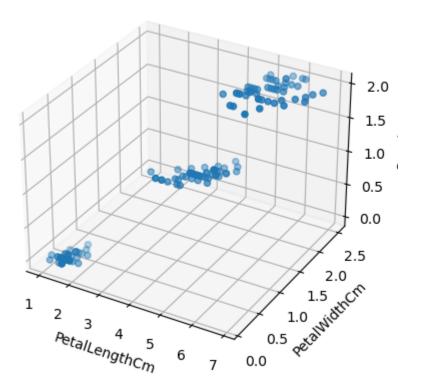
```
In [31]:
               import numpy as np
               import pandas as pd
               from sklearn.cluster import KMeans
               import matplotlib.pyplot as plt
               import seaborn as sns
In [34]:
            In df = pd.read_csv("C:/Users/deepa/Downloads/archive (2)/IRIS.csv")
               df.head()
    Out[34]:
                   sepal_length sepal_width petal_length petal_width
                                                                     species
                           5.1
                                       3.5
                                                    1.4
                                                                   Iris-setosa
                1
                           4.9
                                       3.0
                                                    1.4
                                                               0.2 Iris-setosa
                2
                           4.7
                                       3.2
                                                    1.3
                                                               0.2 Iris-setosa
                3
                           4.6
                                       3.1
                                                    1.5
                                                                   Iris-setosa
                           5.0
                                       3.6
                                                    1.4
                                                               0.2 Iris-setosa
In [35]:
               df['species'] , categories = pd.factorize(df['species'])
               df.head()
    Out[35]:
                   sepal_length sepal_width petal_length petal_width species
                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
                                                    1.3
                                                               0.2
                                                                         0
                3
                                                                         0
                                       3.1
                                                    1.5
                                                               0.2
                           4.6
                                                               0.2
                                                                         0
                           5.0
                                       3.6
                                                    1.4
In [36]:
              df.describe
    Out[36]: <bound method NDFrame.describe of</pre>
                                                            sepal_length sepal_width petal_l
                       petal width
                                      species
               ength
               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
                                                              1.3
                                                                             0.2
                                                                                          0
               3
                               4.6
                                              3.1
                                                              1.5
                                                                             0.2
                                                                                          0
               4
                               5.0
                                              3.6
                                                              1.4
                                                                             0.2
                                                                                          0
                                                                              . . .
               . .
               145
                               6.7
                                              3.0
                                                              5.2
                                                                             2.3
                                                                                          2
               146
                               6.3
                                              2.5
                                                              5.0
                                                                             1.9
                                                                                          2
                                                                                          2
               147
                               6.5
                                              3.0
                                                              5.2
                                                                             2.0
                                                                                          2
               148
                               6.2
                                              3.4
                                                              5.4
                                                                             2.3
               149
                               5.9
                                              3.0
                                                              5.1
                                                                             1.8
                                                                                          2
               [150 rows x 5 columns]>
```

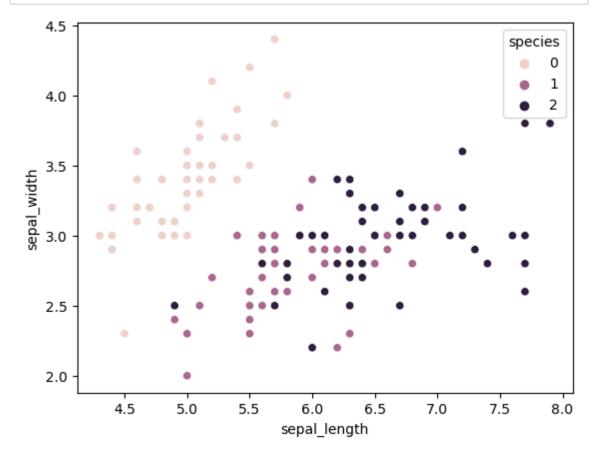
```
    df.isna().sum()

In [37]:
   Out[37]: sepal length
                              0
             sepal_width
                              0
             petal_length
                              0
             petal_width
                              0
                              0
             species
             dtype: int64
          ▶ from mpl_toolkits.mplot3d import Axes3D
In [38]:
             fig = plt.figure()
             ax = fig.add_subplot(111,projection = '3d')
             ax.scatter(df.petal_length ,df.petal_width , df.species)
             ax.set_xlabel('PetalLengthCm')
             ax.set_ylabel('PetalWidthCm')
             ax.set_zlabel('Species')
             plt.title('3D Scatter plot Example')
             plt.show()
```

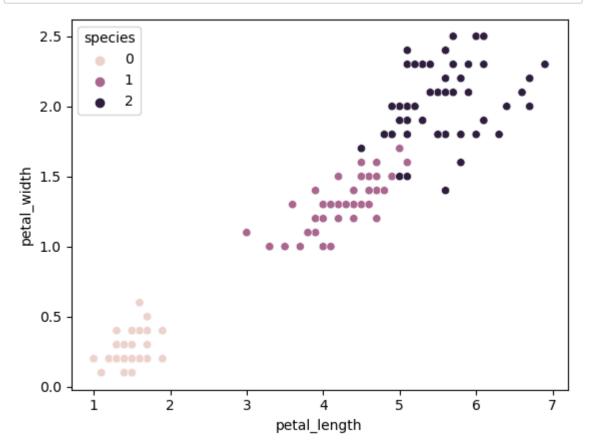
3D Scatter plot Example



In [39]: ▶ sns.scatterplot(data = df, x="sepal_length" ,y ="sepal_width",hue ="species



In [40]: N sns.scatterplot(data= df , x="petal_length", y="petal_width", hue="species")



```
In [46]: M k_rng = range(1,10)
sse=[]

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

```
Task3 - Jupyter Notebook
C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster\ kmeans.py:87
0: FutureWarning: The default value of `n_init` will change from 10 to 'a
uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning
  warnings.warn(
C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster\ kmeans.py:138
2: UserWarning: KMeans is known to have a memory leak on Windows with MK
L, 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\deepa\anaconda3\lib\site-packages\sklearn\cluster\ kmeans.py:87
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C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:138
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C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster\ kmeans.py:138
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 warnings.warn(
C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster\ kmeans.py:87
0: FutureWarning: The default value of `n_init` will change from 10 to 'a
uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning
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C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:138
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setting the environment variable OMP_NUM_THREADS=1.
 warnings.warn(
C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:87
0: FutureWarning: The default value of `n_init` will change from 10 to 'a
uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning
```

warnings.warn(

C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:138

2: UserWarning: KMeans is known to have a memory leak on Windows with MK

L, 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\deepa\anaconda3\lib\site-packages\sklearn\cluster\ kmeans.py:87

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uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning
warnings.warn(

C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster\ kmeans.py:138

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warnings.warn(

C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87

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warnings.warn(

C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster\ kmeans.py:138

2: UserWarning: KMeans is known to have a memory leak on Windows with MK

L, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

In [47]: ▶ sse

Out[47]: [550.6434666666668,

86.40394533571002,

31.387758974358984,

19.499400899685114,

13.93330875790876,

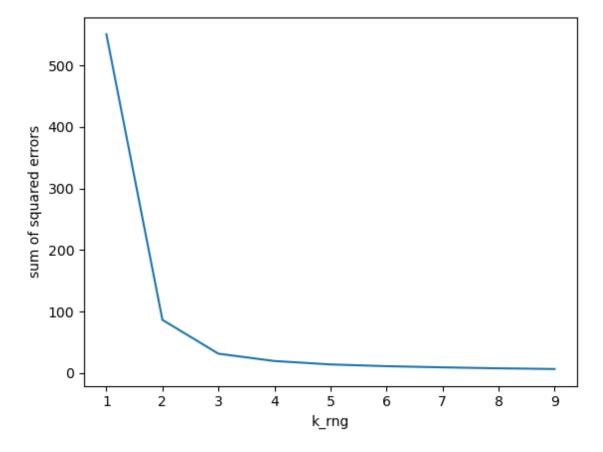
11.089449150654417,

9.228504827285606,

7.631802244955955,

6.496659206692712]

Out[48]: [<matplotlib.lines.Line2D at 0x255a88e8eb0>]



C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:87
0: FutureWarning: The default value of `n_init` will change from 10 to 'a uto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(

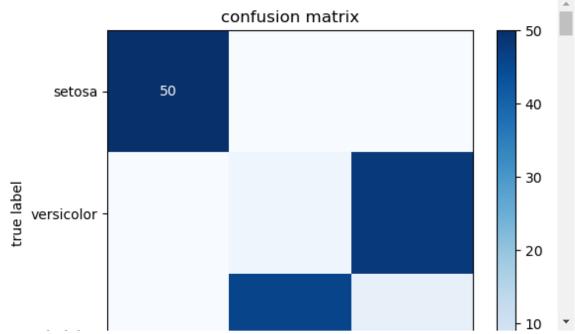
C:\Users\deepa\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:138
2: UserWarning: KMeans is known to have a memory leak on Windows with MK
L, when there are less chunks than available threads. You can avoid it by
setting the environment variable OMP_NUM_THREADS=1.
 warnings.warn(

Out[50]:		sepal_length	sepal_width	petal_length	petal_width	species	cluster
	0	5.1	3.5	1.4	0.2	0	0
	1	4.9	3.0	1.4	0.2	0	0
	2	4.7	3.2	1.3	0.2	0	0
	3	4.6	3.1	1.5	0.2	0	0
	4	5.0	3.6	1.4	0.2	0	0
	145	6.7	3.0	5.2	2.3	2	1
	146	6.3	2.5	5.0	1.9	2	1
	147	6.5	3.0	5.2	2.0	2	1
	148	6.2	3.4	5.4	2.3	2	1
	149	5.9	3.0	5.1	1.8	2	1

150 rows × 6 columns

[0, 46, 4]], dtype=int64)

```
In [53]:
          predicted_labels = df.cluster
            cm = confusion_matrix(true_labels , predicted_labels)
            class_labels = ['setosa','versicolor','virginica']
            plt.imshow(cm,interpolation = 'nearest' , cmap=plt.cm.Blues)
            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)
            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='white
                    plt.xlabel('predicted label')
                    plt.ylabel('true label')
                    plt.show()
```



In []: ▶