

Python-Matplotlib

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matplotlib

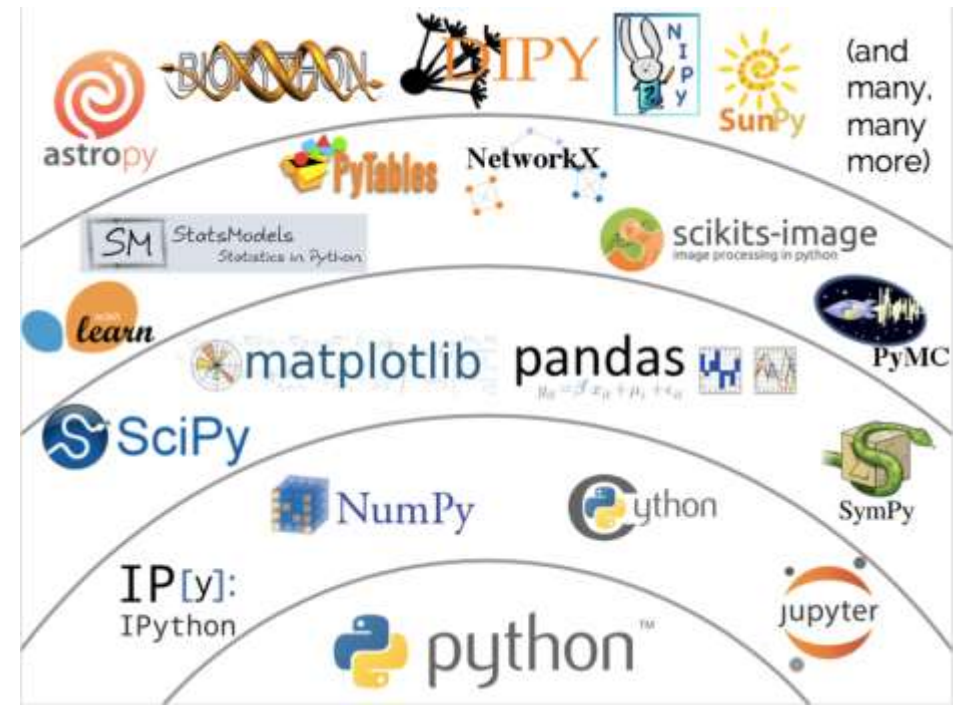


Agenda

- Introduction
- History, usage
- Adjusting the Plot: Line Colors, style, etc.
- Scatter plot
- Density and Contour Plots
- Visualizing 3D



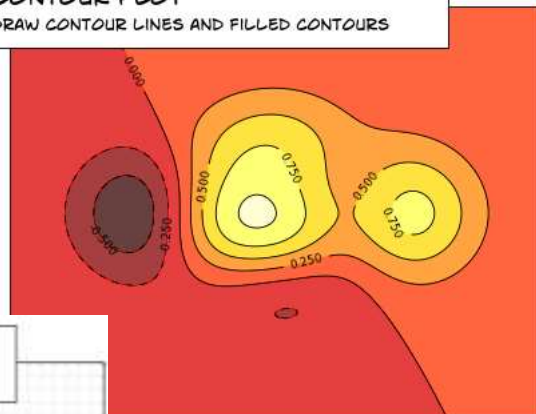
*One guiding principle of Python code is that
“explicit is better than implicit”*





Type of plots

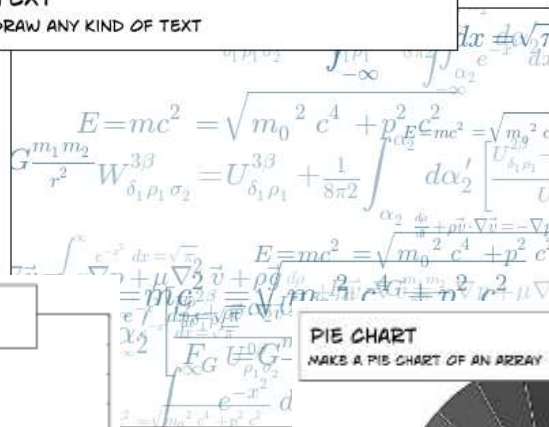
CONTOUR PLOT
DRAW CONTOUR LINES AND FILLED CONTOURS



GRID
DRAW TICKS AND GRID



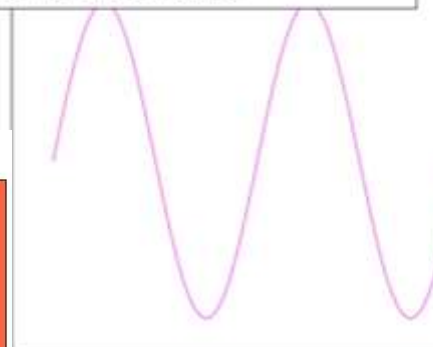
TEXT
DRAW ANY KIND OF TEXT



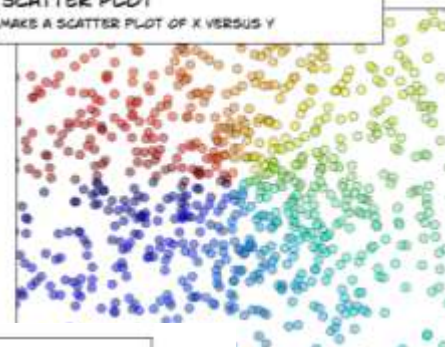
MULTIPLY
PLOT SEVERAL PLOTS AT ONCE



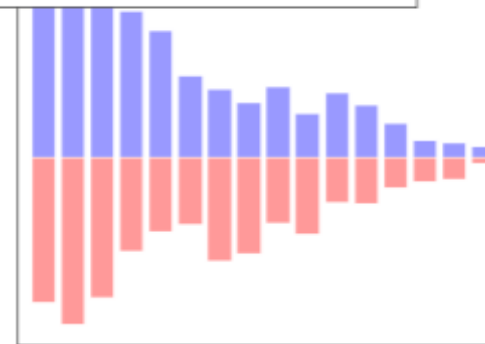
REGULAR PLOT
PLOT LINES AND/OR MARKERS



SCATTER PLOT
MAKE A SCATTER PLOT OF X VERSUS Y



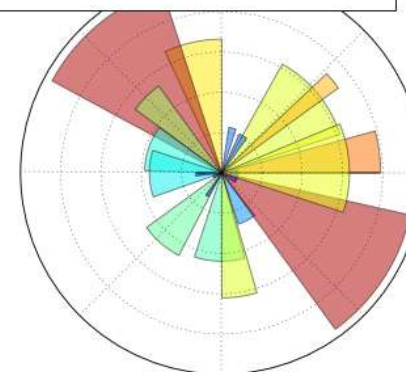
BAR PLOT
MAKE A BAR PLOT WITH RECTANGLES



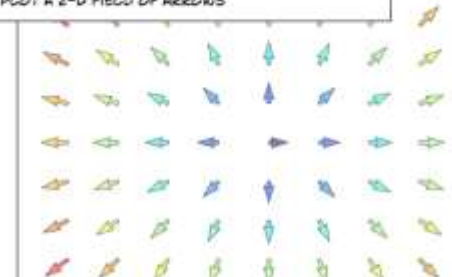
IMSHOW
DISPLAY AN IMAGE TO CURRENT AXIS



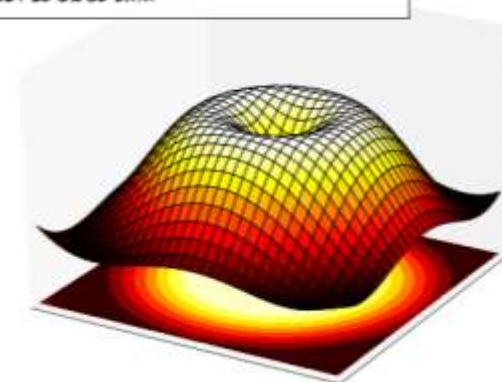
POLAR AXIS
PLOT ANYTHING USING POLAR AXIS



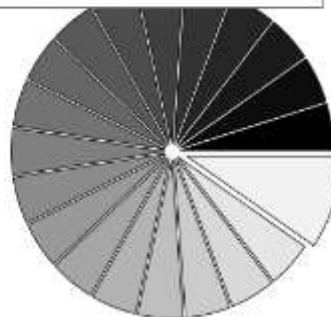
QUIVER PLOT
PLOT A 2-D FIELD OF ARROWS



3D PLOTS
PLOT 2D OR 3D DATA



PIE CHART
MAKE A PIE CHART OF AN ARRAY



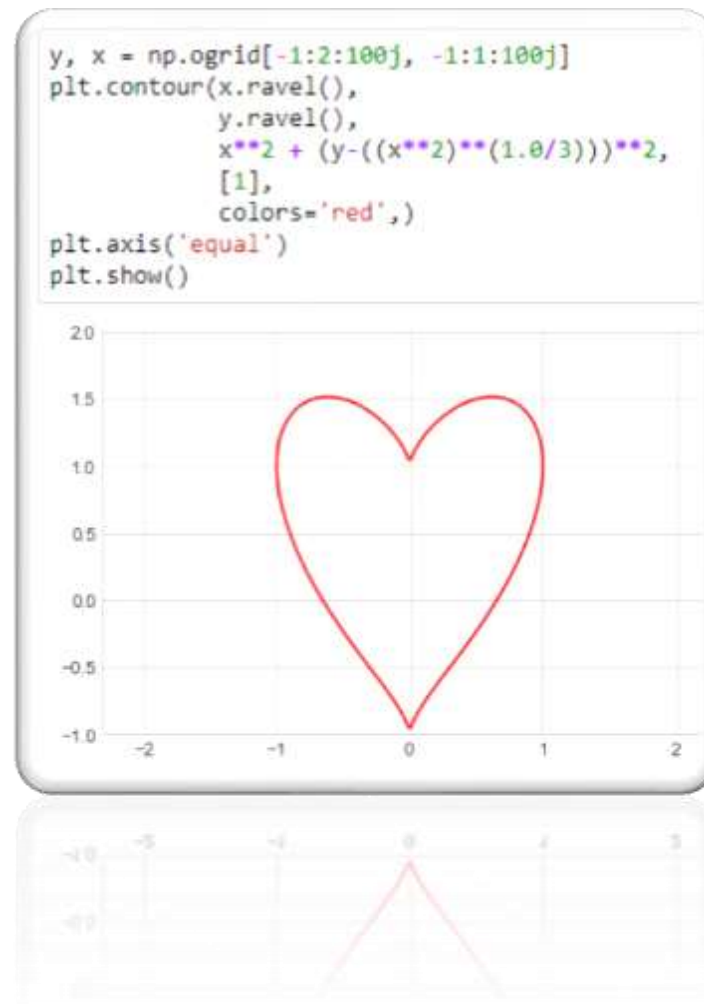
For more information

<https://matplotlib.org/gallery/index.html>



Introduction

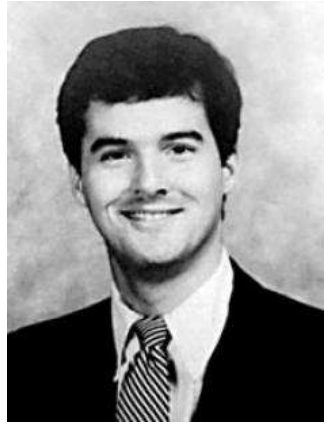
- Matplotlib is a Python 2D plotting library which produces publication quality figures
- Matplotlib is a multiplatform data visualization library built on NumPy arrays, and designed to work with the broader SciPy stack.





History

- Matplotlib is the brainchild of [John Hunter](#) (1968-2012), who, along with its many contributors, have put an immeasurable amount of time and effort into producing a piece of software utilized by thousands of scientists worldwide.
- Alums of Princeton University
- American neurobiologist
- one of the founding directors of NumFOCUS Foundation
- Matplotlib was originally conceived to visualize Electrocorticography (ECoG) data of epilepsy patients during post-doctoral research in Neurobiology

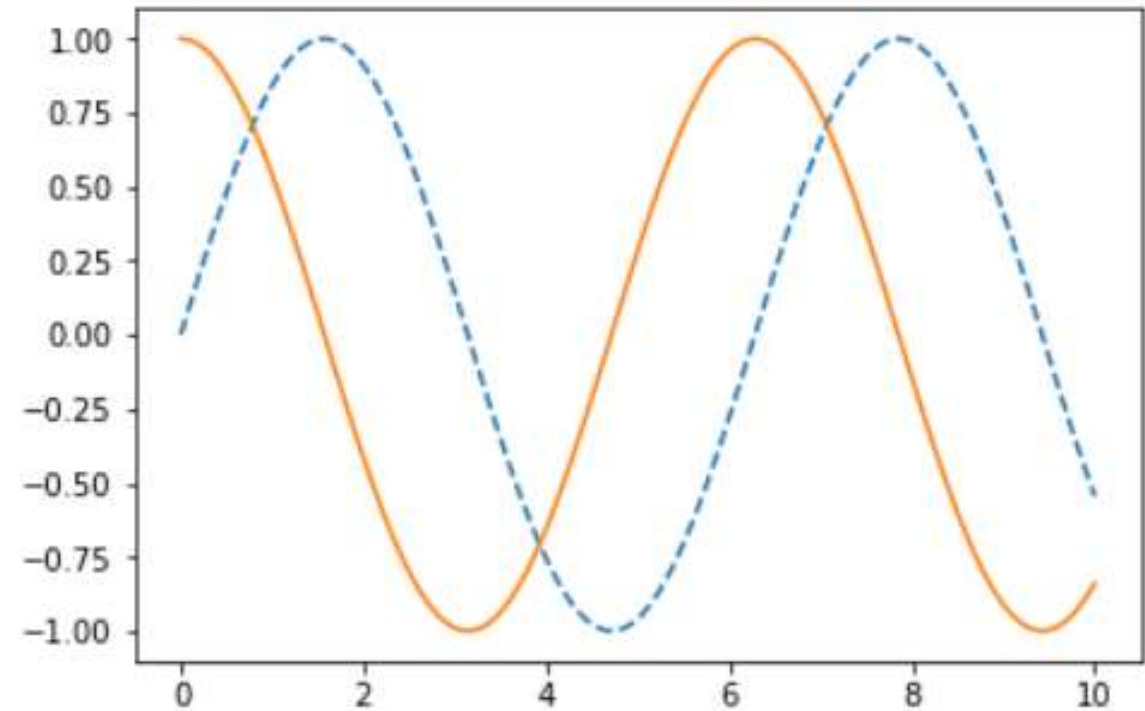




History

- John Hunter in 2002, originally conceived it as a patch to IPython for enabling interactive MATLAB-style plotting via gnuplot from the IPython command line
- IPython's creator, Fernando Perez, was at the time scrambling to finish his PhD, and let John know he wouldn't have time to review the patch for several months. John took this as a cue to set out on his own, and the Matplotlib package was born, with version 0.1 released in 2003.
- It received an early boost when it was adopted as the plotting package of choice of the Space Telescope Science Institute (the folks behind the Hubble Telescope), which financially supported Matplotlib's development and greatly expanded its capabilities

```
import matplotlib.pyplot as plt
import numpy as np
x = np.linspace(0, 10, 100)
plt.plot(x, np.sin(x), '--')
plt.plot(x, np.cos(x))
plt.show()
```





Two Interfaces for the Price of One

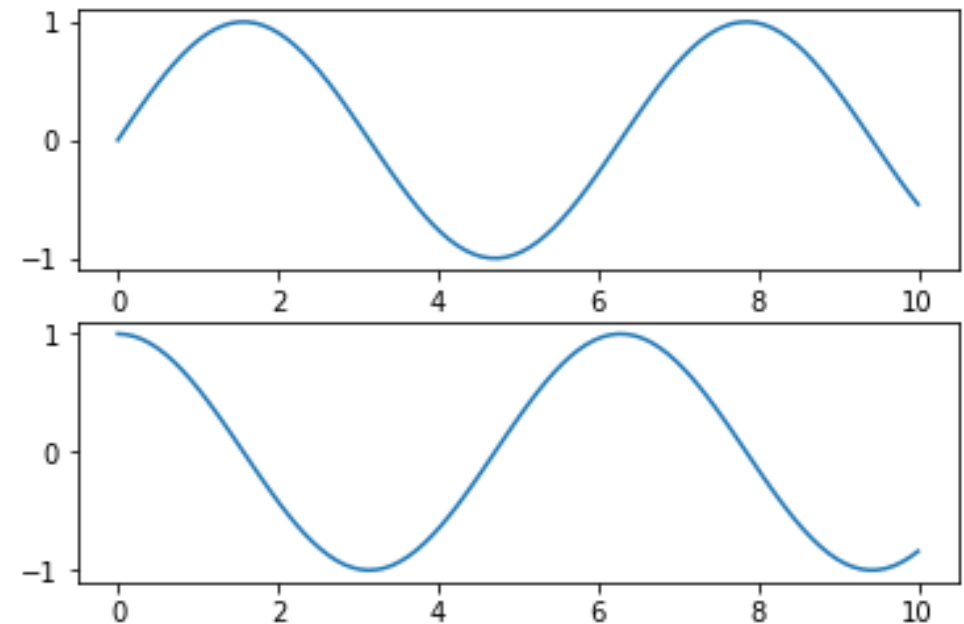
- A potentially confusing feature of Matplotlib is its dual interfaces: *a convenient MATLAB-style state-based interface, and a more powerful object-oriented interface.*
1. **MATLAB-style interface** : Matplotlib was originally written as a Python alternative for MATLAB users, and much of its syntax reflects that fact. The MATLAB-style tools are contained in the pyplot (plt) interface.
 2. **Object-oriented interface** : The object-oriented interface is available for these more complicated situations, and with more control over figure.



MATLAB-style interface

- It's important to note that this interface is **stateful**: it keeps track of the “current” figure and axes.

```
In [12]: plt.figure() # create a plot figure
# create the first of two panels and set current axis
plt.subplot(2, 1, 1) # (rows, columns, panel number)
plt.plot(x, np.sin(x))
# create the second panel and set current axis
plt.subplot(2, 1, 2)
plt.plot(x, np.cos(x));
plt.show();
```

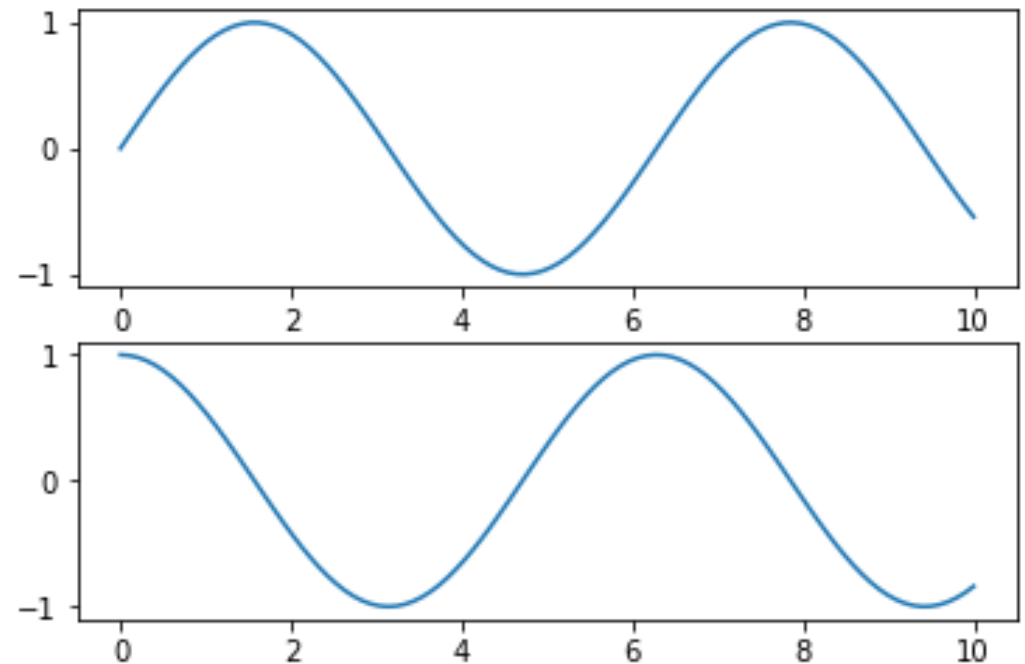




Object-oriented interface

- In the object-oriented interface the plotting functions are methods of explicit Figure and Axes objects.
- With subplot you can arrange plots in a regular grid.

```
# First create a grid of plots  
# ax will be an array of two Axes objects  
fig, ax = plt.subplots(2)  
# Call plot() method on the appropriate object  
ax[0].plot(x, np.sin(x))  
ax[1].plot(x, np.cos(x));  
plt.show();
```



subplot(2,1,1)

subplot(2,1,2)

subplot(1,2,1)

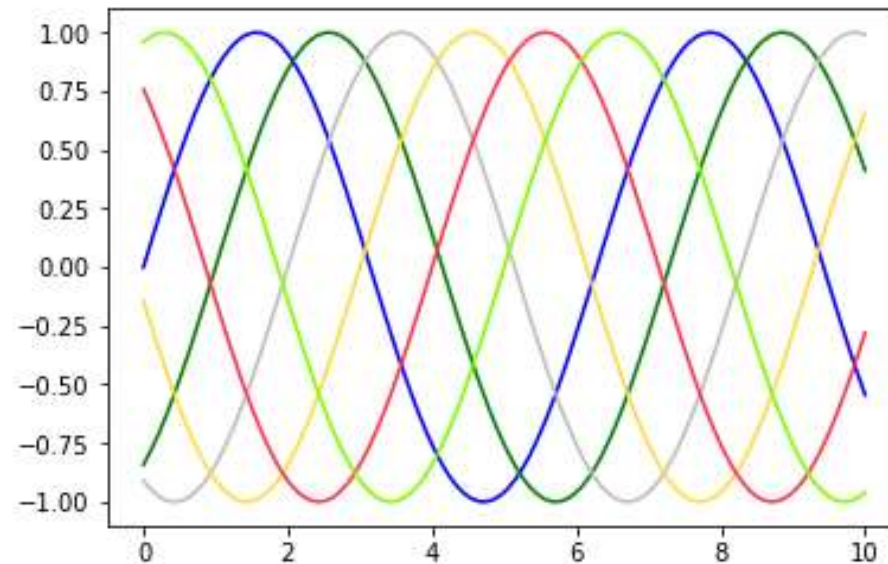
subplot(1,2,2)



Adjusting the Plot: Line Colors

- If no color is specified, Matplotlib will automatically cycle through a set of default colors for multiple lines.

```
plt.plot(x, np.sin(x - 0), color='blue')      # specify color by name
plt.plot(x, np.sin(x - 1), color='g')        # short color code (rgbcmyk)
plt.plot(x, np.sin(x - 2), color='0.75')     # Grayscale between 0 and 1
plt.plot(x, np.sin(x - 3), color='#FFDD44')  # Hex code (RRGGBB from 00 to FF)
plt.plot(x, np.sin(x - 4), color=(1.0,0.2,0.3)) # RGB tuple, values 0 and 1
plt.plot(x, np.sin(x - 5), color='chartreuse'); # all HTML color names supported
plt.show();
```



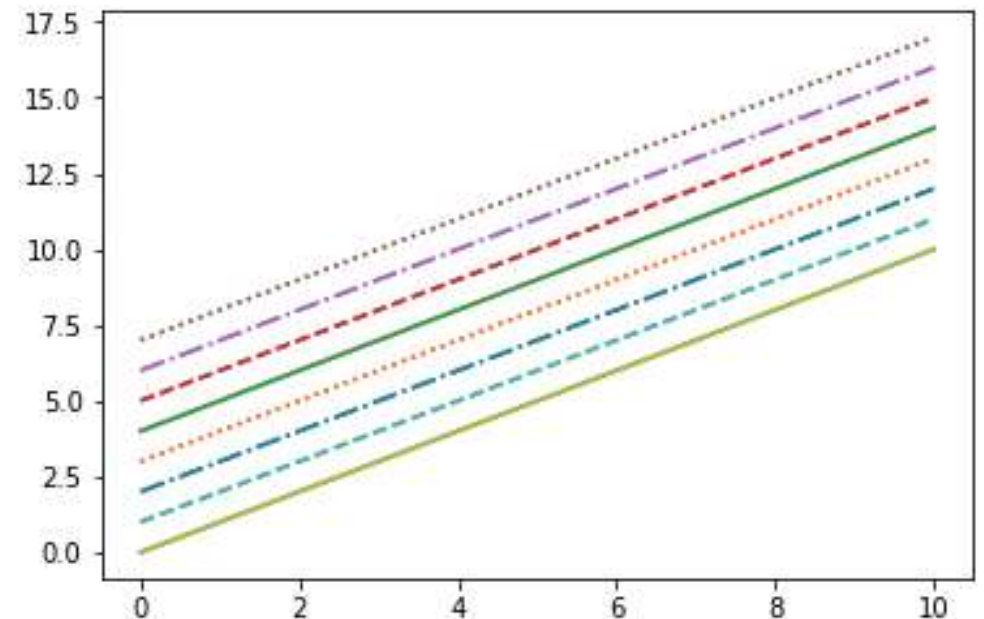


Adjusting the Plot: Styles

- The linestyle and color codes can be combined into a single non-keyword argument

```
plt.plot(x, x + 0, '-g') # solid green  
plt.plot(x, x + 1, '--c') # dashed cyan  
plt.plot(x, x + 2, '-.k') # dashdot black  
plt.plot(x, x + 3, ':r'); # dotted red
```

```
plt.plot(x, x + 0, linestyle='solid')  
plt.plot(x, x + 1, linestyle='dashed')  
plt.plot(x, x + 2, linestyle='dashdot')  
plt.plot(x, x + 3, linestyle='dotted');  
# For short, you can use the following codes:  
plt.plot(x, x + 4, linestyle='-') # solid  
plt.plot(x, x + 5, linestyle='--') # dashed  
plt.plot(x, x + 6, linestyle='-.') # dashdot  
plt.plot(x, x + 7, linestyle=':'); # dotted  
plt.show();
```



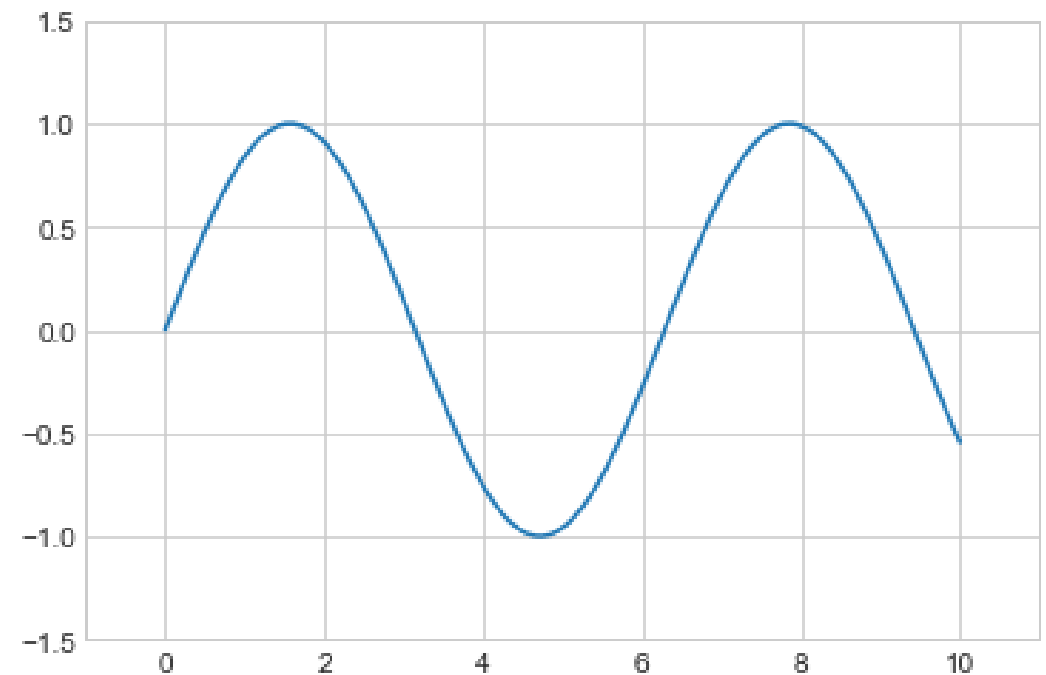


Adjusting the Plot: Axes Limits

- basic way to adjust axis limits is to use the `plt.xlim()` and `plt.ylim()` methods
- The `plt.axis()` method allows you to set the x and y limits with a single call, by passing a list that specifies `[xmin, xmax, ymin, ymax]`
- `plt.axis('tight');` #automatically tighten the bounds around the current plot
- `plt.axis('equal');` #equal aspect ratio

axes is different from axis

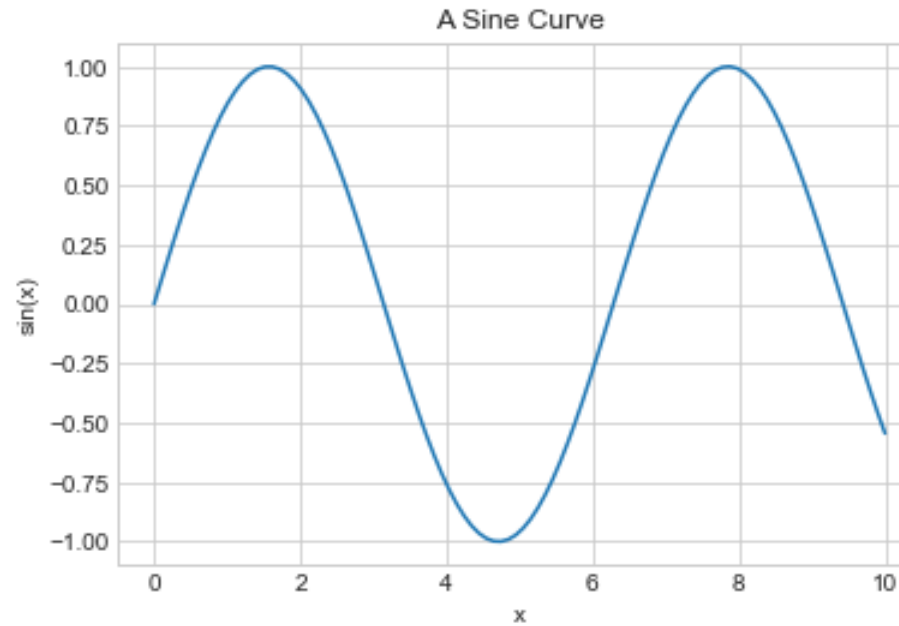
```
plt.style.use('seaborn-whitegrid')
plt.plot(x, np.sin(x))
plt.xlim(-1, 11)
plt.ylim(-1.5, 1.5)
plt.show();
```



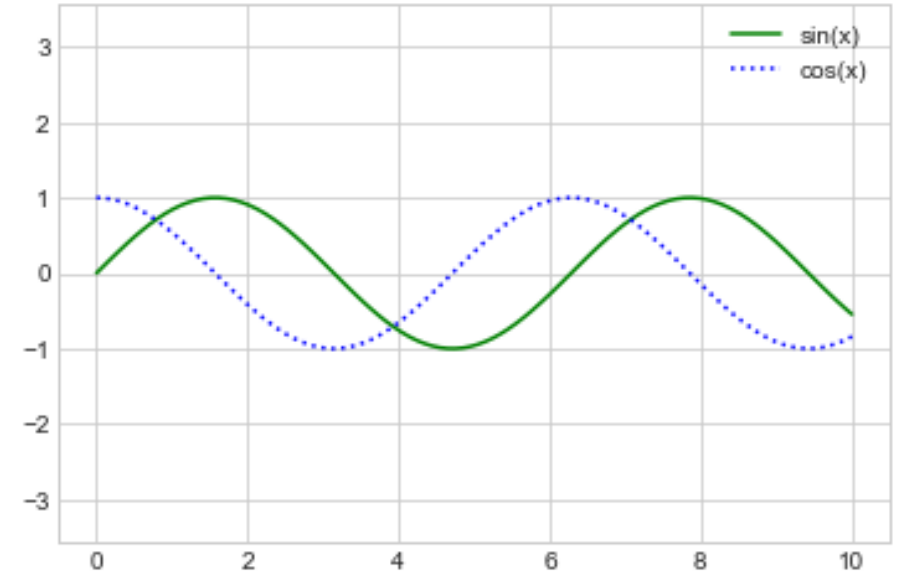


Labeling Plots

```
19]: plt.style.use('seaborn-whitegrid')  
plt.plot(x, np.sin(x))  
plt.title("A Sine Curve")  
plt.xlabel("x")  
plt.ylabel("sin(x)");  
plt.show();
```



```
[20]: plt.plot(x, np.sin(x), '-g', label='sin(x)')  
plt.plot(x, np.cos(x), ':b', label='cos(x)')  
plt.axis('equal')  
plt.legend(); plt.show();
```



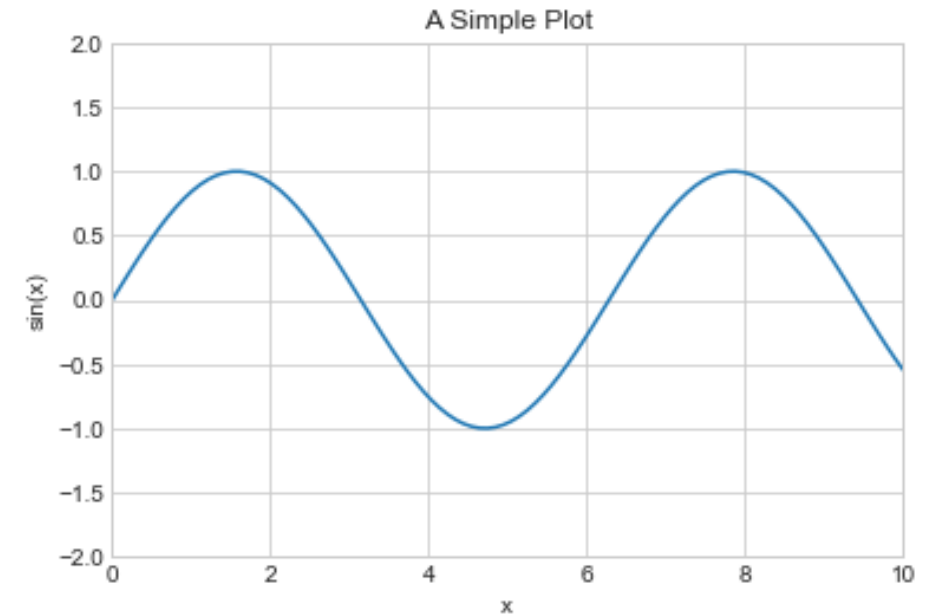


Matplotlib Gotchas

For transitioning between MATLAB-style functions and object-oriented methods, make the following changes:

- `plt.xlabel()` → `ax.set_xlabel()`
- `plt.ylabel()` → `ax.set_ylabel()`
- `plt.xlim()` → `ax.set_xlim()`
- `plt.ylim()` → `ax.set_ylim()`
- `plt.title()` → `ax.set_title()`

```
In [26]: ax = plt.axes()  
ax.plot(x, np.sin(x))  
ax.set(xlim=(0, 10), ylim=(-2, 2), xlabel='x',  
       ylabel='sin(x)', title='A Simple Plot');  
plt.show();
```

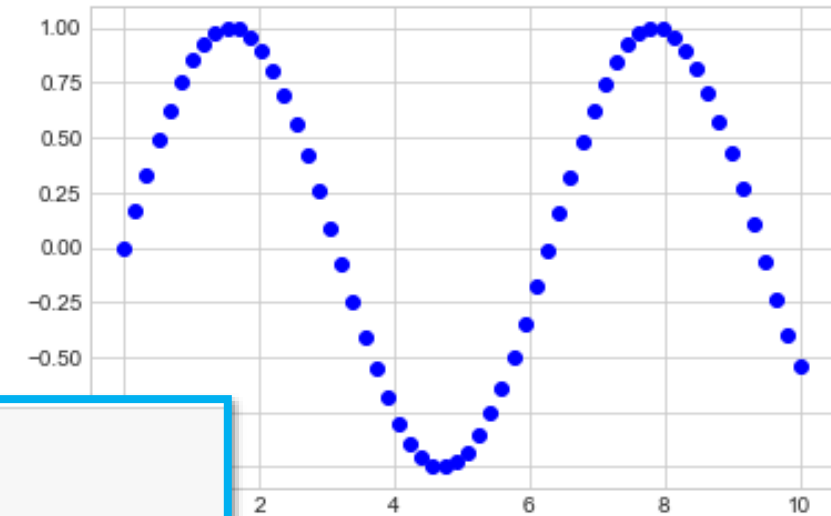




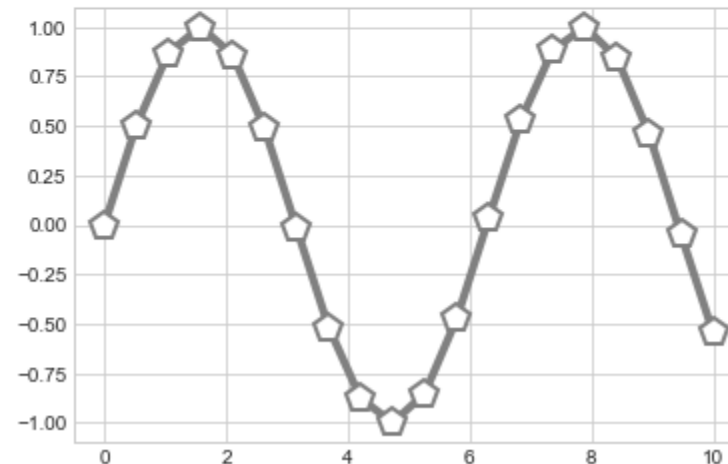
Scatter Plots

- scatter plot, a close cousin of the line plot
- `plt.plot(x, y, '-ok');`
line (-), circle marker (o), black (k)

```
41]: x = np.linspace(0, 10, 60)  
y = np.sin(x)  
plt.plot(x, y, 'o', color='blue');  
plt.show();
```



```
43]: x = np.linspace(0, 10, 20)  
y = np.sin(x)  
plt.plot(x, y, '-p', color='gray',  
         markersize=15, linewidth=4,  
         markerfacecolor='white',  
         markeredgecolor='gray',  
         markeredgewidth=2)  
plt.show();
```

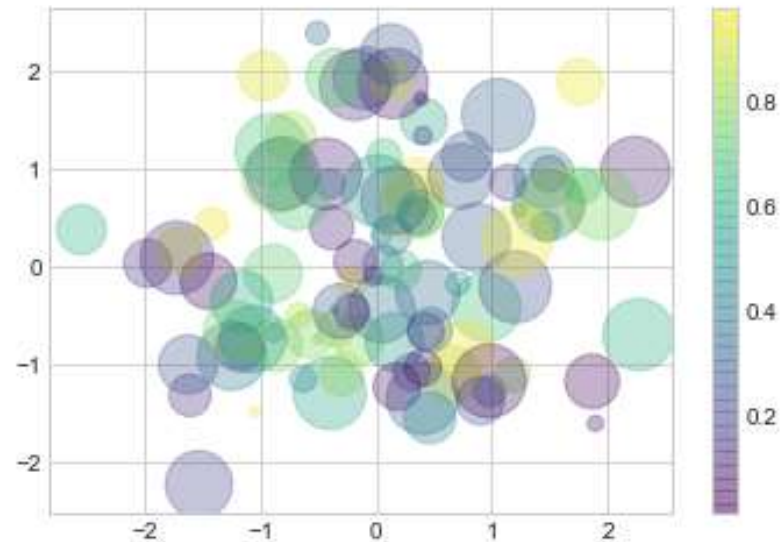




Scatter Plots-efficiency

- as datasets get larger than a few thousand points, plt.plot is noticeably more efficient than plt.scatter.
- plt.scatter has the capability to render a different size and/or color for each point
- plt.plot, the points are always essentially clones of each other
- plt.plot should be preferred over plt.scatter for large datasets.

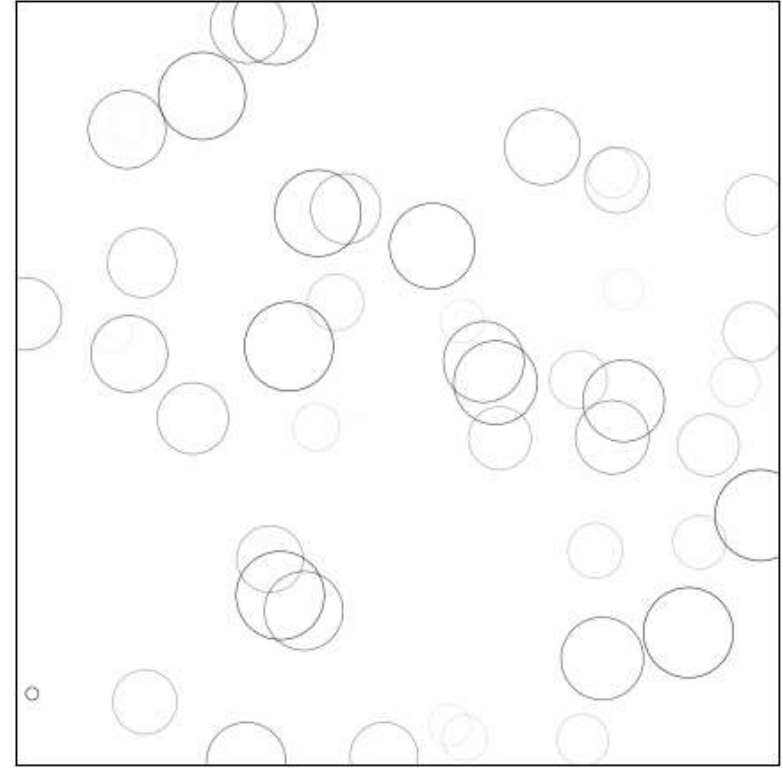
```
n [46]: rng = np.random.RandomState(0)
x = rng.randn(100)
y = rng.randn(100)
colors = rng.rand(100)
sizes = 1000 * rng.rand(100)
plt.scatter(x, y, c=colors, s=sizes, alpha=0.3, cmap='viridis')
plt.colorbar(); # show color scale
plt.show();
```





Drip Drop

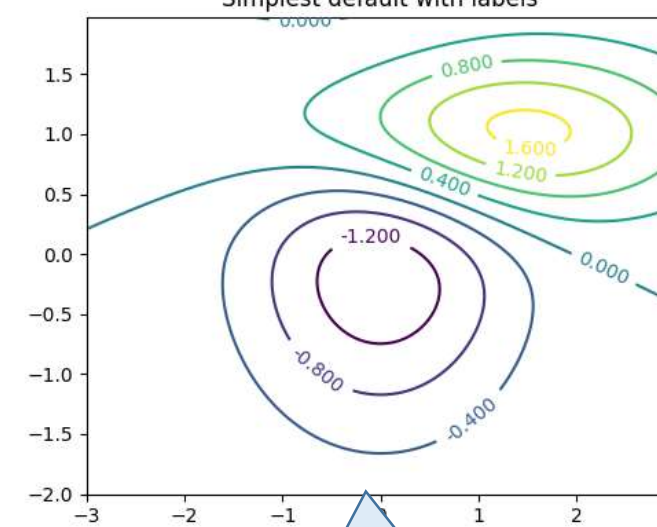
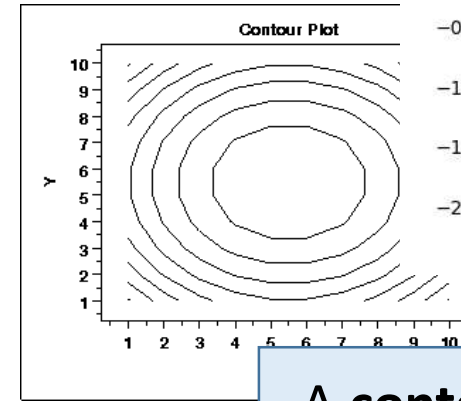
- A very simple rain effect can be obtained by having small growing rings randomly positioned over a figure. Of course, they won't grow forever since the wave is supposed to damp with time.
- To simulate that, we can use a more and more transparent color as the ring is growing, up to the point where it is no more visible. At this point, we remove the ring and create a new one.
- Don't remove the largest ring but re-use it to set a new ring at a new random position, with nominal size and color.
- Keeping count of rings constant





Density and Contour Plots

- It is useful to display three-dimensional data in two dimensions using contours or color-coded regions.
- There are three Matplotlib functions that can be helpful for this task:
 - `plt.contour` for contour plots,
 - `plt.contourf` for filled contour plots,
 - `plt.imshow` for showing images.



A **contour plot** is a graphical technique for representing a 3-dimensional surface by plotting constant z slices, called **contours**, on a 2-dimensional format. That is, given a value for z , lines are drawn for connecting the (x,y) coordinates where that z value occurs.



Density and Contour Plots

- A contour plot can be created with the `plt.contour` function.
- It takes three arguments:
 - a grid of x values, a grid of y values, and a grid of z values.
 - The x and y values represent positions on the plot, and the z values will be represented by the contour levels.
- `np.meshgrid` function is used, which builds two-dimensional grids from one-dimensional arrays



meshgrid working

```
xlist = np.linspace(-3.0, 3.0, 3)
ylist = np.linspace(-3.0, 3.0, 4)
X, Y = np.meshgrid(xlist, ylist)
Z = np.sqrt(X**2 + Y**2)
print("\nxlist value");print(xlist);
print("\nylist value");print(ylist);
print("\nX value");print(X);
print("\nY value");print(Y);
print("\nZ value");print(Z);
```

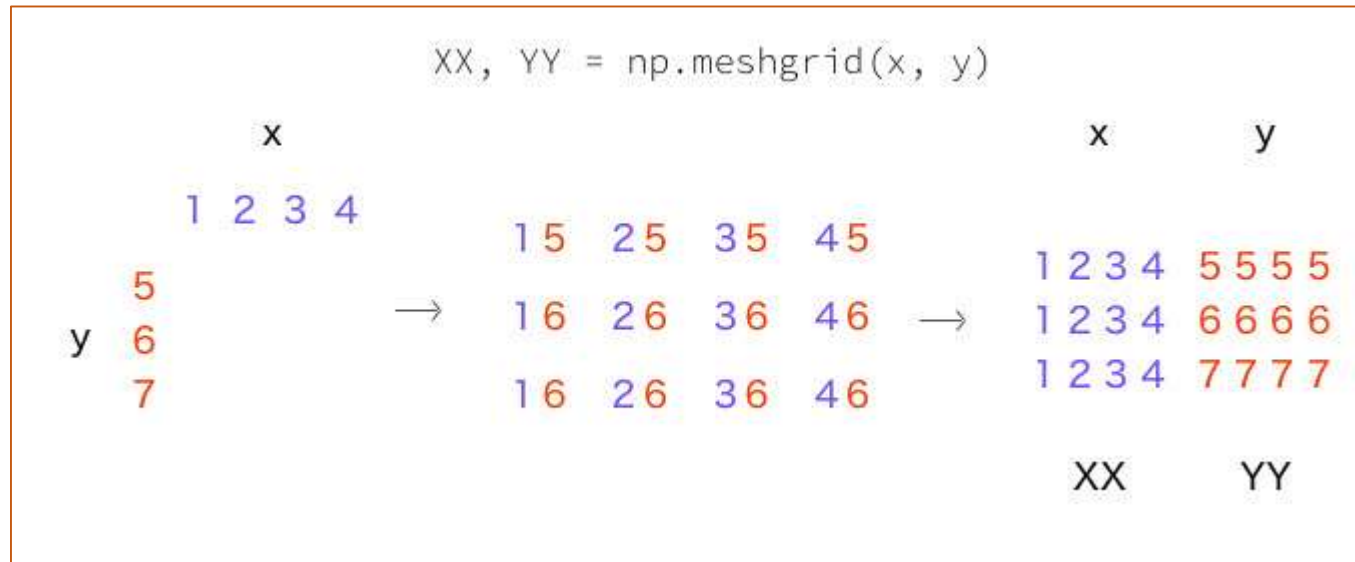
xlist value
[-3. 0. 3.]

ylist value
[-3. -1. 1. 3.]

X value
[[-3. 0. 3.]
 [-3. 0. 3.]
 [-3. 0. 3.]
 [-3. 0. 3.]]

Y value
[[-3. -3. -3.]
 [-1. -1. -1.]
 [1. 1. 1.]
 [3. 3. 3.]]

Z value
[[4.24264069 3. 4.24264069]
 [3.16227766 1. 3.16227766]
 [3.16227766 1. 3.16227766]
 [4.24264069 3. 4.24264069]]

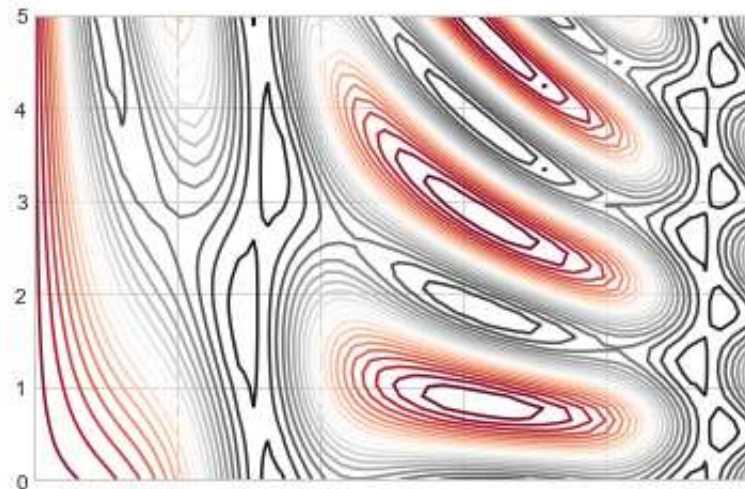




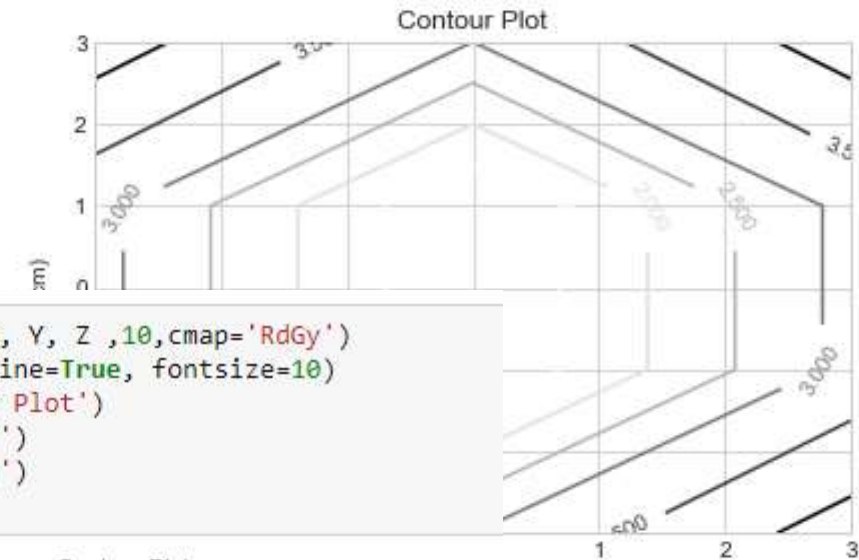
```
3]: def f(x, y):
    return np.sin(x) ** 10 + np.cos(10 + y * x) * np.cos(x)
x = np.linspace(0, 5, 50)
y = np.linspace(0, 5, 40)
X, Y = np.meshgrid(x, y)
Z = f(X, Y)
plt.contour(X, Y, Z, colors='black');
plt.show();
```



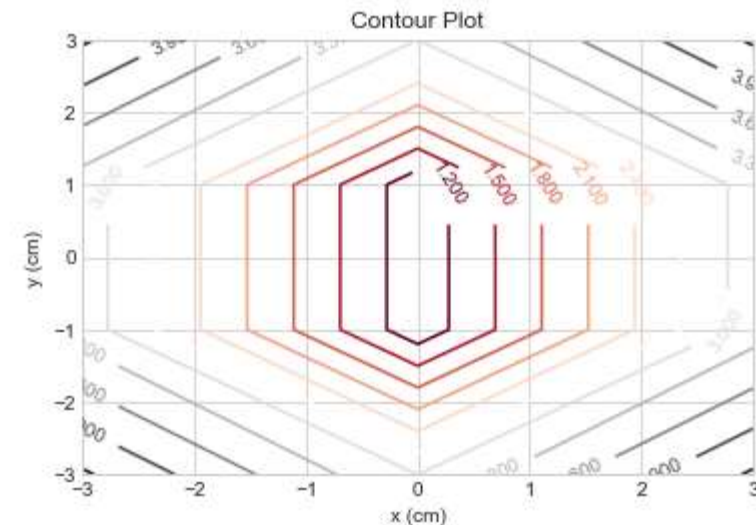
```
]: plt.contour(X, Y, Z, 20, cmap='RdGy');
plt.show();
```



```
[65]: cp = plt.contour(X, Y, Z )
plt.clabel(cp, inline=True,
           fontsize=10)
plt.title('Contour Plot')
plt.xlabel('x (cm)')
plt.ylabel('y (cm)')
plt.show();
```



```
[67]: cp = plt.contour(X, Y, Z ,10,cmap='RdGy')
plt.clabel(cp, inline=True, fontsize=10)
plt.title('Contour Plot')
plt.xlabel('x (cm)')
plt.ylabel('y (cm)')
plt.show();
```

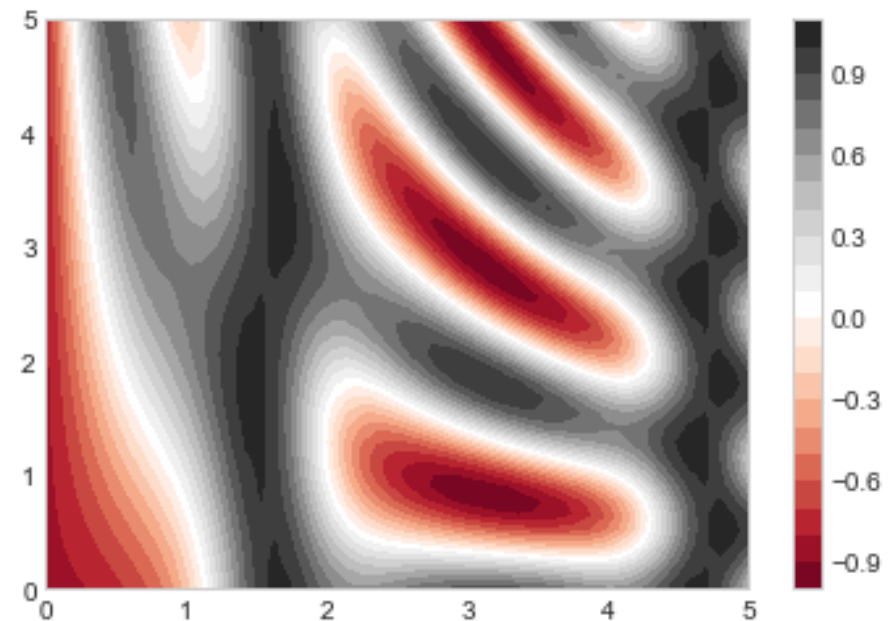




Filled contours

- The colorbar makes it clear that the black regions are “peaks,” while the red regions are “valleys.”

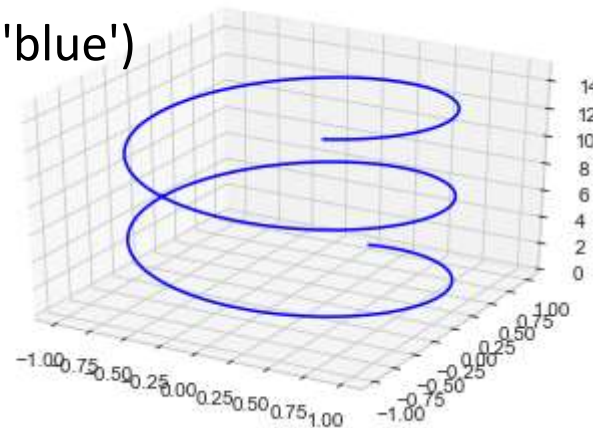
```
[88]: def f(x, y):  
        return np.sin(x) ** 10 + np.cos(10 + y * x) * np.cos(x)  
x = np.linspace(0, 5, 50)  
y = np.linspace(0, 5, 40)  
X, Y = np.meshgrid(x, y)  
Z = f(X, Y)  
plt.contourf(X, Y, Z, 20, cmap='RdGy');  
plt.colorbar();  
plt.show();
```



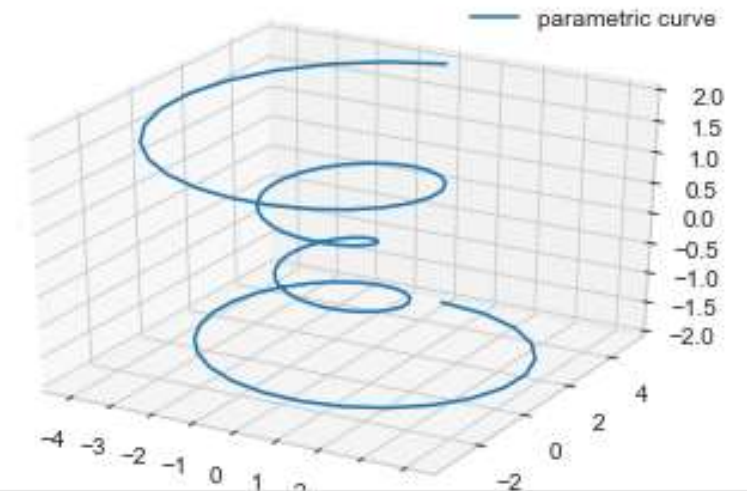


Three-Dimensional Plotting in Matplotlib

```
%matplotlib inline
from mpl_toolkits import mplot3d
import numpy as np
import matplotlib.pyplot as plt
fig = plt.figure()
ax = plt.axes(projection='3d')
# Data for a three-dimensional line
zline = np.linspace(0, 15, 1000)
xline = np.sin(zline)
yline = np.cos(zline)
ax.plot3D(xline, yline, zline, 'blue')
```



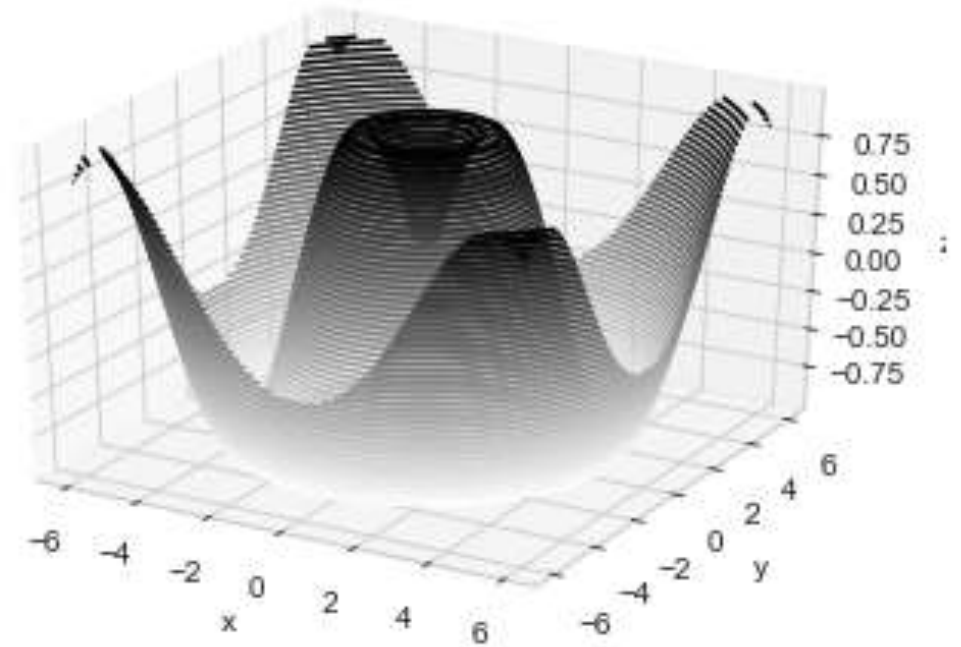
```
import matplotlib as mpl
from mpl_toolkits.mplot3d import Axes3D
import numpy as np
import matplotlib.pyplot as plt
mpl.rcParams['legend.fontsize'] = 10
fig = plt.figure()
ax = fig.gca(projection='3d')
theta = np.linspace(-4 * np.pi, 4 * np.pi, 100)
z = np.linspace(-2, 2, 100)
r = z**2 + 1
x = r * np.sin(theta)
y = r * np.cos(theta)
ax.plot(x, y, z, label='parametric curve')
ax.legend()
plt.show()
```





3D contours

```
def f(x, y):  
    return np.sin(np.sqrt(x ** 2 + y ** 2))  
  
x = np.linspace(-6, 6, 30)  
y = np.linspace(-6, 6, 30)  
X, Y = np.meshgrid(x, y)  
Z = f(X, Y)  
  
fig = plt.figure()  
ax = plt.axes(projection='3d')  
ax.contour3D(X, Y, Z, 50, cmap='binary')  
ax.set_xlabel('x')  
ax.set_ylabel('y')  
ax.set_zlabel('z');  
plt.show();
```



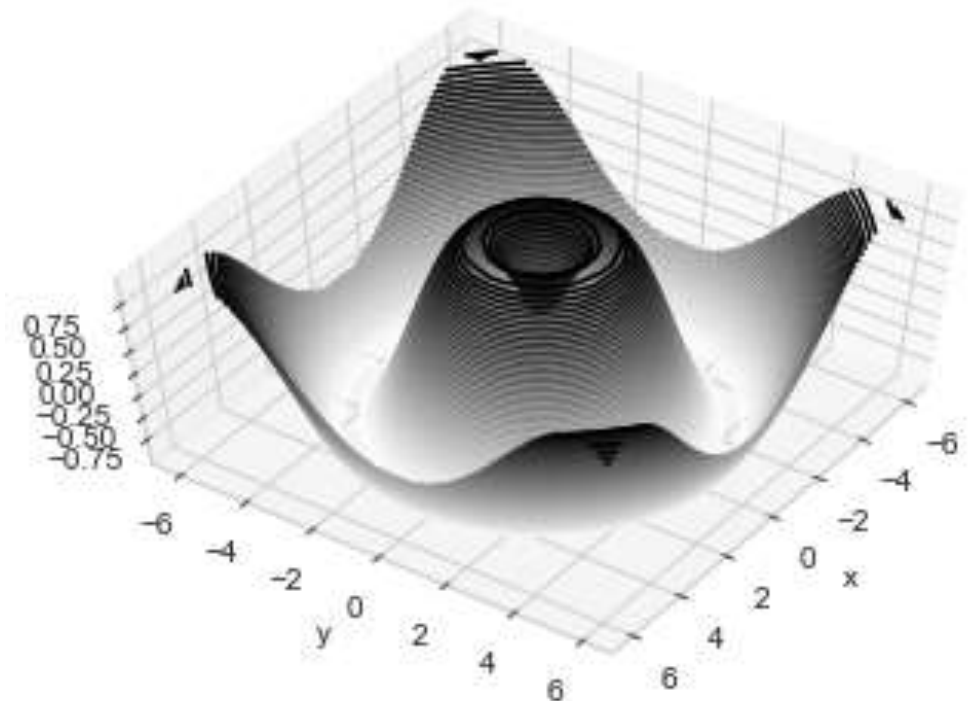
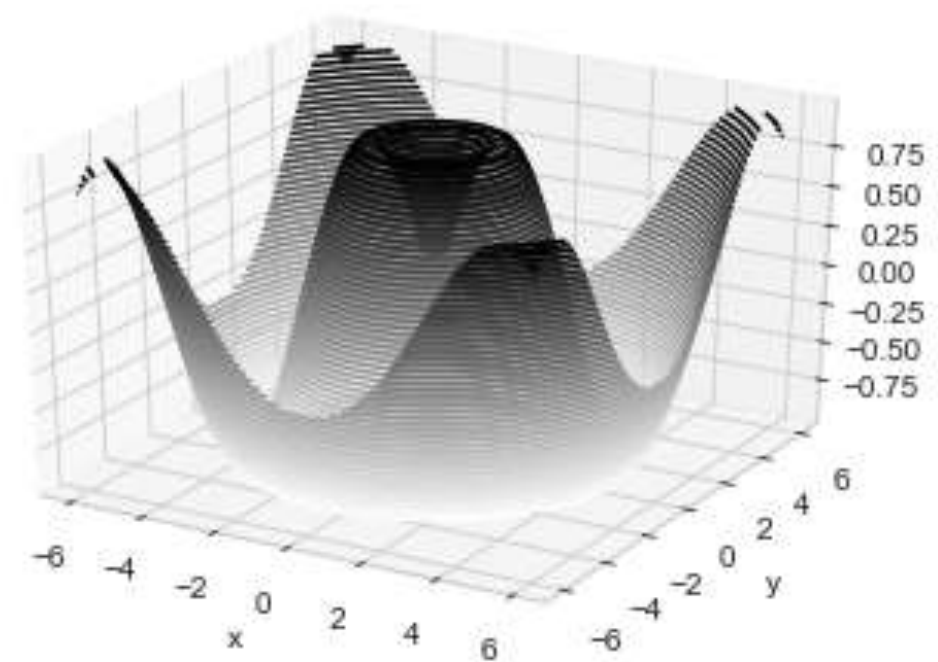


3D contours

view_init method to set the elevation and azimuthal angles

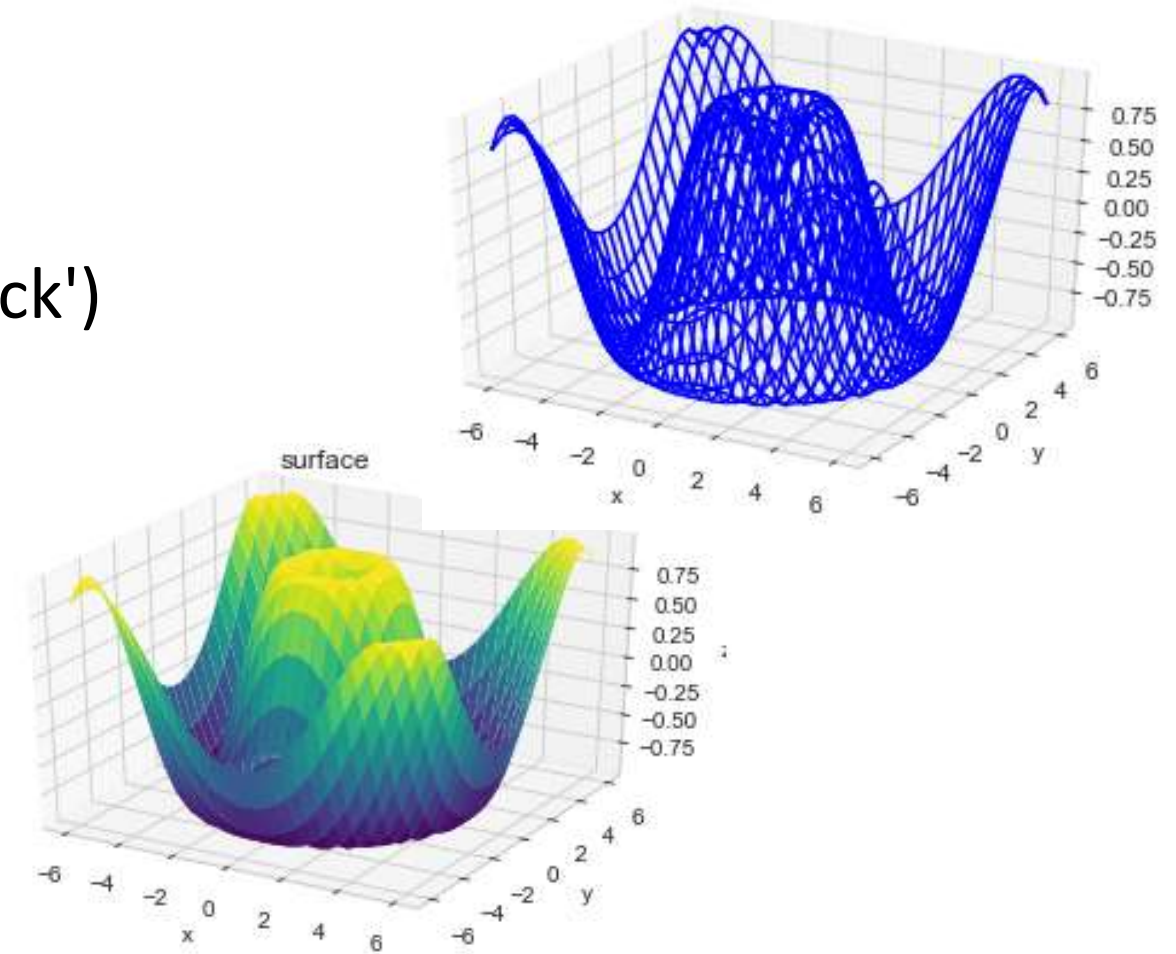
```
ax.view_init(60, 35)
```

elevation of 60 degrees (that is, 60 degrees above the x-y plane) and an azimuth of 35 degrees (that is, rotated 35 degrees counter-clockwise about the z-axis)





- `ax.plot_wireframe(X, Y, Z, color='black')`
- `ax.plot_surface(X, Y, Z, rstride=1, cstride=1, cmap='viridis', edgecolor='none')`
- `ax.set_title('surface')`

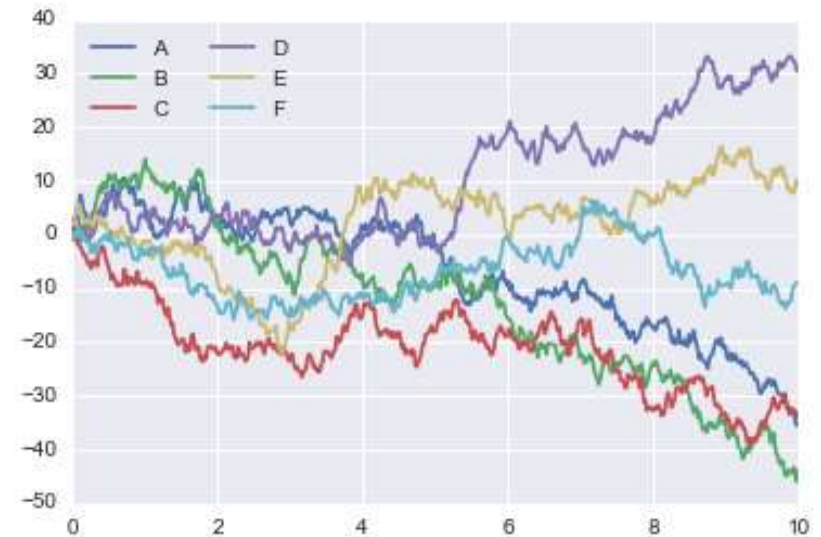
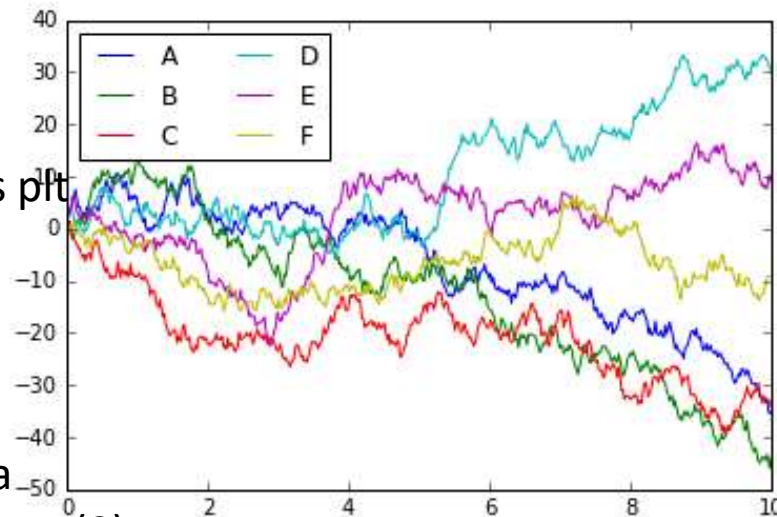




Visualization with Seaborn

- Seaborn provides an API on top of Matplotlib that offers sane choices for plot style and color defaults

```
import matplotlib.pyplot as plt
plt.style.use('classic')
%matplotlib inline
import numpy as np
import pandas as pd
# Create some random data
rng = np.random.RandomState(0)
x = np.linspace(0, 10, 500)
y = np.cumsum(rng.randn(500, 6), 0)
# Plot the data with Matplotlib defaults
plt.plot(x, y)
plt.legend('ABCDEF', ncol=2, loc='upper left');
plt.show();
```



```
import seaborn as sns
sns.set()
plt.plot(x, y)
plt.legend('ABCDEF', ncol=2, loc='upper left');
plt.show();
```