

# Technique Assignment 1: Python

Cogs 109

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```
In [1]: # load matplotlib for plotting
# import datasets so we can use the Fisher's iris dataset
import matplotlib.pyplot as plt
import matplotlib
from sklearn import datasets
```

## 1. Loops in Python

Result:

```
In [2]: s = 0
x = 3
while x<100:
    s += x
    x += 3
print(s)
```

1683

## 2. List operations

Result:

```
In [3]: s = 0
for i in list(range(3,100,3)):
    s+=i
print(s)
```

1683

## 3. Strings

```
In [4]: string = "Hello Python"
sub = string[:5]
sub2 = string[6:]
sub3,sub4 = string.split()
```

```
print(sub,sub2)
print(sub3,sub4)
```

Hello Python  
Hello Python

## 4. Datasets: Fisher's iris data

```
In [5]: # load Fisher's iris dataset
iris = datasets.load_iris()
```

```
In [6]: # check out the data (we can look at small datasets this way, but
        # not big ones)
        #iris.data

        # Once you've looked at the data, be sure to comment out the code
        # above.

        # You should not display unnecessary data in your pdf report,
        # especially if it's a long list of numbers!
```

```
In [7]: # This is part of the dataset too.
        #iris.target

        # Don't display this in your report either!
```

```
In [8]: #print(iris.DESCR)
```

## 5. Plotting data

a. Three classes of iris, 50 of each with 4 numeric, predictive attributes and classes: sepal length in cm, sepal width in cm, petal length in cm, petal width in cm class: Iris-Setosa, Iris-Versicolour, Iris-Virginica

b. 4 variables: sepal length, sepal width, petal length, petal width

```
In [9]: iris.feature_names
```

```
Out[9]: ['sepal length (cm)',
        'sepal width (cm)',
        'petal length (cm)',
        'petal width (cm)']
```

c. 150 samples

```
In [10]: len(iris.target)
```

```
Out[10]: 150
```

d. Yes, three categories: 'setosa', 'versicolor', 'virginica' and labeled samples with 0,1,2

```
In [11]: iris.target_names
```

```
Out[11]: array(['setosa', 'versicolor', 'virginica'], dtype='<U10')
```

```
In [13]: #iris.target
```

e. Not really, they are just raw numbers that looks similar to each other

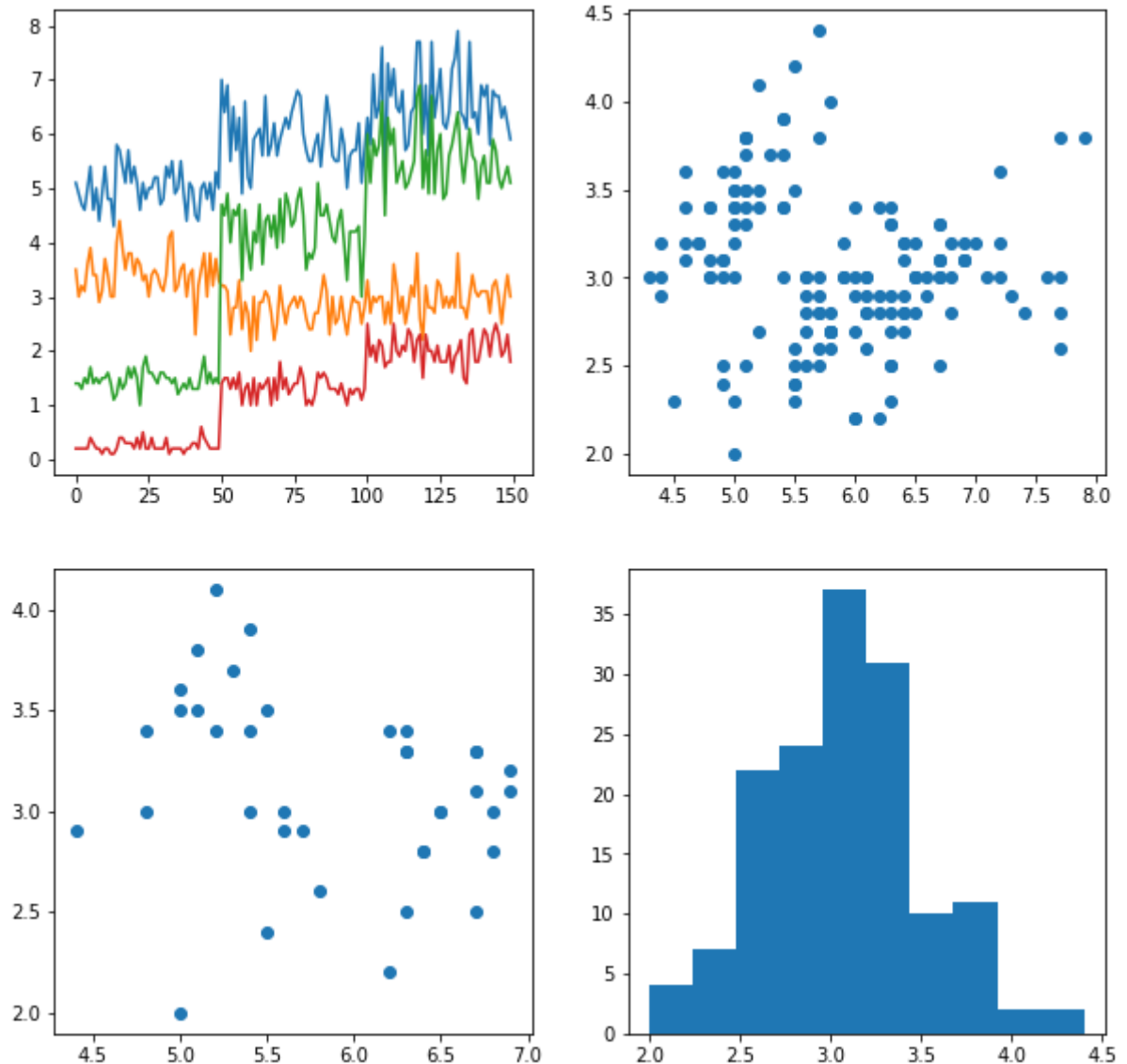
```
In [14]: plt.rcParams['figure.figsize'] = [10,10]
fig, axs = plt.subplots(2,2)
axs[0,0].plot(iris.data)

