```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import cross_val_score
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score
from sklearn.preprocessing import MinMaxScaler
from sklearn.model_selection import train_test_split, KFold
from sklearn.model_selection import KFold, StratifiedKFold, cross_val_score
import matplotlib.pyplot as plt
In [2]: df=pd.read_csv('titanica.csv')
```

							001 00.0					
Out[2]:		PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cab
	0	NaN	0	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	Ná
	1	893.0	1	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	Na
	2	894.0	0	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	Nŧ
	3	895.0	0	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	Ná
	4	896.0	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	Nŧ
	•••		•••							•••		
	413	1305.0	0	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	8.0500	Na
	414	1306.0	1	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	C1
	415	1307.0	0	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	Né
	416	1308.0	0	3	Ware, Mr. Frederick	male	NaN	0	0	359309	8.0500	Na
	417	1309.0	0	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.3583	Ná

418 rows × 12 columns

In [3]: df.head()

Out[3]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Eml
	0	NaN	0	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	NaN	
	1	893.0	1	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	NaN	
	2	894.0	0	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	NaN	
	3	895.0	0	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	NaN	
	4	896.0	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	NaN	
1													

In [4]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 418 entries, 0 to 417
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	PassengerId	417 non-null	float64
1	Survived	418 non-null	int64
2	Pclass	418 non-null	int64
3	Name	418 non-null	object
4	Sex	418 non-null	object
5	Age	332 non-null	float64
6	SibSp	418 non-null	int64
7	Parch	418 non-null	int64
8	Ticket	418 non-null	object
9	Fare	417 non-null	float64
10	Cabin	91 non-null	object
11	Embarked	418 non-null	object
A.L.	C1+C4/2	\ '-+<4/4\	

dtypes: float64(3), int64(4), object(5)

memory usage: 39.3+ KB

In [5]: df.isnull().sum()

```
1
        PassengerId
Out[5]:
        Survived
                         0
        Pclass
                         0
        Name
                         0
        Sex
                         0
                        86
        Age
                         0
        SibSp
        Parch
                         0
        Ticket
                         0
        Fare
                         1
        Cabin
                       327
        Embarked
                         0
        dtype: int64
```

In [6]: df.dropna(axis=0, how='any', subset=None, inplace=False)

												1.1750 E:  2.3750 B! B! B! B! B! B! C.3750 B: B!	
ut[6]:		PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cab	
	12	904.0	1	1	Snyder, Mrs. John Pillsbury (Nelle Stevenson)	female	23.0	1	0	21228	82.2667	B₄	
	14	906.0	1	1	Chaffee, Mrs. Herbert Fuller (Carrie Constance	female	47.0	1	0	W.E.P. 5734	61.1750	E:	
	24	916.0	1	1	Ryerson, Mrs. Arthur Larned (Emily Maria Borie)	female	48.0	1	3	PC 17608	262.3750	B! B(	
	26	918.0	1	1	Ostby, Miss. Helene Ragnhild	female	22.0	0	1	113509	61.9792	B:	
	28	920.0	0	1	Brady, Mr. John Bertram	male	41.0	0	0	113054	30.5000	Αí	
	•••										•••		
	404	1296.0	0	1	Frauenthal, Mr. Isaac Gerald	male	43.0	1	0	17765	27.7208	D₄	
	405	1297.0	0	2	Nourney, Mr. Alfred (Baron von Drachstedt")"	male	20.0	0	0	SC/PARIS 2166	13.8625	D:	
	407	1299.0	0	1	Widener, Mr. George Dunton	male	50.0	1	1	113503	211.5000	C{	
	411	1303.0	1	1	Minahan, Mrs. William Edward (Lillian E Thorpe)	female	37.0	1	0	19928	90.0000	C7	
	414	1306.0	1	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	C1(	

87 rows × 12 columns

```
In [7]: y=df['Survived']
x=df.drop(['Survived'],axis=1)
y
```

```
0
1
Out[7]:
                  1
          2
                  0
          3
                  0
          4
                  1
                  0
          413
          414
                  1
          415
          416
                  0
          417
                  0
```

Name: Survived, Length: 418, dtype: int64

In [8]: x.head()

Out[8]:		PassengerId	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
	0	NaN	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	NaN	Q
	1	893.0	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	NaN	S
	2	894.0	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	NaN	Q
	3	895.0	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	NaN	S
	4	896.0	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	NaN	S

```
In [9]: df=df.drop(['Name','Sex','Embarked','Ticket','Cabin'],axis=1)
In [10]: df
```

Out[10]:		PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare				
	0	NaN	0	3	34.5	0	0	7.8292				
	1	893.0	1	3	47.0	1	0	7.0000				
	2	894.0	0	2	62.0	0	0	9.6875				
	3	895.0	0	3	27.0	0	0	8.6625				
	4	896.0	1	3	22.0	1	1	12.2875				
	•••	•••	•••			•••						
	413	1305.0	0	3	NaN	0	0	8.0500				
	414	1306.0	1	1	39.0	0	0	108.9000				
	415	1307.0	0	3	38.5	0	0	7.2500				
	416	1308.0	0	3	NaN	0	0	8.0500				
	417	1309.0	0	3	NaN	1	1	22.3583				
In [11]:	418 rows × 7 columns  kf=KFold(n_splits=5) print("Data is splitinto following number of folds:") kf.get n splits(df)											
t[ <b>11</b> ]:	Data 5	is splitin	to follow	ving nu	ımber	of fol	ds:					
[12]:	print for t c	ecisionTre ("Accuraci rain_index lf.fit(df. red=clf.pr rint(round	es for each, test_indicated in the second in	ach fol ndex in in_inde .iloc[t	kf.s x], y est_i	split(d /.iloc[ index,:	f,y): train_ ])					
	Accuracies for each fold of data are: 1.0 1.0											

1.0 1.0 1.0

In [13]: print(df)

```
PassengerId Survived Pclass
                                     Age SibSp Parch
                                                             Fare
0
                                     34.5
             NaN
                         0
                                  3
                                               0
                                                      0
                                                           7.8292
                                  3
1
           893.0
                         1
                                     47.0
                                               1
                                                      0
                                                           7.0000
2
           894.0
                         0
                                  2 62.0
                                               0
                                                      0
                                                           9.6875
3
           895.0
                         0
                                  3
                                     27.0
                                               0
                                                      0
                                                           8.6625
4
                         1
                                  3
                                     22.0
           896.0
                                               1
                                                      1
                                                          12.2875
                                      . . .
413
          1305.0
                         0
                                  3
                                      NaN
                                               0
                                                      0
                                                           8.0500
                         1
                                  1
                                     39.0
                                                      0
                                                         108.9000
414
          1306.0
                                               0
415
          1307.0
                         0
                                  3
                                     38.5
                                               0
                                                      0
                                                           7.2500
416
          1308.0
                         0
                                  3
                                     NaN
                                               0
                                                      0
                                                           8.0500
417
          1309.0
                         0
                                  3
                                      NaN
                                               1
                                                      1
                                                          22.3583
```

[418 rows x 7 columns]

```
In [14]: cv =cross_val_score(DecisionTreeClassifier(),df,y,cv=5,scoring='accuracy')
    print("Accuracies of all the folds after the cross validation are:")
    cv
```

Accuracies of all the folds after the cross validation are: array([1., 1., 1., 1.])

In [ ]:

Out[14]: