```
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
```

Out[3]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunctio
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
763	10	101	76	48	180	32.9	0.17
764	2	122	70	27	0	36.8	0.34
765	5	121	72	23	112	26.2	0.24
766	1	126	60	0	0	30.1	0.34
767	1	93	70	31	0	30.4	0.31

768 rows × 9 columns

In [4]: | df.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

```
df=df.dropna()
In [5]:
In [6]: |df.describe()
Out[6]:
                Pregnancies
                              Glucose
                                      BloodPressure SkinThickness
                                                                     Insulin
                                                                                  BMI
                                                                                       Diabetes
                 768.000000 768.000000
                                         768.000000
                                                       768.000000 768.000000
                                                                            768.000000
          count
                   3.845052 120.894531
                                          69.105469
                                                        20.536458
                                                                  79.799479
                                                                             31.992578
          mean
                   3.369578
                            31.972618
                                          19.355807
                                                        15.952218 115.244002
                                                                              7.884160
            std
           min
                   0.000000
                             0.000000
                                           0.000000
                                                         0.000000
                                                                   0.000000
                                                                              0.000000
           25%
                   1.000000
                            99.000000
                                          62.000000
                                                         0.000000
                                                                   0.000000
                                                                             27.300000
           50%
                   3.000000 117.000000
                                          72.000000
                                                        23.000000
                                                                  30.500000
                                                                             32.000000
           75%
                   6.000000
                           140.250000
                                          80.000000
                                                        32.000000 127.250000
                                                                             36.600000
                  17.000000 199.000000
                                                        99.000000 846.000000
                                                                             67.100000
           max
                                         122.000000
In [7]: | df.columns
Out[7]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
                 'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
               dtype='object')
In [8]:
        x=df[['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
                 'BMI', 'DiabetesPedigreeFunction', 'Age']]
         y=df['Outcome']
         x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
         lr=LogisticRegression()
         lr.fit(x train,y train)
         C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear model\ logistic.py:
         763: ConvergenceWarning: lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html (https://sciki
         t-learn.org/stable/modules/preprocessing.html)
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear model.html#logistic-regres
         sion (https://scikit-learn.org/stable/modules/linear model.html#logistic-regr
           n_iter_i = _check_optimize_result(
Out[8]: LogisticRegression()
```

```
In [9]: |lr.predict(x test)
 Out[9]: array([1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1,
                0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
                0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1,
                0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0,
                0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0,
                0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 1, 0,
                0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0,
                0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0,
                0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0,
                0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1,
                0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0], dtype=int64)
In [10]: |lr.score(x_test,y_test)
Out[10]: 0.8051948051948052
In [11]: from sklearn.preprocessing import StandardScaler
         fs=StandardScaler().fit_transform(x)
         logr=LogisticRegression()
         logr.fit(fs,y)
Out[11]: LogisticRegression()
In [13]: o=[[1,2,3,4,5,6,7,8]]
         prediction=logr.predict(o)
         print(prediction)
         [1]
In [14]: logr.classes
Out[14]: array([0, 1], dtype=int64)
In [15]: |logr.predict_proba(o)[0][0]
Out[15]: 0.00029236948687560993
In [16]: logr.predict_proba(o)[0][1]
Out[16]: 0.9997076305131244
 In [ ]:
```