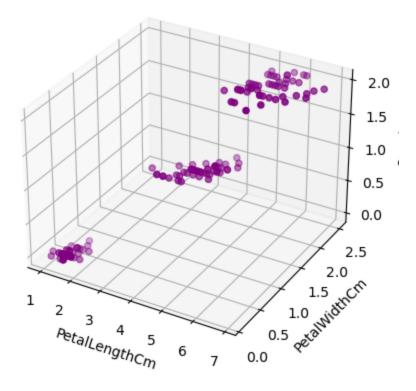
```
In [1]:
         import numpy as np
         import pandas as pd
         from sklearn.cluster import KMeans
         import matplotlib.pyplot as plt
         import seaborn as sns
         df = pd.read csv(r'C:\Users\hp\Desktop\CodSoft\IRIS.csv')
In [4]:
         df.head()
In [5]:
             sepal_length sepal_width petal_length petal_width
Out[5]:
                                                                  species
         0
                     5.1
                                  3.5
                                               1.4
                                                            0.2 Iris-setosa
         1
                                  3.0
                     4.9
                                               1.4
                                                            0.2 Iris-setosa
         2
                     4.7
                                  3.2
                                               1.3
                                                           0.2 Iris-setosa
         3
                     4.6
                                  3.1
                                               1.5
                                                            0.2 Iris-setosa
         4
                     5.0
                                  3.6
                                               1.4
                                                            0.2 Iris-setosa
         df['species'], categories =pd.factorize(df['species'])
In [6]:
         df.head()
             sepal length sepal width petal length petal width species
Out[6]:
         0
                     5.1
                                  3.5
                                               1.4
                                                            0.2
                                                                     0
         1
                     4.9
                                  3.0
                                               1.4
                                                            0.2
                                                                     0
         2
                     4.7
                                  3.2
                                                            0.2
                                               1.3
                                                                     0
         3
                     4.6
                                  3.1
                                               1.5
                                                            0.2
                                                                     0
                                                            0.2
                                                                     0
                     5.0
                                  3.6
                                               1.4
         4
         df.describe
In [7]:
         <bound method NDFrame.describe of</pre>
                                                      sepal_length sepal_width petal_length petal
Out[7]:
         _width species
         0
                                                                                    0
                         5.1
                                        3.5
                                                        1.4
                                                                       0.2
                         4.9
                                        3.0
                                                                       0.2
                                                                                    0
         1
                                                        1.4
                                                                       0.2
         2
                         4.7
                                        3.2
                                                        1.3
                                                                                    0
         3
                         4.6
                                        3.1
                                                        1.5
                                                                       0.2
                                                                                    0
         4
                         5.0
                                                        1.4
                                                                       0.2
                                        3.6
                                                                                    0
                          . . .
                                                                        . . .
                                        . . .
                                                         . . .
                                                                                  . . .
         . .
                                                                                    2
         145
                         6.7
                                        3.0
                                                        5.2
                                                                       2.3
         146
                         6.3
                                        2.5
                                                        5.0
                                                                       1.9
                                                                                    2
                                                                                    2
         147
                         6.5
                                        3.0
                                                        5.2
                                                                       2.0
                                                                       2.3
                                                                                    2
         148
                         6.2
                                        3.4
                                                        5.4
                         5.9
                                        3.0
                                                                                    2
         149
                                                        5.1
                                                                       1.8
         [150 rows x 5 columns]>
         df.isna().sum()
In [8]:
```

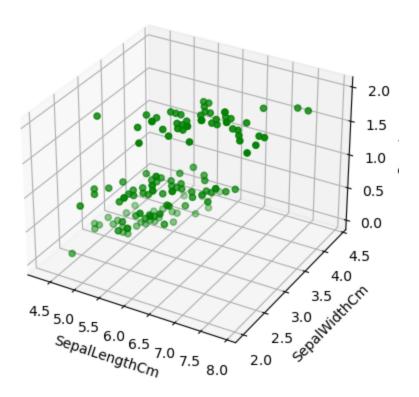
```
sepal_length
Out[8]:
         sepal_width
         petal_length
         petal_width
                         0
         species
         dtype: int64
In [25]: from mpl_toolkits.mplot3d import Axes3D
         fig = plt.figure()
         ax = fig.add_subplot(111, projection='3d')
         ax.scatter(df.petal_length, df.petal_width, df.species,color='purple')
         ax.set_xlabel('PetalLengthCm')
         ax.set_ylabel('PetalWidthCm')
         ax.set_zlabel('Species')
         plt.title('3D Scatter Plot')
         plt.show()
```

3D Scatter Plot

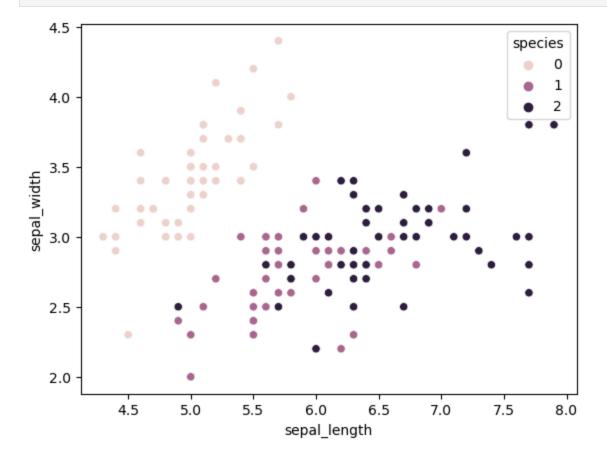


```
In [27]: from mpl_toolkits.mplot3d import Axes3D
    fig = plt.figure()
    ax = fig.add_subplot(111, projection='3d')
    ax.scatter(df.sepal_length, df.sepal_width, df.species,color='green')
    ax.set_xlabel('SepalLengthCm')
    ax.set_ylabel('SepalWidthCm')
    ax.set_zlabel('Species')
    plt.title('3D Scatter Plot')
    plt.show()
```

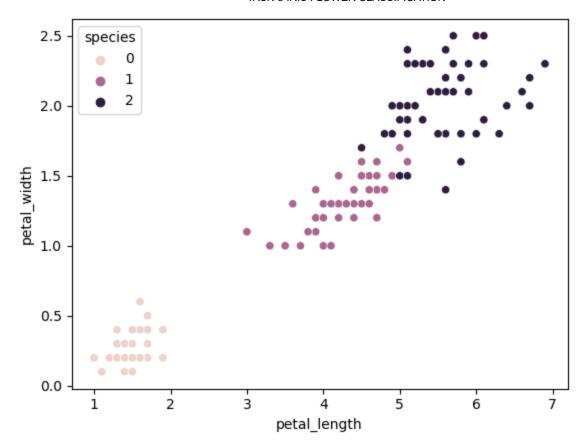
3D Scatter Plot



In [43]: sns.scatterplot(data=df, x="sepal_length", y="sepal_width",hue="species");



In [12]: sns.scatterplot(data=df, x="petal_length", y="petal_width",hue="species");



```
In [66]: iris = range(1,10)
    yash=[]

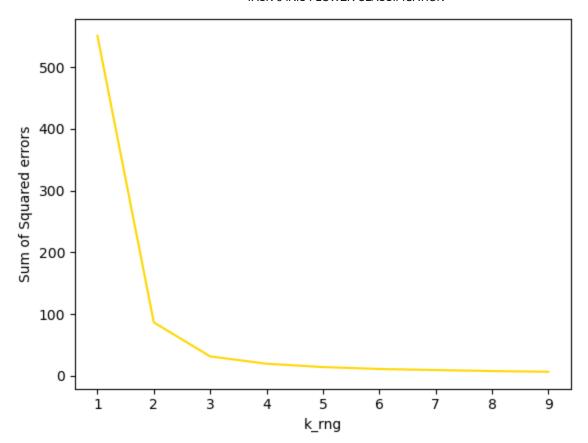
for k in iris:
    km = KMeans(n_clusters=k)
    km.fit(df[[ 'petal_length', 'petal_width']])
    yash.append(km.inertia_)
```

```
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: FutureWarnin
g: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value
of `n init` explicitly to suppress the warning
  super()._check_params_vs_input(X, default_n_init=10)
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1436: UserWarning:
KMeans is known to have a memory leak on Windows with MKL, when there are less chunks
than available threads. You can avoid it by setting the environment variable OMP NUM
THREADS=1.
 warnings.warn(
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: FutureWarnin
g: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value
of `n init` explicitly to suppress the warning
  super(). check params vs input(X, default n init=10)
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1436: UserWarning:
KMeans is known to have a memory leak on Windows with MKL, when there are less chunks
than available threads. You can avoid it by setting the environment variable OMP_NUM_
THREADS=1.
 warnings.warn(
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: FutureWarnin
g: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value
of `n_init` explicitly to suppress the warning
 super()._check_params_vs_input(X, default_n_init=10)
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\ kmeans.py:1436: UserWarning:
KMeans is known to have a memory leak on Windows with MKL, when there are less chunks
than available threads. You can avoid it by setting the environment variable OMP_NUM_
THREADS=1.
 warnings.warn(
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\ kmeans.py:1412: FutureWarnin
g: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value
of `n_init` explicitly to suppress the warning
  super(). check params vs input(X, default n init=10)
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1436: UserWarning:
KMeans is known to have a memory leak on Windows with MKL, when there are less chunks
than available threads. You can avoid it by setting the environment variable OMP_NUM_
THREADS=1.
 warnings.warn(
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: FutureWarnin
g: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value
of `n_init` explicitly to suppress the warning
  super()._check_params_vs_input(X, default_n_init=10)
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\ kmeans.py:1436: UserWarning:
KMeans is known to have a memory leak on Windows with MKL, when there are less chunks
than available threads. You can avoid it by setting the environment variable OMP_NUM_
THREADS=1.
 warnings.warn(
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: FutureWarnin
g: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value
of `n_init` explicitly to suppress the warning
  super()._check_params_vs_input(X, default_n_init=10)
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1436: UserWarning:
KMeans is known to have a memory leak on Windows with MKL, when there are less chunks
than available threads. You can avoid it by setting the environment variable OMP_NUM_
THREADS=1.
 warnings.warn(
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\ kmeans.py:1412: FutureWarnin
g: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value
of `n_init` explicitly to suppress the warning
  super()._check_params_vs_input(X, default_n_init=10)
C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\ kmeans.py:1436: UserWarning:
KMeans is known to have a memory leak on Windows with MKL, when there are less chunks
```

```
than available threads. You can avoid it by setting the environment variable OMP_NUM_
         THREADS=1.
           warnings.warn(
         C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: FutureWarnin
         g: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value
         of `n init` explicitly to suppress the warning
           super(). check params vs input(X, default n init=10)
         C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1436: UserWarning:
         KMeans is known to have a memory leak on Windows with MKL, when there are less chunks
         than available threads. You can avoid it by setting the environment variable OMP_NUM_
         THREADS=1.
           warnings.warn(
         C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: FutureWarnin
         g: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value
         of `n init` explicitly to suppress the warning
           super()._check_params_vs_input(X, default_n_init=10)
         C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1436: UserWarning:
         KMeans is known to have a memory leak on Windows with MKL, when there are less chunks
         than available threads. You can avoid it by setting the environment variable OMP_NUM_
         THREADS=1.
           warnings.warn(
In [67]: yash
         [550.6434666666669,
Out[67]:
          86.40394533571003,
          31.38775897435897,
          19.499400899685114,
          13.933308757908755,
          11.090603568447506,
          9.219714009661839,
          7.631802244955954,
```

[<matplotlib.lines.Line2D at 0x27c758cd810>] Out[45]:

6.5929383963862245]

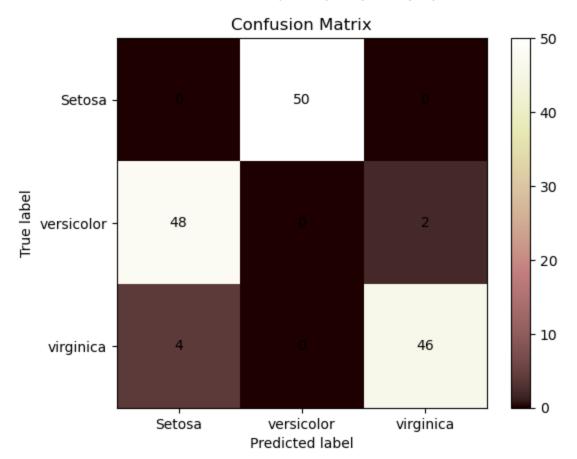


```
km = KMeans(n_clusters=3,random_state=0,)
In [16]:
       y_predicted = km.fit_predict(df[['petal_length','petal_width']])
       y_predicted
       C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: FutureWarnin
       g: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value
       of `n_init` explicitly to suppress the warning
         super()._check_params_vs_input(X, default_n_init=10)
       C:\Users\hp\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1436: UserWarning:
       KMeans is known to have a memory leak on Windows with MKL, when there are less chunks
       than available threads. You can avoid it by setting the environment variable OMP_NUM_
       THREADS=1.
         warnings.warn(
       Out[16]:
             0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 2, 0, 0, 0, 0,
             0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
             2, 2, 2, 2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2, 2, 0, 2, 2, 2, 2,
             2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2])
       df['cluster']=y_predicted
In [17]:
       df.head(150)
```

Out[17]:		sepal_length	sepal_width	petal_length	petal_width	species	cluster
	0	5.1	3.5	1.4	0.2	0	1
	1	4.9	3.0	1.4	0.2	0	1
	2	4.7	3.2	1.3	0.2	0	1
	3	4.6	3.1	1.5	0.2	0	1
	4	5.0	3.6	1.4	0.2	0	1
	•••						
	145	6.7	3.0	5.2	2.3	2	2
	146	6.3	2.5	5.0	1.9	2	2
	147	6.5	3.0	5.2	2.0	2	2
	148	6.2	3.4	5.4	2.3	2	2
	149	5.9	3.0	5.1	1.8	2	2

150 rows × 6 columns

```
In [18]: from sklearn.metrics import confusion matrix
         cm = confusion_matrix(df.species, df.cluster)
         array([[ 0, 50, 0],
Out[18]:
                [48, 0, 2],
                [ 4, 0, 46]], dtype=int64)
In [63]: true_labels = df.species
         predicted_labels= df.cluster
         cm = confusion_matrix(true_labels, predicted_labels)
         class_labels = ['Setosa', 'versicolor', 'virginica']
         # Plot confusion matrix
         plt.imshow(cm, interpolation='nearest', cmap=plt.cm.pink)
         plt.title('Confusion Matrix')
         plt.colorbar()
         tick_marks = np.arange(len(class_labels))
         plt.xticks(tick_marks, class_labels)
         plt.yticks(tick_marks, class_labels)
         # Fill matrix with values
         for i in range(len(class_labels)):
             for j in range(len(class_labels)):
                 plt.text(j, i, str(cm[i][j]), ha='center', va='center', color='black')
         plt.xlabel('Predicted label')
         plt.ylabel('True label')
         plt.show()
```



In []: