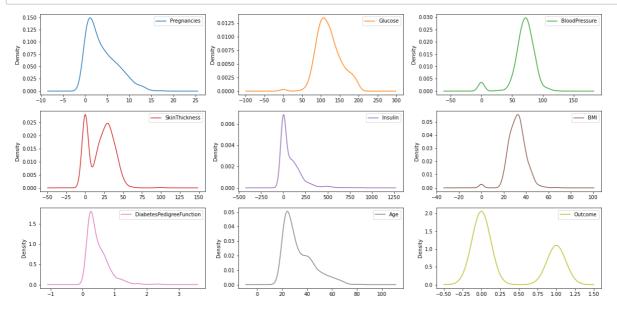
#### In [33]:

```
# Draw histograms for all attributes
pdata.hist()
pyplot.show()
```



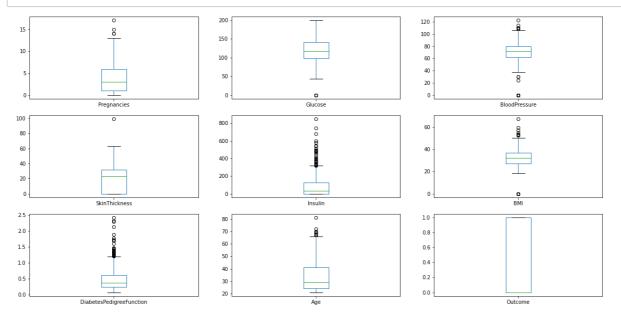
#### In [34]:

# # Density plots for all attributes pdata.plot(kind='density', subplots=True, layout=(3,3), sharex=False) pyplot.show()



#### In [35]:

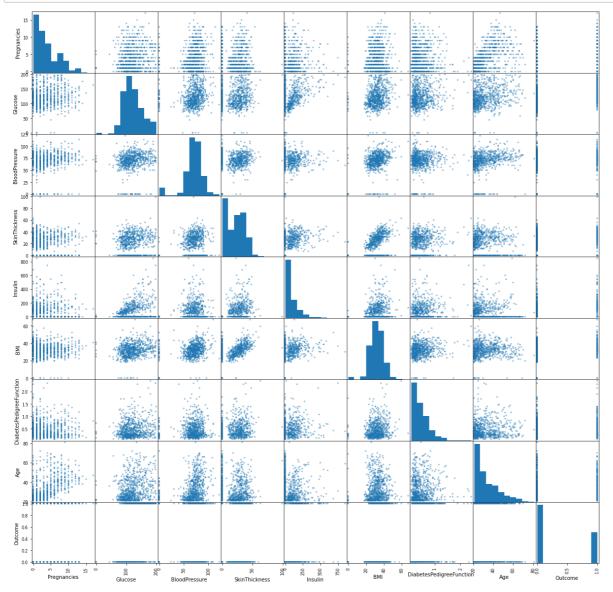
# Draw box and whisker plots for all attributes
pdata.plot(kind= 'box', subplots=True, layout=(3,3), sharex=False, sharey=False)
pyplot.show()



#### In [36]:

```
# Import required package
from pandas.plotting import scatter_matrix
pyplot.rcParams['figure.figsize'] = [20, 20]

# Plotting Scatterplot Matrix
scatter_matrix(pdata)
pyplot.show()
```



In [ ]:			