

```
In [2]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [3]: import pandas as pd

url = "https://raw.githubusercontent.com/mwaskom/seaborn-data/master/iris.csv"

df = pd.read_csv(url)
df.head()
```

```
Out[3]:
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa

```
In [4]: df.columns
```

```
Out[4]: Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',
              'species'],
              dtype='object')
```

```
In [5]: df.sum().isnull()
```

```
Out[5]: sepal_length    False
sepal_width          False
petal_length         False
petal_width          False
species              False
dtype: bool
```

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

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa

```
In [7]: print(df.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
 #   Column          Non-Null Count  Dtype
---  -
 0   sepal_length    150 non-null    float64
 1   sepal_width     150 non-null    float64
 2   petal_length    150 non-null    float64
 3   petal_width     150 non-null    float64
 4   species         150 non-null    object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
None
```

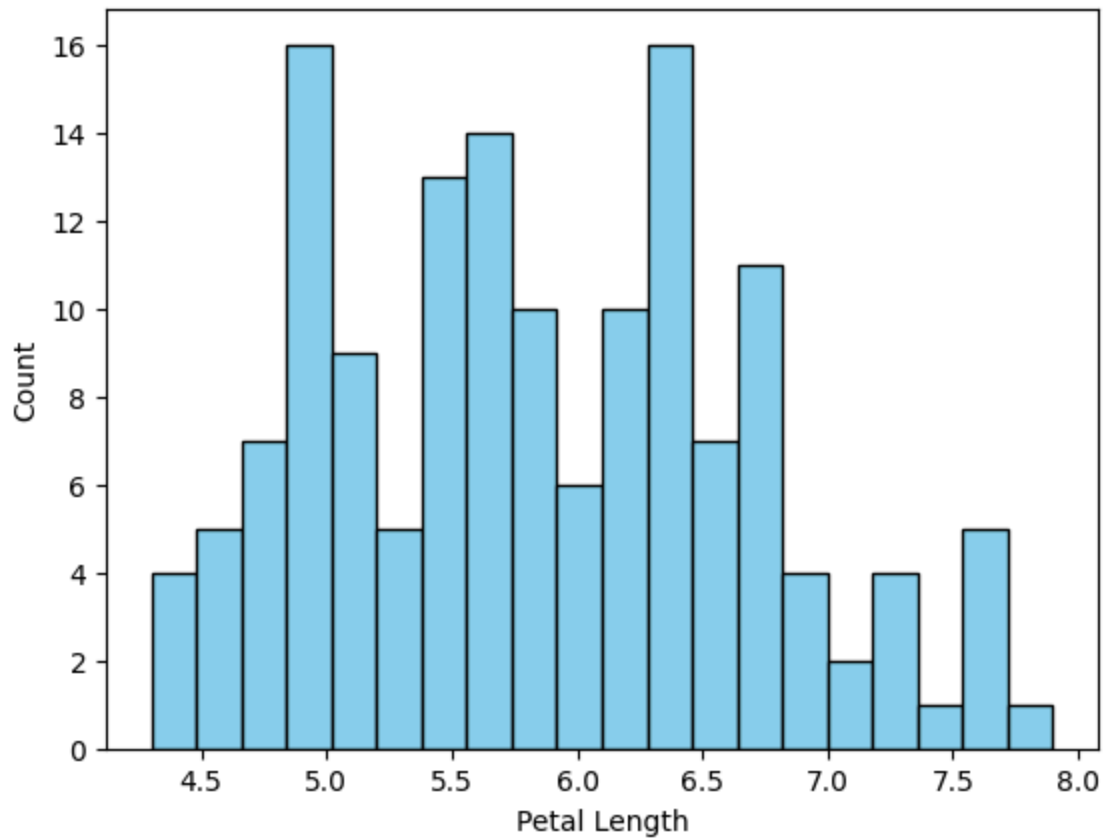
```
In [8]: print(df.describe())
```

```
      sepal_length  sepal_width  petal_length  petal_width
count    150.000000    150.000000    150.000000    150.000000
mean         5.843333         3.057333         3.758000         1.199333
std          0.828066         0.435866         1.765298         0.762238
min          4.300000         2.000000         1.000000         0.100000
25%          5.100000         2.800000         1.600000         0.300000
50%          5.800000         3.000000         4.350000         1.300000
75%          6.400000         3.300000         5.100000         1.800000
max          7.900000         4.400000         6.900000         2.500000
```

```
In [9]: # count species counts
print(df['species'].value_counts())
```

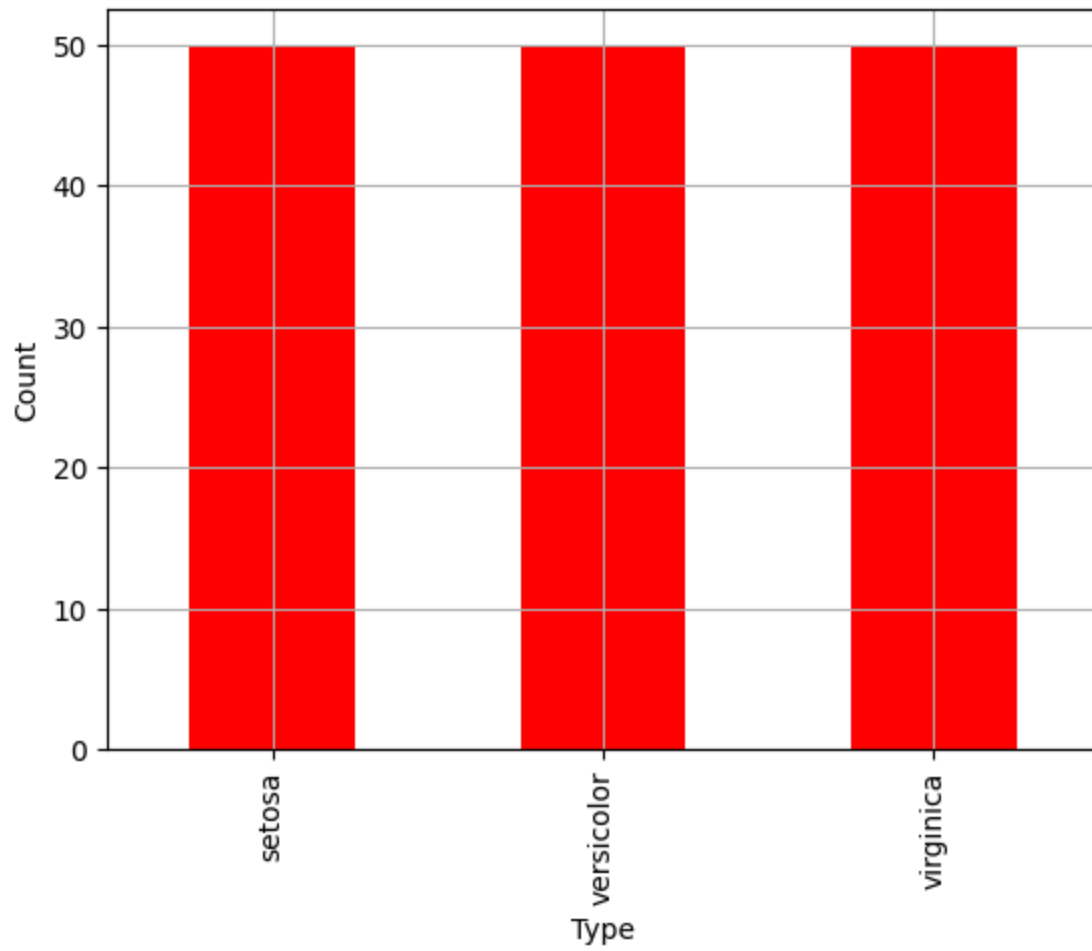
```
species
setosa      50
versicolor  50
virginica   50
Name: count, dtype: int64
```

```
In [10]: plt.hist(df['sepal_length'],bins=20,color='skyblue',edgecolor='black')
plt.xlabel('Petal Length')
plt.ylabel('Count')
plt.show()
```

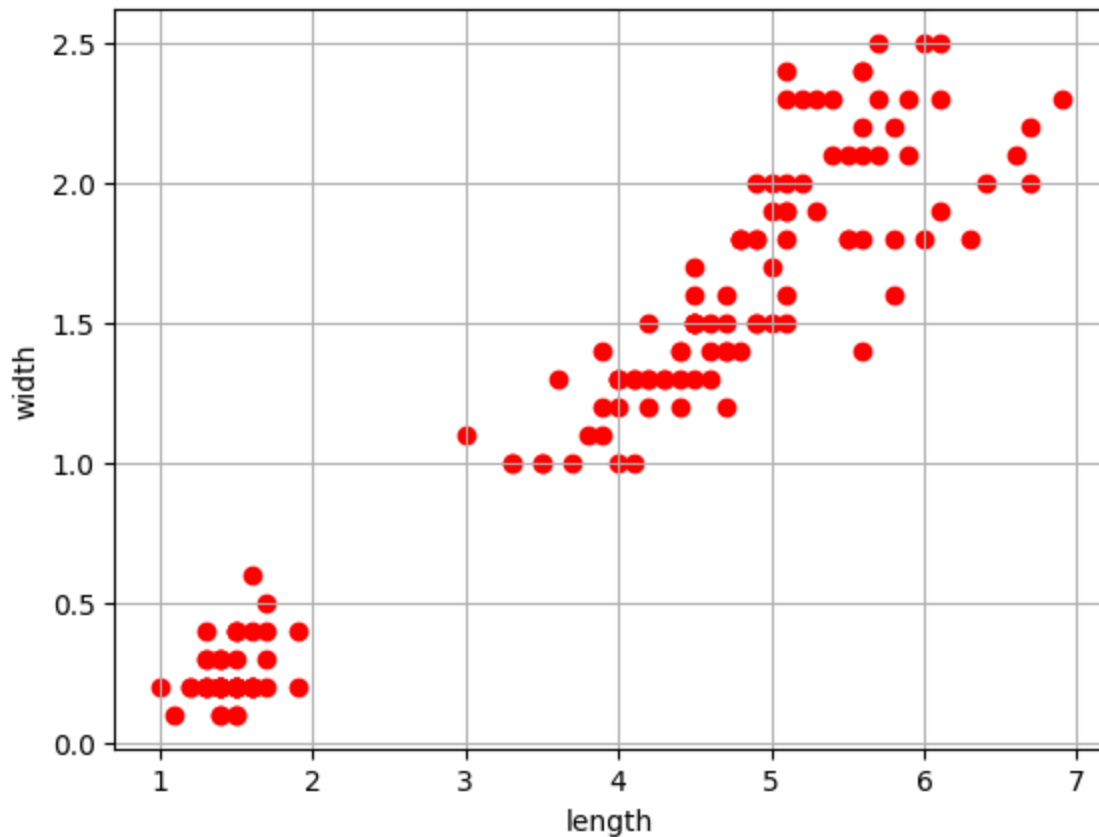


```
In [11]: df['species'].value_counts().plot(kind='bar',color='red')

plt.xlabel('Type')
plt.ylabel('Count')
plt.grid(True)
plt.show()
```



```
In [12]: plt.scatter(df['petal_length'], df['petal_width'], c='red')
plt.xlabel('length')
plt.ylabel('width')
plt.grid(True)
plt.show()
```



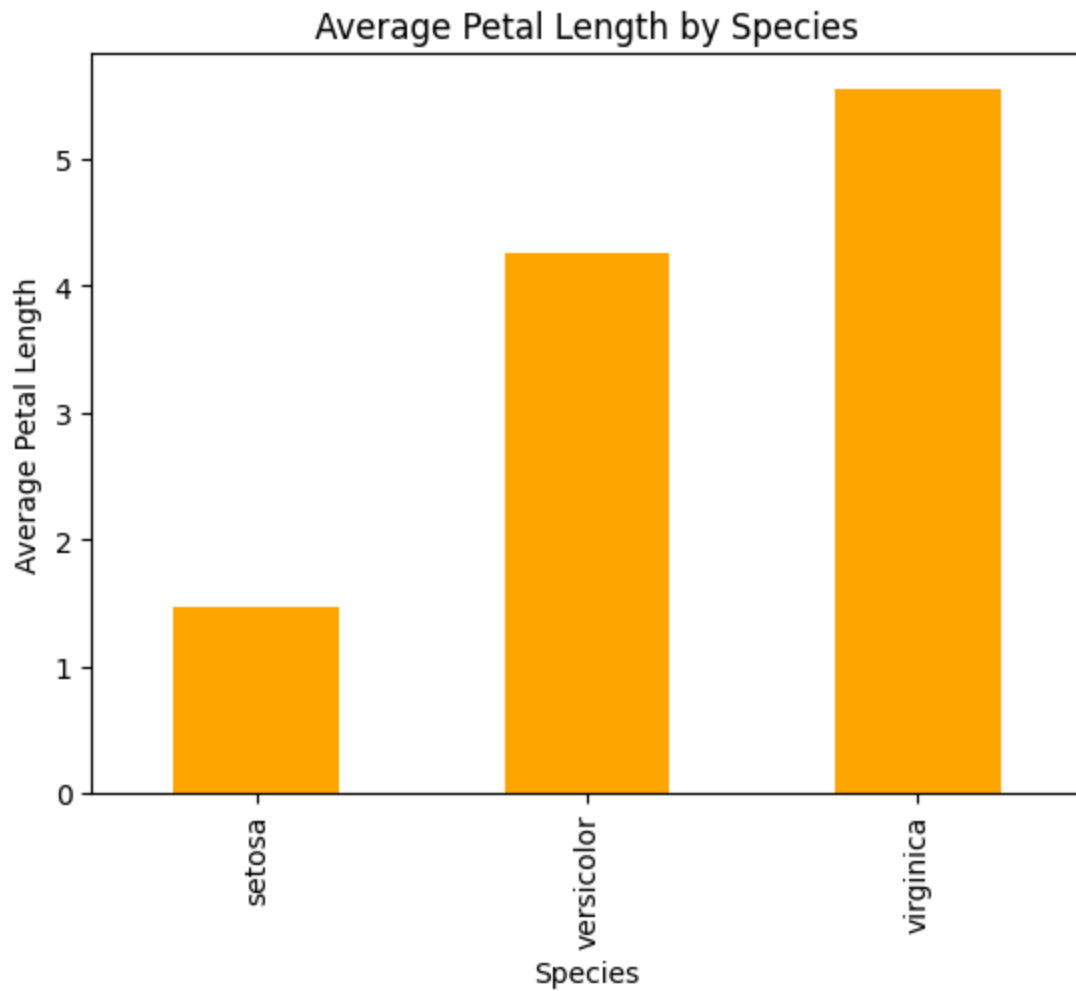
```
In [13]: # Group by species and calculate mean of numeric columns
species_group=df.groupby('species').mean()
print(species_group)
```

	sepal_length	sepal_width	petal_length	petal_width
species				
setosa	5.006	3.428	1.462	0.246
versicolor	5.936	2.770	4.260	1.326
virginica	6.588	2.974	5.552	2.026

```
In [14]: species_petal=df.groupby('species')['petal_length'].agg(['min','max'])
print(species_petal)
```

	min	max
species		
setosa	1.0	1.9
versicolor	3.0	5.1
virginica	4.5	6.9

```
In [15]: species_group['petal_length'].plot(kind='bar', color='orange')
plt.title('Average Petal Length by Species')
plt.xlabel('Species')
plt.ylabel('Average Petal Length')
plt.show()
```



```
In [16]: # Filter flowers with petal length greater than 4.5
long_petals=df[df['petal_length']>4.5]
print("Flowers with petal length > 4.5:")
print(long_petals)

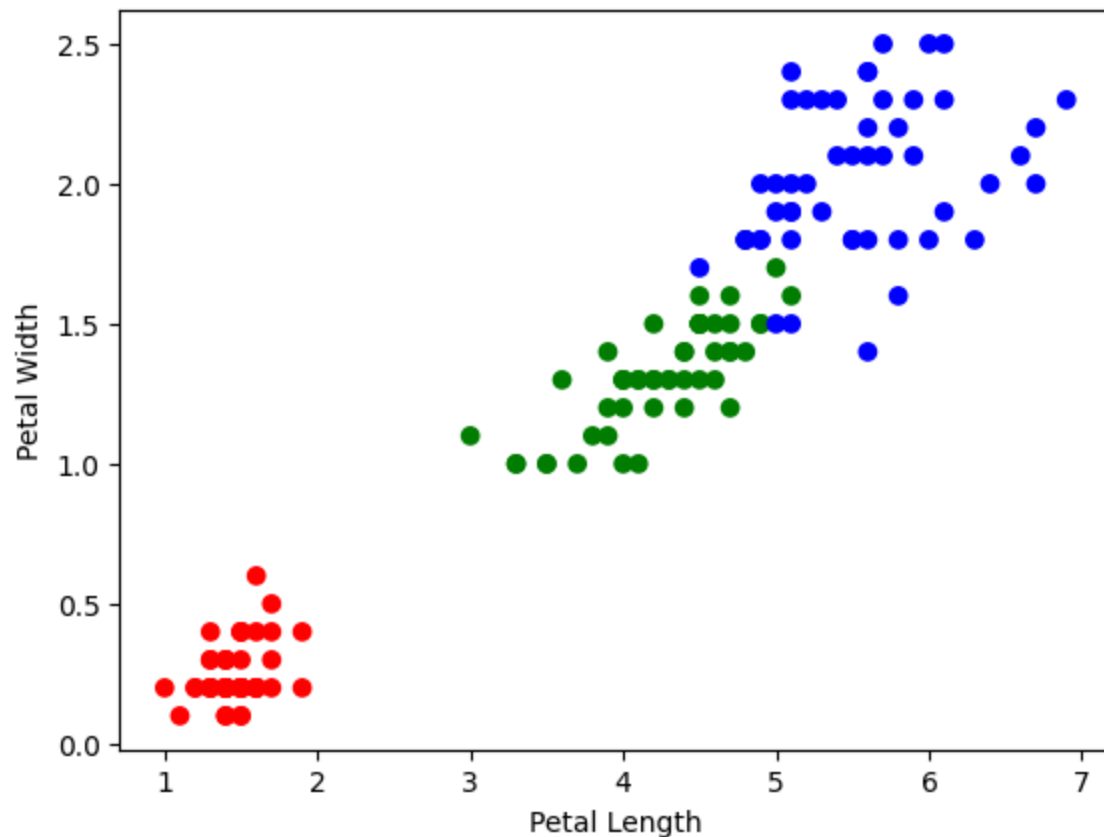
colors = {'setosa':'red', 'versicolor':'green', 'virginica':'blue'}
plt.scatter(df['petal_length'], df['petal_width'],
            c=df['species'].map(colors))

plt.xlabel('Petal Length')
plt.ylabel('Petal Width')
plt.show()
```

Flowers with petal length > 4.5:

	sepal_length	sepal_width	petal_length	petal_width	species
50	7.0	3.2	4.7	1.4	versicolor
52	6.9	3.1	4.9	1.5	versicolor
54	6.5	2.8	4.6	1.5	versicolor
56	6.3	3.3	4.7	1.6	versicolor
58	6.6	2.9	4.6	1.3	versicolor
..
145	6.7	3.0	5.2	2.3	virginica
146	6.3	2.5	5.0	1.9	virginica
147	6.5	3.0	5.2	2.0	virginica
148	6.2	3.4	5.4	2.3	virginica
149	5.9	3.0	5.1	1.8	virginica

[63 rows x 5 columns]



```
In [17]: # Filter flowers above species mean petal length
species_mean_petals = df.groupby('species')['petal_length'].mean()
above_avg = df[df.apply(lambda row: row['petal_length'] > species_mean_petals[row['species']], axis=1)]
print(above_avg.head())

# Count by species
print(above_avg['species'].value_counts())

# Scatter plot with color & size by species
colors = {'setosa':'red', 'versicolor':'green', 'virginica':'blue'}
sizes = df['petal_length']*20
plt.scatter(df['petal_length'], df['petal_width'], c=df['species'].map(colors), s=sizes)
plt.title('Petal Length vs Width (Advanced)')
plt.xlabel('Petal Length')
```

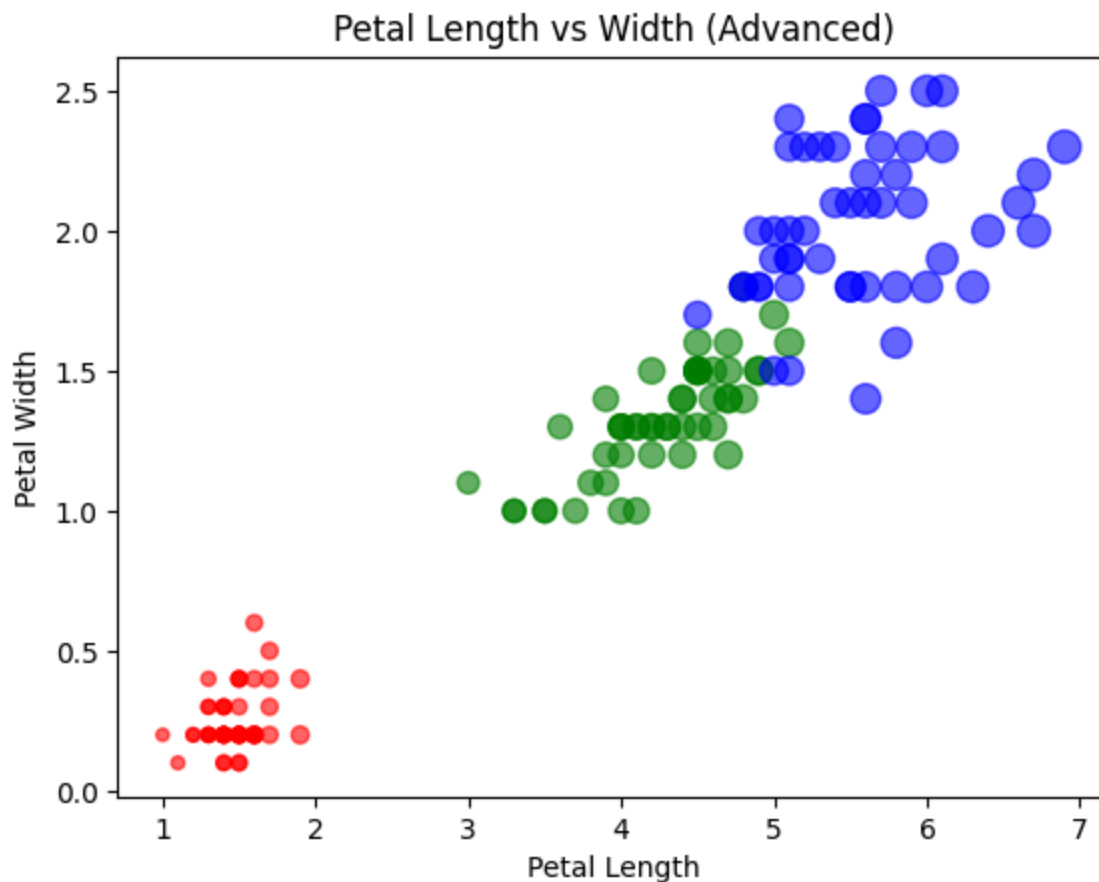
```
plt.ylabel('Petal Width')
plt.show()
```

	sepal_length	sepal_width	petal_length	petal_width	species
3	4.6	3.1	1.5	0.2	setosa
5	5.4	3.9	1.7	0.4	setosa
7	5.0	3.4	1.5	0.2	setosa
9	4.9	3.1	1.5	0.1	setosa
10	5.4	3.7	1.5	0.2	setosa

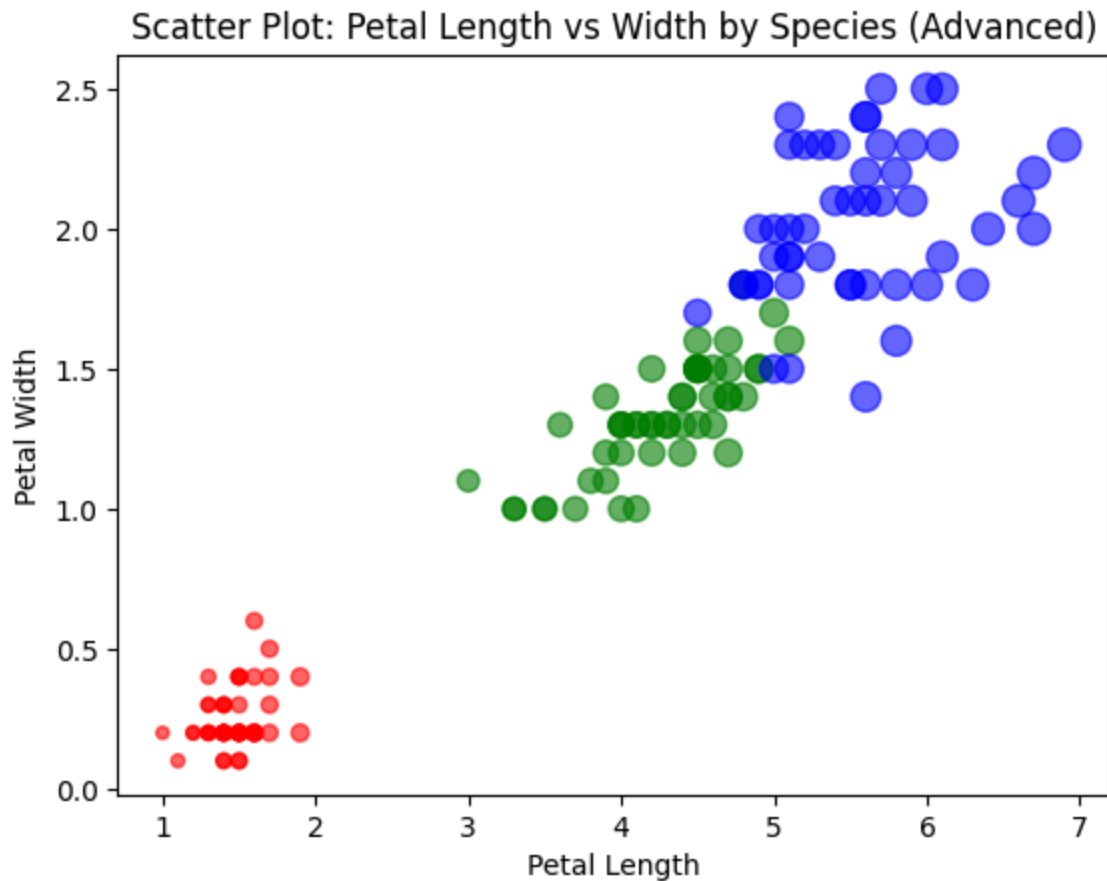
species

versicolor	27
setosa	26
virginica	25

Name: count, dtype: int64



```
In [18]: colors = {'setosa':'red', 'versicolor':'green', 'virginica':'blue'}
          sizes = df['petal_length']*20 # size proportional to petal length
          plt.scatter(df['petal_length'], df['petal_width'],
                      c=df['species'].map(colors), s=sizes, alpha=0.6)
          plt.title('Scatter Plot: Petal Length vs Width by Species (Advanced)')
          plt.xlabel('Petal Length')
          plt.ylabel('Petal Width')
          plt.show()
```

This report analyzes the Iris dataset. The goal is to explore flower features, compare species, and extract useful patterns to better understand the data.

Conclusion

Species differ clearly in petal length and width

Most flowers with longer petals are Virginica

Scatter plots help visualize patterns in the data