

# Object Oriented Programming (Using Python)

## UNIT- III

### Python Libraries:

- Matplotlib cont'd.
  - Coding Styles
  - Styling Artists
  - Labelling plots
  - Axis scales and ticks
  - Color mapped data

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# matplotlib - Quick start guide

[https://matplotlib.org/stable/tutorials/introductory/quick\\_start.html](https://matplotlib.org/stable/tutorials/introductory/quick_start.html)

# Coding Styles

## The explicit and the implicit interfaces

*There are essentially two ways to use `Matplotlib`:*

- Explicitly create `Figures` and `Axes`, and call methods on them (the "object-oriented (OO) style").
- Rely on `pypplot` to implicitly create and manage the `Figures` and `Axes`, and use `pypplot` functions for plotting.

# Coding Styles

## The explicit and the implicit interfaces

Example of **Object-oriented (OO)** style:

```
>>> import matplotlib.pyplot as plt
# Note that even in the OO-style, we use `.pyplot.figure` to
create the Figure.
>>> fig, ax = plt.subplots(figsize=(5, 2.7),
layout='constrained')
>>> ax.plot(x, x, label='linear') # Plot some data on the
axes.
[<matplotlib.lines.Line2D object at 0x000001417FE6F5B0>]
```

# Coding Styles

## The explicit and the implicit interfaces

Example of **Object-oriented (OO) style**: cont'd.

```
>>> ax.plot(x, x**2, label='quadratic') # Plot more data on
the axes...
[<matplotlib.lines.Line2D object at 0x000001417FE6F880>]
>>> ax.plot(x, x**3, label='cubic') # ... and some more.
[<matplotlib.lines.Line2D object at 0x000001417FE6FB20>]
>>> ax.set_xlabel('x label') # Add an x-label to the axes.
Text(0.5, 0, 'x label')
>>> ax.set_ylabel('y label') # Add a y-label to the axes.
Text(0, 0.5, 'y label')
```

# Coding Styles

## The explicit and the implicit interfaces

Example of **Object-oriented (OO) style**: cont'd.

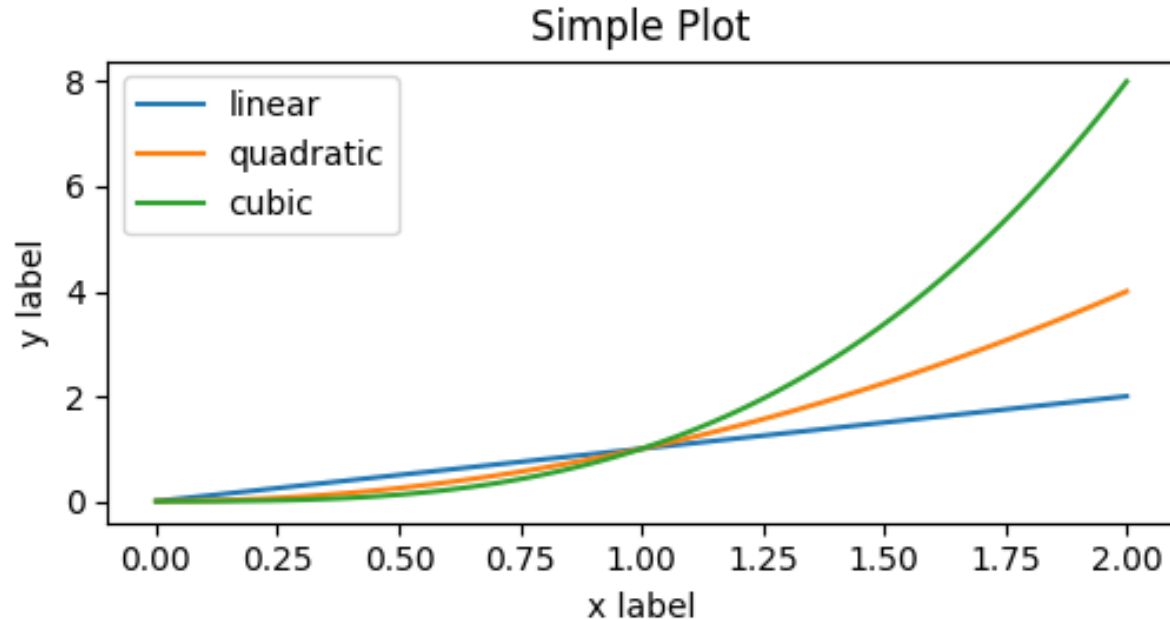
```
>>> ax.set_title("Simple Plot")    # Add a title to the axes.
Text(0.5, 1.0, 'Simple Plot')
>>> ax.legend()    # Add a legend.
<matplotlib.legend.Legend object at 0x000001417FE6FB50>
```

# Coding Styles

## The explicit and the implicit interfaces

Example of **Object-oriented (OO) style**: cont'd.

```
>>> plt.show()
```



# Coding Styles

## The explicit and the implicit interfaces

Example of **pyplot-style**:

```
>>> import numpy as np
>>> import matplotlib.pyplot as plt
>>> x = np.linspace(0, 2, 100) # Sample data.
>>> plt.figure(figsize=(5, 2.7), layout='constrained')
<Figure size 500x270 with 0 Axes>
>>> plt.plot(x, x, label='linear') # Plot some data on
the (implicit) axes.
[<matplotlib.lines.Line2D object at 0x000001410000BA60>]
```



# Coding Styles

## The explicit and the implicit interfaces

Example of **pyplot-style**: cont'd.

```
>>> plt.plot(x, x**2, label='quadratic') # etc.
[<matplotlib.lines.Line2D object at 0x000001410000BD30>]
>>> plt.plot(x, x**3, label='cubic')
[<matplotlib.lines.Line2D object at 0x000001410000BFD0>]
>>> plt.xlabel('x label')
Text(0.5, 0, 'x label')
>>> plt.ylabel('y label')
Text(0, 0.5, 'y label')
```

# Coding Styles

## The explicit and the implicit interfaces

Example of **pyplot-style**: cont'd.

```
>>> plt.title("Simple Plot")
```

```
Text(0.5, 1.0, 'Simple Plot')
```

```
>>> plt.legend()
```

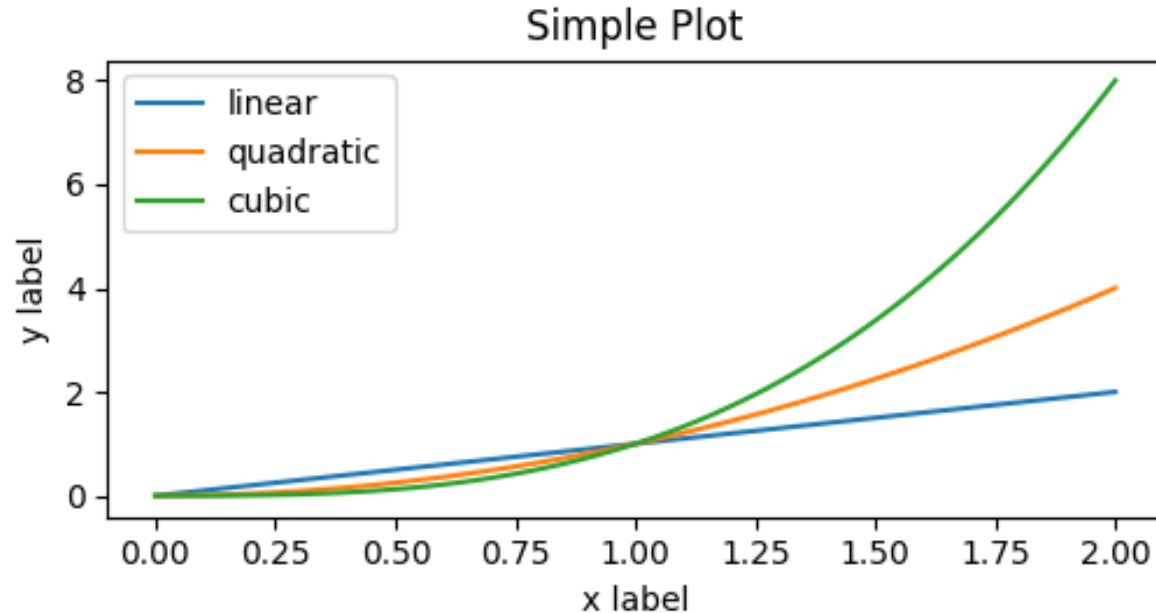
```
<matplotlib.legend.Legend object at 0x000001417FF4B190>
```

# Coding Styles

## The explicit and the implicit interfaces

Example of `pyplot-style`: cont'd.

```
>>> plt.show()
```



# Coding Styles

## Making a helper functions

If we need to make the **same plots** over and over again with different **data sets**, or want to easily wrap **Matplotlib** methods, use the recommended **signature function** below.

```
>>> import matplotlib.pyplot as plt
>>> import numpy as np
>>> def my_plotter(ax, data1, data2, param_dict):
    """
    A helper function to make a graph.
    """
    out = ax.plot(data1, data2, **param_dict)
    return out
```

# Coding Styles

## Making a helper functions

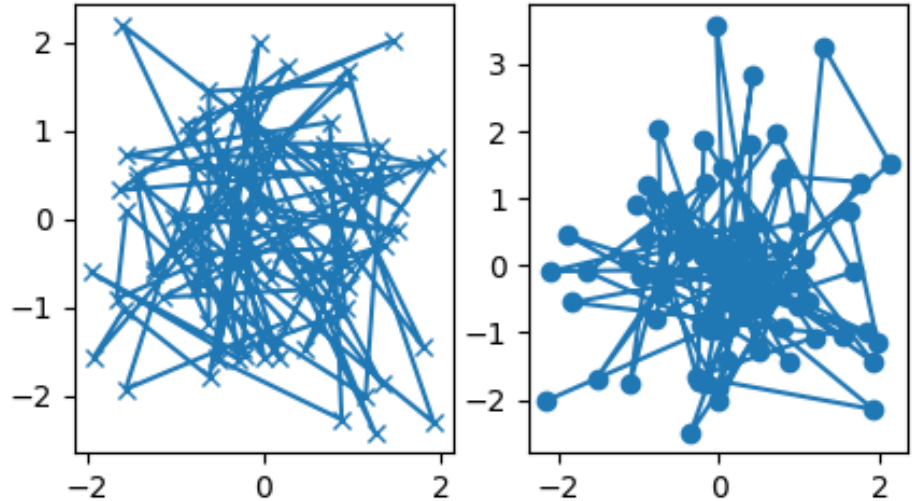
We would then use twice to populate two subplots:

```
>>> data1, data2, data3, data4 = np.random.randn(4, 100)
# make 4 random data sets
>>> fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(5,2.7))
>>> my_plotter(ax1, data1, data2, {'marker': 'x'})
[<matplotlib.lines.Line2D object at 0x000001F3D45F5B70>]
>>> my_plotter(ax2, data3, data4, {'marker': 'o'})
[<matplotlib.lines.Line2D object at 0x000001F3D45F5E70>]
```

# Coding Styles

## Making a helper functions

```
>>> plt.show()
```



**Note** that if you want to install these as a python package, or any other customizations you could use one of the many templates on the web; **Matplotlib** has one at [mpl-cookiercutter](https://github.com/matplotlib/mpl-cookiercutter)

# Styling Artists

- Most `plotting` methods have `styling` options for the `Artists`, accessible either when a `plotting` method is called, or from a "`setter`" on the `Artist`.
- In the `plot` below we `manually` set the `color`, `linewidth`, and `linestyle` of the `Artists` created by `plot`, and we set the `linestyle` of the `second line` after the fact with `set_linestyle`.

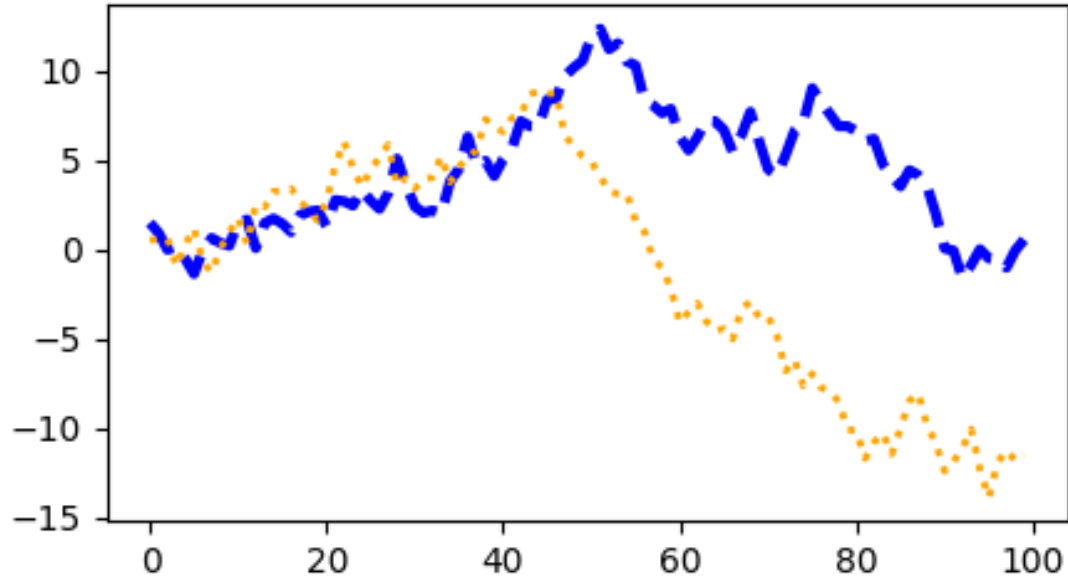
# Styling Artists

```
>>> import matplotlib.pyplot as plt
>>> data1, data2, data3, data4 = np.random.randn(4, 100)
>>> fig, ax = plt.subplots(figsize=(5, 2.7))
>>> x = np.arange(len(data1))
>>> ax.plot(x, np.cumsum(data1), color='blue',
linewidth=3, linestyle='--')
[<matplotlib.lines.Line2D object at 0x000001F3D2E2F580>]
>>> l, = ax.plot(x, np.cumsum(data2), color='orange',
linewidth=2)
>>> l.set_linestyle(':')
```



# Styling Artists

```
>>> plt.show()
```



# Styling Artists

## Colors

- **Matplotlib** has a very flexible array of colors that are accepted for most **Artists**.
- Some **Artists** will take **multiple colors**. i.e. for a **scatter plot**, the **edge** of the markers can be different colors from the **interior**:

# Styling Artists

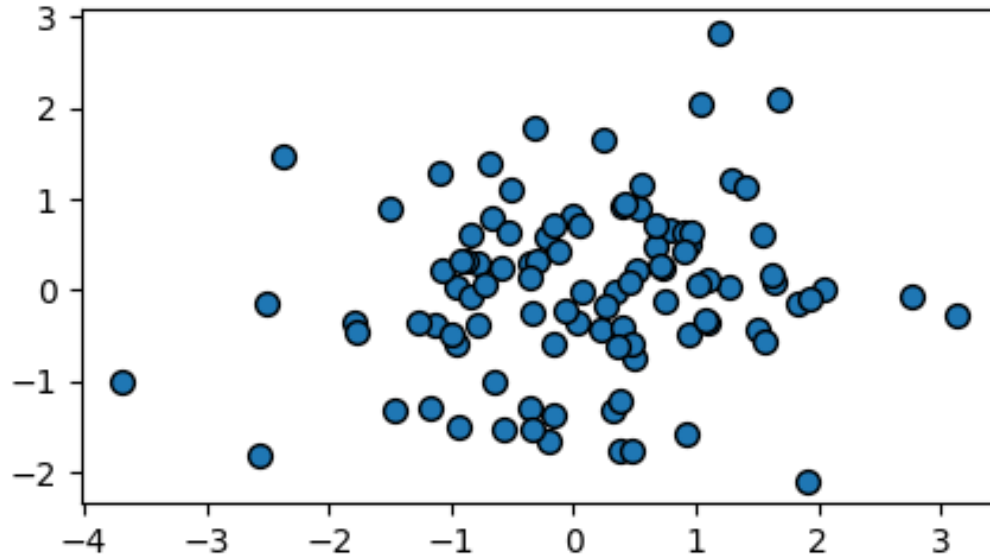
## Colors

```
>>> import matplotlib.pyplot as plt
>>> data1, data2, data3, data4 = np.random.randn(4, 100)
>>> fig, ax = plt.subplots(figsize=(5, 2.7))
>>> ax.scatter(data1, data2, s=50, facecolor='C0',
edgecolor='k')
<matplotlib.collections.PathCollection object at
0x000001ED8D6A4FD0>
```

# Styling Artists

## Colors

```
>>> plt.show()
```



# Styling Artists

## Linewidths, linestyle, and markersizes

- **Line widths** are typically in typographic points ( $1 \text{ pt} = 1/72 \text{ inch}$ ) and available for **Artists** that have **stroked lines**.
- Similarly, **stroked lines** can have a **linestyle**.
- **Marker size** depends on the method being used.
- `plot` specifies **markersize** in points, and is generally the "**diameter**" or **width** of the marker.
- `scatter` specifies **markersize** as approximately proportional to the **visual area** of the marker.
- There is an **array** of **markerstyles** available as **string codes**, or users can define their own `MarkerStyle`.

# Styling Artists

## Linewidths, linestyle, and markersizes

```
>>> import matplotlib.pyplot as plt
>>> data1, data2, data3, data4 = np.random.randn(4, 100)
>>> fig, ax = plt.subplots(figsize=(5, 2.7))
>>> ax.plot(data1, 'o', label='data1')
[<matplotlib.lines.Line2D object at 0x000001ED8BEF8970>]
>>> ax.plot(data2, 'd', label='data2')
[<matplotlib.lines.Line2D object at 0x000001ED8BEF8C40>]
>>> ax.plot(data3, 'v', label='data3')
[<matplotlib.lines.Line2D object at 0x000001ED8BEF8DC0>]
>>> ax.plot(data4, 's', label='data4')
[<matplotlib.lines.Line2D object at 0x000001ED8BEF9060>]
```

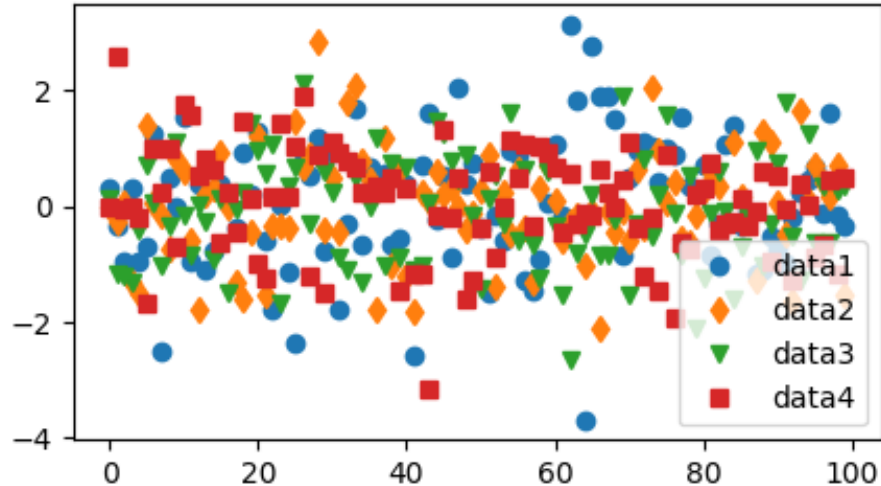
# Styling Artists

## Linewidths, linestyle, and markersizes

```
>>> ax.legend()
```

```
<matplotlib.legend.Legend object at 0x000001ED8BEF91E0>
```

```
>>> plt.show()
```



# Labelling plots

## Axes labels and text

- `set_xlabel`, `set_ylabel`, and `set_title` are used to add text in the indicated locations.
- Text can also be directly added to plots using `text`:



# Labelling plots

## Axes labels and text

```
>>> import matplotlib.pyplot as plt
>>> mu, sigma = 115, 15
>>> x = mu + sigma * np.random.randn(10000)
>>> fig, ax = plt.subplots(figsize=(5, 2.7),
layout='constrained')
>>> n, bins, patches = ax.hist(x, 50, density=True,
facecolor='C0', alpha=0.75)
>>> ax.set_xlabel('Length [cm]')
Text(0.5, 0, 'Length [cm]')
>>> ax.set_ylabel('Probability')
Text(0, 0.5, 'Probability')
```

# Labelling plots

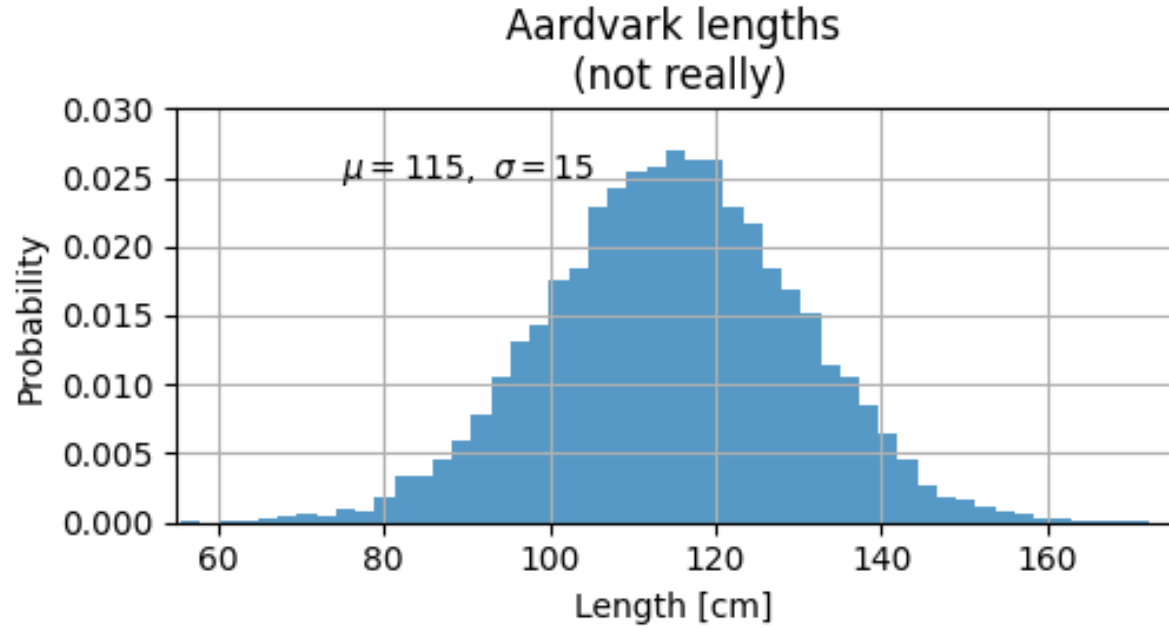
## Axes labels and text

```
>>> ax.set_title('Aardvark lengths\n (not really)')
Text(0.5, 1.0, 'Aardvark lengths\n (not really)')
>>> ax.text(75, .025, r'$\mu=115,\ \sigma=15$')
Text(75, 0.025, '$\mu=115,\ \sigma=15$')
>>> ax.axis([55, 175, 0, 0.03])
(55.0, 175.0, 0.0, 0.03)
>>> ax.grid(True)
```

# Labelling plots

## Axes labels and text

```
>>> plt.show()
```



# Labelling plots

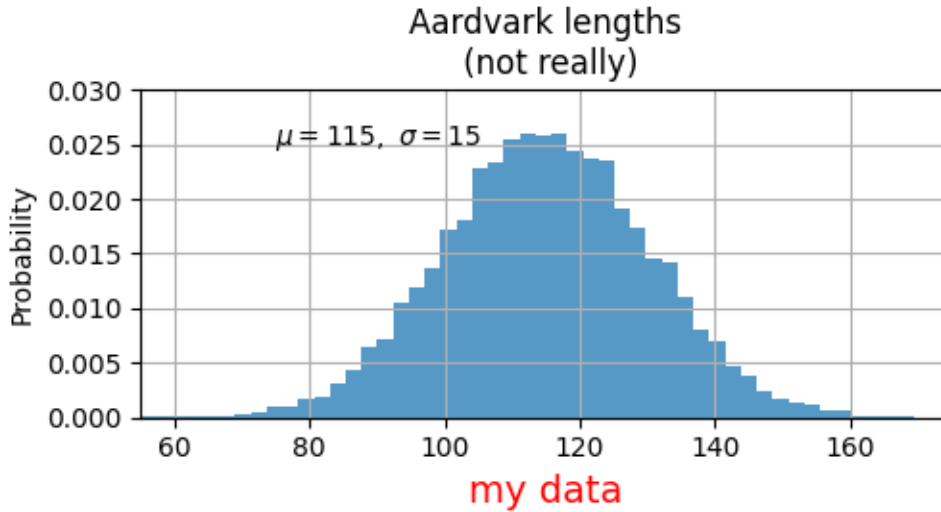
## Axes labels and text

- All of the `text` functions return a `matplotlib.text.Text` instance.
- Just as **with lines above**, you can **customize the properties** by passing **keyword arguments** into the **text functions**:

# Labelling plots

## Axes labels and text

```
>>> ax.set_xlabel('my data', fontsize=14, color='red')  
>>> plt.show()
```



# Labelling plots

## Annotations

- We can also **annotate points** on a **plot**, often by connecting an **arrow** pointing to `xy`, to a piece of text at `xytext`:

```
>>> import matplotlib.pyplot as plt
>>> import numpy as np
>>> fig, ax = plt.subplots(figsize=(5, 2.7))
>>> t = np.arange(0.0, 5.0, 0.01)
>>> s = np.cos(2 * np.pi * t)
>>> line, = ax.plot(t, s, lw=2)
```

# Labelling plots

## Annotations

```
>>> ax.annotate('local max', xy=(2, 1), xytext=(3, 1.5),  
                arrowprops=dict(facecolor='black',  
shrink=0.05))
```

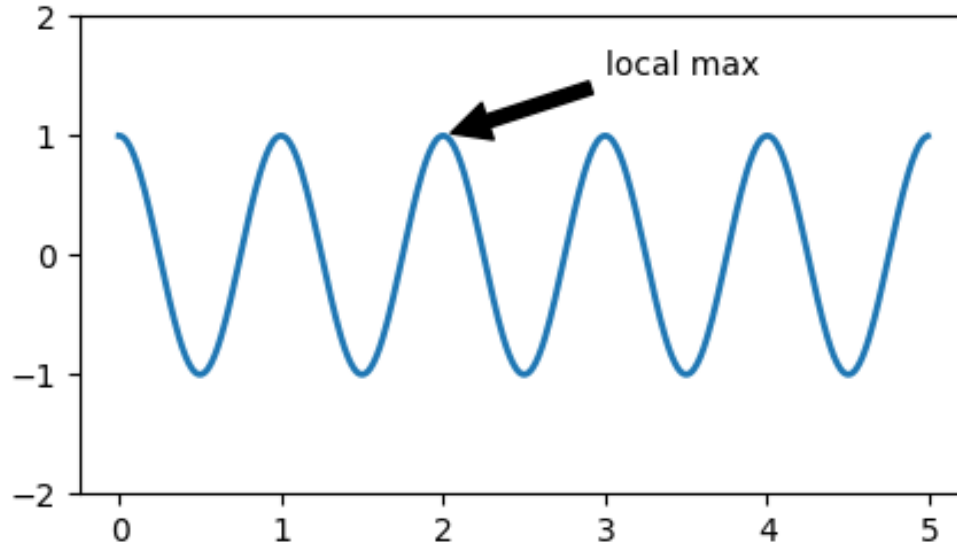
```
Text(3, 1.5, 'local max')
```

```
>>> ax.set_ylim(-2, 2)  
(-2.0, 2.0)
```

# Labelling plots

## Annotations

```
>>> plt.show()
```





# Labelling plots

## Legends

- Often we want to identify **lines** or **markers** with a `Axes.legend`:

```
>>> import matplotlib.pyplot as plt
>>> import numpy as np
>>> fig, ax = plt.subplots(figsize=(5, 2.7))
>>> data1, data2, data3, data4 = np.random.randn(4, 100)
>>> ax.plot(np.arange(len(data1)), data1, label='data1')
[<matplotlib.lines.Line2D object at 0x000001ED9036CAC0>]
```

# Labelling plots

## Legends

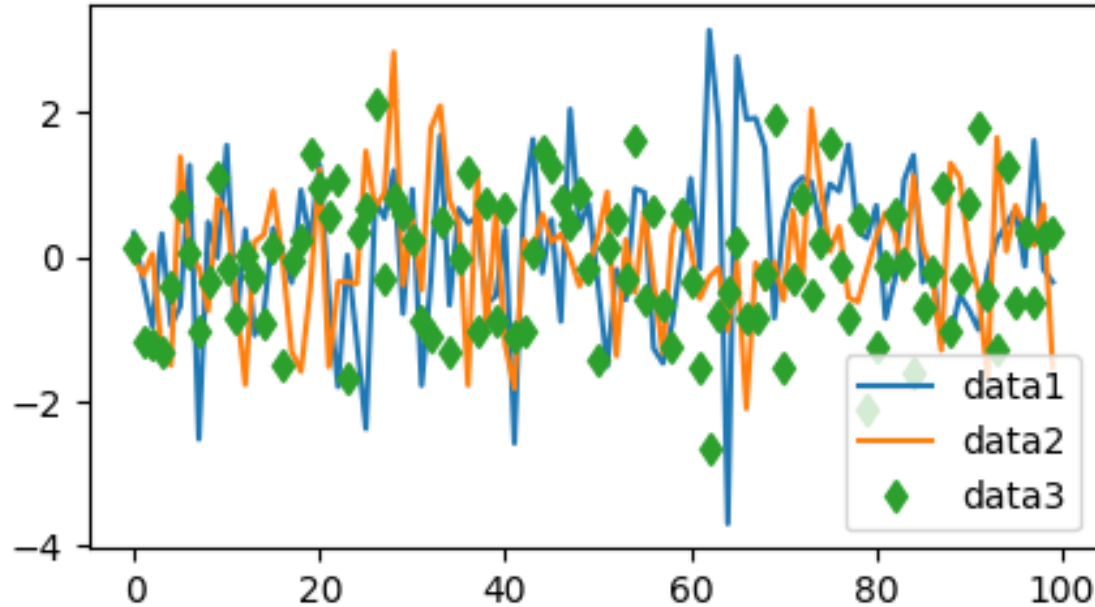
- Often we want to identify **lines** or **markers** with a `Axes.legend`:

```
>>> ax.plot(np.arange(len(data2)), data2, label='data2')
[<matplotlib.lines.Line2D object at 0x000001ED9036CD90>]
>>> ax.plot(np.arange(len(data3)), data3, 'd',
label='data3')
[<matplotlib.lines.Line2D object at 0x000001ED9036D030>]
>>> ax.legend()
<matplotlib.legend.Legend object at 0x000001ED902EC6A0>
```

# Labelling plots

## Legends

```
>>> plt.show()
```



# Axis scales and ticks

- Each Axes has two (or three) `Axis` objects representing the `x`- and `y`-axis.
- These control the *scale* of the Axis, the *tick locators* and the *tick formatters*.
- Additional Axes can be attached to display further Axis objects.

# Axis scales and ticks

## Scales

- In addition to the **linear scale**, **Matplotlib** supplies **non-linear scales**, such as a **log-scale**.
- Since **log-scales** are used so much there are also direct methods like `loglog`, `semilogx`, and `semilogy`.
- There are a number of scales.
- Here we **set the scale manually**:

# Axis scales and ticks

## Scales

```
>>> import matplotlib.pyplot as plt
>>> import numpy as np
>>> fig, axs = plt.subplots(1, 2, figsize=(5, 2.7),
layout='constrained')
>>> data1, data2, data3, data4 = np.random.randn(4, 100)
>>> xdata = np.arange(len(data1)) # make an ordinal for
this
>>> data = 10**data1
```

# Axis scales and ticks

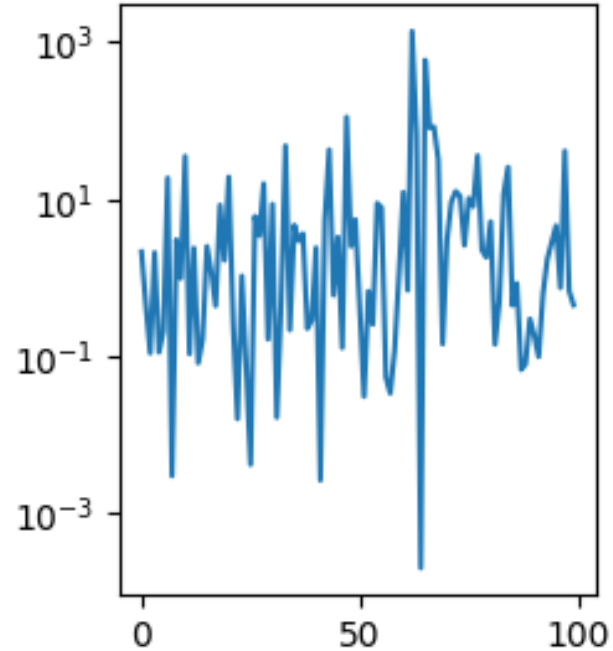
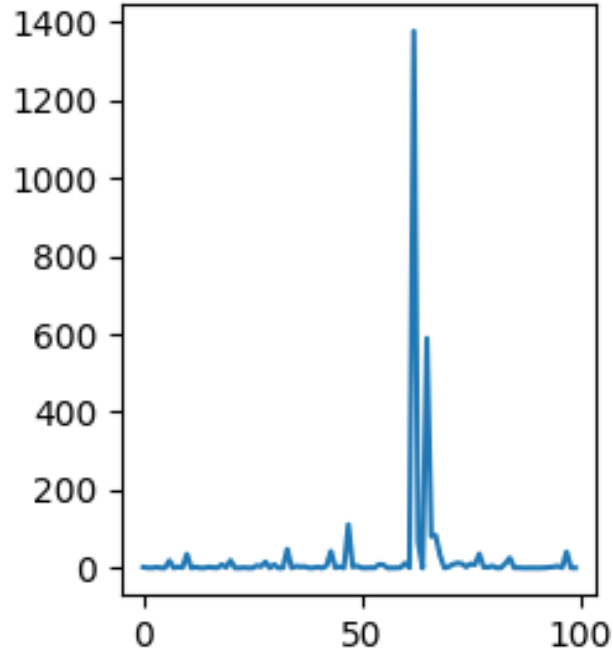
## Scales

```
>>> axs[0].plot(xdata, data)
[<matplotlib.lines.Line2D object at 0x000001ED9041D1E0>]
>>> axs[1].set_yscale('log')
>>> axs[1].plot(xdata, data)
[<matplotlib.lines.Line2D object at 0x000001ED9041D4E0>]
```

# Axis scales and ticks

## Scales

```
>>> plt.show()
```





# Axis scales and ticks

## Scales

- The **scale** sets the mapping from **data values** to **spacing** along the **Axis**.
- This happens in both directions, and gets combined into a *transform*, which is the way that **Matplotlib** maps from **data coordinates** to **Axes**, **Figure**, or **screen coordinates**.

# Axis scales and ticks

## Tick locators and formatters

- Each Axis has a `tick locator` and `formatter` that choose where along the Axis objects to put tick marks.
- A simple interface to this is `set_xticks`:

# Axis scales and ticks

## Tick locators and formatters

```
>>> import matplotlib.pyplot as plt
>>> import numpy as np
>>> fig, axs = plt.subplots(2, 1, layout='constrained')
>>> data1, data2, data3, data4 = np.random.randn(4, 100)
>>> xdata = np.arange(len(data1)) # make an ordinal for
this
>>> axs[0].plot(xdata, data1)
[<matplotlib.lines.Line2D object at 0x000001ED902C60E0>]
```

# Axis scales and ticks

## Tick locators and formatters

```
>>> axs[1].plot(xdata, data1)
[<matplotlib.lines.Line2D object at 0x000001ED8C038A00>]
>>> axs[1].set_xticks(np.arange(0, 100, 30), ['zero',
'30', 'sixty', '90'])
[<matplotlib.axis.XTick object at 0x000001ED8C039FF0>,
<matplotlib.axis.XTick object at 0x000001ED8C039E40>,
<matplotlib.axis.XTick object at 0x000001ED902C4520>,
<matplotlib.axis.XTick object at 0x000001ED8BF90700>]
```

# Axis scales and ticks

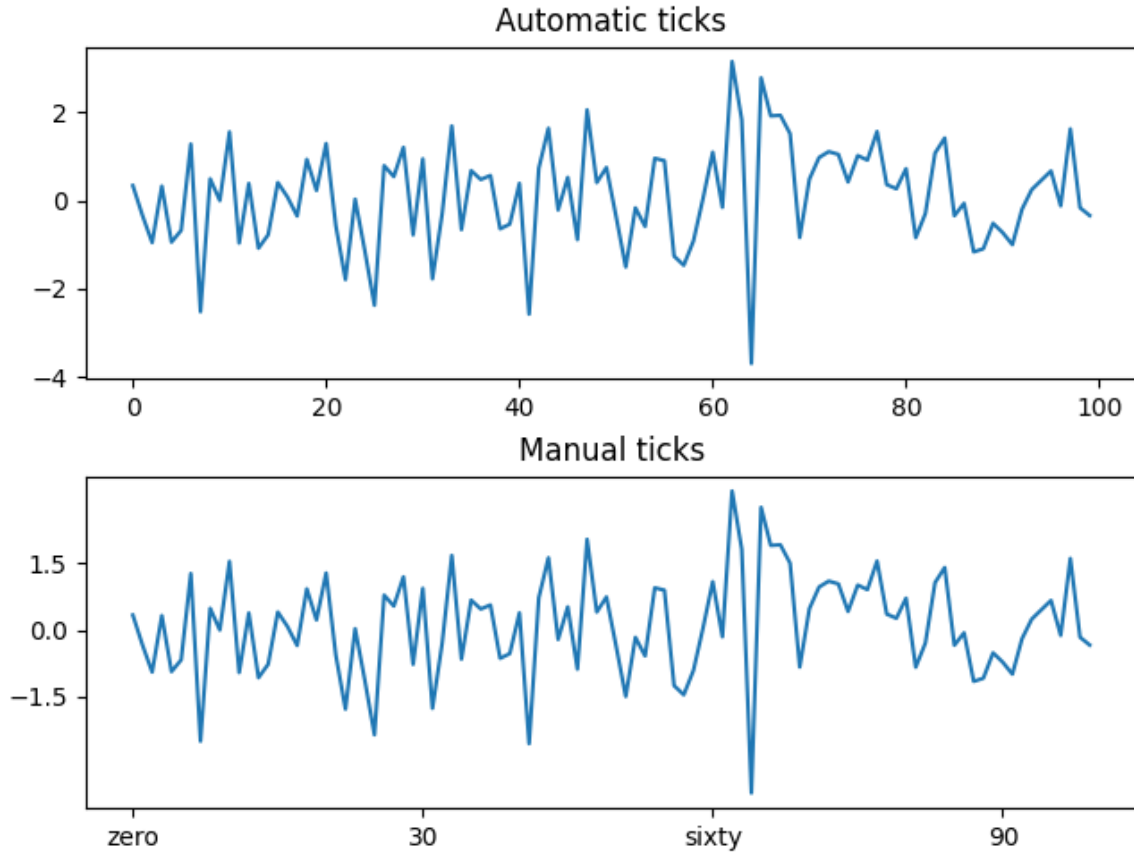
## Tick locators and formatters

```
>>> axs[1].set_yticks([-1.5, 0, 1.5]) # note that we
don't need to specify labels
[<matplotlib.axis.YTick object at 0x0000001ED9045C640>,
<matplotlib.axis.YTick object at 0x0000001ED8C038190>,
<matplotlib.axis.YTick object at 0x0000001ED902C6170>]
>>> axs[1].set_title('Manual ticks')
Text(0.5, 1.0, 'Manual ticks')
```

# Axis scales and ticks

## Tick locators and formatters

```
>>> plt.show()
```



# Axis scales and ticks

## Tick locators and formatters

- Different scales can have different locators and formatters; for instance the **log-scale** above uses `LogLocator` and `LogFormatter`.

# Axis scales and ticks

## Plotting dates and strings

- **Matplotlib** can handle plotting arrays of dates and arrays of strings, as well as floating point numbers.
- These get special locators and formatters as appropriate.
- For dates:



# Axis scales and ticks

## Plotting dates and strings

```
>>> import matplotlib as mpl
>>> import matplotlib.pyplot as plt
>>> import numpy as np
>>> fig, ax = plt.subplots(figsize=(5, 2.7),
layout='constrained')
>>> dates = np.arange(np.datetime64('2023-06-28'),
np.datetime64('2023-07-15'),
np.timedelta64(1, 'h'))
```

# Axis scales and ticks

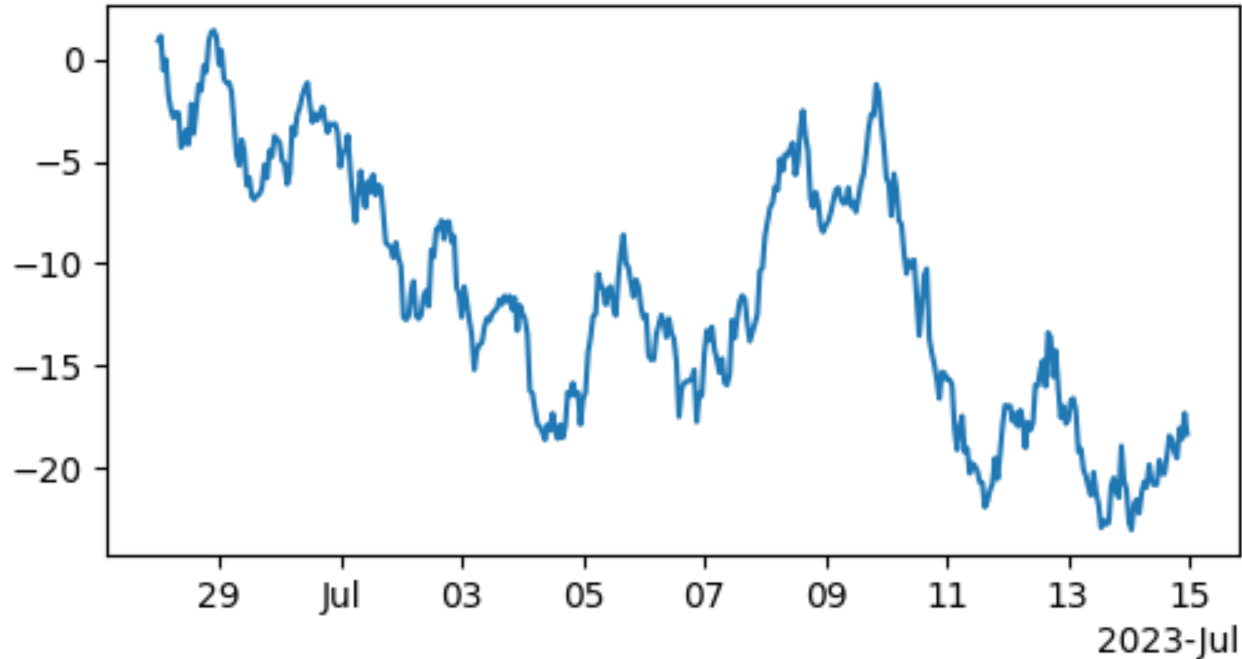
## Plotting dates and strings

```
>>> data = np.cumsum(np.random.randn(len(dates)))
>>> ax.plot(dates, data)
[<matplotlib.lines.Line2D object at 0x000001ED8BF51060>]
>>> cdf =
mpl.dates.ConciseDateFormatter(ax.xaxis.get_major_locator
())
>>> ax.xaxis.set_major_formatter(cdf)
```

# Axis scales and ticks

## Plotting dates and strings

```
>>> plt.show()
```



# Axis scales and ticks

## Plotting dates and strings

- For **strings**: we get **categorical plotting**

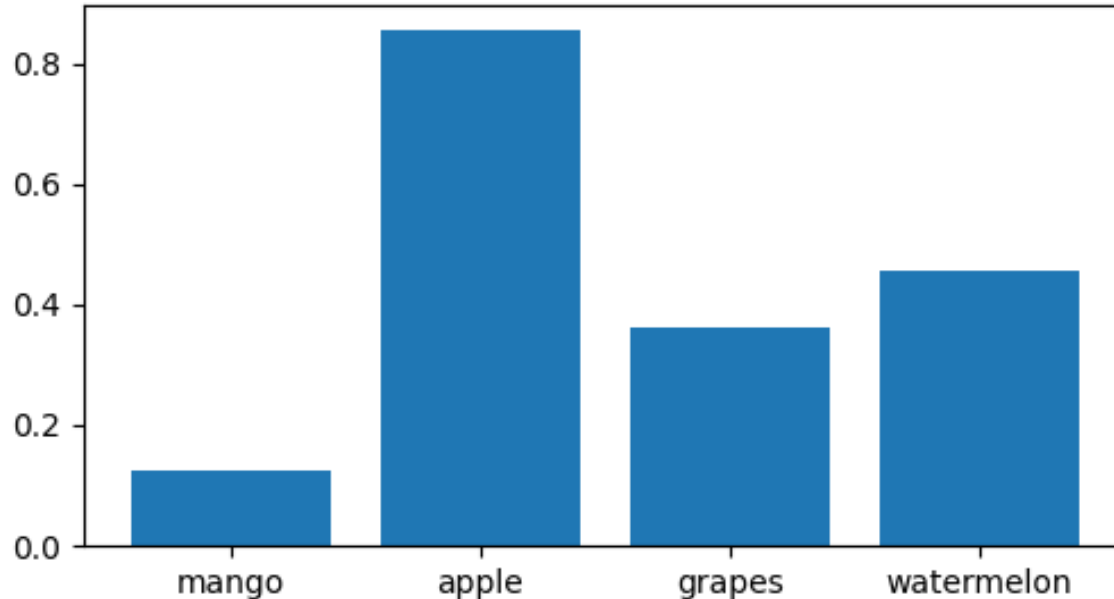
```
>>> import matplotlib.pyplot as plt
>>> import numpy as np
>>> fig, ax = plt.subplots(figsize=(5, 2.7),
layout='constrained')
>>> categories = ['mango', 'apple', 'grapes',
'watermelon']
>>> ax.bar(categories, np.random.rand(len(categories)))
<BarContainer object of 4 artists>
```

# Axis scales and ticks

## Plotting dates and strings

- For **strings**: we get **categorical plotting**

```
>>> plt.show()
```



# Axis scales and ticks

## Plotting dates and strings

- One limitation about **categorical plotting** is that some methods of parsing **text files** return a list of strings, even if the strings all represent **numbers** or **dates**.
- If you pass **1000 strings**, **Matplotlib** will think you meant **1000 categories** and will add **1000 ticks** to your **plot**.

# Axis scales and ticks

## Additional Axis objects

- Plotting data of different magnitude in one chart may require an additional `y-axis`.
- Such an `Axis` can be created by using `twinx` to add a `new Axes` with an invisible `x-axis` and a `y-axis` positioned at the `right` (analogously for `twinx`).
- Similarly, you can add a `secondary_xaxis` or `secondary_yaxis` having a `different scale` than the `main Axis` to represent the data in `different scales` or `units`.

# Axis scales and ticks

## Additional Axis objects

```
>>> import matplotlib.pyplot as plt
>>> import numpy as np
>>> fig, (ax1, ax3) = plt.subplots(1, 2, figsize=(7,
2.7), layout='constrained')
>>> l1, = ax1.plot(t, s)
>>> ax2 = ax1.twinx()
>>> l2, = ax2.plot(t, range(len(t)), 'C1')
>>> ax2.legend([l1, l2], ['Sine (left)', 'Straight
(right)'])
<matplotlib.legend.Legend object at 0x000001ED905501C0>
```



# Axis scales and ticks

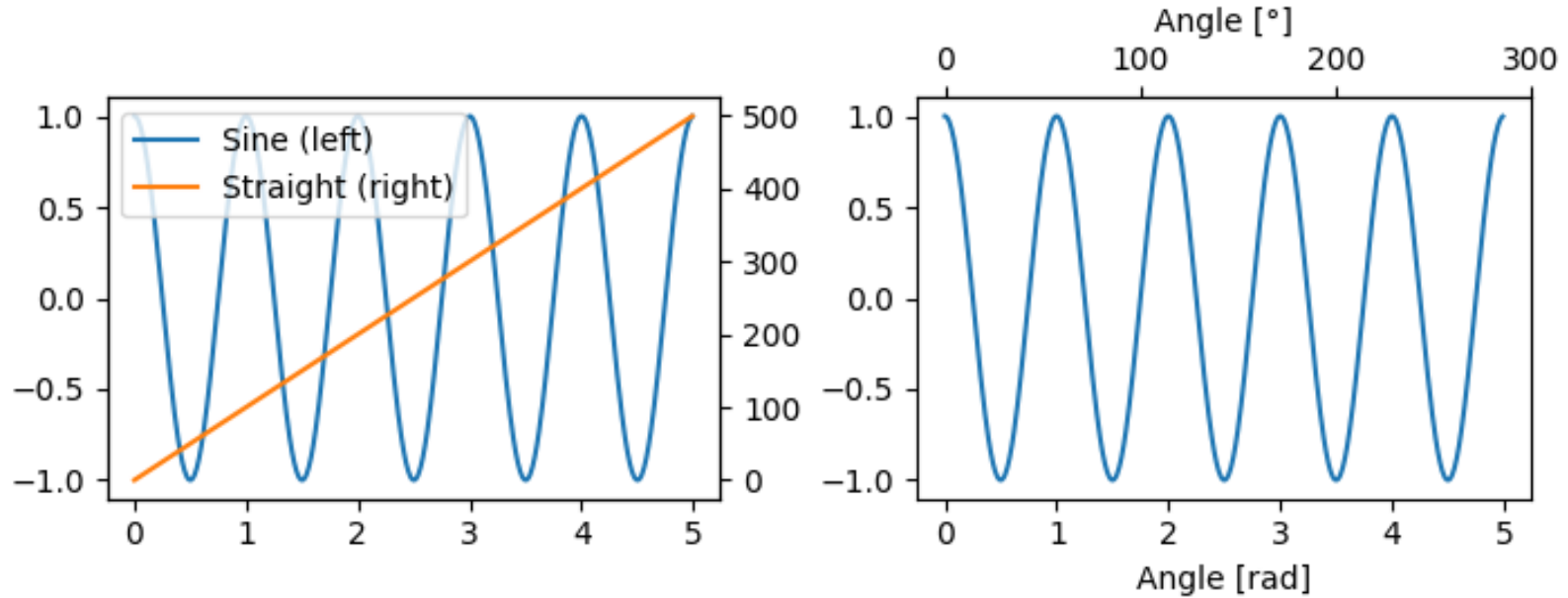
## Additional Axis objects

```
>>> ax3.plot(t, s)
[<matplotlib.lines.Line2D object at 0x000001ED90581C90>]
>>> ax3.set_xlabel('Angle [rad]')
Text(0.5, 0, 'Angle [rad]')
>>> ax4 = ax3.secondary_xaxis('top',
functions=(np.rad2deg, np.deg2rad))
>>> ax4.set_xlabel('Angle [°]')
Text(0.5, 0, 'Angle [°]')
```

# Axis scales and ticks

## Additional Axis objects

```
>>> plt.show()
```



# Color mapped data

- Often we want to have a **third dimension** in a **plot** represented by a **colors** in a **colormap**.
- **Matplotlib** has a number of **plot types** that do this:

# Color mapped data

```
>>> import matplotlib.pyplot as plt
>>> import numpy as np
>>> X, Y = np.meshgrid(np.linspace(-3, 3, 128),
np.linspace(-3, 3, 128))
>>> Z = (1 - X/2 + X**5 + Y**3) * np.exp(-X**2 - Y**2)
>>> fig, axs = plt.subplots(2, 2, layout='constrained')
>>> pc = axs[0, 0].pcolormesh(X, Y, Z, vmin=-1, vmax=1,
cmap='RdBu_r')
>>> fig.colorbar(pc, ax=axs[0, 0])
<matplotlib.colorbar.Colorbar object at
0x000001ED9372DDB0>
```

# Color mapped data

```
>>> axs[0, 0].set_title('pcolormesh()')
Text(0.5, 1.0, 'pcolormesh()')
>>> co = axs[0, 1].contourf(X, Y, Z, levels=np.linspace(-
1.25, 1.25, 11))
>>> fig.colorbar(co, ax=axs[0, 1])
<matplotlib.colorbar.Colorbar object at
0x000001ED9001D3C0>
>>> axs[0, 1].set_title('contourf()')
Text(0.5, 1.0, 'contourf()')
```

# Color mapped data

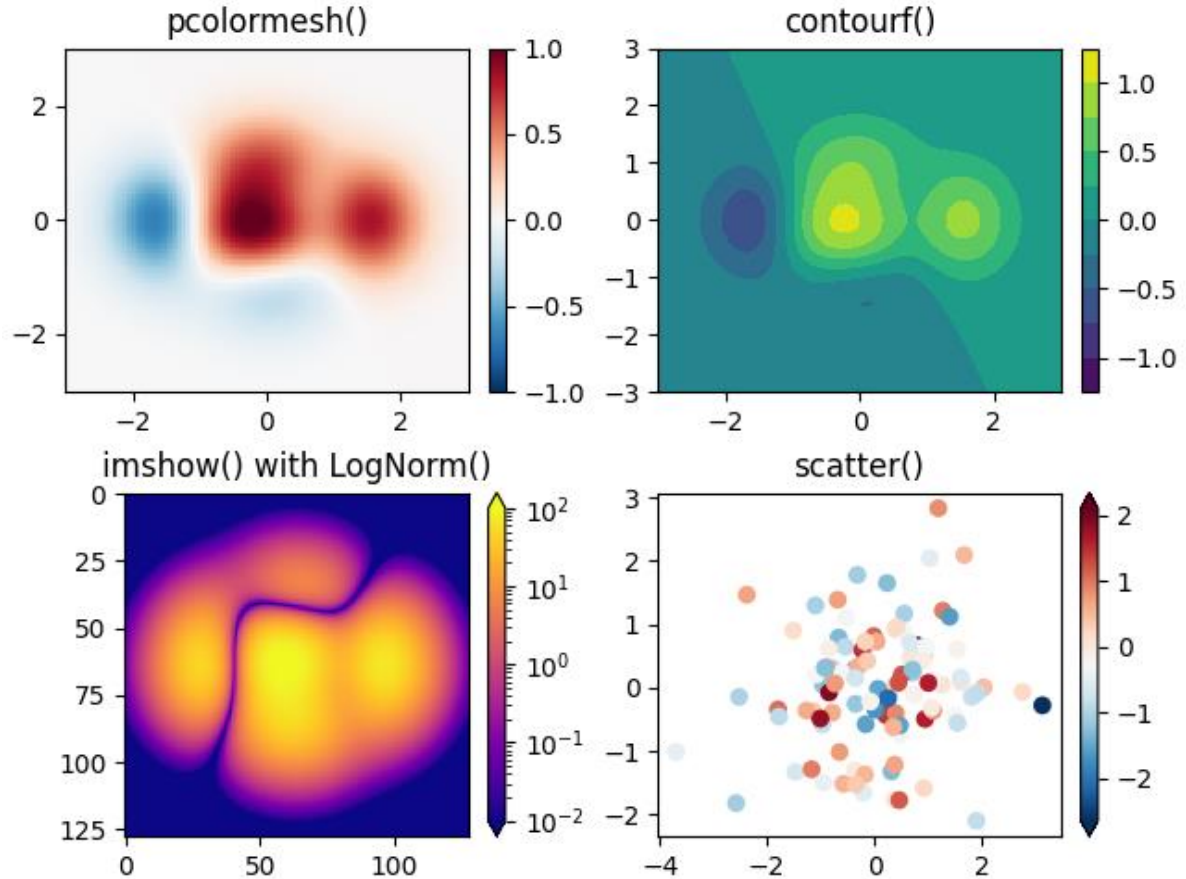
```
>>> pc = axs[1, 0].imshow(Z**2 * 100, cmap='plasma',  
  
norm=mpl.colors.LogNorm(vmin=0.01, vmax=100))  
  
>>> fig.colorbar(pc, ax=axs[1, 0], extend='both')  
  
<matplotlib.colorbar.Colorbar object at  
0x000001ED93696500>  
  
>>> axs[1, 0].set_title('imshow() with LogNorm()')  
Text(0.5, 1.0, 'imshow() with LogNorm()')
```

# Color mapped data

```
>>> pc = axs[1, 1].scatter(data1, data2, c=data3,  
cmap='RdBu_r')  
  
>>> fig.colorbar(pc, ax=axs[1, 1], extend='both')  
  
<matplotlib.colorbar.Colorbar object at  
0x000001ED936C2530>  
  
>>> axs[1, 1].set_title('scatter()')  
  
Text(0.5, 1.0, 'scatter()')
```

# Color mapped data

```
>>> plt.show()
```





# Color mapped data

## Colormaps

- These are all examples of Artists that derive from `ScalarMappable` objects.
- They all can set a linear mapping between `vmin` and `vmax` into the colormap specified by `cmap`.
- `Matplotlib` has many colormaps to choose from the existing, you can make your own or download as third-party packages.

# Color mapped data

## Normalizations

- Sometimes we want a **non-linear mapping** of the data to the **colormap**, as in the `LogNorm` example above.
- We do this by supplying the `ScalarMappable` with the *norm* argument instead of *vmin* and *vmax*.

# Color mapped data

## Colorbars

- Adding a `colorbar` gives a key to relate the color back to the underlying data.
- **Colorbars** are **figure-level Artists**, and are attached to a **ScalarMappable** (where they get their information about the **norm** and **colormap**) and usually steal space from a parent Axes.
- Placement of **colorbars** can be complex.
- You can also change the appearance of **colorbars** with the `extend` keyword to add arrows to the ends, and `shrink` and `aspect` to control the size.
- Finally, the **colorbar** will have default locators and formatters appropriate to the `norm`.
- These can be changed as for other Axis objects.