

# Unit 5: Basic plotting

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November 14, 2020

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## 1 Unit 5: Basic plotting

In this unit, we take a first look at plotting numerical data. Python itself does not have any built-in plotting capabilities, so we will be using *matplotlib* (MPL), the most popular graphics library for Python.

- For details on a plotting function, have a look at the official documentation at <https://matplotlib.org/>
- There is an official introductory [tutorial](#) which you can use along-side this unit.

When using matplotlib in interactive Jupyter notebooks (such as this one), we can enable a more fancy plotting backend that allows us to dynamically adjust the zoom, etc. This is done by adding the line

```
%matplotlib widget
```

For this to work, the `ipyml` package needs to be installed, see [here](#) for details. Note that this is not supported or required (and in fact produces a syntax error) in regular `*.py` Python script files.

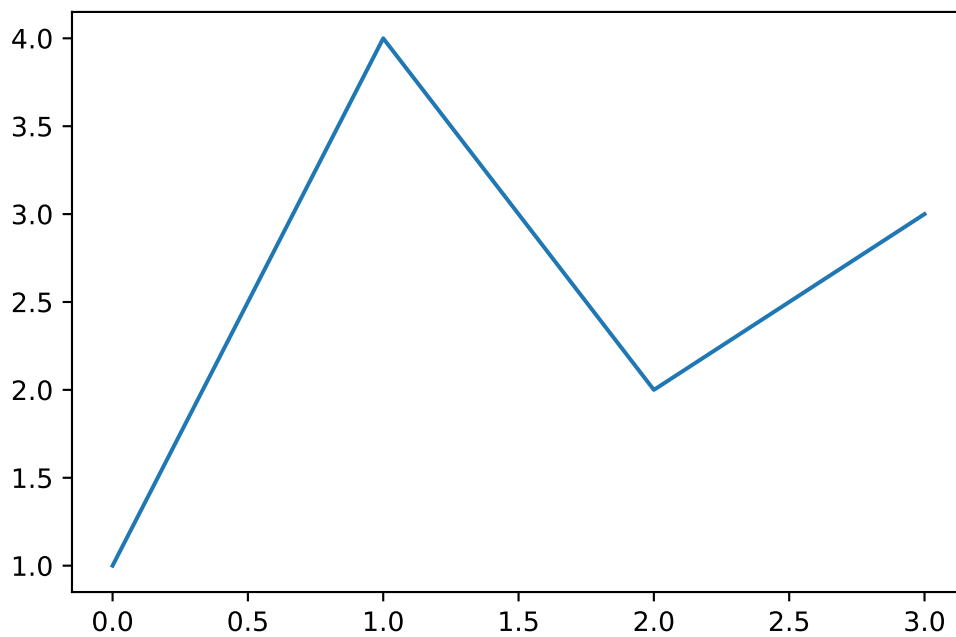
### 1.1 Line and scatter plots

One of the simplest plots we can generate is a line described by a list of points.

```
[1]: # import matplotlib library
import matplotlib.pyplot as plt

# Plot list of integers
yvalues = [1, 4, 2, 3]
plt.plot(yvalues)
```

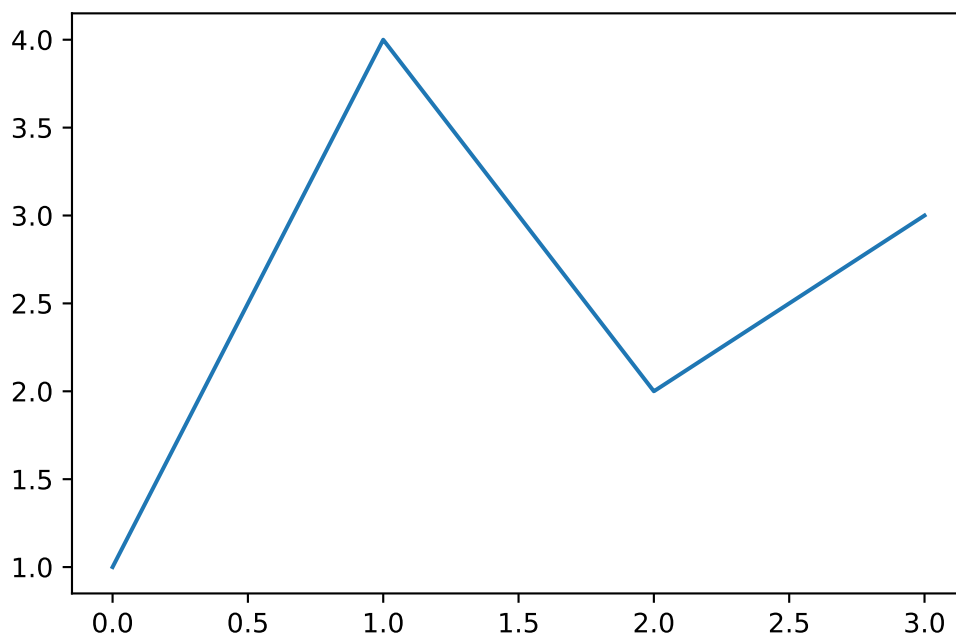
```
[1]: [<matplotlib.lines.Line2D at 0x7f27613a0f10>]
```



We didn't even have to specify the corresponding  $x$ -values, as MPL automatically assumes them to be  $[0, 1, 2, \dots]$

```
[2]: # explicitly specify x-values  
xvalues = [0, 1, 2, 3]  
plt.plot(xvalues, yvalues)
```

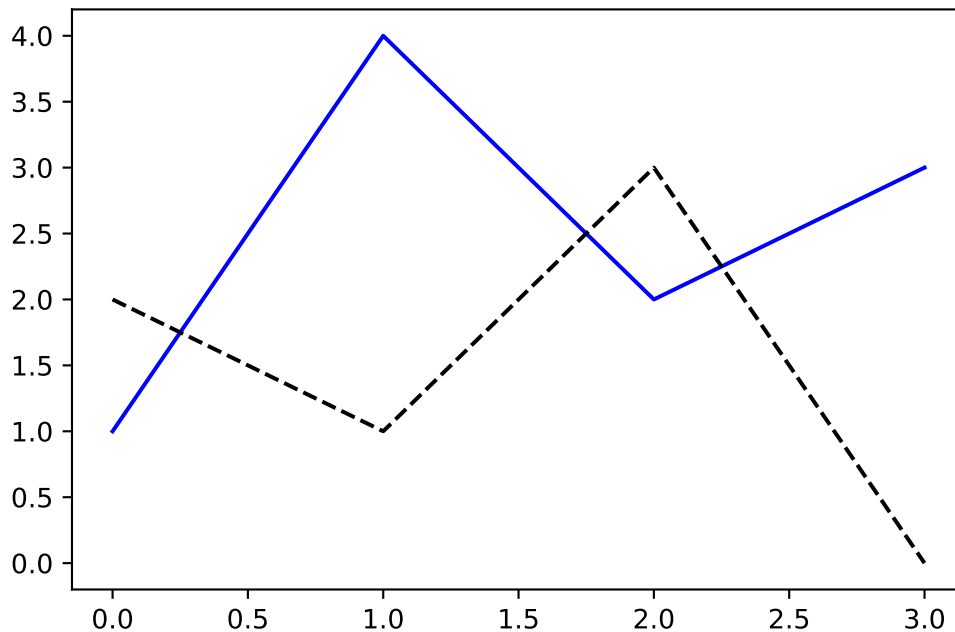
```
[2]: [<matplotlib.lines.Line2D at 0x7f2734e8cbd0>]
```



Similar to Matlab, we can also specify multiple lines to be plotting in a single graph:

```
[3]: yvalues2 = [2.0, 1.0, 3.0, 0.0]
plt.plot(xvalues, yvalues, 'b-', xvalues, yvalues2, 'k--')
```

```
[3]: [<matplotlib.lines.Line2D at 0x7f2734e0c510>,
      <matplotlib.lines.Line2D at 0x7f2734e0c390>]
```



The characters following each set of  $y$ -values are style specifications that are very similar to the ones used in Matlab. More specifically, the letters are short-hand notations for colours: - b: blue - g: green - r: red - c: cyan - m: magenta - y: yellow - k: black - w: white

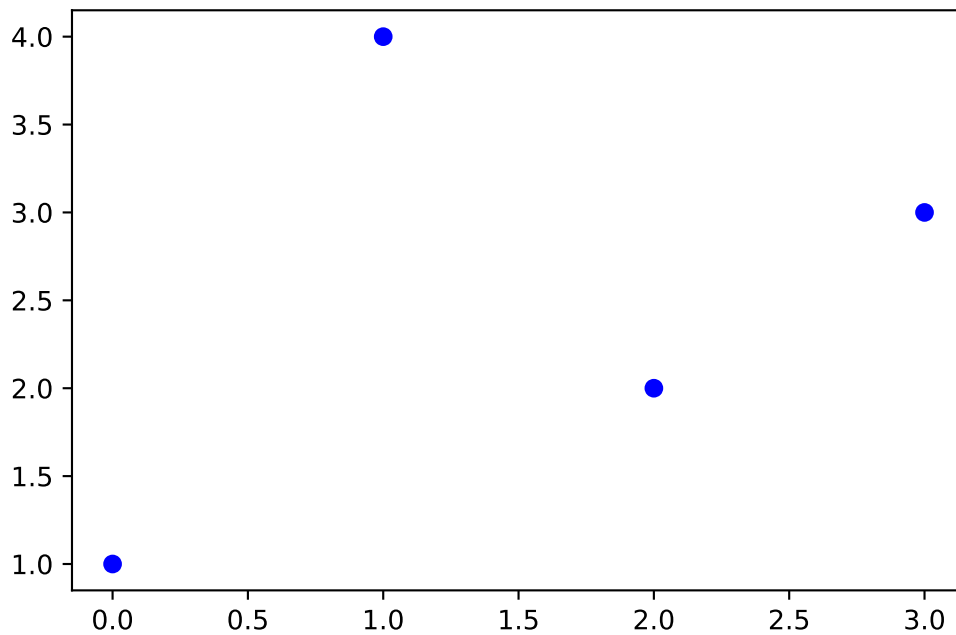
The remaining characters set the line styles. Valid values are -, --, -. and :, with the obvious interpretation.

It is possible to use any RGB colour or used one of the many predefined ones (see [here](#) for details), and to create your own line styles.

We use the `scatter()` routine to create scatter plots in a similar fashion:

```
[4]: plt.scatter(xvalues, yvalues, color='blue')
```

```
[4]: <matplotlib.collections.PathCollection at 0x7f2744809710>
```



## 1.2 Plotting categorical data

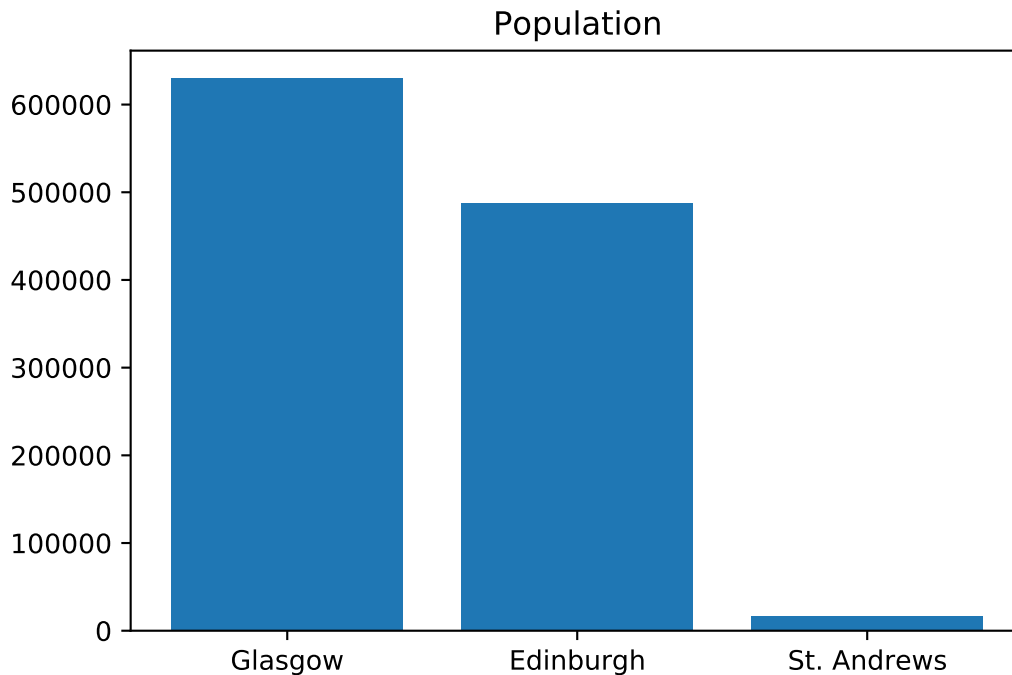
Instead of numerical values on the x-axis, we can also plot categorical variables by passing them directly to the plotting function.

For example, assume we have three categorical “groups” and each has an associated numerical value:

```
[5]: import matplotlib.pyplot as plt

cities = ['Glasgow', 'Edinburgh', 'St. Andrews']
population = [630000, 488000, 16800]
plt.bar(cities, population)
plt.title('Population')
```

```
[5]: Text(0.5, 1.0, 'Population')
```



### 1.3 Adding labels and annotations

Matplotlib has numerous functions to add labels and annotations:

- Use `title()` and `suptitle()` to add titles to your graphs. The latter adds a title for the whole figure, which might span multiple plots (axes).
- We can add axis labels by calling `xlabel()` and `ylabel()`.
- To add a legend, call `legend()`, which in its most simple form takes a list of labels which are in the same order as the plotted data. [\[docs\]](#)
- Use `text()` to add additional text at arbitrary locations. [\[docs\]](#)
- Use `annotate()` to display text next to some data point; it's easier to position correctly than `text()` and you can add arrows! [\[docs\]](#)

```
[6]: import matplotlib.pyplot as plt

xvalues = [0, 1, 2, 3]
yvalues = [1, 4, 2, 3]
yvalues2 = [2.0, 1.0, 3.0, 0.0]

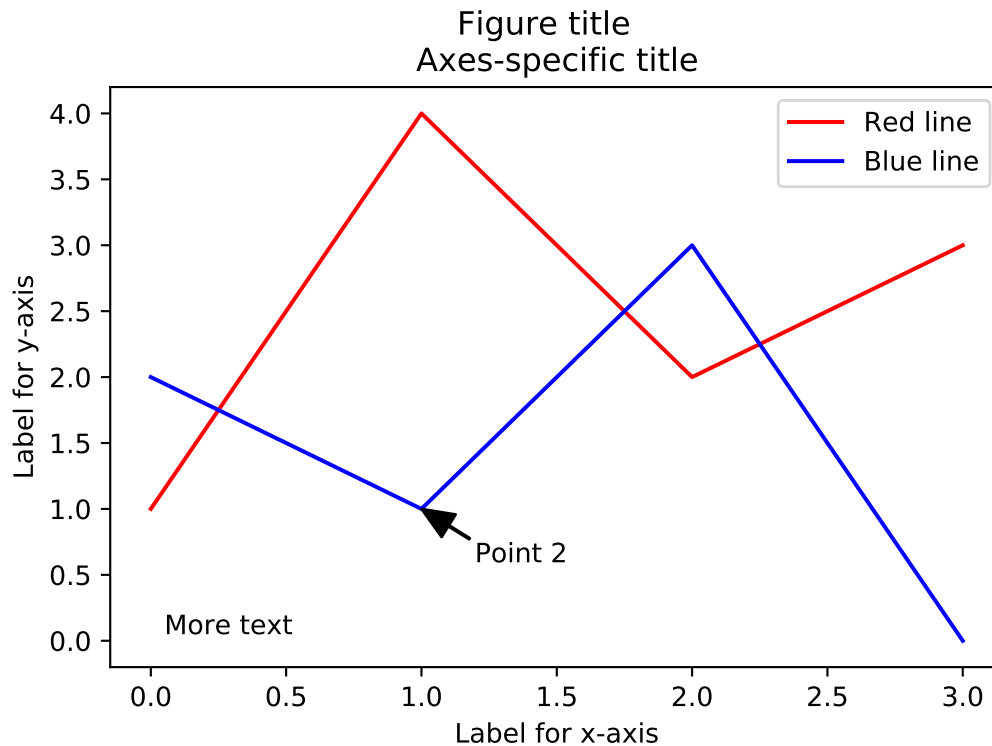
plt.plot(xvalues, yvalues, 'r', xvalues, yvalues2, 'b')
plt.suptitle('Figure title')
plt.title('Axes-specific title')
plt.xlabel('Label for x-axis')
plt.ylabel('Label for y-axis')
plt.legend(['Red line', 'Blue line'])

# Adds text at data coordinates (0.05, 0.05)
plt.text(0.05, 0.05, 'More text')

# Annotate second point
plt.annotate('Point 2', (xvalues[1], yvalues2[1]), (20, -20),
            textcoords='offset points',
```

```
arrowprops={'facecolor': 'black', 'width': 0.5,
            'headwidth': 10.0})
```

```
[6]: Text(20, -20, 'Point 2')
```



## 1.4 Plot limits, ticks and tick labels

We adjust the plot limits, ticks and tick labels as follows:

- Plotting limits are set using the `xlim()` and `ylim()` functions. Each accepts a tuple (min, max) to set the desired range.
- Ticks and tick labels can be set by calling `xticks()` or `yticks()`.

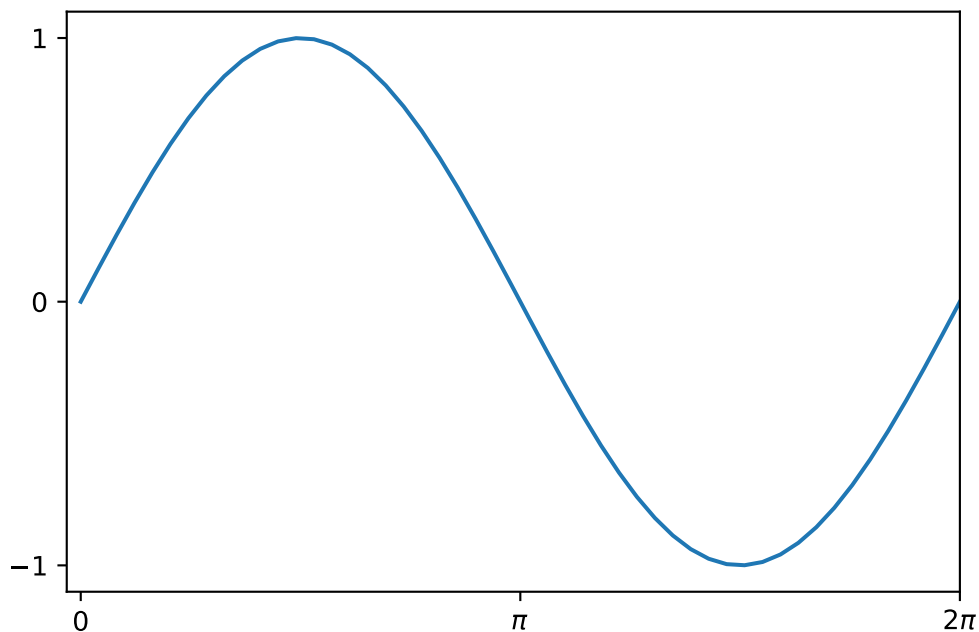
```
[7]: import matplotlib.pyplot as plt
import numpy as np

xvalues = np.linspace(0.0, 2*np.pi, 50)
plt.plot(xvalues, np.sin(xvalues))

# Adjust plot limits in x and y direction
plt.xlim((-0.1, 1.1))
plt.ylim((-1.1, 1.1))

# Set major ticks for x and y axes, and xtick labels.
# We can use LaTeX code in labels!
plt.xticks([0.0, np.pi, 2*np.pi], ['0', r'$\pi$', r'$2\pi$'])
plt.yticks([-1.0, 0.0, 1.0])
```

```
[7]: ([<matplotlib.axis.YTick at 0x7f2734c22350>,
      <matplotlib.axis.YTick at 0x7f2734c1ce10>,
      <matplotlib.axis.YTick at 0x7f2734c1ac90>],
      [Text(0, 0, ''), Text(0, 0, ''), Text(0, 0, '')])
```



## 1.5 Object-oriented interface

So far, we have only used the so-called `pyplot` interface which involves calling *global* plotting functions from the `plt` module. This interface is intended to be similar to Matlab, but is also somewhat limited and less clean.

We can instead use the object-oriented interface (called this way because we call methods of the `Figure` and `Axes` objects instead). While there is not much point in using the object-oriented interface in a Jupyter notebook, it should be the preferred method when writing re-usable code in Python files.

To use the object-oriented interface, we need to get figure and axes objects. The easiest way to accomplish this is using the `subplots()` function, like this:

```
fig, ax = plt.subplots()
```

As an example, we recreate the graph from the section on labels and annotations using the object-oriented interface:

```
[8]: import matplotlib.pyplot as plt

xvalues = [0, 1, 2, 3]
yvalues = [1, 4, 2, 3]
yvalues2 = [2.0, 1.0, 3.0, 0.0]

fig, ax = plt.subplots()
ax.plot(xvalues, yvalues, color='red', label='Red line')
ax.plot(xvalues, yvalues2, color='blue', label='Blue line')
ax.set_xlabel('Label for x-axis')
ax.set_ylabel('Label for y-axis')
ax.legend()
```

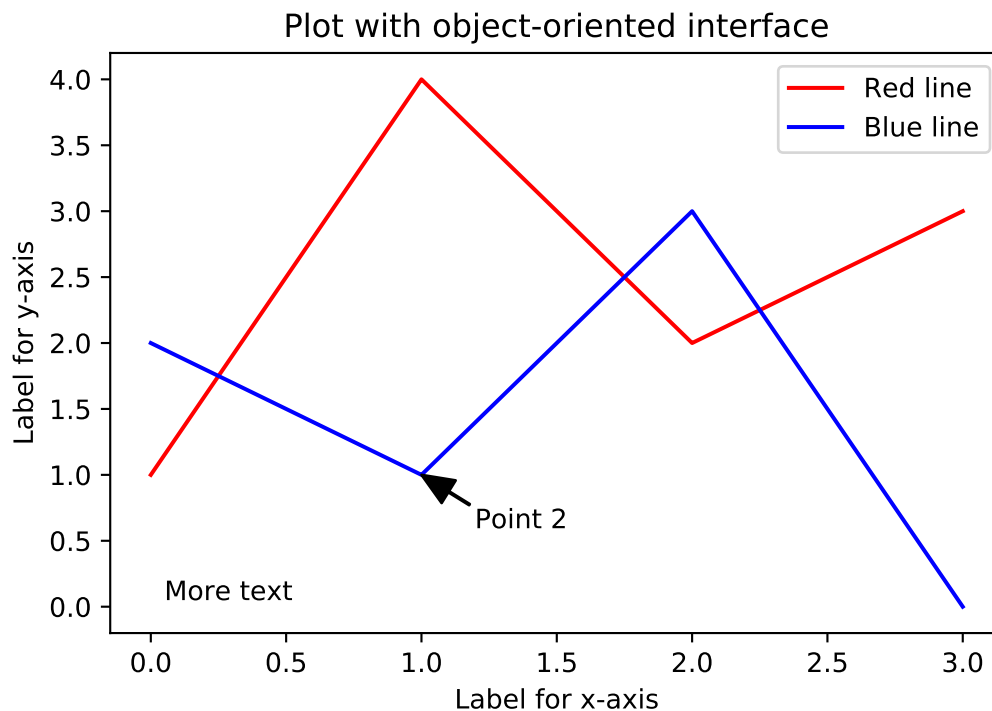
```

ax.set_title('Plot with object-oriented interface')
ax.text(0.05, 0.05, 'More text')

# Annotate second point
plt.annotate('Point 2', (xvalues[1], yvalues2[1]), (20, -20),
             textcoords='offset points',
             arrowprops={'facecolor': 'black', 'width': 0.5,
                         'headwidth': 10.0})

```

[8]: Text(20, -20, 'Point 2')



The code is quite similar, except that attributes are set using the `set_xxx()` methods of the `ax` object. For example, instead of calling `xlim()`, we use `ax.set_xlim()`.

The above example also illustrates how we can specify plot properties such as styles and legend labels explicitly as keyword arguments when calling `plot()`:

```
ax.plot(xvalues, yvalues, property1=value1, property2=value2, ...)
```

## 1.6 Working with multiple plots (axes)

The object-oriented interface becomes particularly useful if we want to create multiple axes (or figures), which is possible with the `pyplot` programming model, but more obscure.

For example, to create a row with two plots, we use:

```

[9]: import matplotlib.pyplot as plt
import numpy as np

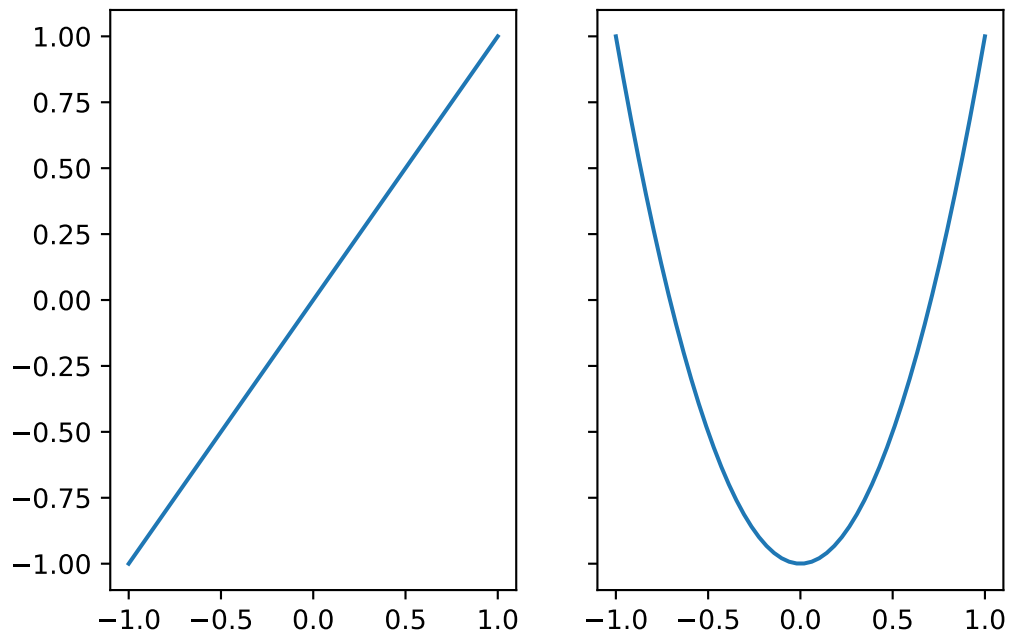
fig, ax = plt.subplots(1, 2, sharex=True, sharey=True)
xvalues = np.linspace(-1.0, 1.0, 50)
ax[0].plot(xvalues, xvalues)

