

```
In [115]: import pandas as pd
import numpy as np
```

Reading Data

```
In [116]: data=pd.read_csv('titanic.csv')

data.info()
```

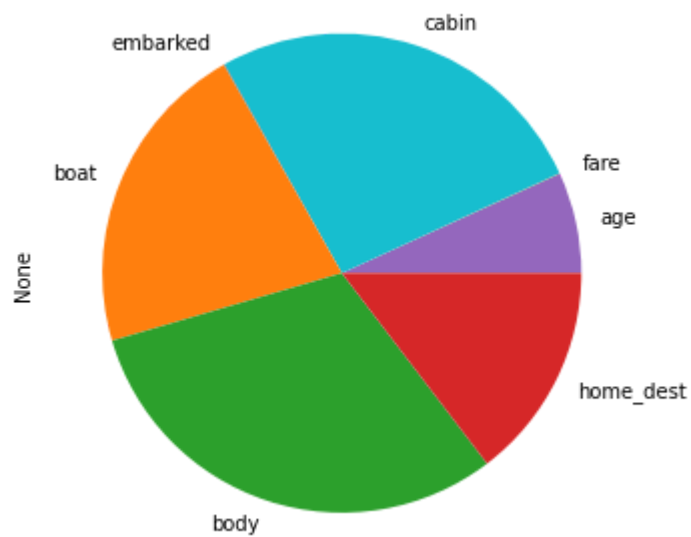
```
RangeIndex: 1309 entries, 0 to 1308
Data columns (total 14 columns):
 #   Column      Non-Null Count  Dtype  
---  -
 0   pclass      1309 non-null   int64  
 1   survived    1309 non-null   bool    
 2   name        1309 non-null   object  
 3   sex         1309 non-null   object  
 4   age         1046 non-null   float64 
 5   sibsp       1309 non-null   int64  
 6   parch       1309 non-null   int64  
 7   ticket      1309 non-null   object  
 8   fare        1308 non-null   float64 
 9   cabin       295 non-null    object  
10   embarked    1307 non-null   object  
11   boat        486 non-null    object  
12   body        121 non-null    float64 
13   home_dest   745 non-null    object  
dtypes: bool(1), float64(3), int64(3), object(7)
memory usage: 98.5+ KB
```

Data Cleaning

Checking Missing Values

```
In [117]: data.isnull().sum().plot.pie(figsize=(5.5,5.5))
```

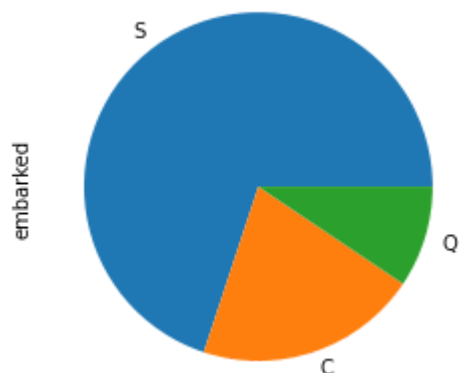
```
Out[117]: <matplotlib.axes._subplots.AxesSubplot at 0x1603eca0>
```



Embarked and Fare have low number of missing data

```
In [118]: pd.value_counts(data['embarked'],sort=True).plot.pie()
```

```
Out[118]: <matplotlib.axes._subplots.AxesSubplot at 0x15fae598>
```



Mostly Embarked from S(Southampton)

```
In [119]: data.embarked=data.embarked.fillna('S')
```

Filling Missing Data with most Frequent value

```
In [120]: data.fare.fillna(data.fare.median(),inplace=True)
```

Filling Fare missing value with Meadian

```
In [121]: data.insert(0,'cabinchar',data.cabin.str[0])
```

Extract First Character from Name

```
In [122]: titleList=[]
for i in range(0,1309):
    f=data.name.values[i].find('.')
    g=data.name.values[i].find(',')
    titleList.append(data.name.values[i][g+2:f])

data['titleList']=titleList
```

Extracted Titles from Name

```
In [123]: titleage =[]
for i in range(0,1309):
    if data.titleList[i] == 'Master':
        titleage.append(1)
    elif data.titleList[i] == 'Miss':
        titleage.append(2)
    else:
        titleage.append(3)

data['titleage']=titleage
```

Numerical weights to Selected Titles in order of Age

```
In [124]: data['Mother']=np.where([data['titleList']=='Miss']and[data['parch']>2], 1, 0).re
```

New column 'Mother' (They had higher chances of Survival)

```
In [125]: data.head(2)
```

Out[125]:

	cabinchar	pclass	survived	name	sex	age	sibsp	parch	ticket	fare	cabin
0	B	1	True	Allen, Miss. Elisabeth Walton	female	29.0000	0	0	24160	211.3375	B5
1	C	1	True	Allison, Master. Hudson Trevor	male	0.9167	1	2	113781	151.5500	C22 C26

```
In [126]: data['family']=data.parch+data.sibsp
data.head(2)
```

Out[126]:

	cabinchar	pclass	survived	name	sex	age	sibsp	parch	ticket	fare	cabin
0	B	1	True	Allen, Miss. Elisabeth Walton	female	29.0000	0	0	24160	211.3375	B5
1	C	1	True	Allison, Master. Hudson Trevor	male	0.9167	1	2	113781	151.5500	C22 C26



New column 'Family'

```
In [127]: data.drop(['boat', 'body', 'home_dest', 'ticket', 'name', 'cabin'],axis=1,inplace=True)
data.head(2)
```

Out[127]:

	cabinchar	pclass	survived	sex	age	sibsp	parch	fare	embarked	titleList	titleage
0	B	1	True	female	29.0000	0	0	211.3375	S	Miss	
1	C	1	True	male	0.9167	1	2	151.5500	S	Master	



Dropping NonRelevant or Used columns

Filling CabinChar and Age using KNN

```
In [128]: from sklearn.neighbors import KNeighborsClassifier,KNeighborsRegressor
```

```
In [129]: traincabin= data[data.cabinchar.notnull()]
testcabin = data[data.cabinchar.isnull()]
```

```
In [130]: clf=KNeighborsClassifier()
clf.fit(traincabin.drop(['cabinchar','age','embarked','titleList','sex'],axis=1),
data.cabinchar[data.cabinchar.isnull()] = clf.predict(testcabin.drop(['cabinchar',
```

<ipython-input-130-6c6c78f15607>:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
data.cabinchar[data.cabinchar.isnull()] = clf.predict(testcabin.drop(['cabinchar', 'age', 'embarked', 'titleList', 'sex'],axis=1))
```

```
In [131]: trainAge=data[data.age.notnull()]
testAge=data[data.age.isnull()]
```

```
In [132]: clfAge=KNeighborsRegressor()
clfAge.fit(trainAge.drop(['cabinchar','age','embarked','titleList','sex'],axis=1),
data.age[data.age.isnull()]=clfAge.predict(testAge.drop(['cabinchar','age','embar
```

<ipython-input-132-acd556305862>:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
data.age[data.age.isnull()]=clfAge.predict(testAge.drop(['cabinchar','age','embarked','titleList','sex'],axis=1))
```

Filled missing values with most probabalistic estimation using KNeighbors Algorithms

```
In [133]: 'NaN' in data.isnull()
```

Out[133]: False

Final Null Check

```
In [134]: data=pd.get_dummies(data)
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1309 entries, 0 to 1308
Data columns (total 40 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   pclass                                1309 non-null   int64
1   survived                             1309 non-null   bool
2   age                                   1309 non-null   float64
3   sibsp                                1309 non-null   int64
4   parch                                1309 non-null   int64
5   fare                                  1309 non-null   float64
6   titleage                             1309 non-null   int64
7   Mother                               1309 non-null   int32
8   family                               1309 non-null   int64
9   cabinchar_A                          1309 non-null   uint8
10  cabinchar_B                          1309 non-null   uint8
11  cabinchar_C                          1309 non-null   uint8
12  cabinchar_D                          1309 non-null   uint8
13  cabinchar_E                          1309 non-null   uint8
14  ...
```

Dummy Encoding of Categorical Data

```
In [135]: target=data.survived
```

Prediction of Survival given as target

Feature Extraction

```
In [136]: from sklearn.decomposition import PCA
data=pd.DataFrame(PCA(n_components=8,random_state=0).fit_transform(data.drop('survived',axis=1))
```

PCA for reducing large variance

```
In [137]: data.head(2)
```

Out[137]:

	0	1	2	3	4	5	6	7
0	177.780755	-9.766099	-2.826855	-0.636045	-0.079205	0.686435	-1.040244	-0.476214
1	116.746087	-35.058689	0.236209	0.331084	0.331235	-1.389845	-0.430874	-1.137526

```
In [138]: from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix, precision_score, recall_score, accuracy_score
tnx,tsx,tny,tsy=train_test_split(data,target,test_size=0.3,random_state=0)
```

Splitting Training and Testing

Training Models

```
In [139]: from sklearn.ensemble import VotingClassifier
from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
from sklearn.naive_bayes import GaussianNB
from sklearn.model_selection import GridSearchCV
```

All necessary packages imported

Decision Tree

```
In [140]: paramsdt = {'criterion':['gini','entropy'],'splitter':['random','best'],'max_depth':10}
griddt = GridSearchCV(DecisionTreeClassifier(random_state=0),param_grid=paramsdt,scoring='accuracy')
griddt.fit(tnx,tny)
griddt.best_params_
```

```
Out[140]: {'criterion': 'gini', 'max_depth': 9, 'splitter': 'random'}
```

```
In [141]: clfdt=DecisionTreeClassifier(criterion='entropy',splitter='random',max_depth=20,random_state=0)
clfdt.fit(tnx,tny)
preddt=clfdt.predict(tsx)
```

```
In [142]: print('Accuracy = ',format(accuracy_score(tsy,preddt),'.4f'))
print('F1 Score = ',format(f1_score(tsy,preddt),'.4f'))
print('Precision = ',format(precision_score(tsy,preddt),'.4f'))
print('Recall = ',format(recall_score(tsy,preddt),'.4f'))
```

```
Accuracy = 0.7430
F1 Score = 0.6406
Precision = 0.6716
Recall = 0.6122
```

Decision Tree Cross Validation with Best parameters and Results

Logistic Regression

In []:

```
In [143]: clfknn=KNeighborsClassifier()  
clfknn.fit(tnx,tny)  
predknn=clfknn.predict(tsx)
```

```
In [144]: print('Accuracy = ',format(accuracy_score(tsy,predknn),'.4f'))  
print('F1 Score = ',format(f1_score(tsy,predknn),'.4f'))  
print('Precision = ',format(precision_score(tsy,predknn),'.4f'))  
print('Recall = ',format(recall_score(tsy,predknn),'.4f'))
```

```
Accuracy = 0.7023  
F1 Score = 0.5895  
Precision = 0.6087  
Recall = 0.5714
```

Logistic Regression Cross Validation with Best parameters and Results

SVM

```
In [145]: paramssvm={'kernel':[ 'poly', 'rbf' ],'C':[4,5,6]}  
gridsvm=GridSearchCV(SVC(random_state=0),param_grid=paramssvm,cv=5)  
gridsvm.fit(tnx,tny)  
gridsvm.best_params_
```

```
Out[145]: {'C': 6, 'kernel': 'rbf'}
```

```
In [146]: clfsvm=SVC(C=6,kernel='rbf')  
clfsvm.fit(tnx,tny)  
predsvm=clfsvm.predict(tsx)
```

```
In [147]: print('Accuracy = ',format(accuracy_score(tsy,predsvm),'.4f'))  
print('F1 Score = ',format(f1_score(tsy,predsvm),'.4f'))  
print('Precision = ',format(precision_score(tsy,predsvm),'.4f'))  
print('Recall = ',format(recall_score(tsy,predsvm),'.4f'))
```

```
Accuracy = 0.7583  
F1 Score = 0.6865  
Precision = 0.6667  
Recall = 0.7075
```

SVM Cross Validation with Best parameters and Results

Ensembling of All Three

Voting Method and Results

```
In [148]: clfvot=VotingClassifier(estimators=[('dt',clfdt),('svm',clfsvm),('nb',clfknn)])
          clfvot.fit(tnx,tny)
          predvot=clfvot.predict(tsx)
```

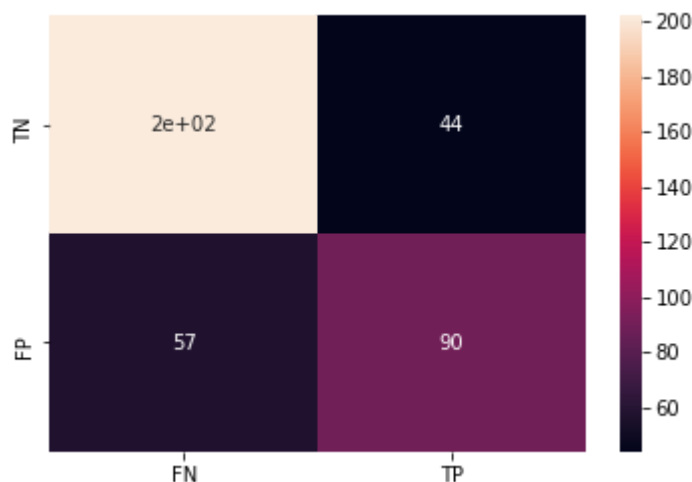
```
In [149]: print('Accuracy = ',format(accuracy_score(tsy,predvot),'.4f'))
          print('F1 Score = ',format(f1_score(tsy,predvot),'.4f'))
          print('Precision = ',format(precision_score(tsy,predvot),'.4f'))
          print('Recall = ',format(recall_score(tsy,predvot),'.4f'))
```

```
Accuracy = 0.7710
F1 Score = 0.6939
Precision = 0.6939
Recall = 0.6939
```

Confusion Matrix

```
In [150]: import seaborn as sns
          sns.heatmap(confusion_matrix(tsy,preddt),annot=True,xticklabels=['FN','TP'],yticklabels=['TN','FP'])
```

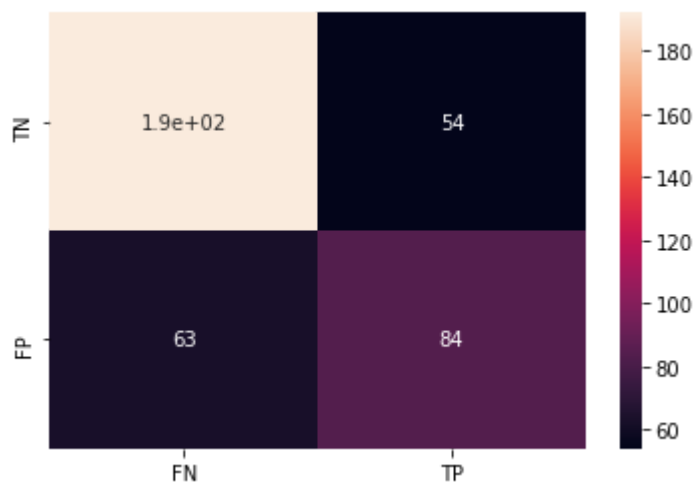
Out[150]: <matplotlib.axes._subplots.AxesSubplot at 0x1614f490>



CM for Decision Tree

```
In [151]: sns.heatmap(confusion_matrix(tsy,predknn),annot=True,xticklabels=['FN','TP'],ytic
```

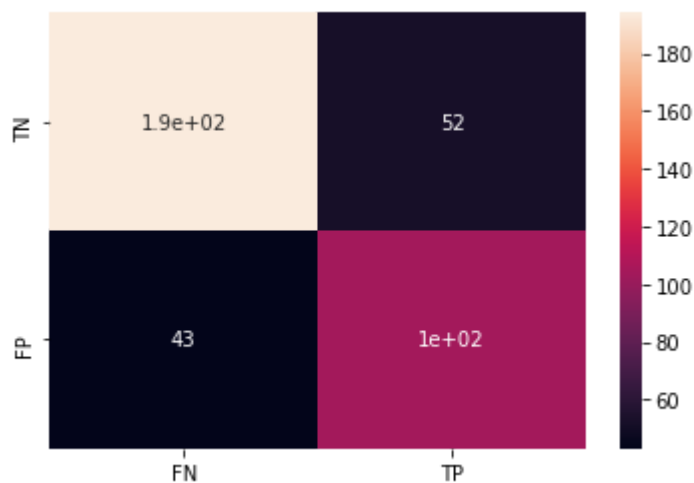
```
Out[151]: <matplotlib.axes._subplots.AxesSubplot at 0x16215970>
```



CM for KNN

```
In [152]: sns.heatmap(confusion_matrix(tsy,predsvm),annot=True,xticklabels=['FN','TP'],ytic
```

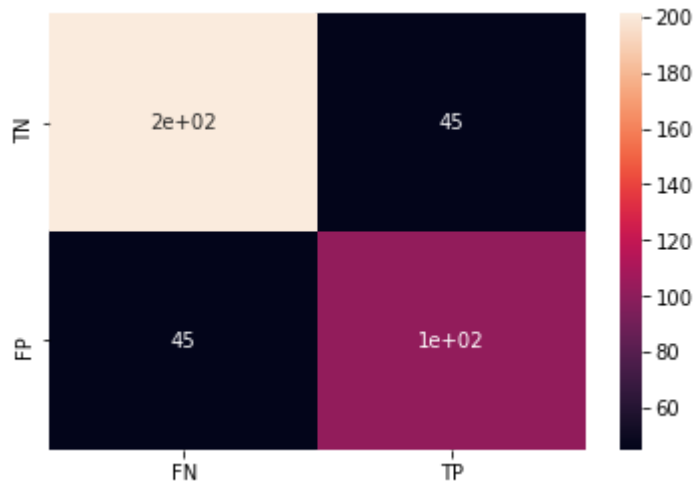
```
Out[152]: <matplotlib.axes._subplots.AxesSubplot at 0x1624eee0>
```



CM for SVM

```
In [153]: sns.heatmap(confusion_matrix(tsy,predvot),annot=True,xticklabels=['FN','TP'],ytic
```

```
Out[153]: <matplotlib.axes._subplots.AxesSubplot at 0x162a0d78>
```



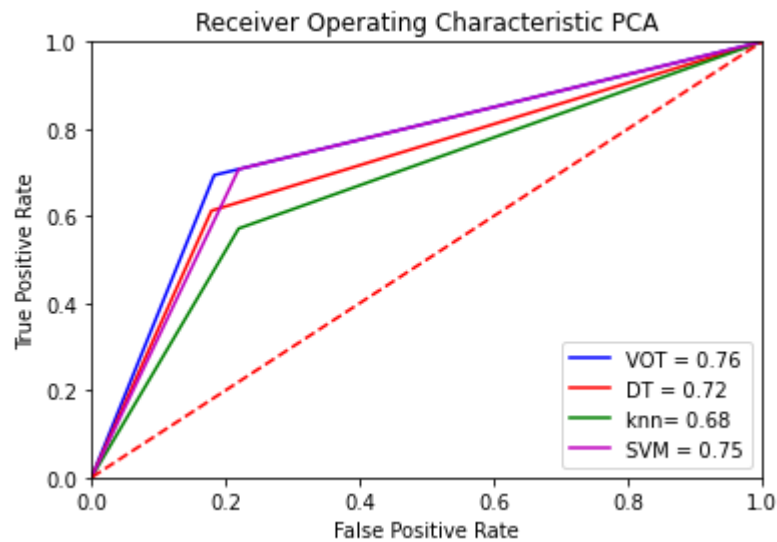
CM for Ensemble Learning

ROC Curve

```
In [154]: fpr,tpr,threshold=roc_curve(tsy,predvot)
roc_auc=auc(fpr,tpr)
fpr1,tpr1,threshold1=roc_curve(tsy,prededt)
roc_auc1=auc(fpr1,tpr1)
fpr2,tpr2,threshold2=roc_curve(tsy,predknn)
roc_auc2=auc(fpr2,tpr2)
fpr3,tpr3,threshold3=roc_curve(tsy,predsvm)
roc_auc3=auc(fpr3,tpr3)
```

Calculating Area Under ROC Curve

```
In [155]: import matplotlib.pyplot as plt
plt.title('Receiver Operating Characteristic PCA')
plt.plot(fpr, tpr, 'b', label = 'VOT = %0.2f' % roc_auc)
plt.plot(fpr1, tpr1, 'r', label = 'DT = %0.2f' % roc_auc1)
plt.plot(fpr2, tpr2, 'g', label = 'knn= %0.2f' % roc_auc2)
plt.plot(fpr2, tpr3, 'm', label = 'SVM = %0.2f' % roc_auc3)
plt.legend(loc = 'lower right')
plt.plot([0, 1], [0, 1], 'r--')
plt.xlim([0, 1])
plt.ylim([0, 1])
plt.ylabel('True Positive Rate')
plt.xlabel('False Positive Rate')
plt.show()
```



Ensemble Learning gives best Predictions and Model among ALL.

In []: 1

In []:

In []:

In []:

In []:

In []:

In []: