

Data Set-up/Exploration

```
# import necessary packages
import warnings
warnings.filterwarnings('ignore')

from plotnine import *

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

from sklearn.model_selection import train_test_split #TTS function
from sklearn import svm #svm model
from sklearn import metrics #scikit-learn metrics module

# load iris data set
iris =
pd.read_csv("https://raw.githubusercontent.com/lannen/iris-1/main/iris-1.csv")

# look at features
iris.head()
```

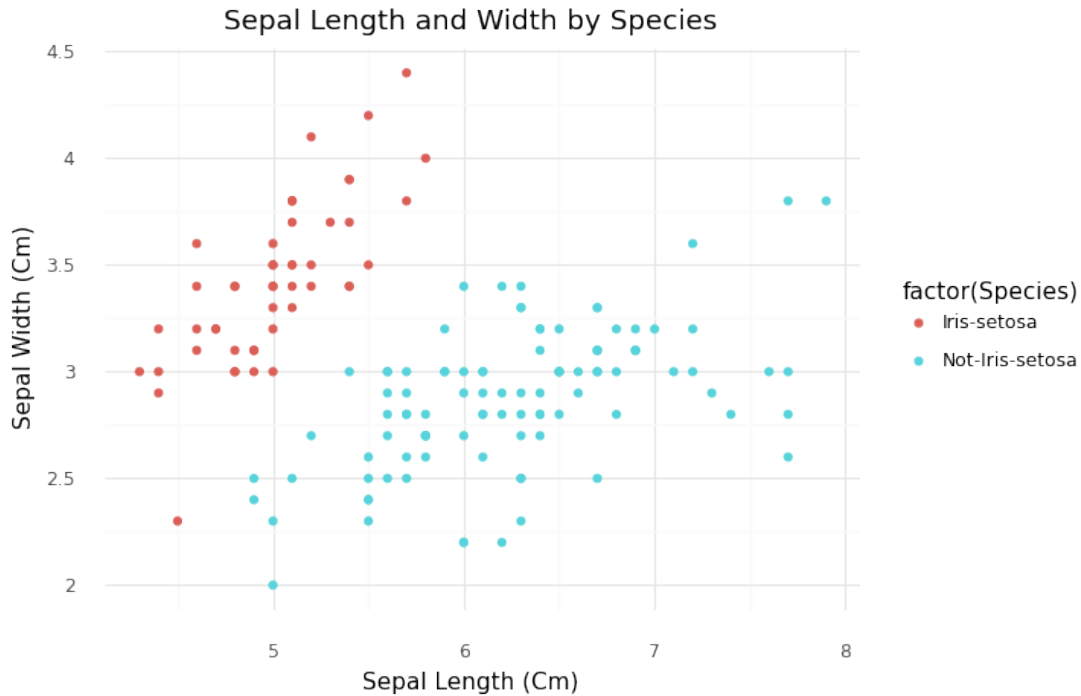
	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	
Species						
0	1	5.1	3.5	1.4	0.2	Iris-
						setosa
1	2	4.9	3.0	1.4	0.2	Iris-
						setosa
2	3	4.7	3.2	1.3	0.2	Iris-
						setosa
3	4	4.6	3.1	1.5	0.2	Iris-
						setosa
4	5	5.0	3.6	1.4	0.2	Iris-
						setosa

```
# look at size
iris.shape
```

```
(150, 6)
```

Sepal & Petal (length/width) Visualizations

```
# comparing sepal length and width by species
(ggplot(iris, aes(x = "SepalLengthCm", y = "SepalWidthCm", color =
"factor(Species)")) +
  geom_point() + theme_minimal() + xlab("Sepal Length (Cm)") +
  ylab("Sepal Width (Cm)") +
  ggtitle("Sepal Length and Width by Species"))
```



```
<ggplot: (8741076909357)>
```

Graph description

- Iris-Setosa: larger Sepal width (~ 3-4.5cm), shorter Sepal length (~ 4-6cm)
- Not-Iris-Setosa: smaller Sepal width (~ 2-3.5cm), longer Sepal length (~ 5-8cm)

comparing petal length and width by species

```
(ggplot(iris, aes(x = "PetalLengthCm", y = "PetalWidthCm", color =  
"factor(Species)")) +  
  geom_point() + theme_minimal() + xlab("Petal Length (Cm)") +  
  ylab("Petal Width (Cm)") +  
  ggtitle("Petal Length and Width by Species"))
```



```
<ggplot: (8741076811881)>
```

Graph description

- Iris-Setosa: smaller Petal width (~ 0-0.5cm), shorter Petal length (~ 0.5-2cm)
- Not-Iris-Setosa: larger Petal width (~ 1-2.5cm), longer Petal length (~ 3-7cm)

Data Transformation

```
# make into a dataframe
```

```
iris_df = pd.DataFrame(iris)
```

```
# change species feature to a dummy variable for analysis
```

```
# target - Iris-setosa OR Not-Iris-setosa
```

```
iris_df.Species = iris_df.Species.eq('Iris-setosa').mul(1)
```

```
# check if species is a dummy variable
```

```
iris_df.head()
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
Species					
0	1	5.1	3.5	1.4	0.2
1					
1	2	4.9	3.0	1.4	0.2
1					
2	3	4.7	3.2	1.3	0.2
1					
3	4	4.6	3.1	1.5	0.2
1					

4	5	5.0	3.6	1.4	0.2
1					

Train, Test, Split & Support Vector Machines

split dataset into 70% training and 30% test

```
X_train, X_test, y_train, y_test = train_test_split(iris_df,  
iris.Species, test_size=0.3, random_state=109)
```

create a svm Classifier

```
clf = svm.SVC(kernel='linear') # Linear Kernel
```

train/fit the model using the training set

```
clf.fit(X_train, y_train)
```

predict the response for test set

```
y_pred = clf.predict(X_test)
```

Accuracy, Precision, Recall

model accuracy: how often is the classifier correct?

```
print("Accuracy:", metrics.accuracy_score(y_test, y_pred))
```

model precision: what percentage of positive tuples are labeled as such?

```
print("Precision:", metrics.precision_score(y_test, y_pred))
```

model recall: what percentage of positive tuples are labeled as such?

```
print("Recall:", metrics.recall_score(y_test, y_pred))
```

Accuracy: 1.0

Precision: 1.0

Recall: 1.0