

# Computational Environments and Toolchains

## Topic 02 — Scientific Computing using Python

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### Lecture 01 — Matplotlib

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#### Outline

- matplotlib (2D and 3D plotting library)
- numpy (high performance matrix library)
- scipy (scientific computation library)

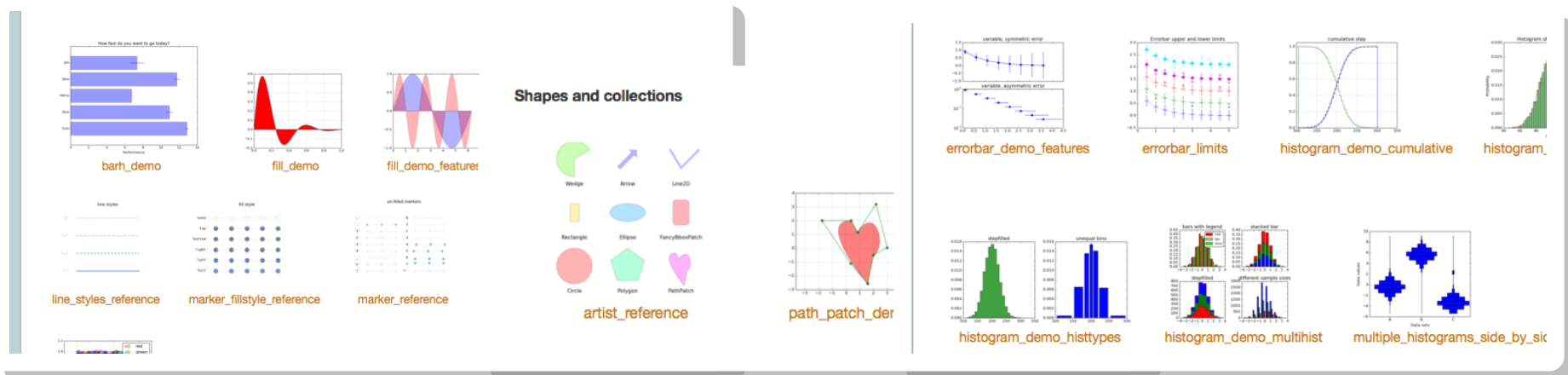
# Outline

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# A 20 Second Intro to Matplotlib

- Matplotlib is the standard plotting library for scientific python
- Design objectives
  - Plots should look great – publication quality.
  - Supports formats suitable for inclusion with T<sub>E</sub>X documents.
  - Embeddable in a graphical user interface for application development.
  - Code should be easy enough to understand and extendable.
- Mostly it is for 2D data (including surface plots of  $f(x, y)$ , etc.)
- Active development with lots of new features
- Best way to figure out how to do something: look at the gallery  
<http://matplotlib.org/gallery.html>



# Importing Matplotlib

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- There are several interfaces to matplotlib that provide varying amounts of access to its underlying functionality

`http://matplotlib.org/faq/usage\_faq.html`

- `matplotlib` is the entire package.
- `pyplot` is a module within matplotlib that provides easy access to the core plotting routines.
- `pylab` combines `pyplot` and `numpy` into a single namespace to give a MatLab like interface
  - This is recommended for interactive work but I tend to just import both `pylab` and `numpy` separately.
- A number of toolkits extend the functionality
  - `basemap` and `cartopy`: mapping (e.g. projecting onto a globe, geographical boundaries)
  - `mplot3d`: basic 3-d plotting.
  - `AxesGrid`: high-level methods for arranging multiple plots together in a figure

# Mathplotlib Backends

`matplotlib` can output to a number of different devices—the backends provide this functionality:

- Interactive backends (allow for plotting to the screen, and updates (animations) with each command (if desired)):
  - `pygtk`, `wxpython`, `tkinter`, `qt`, `macosx`
- Hardcopy backends (saving to external image files):
  - PNG, SVG, PDF, PS

To select a backend:

```
1 import matplotlib
2 matplotlib.use("PS")
3 import matplotlib.pyplot as plt
```

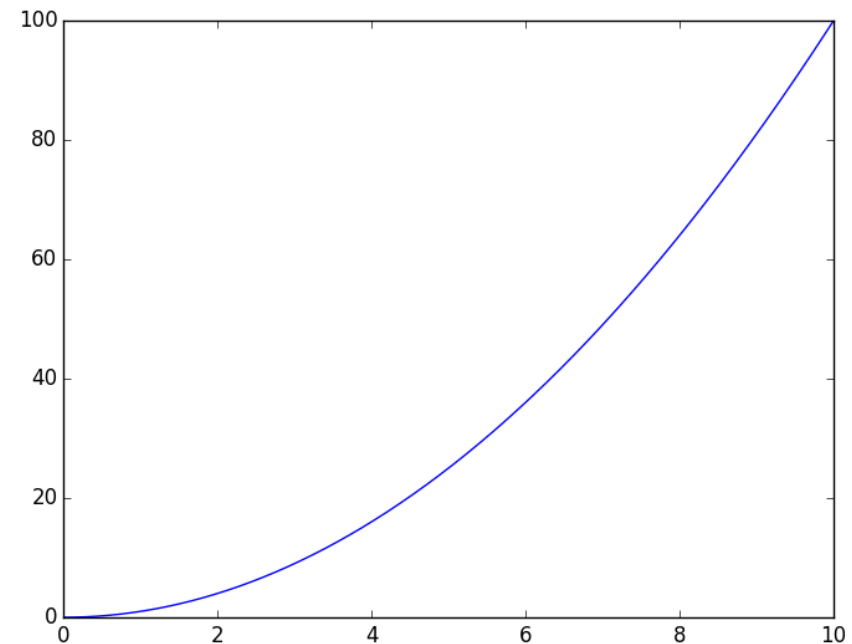
backend.py

In iPython, use the `%matplotlib inline` to get images embedded in the notebooks.

# Line Plot (Basic)

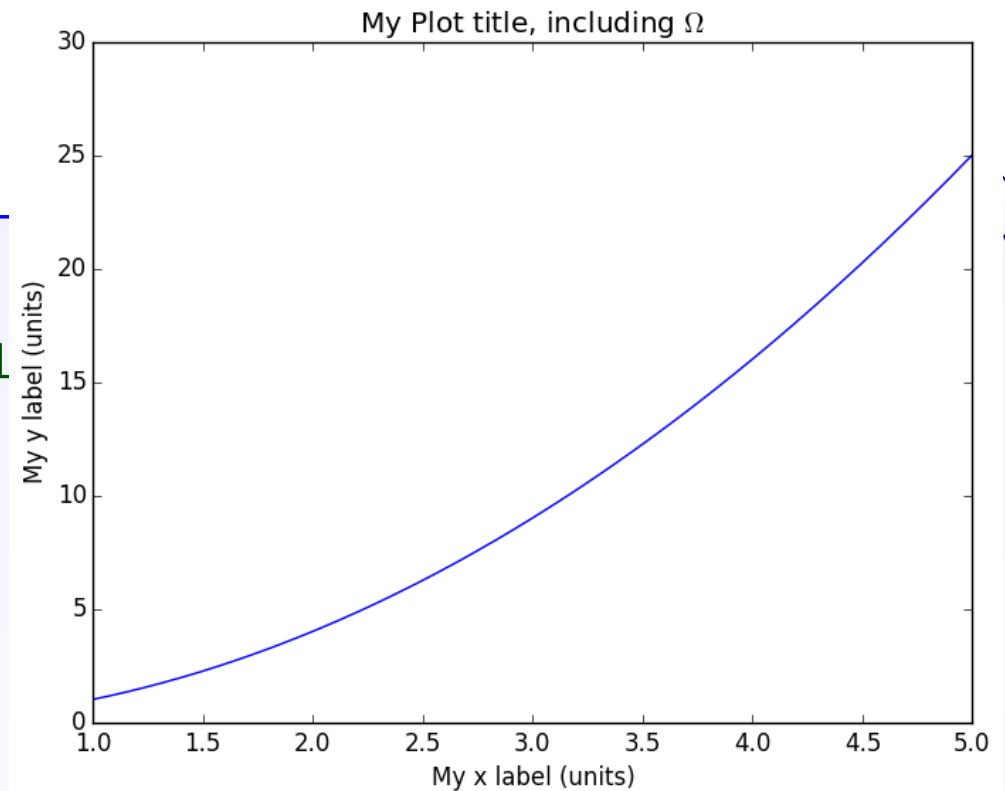
`line_plot .py`

```
5 import numpy as np
6 import matplotlib.pyplot as plt
7
8 x = np.linspace(0, 10, 1000)      # evenly spaced points
9 y = np.power(x, 2)                # element-wise squaring
10
11 plt.plot(x, y)
12
13 plt.savefig("line_plot.png")
```

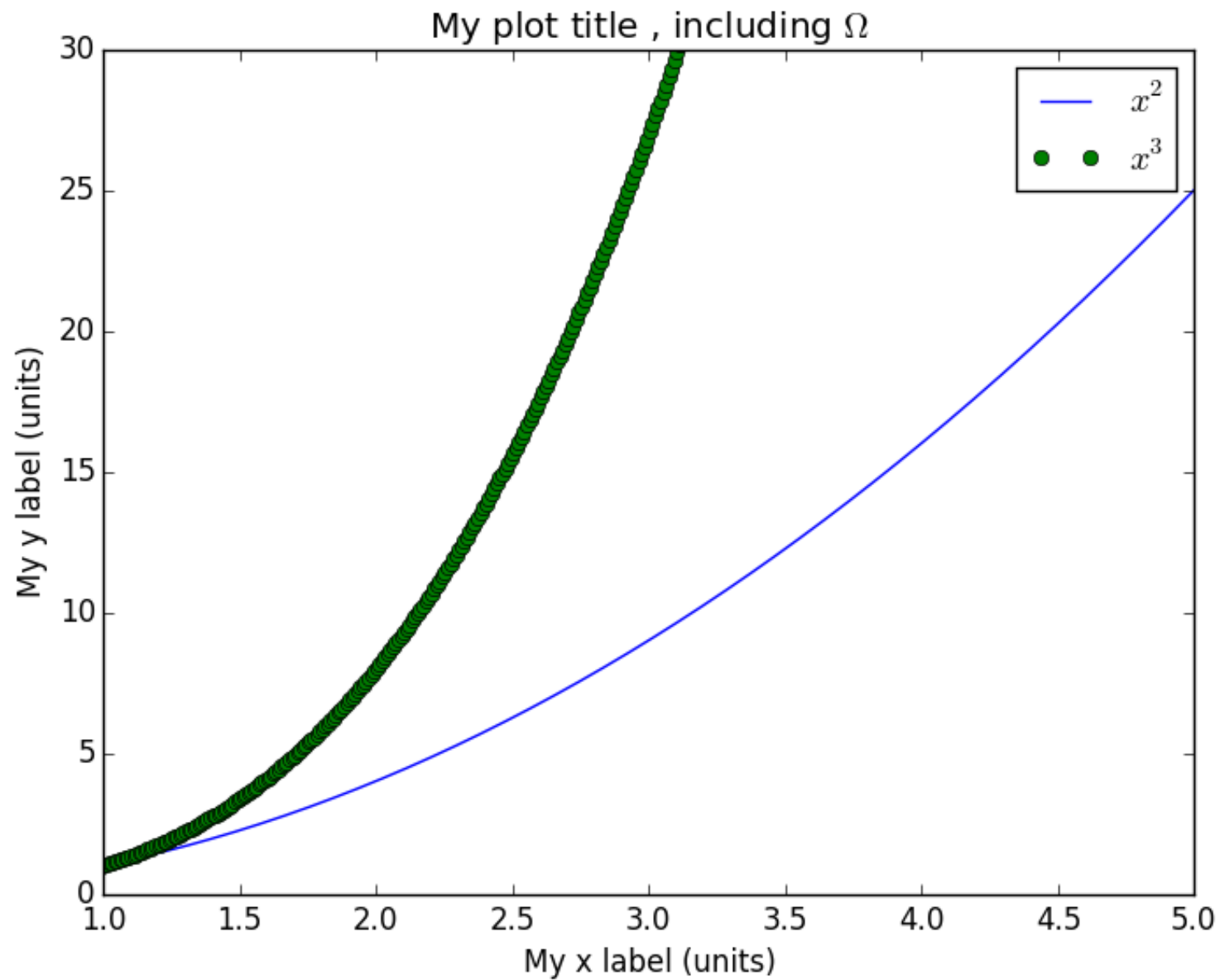


# Line Plot (Proper)

```
5 import numpy as np
6 import matplotlib.pyplot as plt
7
8 x = np.linspace(0, 10, 1000)
9 y = np.power(x, 2)
10
11 plt.plot(x, y)
12 plt.xlim((1, 5))
13 plt.ylim((0, 30))
14 plt.xlabel("My_x_label_(units)")
15 plt.ylabel("My_y_label_(units)")
16 plt.title("My_Plot_title,_including_\\Omega")
17
18 plt.savefig("line_plot_proper.png", bbox_inches="tight")
```



# Line Plot (Fancy)



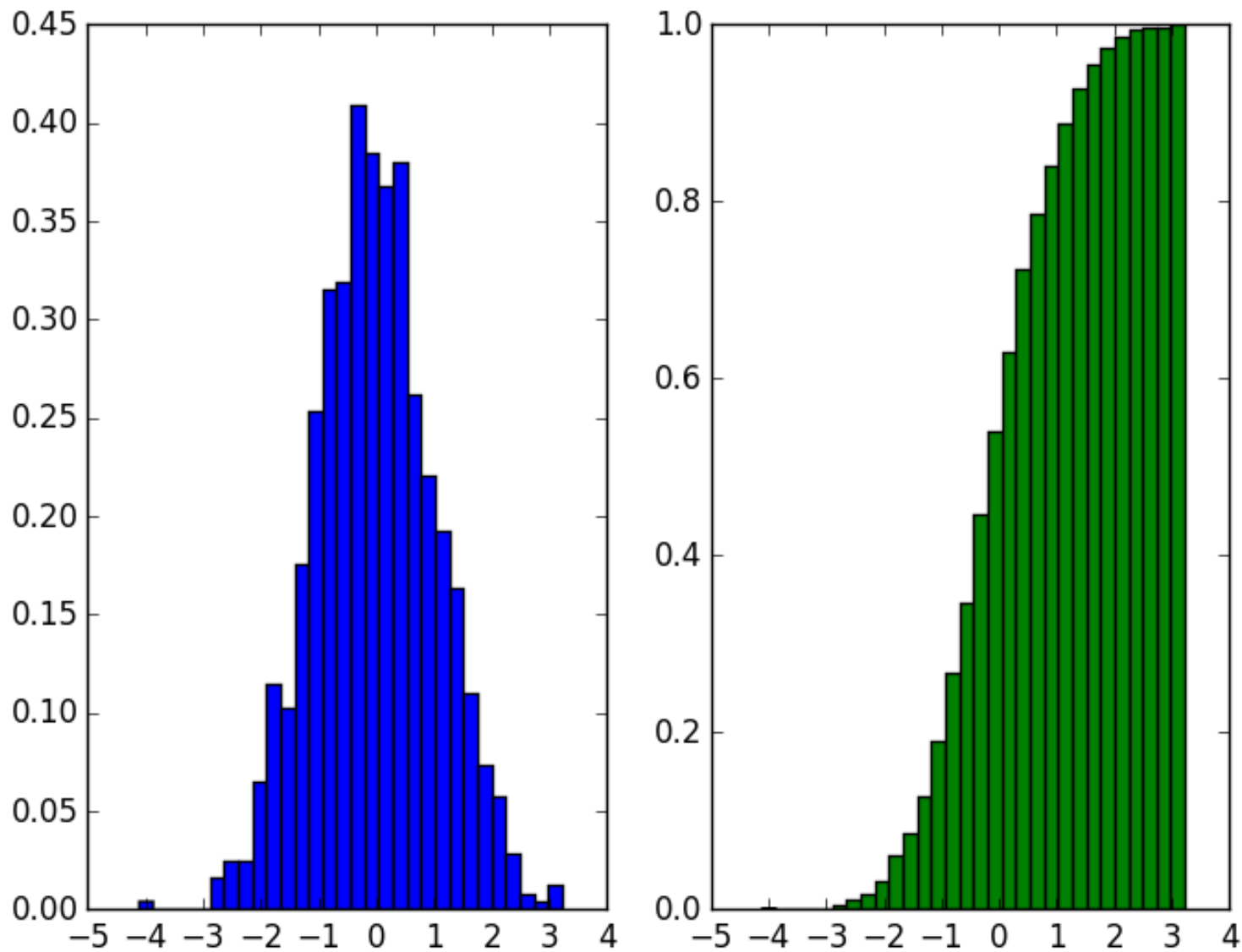


# Line Plot (Fancy) — Code

`line_plot_fancy .py`

```
5 import numpy as np
6 import matplotlib.pyplot as plt
7
8 x = np.linspace(0, 10, 1000)
9 y1 = np.power(x, 2)
10 y2 = np.power(x, 3)
11
12 plt.plot(x, y1, "b-", x, y2, "go")           # look at me !
13 plt.xlim((1, 5))
14 plt.ylim((0, 30))
15 plt.xlabel("My_x_label_(units)")
16 plt.ylabel("My_y_label_(units)")
17 plt.title("My_plot_title_,_including_\Omega$")
18 plt.legend(("x^2$", "x^3$"))                 # and me !
19
20 plt.savefig("line_plot_fancy.png", bbox_inches="tight")
```

# Histogram

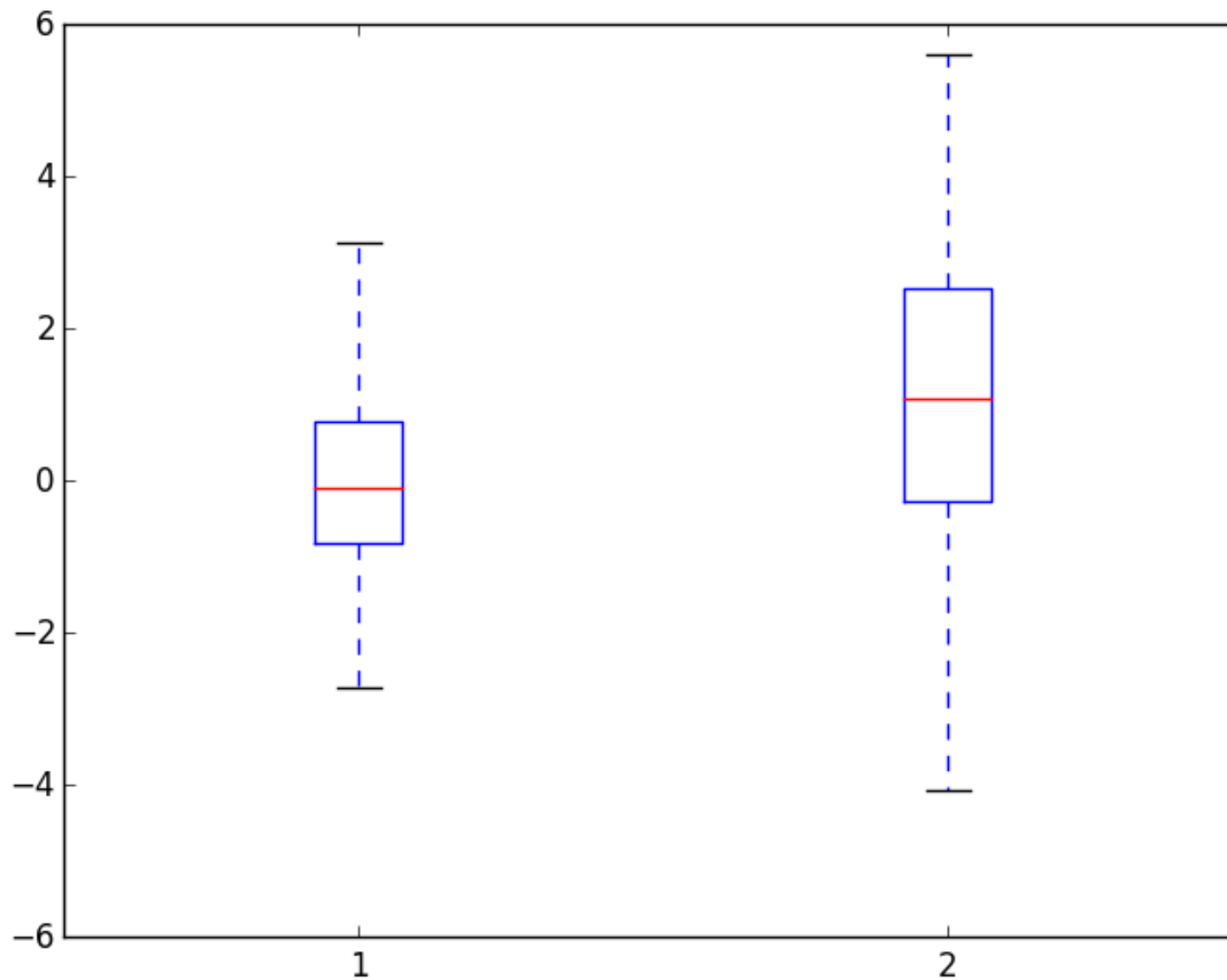


# Histogram — Code

histogram.py

```
5 import numpy as np
6 import matplotlib.pyplot as plt
7
8 data = np.random.randn(1000)
9
10 # histogram (pdf)
11 plt.subplot(1, 2, 1)
12 plt.hist(data, bins=30, normed=True, facecolor="b")
13
14 # empirical cdf
15 plt.subplot(1, 2, 2)
16 plt.hist(data, bins=30, normed=True, color="g", cumulative=True)
17
18 plt.savefig("histogram.png", bbox_inches="tight")
```

# Box Plot



# Box Plot — Code

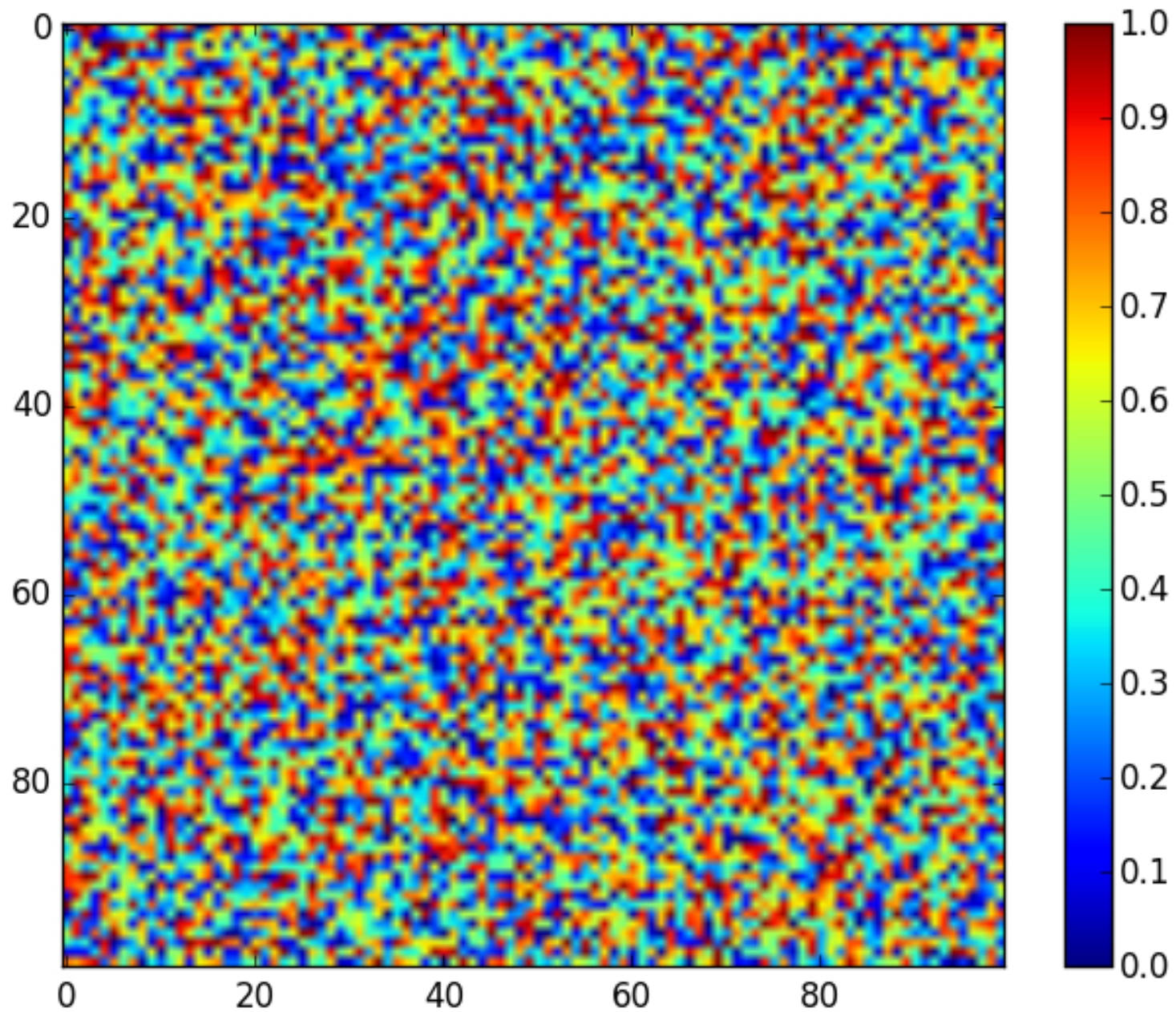
box\_plot.py

```
5 import numpy as np
6 import matplotlib.pyplot as plt
7
8 samp1 = np.random.normal(loc=0., scale=1., size=100)
9 samp2 = np.random.normal(loc=1., scale=2., size=100)
10 plt.boxplot((samp1, samp2))
11
12 plt.savefig("box_plot.png", bbox_inches="tight")
```

In descriptive statistics, a box plot is a convenient way of graphically depicting groups of numerical data through their quartiles — min, first quartile, median (second quartile), third quartile, and max.

[https://en.wikipedia.org/wiki/Box\\_plot](https://en.wikipedia.org/wiki/Box_plot)

# Image Plot

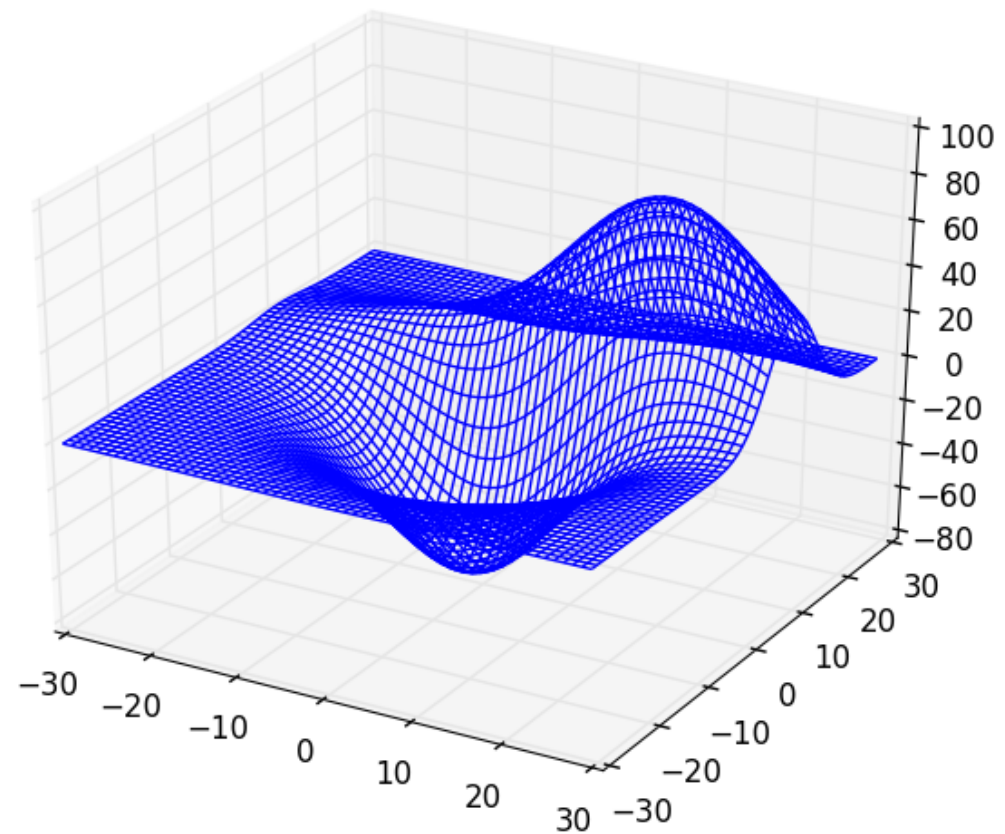


# Image Plot — Code

image\_plot.py

```
5 import numpy as np
6 import matplotlib.pyplot as plt
7
8 A = np.random.random((100, 100))
9 plt.imshow(A)           # Display an image on the axes
10 plt.hot()               # set colormap to hot
11 plt.colorbar()          # Add a colorbar to a plot
12
13 plt.savefig("image_plot.png", bbox_inches="tight")
```

# Wire Plot



```
5 from mpl_toolkits.mplot3d import axes3d
6 import matplotlib.pyplot as plt
7
8 ax = plt.subplot(111, projection="3d")
9
10 X, Y, Z = axes3d.get_test_data(0.1)
11 ax.plot_wireframe(X, Y, Z)
12
13 plt.savefig("wire_plot.png", bbox_inches="tight")
```