

## Titanic\_Dtree\_RF\_Prediction

```
import pandas as pd
import numpy as np
from sklearn import tree
from sklearn import preprocessing
```

### Loading Data and Data Treatment:

```
titanic_train = pd.read_csv("train.csv")
```

```
titanic_train.head()
```

Out[6]:

	PassengerId	Survived	Pclass	...	Fare	Cabin	Embarked
0	1	0	3 ...	7.2500	NaN	S	
1	2	1	1 ...	71.2833	C85	C	
2	3	1	3 ...	7.9250	NaN	S	
3	4	1	1 ...	53.1000	C123	S	
4	5	0	3 ...	8.0500	NaN	S	

[5 rows x 12 columns]

```
titanic_train.isnull().sum()
```

Out[7]:

PassengerId	0
Survived	0
Pclass	0
Name	0
Sex	0
Age	0
SibSp	0
Parch	0
Ticket	0

```
Fare      0
Cabin     687
Embarked   0
dtype: int64
```

```
titanic_train["Cabin"].mode()
```

```
Out[8]:
```

```
0    B96 B98
1    C23 C25 C27
2         G6
dtype: object
```

### **Encoding Categorical Variables**

```
label_encoder = preprocessing.LabelEncoder()

titanic_train["Sex"] = label_encoder.fit_transform(titanic_train["Sex"])

titanic_train["Embarked"] = label_encoder.fit_transform(titanic_train["Embarked"])
```

### **Random Forest Algorithm to find imp Variables**

```
from sklearn.ensemble import RandomForestClassifier
```

```
features = ['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', 'Fare', 'Embarked']
```

```
rf_model = RandomForestClassifier(n_estimators= 1000, max_features= 2, oob_score= True)
```

```
rf_model.fit(X = titanic_train[features], y = titanic_train["Survived"])
```

```
Out[17]:
```

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None,
                        criterion='gini', max_depth=None, max_features=2,
                        max_leaf_nodes=None, max_samples=None,
                        min_impurity_decrease=0.0, min_impurity_split=None,
                        min_samples_leaf=1, min_samples_split=2,
                        min_weight_fraction_leaf=0.0, n_estimators=1000,
```



```
with open("titanic_DTree1.dot","w") as f:
```

```
    f = tree.export_graphviz(tree_model,feature_names=['Sex','Age','Fare'], out_file= f)
```

```
print("DTree Model Accuracy: ", tree_model.score(X = predictors, y = titanic_train['Survived']))
```

***DTree Model Accuracy: 0.8706411698537683***

### **Testing the Model**

```
titanic_test = pd.read_csv("test.csv")
```

```
titanic_test.head()
```

Out[26]:

	PassengerId	Pclass	...	Fare	Embarked
0	892	3	...	7.8292	Q
1	893	3	...	7.0000	S
2	894	2	...	9.6875	Q
3	895	3	...	8.6625	S
4	896	3	...	12.2875	S

[5 rows x 10 columns]

```
titanic_test.isnull().sum()
```

Out[27]:

PassengerId	0
Pclass	0
Name	0
Sex	0
Age	0
SibSp	0
Parch	0

```
Ticket      0
Fare        0
Embarked    0
dtype: int64
```

```
titanic_test['Sex']=label_encoder.fit_transform(titanic_test['Sex'])
```

```
test_features = titanic_test[['Sex','Age','Fare']]
```

```
test_pred = tree_model.predict(X = test_features)
```

```
Predicted_output = pd.DataFrame({"PassengerId": titanic_test["PassengerId"], "Name":
titanic_test["Name"], "Survived": test_pred})
```

```
Predicted_output.to_csv("titanic_testdata_output1.csv", index= False)
```

## Decision Tree



### Inference:

1. Based on the importance value generated with Random forest algorithm, it is seen that the features '**Sex**', '**Age**' and '**Fare**' are more significant for decision tree generation.
2. Decision tree generated with these features and max-depth of 8 provides **87%** accuracy in classifying the record as Survived(Y/N) and also predicting the survival(Y/N) for any unseen record.