Basic plotting with matplotlib.pyplot

Learning goals

- to understand the anatomy of a matplotlib plot
- to learn how to make plot using
 - the pyplot interface
 - the "object oriented" method
- to use plotting themes
- to learn how to construct multi-panel figures

matplotlib (https://matplotlib.org/)

Matplotlib is the fundamental library for plotting in Python. It provides a comprehensive set of tools to create both static and dynamic (e.g., interactive) figures.

Matplotlib works really well with numpy arrays, the libary is very large and we will focus specifically on a submodule matplotlib.pyplot which provides an excellent starting point to learn about the platform.

This tutorial will focus less on data, programming and mathy stuff and more on understanding how to make, interpret and customize a plot using pyplot.

The anatomy of a plot

The basic anatomy of a plot reproduced from the <u>matplotlib</u> <u>startup guide</u> <u>(https://matplotlib.org/stable/tutorials/introductory/quick_start.html)</u> is shown below.

I	Drawing
	?
	<u>—</u>

A figure in matplotlib consists of two key pieces:

- the *figure*
- and one or more axes

Matplotlib uses the object *figure* to handle plotting graphs. A *figure* is like a piece of paper or your canvas. Plots are made inside a figure and the objects used to refer to the plots is *axes*. An *axes* is the area insde a *figure* where plots can be made. A Figure can contain one or more.

Confusingly, once we have made a plot in our **axes**, our **axes** will now have at least one **axis** thingie on it – so, in the land of matplotlib, "axes" is **not** the plural of "axis"! Rather...

An **axes** is a part of a **figure** that contains one or more **axis** lines, as well as other things such as a title, x- and y-axis labels, and one or more plots of data.

Happily, we do all of our plotting and formatting by working with our *axes*, we just have to make sure that when we say "axes", we mean "a plotting area on our figure" rather than "a bunch of 'axises' or 'axi' or whatever."

To avoid confusion, we'll refer to an axes as an axes panel from here on out.

Ways to make plots using matplotlib

There are two main ways to make plots using matplotlib (not counting the library seaborn - we'll revisit seaborn later on). The two main ways are:

- setting directly the properties of each object in a figure. This direct method is also called the "object oriented" interface
- using the pyplot interface

Below two examples of the same plot implemented using either style:

OO plotting style example

```
In [2]: import matplotlib as mpl  # the standard acronym to import matplotlib is `mpl`
import matplotlib.pyplot as plt # the standard acronym to import pyplot is `plt`
import numpy as np
```

We plot a quadratic function as an example:

```
In [3]: x = \text{np.array}([1, 2, 3, 4, 5, 6, 7, 8, 9, 10])

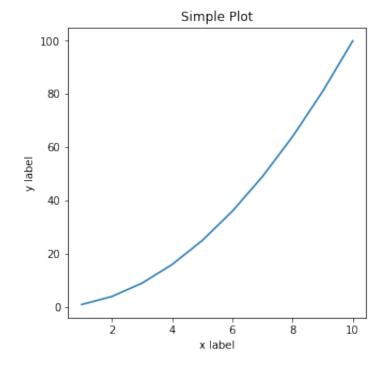
y = x**2
```

Next, w e create the figure (fig) and the axes (ax):

```
In [4]: fig, ax = plt.subplots(figsize=(5, 5)) ## Note here that for semplicity we actually use pyplot to
# Plot the data on the axes
ax.plot(x, y)

# format the axes and add information to them
ax.set_xlabel('x label') # Add an x-label to the axes.
ax.set_ylabel('y label') # Add a y-label to the axes.
ax.set_title("Simple Plot") # Add a title to the axes.
```

Out[4]: Text(0.5, 1.0, 'Simple Plot')

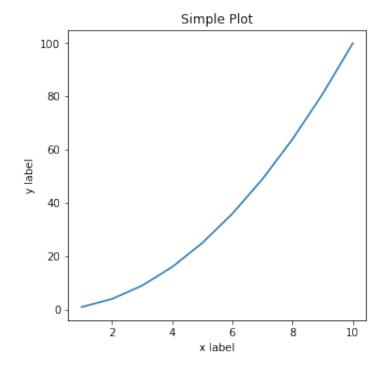


pyplot plotting style

Open the figure using pyplot:

```
In [5]: plt.figure(figsize=(5, 5))
plt.plot(x, y) # Plot some data on the (implicit) axes.
# format the axis and add information to them
plt.xlabel('x label')
plt.ylabel('y label')
plt.title("Simple Plot")
```

Out[5]: Text(0.5, 1.0, 'Simple Plot')



Note the difference with the previous call to pyplot where we returned both fig and ax, here instead we use the implicit method and change values to the parameters contained in the pyplot object plt, everything is set inside plt.

Complete the following exercise.

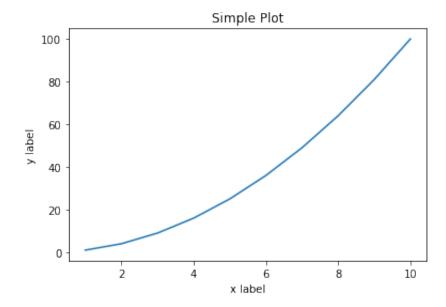
• What happens if you execute the command plt.figure(figsize=(5, 5)) in the first code cell below and the rest of the commands above (from plt.plot(x,y) onwards) in the second code cell below?

[Use the code cell below to show the code]

[Use the markdown cell below to provide your interpretation of what might be happening here]

```
In [7]: plt.plot(x, y) # Plot some data on the (implicit) axes.
# format the axis and add information to them
plt.xlabel('x label')
plt.ylabel('y label')
plt.title("Simple Plot")
```

Out[7]: Text(0.5, 1.0, 'Simple Plot')



[The shape of the figure were changed from 5x5 in the original data to rectangular shape.]

We will focus first on the pyplot interface and go back to the OO interface later. This means that we will learn first how to

import matlibplot.pyplot as plt

and the use the plt.plot() command (which I like to pronounce "plit dot plot") to do our plotting. The plt.plot() automatically creates a **figure** and **axes panel** for us when we start plotting. For simple plots, we don't even have to know that there is an official thing called a "figure" and a another official thing called an "axes", we just start plotting!

In the object-oriented way, we create a **figure** and one or more **axes** (axes panels) explicitly, and then we use the methods (functions) associated with those objects to do our plotting and formatting. (This will make more sense when we do some examples.)

The fact that there are these two different ways to plot can make search results or StackExchange conversations confusing! So we'll go through these in turn, doing the same thing using two different methods.

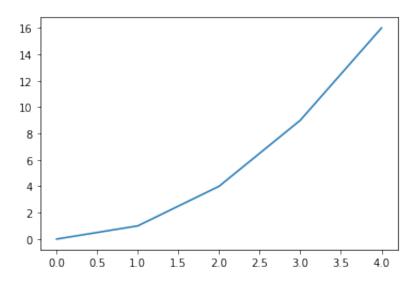
Plotting Python lists using matplotlib's pyplot

Let's breakdown the operations that pyplot performs for us, by making another plot! The steps are simple. We'll:

- import pyplot (we have done that above)
- make (simulate) some data to plot
- plot the data using plt.plot()
- breakdown the conceptual operations plt performs for us

```
In [12]: my_x_vals = [0, 1, 2, 3, 4]
    my_y_vals = [0, 1, 4, 9, 16]
    plt.plot(my_x_vals, my_y_vals)
```

Out[12]: [<matplotlib.lines.Line2D at 0x7fc79ad31700>]



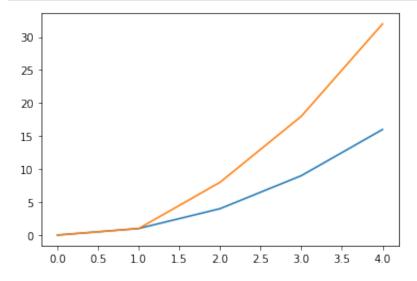
Yay! What plt.plot() has done for us here in a single command is

- 1. created a *figure*
- 2. created a single axes panel on that figure
- 3. made both an x- and a y-axis (with tick marks) on the axes panel
- 4. added x and y tick labels adjacent to each axis (on the axes panel)
- 5. finally, plotted the data using the default line style and color (on the axes panel)

Let's next plot a couple of functions on top of eachother:

```
In [13]: my_x_vals = [0, 1, 2, 3, 4]
    my_y_vals = [0, 1, 4, 9, 16]
    my_y_vals2 = [0, 1, 8, 18, 32]

# Creating 2 lines into one graph
    plt.plot(my_x_vals, my_y_vals);
    plt.plot(my_x_vals, my_y_vals2);
```



Complete the following exercise.

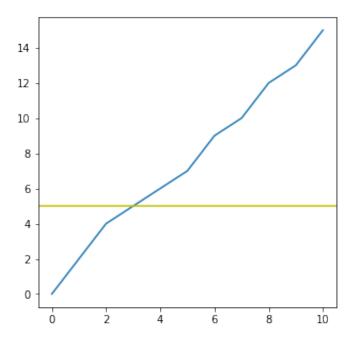
- given the data x = [0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10]
- create a new figure with y and x axis of equal dimensions (hint, we did this above)
- use plt to plot a horizontal line fixed at 5

[Use the code cell below to implement the plot]

```
In [14]: x = [0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10]
y = [0, 2, 4, 5, 6, 7, 9, 10, 12, 13, 15]

plt.figure(figsize=(5, 5))
plt.plot(x,y)
plt.axhline(y=5, color='y', linestyle='-')
```

Out[14]: <matplotlib.lines.Line2D at 0x7fc79aef1190>

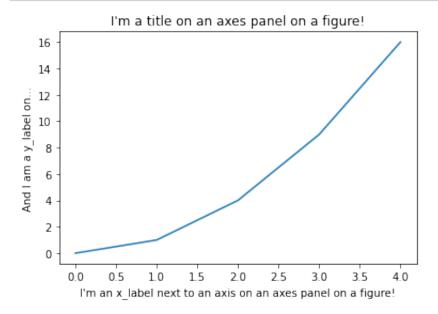


Adding axis labels and titles

Some common things we want to add are labels for the x- and y-axis, and a title for our plot. Handely, pyplot has functions for all of these. They are:

- xlabel()
- ylabel()
- title()

```
In [15]: plt.plot(my_x_vals, my_y_vals);
   plt.xlabel('I\'m an x_label next to an axis on an axes panel on a figure!');
   plt.ylabel('And I am a y_label on...');
   plt.title('I\'m a title on an axes panel on a figure!');
```



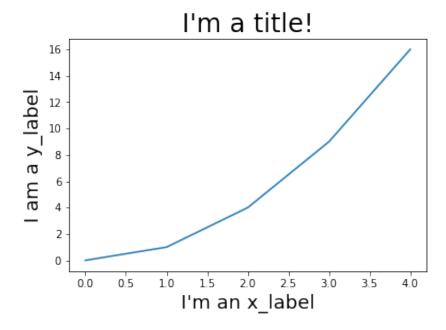
Side note: Hey, what's up with the backslash inside the xlabel and title? In python, strings are defined by either single or double quotes. Like this:

But what happens when you want an apostophe (half a single quote) to be part of your string? This won't work:

So Python (and all programming languages) provide an "escape character" that says "the thing right after me is part of the string!" So if we want to have our apostrophe be part of the sting, we "escape" it with the escape character "" (the backslash). So this works:

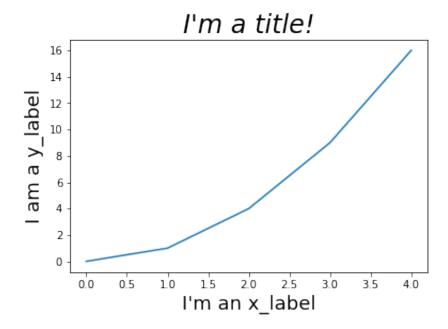
The label and title commands allow us to specify the font size as well (**Pro tip**: Unless you are plotting just for yourself, you almost always have to make the font sizes larger than the default!)

```
In [20]: plt.plot(my_x_vals, my_y_vals);
   plt.xlabel('I\'m an x_label', fontsize = 18);
   plt.ylabel('I am a y_label', fontsize = 18);
   plt.title('I\'m a title!', fontsize = 24);
```



Each of these also allows you to do italics if you wish.

```
In [21]: plt.plot(my_x_vals, my_y_vals);
   plt.xlabel('I\'m an x_label', fontsize = 18);
   plt.ylabel('I am a y_label', fontsize = 18);
   plt.title('I\'m a title!', fontsize = 24, fontstyle = 'italic');
```

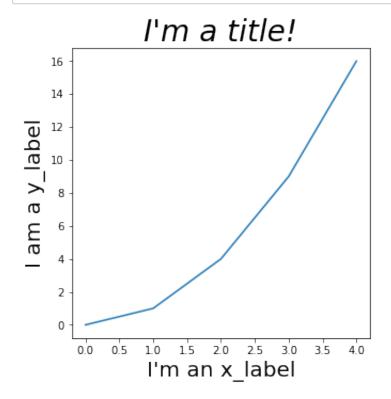


Complete the following exercise.

- repeate the plot in the previous cell but open a new figure of size x=5 and y=5 (hint we have done this above)
- change the font size of the title to 28 and of the axis to 20

[Use the code cell below to implement the plot]

```
In [22]: plt.figure(figsize=(5, 5))
   plt.plot(my_x_vals, my_y_vals);
   plt.xlabel('I\'m an x_label', fontsize = 20);
   plt.ylabel('I am a y_label', fontsize = 20);
   plt.title('I\'m a title!', fontsize = 28, fontstyle = 'italic');
```



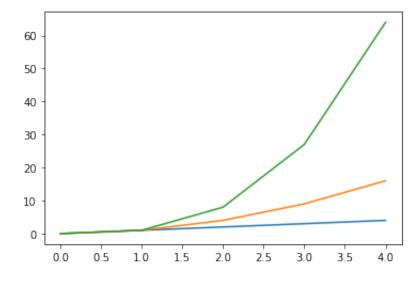
Multiple plots on an axes panel

If we wish to plot multiple things, plt.plot() will change the colors for us automatically.

```
In [23]: my_x = np.array([0, 1, 2, 3, 4])

y_lin = my_x
y_quad = my_x**2
y_cube = my_x**3

plt.plot(my_x, y_lin);
plt.plot(my_x, y_quad);
plt.plot(my_x, y_cube);
```



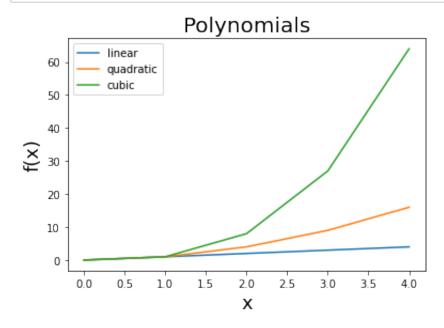
Legends

Crucially, we can give each plot a **label** so that we can add a **legend**.

```
In [24]: plt.plot(my_x, y_lin, label = 'linear');
plt.plot(my_x, y_quad, label = 'quadratic');
plt.plot(my_x, y_cube, label = 'cubic');

plt.xlabel('x', fontsize = 18);
plt.ylabel('f(x)', fontsize = 18);
plt.title('Polynomials', fontsize = 20,);

plt.legend();
```

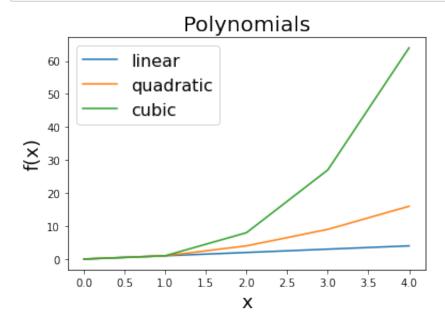


You can change the font size for the legend just like you do the title, etc. You can also specify the location of the legend with the loc= parameter. The default is 'best'.

```
In [25]: plt.plot(my_x, y_lin, label = 'linear');
plt.plot(my_x, y_quad, label = 'quadratic');
plt.plot(my_x, y_cube, label = 'cubic');

plt.xlabel('x', fontsize = 18);
plt.ylabel('f(x)', fontsize = 18);
plt.title('Polynomials', fontsize = 20,);

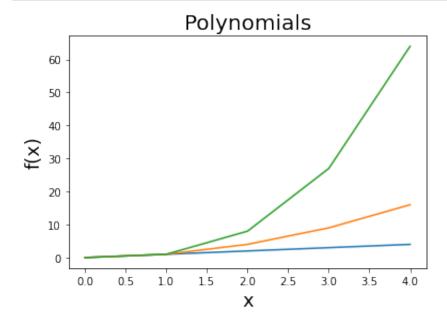
plt.legend(loc = 'upper left', fontsize = 16);
```



Let's take minute to try and guess what some different allowed locations are!

```
In [26]: plt.plot(my_x, y_lin, label = 'linear');
plt.plot(my_x, y_quad, label = 'quadratic');
plt.plot(my_x, y_cube, label = 'cubic');

plt.xlabel('x', fontsize = 18);
plt.ylabel('f(x)', fontsize = 18);
plt.title('Polynomials', fontsize = 20,);
```



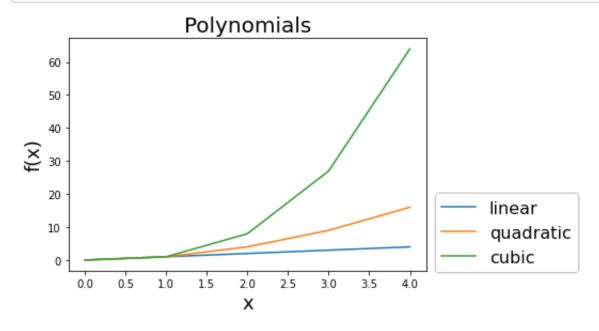
If we want even finer control over legend position (if, for example, we wanted to move the legend outside the axis limits), we can specify the position "manually". To do this, we will need to set the property bbox_to_anchor instead of loc.

Let's try it:

```
In [27]: plt.plot(my_x, y_lin, label = 'linear');
plt.plot(my_x, y_quad, label = 'quadratic');
plt.plot(my_x, y_cube, label = 'cubic');

plt.xlabel('x', fontsize = 18);
plt.ylabel('f(x)', fontsize = 18);
plt.title('Polynomials', fontsize = 20,);

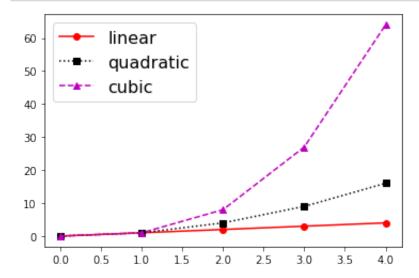
plt.legend(bbox_to_anchor =(1.0, 0.37), fontsize = 16);
```



Adjusting colors, markers, and line styles

We can also specify "by hand" what the *colors*, *lines*, and data point *markers* look like by adding a string containing a code for each. The string has to come right after the data. Like this:

```
In [29]: plt.plot(my_x, y_lin, '-ro', label = 'linear'); # note the cryptic string in the middle!
plt.plot(my_x, y_quad, ':ks', label = 'quadratic');
plt.plot(my_x, y_cube, '--m^', label = 'cubic');
plt.legend(loc = 'upper left', fontsize = 16);
```



There are just 4 linestyles, the three above and '-.' For reference, they are:

code	line style
-	solid
	dashed
:	dotted
	dot-dash

The common colors that are available are:

code	line style
r	red
g	green

b	blue
k	black
у	yellow
m	majenta
С	cyan

There may be others...

Some common markers are:

code	line style
0	circle
	point
s	square
d	diamond
^, >, <, v	triangles

There are a few others, which you can find https://www.mathworks.com/help/matlab/creating-plots/specify-line-and-marker-appearance-in-plots.html).

Plot styles

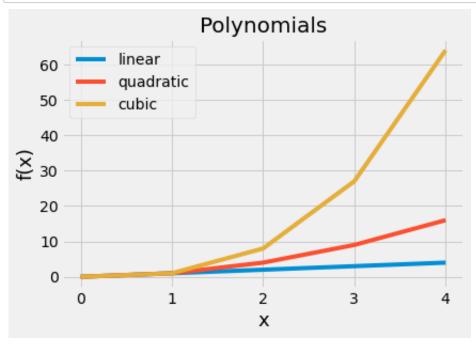
pyplot offers several built-in styles that allow you to change the overall appearance – the look and feel – of your plots. We set these with the plt.style.use() function:

```
In [30]: plt.style.use('fivethirtyeight') # set style to Nate Silver's

plt.plot(my_x, y_lin, label = 'linear');
plt.plot(my_x, y_quad, label = 'quadratic');
plt.plot(my_x, y_cube, label = 'cubic');

plt.xlabel('x', fontsize = 18);
plt.ylabel('f(x)', fontsize = 18);
plt.title('Polynomials', fontsize = 20,);

plt.legend();
```



Nice!!

You can see what styles are available like this:

```
In [31]:
         plt.style.available
Out[31]:
         ['Solarize_Light2',
           '_classic_test_patch',
           '_mpl-gallery',
           '_mpl-gallery-nogrid',
           'bmh',
           'classic',
           'dark_background',
           'fast',
           'fivethirtyeight',
           'ggplot',
           'grayscale',
           'seaborn',
           'seaborn-bright',
           'seaborn-colorblind',
           'seaborn-dark',
           'seaborn-dark-palette',
           'seaborn-darkgrid',
           'seaborn-deep',
           'seaborn-muted',
           'seaborn-notebook',
           'seaborn-paper',
           'seaborn-pastel',
           'seaborn-poster',
           'seaborn-talk',
           'seaborn-ticks',
           'seaborn-white',
           'seaborn-whitegrid',
           'tableau-colorblind10']
```

And you can see what each one looks like here (https://www.dunderdata.com/blog/view-all-available-matplotlib-styles).

You can always go back to the default style by specifying default, like this:

```
In [32]:
    plt.style.use('default') # back to default, please
```

Figures with multiple axis panels

If you open any paper, book, or website, you will see figures with multiple panels showing different but related things. In pyplot, this is accomplished with plt.subplot().

