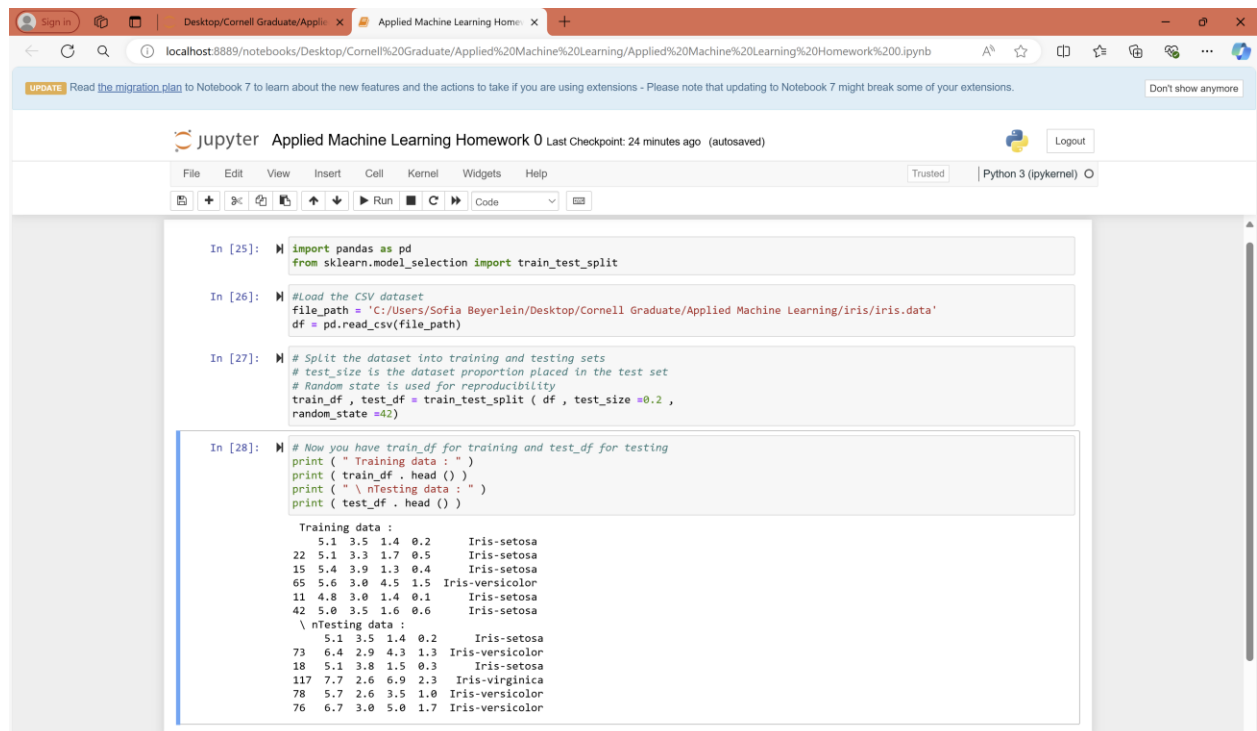


Applied Machine Learning Homework 0

1. There are 4 attributes/features/attributes per sample. There are 3 different species (Iris Setosa, Iris Versicolour, and Iris Virginica) and there's 50 samples per each species so a total of 150 instances.
2. Figure out how to parse the dataset you downloaded. Load the samples into an $N \times p$ array, where N is the number of samples and p is the number of attributes per sample. Additionally, create a N -dimensional vector containing each sample's label (species).



The screenshot shows a Jupyter Notebook titled "Applied Machine Learning Homework 0" with a last checkpoint 24 minutes ago. The notebook contains four code cells. The first cell imports pandas and sklearn.model_selection.train_test_split. The second cell loads the Iris dataset from a local file path. The third cell splits the dataset into training and testing sets with a test size of 0.2 and a random state of 42. The fourth cell prints the training and testing data, showing the first few rows of each.

```
In [25]: import pandas as pd
         from sklearn.model_selection import train_test_split

In [26]: #Load the CSV dataset
         file_path = 'C:/Users/Sofia Beyerlein/Desktop/Cornell Graduate/Applied Machine Learning/iris/iris.data'
         df = pd.read_csv(file_path)

In [27]: # Split the dataset into training and testing sets
         # test_size is the dataset proportion placed in the test set
         # Random state is used for reproducibility
         train_df, test_df = train_test_split(df, test_size=0.2, random_state=42)

In [28]: # Now you have train_df for training and test_df for testing
         print(" Training data : ")
         print(train_df.head())
         print("\nTesting data : ")
         print(test_df.head())

Training data :
   5.1  3.5  1.4  0.2  Iris-setosa
  22  5.1  3.3  1.7  0.5  Iris-setosa
   15  5.4  3.9  1.3  0.4  Iris-setosa
   65  5.6  3.0  4.5  1.5  Iris-versicolor
   11  4.8  3.0  1.4  0.1  Iris-setosa
   42  5.0  3.5  1.6  0.6  Iris-setosa
\nTesting data :
   73  6.4  2.9  4.3  1.3  Iris-versicolor
   18  5.1  3.8  1.5  0.3  Iris-setosa
  117  7.7  2.6  6.9  2.3  Iris-virginica
   78  5.7  2.6  3.5  1.0  Iris-versicolor
   76  6.7  3.0  5.0  1.7  Iris-versicolor
```

