```
In [1]: import pandas as pd
          titanic_data = pd.read_csv('https://raw.githubusercontent.com/zekelabs/data-science-complete-tutorial/master/Data/ti
 In [2]:
          tanic-train.csv.txt')
 In [3]: titanic_data.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 891 entries, 0 to 890
          Data columns (total 12 columns):
                         891 non-null int64
          PassengerId
          Survived
                          891 non-null int64
          Pclass
                          891 non-null int64
                          891 non-null object
          Name
          Sex
                          891 non-null object
                          714 non-null float64
          Age
                          891 non-null int64
          SibSp
          Parch
                          891 non-null int64
         Ticket
                          891 non-null object
          Fare
                          891 non-null float64
          Cabin
                          204 non-null object
          Embarked
                          889 non-null object
          dtypes: float64(2), int64(5), object(5)
          memory usage: 83.6+ KB
In [23]:
         titanic_data.describe()
Out[23]:
                            Survived
                                       Pclass
                                                           SibSp
                                                                     Parch
                PassengerId
                                                   Age
                                                                                Fare
          count
                 891.000000
                           891.000000
                                    891.000000
                                             714.000000
                                                       891.000000
                                                                 891.000000
                                                                          891.000000
                                               29.699118
                 446.000000
                            0.383838
                                      2.308642
                                                         0.523008
                                                                   0.381594
                                                                            32.204208
           mean
                 257.353842
            std
                            0.486592
                                      0.836071
                                               14.526497
                                                         1.102743
                                                                   0.806057
                                                                            49.693429
                  1.000000
                            0.000000
                                      1.000000
                                               0.420000
                                                         0.000000
                                                                   0.000000
                                                                            0.000000
            min
                 223.500000
                            0.000000
                                      2.000000
                                               20.125000
                                                                            7.910400
            25%
                                                         0.000000
                                                                   0.000000
            50%
                 446.000000
                            0.000000
                                      3.000000
                                               28.000000
                                                         0.000000
                                                                   0.000000
                                                                            14.454200
            75%
                 668.500000
                            1.000000
                                      3.000000
                                               38.000000
                                                         1.000000
                                                                   0.000000
                                                                            31.000000
                 891.000000
                            1.000000
                                      3.000000
                                               80.000000
                                                         8.000000
                                                                   6.000000 512.329200
            max

    Splitting feature data & target data

           · Dropping less impotant columns
 In [4]: feature_data = titanic_data.drop(['Survived'], axis=1)
          target_data = titanic_data.Survived
 In [6]: feature_data.drop(['Name', 'Cabin', 'PassengerId'], axis=1, inplace=True)
         feature_data.drop(['Ticket'], axis=1, inplace=True)
 In [9]:
          feature_data.head()
In [10]:
Out[10]:
             Pclass
                     Sex Age SibSp Parch
                                            Fare Embarked
                    male 22.0
                                          7.2500
                                                       S
                 1 female 38.0
                                       0 71.2833
                                                       С
          1
                                 1
                 3 female 26.0
                                 0
                                       0 7.9250
                 1 female 35.0
                                       0 53.1000
                                       0 8.0500
                 3 male 35.0
In [11]: feature_data.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 891 entries, 0 to 890
          Data columns (total 7 columns):
          Pclass
                      891 non-null int64
                      891 non-null object
          Sex
                      714 non-null float64
          Age
          SibSp
                      891 non-null int64
          Parch
                      891 non-null int64
          Fare
                      891 non-null float64
                      889 non-null object
          Embarked
          dtypes: float64(2), int64(3), object(2)
          memory usage: 48.8+ KB
In [14]: | feature_data.Embarked.value_counts()
Out[14]: S
               644
               168
                77
          Name: Embarked, dtype: int64

    Seperating features based on types

In [13]: cat_cols = list(feature_data.select_dtypes(include=['object']))
          num_cols = list(feature_data.select_dtypes(exclude=['object']))

    Realizing missing values, create imputers

In [12]: from sklearn.impute import SimpleImputer
In [15]: | si_cat_cols = SimpleImputer(strategy='constant', fill_value='S')
          si_num_cols = SimpleImputer(strategy='median')
In [17]: | num_data = si_num_cols.fit_transform(feature_data[num_cols])
In [18]: | cat_data = si_cat_cols.fit_transform(feature_data[cat_cols])
In [19]: from sklearn.preprocessing import OneHotEncoder
In [20]: ohe = OneHotEncoder()
In [22]: cat_data_ohe = ohe.fit_transform(cat_data).toarray()
          Combine transformed data
In [24]: import numpy as np
          feature_data_tf = np.hstack([num_data, cat_data_ohe])
In [25]: from sklearn.model_selection import train_test_split
In [26]: | trainx, testx, trainy, testy = train_test_split(feature_data_tf, target_data)
In [30]: from sklearn.linear_model import LogisticRegression
          from sklearn.tree import DecisionTreeClassifier
          from sklearn.ensemble import RandomForestClassifier
In [34]: models = [ LogisticRegression(), DecisionTreeClassifier(), RandomForestClassifier()]
          trained_models = []
          for model in models:
              model.fit(trainx, trainy)
              trained_models.append(model)
          /home/awantik/anaconda3/lib/python3.7/site-packages/sklearn/linear_model/logistic.py:433: FutureWarning: Default solv
          er will be changed to 'lbfgs' in 0.22. Specify a solver to silence this warning.
            FutureWarning)
          /home/awantik/anaconda3/lib/python3.7/site-packages/sklearn/ensemble/forest.py:246: FutureWarning: The default value
          of n_estimators will change from 10 in version 0.20 to 100 in 0.22.
            "10 in version 0.20 to 100 in 0.22.", FutureWarning)
In [35]: for model in trained_models:
              print (model.score(testx, testy))
          0.7354260089686099
          0.7892376681614349
          0.7847533632286996
```

In []: