

iris-flower

July 20, 2024

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import os
import psutil
```

```
[2]: from sklearn.cluster import KMeans
```

```
[3]: df = pd.read_csv("IRIS.csv")
```

```
[4]: df.head()
```

```
[4]:   sepal_length  sepal_width  petal_length  petal_width  species
0           5.1           3.5           1.4           0.2  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           0.2  Iris-setosa
4           5.0           3.6           1.4           0.2  Iris-setosa
```

```
[5]: df['species'],categories= pd.factorize(df['species'])
```

```
[6]: df.head()
```

```
[6]:   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           4.6           3.1           1.5           0.2         0
4           5.0           3.6           1.4           0.2         0
```

```
[7]: df.describe
```

```
[7]: <bound method NDFrame.describe of
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 | 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 |
| 147 | 6.5 | 3.0 | 5.2 | 2.0 | 2 |
| 148 | 6.2 | 3.4 | 5.4 | 2.3 | 2 |
| 149 | 5.9 | 3.0 | 5.1 | 1.8 | 2 |

[150 rows x 5 columns]>

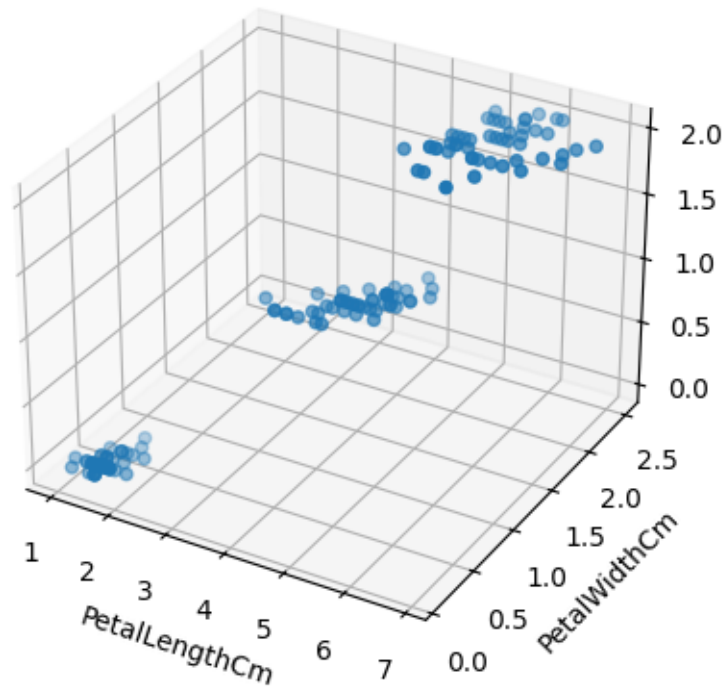
```
[8]: df.isnull().sum()
```

```
[8]: sepal_length    0
     sepal_width     0
     petal_length    0
     petal_width     0
     species         0
     dtype: int64
```

```
[9]: 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)

     ax.set_xlabel('PetalLengthCm')
     ax.set_ylabel('PetalWidthCm')
     ax.set_zlabel('Species')
     plt.title('3D Scatter plot Example')
     plt.show()
```

