Problem Statement:-

We are working on data from Thoracic surgery of primary lung cancer from this we are going to find accuracy of patient will die or survive from after 1yr year of surgery. for that we are going to usee SVM(support vector machine) algorithm to find accuracy of survival.

SVM(Support Vector Machine)

SVM, or Support Vector Machine, is a type of supervised learning algorithm used for classification and regression analysis. It is a powerful tool that works well with both linear and non-linear data, and it is particularly useful in high-dimensional spaces. SVM aims to find a hyperplane that separates the data points into different classes, with the maximum margin between the classes.

We are using sklearn to import SVM. scikit-learn sklearn is a popular Python library for machine learning. It provides a wide variety of machine learning algorithms and tools for data preprocessing, model selection, and evaluation.

sklearn includes a range of supervised and unsupervised learning algorithms, including linear regression, logistic regression, support vector machines (SVMs), k-nearest neighbors (KNN), decision trees, random forests, gradient boosting, clustering algorithms, and many more.

```
#import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
#inserting data set
from google.colab import drive
drive.mount('/content/drive')
data = pd.read csv('/content/drive/MyDrive/kgce lms/expt7/expt7.csv')
#printing dataset
data
Drive already mounted at /content/drive; to attempt to forcibly
remount, call drive.mount("/content/drive", force_remount=True).
      DGN PRE4 PRE5 PRE6 PRE7 PRE8 PRE9 PRE10 PRE11 PRE14 PRE17
PRE19
     DGN2 2.88 2.16 PRZ1
                               F
                                    F
                                         F
                                               Т
                                                     T 0C14
                                                                 F
0
F
1
     DGN3 3.40 1.88 PRZ0
                              F
                                    F
                                         F
                                               F
                                                     F 0C12
                                                                 F
F
```

```
0C11
2
     DGN3
            2.76
                   2.08
                         PRZ1
                                  F
                                        F
                                              F
                                                    Τ
                                                           F
                                                                         F
F
3
                         PRZ0
     DGN3
            3.68
                   3.04
                                  F
                                        F
                                              F
                                                     F
                                                           F
                                                               0C11
                                                                         F
F
                                              F
4
     DGN3
                   0.96
                         PRZ2
                                  F
                                        Т
                                                    Т
                                                           Т
                                                               0C11
                                                                         F
            2.44
F
. .
                           . . .
465
     DGN2
            3.88
                   2.12
                         PRZ1
                                  F
                                        F
                                              F
                                                     Т
                                                           F
                                                               0C13
                                                                         F
466
     DGN3
            3.76
                   3.12
                         PRZ0
                                   F
                                        F
                                              F
                                                     F
                                                           F
                                                               0C11
                                                                         F
                                                                         F
467
     DGN3
            3.04
                   2.08
                                   F
                                        F
                                              F
                                                    Τ
                                                           F
                                                               0C13
                         PRZ1
                                   F
                                        F
                                              F
468
     DGN3
            1.96
                   1.68
                         PRZ1
                                                     Τ
                                                           Т
                                                               0C12
                                                                         F
                                  F
                                        F
                                              F
                                                     F
                                                               0C12
                                                                         F
469
     DGN3
            4.72
                   3.56
                         PRZ0
                                                           F
F
    PRE25 PRE30 PRE32
                         AGE Risk1Yr
0
         F
               Т
                      F
                           60
                                     F
                                     F
         F
               Т
                      F
1
                           51
2
         F
               Т
                      F
                                     F
                           59
3
         F
               F
                      F
                           54
                                     F
         F
4
               Т
                      F
                           73
                                     Τ
       . . .
                    . . .
         F
                      F
465
               Τ
                           63
                                     F
         F
                                     F
               Т
                      F
466
                           61
         F
                                     F
467
               F
                      F
                           52
468
         F
               Τ
                      F
                           79
                                     F
         F
469
               Т
                           51
                                     F
[470 rows x 17 columns]
#import LabelEndcoder from sklearn.preprocessing
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
data['DGN'] = le.fit transform(data['DGN'])
# data['PRE4'] = le.fit transform(data['PRE4'])
# data['PRE5'] = le.fit_transform(data['PRE5'])
data['PRE6'] = le.fit_transform(data['PRE6'])
data['PRE7'] = le.fit transform(data['PRE7'])
data['PRE8'] = le.fit transform(data['PRE8'])
```

```
data['PRE9'] = le.fit transform(data['PRE9'])
data['PRE10'] = le.fit_transform(data['PRE10'])
data['PRE11'] = le.fit transform(data['PRE11'])
data['PRE14'] = le.fit_transform(data['PRE14'])
data['PRE17'] = le.fit transform(data['PRE17'])
data['PRE19'] = le.fit_transform(data['PRE19'])
data['PRE25'] = le.fit_transform(data['PRE25'])
data['PRE30'] = le.fit_transform(data['PRE30'])
data['PRE32'] = le.fit transform(data['PRE32'])
data['Risk1Yr'] = le.fit transform(data['Risk1Yr'])
data
     DGN
          PRE4
                 PRE5
                        PRE6
                              PRE7
                                     PRE8
                                           PRE9
                                                  PRE10
                                                          PRE11
                                                                 PRE14
PRE17
          2.88
                 2.16
                           1
                                  0
                                        0
                                               0
                                                      1
                                                              1
                                                                      3
0
       1
0
1
       2
          3.40
                                  0
                                        0
                                               0
                                                      0
                                                                      1
                 1.88
                           0
                                                              0
0
2
       2
          2.76
                 2.08
                           1
                                  0
                                        0
                                               0
                                                      1
                                                              0
                                                                      0
0
3
          3.68
       2
                 3.04
                           0
                                  0
                                        0
                                               0
                                                      0
                                                              0
                                                                      0
0
4
       2
          2.44
                 0.96
                           2
                                  0
                                        1
                                               0
                                                      1
                                                              1
                                                                      0
0
. .
     . . .
465
          3.88
                 2.12
                           1
                                  0
                                               0
                                                      1
                                                              0
                                                                      2
       1
                                        0
0
466
          3.76
                 3.12
                           0
                                  0
                                        0
                                               0
                                                      0
                                                              0
                                                                      0
       2
467
                           1
                                                      1
                                                                      2
       2
          3.04
                 2.08
                                  0
                                        0
                                               0
                                                              0
468
          1.96
                                                              1
                                                                      1
       2
                 1.68
                           1
                                  0
                                        0
                                               0
                                                      1
0
469
       2
          4.72
                                  0
                                        0
                                               0
                                                      0
                                                              0
                                                                      1
                 3.56
                           0
     PRE19
             PRE25
                    PRE30
                            PRE32
                                    AGE
                                         Risk1Yr
0
                                     60
          0
                 0
                         1
                                0
                                                0
                         1
                                     51
1
         0
                 0
                                0
                                                0
```

2	0	0	1	0	59	0
3	0	0	0	0	54	0
4	0	0	1	0	73	1
465	0	0	1	0	63	0
466	0	0	1	0	61	0
467	0	0	0	0	52	0
468	0	0	1	0	79	0
469	0	0	1	0	51	0

[470 rows x 17 columns]

#allocing data / bindind data to variable for prediction
y = data['RisklYr']

x = data.drop(['Risk1Yr'], axis = 1)

#printing x

Χ

DGN	١	PRE4	PRE5	PRE6	PRE7	PRE8	PRE9	PRE10	PRE11	PRE14	
PRE17 \	•										
_	L	2.88	2.16	1	0	0	0	1	1	3	
	2	3.40	1.88	0	0	0	0	0	0	1	
	2	2.76	2.08	1	0	0	0	1	0	0	
	2	3.68	3.04	0	0	0	0	0	0	0	
0 4 2 0	2	2.44	0.96	2	0	1	0	1	1	0	
465 1 0	L	3.88	2.12	1	Θ	0	0	1	0	2	
-	2	3.76	3.12	0	0	0	0	0	0	0	
	2	3.04	2.08	1	0	0	0	1	0	2	
	2	1.96	1.68	1	0	0	0	1	1	1	
-	2	4.72	3.56	0	0	0	0	Θ	Θ	1	

	PRE19	PRE25	PRE30	PRE32	AGE
0	0	0	1	0	60
1	0	0	1	0	51
2	0	0	1	0	59
3	0	0	0	0	54
4	0	0	1	0	73

```
465
        0
                                63
               0
                      1
                             0
466
        0
               0
                      1
                             0
                                61
467
        0
               0
                      0
                             0
                                52
468
        0
               0
                      1
                             0
                                79
                      1
               0
                             0
                                51
469
        0
[470 rows x 16 columns]
#printing y
У
0
      0
1
      0
2
      0
3
      0
4
      1
      . .
465
      0
466
      0
467
      0
468
      0
469
      0
Name: Risk1Yr, Length: 470, dtype: int64
#Import the train_test_split from the sklearn
#train test split is a method used in machine learning to split a
dataset into two subsets
from sklearn.model selection import train test split
x train, x test, y train, y test = train test split(x, y, test size =
0.3)
print(x train.shape)
print(x test.shape)
print(y train.shape)
print(y_test.shape)
(329, 16)
(141, 16)
(329,)
(141,)
#importing Support Vector Classification from svm
#SVC stands for Support Vector Classification, which is a type of SVM
algorithm used for classification tasks
from sklearn.svm import SVC
classifier = SVC(kernel = 'poly', random state = 1 )
classifier.fit(x train, y train)
y pred = classifier.predict(x test)
y pred
0,
```

#Import the accuracy_score from the sklearn
#accuracy_score is a performance metric used in classification tasks
to evaluate the accuracy of a machine learning mode
from sklearn.metrics import accuracy_score
print('Accuracy of survival after one year of surgery is')
accuracy score(y test,y pred)*100

Accuracy of survival after one year of surgery is 85.1063829787234