

10) Program to show Cross Validation

#This program shows Cross Validation

```
print('This program shows Cross Validation')
```

```
from sklearn import datasets
```

```
import pandas as pd
```

```
iris=datasets.load_iris()
```

df will fold dataset as a table

```
df=pd.DataFrame(
```

```
    iris.data,
```

```
    columns=iris.feature_names
```

```
)
```

#labels are assigned to df[target] table or array

```
df['target']=pd.Series(
```

```
    iris.target
```

```
)
```

```
from sklearn.model_selection import train_test_split
```

```
from sklearn.tree import DecisionTreeClassifier
```

```
from sklearn.metrics import accuracy_score
```

```
from sklearn.model_selection import ShuffleSplit
```

```
from sklearn.model_selection import cross_validate
```

```
clf=DecisionTreeClassifier()
```

```
rs=ShuffleSplit(n_splits=100,test_size=0.3)
```

```
x=df[iris.feature_names]
```

```
y=df['target']
```

```
print('Accuracy for different iterations')
```

```
cv_results=cross_validate(
```

```
    clf,x,y,cv=rs,scoring='accuracy'
```

```
    ) # put scoring='precision_macro' scoring='recall_macro'
```

```
accuracy_scores=pd.Series(cv_results['test_score'])
```

```
print(accuracy_scores)
```

```
print('Precision for different iterations')
```

```
cv_results=cross_validate(
```

```
    clf,x,y,cv=rs,scoring='precision_macro'
```

```
    ) # put scoring='precision_macro' scoring='recall_macro'
```

```
accuracy_scores=pd.Series(cv_results['test_score'])
```

```
print(accuracy_scores)
```

```
print('Recall for different iterations')
```

```
cv_results=cross_validate(
```

```
    clf,x,y,cv=rs,scoring='recall_macro'
```

```
    ) # put scoring='precision_macro' scoring='recall_macro'
```

```
accuracy_scores=pd.Series(cv_results['test_score'])
```

```
print(accuracy_scores)
```

Output

This program shows Cross Validation

Accuracy for different iterations

0 0.955556

1 0.977778

2 0.977778

3 0.955556

4 0.977778

...

95 0.955556

96 0.933333

97 0.933333

98 0.933333

99 0.977778

Length: 100, dtype: float64

Precision for different iterations

0 0.950549

1 0.948718

2 0.981481

3 0.979167

4 0.958333

...

95 0.936508

96 0.916340

97 0.921569

98 0.984127

99 0.941176

Length: 100, dtype: float64

Recall for different iterations

0 0.972222

1 0.962963

```
2  0.944079
3  0.933333
4  0.911111
...
95 0.948718
96 0.952137
97 0.907407
98 0.944056
99 0.917065
Length: 100, dtype: float64
>>>
```

11) #Program on regression

```
import numpy as np
```

```
n=200 #200 samples
```

```
height_pop1_f=np.random.normal(loc=155,scale=10,size=n)
```

```
height_pop1_m=np.random.normal(loc=175,scale=5,size=n)
```

```
height_pop2_f=np.random.normal(loc=165,scale=10,size=n)
```

```
height_pop2_m=np.random.normal(loc=185,scale=5,size=n)
```

```
height_f=np.concatenate([height_pop1_f,height_pop2_f])
```

```
height_m=np.concatenate([height_pop1_m,height_pop2_m])
```

```
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
from sklearn.tree import export_text
```

```
from sklearn.tree import DecisionTreeRegressor
```

```
from sklearn.model_selection import train_test_split
```

```

df_height=pd.DataFrame(
    {
        'Gender':[1 for i in range(height_f.size)]+
                [2 for i in range(height_m.size)],
        'Height':np.concatenate((height_f,height_m))
    }
)
fig, ax=plt.subplots(1,1,figsize=(10,5))

df_height[df_height['Gender']==1]['Height'].plot(
    label='Female',kind='hist',
    bins=10,alpha=0.7,ax=ax
)
df_height[df_height['Gender']==2]['Height'].plot(
    label='Male',kind='hist',
    bins=10,alpha=0.7,ax=ax
)
ax.legend()
fig.show()

# to calculate mean and median of height
df_height.groupby('Gender')[['Height']].agg([np.mean,np.median]).round(1)
df_train,df_test=train_test_split(df_height,test_size=0.3)
x_train,x_test=df_train[['Gender']],df_test[['Gender']]
y_train,y_test=df_train['Height'],df_test['Height']

for criterion in['mse','mae']:
    rgrsr=DecisionTreeRegressor(criterion=criterion)
    rgrsr.fit(x_train,y_train)
    print(f'criterion={criterion}:\n')
    print(export_text(rgrsr,feature_names=['Gender'],spacing=3,decimals=1))

```

```
print('Program Executed successfully')
```

OUTPUT

criterion=mse:

```
|--- Gender <= 1.5  
| |--- value: [159.9]  
|--- Gender > 1.5  
| |--- value: [180.1]
```

criterion=mae:

```
|--- Gender <= 1.5  
| |--- value: [160.0]  
|--- Gender > 1.5  
| |--- value: [180.2]
```

Program Executed successfully