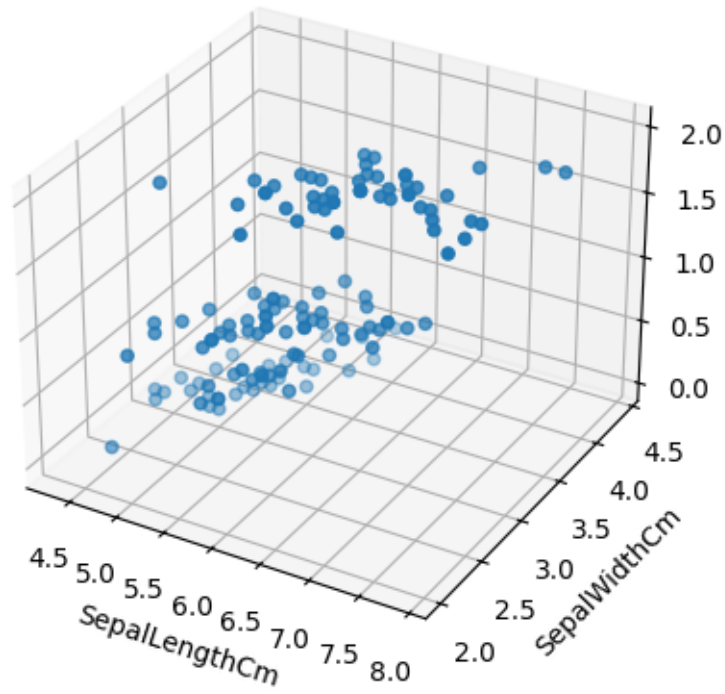
3D Scatter plot Example



```
[10]: 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)

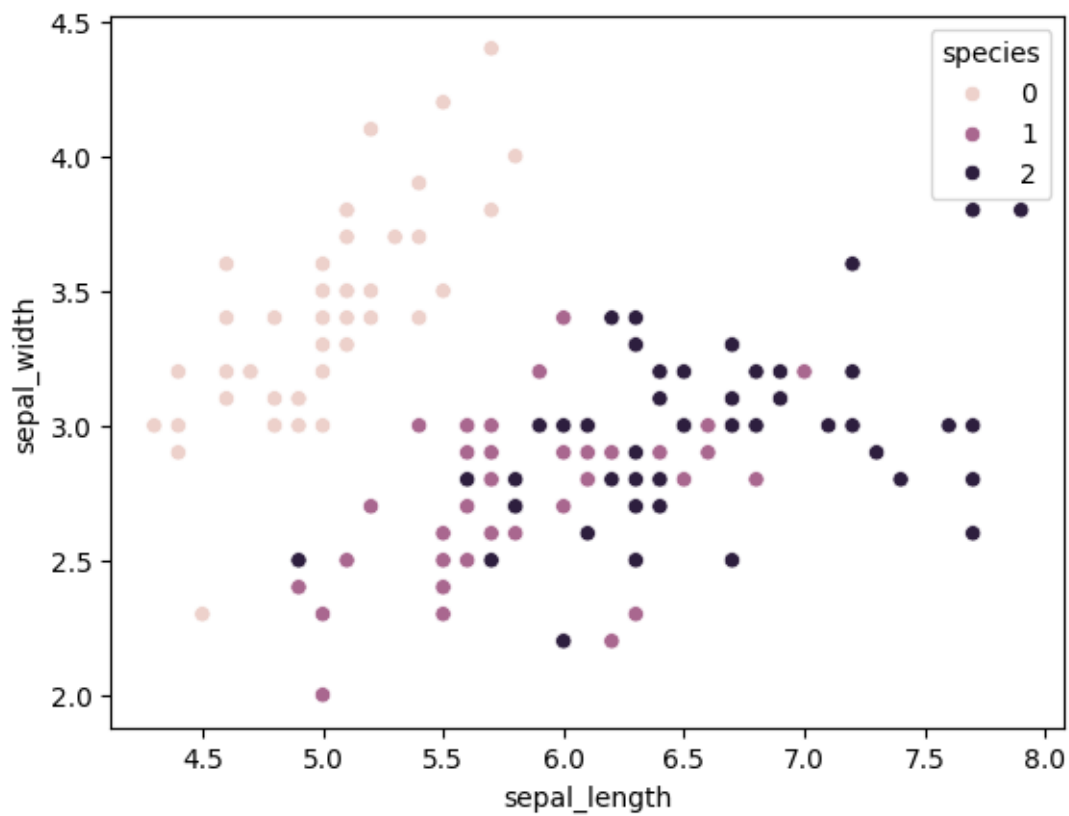
ax.set_xlabel('SepalLengthCm')
ax.set_ylabel('SepalWidthCm')
ax.set_zlabel('Species')
plt.title('3D Scatter plot Example')
plt.show()
```

3D Scatter plot Example



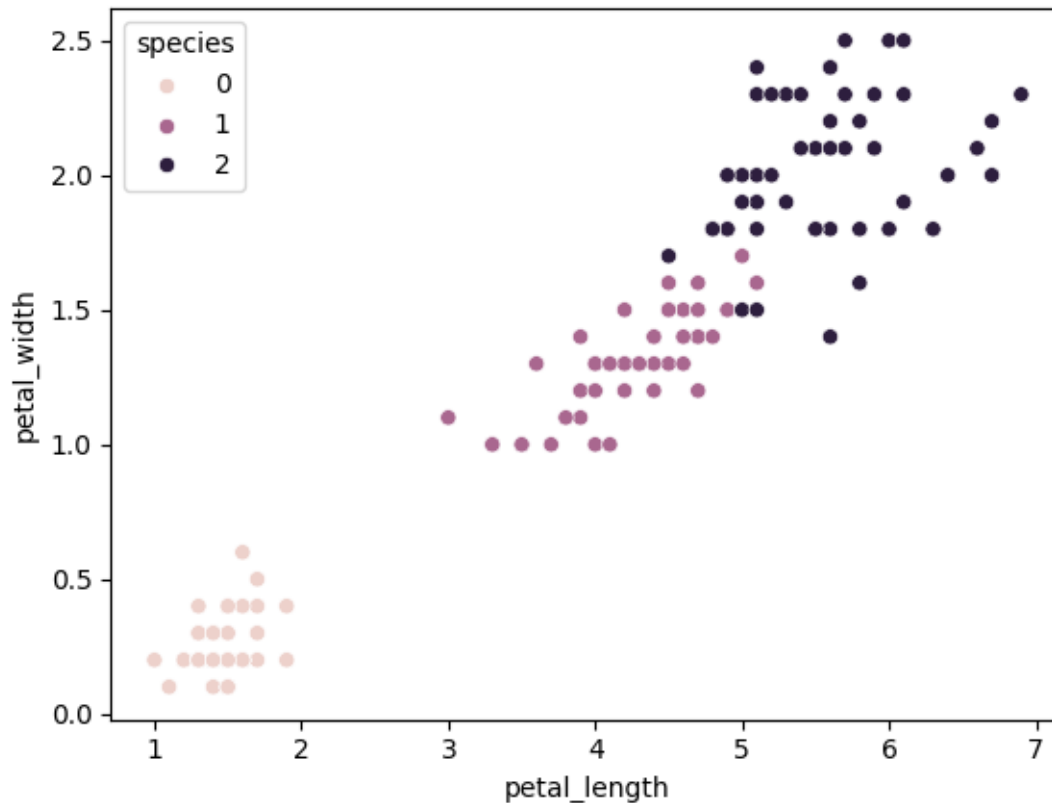
```
[11]: sns.scatterplot(data=df, x='sepal_length',y='sepal_width',hue='species')
```

```
[11]: <Axes: xlabel='sepal_length', ylabel='sepal_width'>
```



```
[12]: sns.scatterplot(data=df, x='petal_length',y='petal_width',hue='species')
```

```
[12]: <Axes: xlabel='petal_length', ylabel='petal_width'>
```



Applying Elbow Technique

```
[13]: 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_)
```

```
d:\Path\Lib\site-packages\joblib\externals\loky\backend\context.py:136:
UserWarning: Could not find the number of physical cores for the following
reason:
[WinError 2] The system cannot find the file specified
Returning the number of logical cores instead. You can silence this warning by
setting LOKY_MAX_CPU_COUNT to the number of cores you want to use.
  warnings.warn(
    File "d:\Path\Lib\site-packages\joblib\externals\loky\backend\context.py",
line 257, in _count_physical_cores
    cpu_info = subprocess.run(
    ~~~~~
    File "d:\Path\Lib\subprocess.py", line 548, in run
```

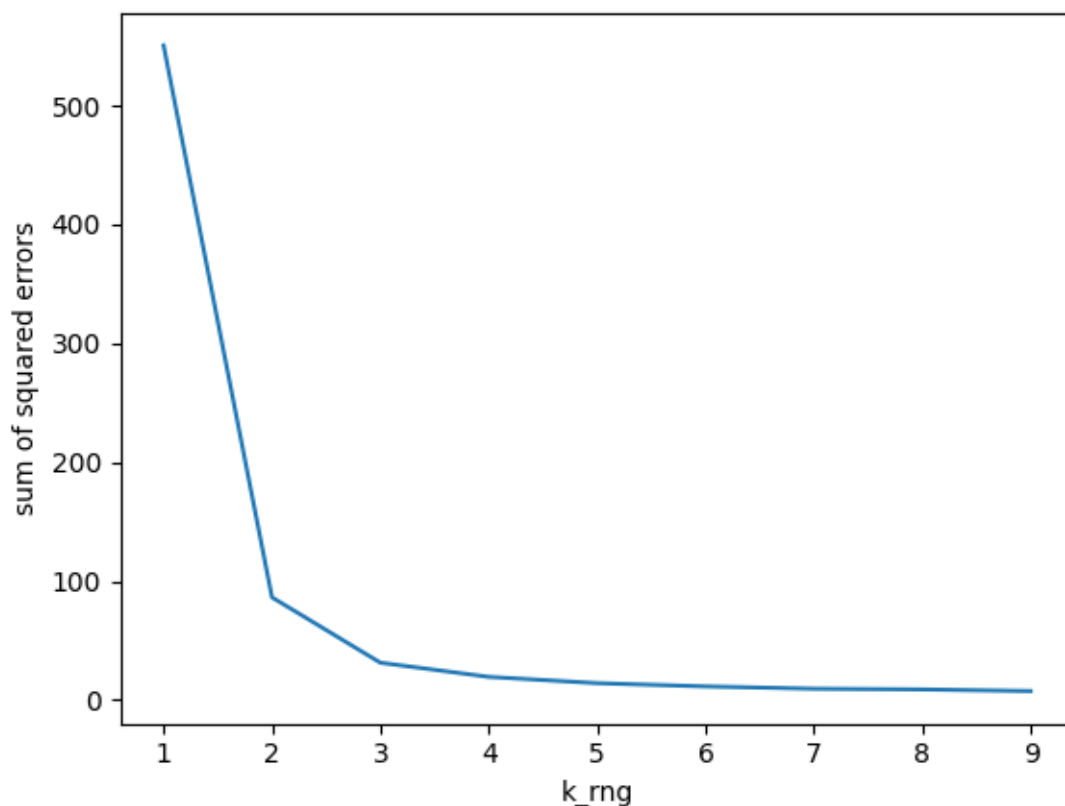
```

    with Popen(*popenargs, **kwargs) as process:
        ~~~~~
File "d:\Path\Lib\subprocess.py", line 1026, in __init__
    self._execute_child(args, executable, preexec_fn, close_fds,
File "d:\Path\Lib\subprocess.py", line 1538, in _execute_child
    hp, ht, pid, tid = _winapi.CreateProcess(executable, args,
    ~~~~~

```

```
[14]: plt.xlabel('k_rng')
plt.ylabel('sum of squared errors')
plt.plot(k_rng,sse)
```

```
[14]: []
```



```
[15]: km=KMeans(n_clusters=3,random_state=0,)
      y_predicted=km.fit_predict(df[['petal_length','petal_width']])
      y_predicted
```

```
[15]: array([1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,  
            1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,  
            1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
```

```
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 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, 0, 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, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2])
```

```
[16]: df['cluster']=y_predicted
df.head(150)
```

```
[16]:      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 x 6 columns]

Accuracy measure

```
[17]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(df.species,df.cluster)
cm
```

```
[17]: array([[ 0, 50,  0],
          [48,  0,  2],
          [ 4,  0, 46]], dtype=int64)
```

```
[18]: true_labels = df.species
predicted_labels=df.cluster

cm=confusion_matrix(true_labels,predicted_labels)
class_labels=['Setosa','versicolor','virginica']

# plt confusin matrix
plt.imshow(cm,interpolation='nearest',cmap=plt.cm.Blues)
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)):
```

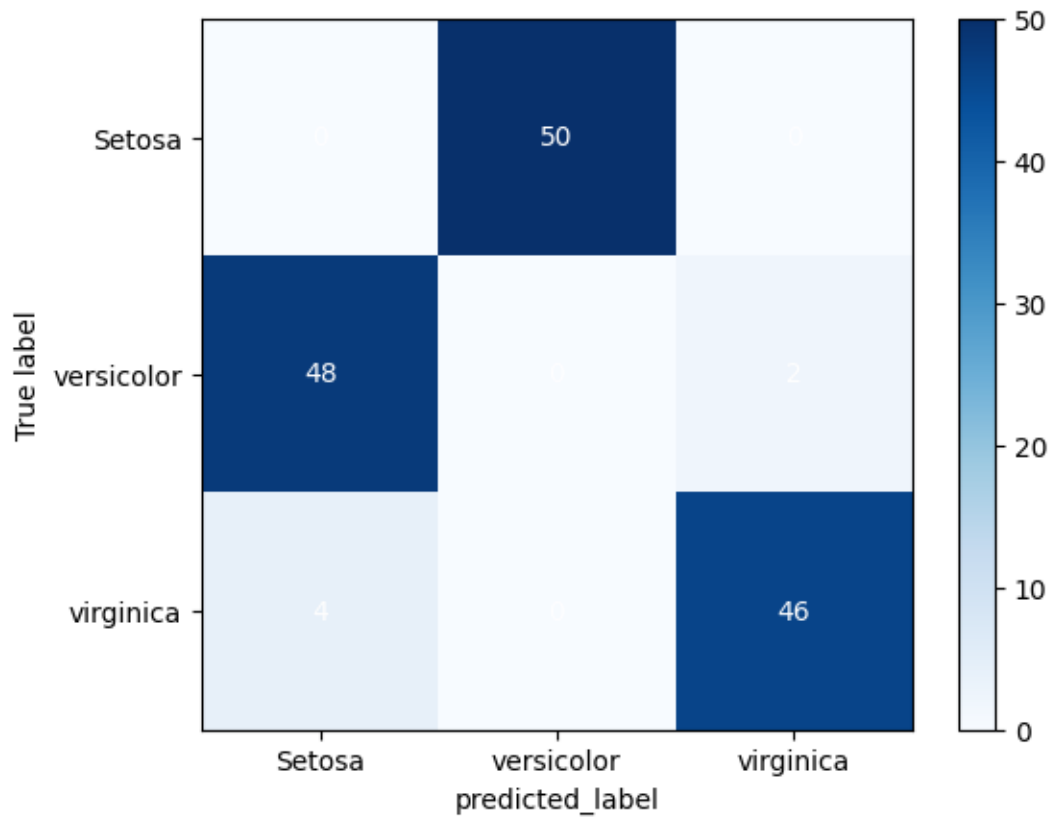


```

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()

```



[]: