## Simulating data:

```
#importing numpy and pandas
import random
import numpy as np
import pandas as pd
from scipy.stats import bernoulli
#opening file in write mode
file1 = open("myfile.csv", "w")
#simulation of data
#here we
file1.write("no,gender,age,diabates,oxygen level,covid test result\n")
for i in range(128):
    b=i+1
    a=random.randint(0,1)
    c=random.randint(12,80)
    d=np.random.normal(80,20)
    e=np.random.normal(95,2.25)
    f=bernoulli.rvs(p=0.3)
    file1.write(str(b)+","+str(a)+","+str(c)+","+str(d)+","+str(e)+","+str(f)+"\n")
text=open("myfile.csv", "r")
print(text.read(1000))
print("success")
no,gender,age,diabates,oxygen level,covid test result
1,1,64,107.37827291412023,92.80400861920597,0
2,1,15,85.30914453816627,100.30653125474808,1
3,1,49,102.52960614931592,95.98632162241087,0
4,1,80,79.99742342207229,93.91341744043042,0
5,1,21,106.04354550779513,95.0556068531797,1
6,0,47,116.17681278036278,95.74328790865383,0
7,1,25,109.39763079347216,93.2255830505931,0
8,1,26,107.91941529603767,96.67364171096546,0
9,1,75,89.22210464392325,96.01812885387163,0
10,0,70,81.62930364754043,92.37178636872265,0
11,0,60,65.1467981016061,92.04127632245735,0
12,1,17,59.45983400765493,97.99754385396382,0
13,0,55,98.827402017759,93.73565005917814,0
14,1,21,55.32516631158924,95.55529537104569,0
15,1,27,107.74568853800406,89.93901974786228,1
16,1,27,97.56540244343844,92.82020451339822,0
17,1,51,83.15769858770119,96.58277254576515,0
18,0,18,73.32795462759937,90.99269256161507,0
19,1,41,19.250354643719973,91.74234770206274,0
20,1,70,97.40423993058822,95.0169428196571,0
21,0,36,48.410807054762365,94.97
success
```

```
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix
import matplotlib.pyplot as plt
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis as LDA
import pandas as pd
#To read the dataset
df=pd.read_csv("myfile.csv")
#To know more about the dataset
print(df.describe())
#Define the independent and dependent variables
#dependent variable is Decision
y= df['covid_test_result']
x= df.drop(['covid_test_result'], axis=1)
# splitting the data
x_train, x_test, y_train, y_test = train_test_split(x,y, test_size= 0.2)
#Implementing Logistic Regression using sklearn
modelLogistic = LogisticRegression()
modelLogistic.fit(x_train,y_train)
#print the regression coefficients
print("The intercept b0= ", modelLogistic.intercept_)
print("The coefficient b1= ", modelLogistic.coef_)
#Make prediction for the test data
y_pred= modelLogistic.predict(x_test)
#Creating confusion matrix
ConfusionMatrix = confusion_matrix(y_test, y_pred)
print(ConfusionMatrix)
lda = LDA(n_components=1)
X_train = lda.fit_transform(x_train, y_train)
X_test = lda.transform(x_test)
plt.scatter(X_test,y_test)
plt.show()
              no
                      gender
                                     age
                                            diabates oxygen_level
count 128.000000 128.000000 128.000000 128.000000
                                                        128.000000
       64.500000
                                           82.048350
mean
                    0.531250
                              43.234375
                                                         95.328219
std
        37.094474
                    0.500983
                               21.297105
                                           20.990092
                                                          2.349439
min
        1.000000
                    0.000000
                               12.000000
                                           35.472896
                                                         88.762674
                    0.000000 22.000000
25%
       32.750000
                                           68.053545
                                                         93.617270
50%
       64.500000
                    1.000000 44.000000
                                           83.375346
                                                         95.064723
75%
       96.250000
                    1.000000 59.000000
                                           95.443831
                                                         96.855583
       128.000000
                    1.000000 80.000000 128.811930
                                                        101.302784
max
```

```
covid_test_result
count 128.000000
mean 0.250000
std 0.434714
min 0.000000
```

localhost:8888/notebooks/PROJECT WORK.ipynb#

```
21/07/2022, 20:40
                                               PROJECT WORK - Jupyter Notebook
  25%
                   0.000000
  50%
                   0.000000
  75%
                   0.250000
                   1.000000
  max
  The intercept b0= [-20.63978702]
  The coefficient b1= [[ 0.00591778 -0.51003829 -0.00993328 0.00293731 0.
  20431669]]
  [[19 0]
   [7 0]]
  1.0
   0.8
   0.6
   0.4
   0.2
   0.0
```

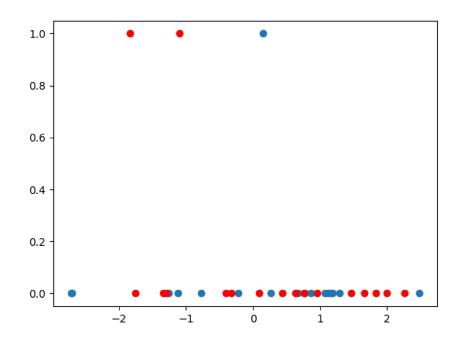
Now we simulate data for pool sample:

Here we've simulated data on the basis of the covariates used in previous one, but here we do not know if an individual is covid positive or not.

Then we divide the total data into three parts, (pools) and test each pools.

Testing the three pools we get two of them are positive, so we divide this two pools into another three or more pool and continue the testing.

## Result:



which is similar to previous scatter plot. The two colors indicate data from two pools. But in second one the cost is low.

## **Discussion:**

Pools sample are useful if the disease prevalence is low and the dataset is too large. But in case of high rates of disease prevalence and low number of samples this method sometimes fails too meet it's desired purpose of cost efficient.