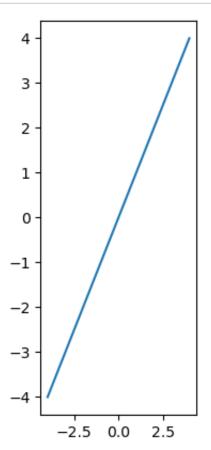
(Let's remake the polynomial so they look more interesting.)

```
In [33]: my_x = np.array([-4, -3, -2, -1, 0, 1, 2, 3, 4])

y_lin = my_x
y_quad = my_x**2  # ** is Python for exponentiation
y_cube = my_x**3
```

Now we'll plot our polynomials in separate panels. We'll start with just one.

```
In [34]: plt.subplot(131)
  plt.plot(my_x, y_lin, label = 'linear');
```



I know it looks funny, but hang on. What happened is that plt.subplot(131) made a figure for us. Then, the 131 told it to divide the figure into 1 row by 3 columns of axes panels, and make the 1st axes panel the "active" one.

So the three numbers are ("how many rows" "how many columns" "which one to plot in").

So now let's plot all three of our polynomials.

In [42]:

```
plt.style.use('default')

# For plt.subplt(#1#2#3), Ex: plt.subplot(131) means "1" = 1 axes panel y; "3" = axes panel x; "...

plt.subplot(131)  # make the first (left hand) axes panel the active one
plt.plot(my_x, y_lin, label = 'linear');

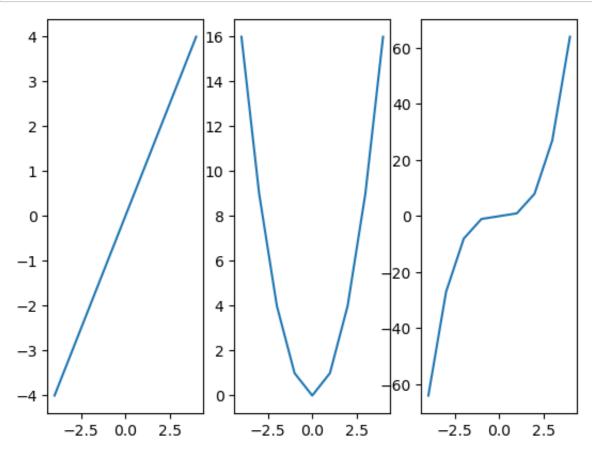
#plt.axis('square')

plt.subplot(132)  # make the second panel active
plt.plot(my_x, y_quad, label = 'quadratic');

#plt.axis('square')

plt.subplot(133)  # make the third panel active
plt.plot(my_x, y_cube, label = 'cubic');

#plt.axis('square')
```



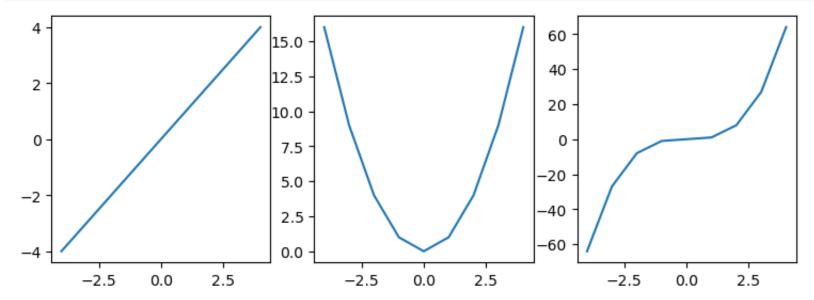
It still looks funny – the panels are tall and skinny and figure only takes up part of the page horizontally. This is because some default figure size is being used. But fear not! We can explicitly make a the figure and set its size (in inches) ourselves with the plt.figure() command. This way we can make our figure have a rational size and aspect ratio for whatever we're trying to plot. So this time, let's make a figure that's 9"x3" for our plotting.

```
In [37]: plt.style.use('default')
    plt.figure(figsize=(9, 3)) # make a wide, short figure

    plt.subplot(131)
    plt.plot(my_x, y_lin, label = 'linear');

    plt.subplot(132)
    plt.plot(my_x, y_quad, label = 'quadratic');

    plt.subplot(133)
    plt.plot(my_x, y_cube, label = 'cubic');
```

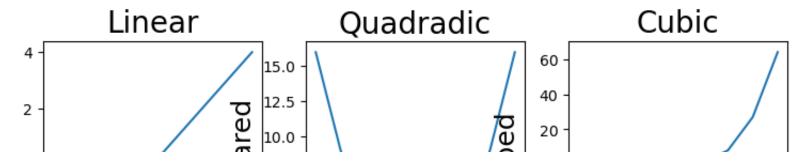


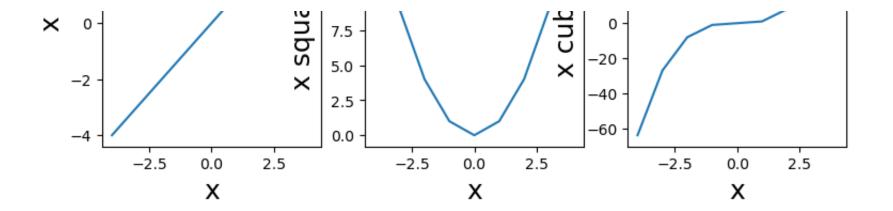
Ah, much better!

One thing that might be a little annoying is that everything else in Python has the row (height) come first, but here we specify the width first in the figsize argument. The size argument is this way because that's how it works in the printing world (e.g. 8 1/2" by 11" paper). Oh well, we'll deal.

While each plot is active, we can take the opportunity to add axis labels and titles:

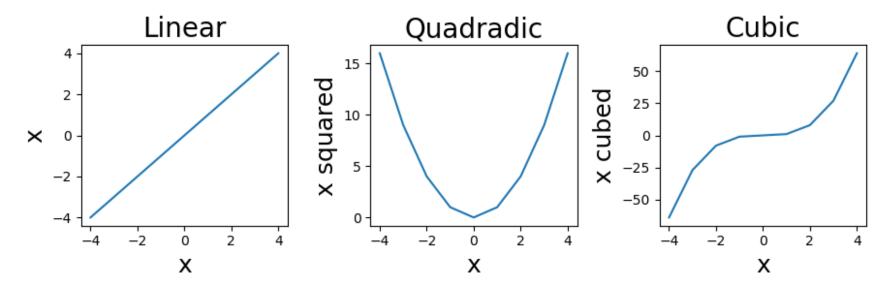
```
In [38]: plt.style.use('default')
         plt.figure(figsize=(9, 3))
                                                   # make the first plot active
         plt.subplot(131)
         plt.plot(my_x, y_lin, label = 'linear'); # do the actual plotting
         plt.xlabel('x', fontsize = 18);
                                              # and while the plot is active
         plt.ylabel('x', fontsize = 18);
                                                   # make it fancy like
         plt.title('Linear', fontsize = 20,);
         plt.subplot(132)
         plt.plot(my_x, y_quad, label = 'quadratic');
         plt.xlabel('x', fontsize = 18);
         plt.ylabel('x squared', fontsize = 18);
         plt.title('Quadradic', fontsize = 20,);
         plt.subplot(133)
         plt.plot(my x, y cube, label = 'cubic');
         plt.xlabel('x', fontsize = 18);
         plt.ylabel('x cubed', fontsize = 18);
         plt.title('Cubic', fontsize = 20,);
```





Okay, all good, right? Oh... wait... our y axis labels are weird, like there's not enough room in the figure to spread the axes panels out. We could play around with making the figure wider, but there is also a function, plt.tight_layout() which attempts to arrange everything for us with no weird clipping or overlapping. So let's try the exact same code with that at the end:

```
In [39]: plt.style.use('default')
         plt.figure(figsize=(9, 3))
         plt.subplot(131)
                                                   # make the first plot active
         plt.plot(my_x, y_lin, label = 'linear'); # do the actual plotting
         plt.xlabel('x', fontsize = 18);
                                                   # and while the plot is active
         plt.ylabel('x', fontsize = 18);
         plt.title('Linear', fontsize = 20,);
         plt.subplot(132)
         plt.plot(my_x, y_quad, label = 'quadratic');
         plt.xlabel('x', fontsize = 18);
         plt.ylabel('x squared', fontsize = 18);
         plt.title('Quadradic', fontsize = 20,);
         plt.subplot(133)
         plt.plot(my_x, y_cube, label = 'cubic');
         plt.xlabel('x', fontsize = 18);
         plt.ylabel('x cubed', fontsize = 18);
         plt.title('Cubic', fontsize = 20,);
         plt.tight_layout()
```

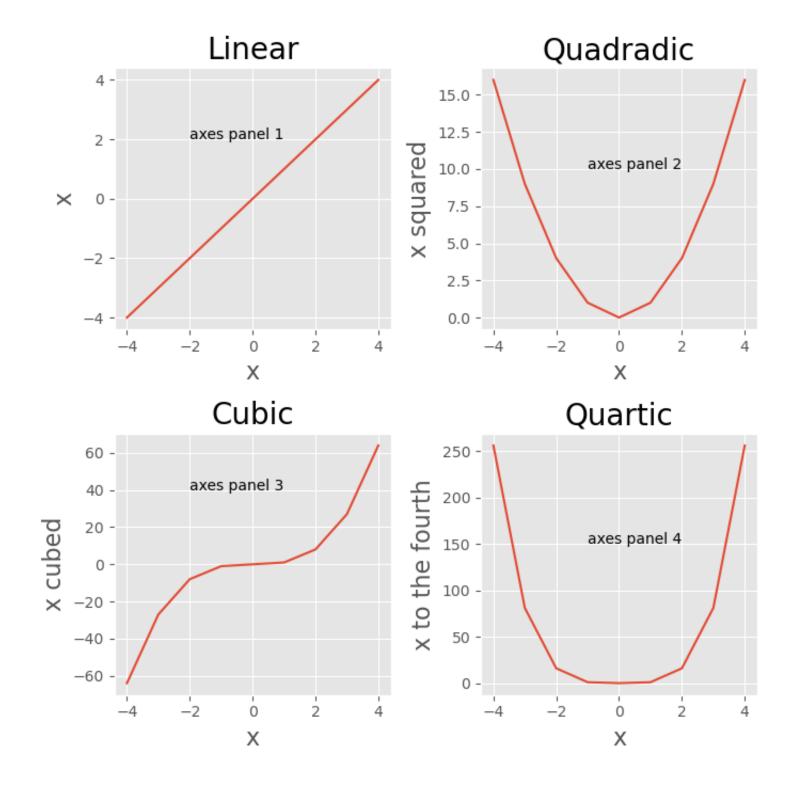


Sweet!

Finally, the subplots – the axes panels – are numbered "row wise", i.e. left-to-right, top to bottom. So if we wanted to write a self-contained piece of code plot four polynomials in a 2x2 grid, it would look like this:

```
In [58]: # compute our polynomials
        my_x = np.array([-4, -3, -2, -1, 0, 1, 2, 3, 4])
         y_lin = my_x
         y_quad = my_x**2 # ** is Python for exponentiation
         y_cube = my_x**3
         y_quart = my_x**4
         # set our style
         plt.style.use('ggplot')
         axisLabelSize = 16
         titleSize = 20
         # make a big square figure
         plt.figure(figsize=(7, 7))
        # Now make each of the plots!
         # first is top left
         plt.subplot(221)
                                                       \# '2' = 2 axes panel v; '2' = 2 axes panl x; '1' = 2
         plt.plot(my_x, y_lin, label = 'linear');
         plt.xlabel('x', fontsize = axisLabelSize);
         plt.ylabel('x', fontsize = axisLabelSize);
         plt.title('Linear', fontsize = titleSize);
         plt.text(-2, 2, 'axes panel 1')
         # then top right
         plt.subplot(222)
         plt.plot(my_x, y_quad, label = 'quadratic');
         plt.xlabel('x', fontsize = axisLabelSize);
         plt.ylabel('x squared', fontsize = axisLabelSize);
         plt.title('Quadradic', fontsize = titleSize,);
         plt.text(-1, 10, 'axes panel 2')
```

```
# TN1ra 15 DOTTOM LETT
plt.subplot(223)
plt.plot(my_x, y_cube, label = 'cubic');
plt.xlabel('x', fontsize = axisLabelSize);
plt.ylabel('x cubed', fontsize = axisLabelSize);
plt.title('Cubic', fontsize = titleSize,);
plt.text(-2, 40, 'axes panel 3')
# then bottom right
plt.subplot(224)
plt.plot(my_x, y_quart, label = 'quartic');
plt.xlabel('x', fontsize = axisLabelSize);
plt.ylabel('x to the fourth', fontsize = axisLabelSize);
plt.title('Quartic', fontsize = titleSize,);
plt.text(-1, 150, 'axes panel 4')
# finally, make everything automagically fit
plt.tight_layout()
```



Note that we snuck in an additional useful goodie, which is the plt.text() function. It allows you to add text (in axis-axis coordinates) to your figure!

So that does it for our overview of plotting the pyplot way! It may seem like a lot right now, but it's really not. Here's a handy summary.

Summary of pyplot plotting functions

Cast, in order of appearance (all are prefixed with plt. assuming you have imported matplotlib.pyplot as plt)

function	use
plot()	make a plot
xlabel()	add x label (can set font size)
ylabel()	add y label (can set font size)
title()	add a title (can set font size)
legend()	add a legend (can set position)
style.use()	use a plot style
style.available	list available styles
subplot()	make a plot matrix and set active plot
figure()	make a figure (can specify size)
tight_layout()	try to make everything fit neatly
text()	add text in x,y plotting coordinates

Now we shall briefly look at plotting in "object oriented" style.

Complete the following exercise.

```
x = np.random.randn(100,1)
y_A = x**2
y_B = y_A**2
y_C = y_B**2
```

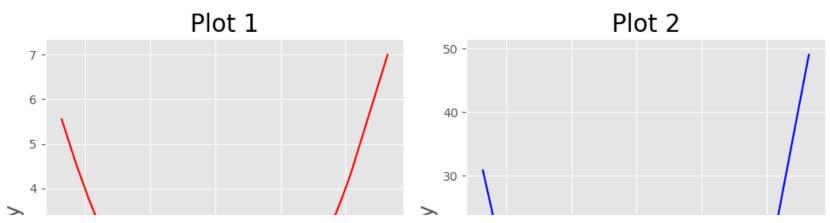
- given the data above:
 - Make 3 plots, one per y values
 - use subplot to plot the 3 plots in 2x2 grid of plots
 - set the style to one of the seaborn styles
 - set the color of each line to a different color say red, blue and green
 - set axis labels, and plot titles

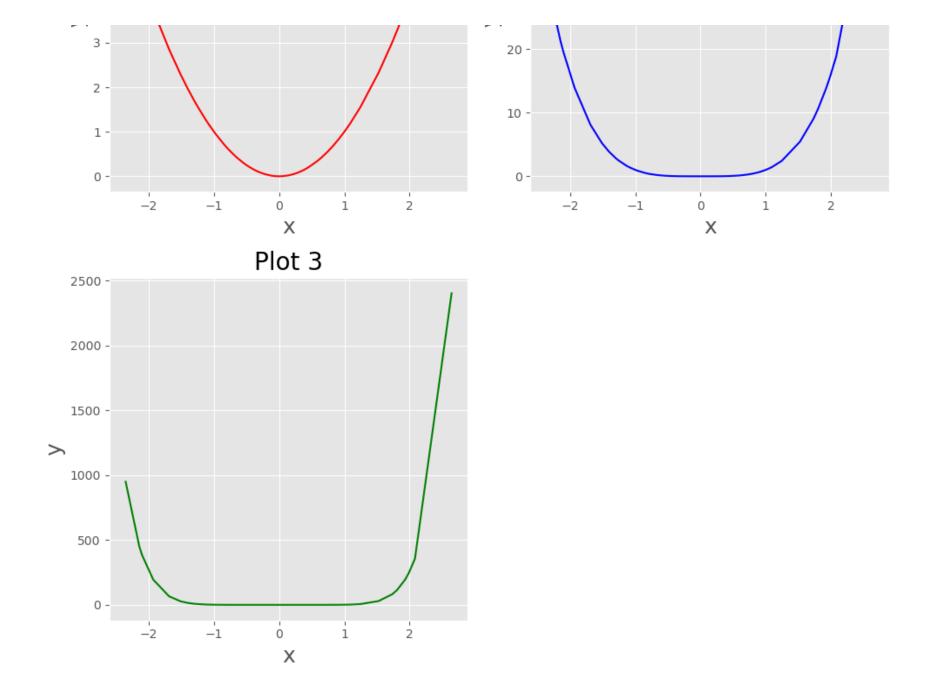
Note that most of what asked to do here is in the previous cells. The only more tricky thing will be to figure out how to find out which styles are available, remeber sometimes you can use plt.<TAB>, sometimes plt.style.<TAB> and see which options are available

[Use the code cell below to implement the plot]

In [76]:

```
x = np.random.randn(100,1)
x = np.sort(x, axis = 0)
y_A = x**2
y_B = y_A**2
y_C = y_B**2
plt.style.use('seaborn-bright')
plt.figure(figsize = (10,10))
plt.subplot(221)
plt.plot(x, y_A, color = 'r');
plt.xlabel('x', fontsize = 18);
plt.ylabel('y', fontsize = 18);
plt.title('Plot 1', fontsize = 20,);
plt.subplot(222)
plt.plot(x, y_B, color = 'b');
plt.xlabel('x', fontsize = 18);
plt.ylabel('y', fontsize = 18);
plt.title('Plot 2', fontsize = 20,);
plt.subplot(223)
plt.plot(x, y_C, color = 'g');
plt.xlabel('x', fontsize = 18);
plt.ylabel('y', fontsize = 18);
plt.title('Plot 3', fontsize = 20,);
plt.tight_layout()
```





Plotting using object oriented (OO) methods

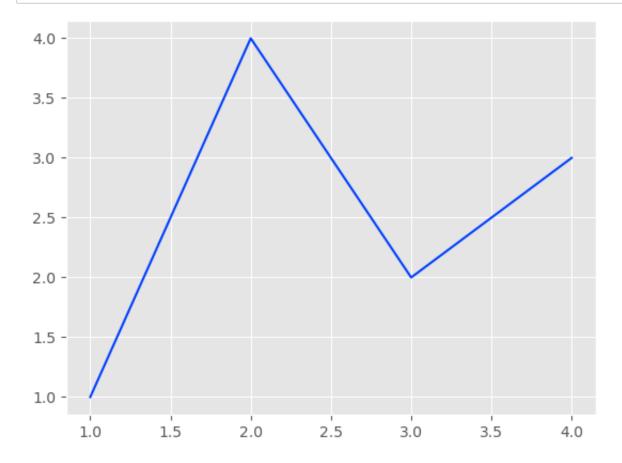
Now, let's go back to our OO plotting method. The difference between plotting using pyplot and using the object oriented approach is that, with the latter, we use the ability of two kinds of objects, **figures** and **axes**, to do stuff. The basic steps are:

- Make a figure and one or more axes with plt.subplots() (note it's plural!!)
- use the ability of an axes to do stuff like plotting
- (sometimes) use the figure's ability to do stuff

A simple object oriented plot

As usual, a simple example will make this more clear. Here's one lifted straight from the matplotlib documentation:

In [60]: fig, ax = plt.subplots() # Create a figure containing a single axes.
ax.plot([1, 2, 3, 4], [1, 4, 2, 3]); # Plot some data on the axes.



The call to plt.subplots() made a figure with a single axes panel in it. It returned a "handle" to the figure named fig, and a handle to the axes panel named ax.

We then used the axes panel's name, ax, to make the axes panel plot data in itself.

Note: we could have named these anything we wanted – Velma & Daphnie, Shaggy & Scooby, Bugs & Daffy, whatever – but Python / Data Science convention is to name the figure "fig" and the axes "ax". Booorrriiing, but handy for writing code that others, including future you, can easily read.

Adding stuff

```
In [61]: x = np.linspace(0, 2, 100) # Make new x values — again!
```

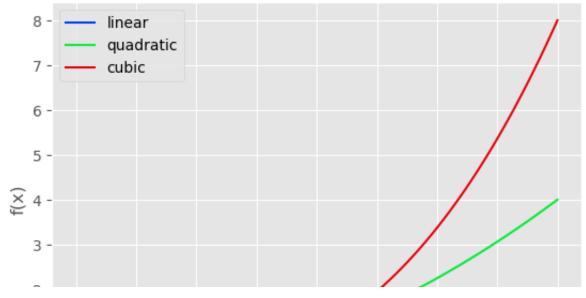
Create the figure and the axes panel. Note that even in the OO-style, we use <code>.pyplot.figure</code> to create these!

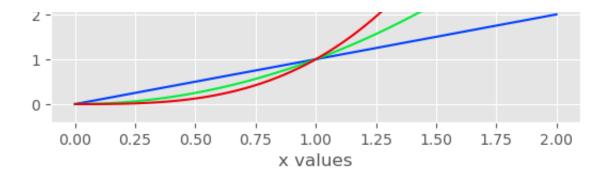
```
In [62]: # make the figure, fig, and the axes panel, ax
fig, ax = plt.subplots()

# now use ax's skills to plot stuff
ax.plot(x, x, label='linear') # Plot a line
ax.plot(x, x**2, label='quadratic') # and a parabola
ax.plot(x, x**3, label='cubic') # and a cubic curvy thingie

# and now use ax's skills add stuff
ax.set_xlabel('x values') # Add an x-label
ax.set_ylabel('f(x)') # and a y-label
ax.set_title("Plot made 00 style!") # and a title
ax.legend(); # and, finally, a legend
```

Plot made OO style!





Notice that, in terms of process, the only really differences are:

- we call subplots() (instead of subplot()), which makes
 - a figure named "fig"
 - an axes panel named "ax"
- we plot with ax.plot() instead of plt.plot()
- we use ax.set_... to do xlabel, ylabel, and title, and
- ax.legend() to make the legend

Happily, all of the ax. functions work just like their plt. counterparts.

OO Figures with multiple axis panels

So if the OO method of making figures is pretty much the same as the pyplot method of making figures (except for some annoying differences in terminology), what's the point in having both? Why would you use one vs. the other? The short answer is that:

- pyplot is easier you can make simple figures with less code
- the OO method is more powerful and flexible for working with complicated figures

Let's return to our four panel figure example above for a quick illustration of this. We'll break up the code so it's a bit more digestable.

Compute the polynomials.

```
In [63]: # compute our polynomials
    my_x = np.array([-4, -3, -2, -1, 0, 1, 2, 3, 4])
    y_lin = my_x
    y_quad = my_x**2
    y_cube = my_x**3
    y_quart = my_x**4
```

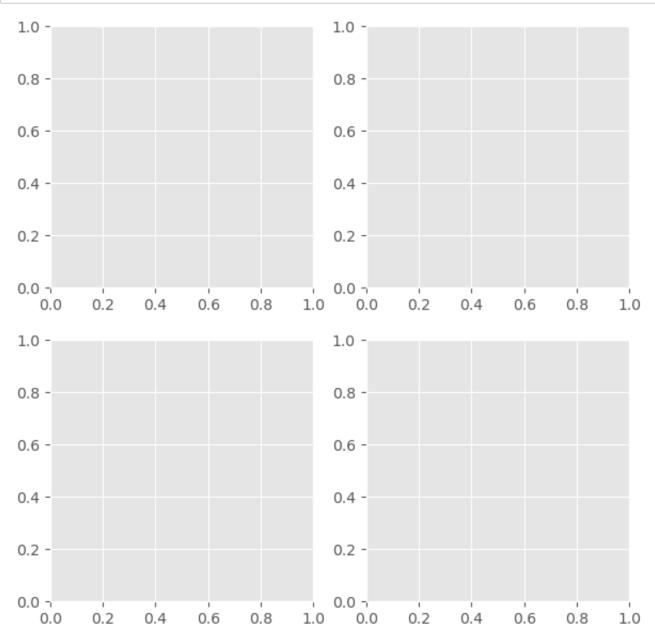
Next, we'll make sure we're stylin'

```
In [64]: # set our style
    plt.style.use('ggplot')
    axisLabelSize = 16
    titleSize = 20
```

Now, here is where the difference begins! We will call plt.subplots() and specify the how many rows and columns of axes panels we want. We will still name our figure "fig".

But instead of getting back a single handle to an axes panel that we would name "ax", we are going to get back an array of handles that, by convention, we'll name "axs".

In [65]: # make a big square figure
fig, axs = plt.subplots(ncols=2, nrows=2, figsize=(7, 7))



So what is "axs"? It is an array, like a numpy array, but contains the "handles" to each of our plots. Like this:

[[upper left plot, upper right plot],

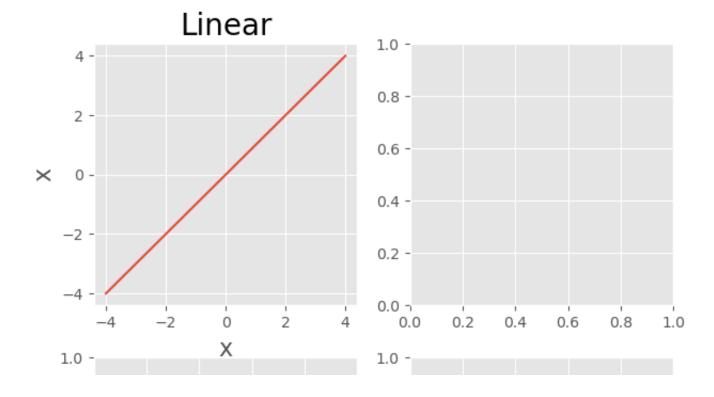
[lower left plot, lower right plot]]

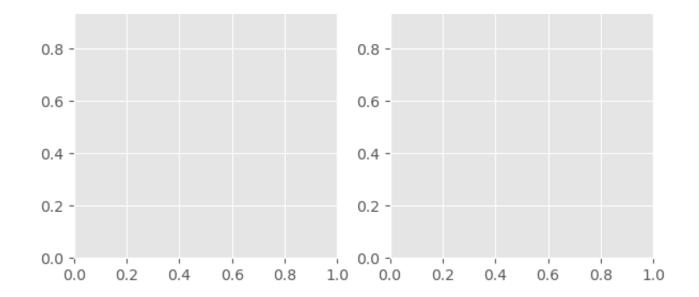
So instead of saying ax.plot() like we did above, we would say axs[0,0].plot() to plot in the upper left axes panel, axs[0,1].plot() to plot in the upper right axes panel, etc.

So let's make the first subplot! (We have to re-make the figure or there will be nowhere for us to plot.)

```
In [66]: fig, axs = plt.subplots(ncols=2, nrows=2, figsize=(7, 7))

axs[0,0].plot(my_x, y_lin, label = 'linear');
axs[0,0].set_xlabel('x', fontsize = axisLabelSize);
axs[0,0].set_ylabel('x', fontsize = axisLabelSize);
axs[0,0].set_title('Linear', fontsize = titleSize);
```





If you think about it, this actually more clear and straightforward than plt.subplot(221), etc. axs[0,0] makes it immediately clear that your are plotting in the [first row, first column] of your figure.

Now let's make the full figure:

```
In [67]:
```

```
# make the figure
fig, axs = plt.subplots(ncols=2, nrows=2, figsize=(7, 7))
# Now make each of the plots!
# first is top left
# plt.subplot(221) we don't neet this anymore
axs[0,0].plot(my_x, y_lin, label = 'linear');
axs[0,0].set xlabel('x', fontsize = axisLabelSize);
axs[0,0].set_ylabel('x', fontsize = axisLabelSize);
axs[0,0].set title('Linear', fontsize = titleSize);
# then top right
axs[0,1].plot(my_x, y_quad, label = 'quadratic');
axs[0,1].set xlabel('x', fontsize = axisLabelSize);
axs[0,1].set ylabel('x squared', fontsize = axisLabelSize);
axs[0,1].set title('Quadradic', fontsize = titleSize,);
# third is bottom left
axs[1,0].plot(my_x, y_cube, label = 'cubic');
axs[1,0].set_xlabel('x', fontsize = axisLabelSize);
axs[1,0].set ylabel('x cubed', fontsize = axisLabelSize);
axs[1,0].set title('Cubic', fontsize = titleSize,);
# then bottom right
axs[1,1].plot(my_x, y_quart, label = 'quartic');
axs[1,1].set_xlabel('x', fontsize = axisLabelSize);
axs[1,1].set ylabel('x to the fourth', fontsize = axisLabelSize);
axs[1,1].set title('Quartic', fontsize = titleSize,);
# finally, make everything automagically fit
fig.tight layout() # notice that now it's fig. instead of plt.
```

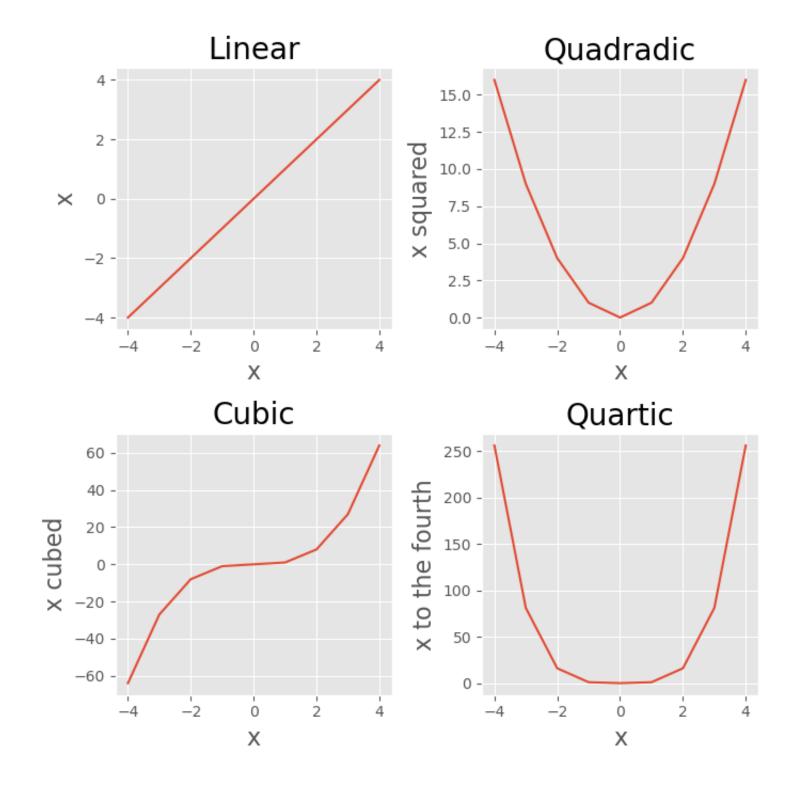


Figure-level titles and labels

And now we can use a little more of the power of the OO method to make the figure a bit more appealing. Notice that

- the x axis labels are redundant, one overall label would be better
- a global y axis label, like "f(x)" would be cool
- the figure could use an overall title, like "Polynomials!" or something.

We can do all these things because our figure named "fig" knows how to do stuff to! We'll use

- fig.suptitle() to make a "supra title"
- fig.supylabel() to make a "supra y label", and
- fig.supxlabel() to make a "supra x label", and

We'll do this all in one cell of code, but the only thing that's different from above is the little chunk at line 20 where we do the figure level stuff – the "supra stuff"!

```
In [68]: # compute our polynomials
    my_x = np.array([-4, -3, -2, -1, 0, 1, 2, 3, 4])
    y_lin = my_x
    y_quad = my_x**2  # ** is Python for exponentiation
    y_cube = my_x**3
    y_quart = my_x**4

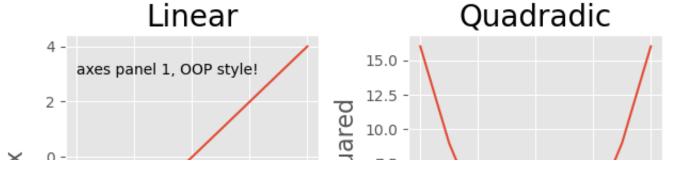
# set our style
    plt.style.use('ggplot')
    axisLabelSize = 16
    titleSize = 20

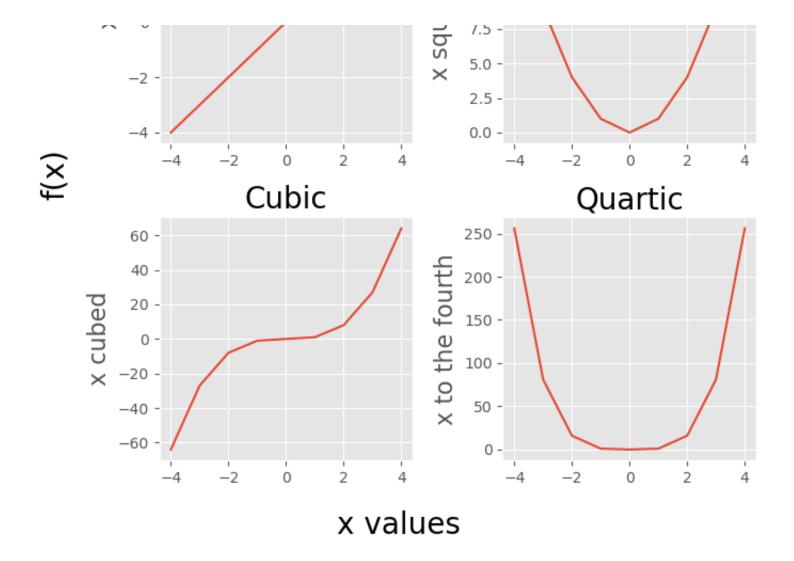
# make a big square figure
    fig, axs = plt.subplots(ncols=2, nrows=2, figsize=(7, 7))

# set the "supra stuff" for the figure
    fig.suptitle('Polynomials!', ha='left', fontsize=24)
    fig.supylabel('f(x)', fontsize=20)
```

```
fig.supxlabel('x values', fontsize=20)
# Now make each of the plots!
# first is top left
# plt.subplot(221) we don't neet this anymore
axs[0,0].plot(my x, y lin, label = 'linear');
axs[0,0].set ylabel('x', fontsize = axisLabelSize);
axs[0,0].set title('Linear', fontsize = titleSize);
axs[0,0].text(-4, 3, 'axes panel 1, 00P style!')
# then top right
axs[0,1].plot(my_x, y_quad, label = 'quadratic');
axs[0,1].set_ylabel('x squared', fontsize = axisLabelSize);
axs[0,1].set title('Quadradic', fontsize = titleSize,);
# third is bottom left
axs[1,0].plot(my_x, y_cube, label = 'cubic');
axs[1,0].set_ylabel('x cubed', fontsize = axisLabelSize);
axs[1,0].set_title('Cubic', fontsize = titleSize,);
# then bottom right
axs[1,1].plot(my_x, y_quart, label = 'quartic');
axs[1,1].set ylabel('x to the fourth', fontsize = axisLabelSize);
axs[1,1].set title('Quartic', fontsize = titleSize,);
# finally, make everything automagically fit
fig.tight_layout()
```

Polynomials!





And there we go! We can do all the same stuff using the OO approach to making plots, but allows us to

- address the axes panels in a more transparent way axs[row, col] vs. subplot(rcn)
- put global labels and titles on our figures

Which method do I choose?

It's largely up to you. A good rule of thumb is

- use pyplot for simple, one panel figures
- use the OO method for multi-panel figures

That having been said, some people feel that "If I have to use the OO method for some figures, why not just use it for every figure?"

It's your call!

Complete the following exercise.

Describe in your own words the two methods by which matplot lib works

[Although both methods are slightly different in terms of coding, they both can offer the same reuslts. However, I think that plt.plot is easier when making a quicl/simple plot. Meanwhile, I think that axs.plot is easier when arranging different grids.]

• Describe the difference between the 00 and plt approach to using matplotlib

[Although they both functioning the same on rearranging grids, the differences are 00 is for axes.plt and plt is for plt.plot options.]

• Make a new plot using your method of choice. The plot should plot at least 30 data points in x transofrmed using 3 of the math operators introduced in the previous tutorials. Label the axis, the figure, the colors of the lines, add titles to the figure and a legend to the plotted data. Also customize the style of the plot and font size of all text.

[Use the markdown cell below to implement your code]

```
x = np.linspace(0,3,50)
y_1 = x**2
y_2 = x**3
y_3 = x**4

plt.style.use('seaborn-white')
plt.plot(x,y_1, ':r', label = 'Y1');
plt.plot(x, y_2, '-b', label = 'Y2');
plt.plot(x, y_3, '-.g', label = "Y3");

plt.title("My Plot", fontsize = 20)
plt.xlabel('x', fontsize = 18)
plt.ylabel('y', fontsize = 18)
plt.legend()
```

Out[96]: <matplotlib.legend.Legend at 0x7fc787730700>