3. I plotted all the graphs using this snippet of code but replacing `plt.scatter(sepal_length, sepal_width, color)` with the other attribute arrays (e.g. `plt.scatter(sepal_length, petal_length, color)`). I also want to clarify that I am

plotting the training data.

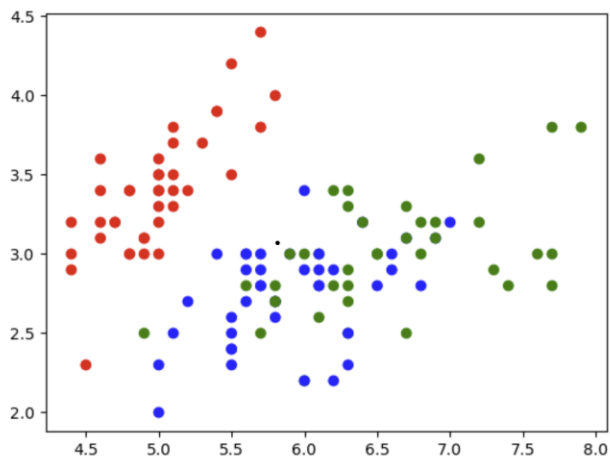
```
In [57]: ▶ #saving the attributes of training data
sepal_length = train_df.iloc[:, 0]
sepal_width = train_df.iloc[:, 1]
petal_length = train_df.iloc[:, 2]
petal_width = train_df.iloc[:, 3]

species = train_df.iloc[:, 4]
color = []

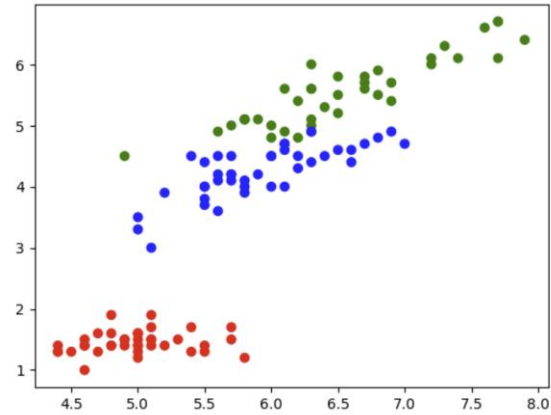
#assigning an rgb value to the species in the array
for specie in species:
    if specie == "Iris-setosa":
        color.append("r")
    if specie == "Iris-versicolor":
        color.append("b")
    if specie == "Iris-virginica":
        color.append("g")

plt.scatter(sepal_length, sepal_width, c=color)
plt.savefig("plot.png")
```

