

DM561  
Linear Algebra with Applications

## Intoduction to Python - Part 3

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*[Based on booklet Python Essentials]*

# Outline

Matplotlib  
Other Data Visualization Libraries  
Pandas

1. Matplotlib

2. Other Data Visualization Libraries

3. Pandas

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**Matplotlib**  
Other Data Visualization Libraries  
Pandas

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2. Other Data Visualization Libraries

3. Pandas

# Matplotlib Library

Matplotlib is a low level graph plotting library in Python that serves as a visualization utility. Most of the Matplotlib utilities lies under the `pypplot` submodule, and are usually imported under the `plt` alias

```
>>> %matplotlib inline  
>>> import matplotlib.pyplot as plt
```

In IPython, Jupyter and Jupyter Lab, functions with `%` are extra functions of IPython that add functionalities to the environment. They are called magic functions.

- `%matplotlib inline` shows the plot
- `%matplotlib notebook` shows the plot and provides controls to interact with the plot

Other useful magic function are:

- `%timeit` to determine running time of a command and
- `%run` to run a script from a file.

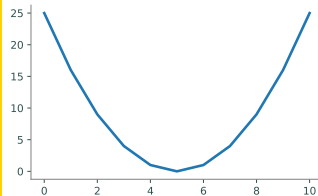
# Line Plots

```
>>> import numpy as np
>>> from matplotlib import pyplot as plt

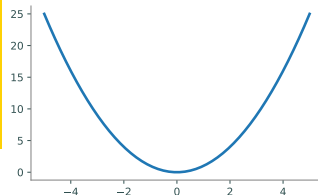
>>> y = np.arange(-5,6)**2
>>> y
array([25, 16, 9, 4, 1, 0, 1, 4, 9, 16, 25])
```

```
# Visualize the plot.
```

```
>>> plt.plot(y) # default values x-values 0, 1, 2, 3,
[<matplotlib.lines.Line2D object at 0x10842d0>]
>>> plt.show() # Reveal the resulting plot.
```



```
>>> x = np.linspace(-5, 5, 50)
>>> y = x**2 # Calculate the range of f(x) = x**2.
>>> plt.plot(x,y)
>>> plt.show()
```



# Plotting a Polynomial Function

Let's plot the following polynomial of degree 3:

$$P_3(x) = x^3 - 7x + 6 = (x - 1)(x - 2)(x + 3)$$

The numpy function `numpy.poly1d` takes an array of coefficients of length `n+1` (try `numpy.polyfit?`):

`a[0] * x**n + a[1] * x**(n-1) + ... + a[n-1]*x + a[n]`

```
>>> import numpy as np
>>> a=[1,0,-7,6]
>>> P=np.poly1d(a)
>>> print(P)
      3
  1 x - 7 x + 6

>>> x = np.linspace(-3.5, 3.5, 500)
>>> plt.plot(x, P(x), '-')
>>> plt.axhline(y=0)
>>> plt.title('A polynomial of order 3');
```

# Interactive Plotting

In non-interactive mode (default):

- newly created figures and changes to figures will not be reflected until explicitly asked to be;
- `.pyplot.show` will block by default.

The interactive mode is mainly useful to build plots from the command line and see the effect of each command while building the figure.

In interactive mode:

- newly created figures will be shown immediately;
- figures will automatically redraw on change;
- `.pyplot.show` will not block by default.

<code>plt.ion()</code>	Enable interactive mode
<code>plt.clf()</code>	Clear figure
<code>plt.ioff()</code>	Disable interactive mode
<code>plt.show()</code>	Show all figures (and maybe block)
<code>plt.pause()</code>	Show all figures, and block for a time.

# Plot Customization

For `plt.plot()`

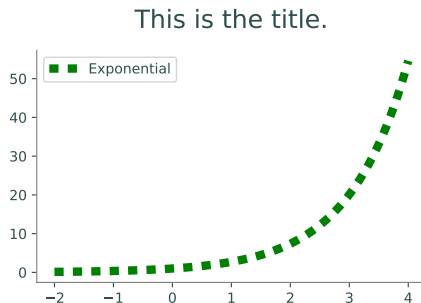
marker	Style	linestyle	Style	color	Color
'.'	point marker	'-'	solid line	'b'	blue
'o'	circle marker	'--'	dashed line	'g'	green
'*'	star marker	'-.'	dash-dot line	'r'	red
'+'	plus marker	':'	dotted line	'c'	cyan
...	...	...	...	'k'	black

The parameter `fmt` is written with this syntax `marker|line|color`  
`markersize` (or `ms`) to set the size of the markers. Other functions

Function	Description
<code>legend()</code>	Place a legend in the plot
<code>title()</code>	Add a title to the plot
<code>xlim()</code> / <code>ylim()</code>	Set the limits of the <code>x</code> - or <code>y</code> -axis
<code>xlabel()</code> / <code>ylabel()</code>	Add a label to the <code>x</code> - or <code>y</code> -axis
<code>grid()</code>	Add the grid lines



```
>>> x1 = np.linspace(-2, 4, 100)
>>> plt.plot(x1, np.exp(x1), 'g:', ↪
↪ linewidth=6, label="Exponential"↪
↪)
>>> plt.title("My title", fontsize=18)
>>> plt.legend(loc="upper left")
>>> plt.show()
```



Function	Description
<code>figure()</code>	Create a new figure or grab an existing figure
<code>axes()</code>	Add an axes to the current figure
<code>gca()</code>	Get the current axes
<code>gcf()</code>	Get the current figure
<code>subplot()</code>	Add a single subplot to the current figure
<code>subplots()</code>	Create a figure and add several subplots to it

`subplot` takes three arguments: the layout is organized in rows and columns, which are represented by the first and second argument. The third argument represents the index of the current plot.

```
# 3. Use plt.subplots() to get the figure and all subplots simultaneously.  
>>> fig, axes = plt.subplots(1, 2)  
>>> axes[0].plot(x, 2*x)  
>>> axes[1].plot(x, x**2)
```

Compare `axes()` vs `axis()` (access properties of the current plot)

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

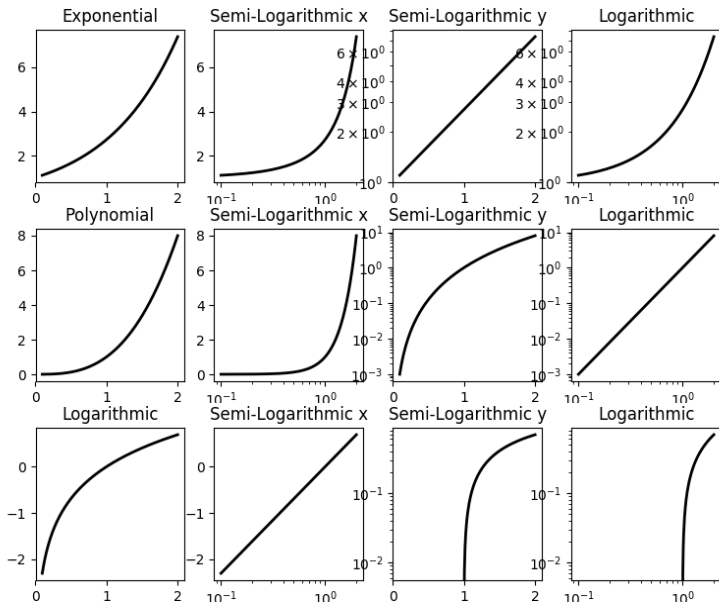
def make_figure(x,f,name):
    plt.figure(figsize=(9,2))
    ax1 = plt.subplot(141)
    ax1.plot(x, f(x), 'k', lw=2)
    plt.title(name)

    ax2 = plt.subplot(1,4,2)
    ax2.semilogx(x, f(x), 'k', lw=2)
    ax2.set_title("Semi-Logarithmic x")

    ax3 = plt.subplot(143)
    ax3.semilogy(x, f(x), 'k', lw=2)
    ax3.set_title("Semi-Logarithmic y")

    ax4 = plt.subplot(144)
    ax4.loglog(x, f(x), 'k', lw=2)
    ax4.set_title("Logarithmic")
    plt.savefig(name+".png")
```

```
xx = np.linspace(.1, 2, 200)
make_figure(xx, lambda xx: np.exp(xx), ↪
            ↪"Exponential")
make_figure(xx, lambda xx: xx**3, "↪
            ↪Polynomial")
make_figure(xx, lambda xx: np.log(xx), ↪
            ↪"Logarithmic")
```

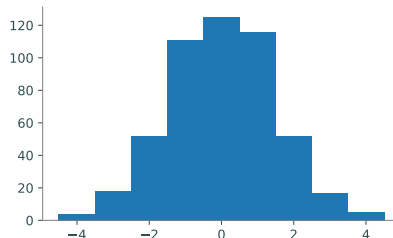
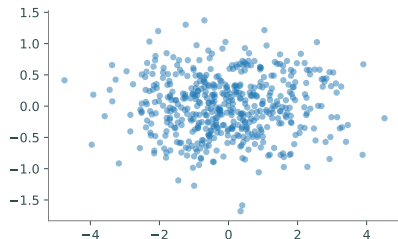


# Scatter Plots and Histograms

```
>>> x = np.random.normal(scale=1.5, size=500)
>>> y = np.random.normal(scale=0.5, size=500)

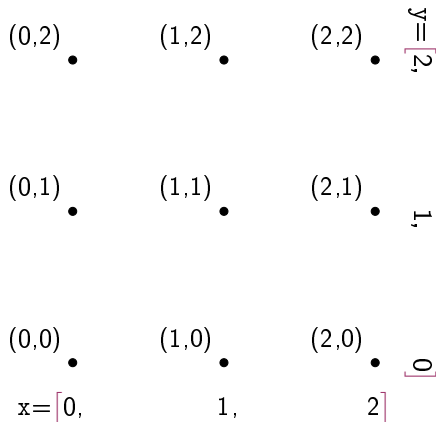
>>> ax1 = plt.subplot(121)
>>> ax1.plot(x, y, 'o', markersize=5, alpha=.5) # transparent circles

>>> ax2 = plt.subplot(122)
>>> ax2.hist(x, bins=np.arange(-4.5, 5.5))
>>> plt.show()
```



# 3D Surfaces

`np.meshgrid()` given two 1-dimensional coordinate arrays, creates two corresponding coordinate matrices:  $(X[i,j], Y[i,j]) = (x[i], y[j])$ .



$$X = \begin{bmatrix} 0 & 1 & 2 \\ 0 & 1 & 2 \\ 0 & 1 & 2 \end{bmatrix}$$
$$Y = \begin{bmatrix} 2 & 2 & 2 \\ 1 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

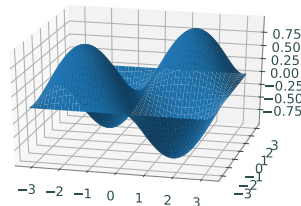
```
>>> x, y = [0, 1, 2], [3, 4, 5]      # A rough domain over [0,2]x[3,5].
>>> X, Y = np.meshgrid(x, y)        # Combine the 1-D data into 2-D data.
>>> for trows in zip(X,Y):
...     print(trows)
...
(array([0 1 2]), array([3 3 3]))
(array([0 1 2]), array([4 4 4]))
(array([0 1 2]), array([5 5 5]))
```

# 3D Surfaces

$$g(x, y) = \sin(x) \sin(y)$$

```
>>> x = np.linspace(-np.pi, np.pi, 200)
>>> y = np.copy(x)
>>> X, Y = np.meshgrid(x, y)
>>> Z = np.sin(X) * np.sin(Y)

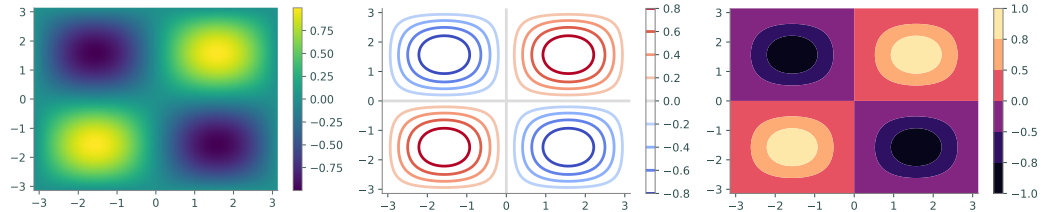
# Draw the corresponding 3-D plot using some →
  →extra tools.
>>> from mpl_toolkits.mplot3d import Axes3D
>>> fig = plt.figure()
>>> ax = fig.add_subplot(1,1,1, projection='3d')
>>> ax.plot_surface(X, Y, Z)
>>> plt.show()
```





# Heat Map and Contour Plot

```
>>> x = np.linspace(-np.pi, np.pi, 100)
>>> y = x.copy()
>>> X, Y = np.meshgrid(x, y)
>>> Z = np.sin(X) * np.sin(Y)          # Calculate  $g(x,y) = \sin(x)\sin(y)$ .
# Plot the heat map of f over the 2-D domain.
>>> plt.subplot(131)
>>> plt.pcolormesh(X, Y, Z, cmap="viridis")
>>> plt.colorbar()
>>> plt.xlim(-np.pi, np.pi)
>>> plt.ylim(-np.pi, np.pi)
# Plot a contour map of f with 10 level curves.
>>> plt.subplot(132)
>>> plt.contour(X, Y, Z, 10, cmap="coolwarm")
>>> plt.colorbar()
# Plot a filled contour map, specifying the level curves.
>>> plt.subplot(133)
>>> plt.contourf(X, Y, Z, [-1, -.8, -.5, 0, .5, .8, 1], cmap="magma")
>>> plt.colorbar()
>>> plt.show()
```



1. Calculate all data that is needed for the animation.
2. Define a figure explicitly with `plt.figure()` and set its window boundaries.
3. Draw empty objects that can be altered dynamically.
4. Define a function to update the drawing objects.
5. Use `matplotlib.animation.FuncAnimation()`.

```
from matplotlib.animation import FuncAnimation
from mpl_toolkits.mplot3d import Axes3D

def sine_animation():
    # 1. Calculate the data to be animated.
    x = np.linspace(0, 2*np.pi, 200)[::-1]
    y = np.sin(x)      #
    # 2. Create a figure and set the window boundaries of the axes.
    fig = plt.figure()
    plt.xlim(0, 2*np.pi)
    plt.ylim(-1.2, 1.2) #
    # 3. Draw an empty line. The comma after 'drawing' is crucial.
    drawing, = plt.plot([],[]) #
    # 4. Define a function that updates the line data.
    def update(index):
        drawing.set_data(x[:index], y[:index])
        return drawing,          # Note the comma!
    # 5.
    a = FuncAnimation(fig, update, frames=len(x), interval=10)
```

# Further Reading and Tutorials

- [https://www.w3schools.com/python/matplotlib\\_intro.asp](https://www.w3schools.com/python/matplotlib_intro.asp)
- <https://www.labri.fr/perso/nrougier/teaching/matplotlib/>.
- [https://matplotlib.org/users/pyplot\\_tutorial.html](https://matplotlib.org/users/pyplot_tutorial.html).
- <http://www.scipy-lectures.org/intro/matplotlib/matplotlib.html>.
- <https://matplotlib.org/2.0.0/examples/animation/index.html>

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Matplotlib  
**Other Data Visualization Libraries**  
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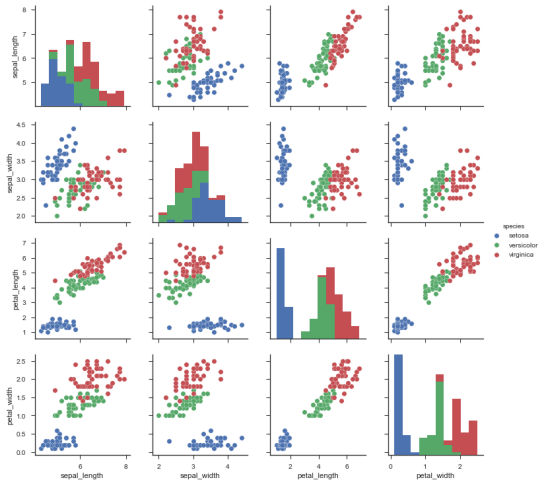
1. Matplotlib

2. Other Data Visualization Libraries

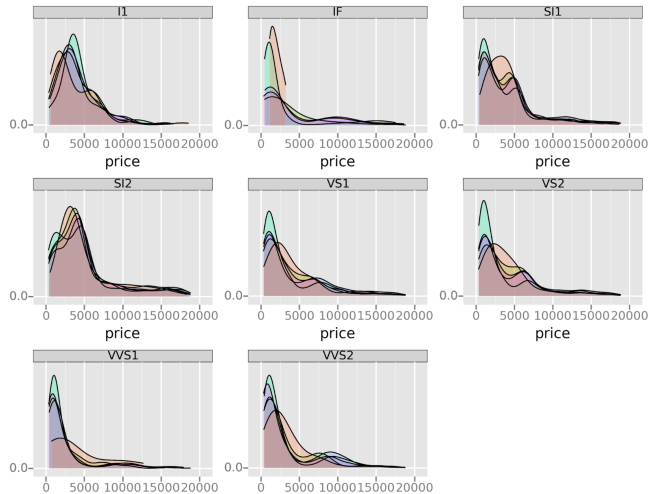
3. Pandas

# Seaborn

A high-level library on the top of Matplotlib. It's easier to generate certain kinds of plots: eg, heat maps, time series, and violin plots.



Based on R's ggplot2 and the Grammar of Graphics



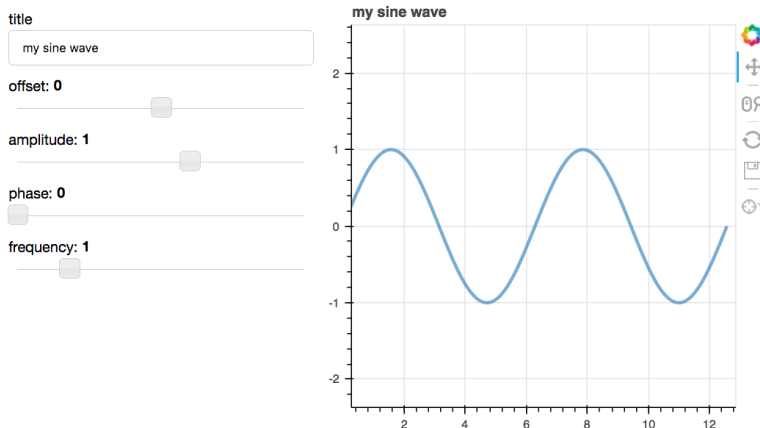


# Bokeh, Plotly, Gleam, Dash and Altair

Create interactive, web-ready plots, as JSON objects, HTML documents, or interactive web applications.

Bokeh is based on the Grammar of Graphics like ggplot.

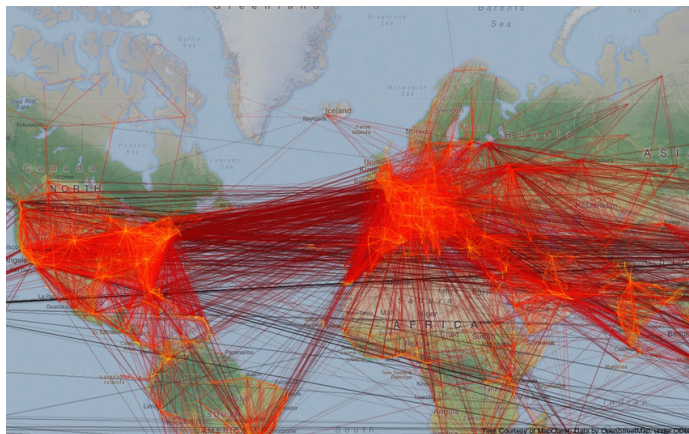
Gleam is inspired by R's Shiny package.



# Geoplotlib, Leaflet and MapBox

Matplotlib  
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Toolbox for plotting geographical data: map-type plots, like choropleths, heatmaps, and dot density maps.



# Graph Algorithms, Graph Drawings

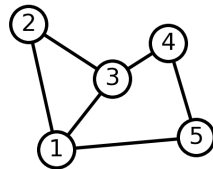
```
import matplotlib.pyplot as plt

G = nx.Graph()
G.add_edge(1, 2)
G.add_edge(1, 3)
G.add_edge(1, 5)
G.add_edge(2, 3)
G.add_edge(3, 4)
G.add_edge(4, 5)

# explicitly set positions
pos = {1: (0, 0), 2: (-1, 0.3), 3: (2, 0.17), 4: (4, 0.255), 5: (5, 0.03)}

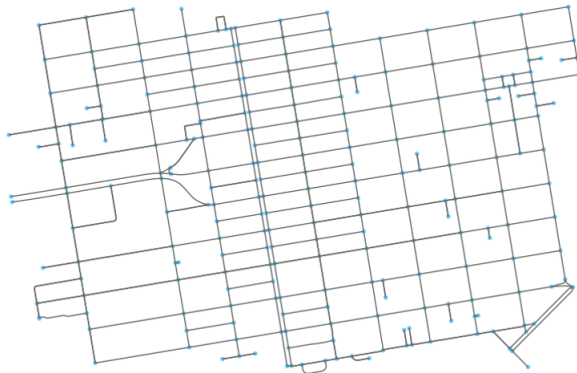
nx.draw_networkx(G, pos)

# Set margins for the axes so that nodes aren't clipped
ax = plt.gca()
ax.margins(0.20)
plt.axis("off")
plt.show()
```



# Street Networks

```
import osmnx as ox
G = ox.graph_from_bbox(37.79, 37.78, -122.41, -122.43, network_type='drive')
G_projected = ox.project_graph(G)
ox.plot_graph(G_projected)
```



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# Pandas Data Structures: Series

**Pandas** library for data management and analysis that combines functionality of NumPy, Matplotlib, and SQL

- **Series** is a one-dimensional array that can hold any datatype, similar to a `ndarray` but with an `index` that gives a label to each entry.

```
>>> import pandas as pd
>>>
>>> # Initialize Series of student grades
>>> math = pd.Series(np.random.randint(0,100,4), ['Mark', 'Barbara', 'Eleanor', ↵
↵ 'David'])
>>> english = pd.Series(np.random.randint(0,100,5), ['Mark', 'Barbara', 'David', ↵
↵ 'Greg', 'Lauren'])
>>> math
Mark      30
Barbara   71
Eleanor   94
David     41
dtype: int64
```

# Pandas Data Structures: Data Frames

- `DataFrame` is a collection of multiple `Series`. It can be thought of as a 2-dimensional array, where each row is a separate datapoint and each column is a feature of the data. The rows are labelled with an `index` (as in a `Series`) and the columns are labelled in the attribute `columns`.

```
>>> # Create a DataFrame of student grades
>>> grades = pd.DataFrame({"Math": math, "English": english})
>>> grades
```

	Math	English
Barbara	52.0	73.0
David	10.0	39.0
Eleanor	35.0	NaN
Greg	NaN	26.0
Lauren	NaN	99.0
Mark	81.0	68.0

# Summary

1. Matplotlib

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