```



```
ax[1].plot(xvalues, 2*xvalues**2.0 - 1)
```

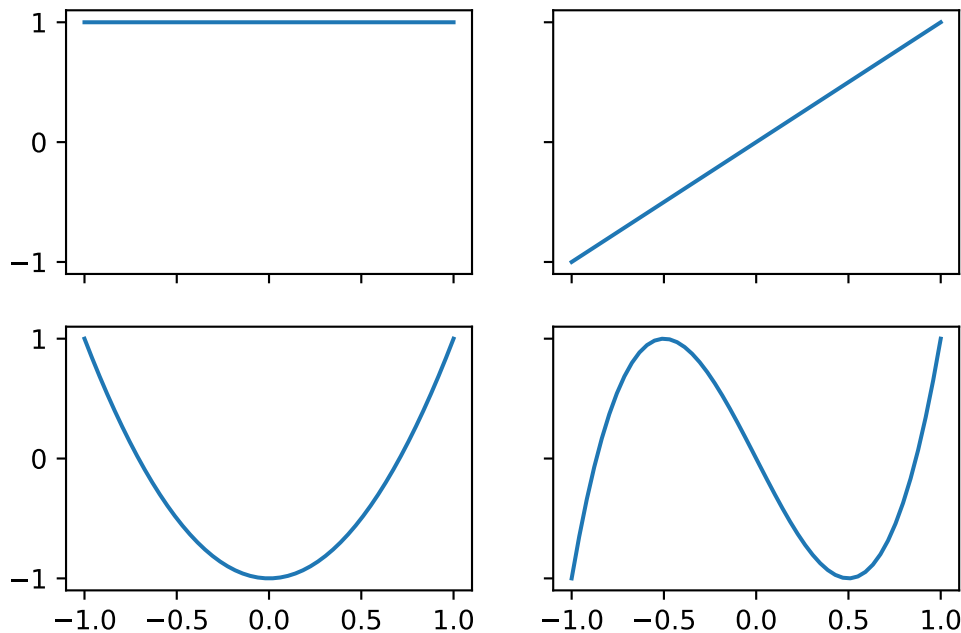
```
[9]: [<matplotlib.lines.Line2D at 0x7f2734dbefd0>]
```



With multiple axes, the second object returned by `subplots()` is actually a NumPy array with a shape that corresponds to the number of plots. If we request a  $2 \times 2$  plot, `ax[0, 0]` will be the top-left axes object, `ax[0, 1]` the top-right one, and so on.

```
[10]: # Plot the first four Chebyshev polynomials on the interval [-1,1]
fig, ax = plt.subplots(2, 2, sharex=True, sharey=True)

xvalues = np.linspace(-1.0, 1.0, 50)
for i in range(2):
    for j in range(2):
        yvalues = np.cos((j + i*2) * np.arccos(xvalues))
        ax[i, j].plot(xvalues, yvalues)
```



Note the use of `sharex=True` and `sharey=True`. This tells matplotlib that all axes share the same plot limits, so the tick labels can be omitted in the figure's interior to preserve space.

## 2 Exercises

### 2.1 Exercise 1: Trigonometric functions

Plot the functions  $\sin(x)$  and  $\cos(x)$  on the interval  $[-\pi, \pi]$ , each in a separate graph. Include a legend for each plot, and add pretty tick labels at  $[-\pi, 0, \pi]$  which use the  $\LaTeX$  symbol for  $\pi$ .

*Hint:* NumPy defines the functions `np.sin()` and `np.cos()` as well as the value `np.pi`.

### 2.2 Exercise 2: Logarithmic scaling

In economics and finance, we often plot using the  $\log_{10}$  scale if the plotted data is of very different orders of magnitude.

Create a figure with two sub-plots, each plotting the function  $f(x) = 10^x$  on a uniformly-spaced interval  $[-5, 5]$  with 100 points. Use the (default) linear scale in the first plot, but apply the  $\log_{10}$  scale in the second.

*Hint:* You can set the axis scale to log by calling `yscale('log')`, or `set_yscale('log')` when using the object-oriented interface.

## 3 Solutions

### 3.1 Solution for exercise 1

```
[11]: import matplotlib.pyplot as plt
import numpy as np

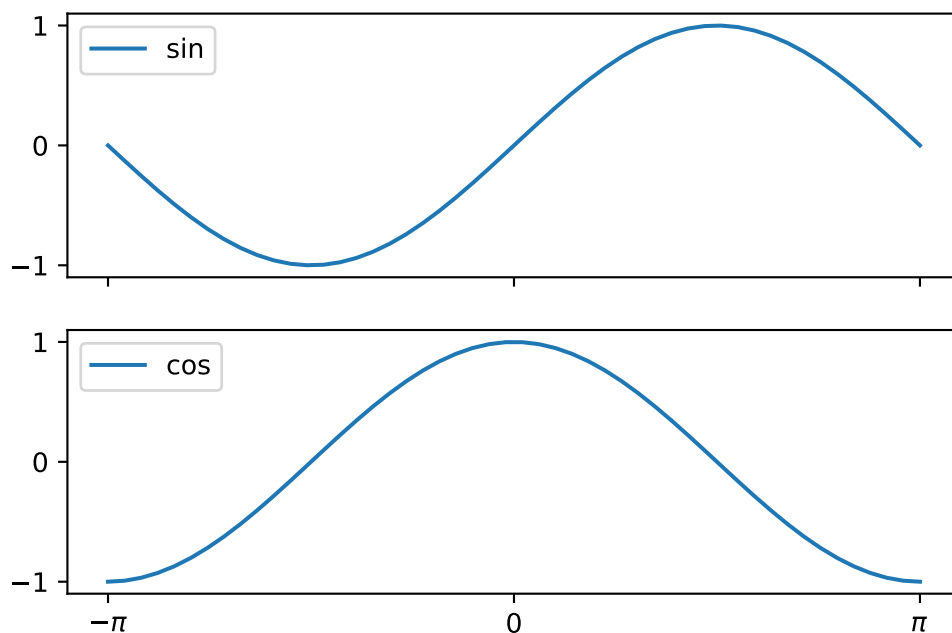
xvalues = np.linspace(-np.pi, np.pi, 50)
# Create figure with two rows, one column
fig, ax = plt.subplots(2, 1, sharey=True, sharex=True)

xticks = [-np.pi, 0.0, np.pi]
xticklabels = [r'$-\pi$', '0', r'$\pi$']
yticks = [-1.0, 0.0, 1.0]

# Create sin() plot using first axes object
ax[0].plot(xvalues, np.sin(xvalues), label='sin')
ax[0].set_xticks(xticks)
ax[0].set_xticklabels(xticklabels)
ax[0].set_yticks(yticks)
ax[0].legend(loc='upper left')

# Create cos() plot using second axes object
ax[1].plot(xvalues, np.cos(xvalues), label='cos')
ax[1].set_xticks(xticks)
ax[1].set_xticklabels(xticklabels)
ax[1].set_yticks(yticks)
ax[1].legend(loc='upper left')
```

```
[11]: <matplotlib.legend.Legend at 0x7f2734a02b90>
```



### 3.2 Solution for exercise 2

```
[12]: import matplotlib.pyplot as plt
import numpy as np

xvalues = np.linspace(-5.0, 5.0, 100)
fig, ax = plt.subplots(1, 2, sharex=True)
ax[0].plot(xvalues, 10.0**xvalues)
ax[0].set_title('Linear scale')

ax[1].plot(xvalues, 10.0**xvalues)
# Set y-axis to log scale (assumes base-10 log)
ax[1].set_yscale('log')
ax[1].set_title('Log scale')
```

```
[12]: Text(0.5, 1.0, 'Log scale')
```

