

# **ASTR 119: Session 6**

## **Matplotlib, Multipanel plots; Line fitting**

# Outline

- 1) Homework due 10/22 at 8:00am
- 2) Attend your assigned lecture, and some comments about Slack
- 3) Visualization of the Day
- 4) Matplotlib basics
- 5) Our first figure, saving figures to a file
- 6) Multipanel figures, aspect ratios, legends
- 7) Line fitting
- 8) Save your work to GitHub

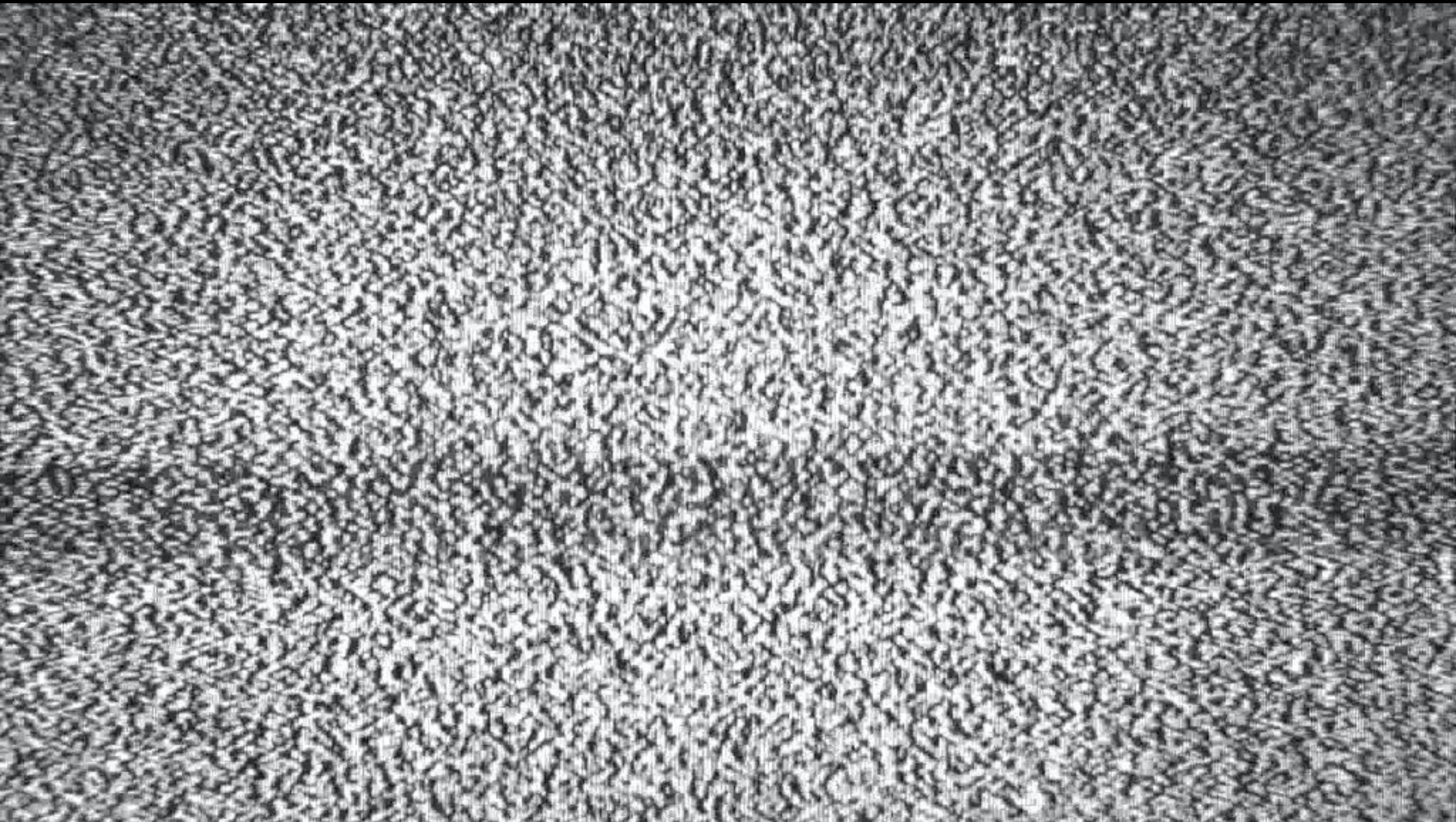
# Homework, due Oct 22, 8:00am

- 1) Make a Jupyter notebook, import numpy and matplotlib, and in a single line use numpy to create an array  $x$  running  $x = [0, 2 * \pi]$  inclusive with 1000 values. Then use matplotlib to plot the following functions on a single plot, using the  $x$  range  $x = [0, 2 * \pi]$  and the vertical range  $y = [-1, 10]$ :
  - a).  $y(x) = 5.5 \cos(2 * x) + 5.5$
  - b).  $y(x) = 0.02 * \exp(x)$
  - c).  $y(x) = 0.25 * x^2 + 0.1 \sin(10 * x)$
- 2) Make the plot's  $y$  label "Measures of Awesomeness" and the  $x$  label "Time in ASTR / EART 119".
- 3) Create an issue for your repository and tag your TAs. CLEAR ALL THE CELLS BEFORE YOU COMMIT THE NOTEBOOK.
- 4) Your TA will clone your code and email you commented version of the code and a grade. To get the full grade possible, all the notebooks will need to run to completion without errors and produce the requested plots.
- 5) Call the repository "astr-119-hw-3" and the notebook "hw-3.ipynb".

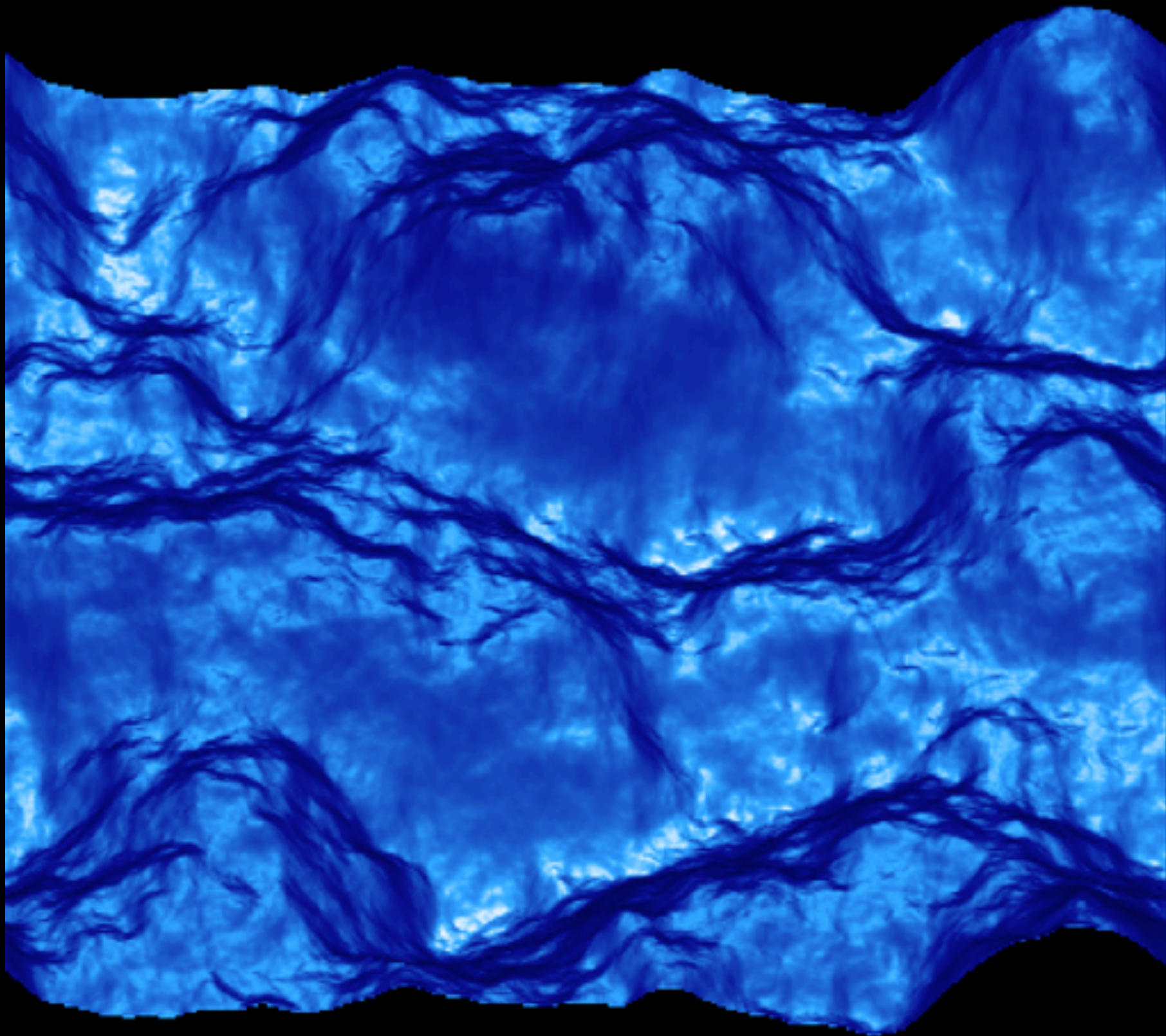
# Lecture, Slack, and Help

- 1) You may only attend the lecture for which you are registered.
- 2) Technical and homework help over Slack will be limited. Please come to office hours and attend your section for better support.
- 3) When asking for help, please provide as much detail as possible about the issue you encounter. Is there an error message? Did you do anything “custom” when configuring your software?









# matplotlib.pyplot



Fork me on GitHub

[home](#) | [examples](#) | [tutorials](#) | [API](#) | [docs](#) » [The Matplotlib API](#) » [Pyplot function overview](#) »

[previous](#) | [next](#) | [modules](#) | [index](#)

## matplotlib.pyplot

`matplotlib.pyplot` is a state-based interface to matplotlib. It provides a MATLAB-like way of plotting.

pyplot is mainly intended for interactive plots and simple cases of programmatic plot generation:

```
import numpy as np
import matplotlib.pyplot as plt

x = np.arange(0, 5, 0.1)
y = np.sin(x)
plt.plot(x, y)
```

The object-oriented API is recommended for more complex plots.

## Functions

<code>acorr(x, *[, data])</code>	Plot the autocorrelation of <code>x</code> .
<code>angle_spectrum(x[, Fs, Fo, window, pad_to, ...])</code>	Plot the angle spectrum.
<code>annotate(text, xy, *args, **kwargs)</code>	Annotate the point <code>xy</code> with text <code>s</code> .
<code>arrow(x, y, dx, dy, **kwargs)</code>	Add an arrow to the axes.
<code>autoscale([enable, axis, tight])</code>	Autoscale the axis view to the data (toggle).
<code>axhline(y, *[, data])</code>	Plot the horizontal line at <code>y</code> .

Quick search

Go

Table Of Contents

`matplotlib.pyplot`

- [Functions](#)

Related Topics

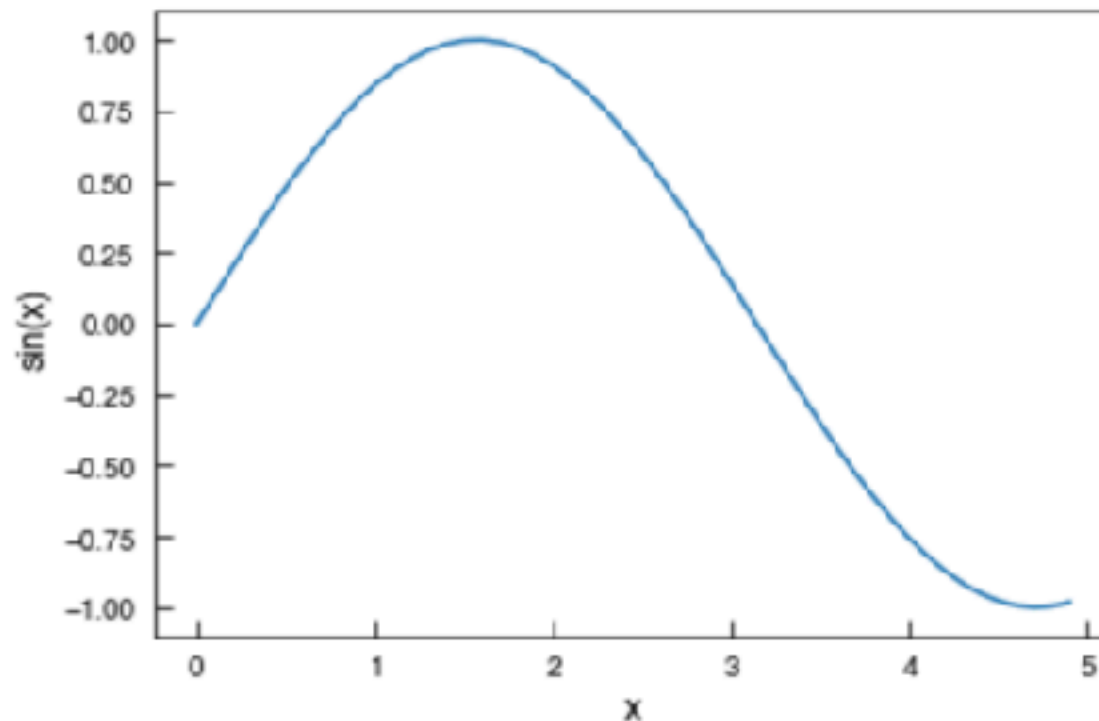
Documentation overview

- [The Matplotlib API](#)
  - [Pyplot function overview](#)
    - [Previous: Pyplot function overview](#)
    - [Next: matplotlib.pyplot.acorr](#)

[Show Page Source](#)

```
In [1]: %matplotlib inline
import numpy as np
import matplotlib.pyplot as plt
```

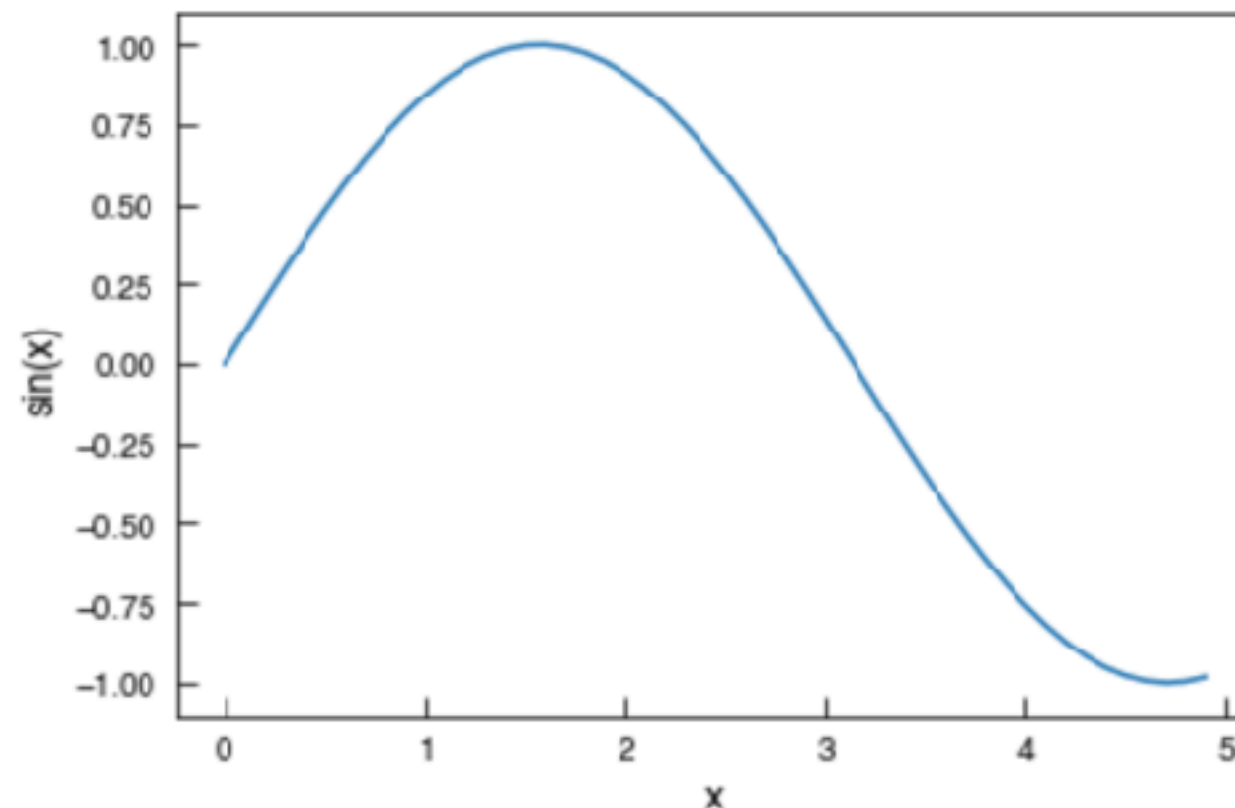
```
In [3]: x = np.arange(0, 5, 0.1) #create x=[0..5] in 0.1 increments
y = np.sin(x) # y = sin(x)
plt.plot(x,y) # make a plot
plt.xlabel('x') # label x axis
plt.ylabel('sin(x)') # label y axis
plt.show() # show the plot on the screen
```



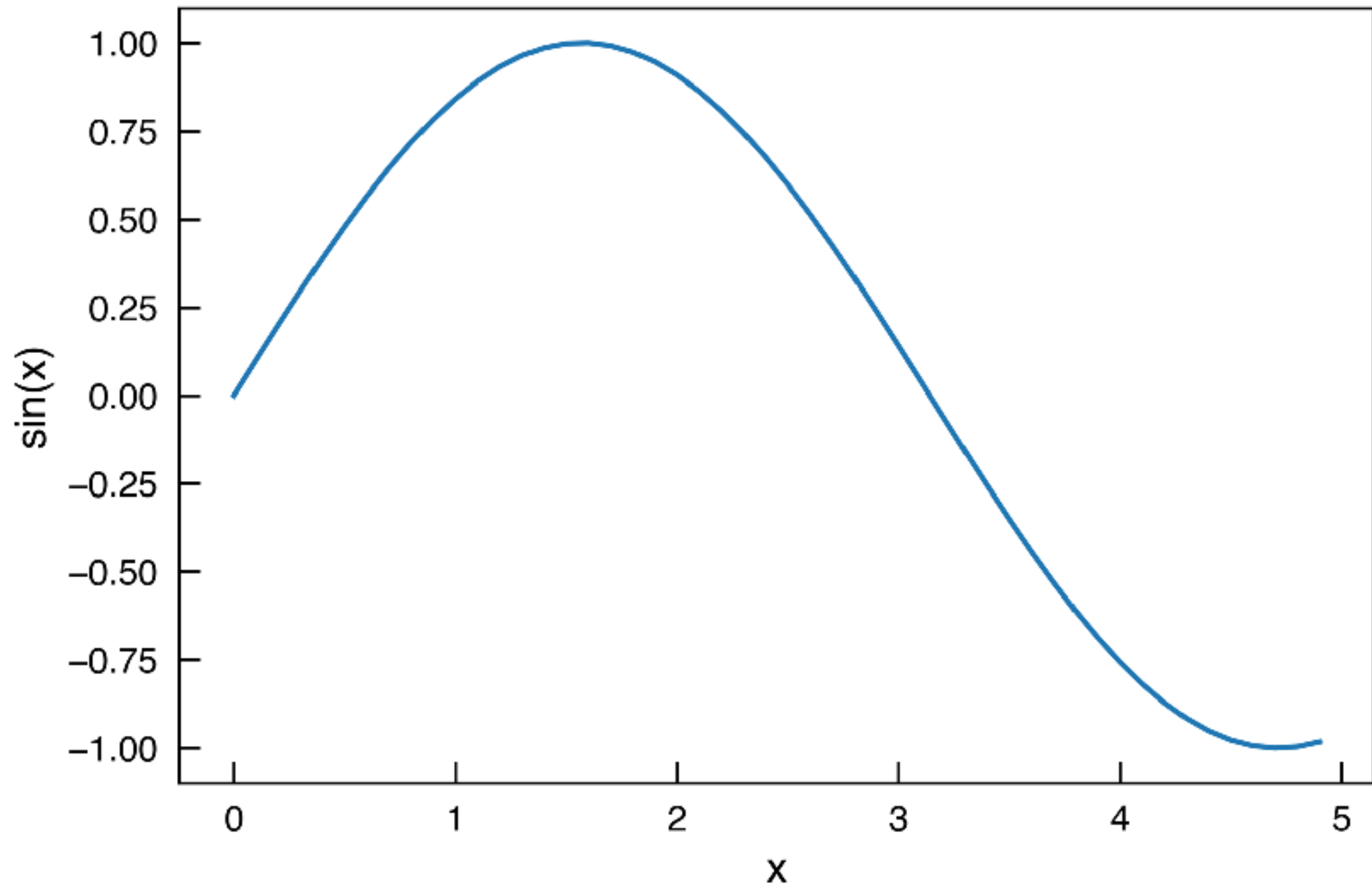
Matplotlib.pyplot



```
In [5]: x = np.arange(0, 5, 0.1) #create x=[0..5] in 0.1 increments
        y = np.sin(x)           # y = sin(x)
        plt.plot(x,y)           # make a plot
        plt.xlabel('x')         # label x axis
        plt.ylabel('sin(x)')    # label y axis
        plt.savefig('sinx.png',bbox_inches="tight",dpi=600)
```



plt.savefig()



`plt.savefig()`

# Multipanel Figures

Using Matplotlib, we can make multi panel figures easily.

## Making multipanel plots with matplotlib

First, we import numpy and matplotlib as usual

```
In [2]: %matplotlib inline
import numpy as np
import matplotlib.pyplot as plt
```

Then we define an array of angles, and their sines and cosines using numpy. This time we will use linspace

```
In [30]: x = np.linspace(0, 2*np.pi, 100)
print(x[-1], 2*np.pi)

y = np.sin(x)
z = np.cos(x)
w = np.sin(4*x)
v = np.cos(4*x)

6.28318530718 6.283185307179586
```



# Multipanel Figures

Using Matplotlib, we can make multi panel figures easily.

Now, let's make a two panel plot side-by-side

```
In [ ]: #call subplots to generate a multipanel figure. This means 1 row, 2 columns of figures
f, axarr = plt.subplots(1, 2)

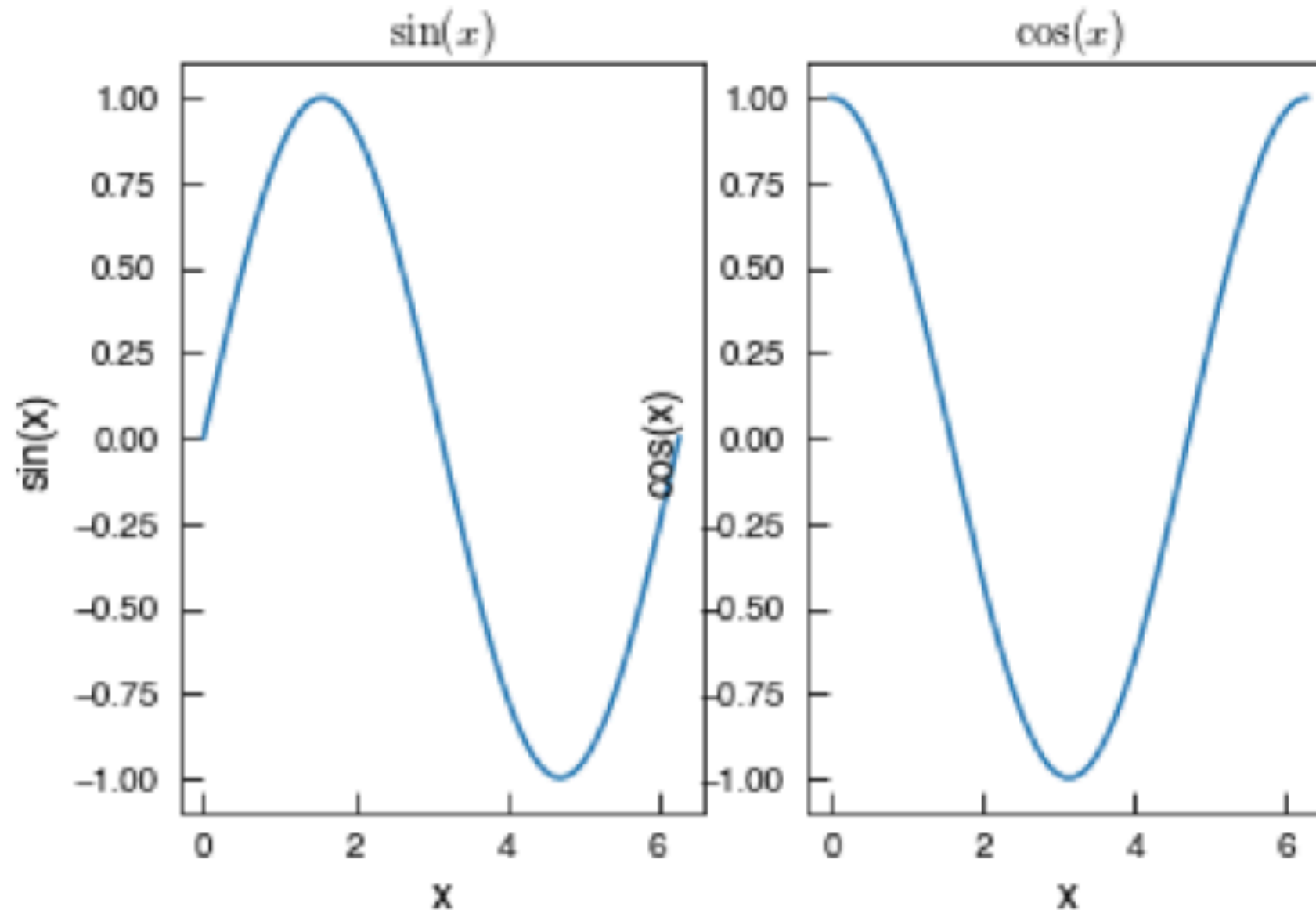
#treat axarr as an array, from left to right

#first panel
axarr[0].plot(x, y)
axarr[0].set_xlabel('x')
axarr[0].set_ylabel('sin(x)')
axarr[0].set_title(r'$\sin(x)$')
#second panel
axarr[1].plot(x, z)
axarr[1].set_xlabel('x')
axarr[1].set_ylabel('cos(x)')
axarr[1].set_title(r'$\cos(x)$')
```

# Multipanel Figures

Using Matplotlib, we can make multi panel figures easily.

```
Out[54]: Text(0.5, 1, '$\\cos(x)$')
```



# Multipanel Figures

`subplots_adjust()` enables us to move the panels further apart

Here we can see that matplotlib has the panels too close together.

We can adjust that using the `subplots_adjust()` function.

```
In [ ]: #call subplots to generate a multipanel figure. This means 1 row, 2 columns of figures
f, axarr = plt.subplots(1, 2)

#treat axarr as an array, from left to right

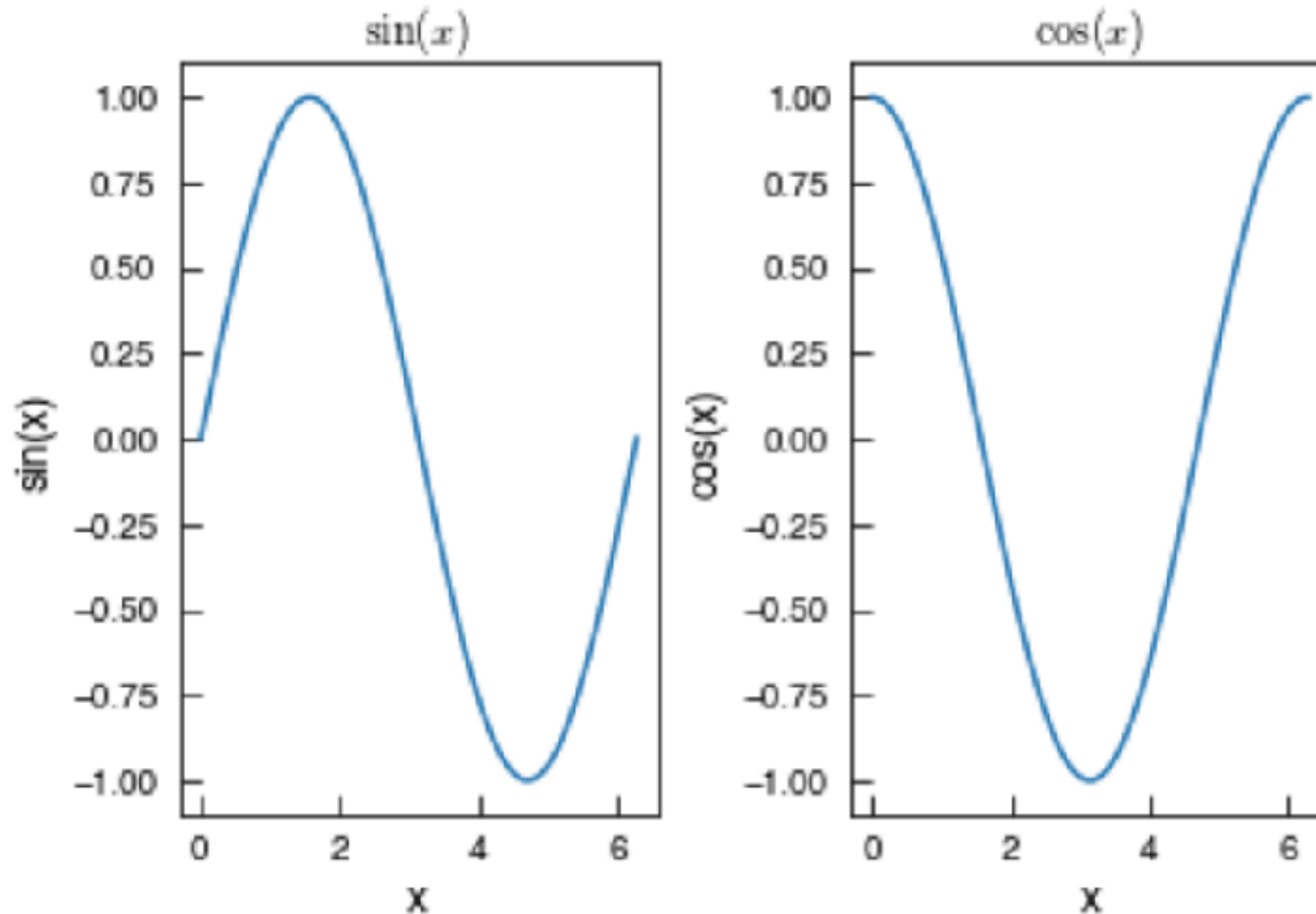
#first panel
axarr[0].plot(x, y)
axarr[0].set_xlabel('x')
axarr[0].set_ylabel('sin(x)')
axarr[0].set_title(r'$\sin(x)$')
#second panel
axarr[1].plot(x, z)
axarr[1].set_xlabel('x')
axarr[1].set_ylabel('cos(x)')
axarr[1].set_title(r'$\cos(x)$')

#add more space between the figures
f.subplots_adjust(wspace=0.4)
```



# Multipanel Figures

`subplots_adjust()` enables us to move the panels further apart



# Multipanel Figures

We can use `set_aspect()` to change the axis ratio.

**OK, but the axis ratios are all squished. Let's fix that too.**

```
In [ ]: #call subplots to generate a multipanel figure. This means 1 row, 2 columns of figures
f, axarr = plt.subplots(1, 2)

#treat axarr as an array, from left to right

#first panel
axarr[0].plot(x, y)
axarr[0].set_xlabel('x')
axarr[0].set_ylabel('sin(x)')
axarr[0].set_title(r'$\sin(x)$')
#second panel
axarr[1].plot(x, z)
axarr[1].set_xlabel('x')
axarr[1].set_ylabel('cos(x)')
axarr[1].set_title(r'$\cos(x)$')

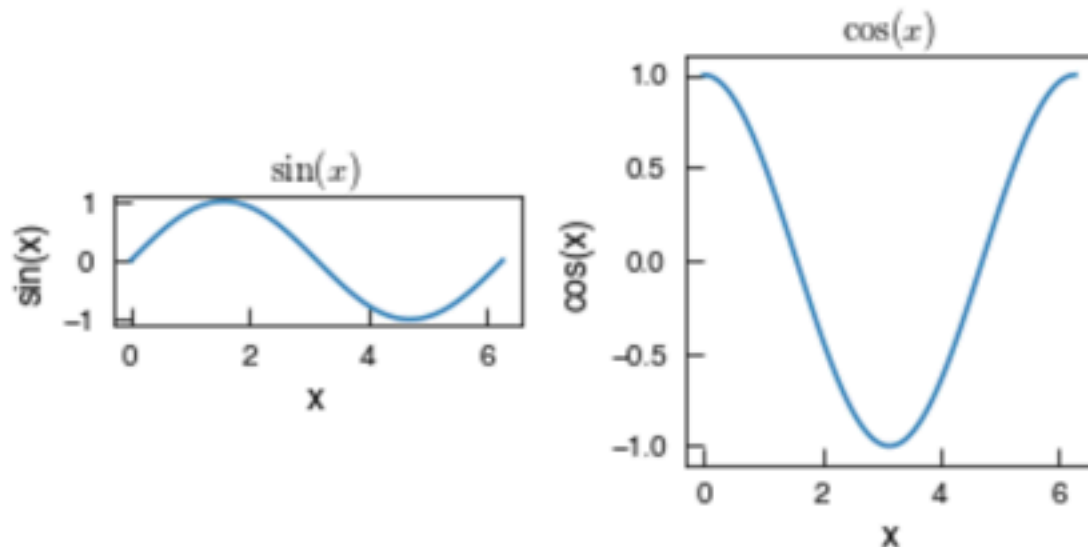
#add more space between the figures
f.subplots_adjust(wspace=0.4)

#fix the axis ratio
#here are two possible options
axarr[0].set_aspect('equal') #make the ratio of the tick units equal, a bit counter intuitive
axarr[1].set_aspect(np.pi) #make a square by setting the aspect to be the ratio of the tick unit range
```

# Multipanel Figures

We can use `set_aspect()` to change the axis ratio.

```
#fix the axis ratio
#here are two possible options
axarr[0].set_aspect('equal') #make the ratio of the tick units equal, a bit counter intuitive
axarr[1].set_aspect(np.pi)  #make a square by setting the aspect to be the ratio of the tick unit range
```





# Legends

Legends are an easy way to notate a complicated figure.