x = iris.data[:,0]
y = iris.data[:,1]
axs[0,1].scatter(x,y)

x1 = x[0::4]
y1 = y[0::4]
axs[1,0].scatter(x1,y1)

axs[1,1].hist(y)
```

```
Out[14]: (array([ 4.,  7., 22., 24., 37., 31., 10., 11.,  2.,  2.]),
array([2.   , 2.24, 2.48, 2.72, 2.96, 3.2  , 3.44, 3.68, 3.92, 4.16, 4.4  ]),
<BarContainer object of 10 artists>)
```



*a.* x: different samples, 150 in total; different lines represent different features, which are sepal length, sepal width, petal length, petal width.

*b.* Since there are 50 samples for each kind of iris, there are dramatic rises in features at the 50, 100 of x axis.

*c.* above

*d.* There seems to be two clusters/lines in the plot, which may imply the connection between the first and second feature.

*e.* above

*f.* It has the shape of a normal distribution, but slightly left-skewed, which implies more iris has size around 3.0-3.5, and very few smaller than 2.5 and bigger than 4.0.

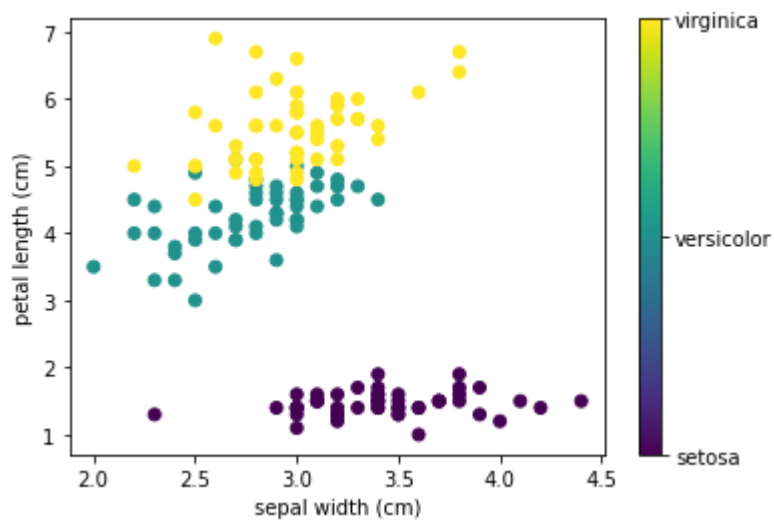
## 6. Plotting with labels

```
In [15]: x = 1
          y = 2

          formatter = plt.FuncFormatter(lambda i, args:
```

```
iris.target_names[i])
plt.figure(figsize=(6, 4))
plt.scatter(iris.data[:, x], iris.data[:, y], c = iris.target)
plt.xlabel(iris.feature_names[x])
plt.ylabel(iris.feature_names[y])

plt.colorbar(ticks=[0, 1, 2], format=formatter)
plt.show()
```



In [ ]: