

# matplotlib

## Lecture 10

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# matplotlib

# matplotlib vs. pyplot

matplotlib is a comprehensive library for creating static, animated, and interactive visualizations in Python.

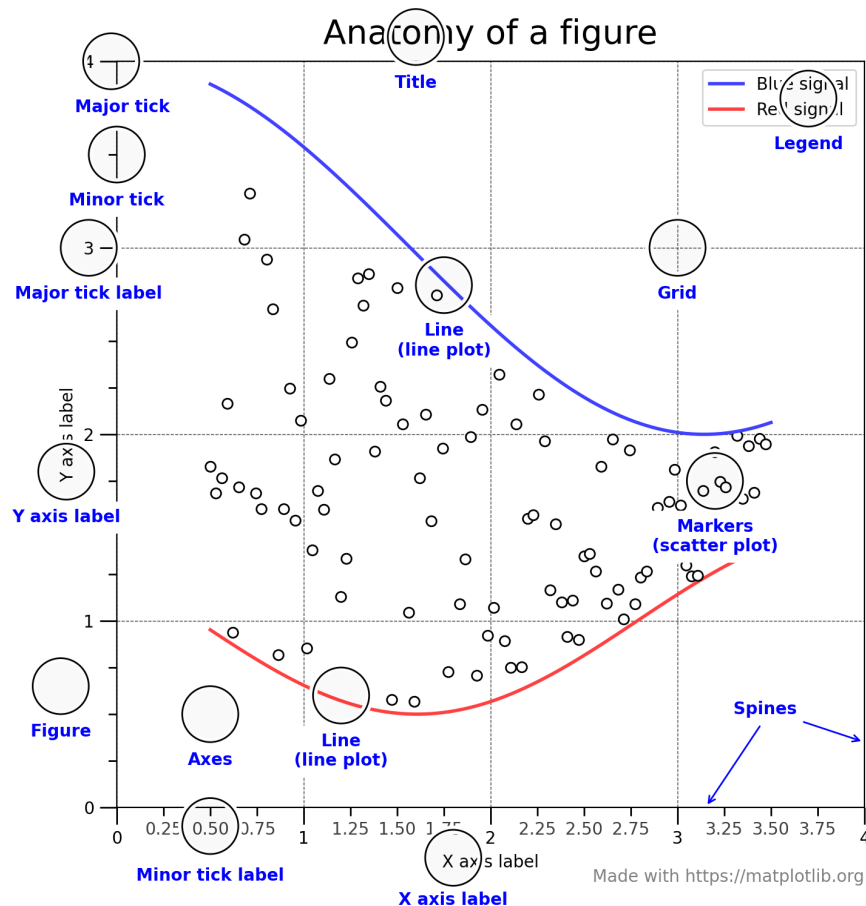
```
1 import matplotlib as mpl
2 import matplotlib.pyplot as plt
```

Why do we usually import only pyplot then?

Matplotlib is the whole package; matplotlib.pyplot is a module in matplotlib; and pylab is a module that gets installed alongside matplotlib.

Pyplot provides the state-machine interface to the underlying object-oriented plotting library. The state-machine implicitly and automatically creates figures and axes to achieve the desired plot.

# Plot anatomy

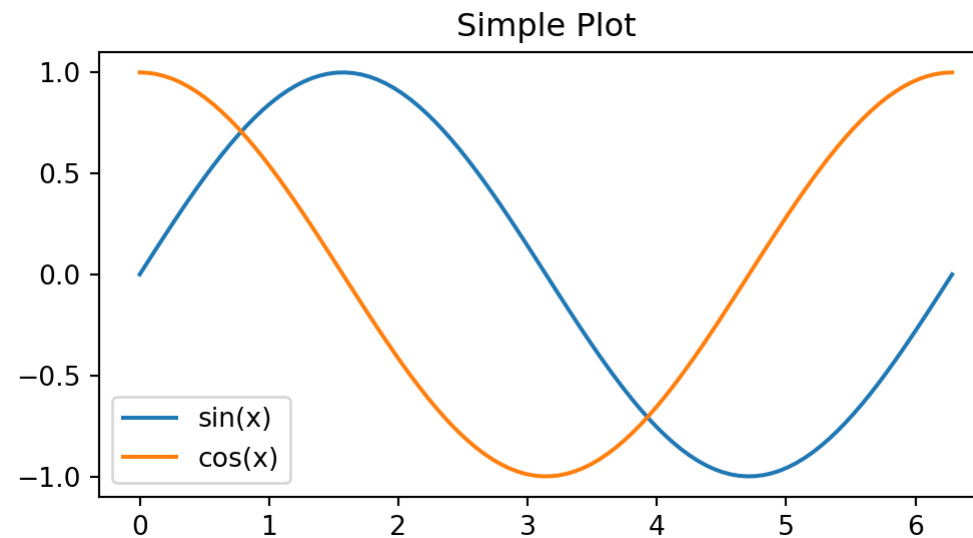


## Important terminology:

- **Figure** - The entire plot (including subplots)
- **Axes** - Subplot attached to a figure, contains the region for plotting data and axis'
- **Axis** - Set the scale and limits, generate ticks and ticklabels
- **Artist** - Everything visible on a figure: text, lines, axis, axes, etc.

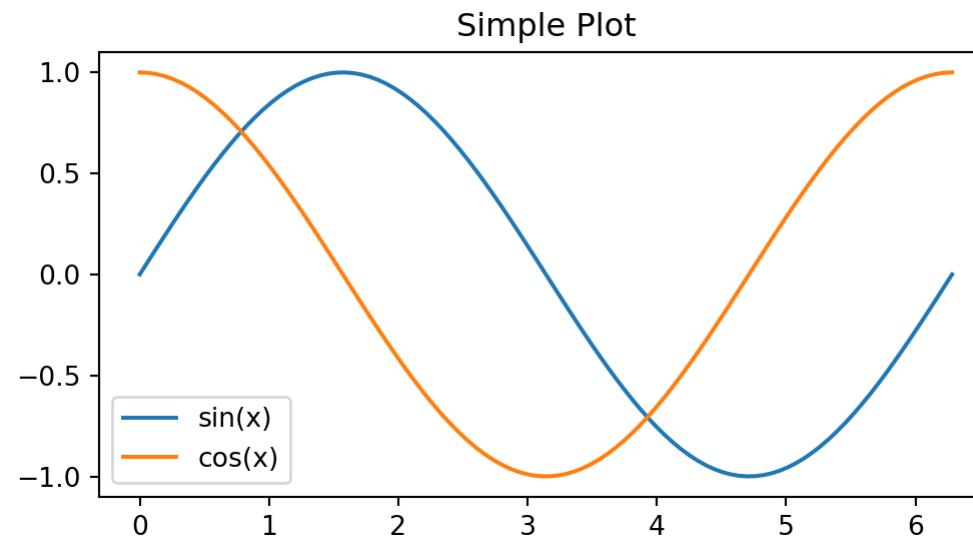
# Basic plot - OO style

```
1 x = np.linspace(0, 2*np.pi, 100)
2 y1 = np.sin(x)
3 y2 = np.cos(x)
4
5 fig, ax = plt.subplots(figsize=(6, 3))
6 ax.plot(x, y1, label="sin(x)")
7 ax.plot(x, y2, label="cos(x)")
8 ax.set_title("Simple Plot")
9 ax.legend()
```



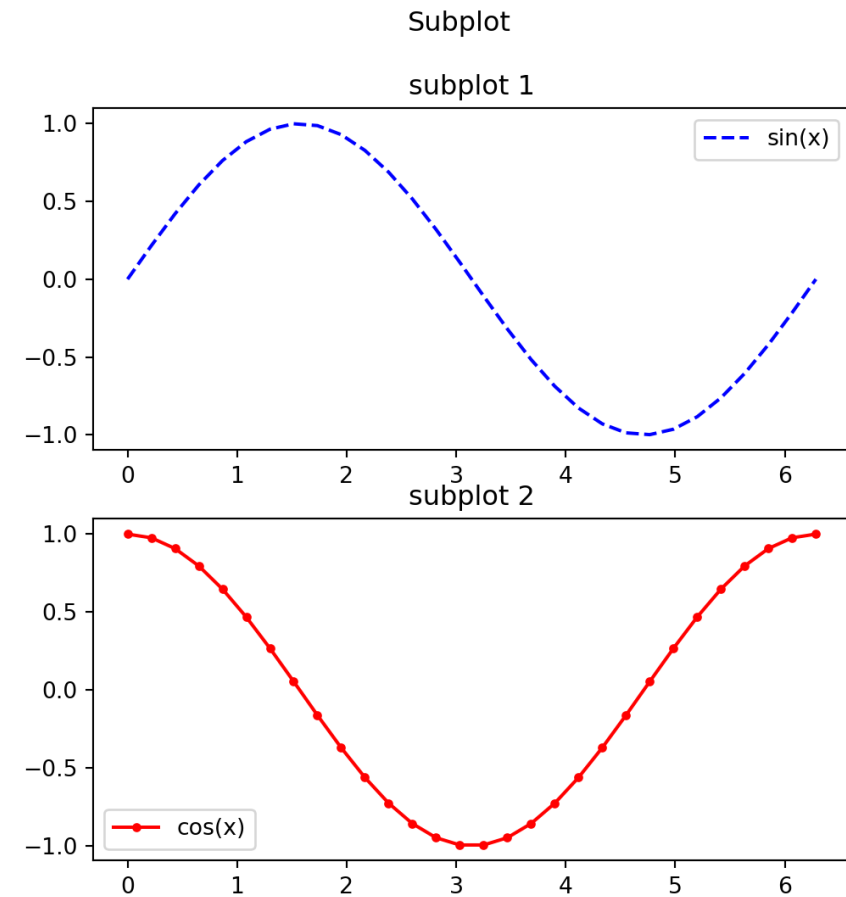
# Basic plot - pyplot style

```
1 x = np.linspace(0, 2*np.pi, 100)
2 y1 = np.sin(x)
3 y2 = np.cos(x)
4
5 plt.figure(figsize=(6, 3))
6 plt.plot(x, y1, label="sin(x)")
7 plt.plot(x, y2, label="cos(x)")
8 plt.title("Simple Plot")
9 plt.legend()
```



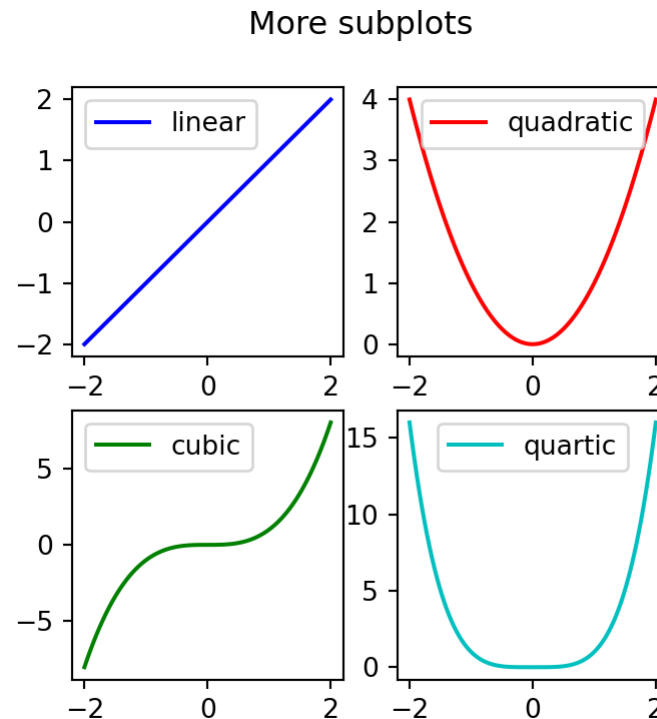
# Subplots

```
1 x = np.linspace(0, 2*np.pi, 30)
2 y1 = np.sin(x)
3 y2 = np.cos(x)
4
5 fig, (ax1, ax2) = plt.subplots(
6     2, 1, figsize=(6, 6)
7 )
8
9 fig.suptitle("Subplot")
10
11 ax1.plot(x, y1, "--b", label="sin(x)")
12 ax1.set_title("subplot 1")
13 ax1.legend()
14
15 ax2.plot(x, y2, "-.r", label="cos(x)")
16 ax2.set_title("subplot 2")
17 ax2.legend()
```



# More subplots

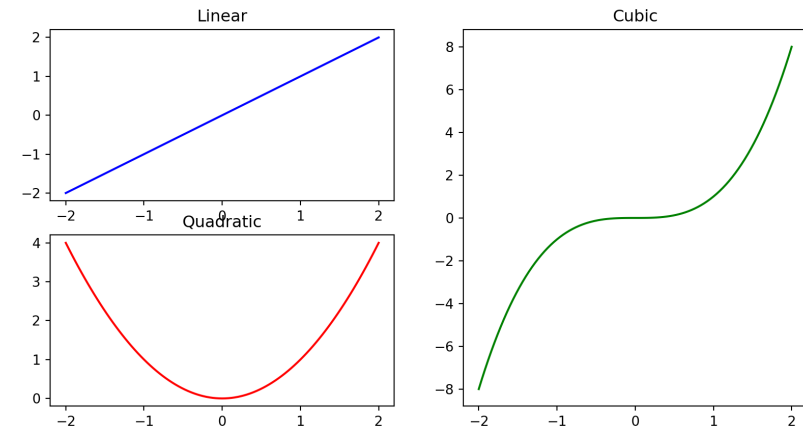
```
1 x = np.linspace(-2, 2, 101)
2
3 fig, axs = plt.subplots(2, 2, figsize=(4, 4))
4 axs[0,0].plot(x, x, "b", label="linear")
5 axs[0,1].plot(x, x**2, "r", label="quadratic")
6 axs[1,0].plot(x, x**3, "g", label="cubic")
7 axs[1,1].plot(x, x**4, "c", label="quartic")
8 [ax.legend() for row in axs for ax in row]
9 fig.suptitle("More subplots")
```





# Fancy subplots (mosaic)

```
1 x = np.linspace(-2, 2, 101)
2
3 fig, axd = plt.subplot_mosaic([['upleft', 'right'],
4                                ['lowleft', 'right']])
5
6 axd['upleft'].plot(x, x, "b", label="linear")
7 axd['lowleft'].plot(x, x**2, "r", label="quadratic")
8 axd['right'].plot(x, x**3, "g", label="cubic")
9
10 axd['upleft'].set_title("Linear")
11 axd['lowleft'].set_title("Quadratic")
12 axd['right'].set_title("Cubic")
```



# Format strings

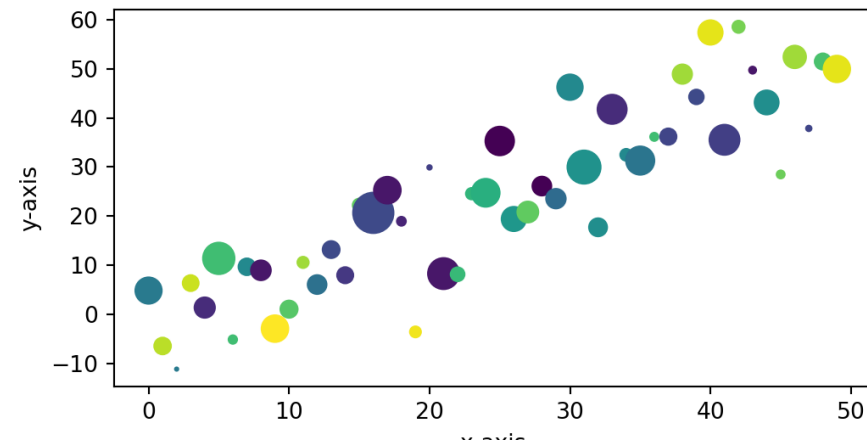
For quick formatting of plots (scatter and line) format strings are a useful shorthand, generally they use the format '`[marker][line][color]`',

character	shape	character	line style	character	color
.	point	—	solid	b	blue
,	pixel	--	dashed	g	green
o	circle	-.	dash-dot	r	red
v	triangle down	:	dotted	c	cyan
^	triangle up			m	magenta
<	triangle left			y	yellow
>	triangle right			k	black
...	+ more			w	white

# Plotting data

Beyond creating plots for arrays (and lists), addressable objects like dicts and DataFrames can be used via `data`,

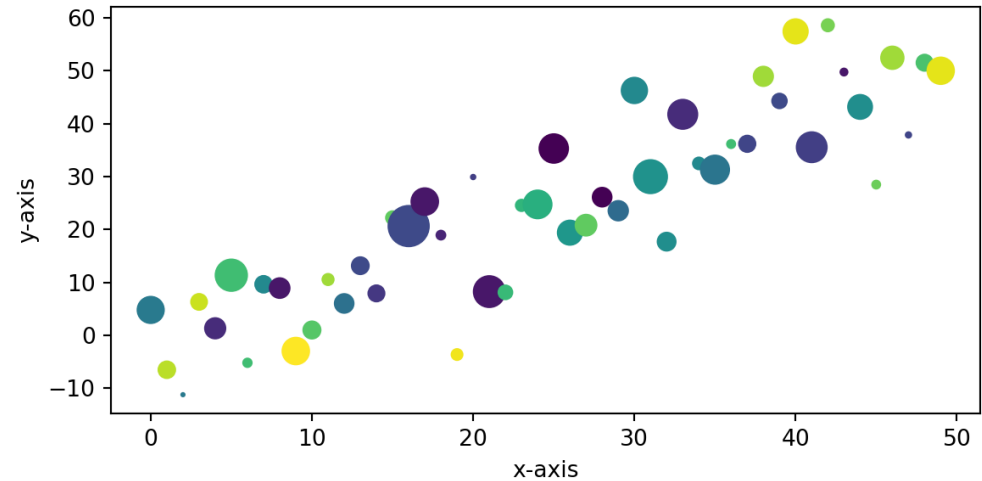
```
1 np.random.seed(19680801)
2 d = {'x': np.arange(50),
3      'color': np.random.randint(0, 50, 50),
4      'size': np.abs(np.random.randn(50)) * 100}
5 d['y'] = d['x'] + 10 * np.random.randn(50)
6
7
8 plt.figure(figsize=(6, 3))
9 plt.scatter(
10     'x', 'y', c='color', s='size',
11     data=d
12 )
13 plt.xlabel("x-axis")
14 plt.ylabel("y-axis")
```



# Constrained layout

To fix the legend clipping we can use the “constrained” layout to adjust automatically,

```
1 np.random.seed(19680801)
2 d = {'x': np.arange(50),
3      'color': np.random.randint(0, 50, 50),
4      'size': np.abs(np.random.randn(50)) * 100}
5 d['y'] = d['x'] + 10 * np.random.randn(50)
6
7
8 plt.figure(
9     figsize=(6, 3),
10    layout="constrained"
11 )
12 plt.scatter(
13     'x', 'y', c='color', s='size',
14     data=d
15 )
16 plt.xlabel("x-axis")
17 plt.ylabel("y-axis")
```

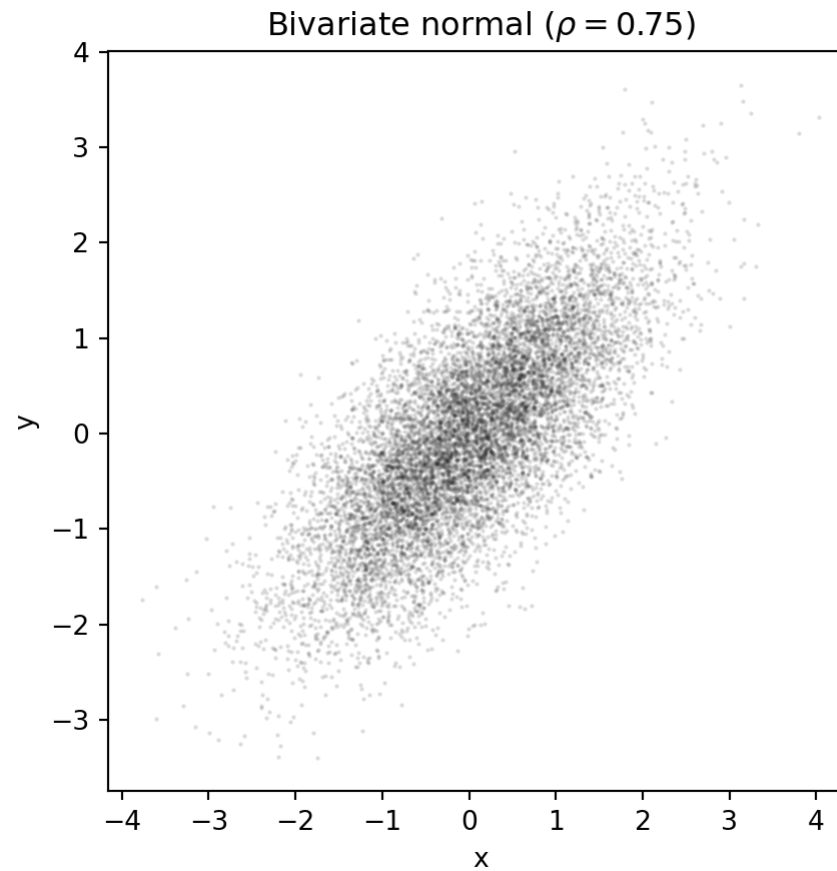


# pyplot w/ pandas data

Data can also come from DataFrame objects or series,

```
1 df = pd.DataFrame({
2     "x": np.random.normal(size=10000)
3 }).assign(
4     y = lambda d: np.random.normal(0.75*d.x, np.sqrt(1-0.75**2), size=10000)
5 )
6
7 fig, ax = plt.subplots(figsize=(5,5))
8 ax.scatter('x', 'y', c='k', data=df, alpha=0.1, s=0.5)
9 ax.set_xlabel('x')
10 ax.set_ylabel('y')
11 ax.set_title("Bivariate normal ( $\rho=0.75$ )")
```

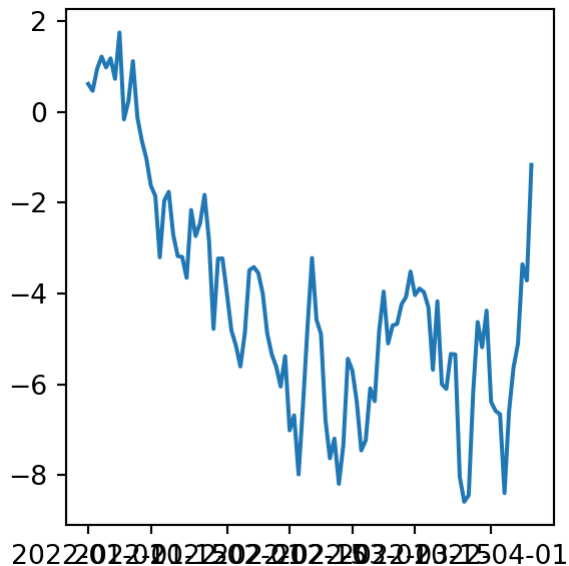
# pyplot w/ pandas data



# pyplot w/ pandas series

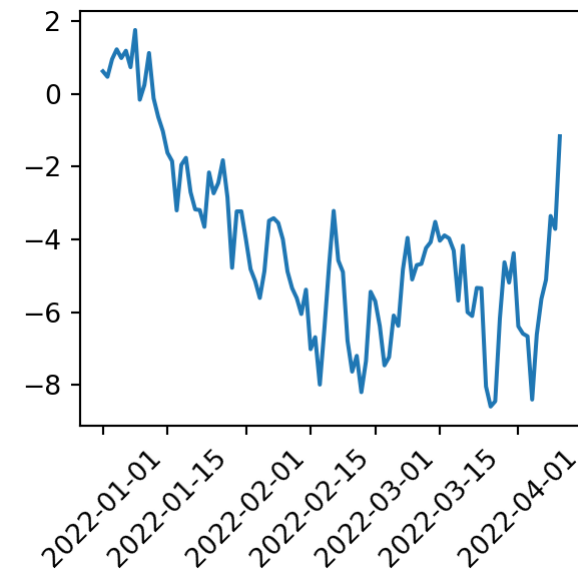
Series objects can also be plotted directly, the index is used as the  $x$  axis values,

```
1 s = pd.Series(  
2     np.cumsum( np.random.normal(size=100) ),  
3     index = pd.date_range("2022-01-01", periods=100)  
4 )  
5 plt.figure(figsize=(3, 3), layout="constrained")  
6 plt.plot(s)  
7 plt.show()
```



```
1 plt.figure(figsize=(3, 3), layout="constrained")  
2 plt.plot(s.index, s.values)  
3 plt.xticks(rotation=45)
```

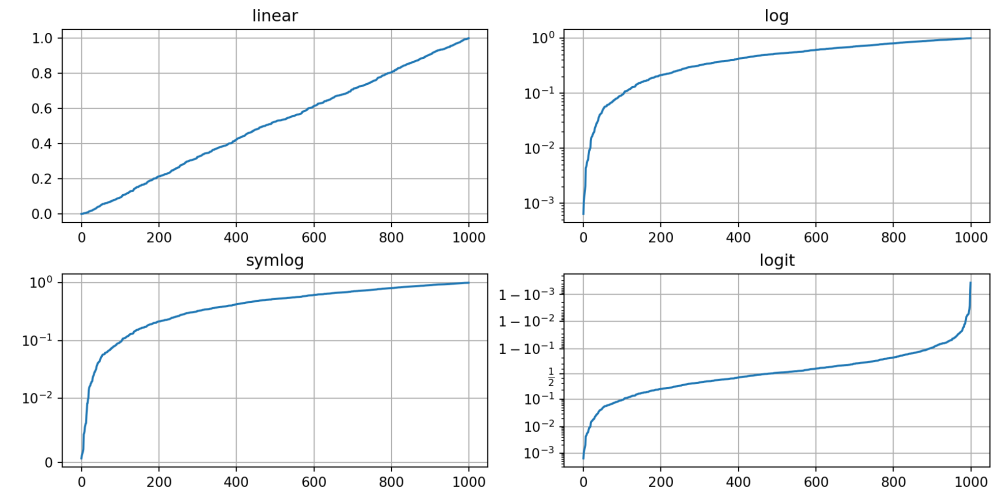
```
1 plt.show()
```



# Scales

Axis scales can be changed via `plt.xscale()`, `plt.yscale()`, `ax.set_xscale()`, or `ax.set_yscale()`, supported values are “linear”, “log”, “symlog”, and “logit”.

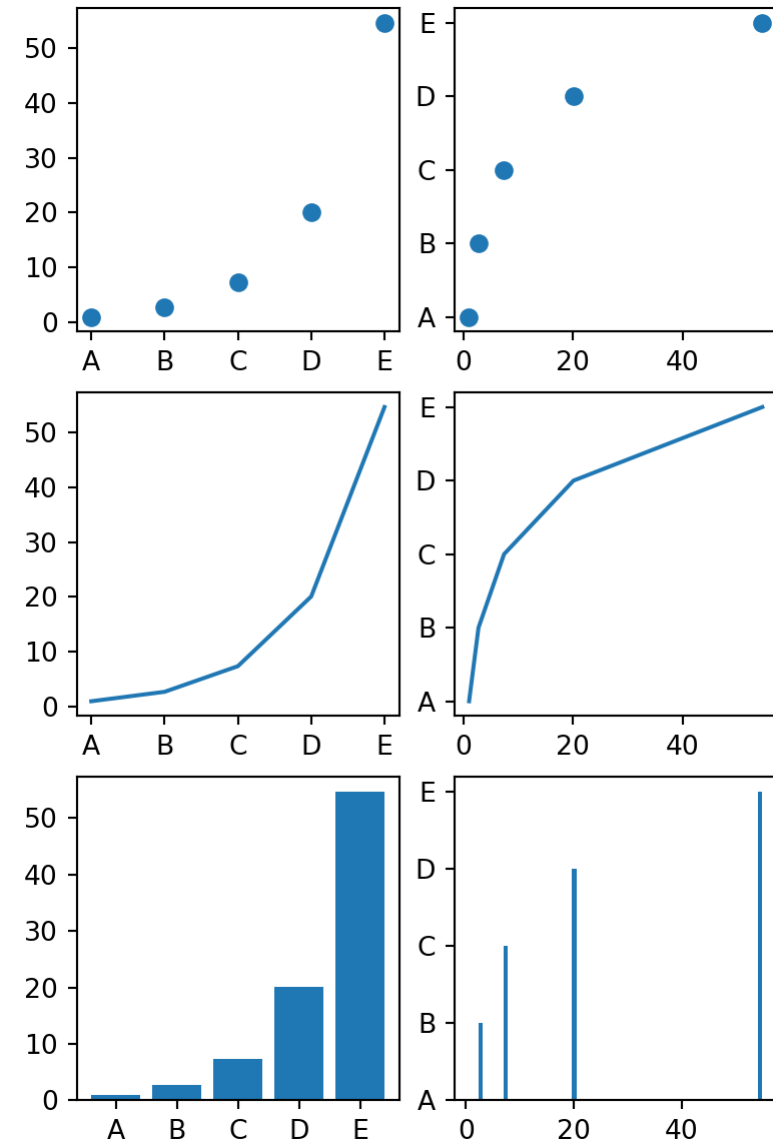
```
1 y = np.sort( np.random.sample(size=1000) )
2 x = np.arange(len(y))
3
4 plt.figure(layout="constrained")
5
6 scales = ['linear', 'log', 'symlog', 'logit']
7 for i, scale in zip(range(4), scales):
8     plt.subplot(221+i)
9     plt.plot(x, y)
10    plt.grid(True)
11    if scale == 'symlog':
12        plt.yscale(scale, linthresh=0.01)
13    else:
14        plt.yscale(scale)
15    plt.title(scale)
16
17
18 plt.show()
```





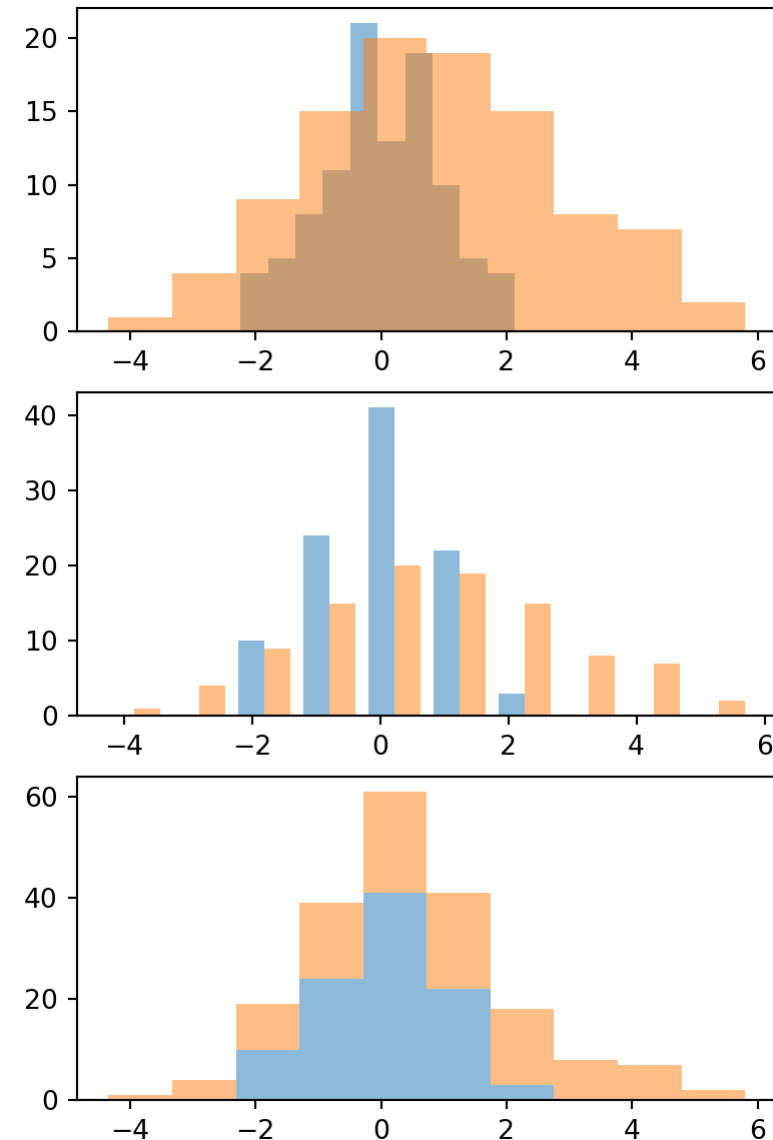
# Categorical data

```
1 df = pd.DataFrame({
2     "cat": ["A", "B", "C", "D", "E"],
3     "value": np.exp(range(5))
4 })
5
6 plt.figure(figsize=(4, 6), layout="constrained")
7
8 plt.subplot(321)
9 plt.scatter("cat", "value", data=df)
10 plt.subplot(322)
11 plt.scatter("value", "cat", data=df)
12
13 plt.subplot(323)
14 plt.plot("cat", "value", data=df)
15 plt.subplot(324)
16 plt.plot("value", "cat", data=df)
17
18 plt.subplot(325)
19 b = plt.bar("cat", "value", data=df)
20 plt.subplot(326)
21 b = plt.bar("value", "cat", data=df)
22
23 plt.show()
```



# Histograms

```
1 df = pd.DataFrame({
2     "x1": np.random.normal(size=100),
3     "x2": np.random.normal(1,2, size=100)
4 })
5
6 plt.figure(figsize=(4, 6), layout="constrained")
7
8 plt.subplot(311)
9 h = plt.hist("x1", bins=10, data=df, alpha=0.5)
10 h = plt.hist("x2", bins=10, data=df, alpha=0.5)
11
12 plt.subplot(312)
13 h = plt.hist(df, alpha=0.5)
14
15 plt.subplot(313)
16 h = plt.hist(df, stacked=True, alpha=0.5)
17
18 plt.show()
```



# Boxplots

```
1 df = pd.DataFrame({
2     "x1": np.random.normal(size=100),
3     "x2": np.random.normal(1,2, size=100),
4     "x3": np.random.normal(-1,3, size=100)
5 }).melt()
6
7 plt.figure(figsize=(4, 4), layout="constrained")
8
9 plt.boxplot("value", positions="variable", data=df)
```

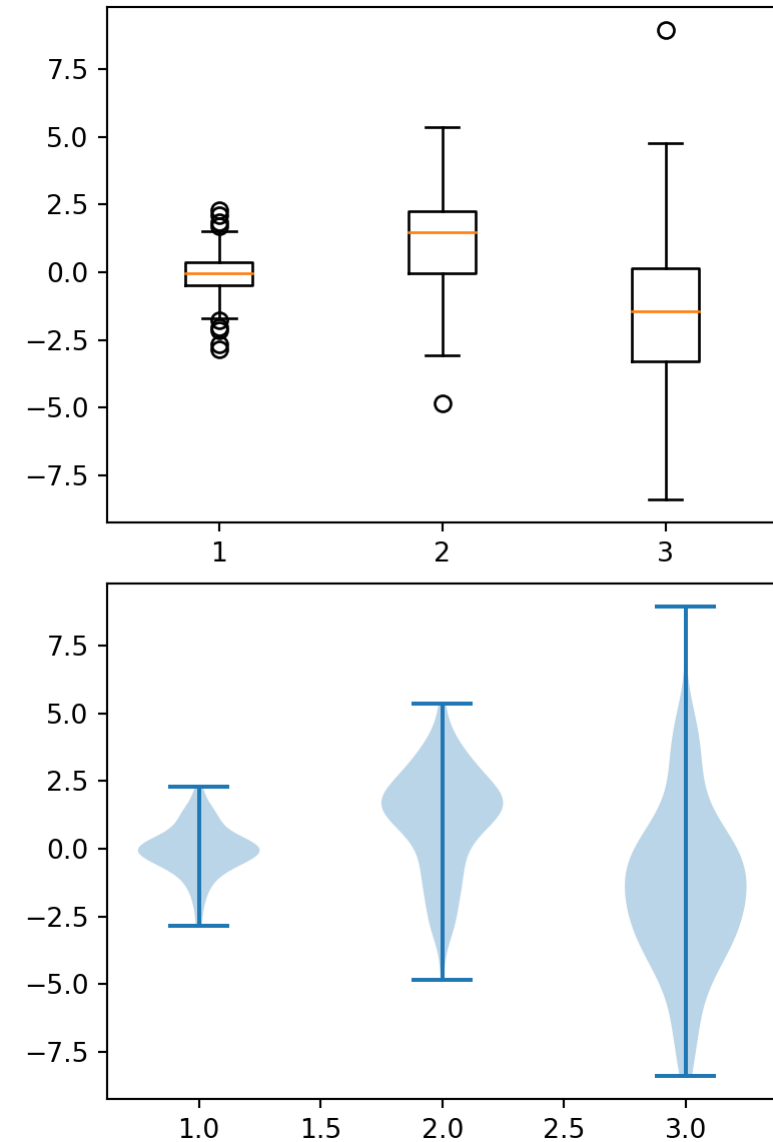
Error: ValueError: List of boxplot statistics and `positions` values must have same the length

```
1 plt.boxplot(df.value, positions=df.variable)
```

Error: ValueError: List of boxplot statistics and `positions` values must have same the length

# Boxplots (cont.)

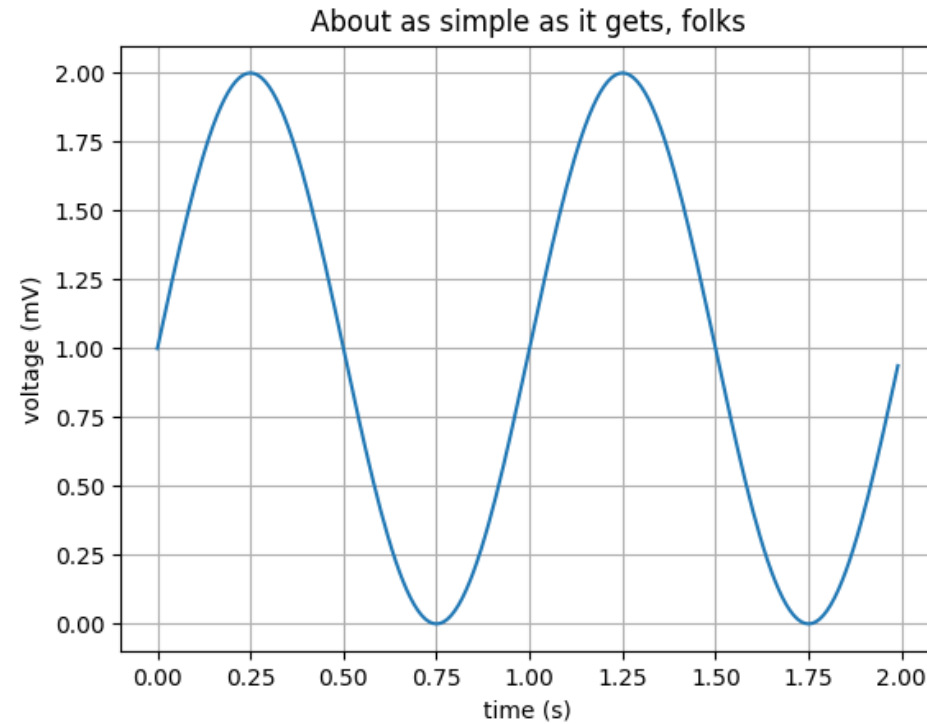
```
1 df = pd.DataFrame({
2     "x1": np.random.normal(size=100),
3     "x2": np.random.normal(1,2, size=100),
4     "x3": np.random.normal(-1,3, size=100)
5 })
6
7 plt.figure(figsize=(4, 6), layout="constrained")
8
9 plt.subplot(211)
10 b = plt.boxplot(df)
11
12 plt.subplot(212)
13 v = plt.violinplot(df)
14
15 plt.show()
```



# Other Plot Types

# Exercise 1

To the best of your ability recreate the following plot,

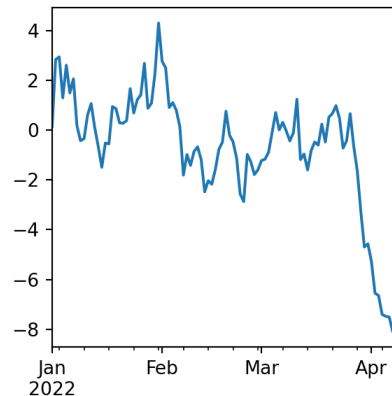


# Plotting with pandas

# plot methods

Both Series and DataFrame objects have a plot method which can be used to create visualizations - dtypes determine the type of plot produced. Note these are just pyplot plots and can be formatted as such.

```
1 s = pd.Series(  
2     np.cumsum( np.random.normal(size=100) ),  
3     index = pd.date_range("2022-01-01", periods=100, freq="D")  
4 )  
5  
6 plt.figure(figsize=(3,3), layout="constrained")  
7 s.plot()  
8 plt.show()
```

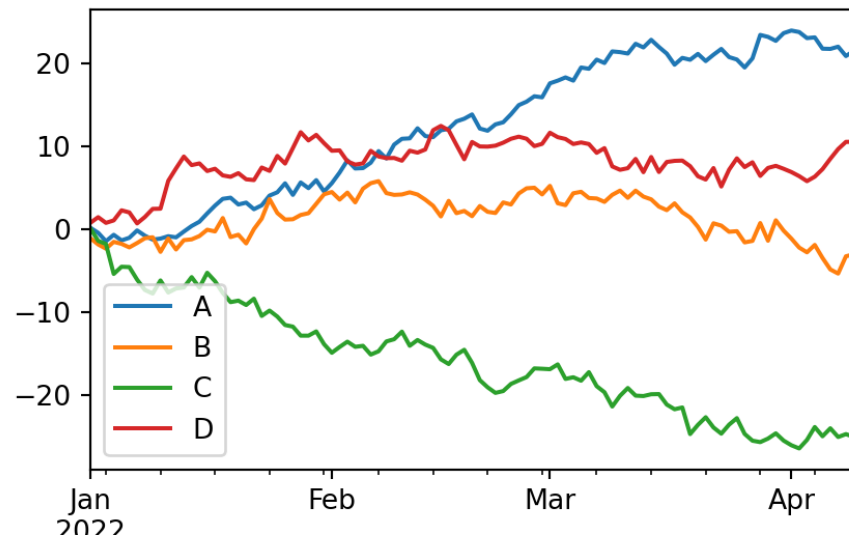


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# DataFrame plot

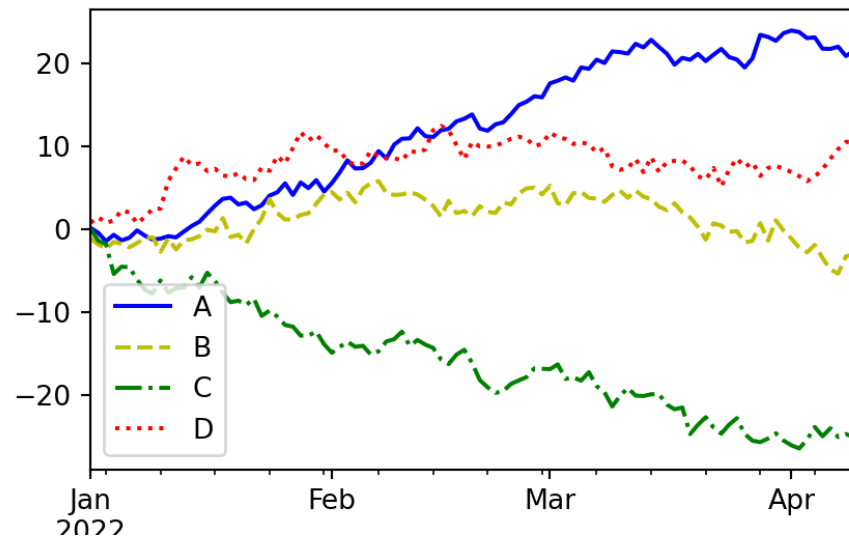
```
1 df = pd.DataFrame(  
2     np.cumsum( np.random.normal(size=(100,4)), axis=0),  
3     index = pd.date_range("2022-01-01", periods=100, freq="D"),  
4     columns = list("ABCD")  
5 )  
6  
7 plt.figure(layout="constrained")  
8 df.plot(figsize=(5,3))  
9 plt.show()
```



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# DataFrame line styles

```
1 df.plot(  
2     figsize=(5,3),  
3     style = {  
4         "A": "-b",  
5         "B": "--y",  
6         "C": "-.g",  
7         "D": ":r"  
8     }  
9 )
```



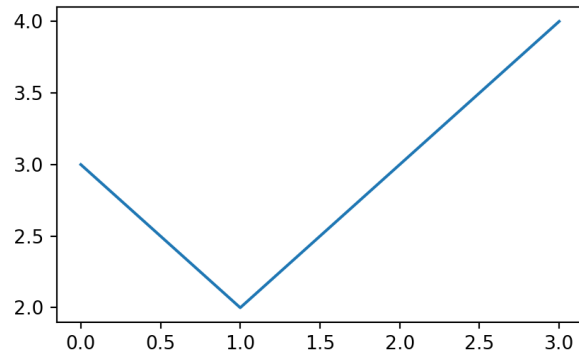
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# DataFrame plot - categorical

```

1 df = pd.DataFrame({
2     "x": list("ABCD"),
3     "y": np.random.poisson(lam=2, size=4)
4 })
5
6 df.plot(figsize=(5,3), legend=False)

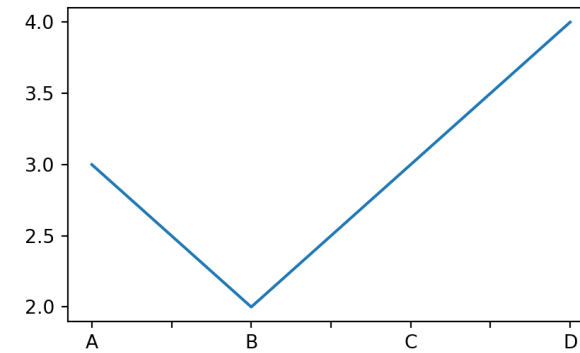
```



```

1 df.set_index("x").plot(figsize=(5,3), legend=False)

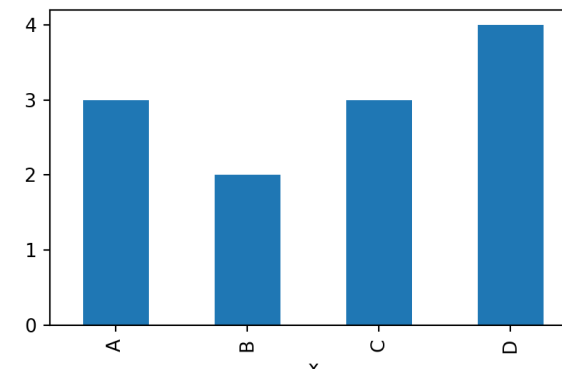
```



```

1 df.set_index("x").plot(
2     figsize=(5,3), kind="bar", legend=False
3 )

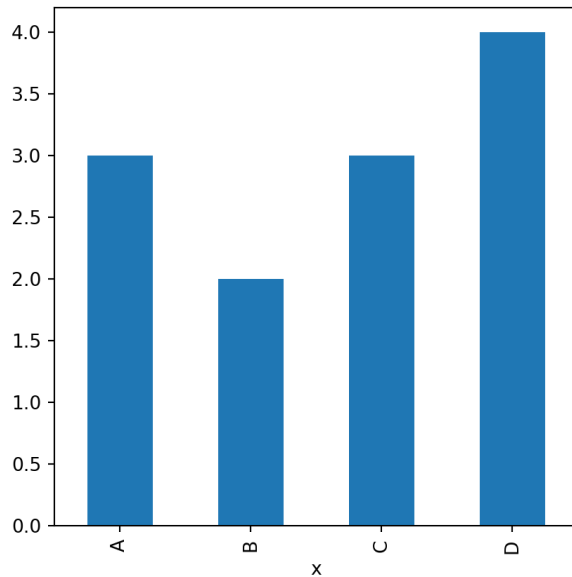
```



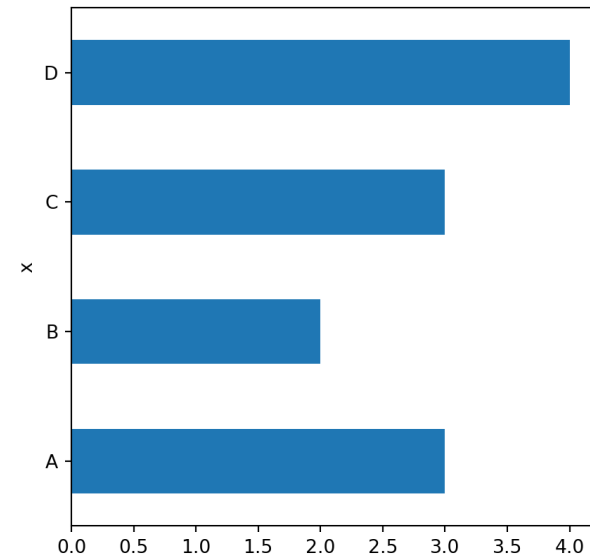
# Other plot types

Plot types can be changed via the `kind` argument or using one of the `DataFrame.plot.<kind>` method,

```
1 df.set_index("x").plot.bar(  
2     legend=False, figsize=(5,5)  
3 )
```



```
1 df.set_index("x").plot.barh(  
2     legend=False, figsize=(5,5)  
3 )
```

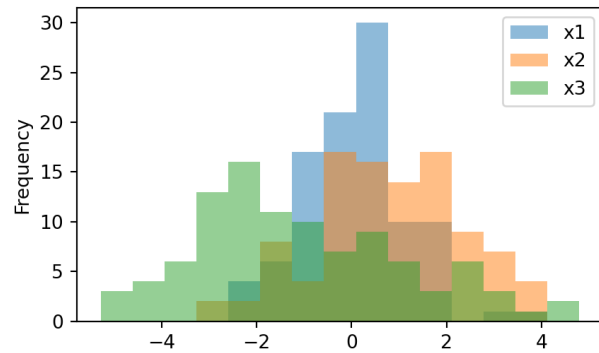


# Wide vs long - histograms

```

1 df = pd.DataFrame({
2     "x1": np.random.normal(size=100),
3     "x2": np.random.normal(1,1.5, size=100),
4     "x3": np.random.normal(-1,2, size=100)
5 })
6
7 df.plot.hist(figsize=(5,3), alpha=0.5, bins=15)

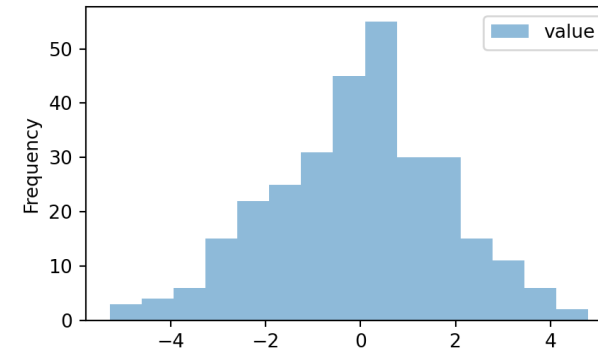
```



```

1 df_wide = df.melt()
2 df_wide.plot.hist(figsize=(5,3), alpha=0.5, bins

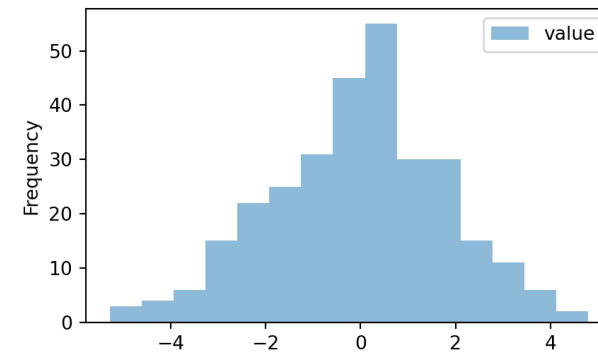
```



```

1 df_wide.set_index("variable").plot.hist(
2     figsize=(5,3), alpha=0.5, bins=15
3 )

```



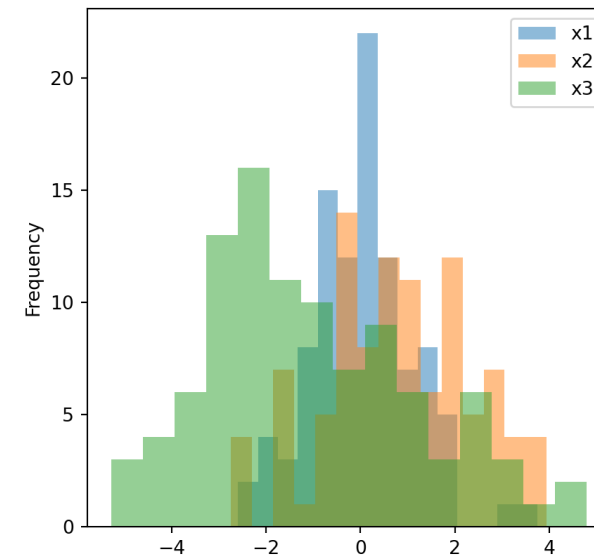
# plot and groupby



```
1 df_wide
```

	variable	value
0	x1	1.471225
1	x1	-0.178315
2	x1	0.156732
3	x1	0.291983
4	x1	1.593502
..	...	...
295	x3	-3.202525
296	x3	-4.066616
297	x3	0.095091
298	x3	-2.446253
299	x3	-1.810505

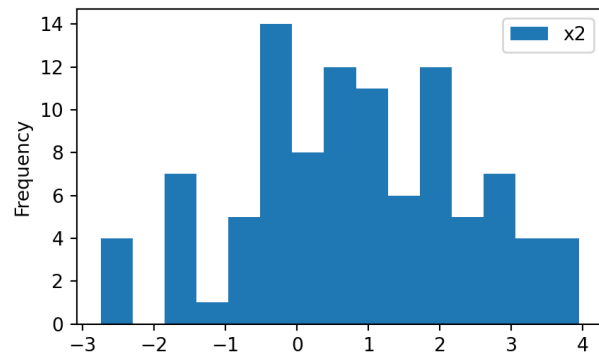
```
1 plt.figure(figsize=(5,5))
2
3 h = ( df_wide
4       .groupby("variable")["value"]
5       .plot.hist(
6         alpha=0.5, legend=True, bins=15
7       ) )
8
9
10 plt.show()
```



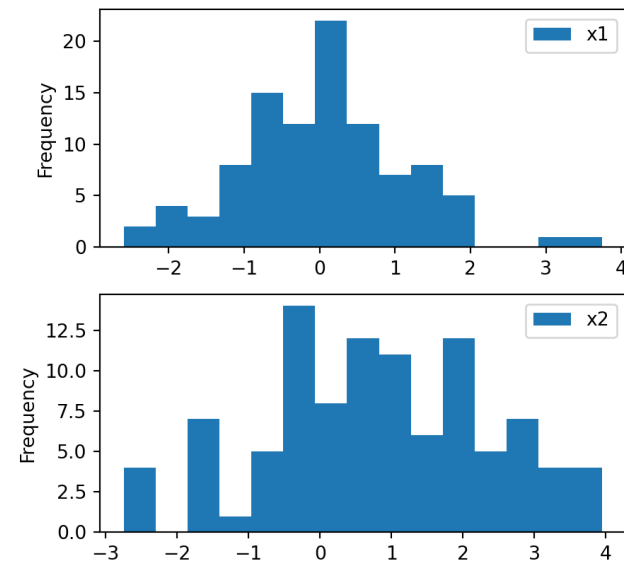
Here we are plotting Series objects hence the need to use

# pandas and subplots

```
1 plt.figure(figsize=(5,3))
2 plt.subplot(211)
3 df[["x1"]].plot.hist(bins=15, figsize=(5,3))
4 plt.subplot(212)
5 df[["x2"]].plot.hist(bins=15, figsize=(5,3))
6
7 plt.show()
```



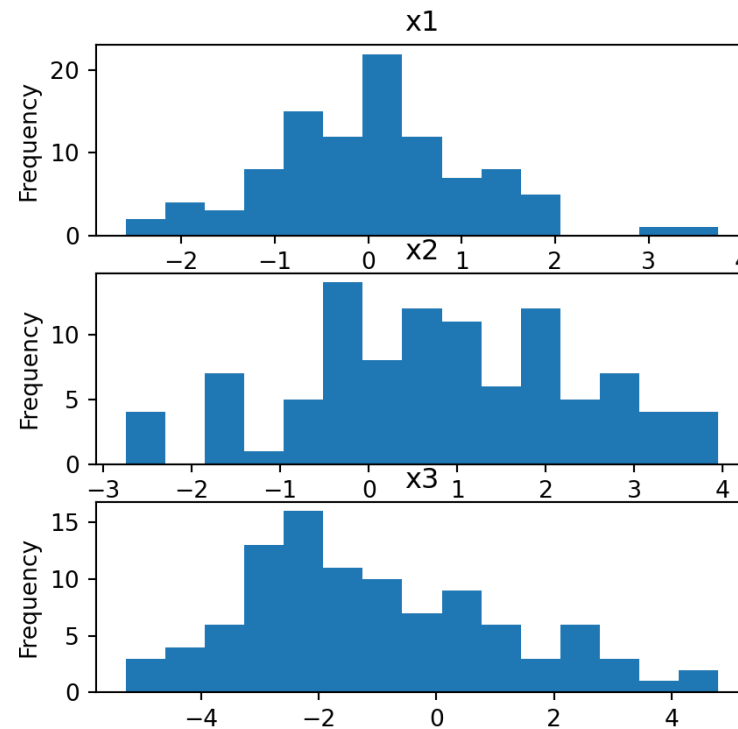
```
1 fig, (ax1, ax2) = plt.subplots(2,1, figsize=(5,5))
2
3 df[["x1"]].plot.hist(ax = ax1, bins=15)
4 df[["x2"]].plot.hist(ax = ax2, bins=15)
5
6 plt.show()
```



# Using by

```
1 df_wide.plot.hist(bins=15, by="variable", legend=False, figsize=(5,5))
```

```
1 plt.show()
```



# Higher level plots - pair plot

The pandas library also provides the `plotting` submodule with several useful higher level plots,

```

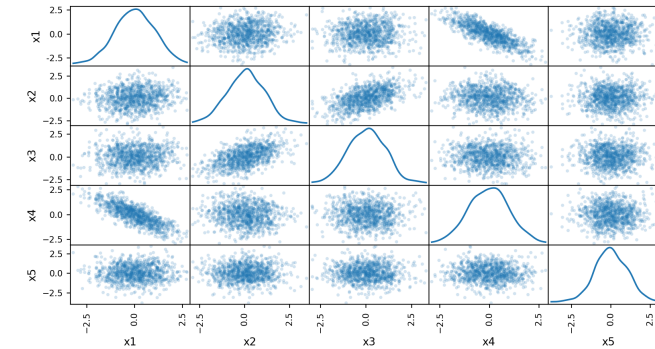
1 cov = np.identity(5)
2 cov[1,2] = cov[2,1] = 0.5
3 cov[3,0] = cov[0,3] = -0.8
4
5 df = pd.DataFrame(
6     np.random.multivariate_normal(mean=[0]*5, cov=
7     columns = ["x1", "x2", "x3", "x4", "x5"]
8 )
9
10 df

```

	x1	x2	x3	x4	x5
0	-0.675512	-0.072846	0.536075	-0.480851	0.828583
1	0.867784	-0.099565	0.014922	-1.403662	0.465785
2	0.028221	-1.572683	-2.679542	-1.030949	-0.655153
3	0.434528	-0.570881	0.446828	-0.424219	-1.336715
4	0.320779	0.294548	0.834556	-0.261610	-0.648069
..	...	...	...	...	...
...					
995	-0.642675	1.500878	0.244987	0.472639	0.444908
996	1.481997	0.902982	1.271029	-1.003460	

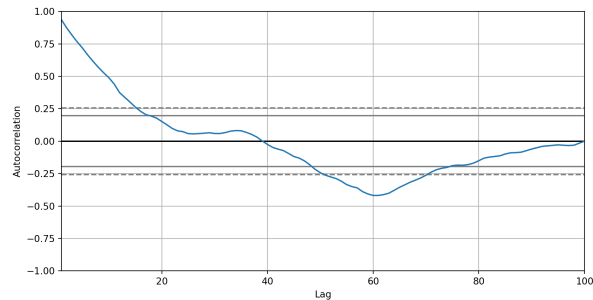
```
1 pd.plotting.scatter_matrix(df, alpha=0.2, diagonal
```

```
1 plt.show())
```

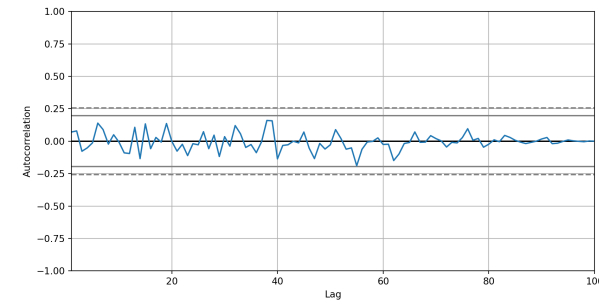


# Autocorrelation plots

```
1 rw = pd.Series(  
2     np.cumsum( np.random.normal(size=  
3 )  
4  
5 pd.plotting.autocorrelation_plot(rw  
6 plt.show()
```



```
1 wn = pd.Series(  
2     np.random.normal(size=100),  
3 )  
4  
5 pd.plotting.autocorrelation_plot(wn  
6 plt.show()
```



# Other plots

```
1 print(  
2     dir(pd.plotting)  
3 )
```

```
['PlotAccessor', '__all__', '__builtins__', '__cached__', '__doc__', '__file__',  
 '__loader__', '__name__', '__package__', '__path__', '__spec__', '_core',  
 '_matplotlib', '_misc', 'andrews_curves', 'autocorrelation_plot',  
 'bootstrap_plot', 'boxplot', 'boxplot_frame', 'boxplot_frame_groupby',  
 'deregister_matplotlib_converters', 'hist_frame', 'hist_series', 'lag_plot',  
 'parallel_coordinates', 'plot_params', 'radviz',  
 'register_matplotlib_converters', 'scatter_matrix', 'table']
```

