# Plots, functions, and conditional programming

AUTHOR Karl Gregory

### **Making plots**

We'll discover in using Python that you will need to import a library to do whatever the thing is that you want to do. In this note the thing that we want to do is to make plots, so we will import a library called matplotlib, which has this website: matplotlib.org. This is not the only plotting library for Python, but it seems to be "the hottest" right now.

```
import matplotlib.pyplot as plt
```

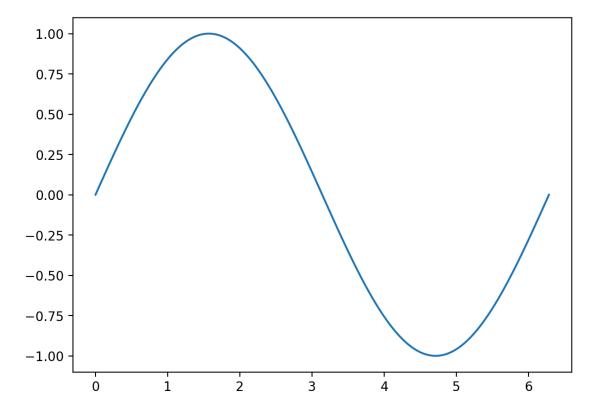
The above is the standard way of importing this library of functions. On the website for the library there are millions of examples of how to make use of all the functionality of this package; to these I will add my own humble offering here.

The first thing about matplotlib is that it plays well with NumPy, so let's import the NumPy library also with the typical alias np:

```
import numpy as np
```

Here is a demo of how to make a plot:

```
x = np.linspace(0,2*np.pi,361)
sinx = np.sin(x)
plt.plot(x,sinx)
plt.show()
```

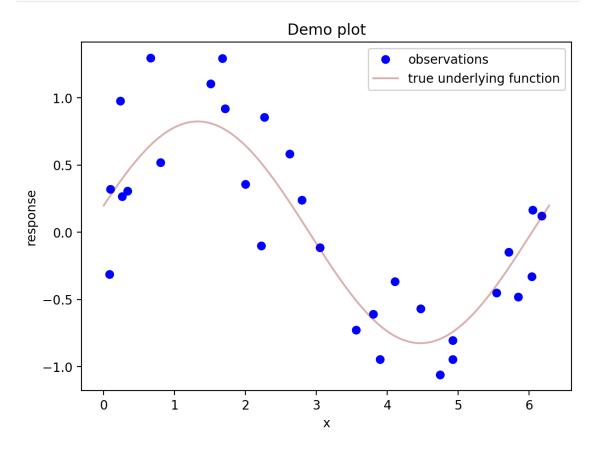


Here is a plot with a few more features, including a legend. Legends can be created in an automatic kind of way if a label= is supplied for each feature added to the plot.

```
# generate some scatterplot data such that the y values are equal to a function plus noise
u = np.random.random(30)*2*np.pi
v = 0.8*np.sin(u) + 0.2*np.cos(u) + np.random.normal(0,1/3,u.size)

# prepare to plot the true function
x = np.linspace(0,2*np.pi,361)
fx = 0.8*np.sin(x) + 0.2*np.cos(x)

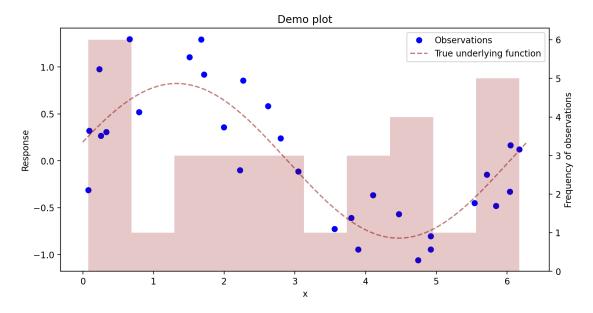
plt.figure() # make a new empty figure (so new plots don't keep going on top of the previo
plt.plot(u,v,'bo',label = 'observations')
plt.plot(x,fx,label = 'true underlying function',color=(0.545,0,0,.3)) # can put an rgb tu
plt.xlabel("x")
plt.ylabel("response")
plt.title("Demo plot")
plt.legend()
plt.show()
```



You can also spend some time setting up the axes and customizing the plot's appearance before you *show* the plot. In the matplotlib tutorials, this seems to be the preferred way to do things. Here is how it works: You have to set up a figure as well as a set of axes. Then you add things to the axes with ax. commands. When you do things this way, a new figure is set up, so you don't need to worry about your plotting commands simply adding things to your previous plot.

The code below also demonstrates how to set up a second vertical axis on the same horizontal axis.

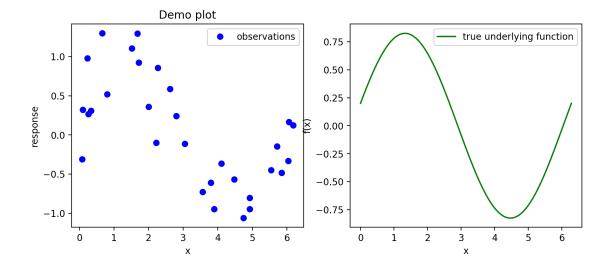
```
fig, ax = plt.subplots(figsize= (10,5)) # set up figure and axes. You can here change the ax.plot(u,v,'bo',label = 'Observations') ax.plot(x,fx,label = 'True underlying function',linestyle="--",color=(0.545,0,0,.5)) ax.set_xlabel("x")
```



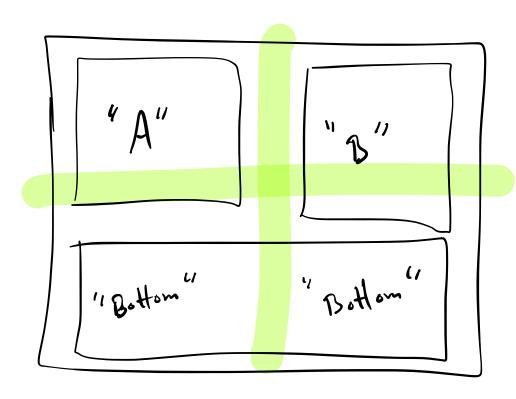
Using the latter approach of setting up a figure first allows one to create multi-plot figures. Each individual plot in a figure is referred to as an *axes*, or a set of axes.

```
fig, (ax1, ax2) = plt.subplots(1,2,figsize = (10,4)) # put in a 1 by 2 grid of plots, name
ax1.plot(u,v,'bo',label = 'observations')
ax1.set_xlabel("x")
ax1.set_ylabel("response")
ax1.set_title("Demo plot")
ax1.legend() # gets placed automatically

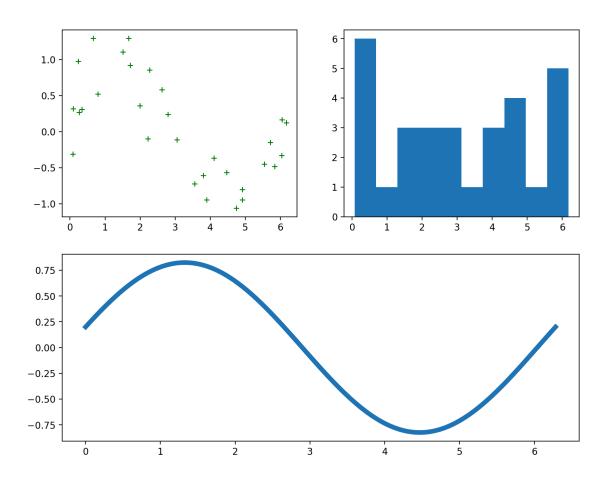
ax2.plot(x,fx,'g',label = 'true underlying function')
ax2.set_xlabel("x")
ax2.set_ylabel("f(x)")
ax2.legend()
plt.show()
```



We can set up plots in a "mosaic" arrangement as follows:



## Three amazing plots!



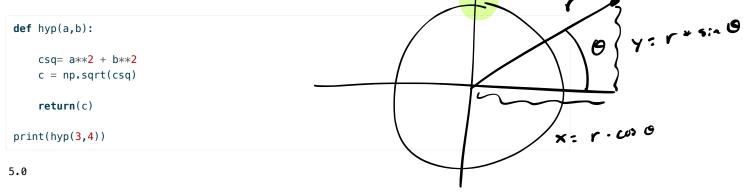
# **Defining new functions**

Here is the syntax for defining a simple, one-liner function:

```
def f(x): return(x*np.sin(x))
print(f(1))
```

#### 0.8414709848078965

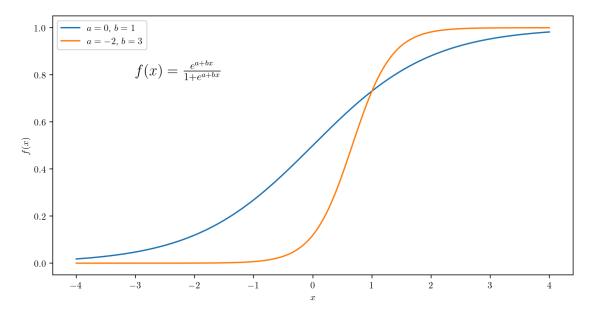
We must tell the function explicitly to return a value, or it will not return anything. If there is more than one line of code, we indent the lines, as shown below (no curly braces like in R). When we stop indenting, Python knows that the function definition has come to an end.



A function can be made to return more than one value. Just put every object to be returned in the return function.

We specify default values in Python just as we did in R:

```
def logistic(x,a=0,b=1):
    l = a + x*b
    val = np.exp(l) / (1 + np.exp(l))
    return(val)
plt.rcParams['text.usetex'] = True # ask to use LaTex in plot labels
x = np.linspace(-4,4,200)
fx1 = logistic(x) # use default values for a and b
fx2 = logistic(x,a = -2, b = 3)
fig, ax = plt.subplots(figsize=(10,5))
ax.plot(x,fx1,label = '$a = 0$, $b = 1$')
ax.plot(x,fx2,label = '$a = -2$, $b = 3$')
ax.set_xlabel('$x$')
ax.set_ylabel('$f(x)$')
ax.text(-3,.8,"f(x) = \frac{e^{a + bx}}{1 + e^{a + bx}}, fontsize = 16)
ax.legend()
plt.show()
```



# **Conditional programming**

The logic of conditional programming is the same across all programming languages. One has only to learn new syntax with each language. Here are some examples of conditional programs in Python:

```
def rolldice():
    r1 = np.random.choice(np.arange(1,7),1) # draw a number from 1 to 6
    r2 = np.random.choice(np.arange(1,7),1)

roll = str(r1) + str(r2)
    if( r1 == r2 ):
        if(r1 == 1):
            print("Hooo-eee, snake-eyes!")
        else:
            print("Yowza, doubles!")

print(roll)
```

#### [1][2]

The else if syntax in Python is a little different: Instead of typing "else if" you type elif:

```
def rolldice():
    r1 = np.random.choice(np.arange(1,7),1)
    r2 = np.random.choice(np.arange(1,7),1)
    roll = str(r1) + str(r2)
    if( r1 == r2 ):
        if(r1 == 1):
            print("Hooo-eee, snake-eyes!")
        elif(r1 == 6):
            print("Can't beat it!!!")
        else:
            print("Yowza, doubles!")
    elif((r1 + r2) <= 7):</pre>
        print("Ahh, sad rollin'.")
   elif((r1 + r2) > 10):
        print("Mighty fine!")
    else:
        print("Well now...")
    print(roll)
```

```
rolldice()
```

Yowza, doubles! [5][5]

For what it's worth, there is a function called bool() which coerces its argument to a boolean value. It may come in handy at some point:

```
print(bool(1))
print(bool(0))
```

True False

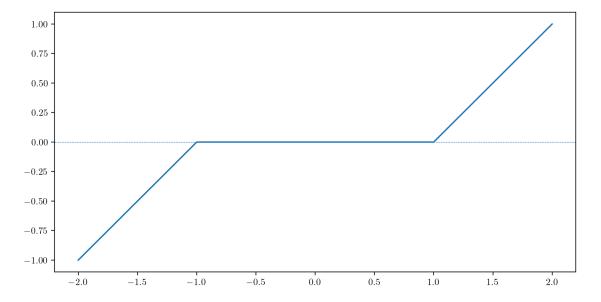
As we did before in R, we can take advantage of the coercion of a logical value to a numerical value to simplify the definition of piecewise defined functions. For example we can define the function

$$S(x,\lambda) = egin{cases} x+\lambda, & x<-\lambda \ 0, & -\lambda \leq x \leq \lambda \ x-\lambda, & \lambda < x \end{cases}$$

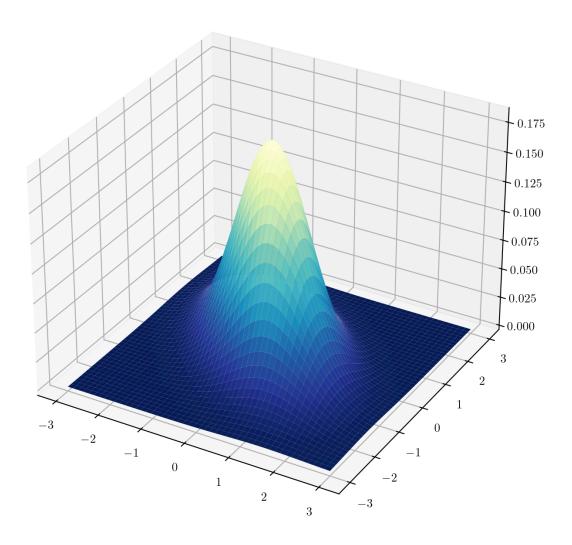
for  $\lambda > 0$  as below:

```
def softthresh(x,lam):
    return((x + lam) * (x < -lam) + (x - lam)*(x > lam))

x = np.linspace(-2,2,5)
lam = 1
fx = softthresh(x,lam)
fig, ax = plt.subplots(figsize = (10,5))
ax.plot(x,fx)
ax.axhline(0,linestyle="--",linewidth=.5)
plt.show()
```



Next we define a function to evaluate the bivariate standard normal density. Then we make a surface plot of the function:



### **Practice**

Practice writing code and reading code with the following exercises.

### Write code

1. Let  $x_1,\ldots,x_n$  be a set of real numbers and denote by  $x_{(1)}\leq\cdots\leq x_{(n)}$  the same set of values sorted in increasing order. The the quantile function of the empirical distribution of the set of points  $x_1,\ldots,x_n$  is given

$$Q_n(u) = x_{(|un|)},$$

for  $u\in(0,1)$ , where  $\lfloor\cdot\rfloor$  is the floor function. Write a function which will compute the interquartile range  $Q_n(3/4)-Q_n(1/4)$  according to the quantile function defined above for a set of values  $x_1,\ldots,x_n$  when these are stored in a one-dimensional NumPy array. You may make use of the <code>np.sort()</code> function. You can check your function against the output below:

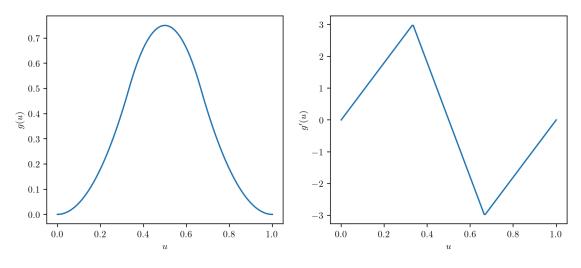
$$x = np.array([ 1.29, 1.36, 0.3, -1.17, -0.70, 0.43, 1.02, 1.02, 0.09, 1.48])$$
  
 $print(IQR(x))$ 

#### 1.72

2. Define g(u) and g'(u) as below and plot them over  $u \in [0,1]$  on side-by-side panels in a single figure as shown.

$$g(u) = egin{cases} (9/2)u^2, & 0 \leq u < 1/3 \ 9u - 9u^2 - 3/2, & 1/3 \leq u < 2/3 \ (9/2)(1-u)^2, & 2/3 \leq u < 1. \end{cases}$$

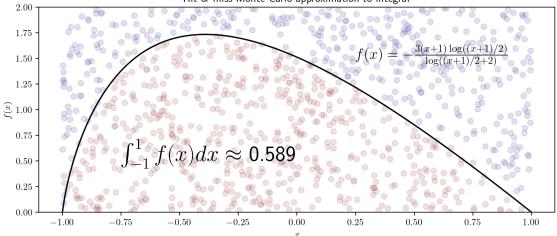
$$g'(u) = egin{cases} 9u, & 0 \leq u < 1/3 \ 9 - 18u, & 1/3 \leq u < 2/3 \ 9(1-u), & 2/3 \leq u < 1. \end{cases}$$



 ${\it 3. Obtain a Monte Carlo approximation using the hit-or-miss method to the integral}\\$ 

$$I = \int_{-1}^{1} -\frac{3(x+1)\log((x+1)/2)}{\log((x+1)/2+2)} dx$$

and make a plot like the one shown below:



4. Write a function to convert a set of Euclidean (x,y) coordinates stored in NumPy arrays  $\times$  and y to polar coordinates  $(r,\theta)$  via

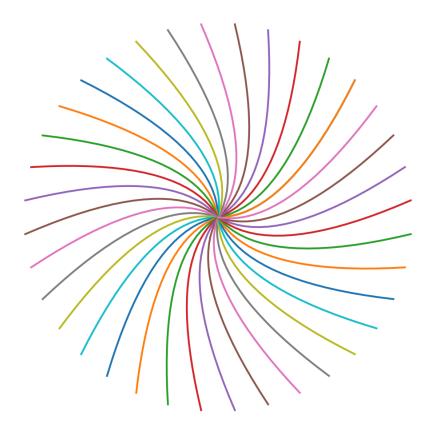
$$r = \sqrt{x^2 + y^2} \ ext{tan}^{-1}(y/x), \qquad x > 0 \ ext{$\pi + ext{tan}^{-1}(y/x)$, } \quad x < 0 \ ext{$\pi/2$, } \quad x = 0, y > 0 \ ext{$-\pi/2$, } \quad x = 0, y < 0$$

as well as a function to convert a point  $(r,\theta)$  given in Polar coordinates to Euclidean coordinates (x,y) via

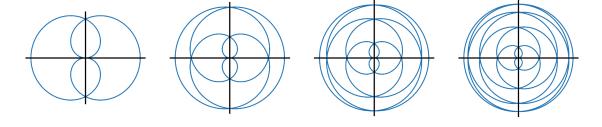
$$x = r\cos\theta$$
$$y = r\sin\theta.$$

Moreover, write a function to rotate a set of points given as (x,y) coordinates in the Euclidean plan (in some NumPy vectors  $\times$  and y) a number of degrees counter-clockwise; you can achieve this by converting the points to polar coordinates and then adding to the value of the angle  $\theta$ . Afterwards convert the points back to Euclidean coordinates. Use your functions to make the plot below of the points created in this code chunk rotated around the origin and plotted at every ten degrees.

```
x = np.linspace(0,1,21)
y = x**2
```



5. For  $j\in\{2,4,6,8\}$ , make a plot of the polar function defined by  $r=\sin(\theta/j)$  for  $\theta\in[0,4j\pi)$ . Evaluate the radius r over a grid of  $\theta$  values and then convert the polar coordinates  $(r,\theta)$  to Euclidean coordinates (x,y) for plotting. The plots should look like this:



### **Read code**

Anticipate the output of the following code chunks:

1.

```
def star(p,s):
    th = np.linspace(np.pi/2,np.pi/2 + 2*np.pi*s,s*p+1)
    x = np.cos(th[::s])
    y = np.sin(th[::s])
```

```
return x, y

x, y = star(5,2)

fig, ax = plt.subplots(figsize=(4,4))
ax.set_aspect('equal')
ax.axis('off')
ax.plot(x,y)
plt.show()
```