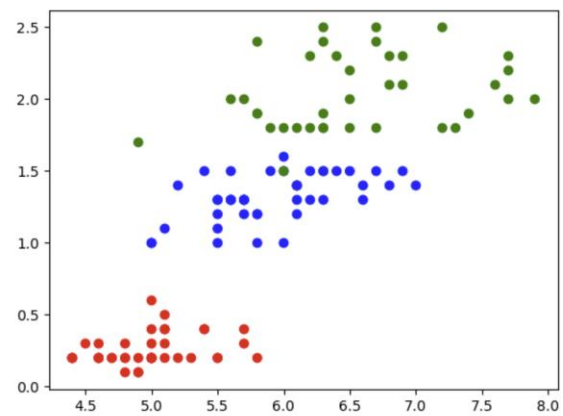
Sepal Length vs. Sepal Width



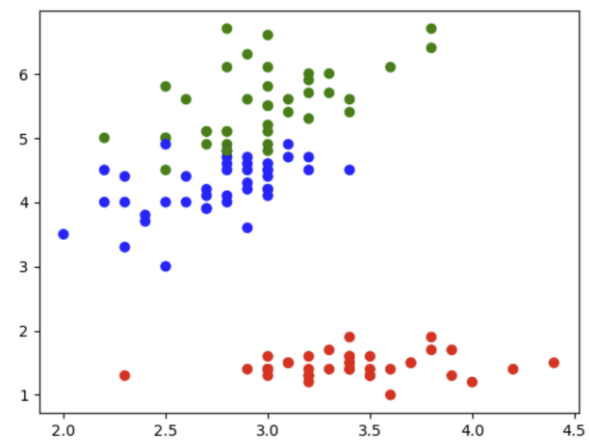
Sepal Length vs. Petal Length



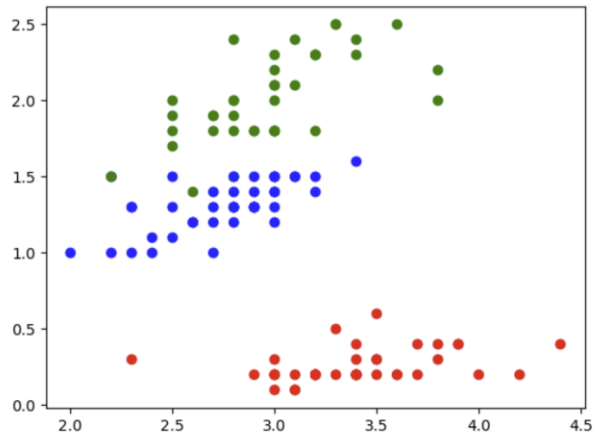
Sepal Length vs. Petal Width



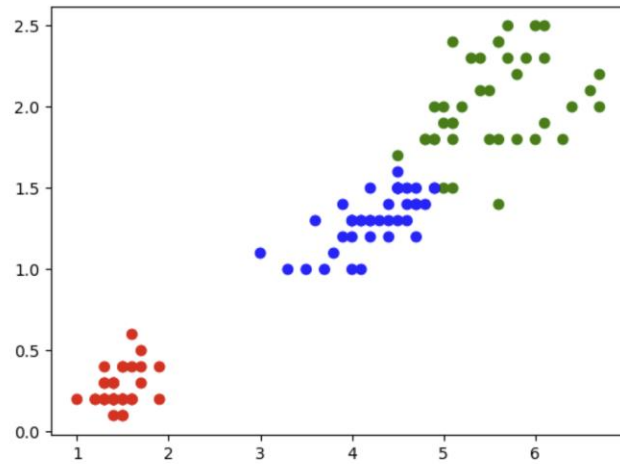
Sepal Width vs. Petal Length



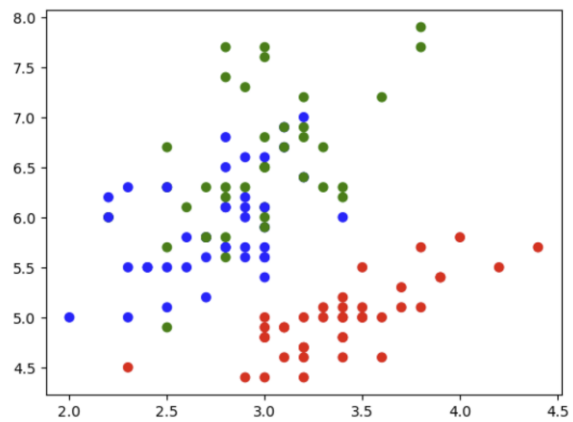
Sepal Width vs. Petal Width



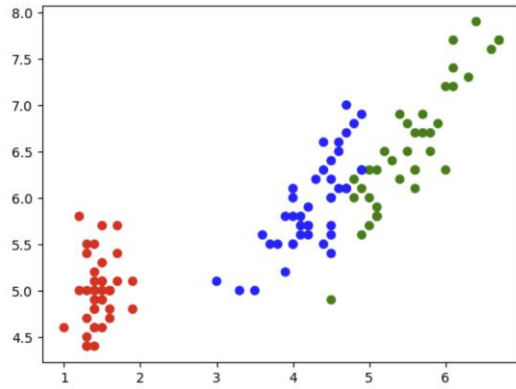
Petal Length vs. Petal Width



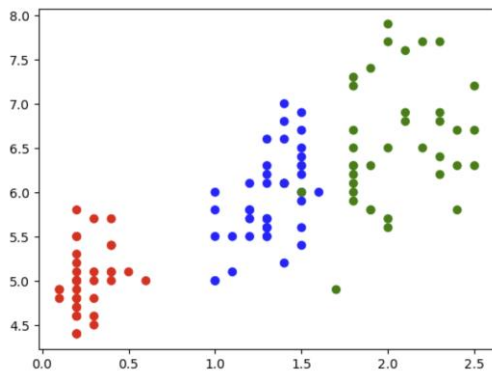
Sepal Width vs. Sepal Length



