In [1]: import numpy as np
 import pandas as pd
 import matplotlib.pyplot as plt
 import seaborn as sns
 from sklearn.linear_model import LogisticRegression
 from sklearn.preprocessing import StandardScaler

In [2]: from sklearn.linear_model import LogisticRegression

In [3]: df=pd.read_csv("health.csv")
df

Out[3]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	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.head()

Out[4]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction
0	6	148	72	35	0	33.6	0.627
1	1	85	66	29	0	26.6	0.351
2	8	183	64	0	0	23.3	0.672
3	1	89	66	23	94	28.1	0.167
4	0	137	40	35	168	43.1	2.288
4.0							

```
In [5]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 768 entries, 0 to 767
         Data columns (total 9 columns):
              Column
                                           Non-Null Count Dtvpe
          0
              Pregnancies
                                                             int64
                                           768 non-null
          1
              Glucose
                                           768 non-null
                                                             int64
              BloodPressure
          2
                                           768 non-null
                                                             int64
          3
              SkinThickness
                                           768 non-null
                                                             int64
          4
              Insulin
                                           768 non-null
                                                             int64
          5
              BMI
                                           768 non-null
                                                             float64
          6
              DiabetesPedigreeFunction
                                                             float64
                                           768 non-null
          7
                                           768 non-null
                                                             int64
          8
                                           768 non-null
              Outcome
                                                             int64
         dtypes: float64(2), int64(7)
         memory usage: 54.1 KB
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
          mean
                                                                               31.992578
                   3.369578
                             31.972618
                                                         15.952218 115.244002
            std
                                           19.355807
                                                                                7.884160
           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%
                                           80.000000
                                                                               36.600000
                   6.000000
                            140.250000
                                                         32.000000 127.250000
                  17.000000 199.000000
                                          122.000000
                                                         99.000000 846.000000
                                                                               67.100000
           max
In [7]: df.columns
Out[7]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
                 'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
               dtype='object')
In [8]: | feature_matrix = df.iloc[:,0:8]
         target_vector = df.iloc[:,-1]
In [9]:
         fs=StandardScaler().fit_transform(feature_matrix)
         logr=LogisticRegression()
```

Out[9]: LogisticRegression()

logr.fit(fs,target vector)

Random Forest

```
In [17]: g1={'Outcome':{'True':1, "False":2}}
    df=df.replace(g1)
    df
```

Out[17]:

	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 [18]: from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.70)
```

```
In [19]: from sklearn.ensemble import RandomForestClassifier
    rfc = RandomForestClassifier()
    rfc.fit(x_train,y_train)
```

Out[19]: RandomForestClassifier()

```
In [21]: from sklearn.model_selection import GridSearchCV
    grid_search = GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring="acgrid_search.fit(x_train,y_train)
```

```
In [ ]:
```