```
In [37]:
```

```
from sklearn.cluster import KMeans
from sklearn.datasets import load_iris
import matplotlib.pyplot as plt
import pandas as pd
```

# In [4]:

```
iris = load_iris()
dir(iris)
```

# Out[4]:

['DESCR', 'data', 'feature names', 'filename', 'target', 'target names']

#### In [39]:

```
# we will only observe the pattern between petal length and width features for this clust
ering problem
data = pd.DataFrame(data = iris.data, columns = iris.feature_names)
data.head()
```

# Out[39]:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
0	5.1	3.5	1.4	0.2
1	4.9	3.0	1.4	0.2
2	4.7	3.2	1.3	0.2
3	4.6	3.1	1.5	0.2
4	5.0	3.6	1.4	0.2

# In [42]:

```
# removing sepal length and sepal width
data.drop(['sepal length (cm)', 'sepal width (cm)'], axis = 'columns', inplace = True)
data
```

#### Out[42]:

	petal length (cm)	petal width (cm)	
0	1.4	0.2	
1	1.4	0.2	
2	1.3	0.2	
3	1.5	0.2	
4	1.4	0.2	
145	5.2	2.3	
146	5.0	1.9	
147	5.2	2.0	
148	5.4	2.3	
149	5.1	1.8	

# 150 rows × 2 columns

# In [43]:

# plotting the data

```
plt.scatter(data['petal width (cm)'], data['petal length (cm)'])
Out[43]:
<matplotlib.collections.PathCollection at 0xd789d3e208>
 7
 6
 5
 4
 3
                        1.5
  0.0
         0.5
                 1.0
                                2.0
                                        2.5
In [44]:
# using elbow methods to find best number of clusters
k list = range(1,11)
sse = []
for k in k list:
    km = KMeans(n clusters = k)
    km.fit(data)
    sse.append(km.inertia )
sse
Out[44]:
[550.8953333333334,
 86.39021984551397,
 31.371358974358973,
 19.48300089968511,
 13.916908757908757,
 11.07065234189628,
 9.244938551786376,
 7.745948930296759,
 6.545263125763126,
 5.704365024006968]
In [45]:
# plotting the elbow
plt.plot(k list, sse)
plt.xlabel("k")
plt.ylabel("Sum of Squared Error")
plt.xticks(k_list)
# we can see the best number of clusters are 3
Out[45]:
([<matplotlib.axis.XTick at 0xd789da3668>,
  <matplotlib.axis.XTick at 0xd789d919b0>,
  <matplotlib.axis.XTick at 0xd789d91e48>,
  <matplotlib.axis.XTick at 0xd789dcb320>,
  <matplotlib.axis.XTick at 0xd789dcb7f0>,
  <matplotlib.axis.XTick at 0xd789dcbcc0>,
  <matplotlib.axis.XTick at 0xd789dd2240>,
```

<matplotlib.axis.XTick at 0xd789dd27b8>,
<matplotlib.axis.XTick at 0xd789dd2d30>,
<matplotlib.axis.XTick at 0xd789dd92e8>],
<a list of 10 Text xticklabel objects>)

```
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```

#### In [46]:

```
# clustering with 3 clusters
km = KMeans(n_clusters = 3)
y_pred = km.fit_predict(data)
y_pred
```

#### Out[46]:

# In [48]:

```
# adding cluster predictions to our data
data['cluster'] = y_pred
data.head()
```

# Out[48]:

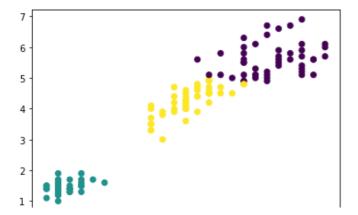
	petal length (cm)	petal width (cm)	cluster
0	1.4	0.2	1
1	1.4	0.2	1
2	1.3	0.2	1
3	1.5	0.2	1
4	1.4	0.2	1

#### In [55]:

```
# plotting the data according to clusters
plt.scatter(data['petal width (cm)'], data['petal length (cm)'], c = data.cluster)
# we can see that the data has been clustered well
```

# Out[55]:

<matplotlib.collections.PathCollection at 0xd78a0da048>



```
0.0 0.5 1.0 1.5 2.0 2.5
```

# In [59]:

```
# cross checking clustering using available target in the iris data
data['target'] = iris.target
plt.scatter(data['petal width (cm)'], data['petal length (cm)'], c = data.target)

# both the plots show similar grouping thus we can conclude clustering is good
# we can say that the three classes can be correctly differentiated using clustering
```

# Out[59]:

<matplotlib.collections.PathCollection at 0xd78a137ac8>

