

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 use 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 0 F 1 F	\ DGN2	2.88	2.16	PRZ1	F	F	F	T	T	OC14	F
	DGN3	3.40	1.88	PRZ0	F	F	F	F	F	OC12	F

2	DGN3	2.76	2.08	PRZ1	F	F	F	T	F	0C11	F
F											
3	DGN3	3.68	3.04	PRZ0	F	F	F	F	F	0C11	F
F											
4	DGN3	2.44	0.96	PRZ2	F	T	F	T	T	0C11	F
F											
..
..											
465	DGN2	3.88	2.12	PRZ1	F	F	F	T	F	0C13	F
F											
466	DGN3	3.76	3.12	PRZ0	F	F	F	F	F	0C11	F
F											
467	DGN3	3.04	2.08	PRZ1	F	F	F	T	F	0C13	F
F											
468	DGN3	1.96	1.68	PRZ1	F	F	F	T	T	0C12	F
F											
469	DGN3	4.72	3.56	PRZ0	F	F	F	F	F	0C12	F
F											

	PRE25	PRE30	PRE32	AGE	Risk1Yr
0	F	T	F	60	F
1	F	T	F	51	F
2	F	T	F	59	F
3	F	F	F	54	F
4	F	T	F	73	T
..
465	F	T	F	63	F
466	F	T	F	61	F
467	F	F	F	52	F
468	F	T	F	79	F
469	F	T	F	51	F

[470 rows x 17 columns]

```

#import LabelEncoder 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'])

```

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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 \											
0	1	2.88	2.16	1	0	0	0	1	1	3	
0											
1	2	3.40	1.88	0	0	0	0	0	0	1	
0											
2	2	2.76	2.08	1	0	0	0	1	0	0	
0											
3	2	3.68	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	1	3.88	2.12	1	0	0	0	1	0	2	
0											
466	2	3.76	3.12	0	0	0	0	0	0	0	
0											
467	2	3.04	2.08	1	0	0	0	1	0	2	
0											
468	2	1.96	1.68	1	0	0	0	1	1	1	
0											
469	2	4.72	3.56	0	0	0	0	0	0	1	
0											
	PRE19	PRE25	PRE30	PRE32	AGE	Risk1Yr					
0	0	0	1	0	60	0					
1	0	0	1	0	51	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]

#allocating data / bindind data to variable for prediction

y = data['Risk1Yr']

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

#printing x

x

	DGN	PRE4	PRE5	PRE6	PRE7	PRE8	PRE9	PRE10	PRE11	PRE14
PRE17 \										
0	1	2.88	2.16	1	0	0	0	1	1	3
0										
1	2	3.40	1.88	0	0	0	0	0	0	1
0										
2	2	2.76	2.08	1	0	0	0	1	0	0
0										
3	2	3.68	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	1	3.88	2.12	1	0	0	0	1	0	2
0										
466	2	3.76	3.12	0	0	0	0	0	0	0
0										
467	2	3.04	2.08	1	0	0	0	1	0	2
0										
468	2	1.96	1.68	1	0	0	0	1	1	1
0										
469	2	4.72	3.56	0	0	0	0	0	0	1
0										

	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
...