**Alright, let's keep the square figure, merge them into one, remove the titles and add legends**

```
In [ ]: #adjust the size of the figure
fig = plt.figure(figsize=(6,6))

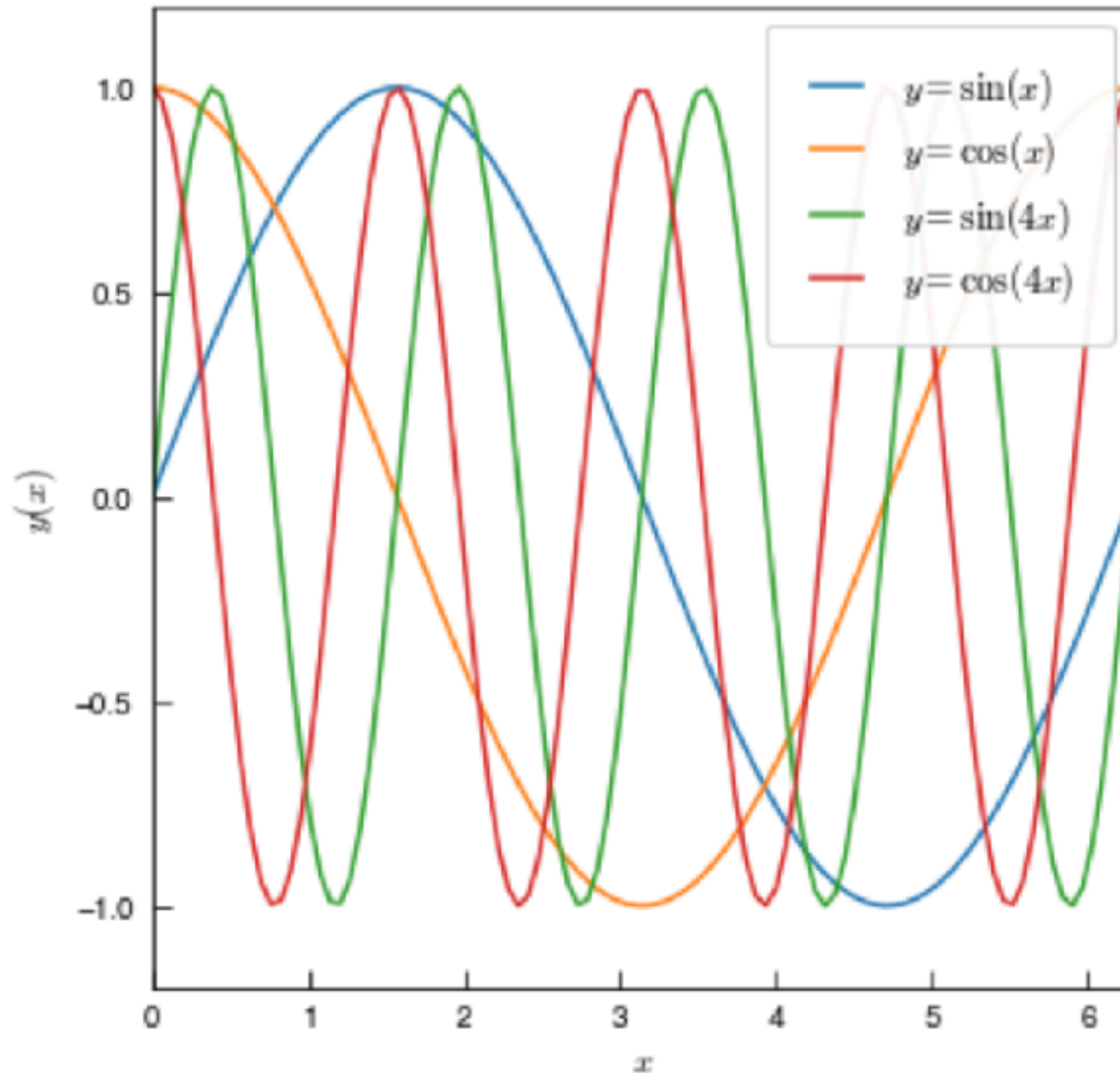
plt.plot(x, y, label=r'$y = \sin(x)$') #add a label to the line
plt.plot(x, z, label=r'$y = \cos(x)$') #add a label to the second line
plt.plot(x, w, label=r'$y = \sin(4x)$') #add a label to the third line
plt.plot(x, v, label=r'$y = \cos(4x)$') #add a label to the fourth line

plt.xlabel(r'$x$') #note set_xlabel vs. xlabel
plt.ylabel(r'$y(x)$') #note set_ylabel vs. ylabel
plt.xlim([0, 2*np.pi]) #note set_xlim vs. xlim
plt.ylim([-1.2, 1.2]) #note set_ylim vs. ylim
plt.legend(lcc=1, framealpha=0.95) #add a legend with a semi-transparent frame in the upper RH corner

#fix the axis ratio
plt.gca().set_aspect(np.pi/1.2) #use "gca" to get current axis()
```

# Legends

Legends are an easy way to notate a complicated figure.



# Line Fitting

Using python, we can quickly perform least squares line fitting

## Example of performing linear least squares fitting

First we import numpy and matplotlib as usual.

```
In [1]: %matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
```

Now, let's generate some random data about a trend line.

```
In [20]: #set a random number seed
np.random.seed(119)

#set number of data points
npoints = 50

#set x
x = np.linspace(0,10.,npoints)

#set slope, intercept, and scatter rms
m = 2.0
b = 1.0
sigma = 2.0

#generate y points
y = m*x + b + np.random.normal(scale=sigma,size=npoints)
y_err = np.full(npoints,sigma)
```

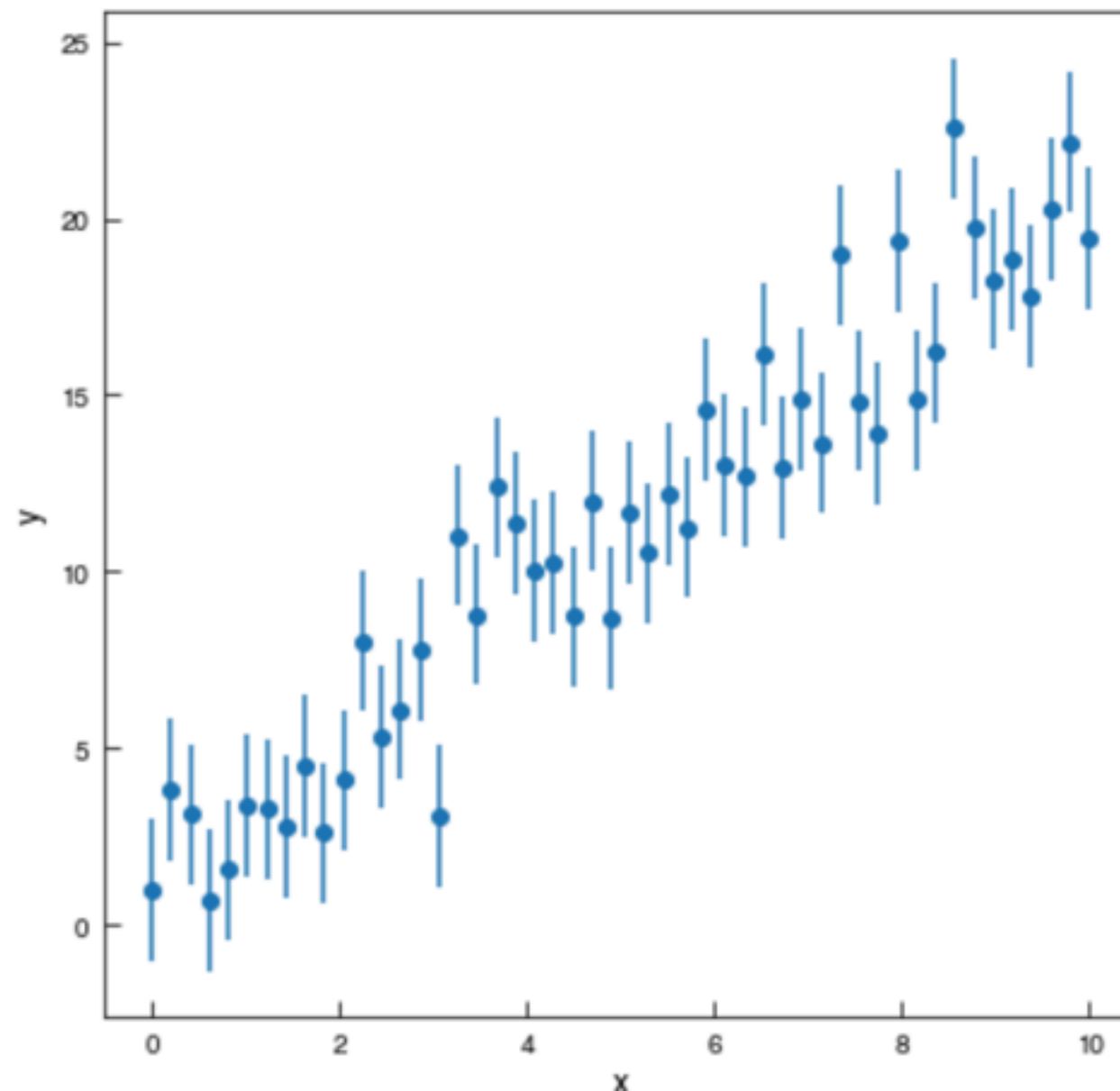


# Line Fitting

Let's just plot the data first

```
In [14]: f = plt.figure(figsize=(7,7))  
plt.errorbar(x,y,sigma,fmt='o')  
plt.xlabel('x')  
plt.ylabel('y')
```

```
Out[14]: Text(0,0.5,'y')
```



# Line Fitting

## Method #1, polyfit()

```
In [28]: m_fit, b_fit = np.polyld(np.polyfit(x, y, 1, w=1./y_err)) #weight with uncertainties
print(m_fit, b_fit)

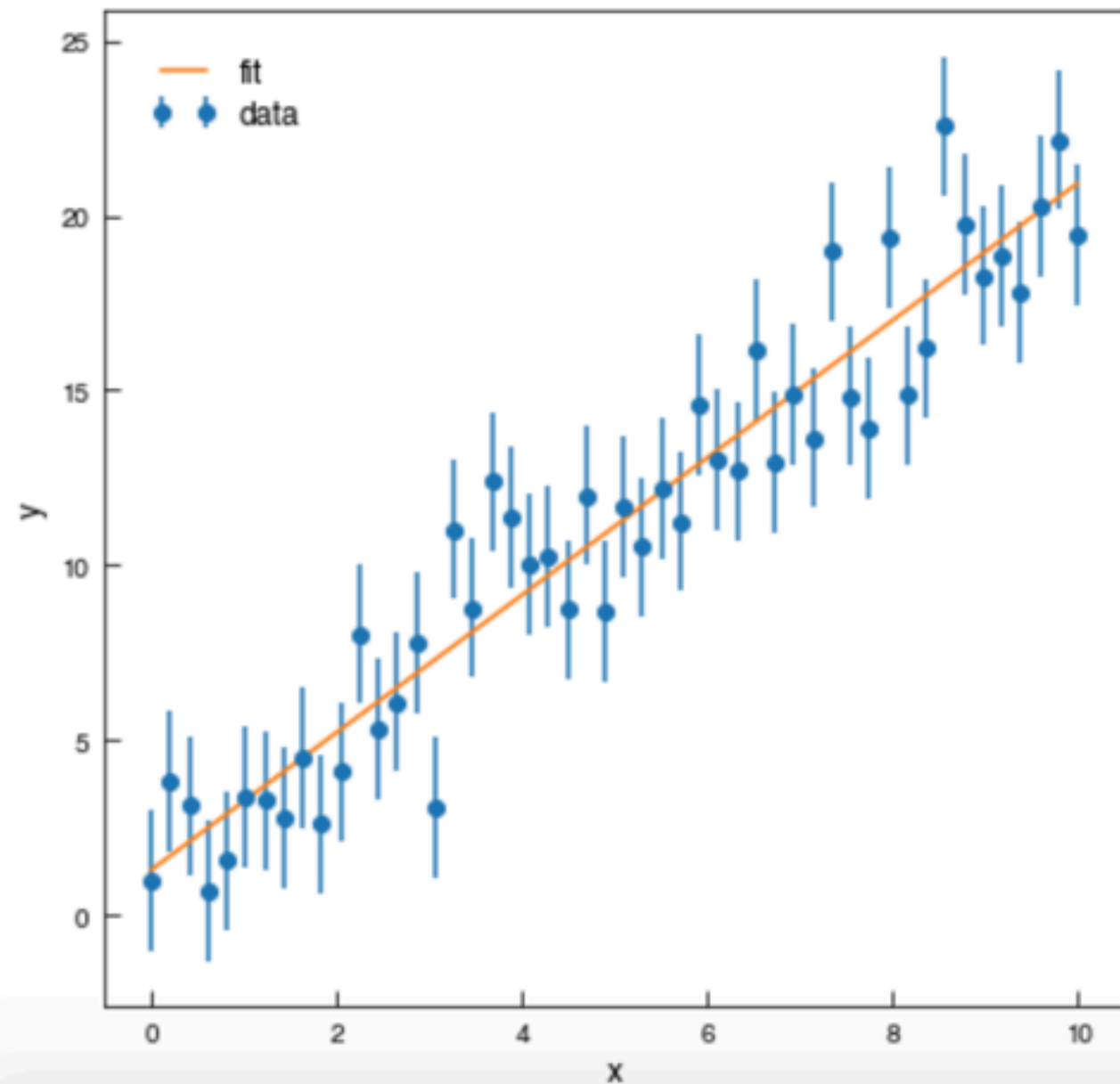
y_fit = m_fit * x + b_fit
1.96340434704 1.2830106813
```

# Line Fitting

## Plot result

```
In [29]: f = plt.figure(figsize=(7,7))  
plt.errorbar(x,y,yerr=y_err,fmt='o',label='data')  
plt.plot(x,y_fit,label='fit')  
plt.xlabel('x')  
plt.ylabel('y')  
plt.legend(loc=2,frameon=False)
```

Out[29]: <matplotlib.legend.Legend at 0x10fbbcef0>



# Line Fitting

## Method #2, scipy + optimize

```
In [31]: #import optimize from scipy
          from scipy import optimize

          #define the function to fit
          def f_line(x, m, b):
              return m*x + b

          #perform the fit
          params, params_cov = optimize.curve_fit(f_line,x,y,sigma=y_err)

          m_fit = params[0]
          b_fit = params[1]
          print(m_fit,b_fit)

1.96340434575 1.28301068905
```

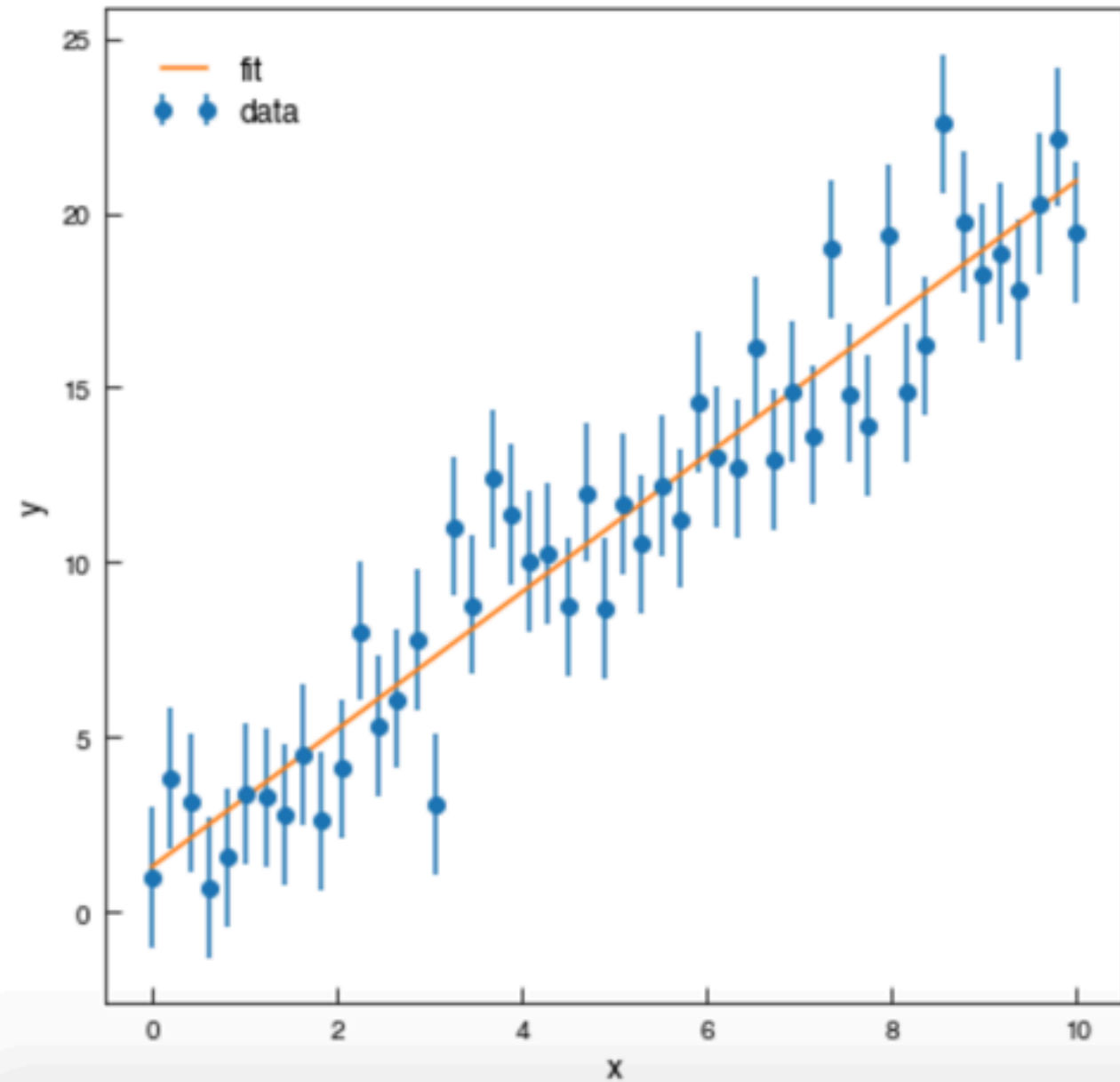


# Line Fitting

## Plot the result

```
In [33]: f = plt.figure(figsize=(7,7))  
plt.errorbar(x,y,yerr=y_err,fmt='o',label='data')  
plt.plot(x,y_fit,label='fit')  
plt.xlabel('x')  
plt.ylabel('y')  
plt.legend(loc=2,frameon=False)
```

Out[33]: <matplotlib.legend.Legend at 0x112928fd0>



# Using pip to install scipy

**Mac OSX / Linux:**

```
$ pip3 install scipy
```

**Windows:**

```
$ py -3 -m pip install scipy
```

# Line Fitting

## Method #2, scipy + optimize

```
In [31]: #import optimize from scipy
          from scipy import optimize

          #define the function to fit
          def f_line(x, m, b):
              return m*x + b

          #perform the fit
          params, params_cov = optimize.curve_fit(f_line,x,y,sigma=y_err)

          m_fit = params[0]
          b_fit = params[1]
          print(m_fit,b_fit)

1.96340434575 1.28301068905
```

# Line Fitting

**We can perform much more complicated fits....**

```
In [38]: #redefine x and y
npoints = 50
x = np.linspace(0., 2*np.pi, npoints)

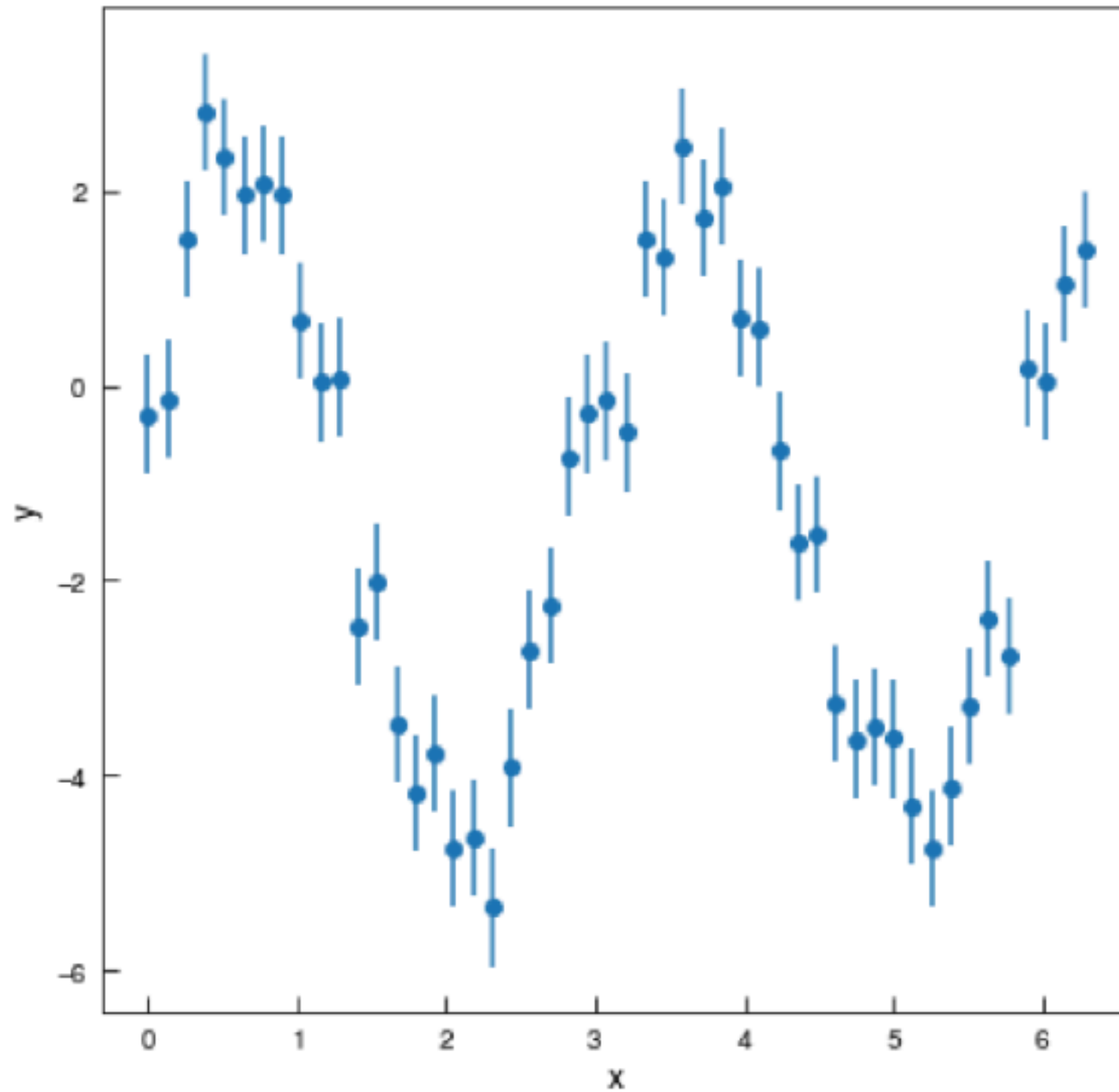
#make y a complicated function
a = 3.4
b = 2.1
c = 0.27
d = -1.3
sig = 0.6

y = a * np.sin( b*x + c ) + d + np.random.normal(scale=sig, size=npoints)
y_err = np.full(npoints, sig)

f = plt.figure(figsize=(7, 7))
plt.errorbar(x, y, yerr=y_err, fmt='o')
plt.xlabel('x')
plt.ylabel('y')
```

# Line Fitting

```
Out[38]: Text(0,0.5,'y')
```





# Line Fitting

## Perform a fit using `scipy.optimize.curve_fit()`

```
In [45]: #import optimize from scipy
          from scipy import optimize

          #define the function to fit
          def f_line(x, a, b, c, d):
              return a * np.sin( b*x + c) + d

          #perform the fit
          params, params_cov = optimize.curve_fit(f_line,x,y,sigma=y_err,p0=[1,2.,0.1,-0.1])

          a_fit = params[0]
          b_fit = params[1]
          c_fit = params[2]
          d_fit = params[3]

          print(a_fit,b_fit,c_fit,d_fit)

          y_fit = a_fit * np.sin(b_fit * x + c_fit) + d_fit

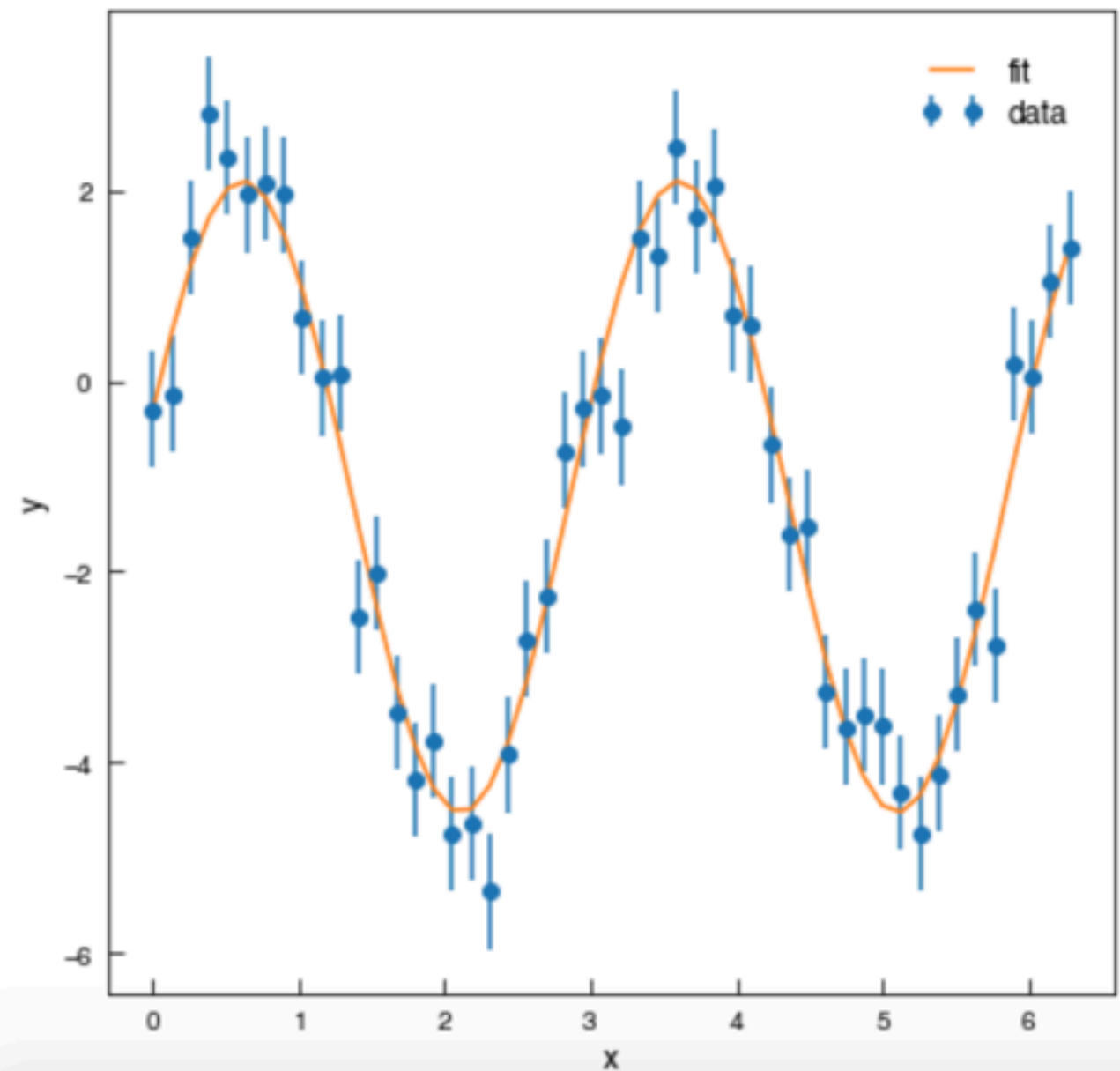
          3.31470667373 2.10036419339 0.278528774808 -1.21522166095
```

# Line Fitting

## Plot the fit

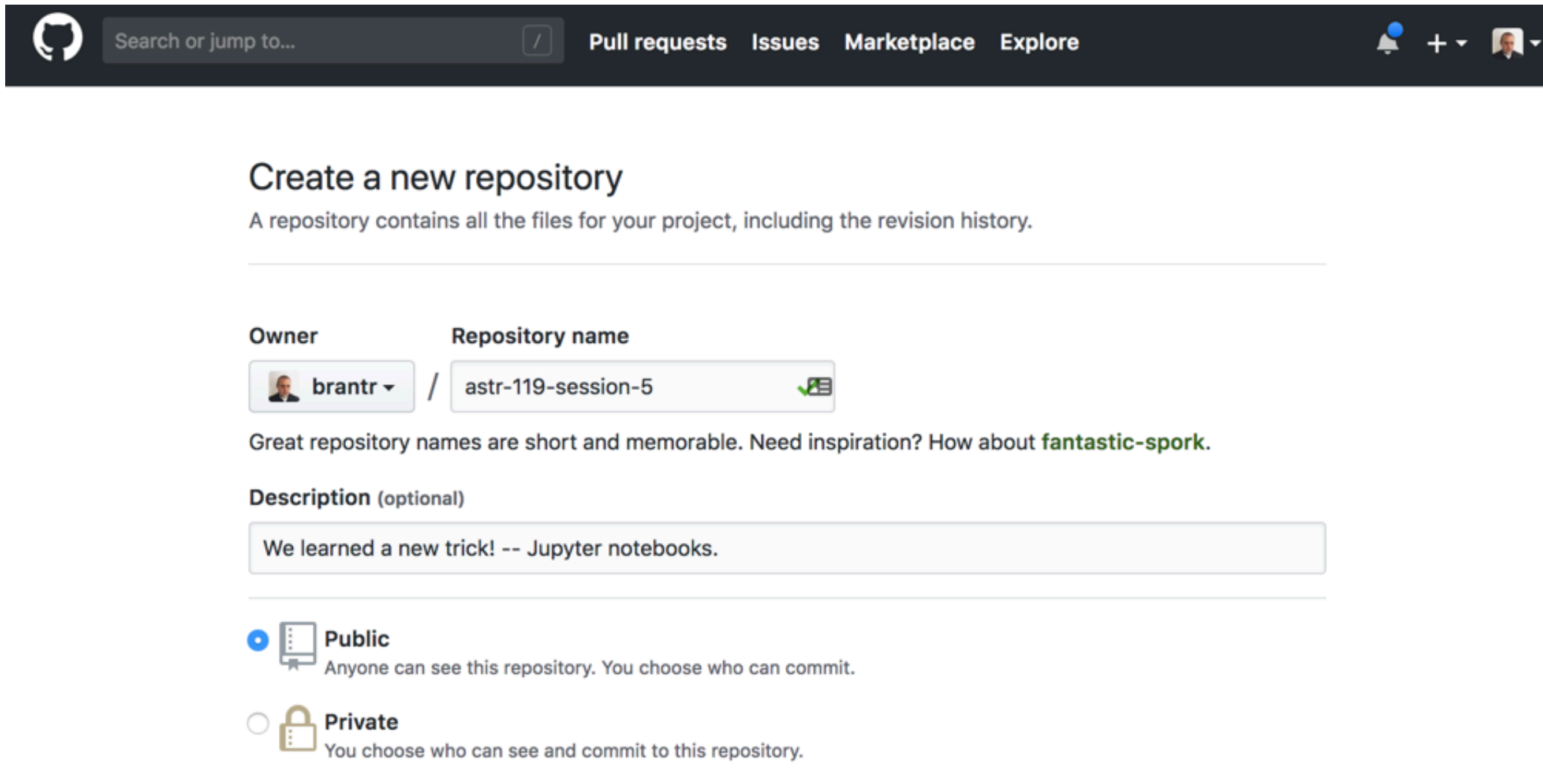
```
In [48]: f = plt.figure(figsize=(7,7))  
plt.errorbar(x,y,yerr=y_err,fmt='o',label='data')  
plt.plot(x,y_fit,label='fit')  
plt.xlabel('x')  
plt.ylabel('y')  
plt.legend(loc=0,frameon=False)
```

Out[48]: <matplotlib.legend.Legend at 0x11346b198>



# Save Your Work

Make a GitHub project “astr-119-session-6”, and commit the programs `my_first_jupyter_notebook.ipynb` and `test_matplotlib.ipynb` you made today.



The screenshot shows the GitHub interface for creating a new repository. The top navigation bar includes the GitHub logo, a search bar, and links to Pull requests, Issues, Marketplace, and Explore. The main heading is "Create a new repository" with a subtext explaining that a repository contains all files for a project, including revision history.

The form fields are as follows:

- Owner:** A dropdown menu showing the user "brantr".
- Repository name:** A text input field containing "astr-119-session-5", which is marked as valid with a green checkmark.
- Description (optional):** A text area containing the text "We learned a new trick! -- Jupyter notebooks."
- Visibility:** Two radio button options: "Public" (selected) and "Private". The "Public" option is accompanied by a document icon and the text "Anyone can see this repository. You choose who can commit." The "Private" option is accompanied by a lock icon and the text "You choose who can see and commit to this repository."