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**CS 5875 - Applied Machine Learning**  
**Homework 0**  
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## Iris Flowers

### 1.1 Downloading the dataset

In order to download the dataset, we called the `pd.read_csv`<sup>1</sup> function on the iris dataset URL, which contains CSV data on the iris plants database.

```
1 url = 'http://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data'
2 df = pd.read_csv(url, header=None)
3 df.columns = ['sepal length (cm)', 'sepal width (cm)', 'petal length (cm)', 'petal width (cm)',
               'class']
```

Taking a look at the `iris.names` file, we can see that there are 150 total samples/instances, with 5 features/attributes for each sample—sepal length (cm), sepal width (cm), petal length (cm), petal width (cm), species. We also observe that there are 3 different species types (*iris setosa*, *iris versicolour*, *iris virginica*), and 50 samples of each species.

### 1.2 Parsing the dataset

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	class
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

The above screenshot illustrates a sample of the  $N \times p$  dimensional data frame.

In the code, the line `label_vec = df['class'].to_numpy()`<sup>2</sup>, creates the  $N$ -dimensional array of the class type.

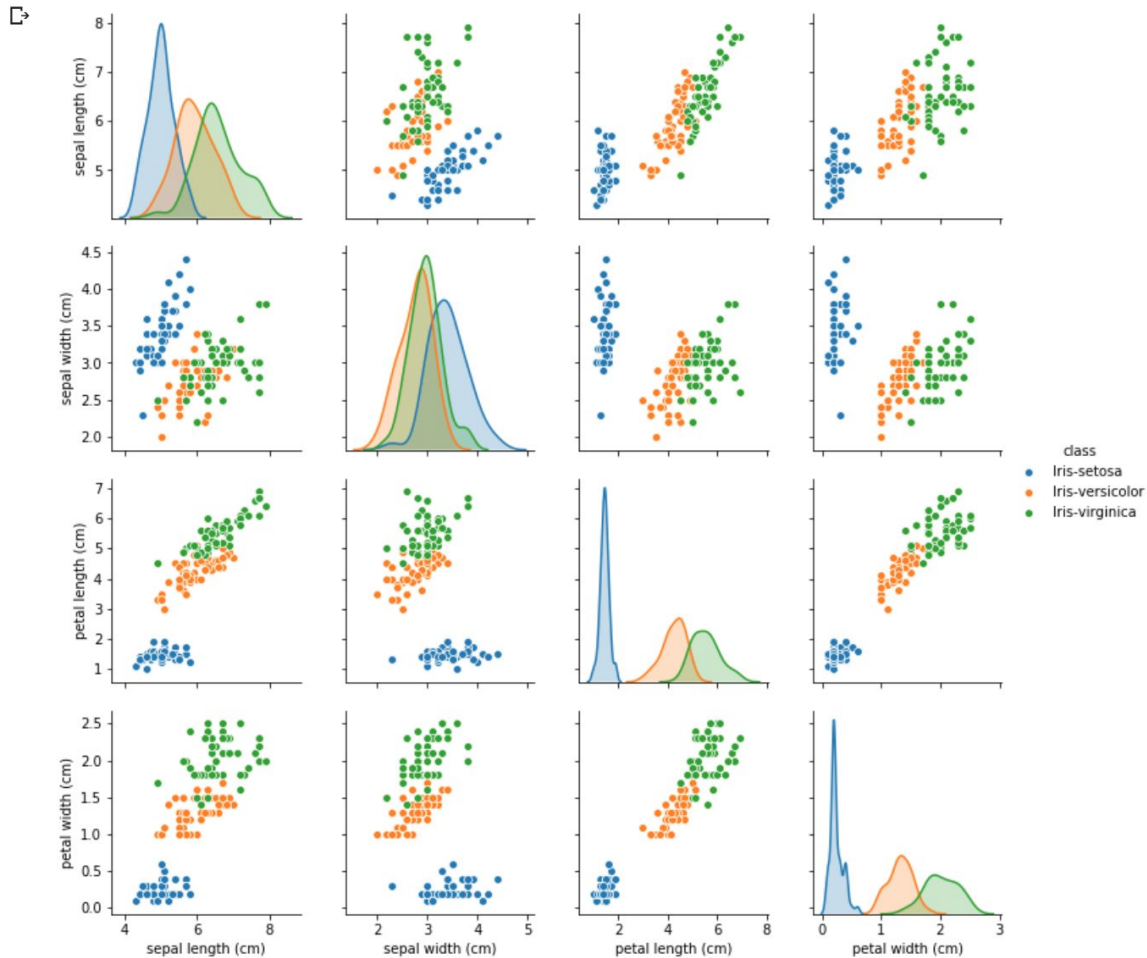
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<sup>1</sup> [https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.read\\_csv.html](https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.read_csv.html)

<sup>2</sup> [https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.to\\_numpy.html](https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.to_numpy.html)

### 1.3 Visualizing the dataset

```
[ ] pairwise_graph = sns.pairplot(df, hue="class").
```



From the above scatterplots, generated using the seaborn library<sup>3</sup>, we can see that petal length and petal width provide the most information about the class/species type. We can also verify this with the high class correlation coefficients found in `iris.names` (0.9490 for petal length, 0.9565 for petal width).

<sup>3</sup> [http://seaborn.pydata.org/examples/scatterplot\\_matrix.html](http://seaborn.pydata.org/examples/scatterplot_matrix.html)