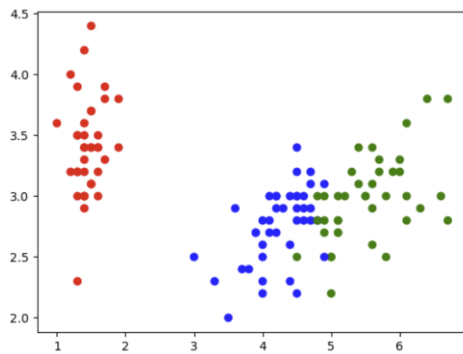
Petal Length vs. Sepal Length



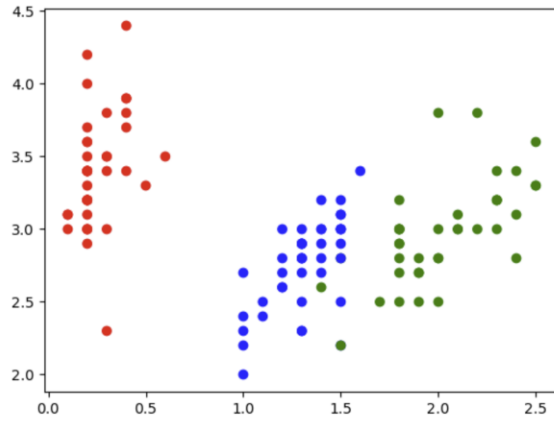
Petal Width vs. Sepal Length



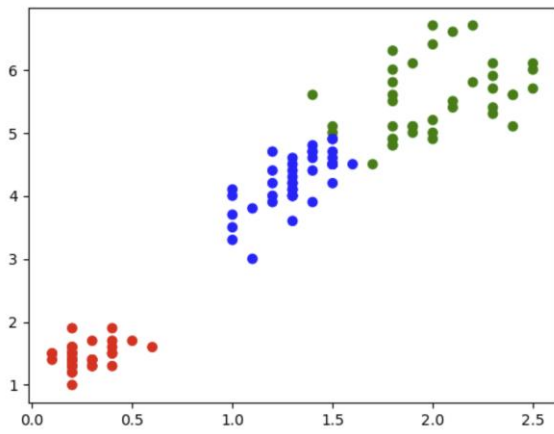
Petal Length vs. Sepal Width



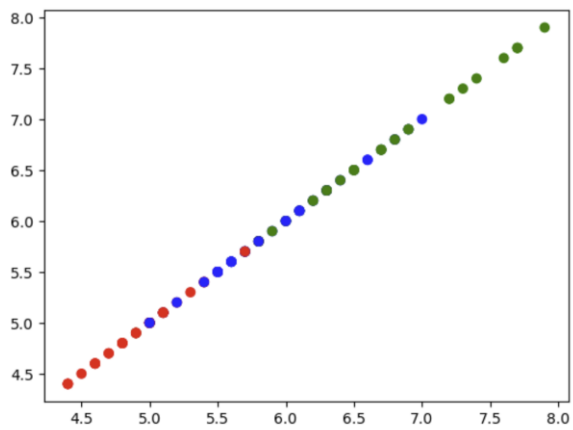
Petal Width vs. Sepal Width



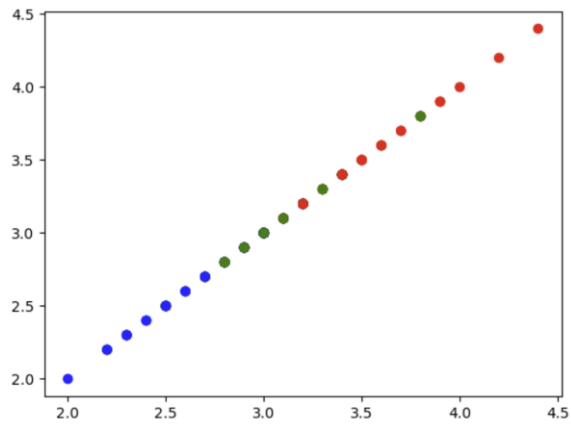
Petal Width vs. Petal Length



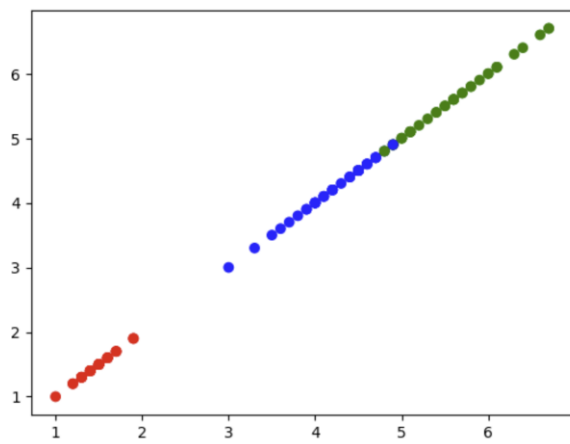
Sepal Length vs. Sepal Length



Sepal Width vs. Sepal Width



Petal Length vs. Petal Length



Petal Width vs. Petal Width

