

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

```
In [24]: df = pd.read_csv('cars.csv')
```

```
In [25]: df.head()
```

```
Out[25]:
   passedemissions  mpg  displacement  horsepower  cylinders  weight  acceleration  modelyear      carname
0            False  18.0        307.0       130.0         8.0    1.7520        12.0        70.0  chevrolet chevelle malibu
1            False  15.0        350.0       165.0         8.0    1.8465        11.5        70.0      buick skylark 320
2            False  18.0        318.0       150.0         8.0    1.7180        11.0        70.0  plymouth satellite
3            False  16.0        304.0       150.0         8.0    1.7165        12.0        70.0      amc rebel sst
4            False  17.0        302.0       140.0         8.0    1.7245        10.5        70.0      ford torino
```

```
In [26]: df.dropna(inplace=True)
```

```
In [27]: df.drop(['passedemissions', 'carname', 'cylinders', 'acceleration', 'modelyear'], axis=1, inplace=True)
```

```
In [28]: #creating a function to replace all numeric values in mpg column as high, low or medium fuel efficiency
def change_mpg_to_effi(cols):
    a=cols[0]
    if a<=17:
        return 'low'
    elif (a>17 and a<=29):
        return 'medium'
    else:
        return 'high'
```

```
In [29]: #applying the above function on that column
df['mpg']=df[['mpg', 'displacement']].apply(change_mpg_to_effi, axis=1)
```

```
In [30]: df.head()
```

```
Out[30]:
   mpg  displacement  horsepower  weight
0  medium     307.0       130.0  1.7520
1    low     350.0       165.0  1.8465
2  medium     318.0       150.0  1.7180
3    low     304.0       150.0  1.7165
4    low     302.0       140.0  1.7245
```

```
In [31]: df['mpg'].value_counts()
```

```
Out[31]: medium    198
low       99
high      95
Name: mpg, dtype: int64
```

```
In [32]: #visualizing the dataframe
sns.pairplot(df,hue='mpg')
```

```
Out[32]: <seaborn.axisgrid.PairGrid at 0x26ab7d35780>
```

```
In [33]: from sklearn.model_selection import train_test_split
```

```
In [34]: #splitting our data into test and training
X=df.drop('mpg',axis=1)
y=df['mpg']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=101)
```

```
In [35]: #importing support vector machine model from scikit learn
from sklearn.svm import SVC
```

```
In [36]: svc_model = SVC()
```

```
In [38]: #fitting our model on training data
svc_model.fit(X_train,y_train)

C:\Users\Chandan\Anaconda3\lib\site-packages\sklearn\svm\base.py:196: FutureWarning: The default value of gamma will change from 'auto' to 'scale' in version 0.22 to account better for unscaled features. Set gamma explicitly to 'auto' or 'scale' to avoid this warning.
  "avoid this warning.", FutureWarning)
```

```
Out[38]: SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
           decision_function_shape='ovr', degree=3, gamma='auto_deprecated',
           kernel='rbf', max_iter=-1, probability=False, random_state=None,
           shrinking=True, tol=0.001, verbose=False)
```

```
In [39]: #predicting categories for test data
predictions=svc_model.predict(X_test)
```

```
In [40]: #analyzing our findings
from sklearn.metrics import classification_report, confusion_matrix
```

```
In [41]: print(confusion_matrix(y_test,predictions))

[[11  0  9]
 [ 0 18 13]
 [ 4  0 63]]
```

```
In [42]: print(classification_report(y_test,predictions))
```

	precision	recall	f1-score	support
high	0.73	0.55	0.63	20
low	1.00	0.58	0.73	31
medium	0.74	0.94	0.83	67
micro avg	0.78	0.78	0.78	118
macro avg	0.82	0.69	0.73	118
weighted avg	0.81	0.78	0.77	118