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*Python for Data Analytics*

# Matplotlib



# Outline

- Introduction to Data Visualization
- What is Matplotlib?
- Pyplot submodule functions
- What is Seaborn?

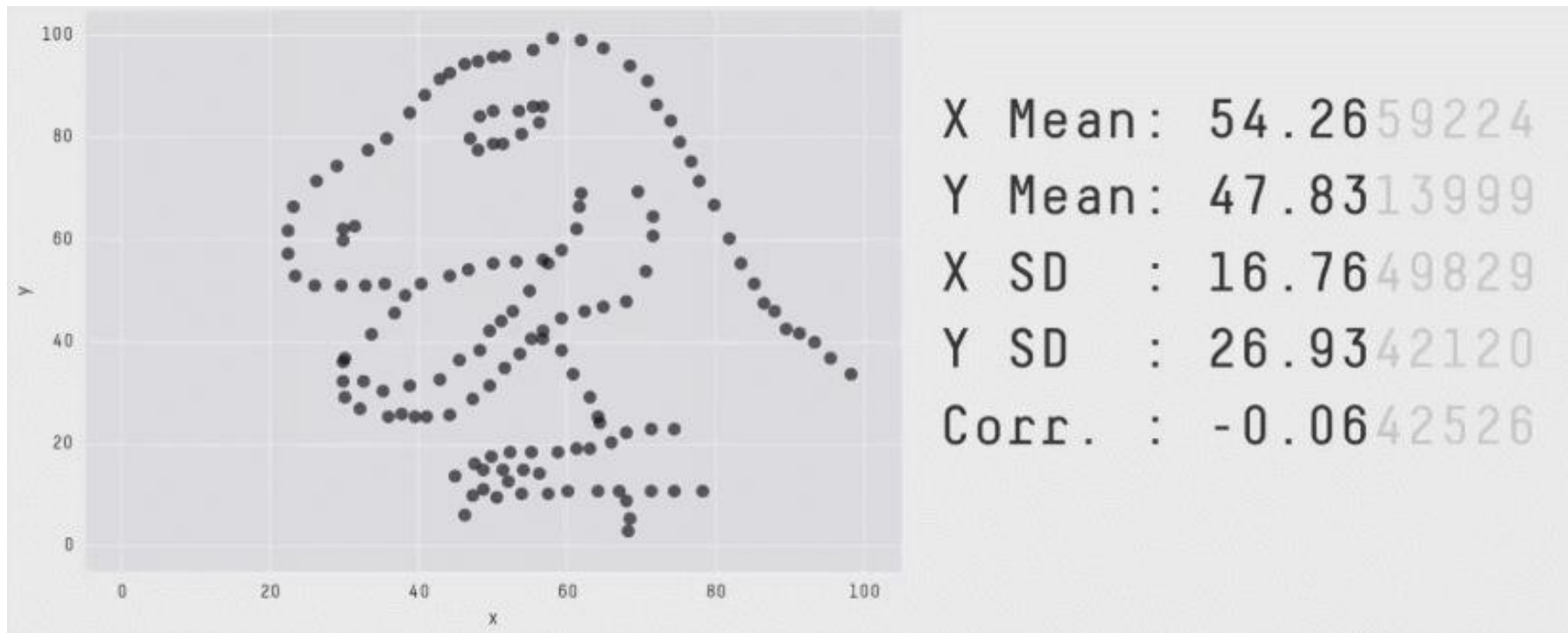
# Introduction to Data Visualization

# Visualization

- Data exploration visualization: figuring out what is true
- Data presentation visualization: convincing other people it is true
- Before you run any analysis, build any machine learning system, etc., always visualize your data
- If you can't identify a trend or make a prediction for your dataset, neither will an automated algorithm

# Visualization vs. Statistics

- Visualization almost always presents a more informative (though less quantitative) view of your data than statistics
  - This is a mathematical property:  $n$  data points and  $m$  equations to satisfy, with  $n > m$



# Data Types

## ■ Nominal

- Categorical data
- No quantitative value, no ordering
- Example – Pet: {dog, cat, rabbit, ...}
- Operations:  $=, \neq$

## ■ Ordinal

- Categorical data, with ordering
- Example – Rating: {1, 2, 3, 4, 5}
- Operations:  $=, \neq, \geq, \leq, >, <$

## ■ Interval

- Numerical data
- Zero has no fixed meaning
- Example – Temperature Celsius
- Operations:  $=, \neq, \geq, \leq, >, <, +, -$

## ■ Ratio

- Numerical data
- Zero has special meaning
- Example – Height, Weight, ...
- Operations:  $=, \neq, \geq, \leq, >, <, +, -, \times, \div$

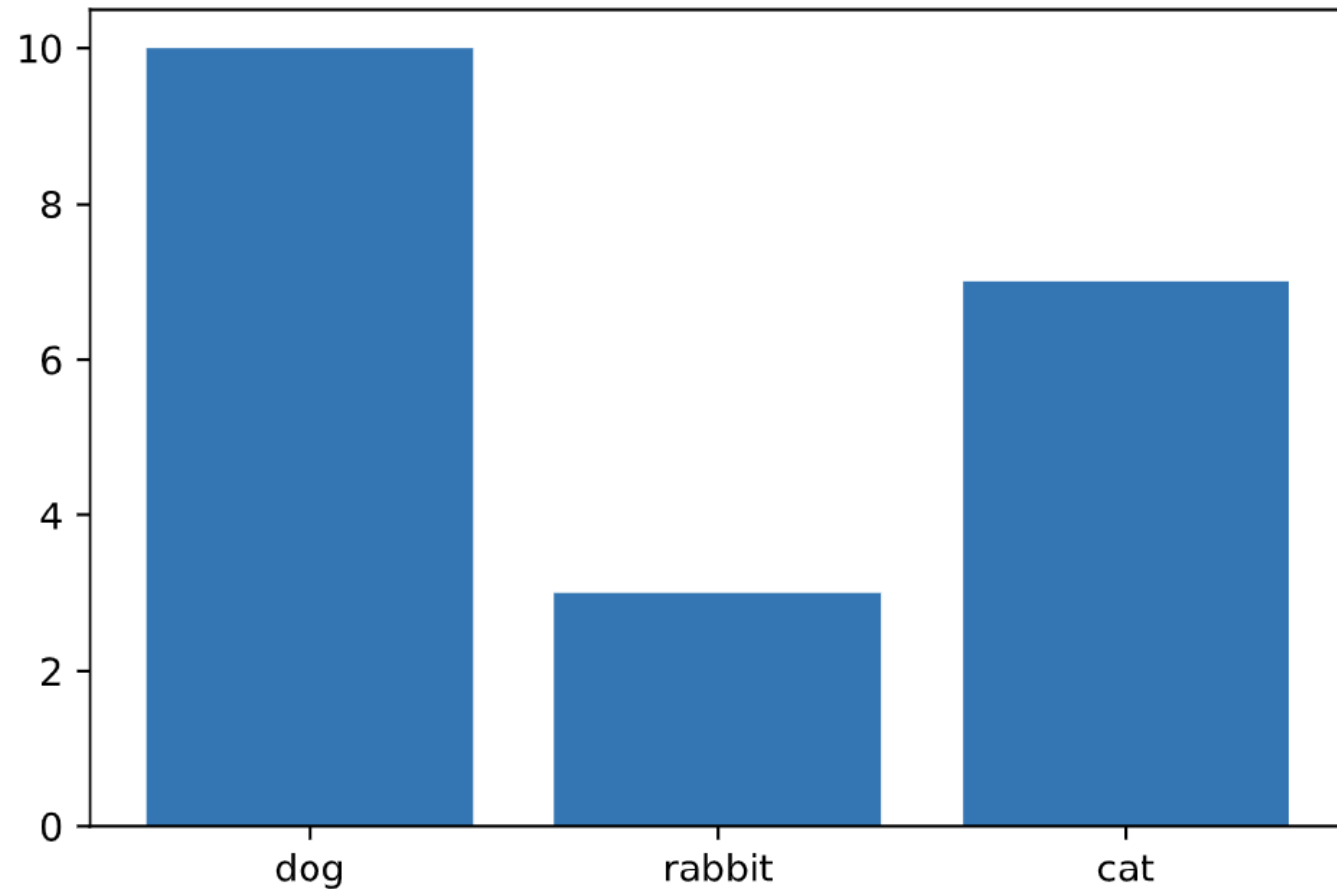
# Visualization Types

- 1D
  - Bar chart, pie chart, histogram
- 2D
  - Scatter plot, line plot, box and whisker plot, heatmap
- 3D
  - Scatter matrix, bubble chart

# ID: Bar Chart

	Data
Nominal	✓
Ordinal	✓
Interval	X
Ratio	X

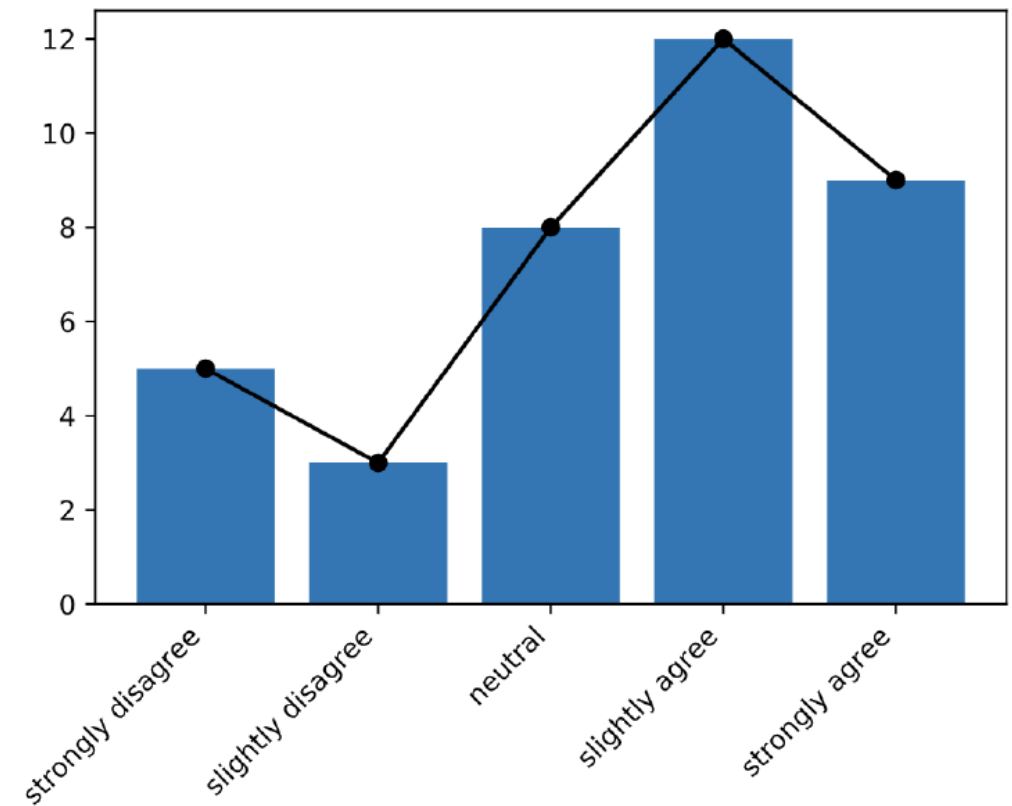
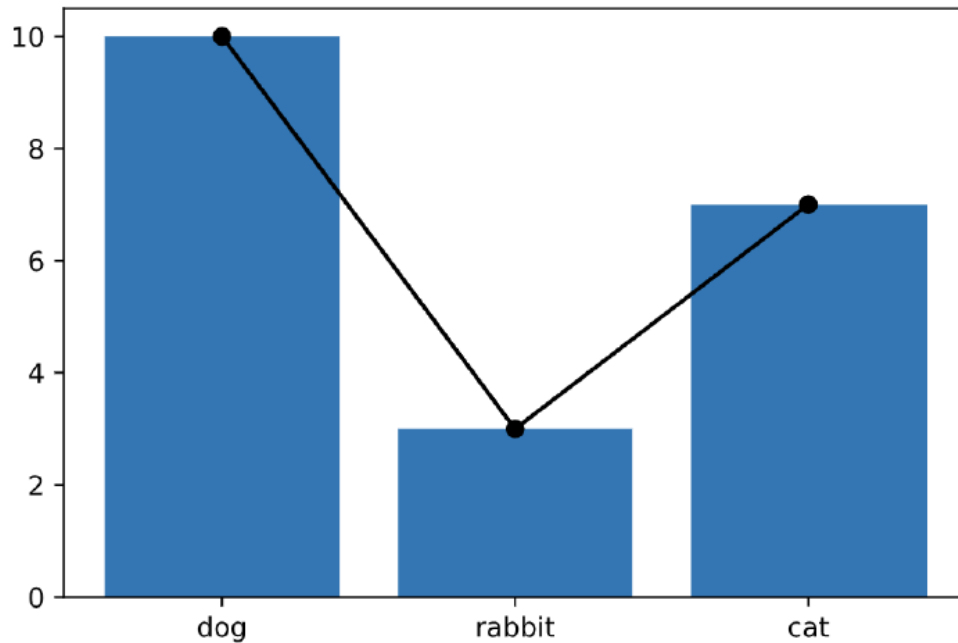
↑  
Suggestions, not rules





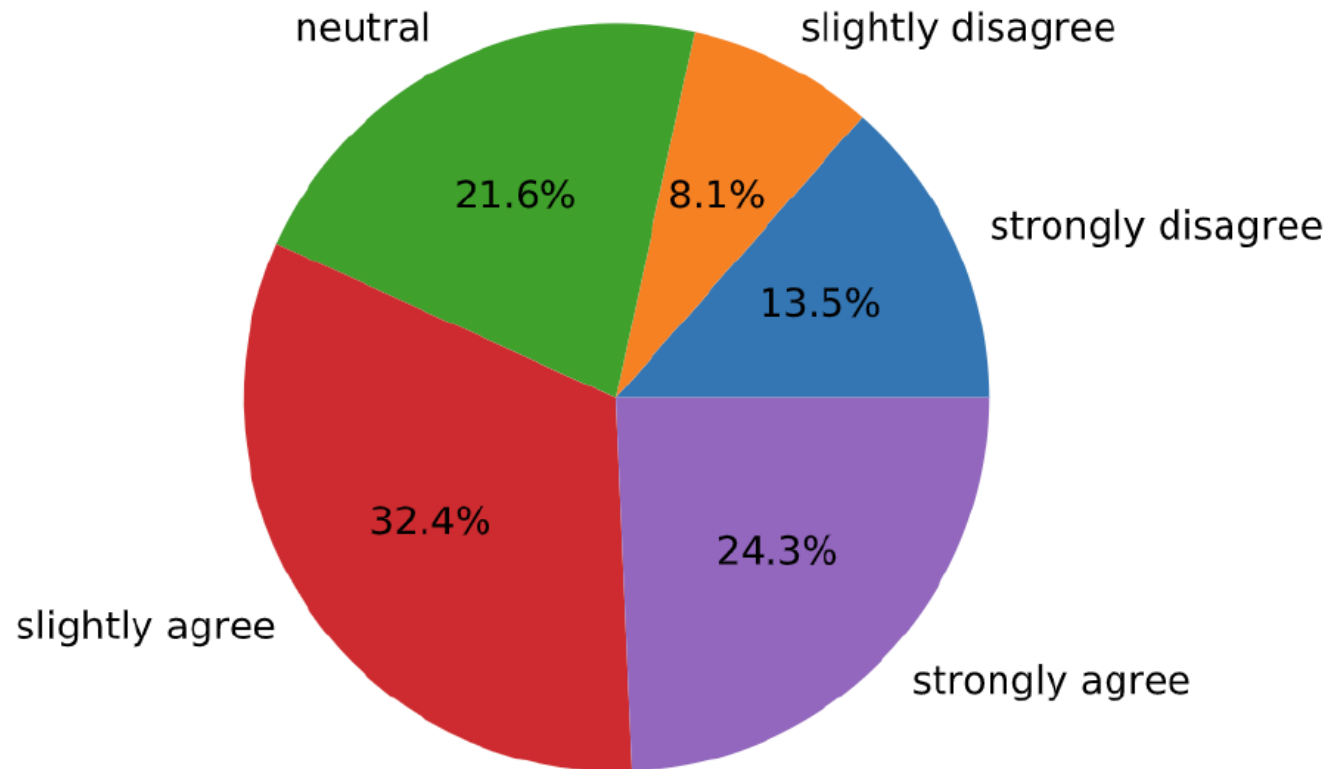
# ID: Bar Chart (bad)

- Don't use lines within a bar chart categorical or ordinal features



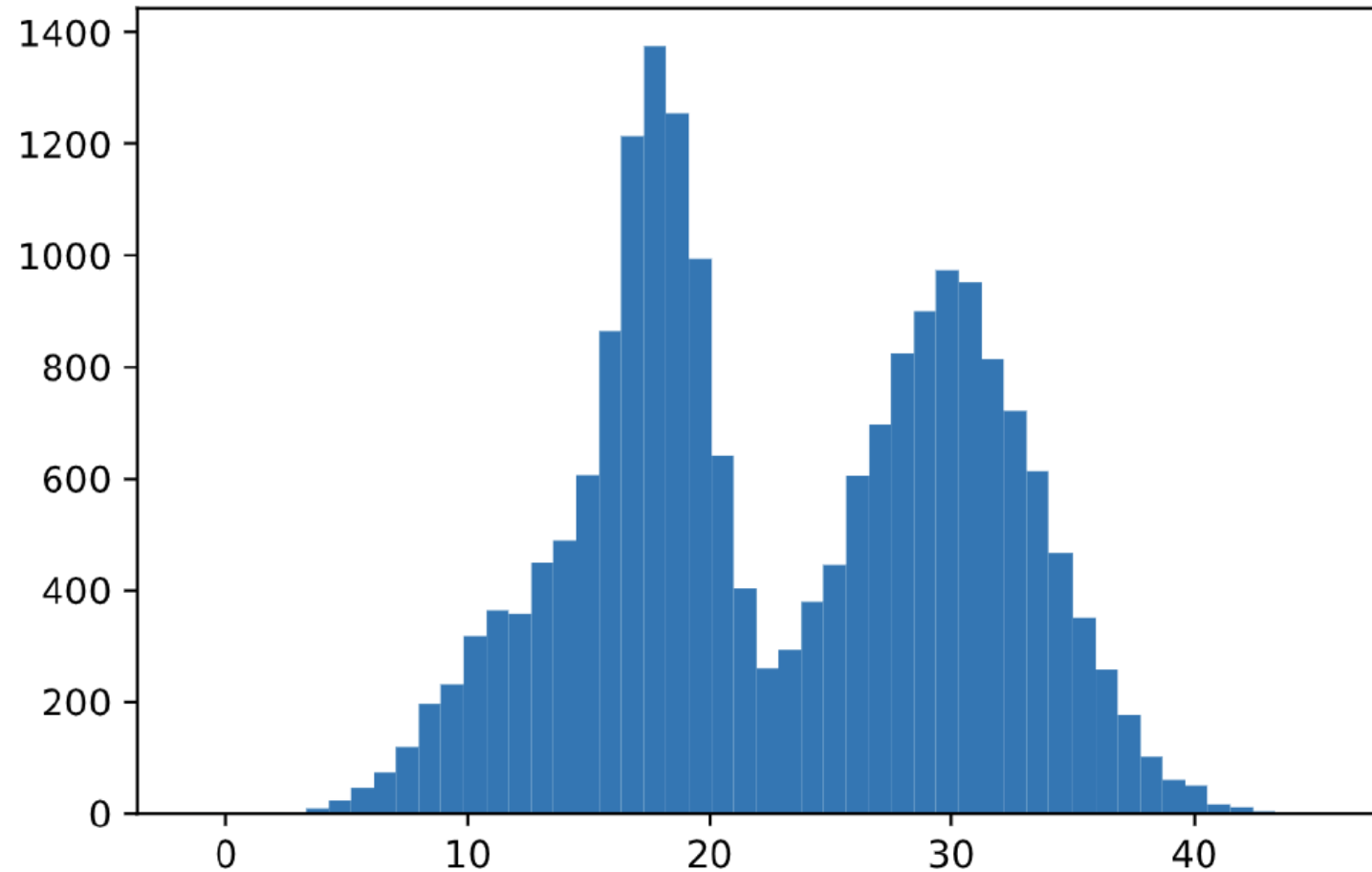
# ID: Pie Chart

	Data
Nominal	✓
Ordinal	✓
Interval	X
Ratio	X



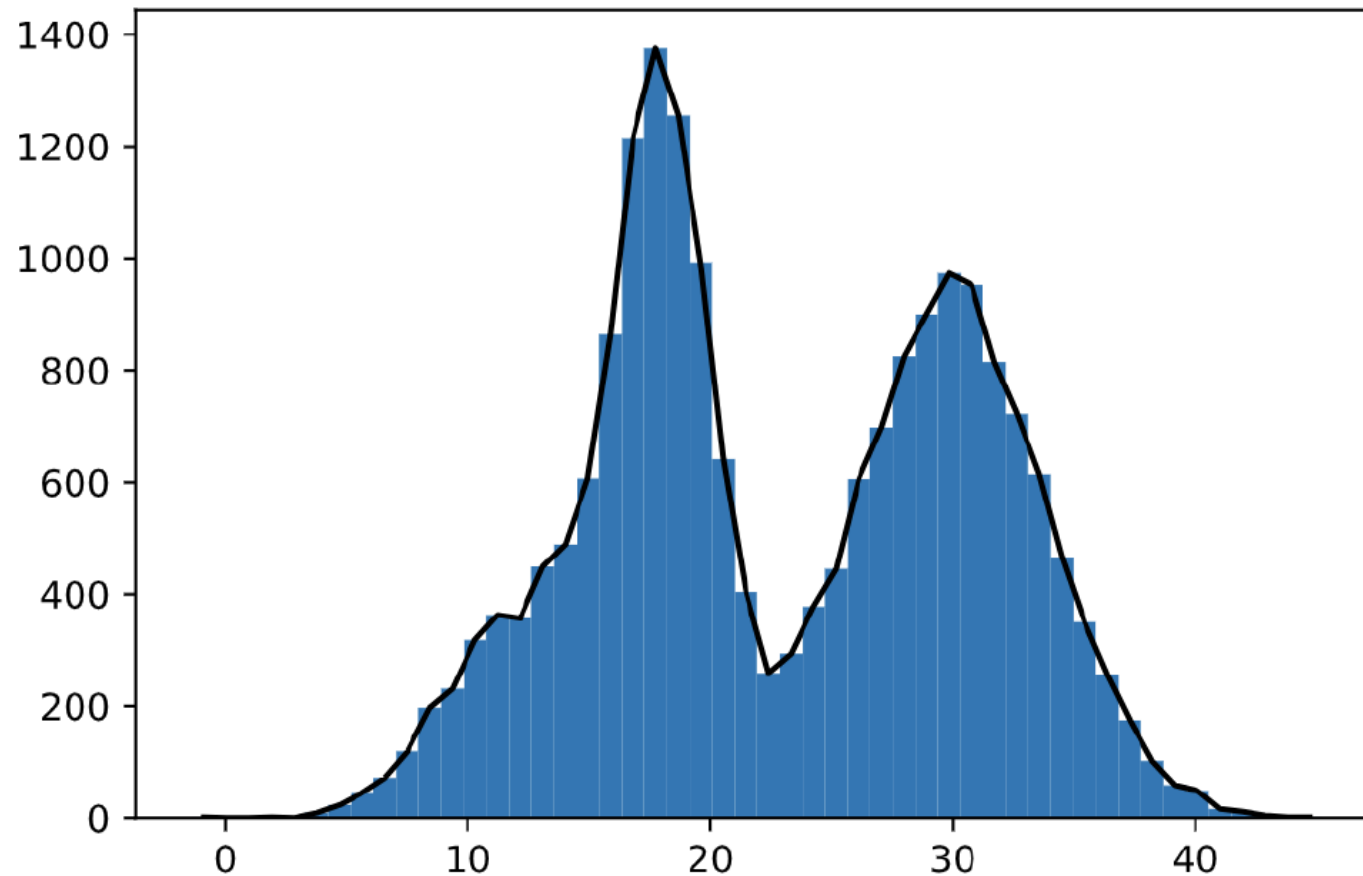
# ID: Histogram

	Data
Nominal	X
Ordinal	X
Interval	✓
Ratio	✓



# ID: Histogram

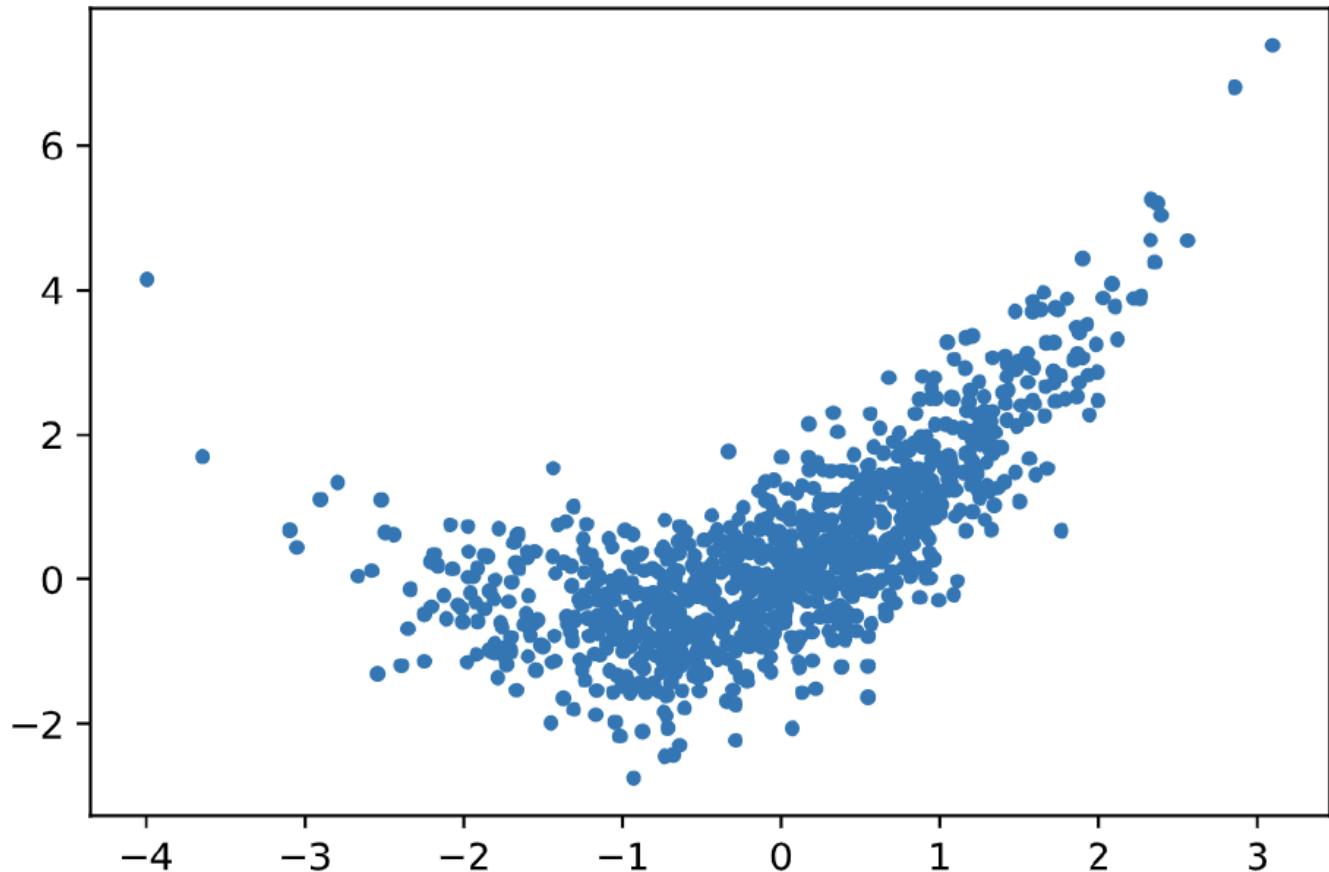
- OK to use lines within a histogram (but not very informative)



# 2D: Scatter Plot

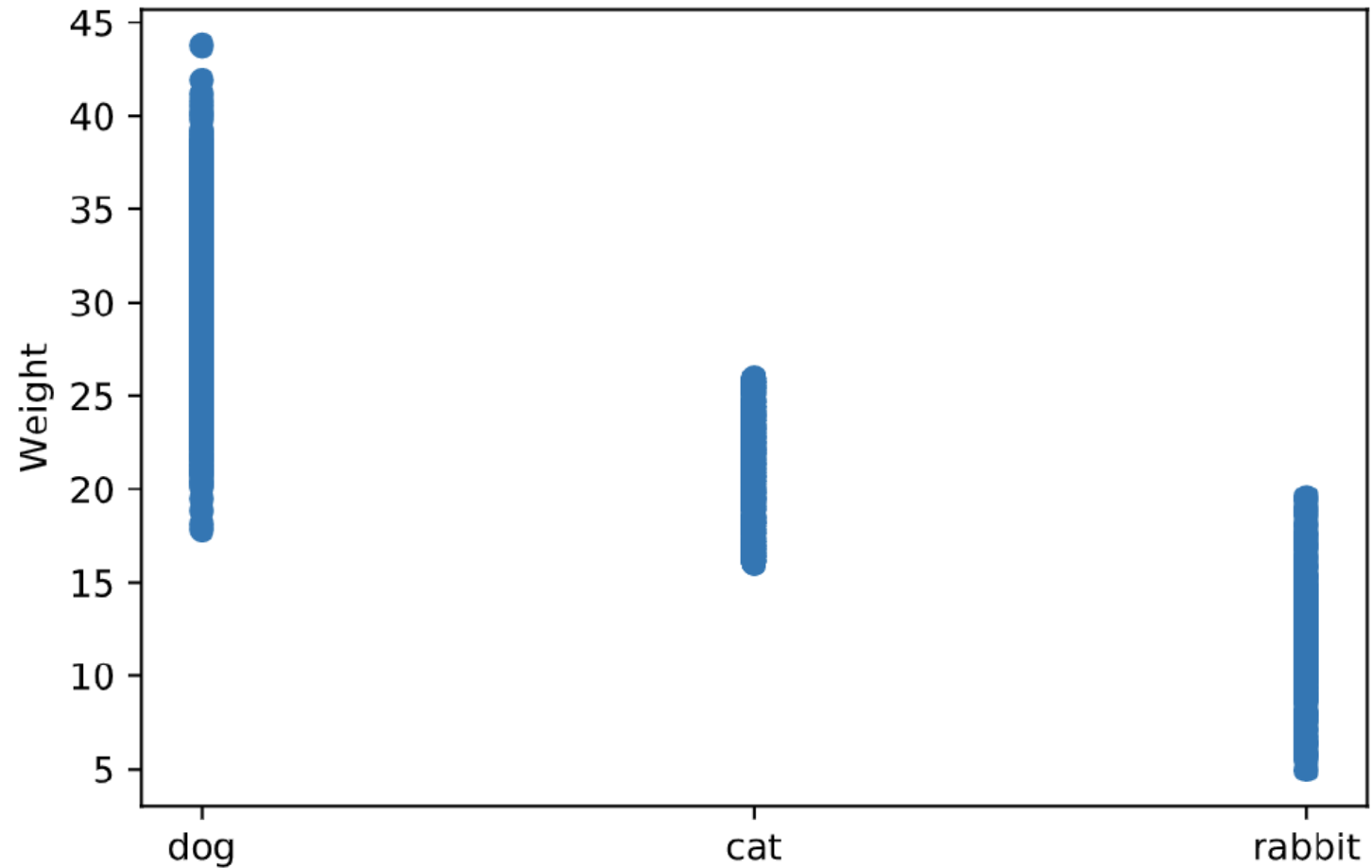
	Dim 1	Dim 2
Nominal	X	X
Ordinal	X	X
Interval	✓	✓
Ratio	✓	✓

Why not ordinal data in first dimension?



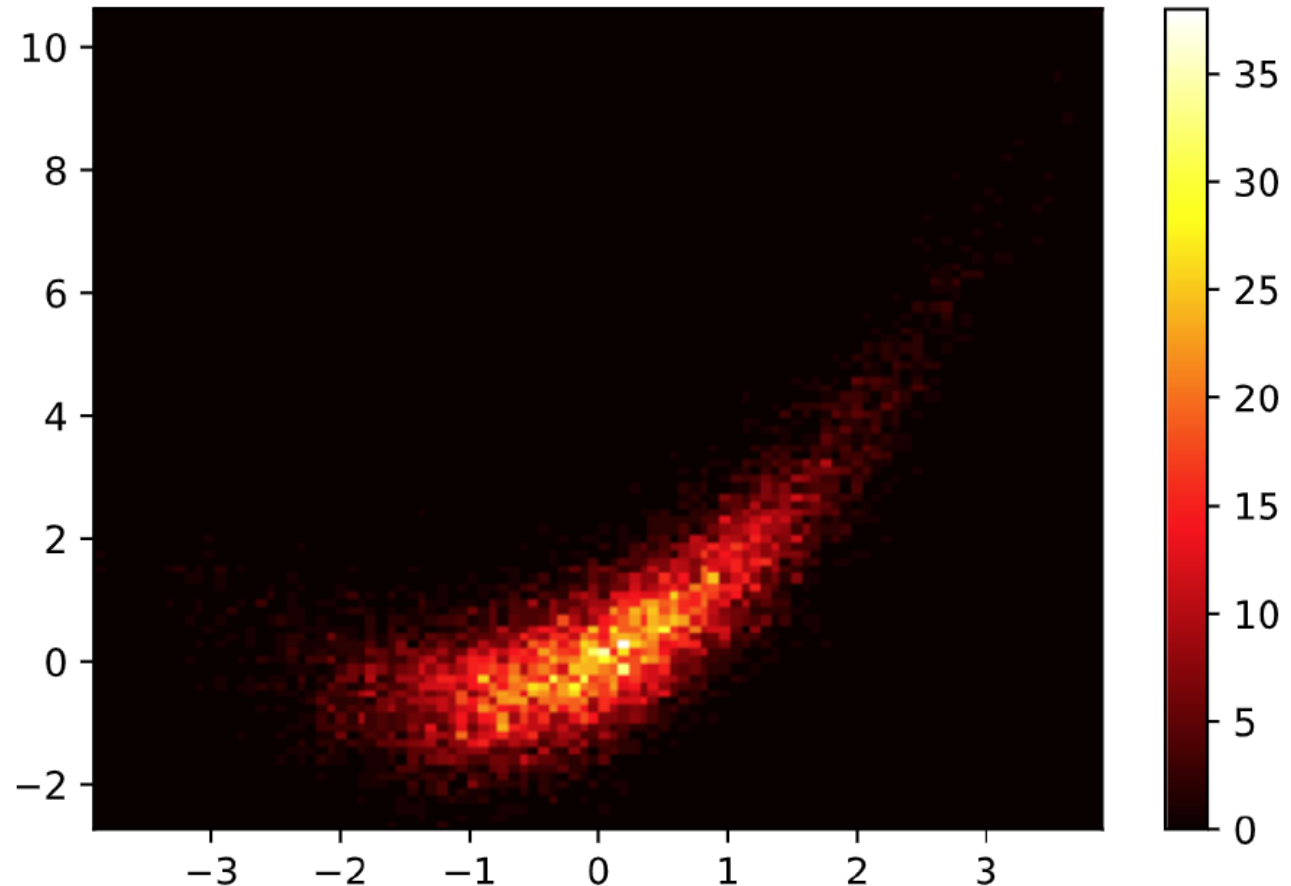
# 2D: Scatter Plot (bad)

	Dim 1	Dim 2
Nominal	X	X
Ordinal	X	X
Interval	✓	✓
Ratio	✓	✓



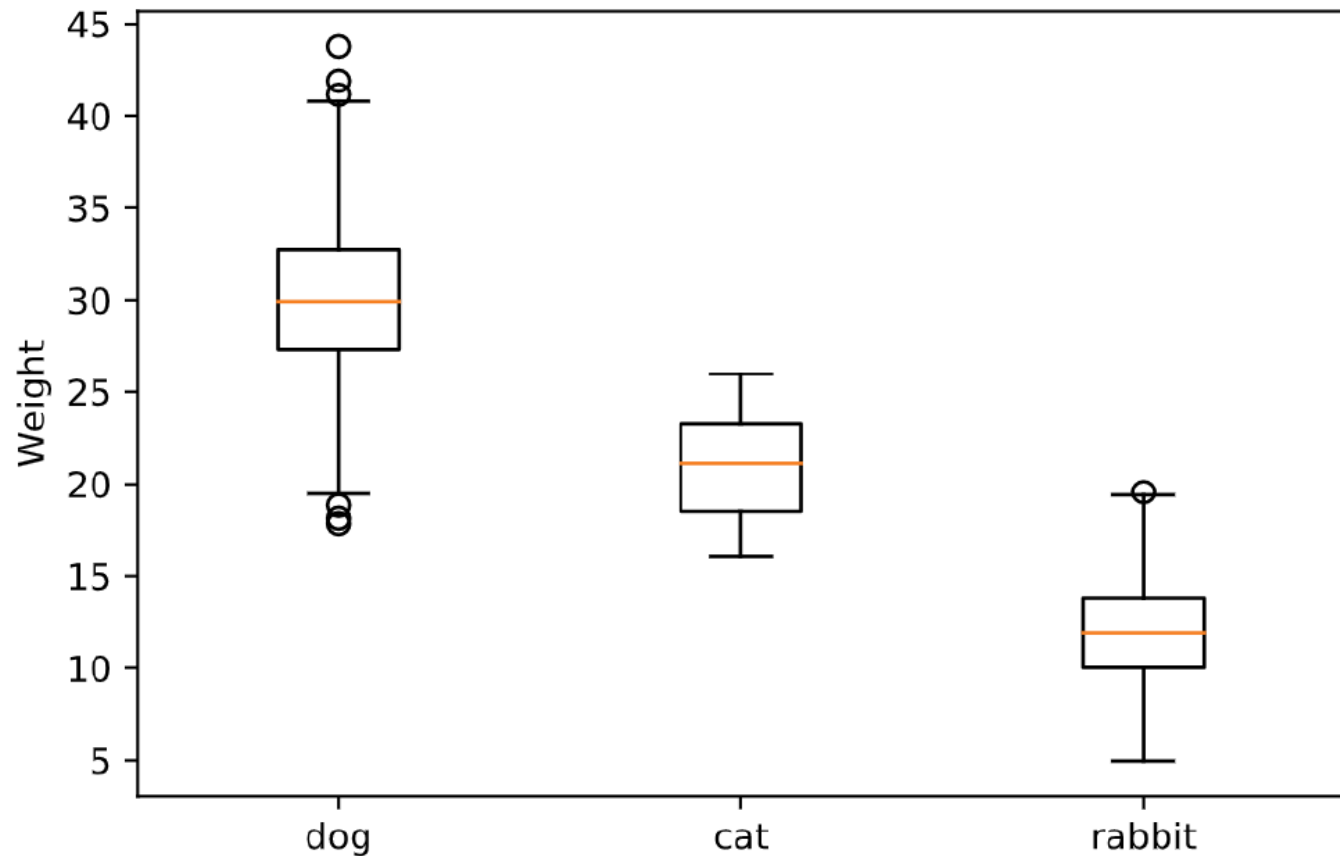
# 2D: Heatmap (density or 2D histogram)

	Dim 1	Dim 2
Nominal	X	X
Ordinal	X	X
Interval	✓	✓
Ratio	✓	✓



# 2D: Box and Whiskers

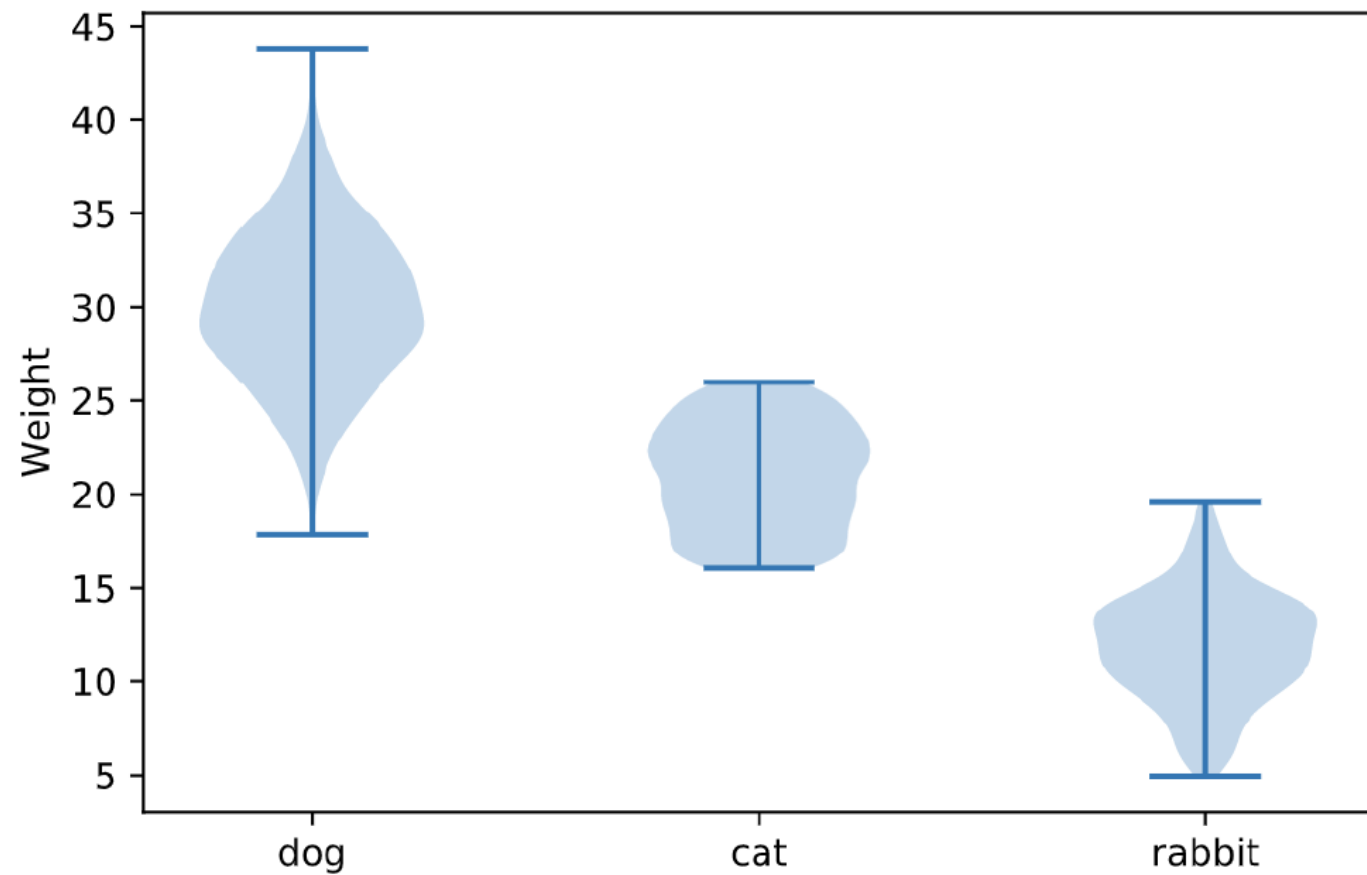
	Dim 1	Dim 2
Nominal	✓	X
Ordinal	✓	X
Interval	X	✓
Ratio	X	✓





# 2D: Violin Plot

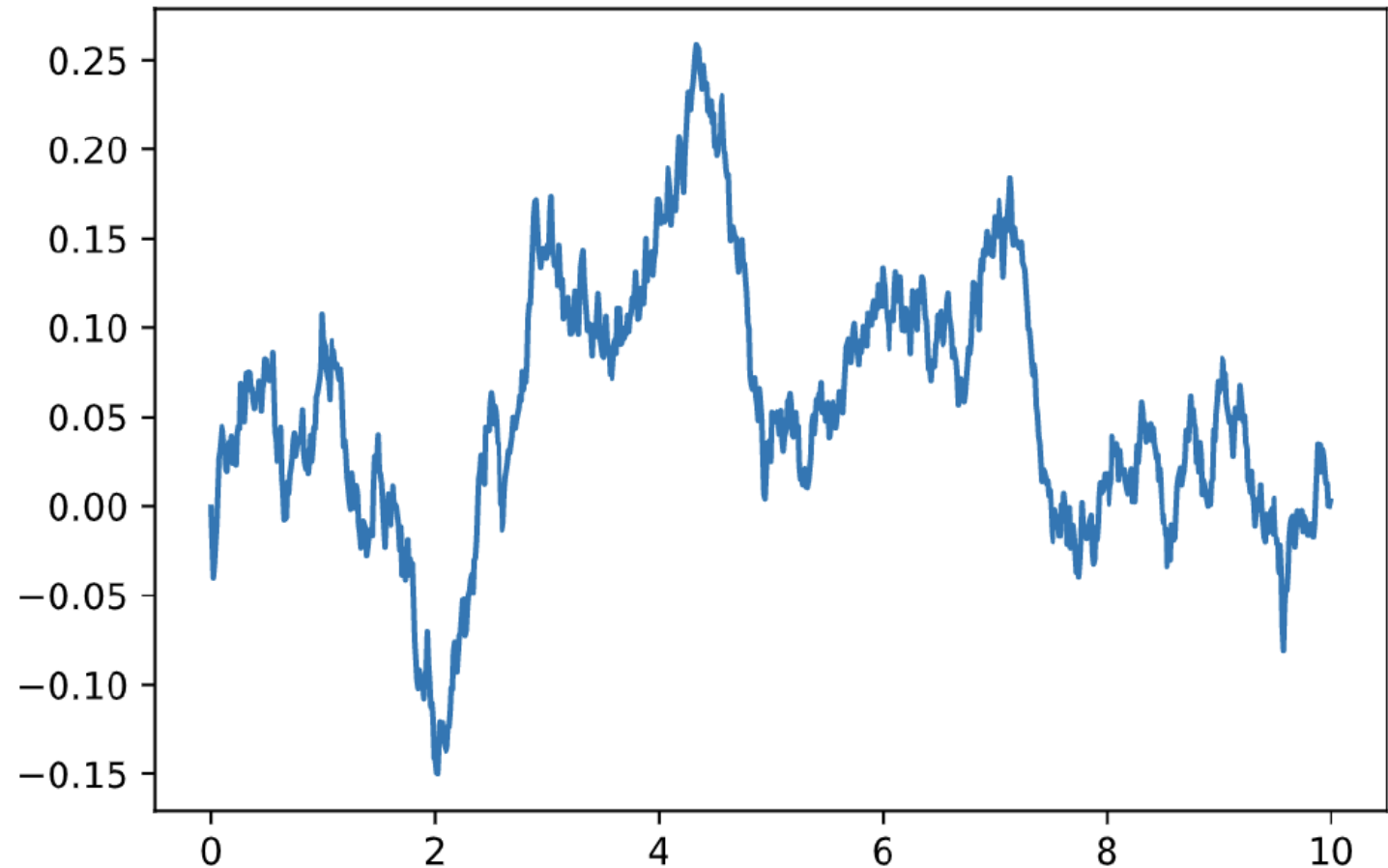
	Dim 1	Dim 2
Nominal	✓	X
Ordinal	✓	X
Interval	X	✓
Ratio	X	✓



# 2D: Line Plot

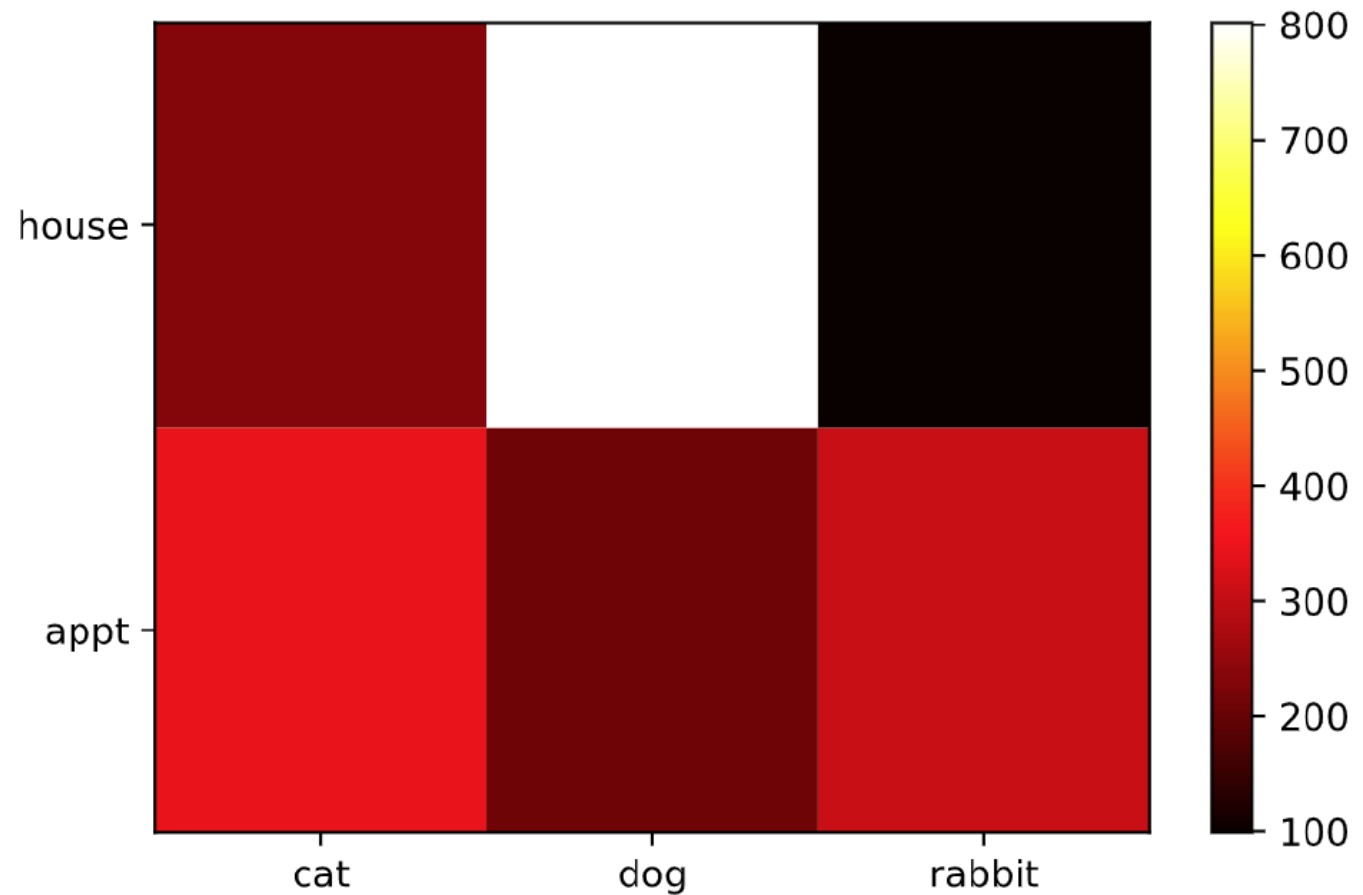
	Dim 1	Dim 2
Nominal	X	X
Ordinal	X	X
Interval	✓	✓
Ratio	✓	✓

Why not ordinal data in first dimension?



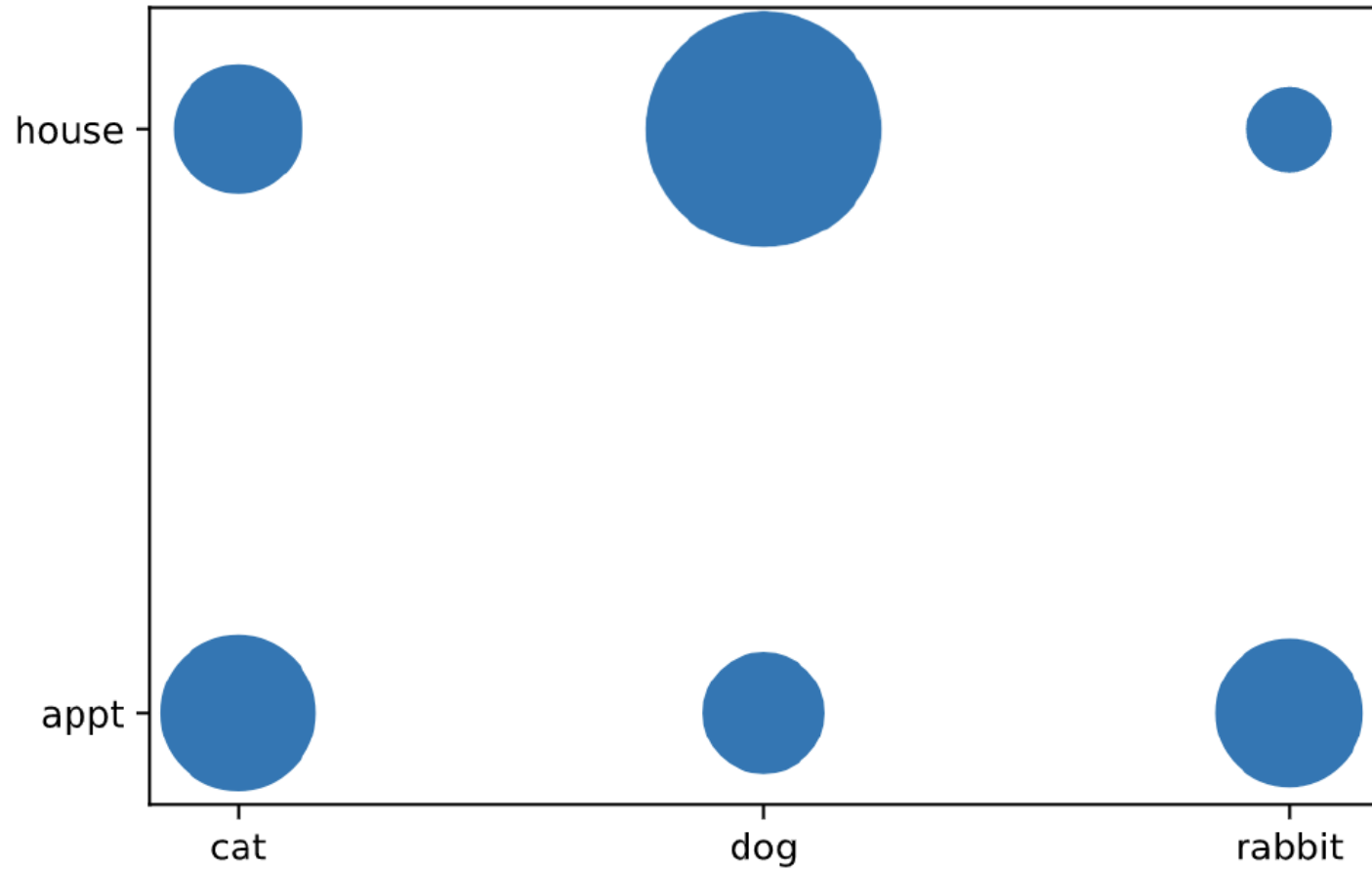
# 2D: Heatmap (matrix)

	Dim 1	Dim 2
Nominal	✓	✓
Ordinal	✓	✓
Interval	X	X
Ratio	X	X



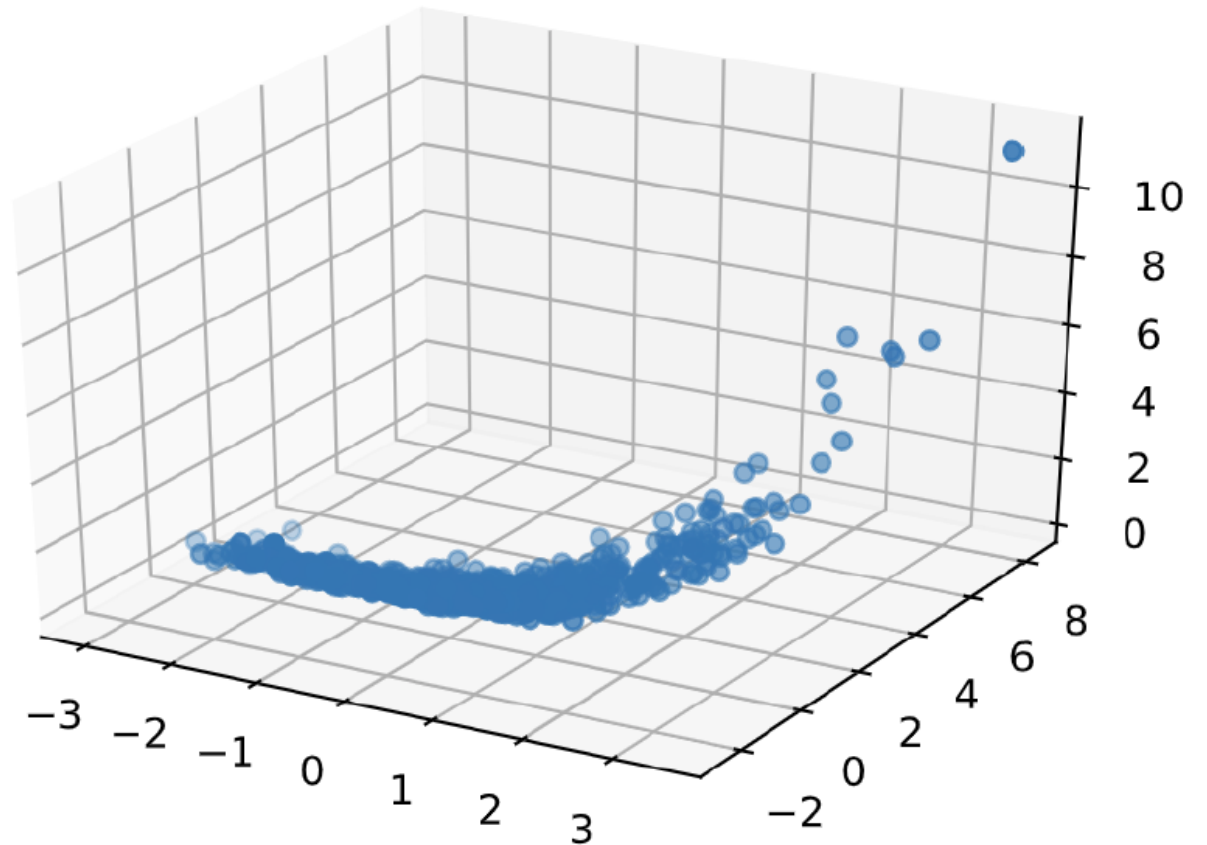
# 2D: Bubble Plot

	Dim 1	Dim 2
Nominal	✓	✓
Ordinal	✓	✓
Interval	X	X
Ratio	X	X



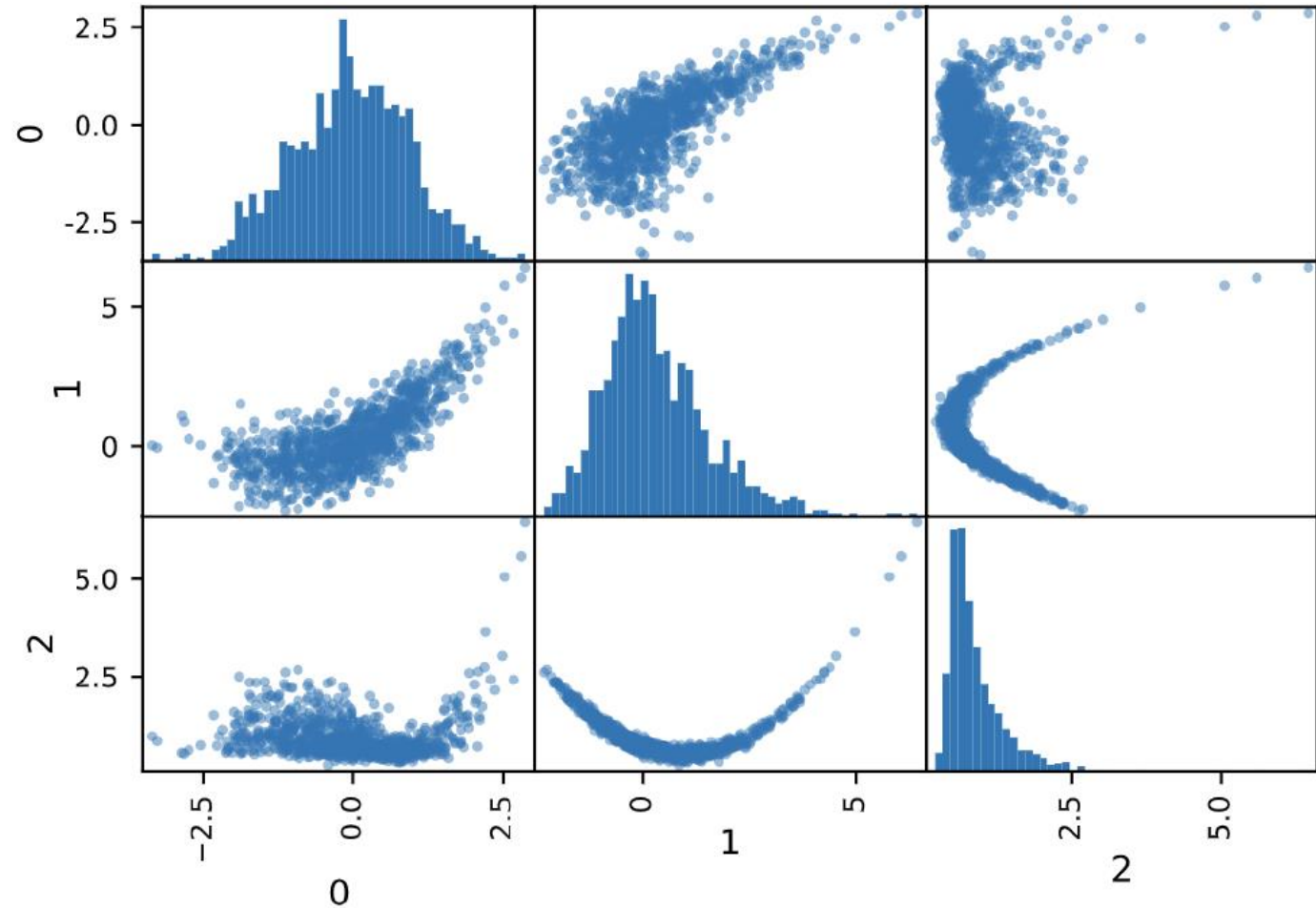
# 3D+: 3D Scatter Plot

	Dim 1	Dim 2	Dim 3
Nominal	X	X	X
Ordinal	X	X	X
Interval	✓	✓	✓
Ratio	✓	✓	✓



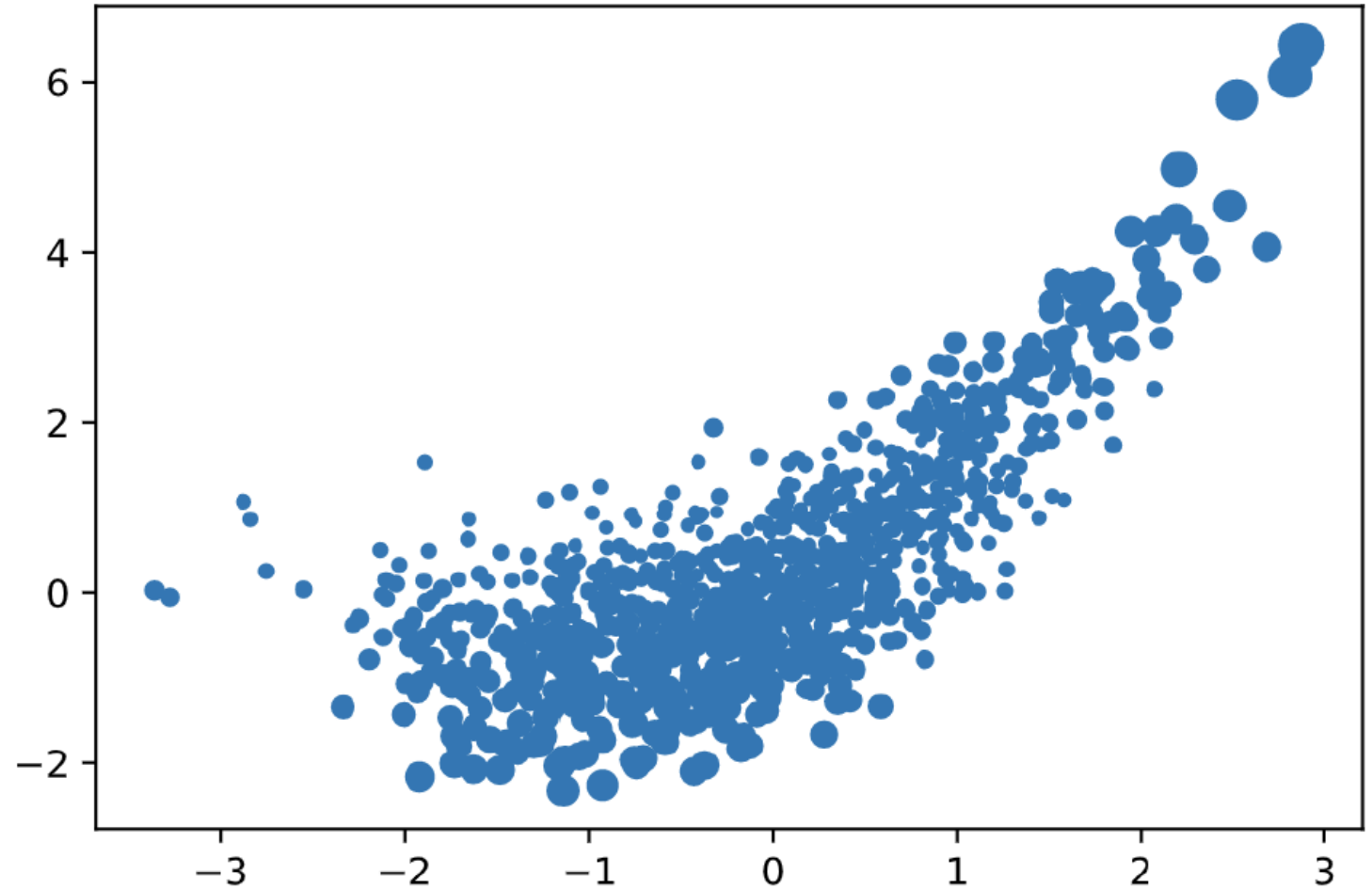
# 3D+: Scatter Plot Matrix

	Dim 1	Dim 2	Dim 3
Nominal	X	X	X
Ordinal	X	X	X
Interval	✓	✓	✓
Ratio	✓	✓	✓



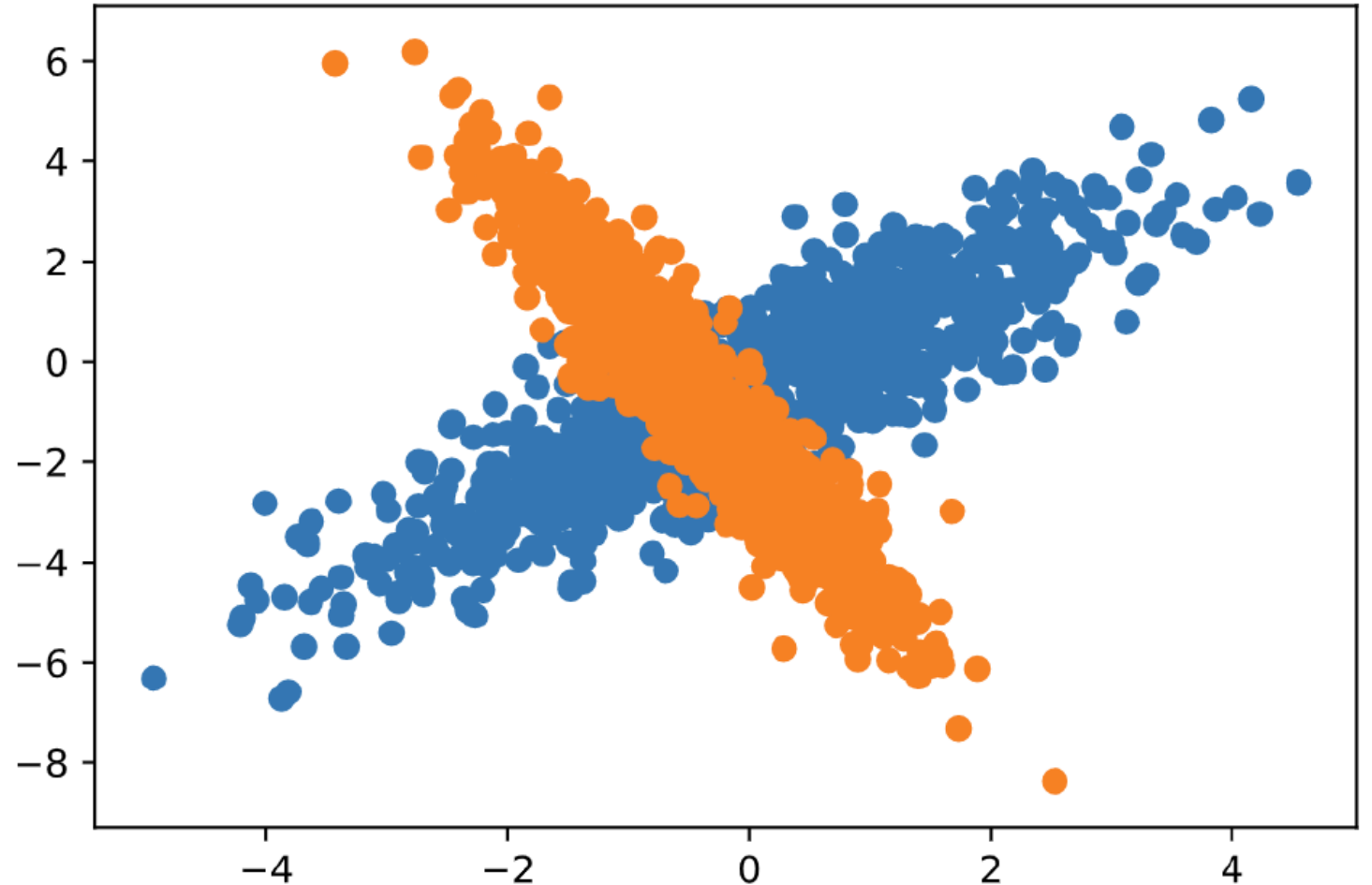
# 3D+: Bubble Plot

	Dim 1	Dim 2	Dim 3
Nominal	X	X	X
Ordinal	X	X	X
Interval	✓	✓	✓
Ratio	✓	✓	✓



# 3D+: Color Scatter Plot

	Dim 1	Dim 2	Dim 3
Nominal	X	X	✓
Ordinal	X	X	✓
Interval	✓	✓	X
Ratio	✓	✓	X



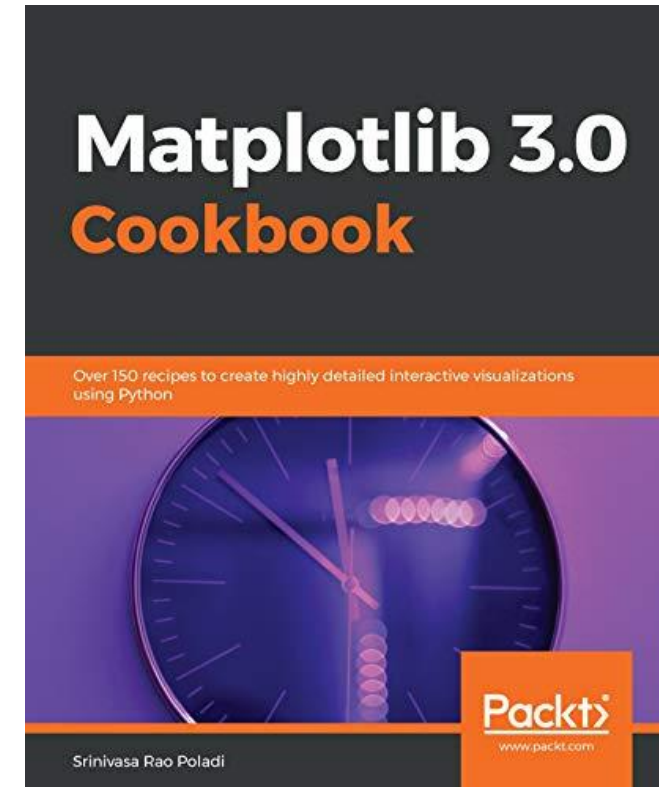
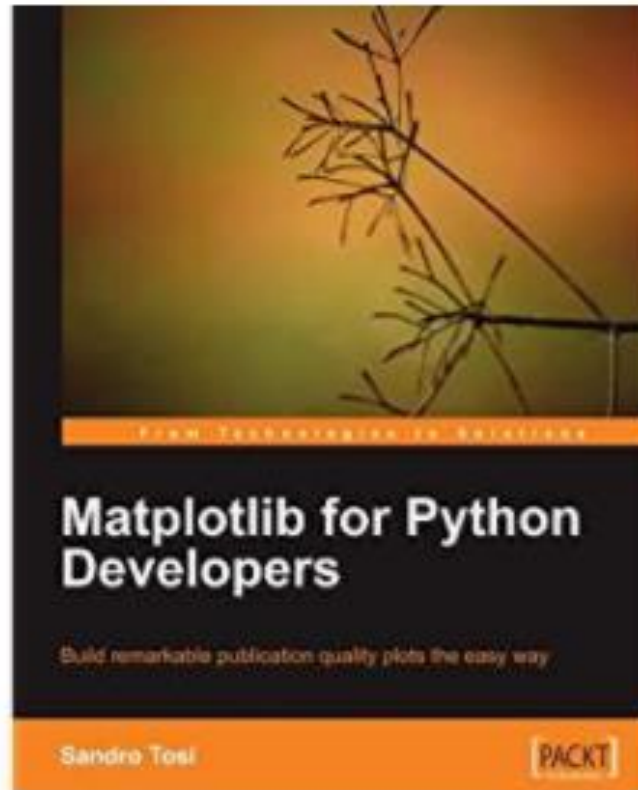
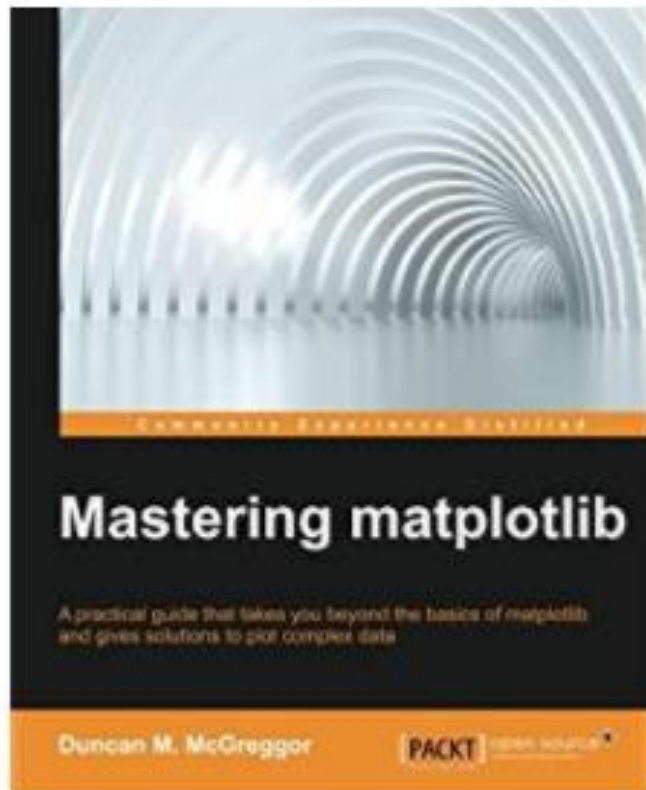


# Matplotlib

- Matplotlib is the standard for plotting in Python / Jupyter Notebook
- Matplotlib used to generate fairly ugly plots by default, but in recent versions this is no longer the case, so minimal need for additional libraries
- It is aimed at generating static plots, not very good for interacting with data (with a few exceptions)
- A number of additional libraries provide some level of interactive plot (and static plots), but matplotlib is enough of a standard

What is Matplotlib?

# Many Matplotlib Books



# What is "Matplotlib" Module?

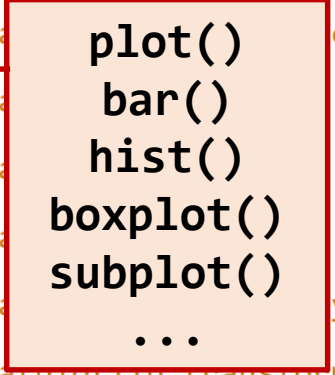
- Plotting library for Python
  - Reliant on NumPy
- **MATLAB-style plotting interface** with hierarchical organized elements
  - pyplot module: provides state-machine environment at the top of hierarchy
- Open source (<http://matplotlib.org>)
  - Original author: John D. Hunter (2003) → Michael Droettboom (2012)
- pyplot tutorials
  - [https://matplotlib.org/users/pyplot\\_tutorial.html](https://matplotlib.org/users/pyplot_tutorial.html)

# Overview

- matplotlib submodules
  - pyplot: provides MATLAB-like plotting framework
    - Closely related with actual plotting functionality
  - image: image loading, rescaling and display operations
  - colors: converting numbers or color arguments to RGB or RGBA
  - cm(colormap): provides function for color mapping functionality
  - collections: for efficient drawing of large collections
- We'll focus on pyplot module because it is in charge of creating plotting
- Other submodules support better plotting environment

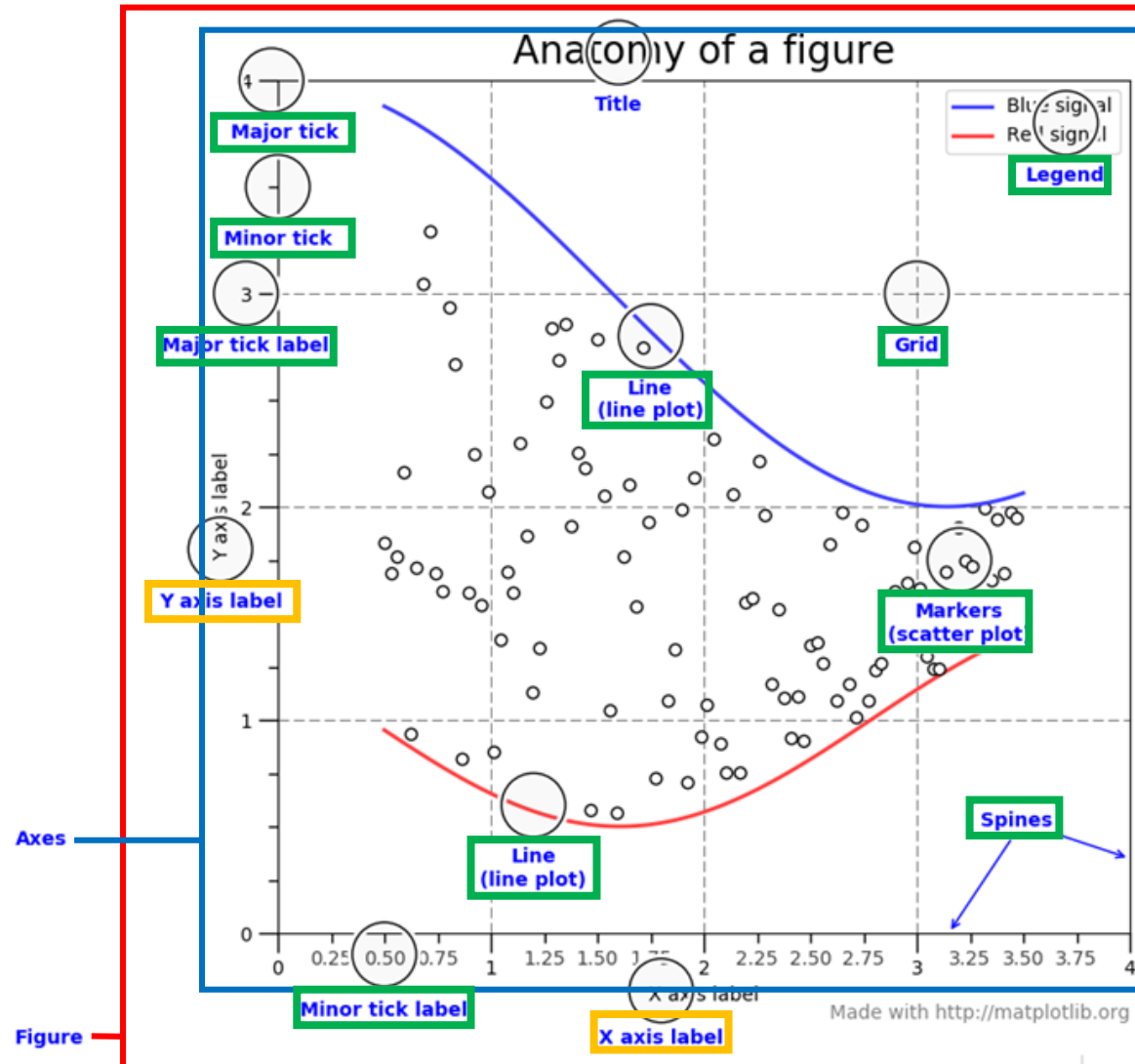
# Matplotlib Submodule List

matplotlib.afm	matplotlib.blocking_input	matplotlib.legend_handler	matplotlib.style
matplotlib.animation	matplotlib.category	matplotlib.lines	matplotlib.table
matplotlib.artist	matplotlib.cbook	matplotlib.markers	matplotlib.testing
matplotlib.axes	matplotlib.cm	matplotlib.mathtext	matplotlib.testing.compare
matplotlib.axis	matplotlib.collections	matplotlib.mlab	matplotlib.testing.decorators
matplotlib.backend_bases	matplotlib.colorbar	matplotlib.offsetbox	matplotlib.testing.disable_internet
matplotlib.backend_managers	matplotlib.colors	matplotlib.patches	matplotlib.exceptions
matplotlib.backend_tools	matplotlib.container	matplotlib.path	matplotlib.figure
matplotlib.backends.backend_agg	matplotlib.contour	matplotlib.patheffects	matplotlib.figure_factory
matplotlib.backends.backend_cairo	matplotlib.dates	matplotlib.projections	matplotlib.figure.mpl_toolkits
matplotlib.backends.backend_mixed	matplotlib.dviread	matplotlib.projections.polar	matplotlib.grab
matplotlib.backends.backend_nbagg	matplotlib.figure	<b>matplotlib.pyplot</b>	matplotlib.image
matplotlib.backends.backend_pdf	matplotlib.font_manager	matplotlib.rcsetup	matplotlib.transforms
matplotlib.backends.backend_pgf	matplotlib.fontconfig_pattern	matplotlib.sankey	matplotlib.tri
matplotlib.backends.backend_ps	matplotlib.gridspec	matplotlib.scale	matplotlib.type1font
matplotlib.backends.backend_svg	matplotlib.image	matplotlib.sphinxext.plot_directive	matplotlib.units
matplotlib.backends.backend_tkagg	matplotlib.legend	matplotlib.spines	matplotlib.widgets



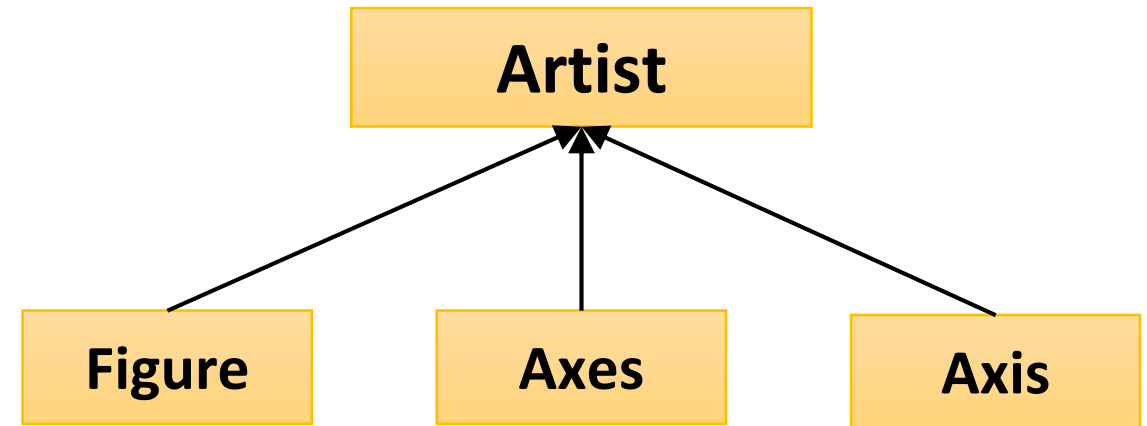
**plot()**  
**bar()**  
**hist()**  
**boxplot()**  
**subplot()**  
**...**

# Pyplot Plotting Example



# Matplotlib Classes

- 4 important classes in matplotlib
- **Artist**
  - Basically everything you can see on the figure
- **Figure**
  - Consider as a window
  - A figure may contain several child Axes and keep track of them
  - Several figures can be created
- **Axes**: region of an image or a plot
- **Axis**: strings labeling ticks





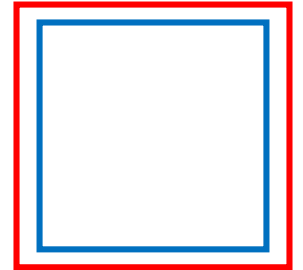
# Using Pyplot

## ■ Import library

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

For jupyter notebook:  
plot graphs without show()

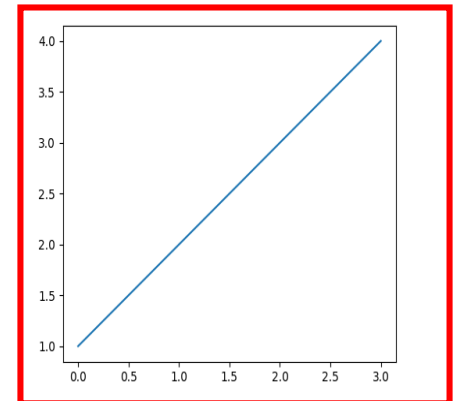
- Pyplot uses **Figure** with **Axes** by default and use it as current Axes
- Each objects can be added as needed



## ■ Plot on current Axes

```
plt.plot([1, 2, 3, 4])
```

- Automatically set the x, y ranges and tick labels



## ■ Display figure

```
plt.show()
```

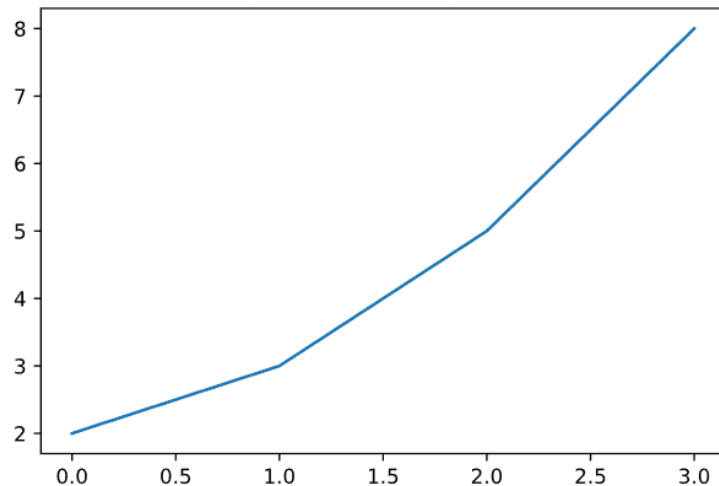
- For actual display of figures in pyplot, additional statement is necessary

# Pyplot Basics

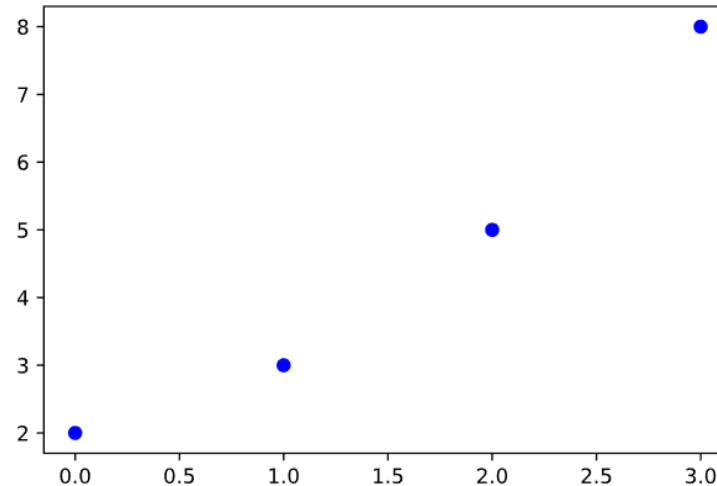
# Line and Scatter Plot

- `plt.plot([x], y, [fmt], [x2], y2, [fmt2], ...)`
  - The coordinates of the points or line nodes given by `x, y`
  - If `x` is omitted, index array `[0, 1, ..., N-1]` is used as `x`
  - `fmt`: defines basic formatting like color, marker and linestyle (default: blue line graph)

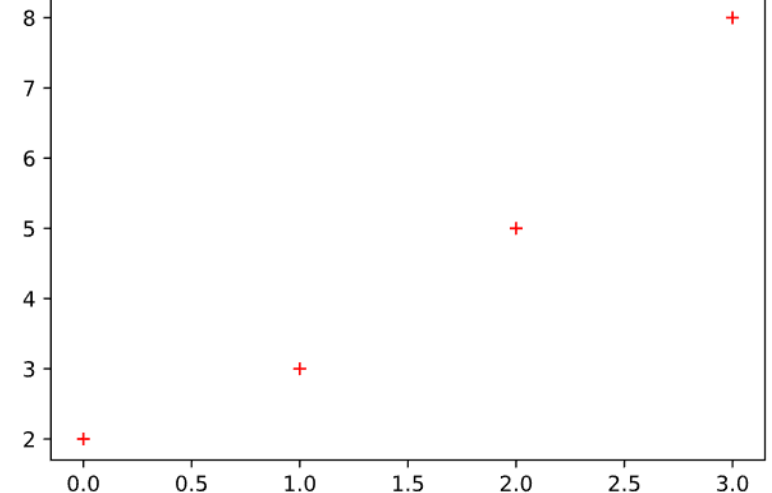
```
plt.plot([2,3,5,8])
```



```
plt.plot([2,3,5,8], 'bo')
```



```
plt.plot([2,3,5,8], 'r+')
```



# Markers

'.'	point marker	's'	square marker
','	pixel marker	'p'	pentagon marker
'o'	circle marker	'*'	star marker
'v'	triangle_down marker	'h'	hexagon1 marker
'^'	triangle_up marker	'H'	hexagon2 marker
'<'	triangle_left marker	'+'	plus marker
'>'	triangle_right marker	'x'	x marker
'1'	tri_down marker	'D'	diamond marker
'2'	tri_up marker	'd'	thin_diamond marker
'3'	tri_left marker	' '	vline marker
'4'	tri_right marker	'_'	hline marker

# Line Styles and Colors

' - '	solid line style
' - - '	dashed line style
' - . '	dash-dot line style
' : '	dotted line style

' b '	blue
' g '	green
' r '	red
' c '	cyan
' m '	magenta
' y '	yellow
' k '	black
' w '	white

**fmt = ' [marker][line][color] '**

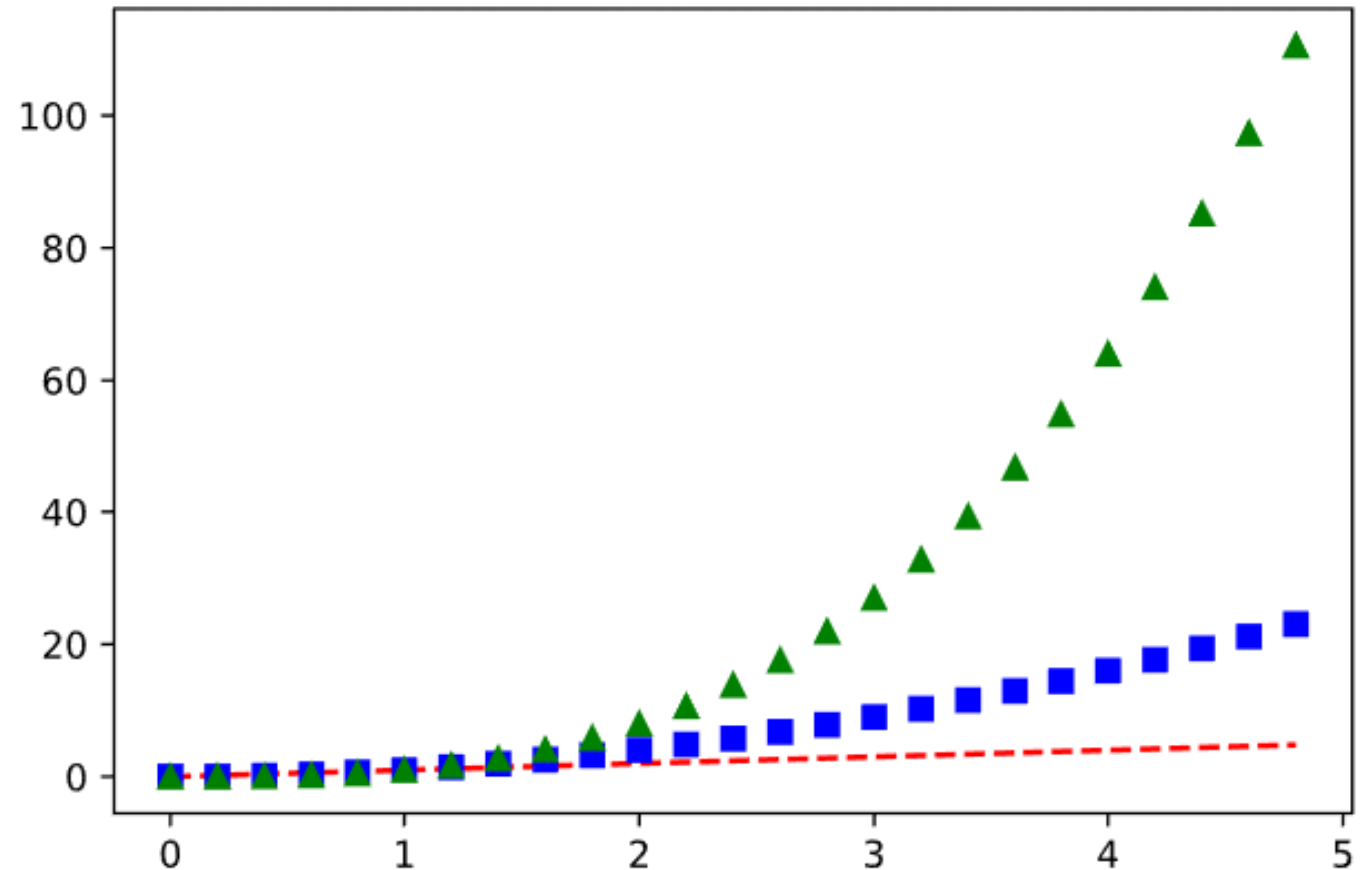
**(Same color for line and marker)**

# Plotting Multiple Graphs

- Plotting 3 different data at once

```
t = np.arange(0.0, 5.0, 0.2)
plt.plot(t, t, 'r--',
         t, t**2, 'bs',
         t, t**3, 'g^')
```

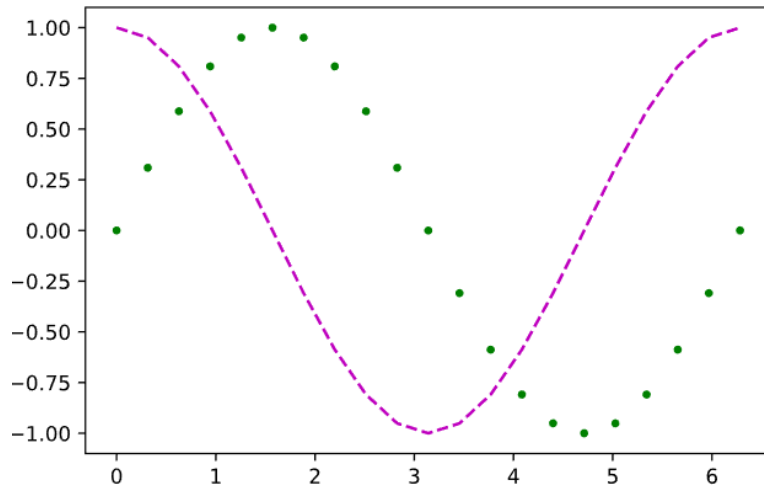
- 'r--': red dashed
- 'bs': blue square
- 'g^': green triangle



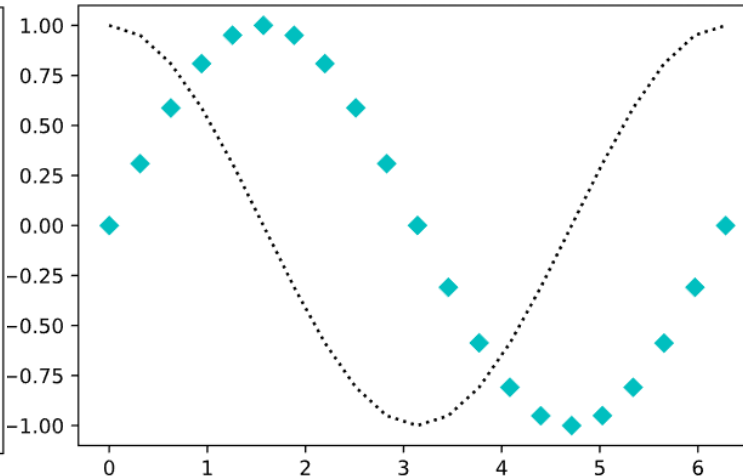
# More Examples

```
x = np.linspace(0, 2*np.pi, 21)
y = np.sin(x)
y2 = np.cos(x)
```

```
plt.plot
(x, y, 'g.', x, y2, 'm--')
```

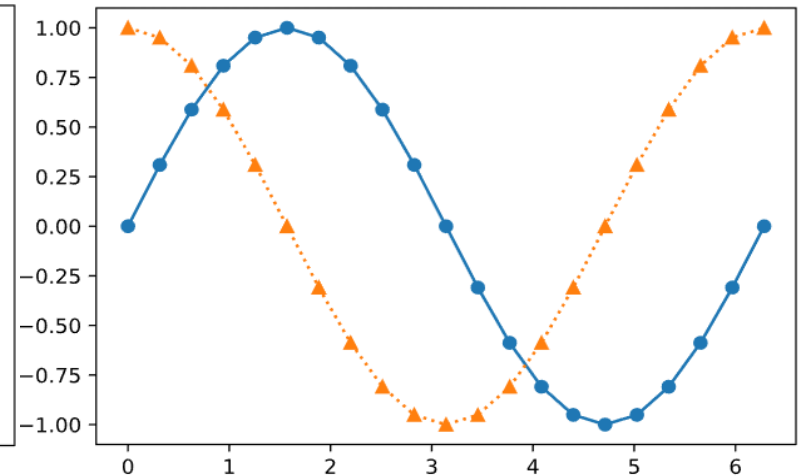


```
plt.plot
(x, y, 'cD', x, y2, 'k-.')
```



*Choose colors automatically*

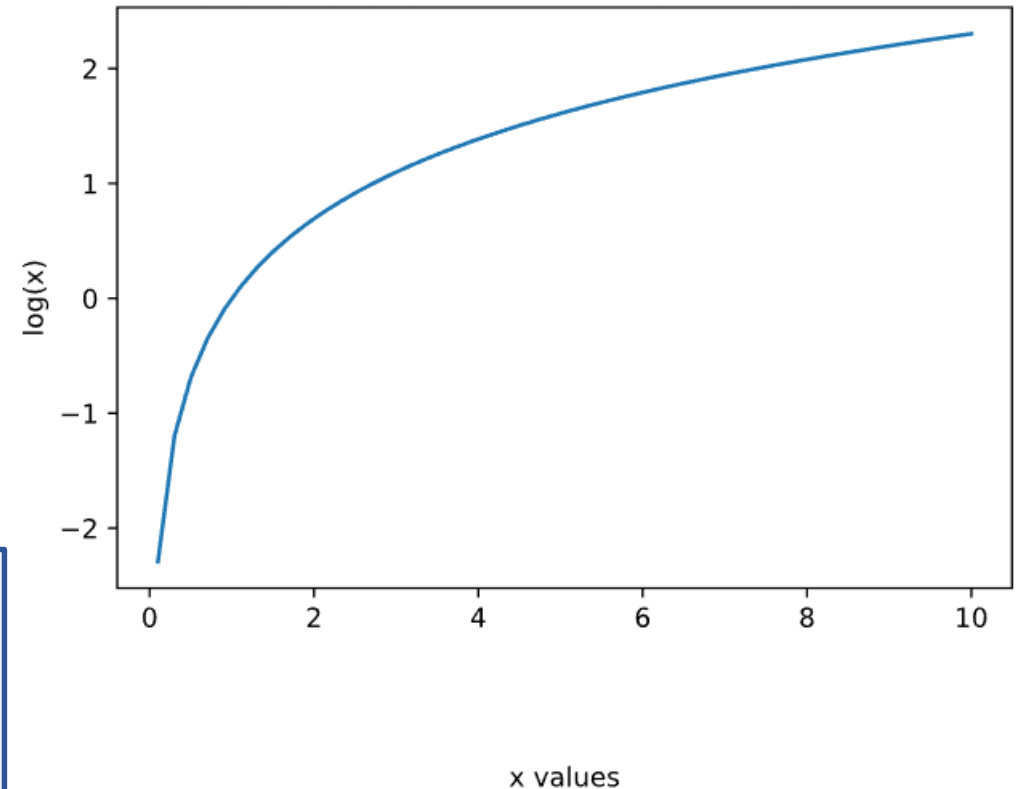
```
plt.plot
(x, y, '-o', x, y2, '^:')
```



# Axis Labels

- `plt.xlabel(xlabel, [labelpad], ...)`
- `plt.ylabel(ylabel, [labelpad], ...)`
  - Set the label for the x-axis (or y-axis)
  - `xlabel, ylabel`: string for the label
  - `labelpad`: spacing in points from the axes bounding box including ticks and tick labels

```
x = np.linspace(0.1, 10, 100)
y = np.log(x)
plt.xlabel('x values', labelpad=50)
plt.ylabel('log(x)')
plt.plot(x, y)
```

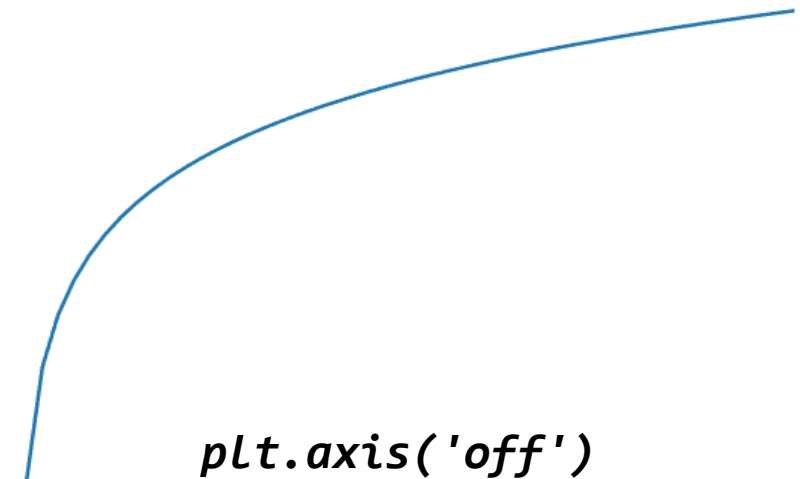
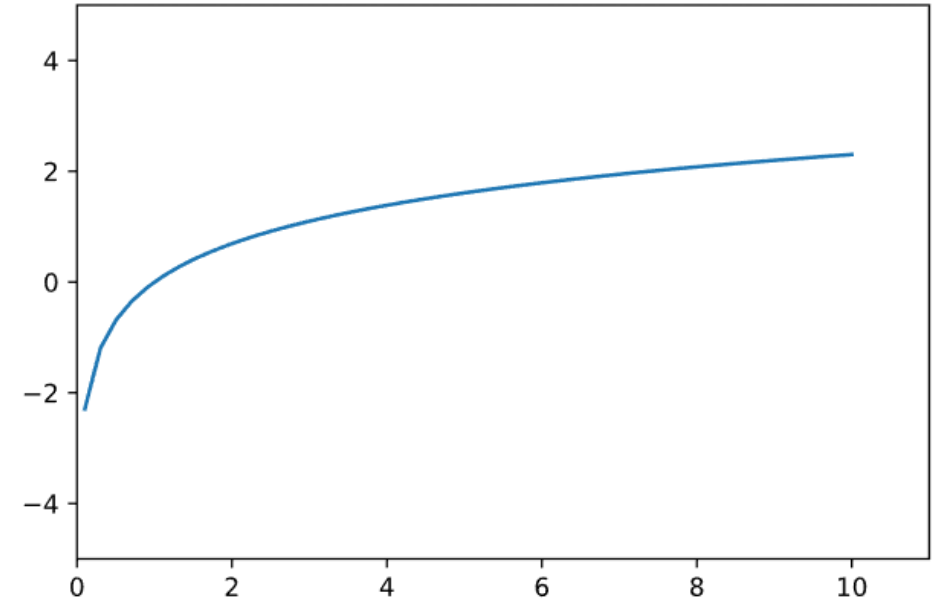




# Axis

- `plt.axis(limits, ..)`
  - Set (or get) some axis properties
  - `limits`: a list with the axis limits to be set.  
[xmin, xmax, ymin, ymax]
  - Use 'off' to hide all the axes

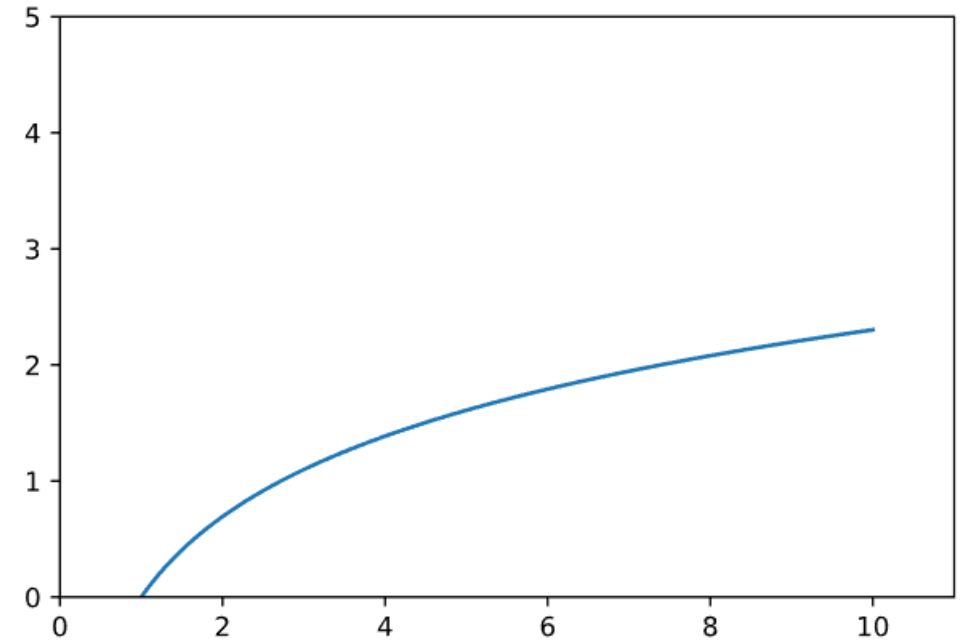
```
x = np.linspace(0.1, 10, 100)
y = np.log(x)
plt.axis([0, 11, -5, 5])
# plt.axis('off')
plt.plot(x, y)
```



# Axis Limit

- `plt.xlim(left, right)`
- `plt.ylim(bottom, top)`
  - Set (or get) the x-limits (or y-limits) of the current axes
  - *left, right*: x-limits
  - *bottom, top*: y-limits

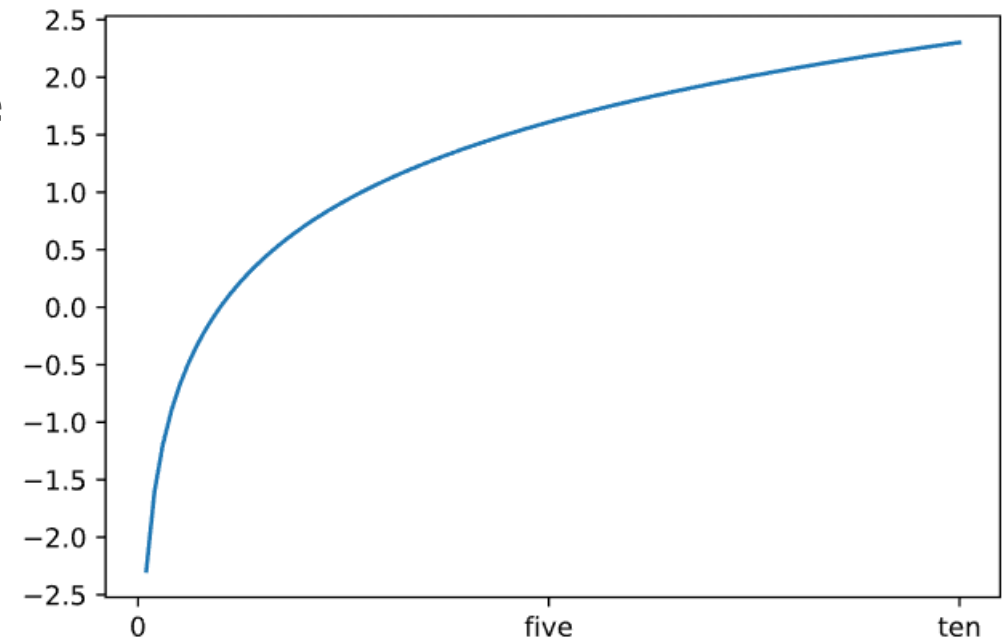
```
x = np.linspace(0.1, 10, 100)
plt.xlim(0, 11)
plt.ylim(top=5)
plt.plot(x, np.log(x))
```



# Ticks

- `plt.xticks(ticks, [labels], ...)`
- `plt.yticks(ticks, [labels], ...)`
  - Set the current tick locations and labels of the x-axis (or y-axis)
  - *ticks*: A list of positions ticks should be placed
  - *labels*: A list of explicit labels to place at the given locations

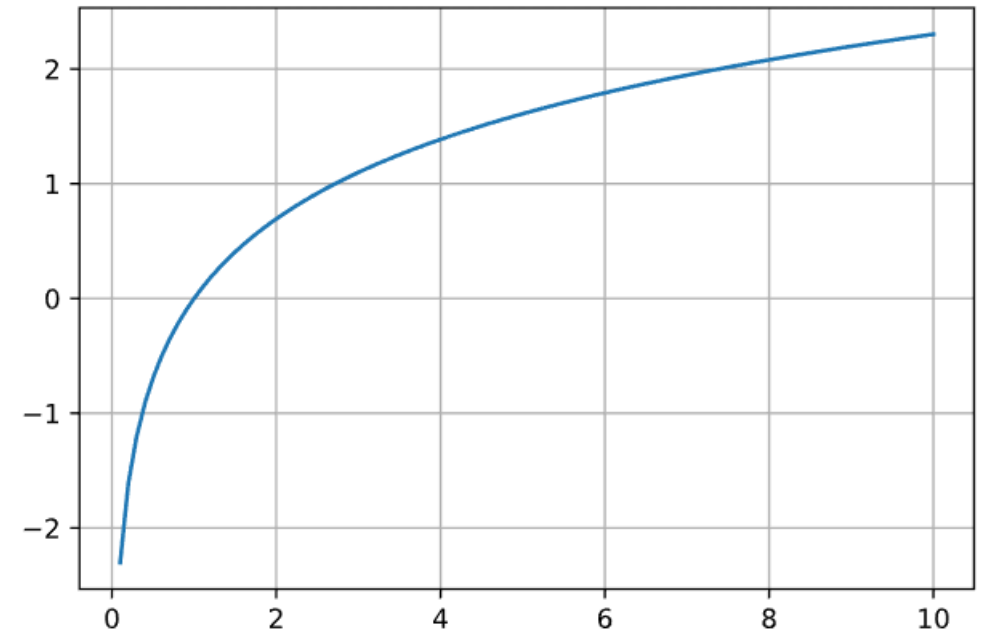
```
x = np.linspace(0.1, 10, 100)
plt.xticks(np.arange(0, 11, 5),
           ('0', 'five', 'ten'))
plt.yticks(np.arange(-3, 3, 0.5))
plt.plot(x, np.log(x))
```



# Grids

- `plt.grid([b], [which], [axis], ...)`
  - Configure the grid lines
  - `b`: if True, show the grid lines (toggle if no argument given)
  - `which`: the grid lines to apply the changes on ('major', 'minor', or 'both')
  - `axis`: the axis to apply ('x', 'y', or 'both')

```
x = np.linspace(0.1, 10, 100)
plt.grid()
plt.plot(x, np.log(x))
```

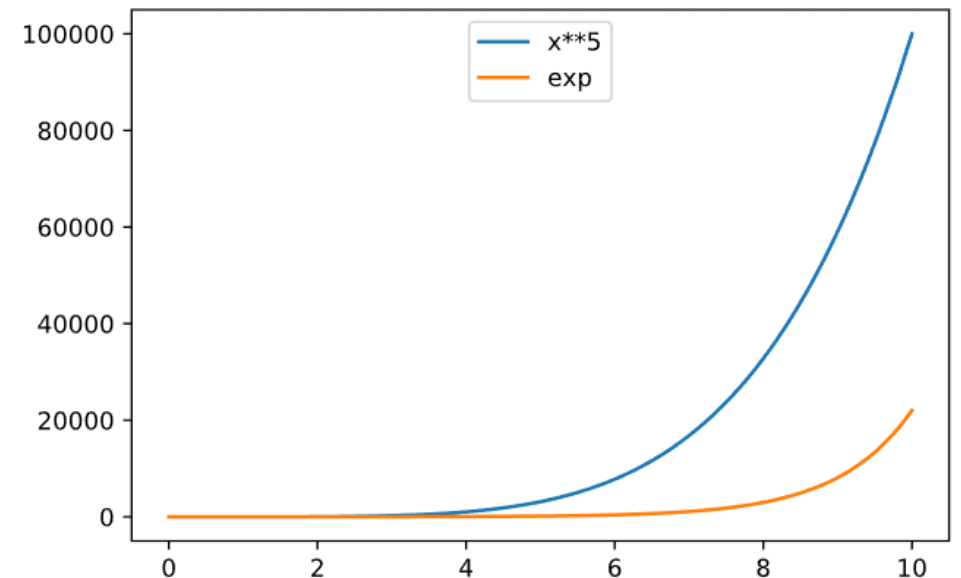


# Legend

- `plt.legend([loc], [ncol], ...)`
  - Place a legend on the axes
  - If no argument is given, automatically determine and show the legend
  - `loc`: the location of the legend (default: 'best')
  - `ncol`: the number of columns (default: 1)

```
x = np.linspace(0, 10, 100)
plt.plot(x, x**5, label='x**5')
plt.plot(x, np.exp(x), label='exp')
plt.legend(loc='upper center')
```

Location String	Location Code
'best'	0
'upper right'	1
'upper left'	2
'lower left'	3
'lower right'	4
'right'	5
'center left'	6
'center right'	7
'lower center'	8
'upper center'	9
'center'	10

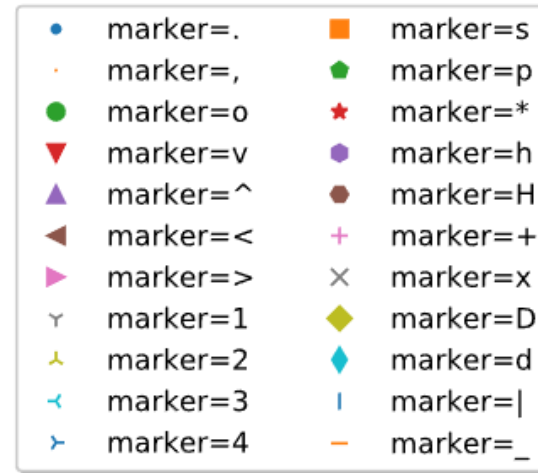
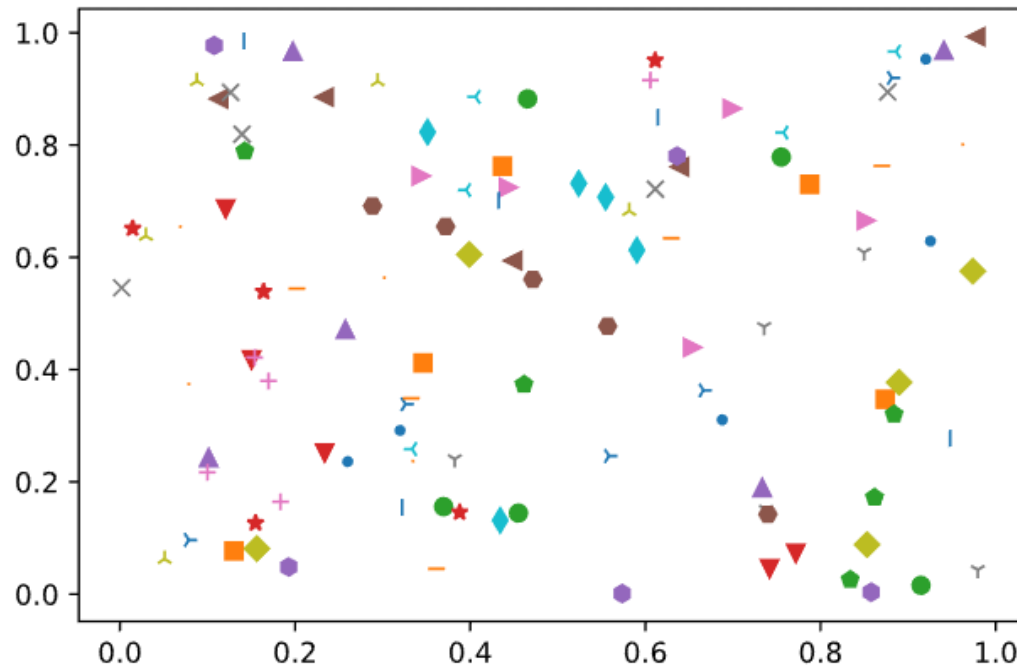


# Legend: More Options

- `plt.legend(...)`
  - `bbox_to_anchor`: box that is used to position the legend in the arbitrary location with the form (x, y) or (x, y, width, height)
  - `frameon`: control whether the legend should be drawn on a frame (default: True)
  - `fancybox`: control whether round edges should be enabled (default: True)
  - `shadow`: control whether to draw a shadow behind the legend (default: False)
  - `fontsize`: control the font size of the legend
  - `title`: the legend's title
  - `markerscale`: the relative size of legend markers compared
  - `markerfirst`: if True, legend marker is placed to the left (default: True)
  - ...

# Legend: Example

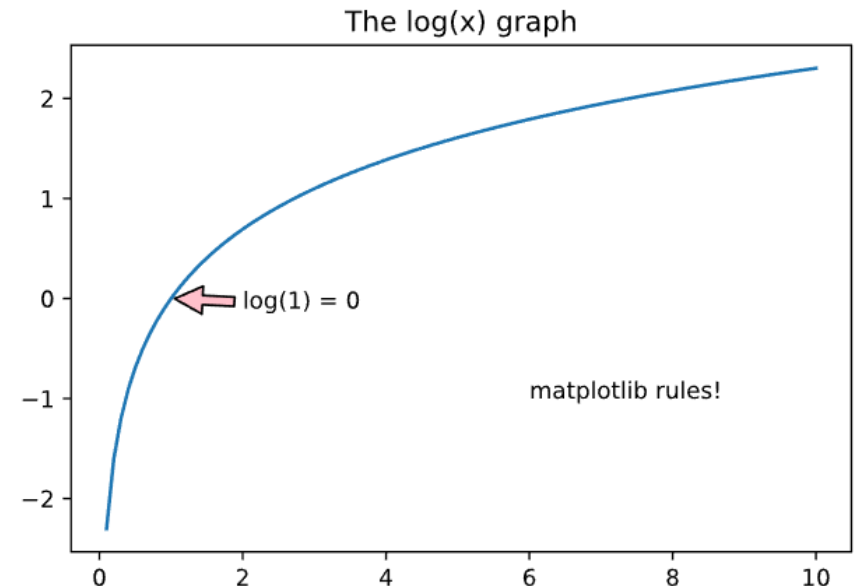
```
markers = '.,ov^<>1234sp*hH+xDd|_'  
for m in markers:  
    plt.plot(np.random.rand(5), np.random.rand(5), m, label='marker=%c' % m)  
plt.legend(ncol=2, bbox_to_anchor=(1.05, 0.9), shadow=True)
```



# Texts

- `plt.title(label, [loc], [pad], ...)`
  - label: text to use for the title
  - loc: which title to set ('center', 'left', or 'right')
  - pad: the offset from the top of the axes
- `plt.text(x, y, s, [loc], [pad], ...)`
  - x, y: the position to place the text
  - s: the text to display
- `plt.annotate(s, [xy], [xytext], ...)`
  - s: the text to display
  - xy: the point to annotate
  - xytext: the position to place the text at

```
x = np.linspace(0.1, 10, 100)
plt.plot(x, np.log(x))
plt.title("The log(x) graph")
plt.text(6, -1, 'matplotlib rules!')
plt.annotate('log(1) = 0', xy=(1,0),
             xytext=(2,-0.1),
             arrowprops=dict(facecolor='pink',
                             shrink=0.05))
```

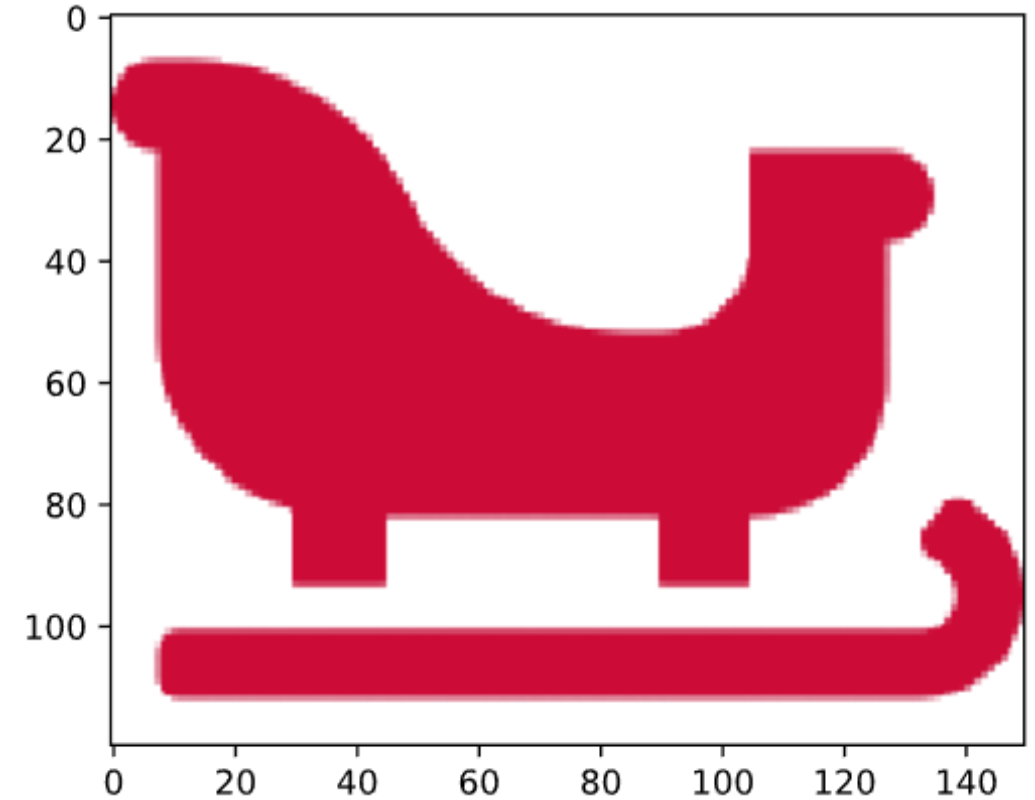




# Image Submodule

```
%matplotlib inline
import matplotlib.pyplot as plt
import matplotlib.image as mpimg

image = mpimg.imread('sleds.png')
plt.imshow(image)
```



# Saving Plots

- `plt.savefig(fname, [format], [dpi], [transparent], ...)`
  - Save the current figure
  - `fname`: file name to save. If format is not given, the output format is inferred from the extension of the fname
  - `format`: the file format (e.g., 'png', 'pdf', 'svg', 'jpg', ...)
  - `dpi`: the resolution in dots per inch
  - `transparent`: if True, make the plot transparent (default: False)

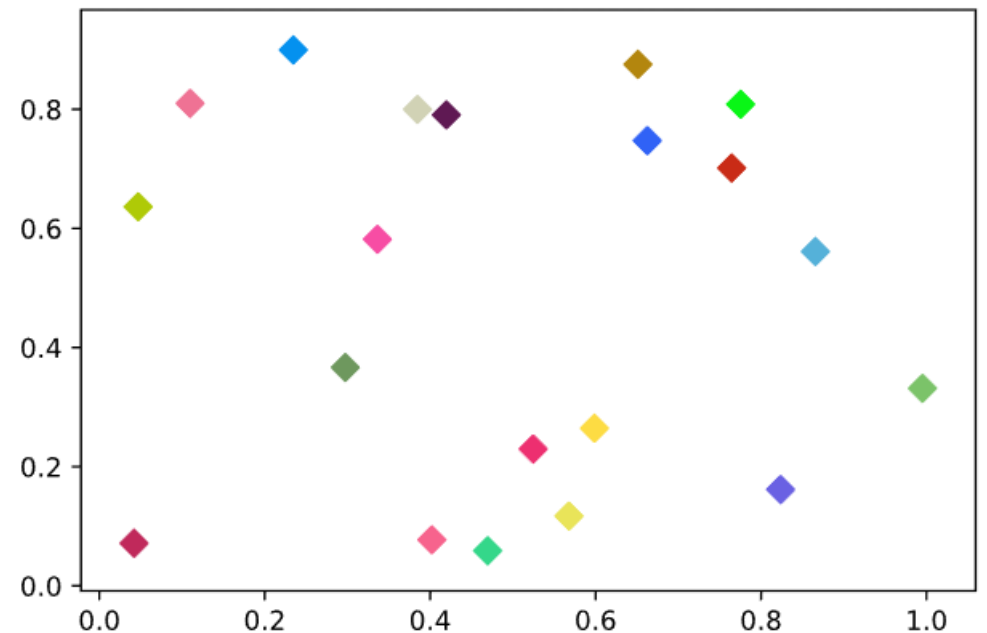
```
x = np.linspace(0.1, 10, 100)
plt.plot(x, np.log(x))
plt.savefig('log.png', dpi=300, transparent=True)
```

# Scatter Plot

# scatter()

- `plt.scatter(x, y, [s], [c], [marker], [alpha], ...)`
  - A scatter plot of y vs. x with varying marker size and/or color
  - `x, y`: data positions
  - `s, c`: the marker size (in points\*\*2) and its color
  - `marker`: the marker style
  - `alpha`: the alpha blending value (0: transparent ~ 1: opaque)

```
x = np.random.rand(20)
y = np.random.rand(20)
colors=['#'+'%06x'%np.random.randint(256**3)
        for _ in range(20)]
plt.scatter(x, y, s=50, c=colors, marker='D')
```



# scatter() vs. plot()

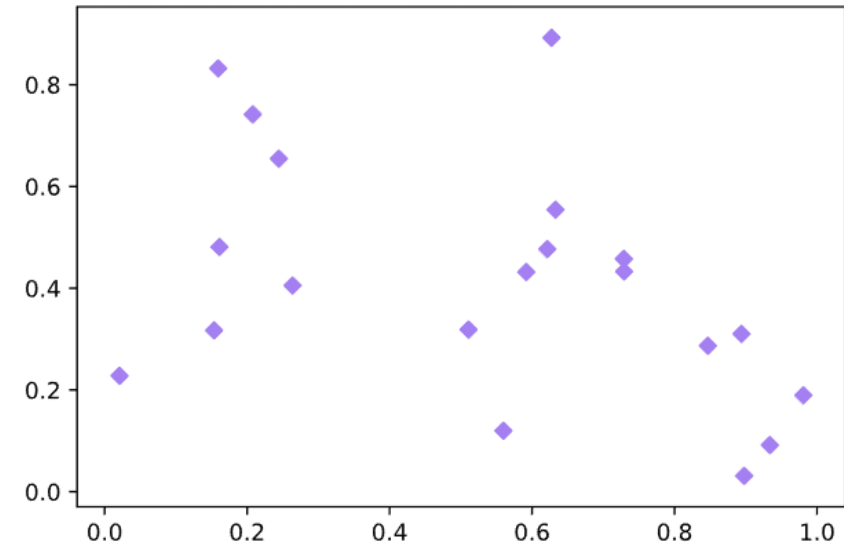
## ▪ pyplot.plot()

```
np.random.seed(20200101)
N=20
x = np.random.rand(N)
y = np.random.rand(N)
plt.plot(x,y,'D',c='#a37ff1',markersize=5)
```

## ▪ pyplot.scatter()

```
np.random.seed(20200101)
N=20
x = np.random.rand(N)
y = np.random.rand(N)
plt.scatter(x,y,s=25,c='#a37ff1',marker='D')
```

- Both can be used for simple cases  
→ same result
- scatter() is more comfortable to configure each data point
- plot() allows to plot the lines connecting data points

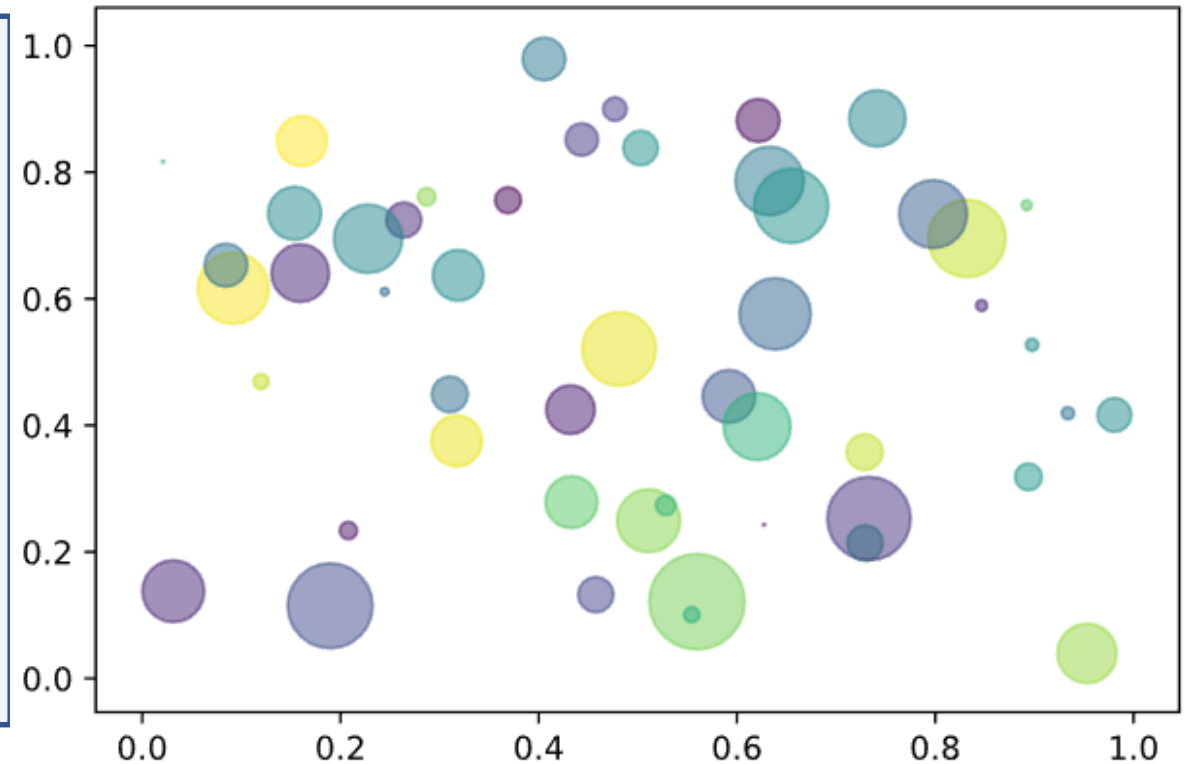


# Scatter Plot with Different Marker Sizes

```
np.random.seed(20200101)

N = 50
x = np.random.rand(N)
y = np.random.rand(N)
colors = np.random.rand(N)
area=(30 * np.random.rand(N))**2

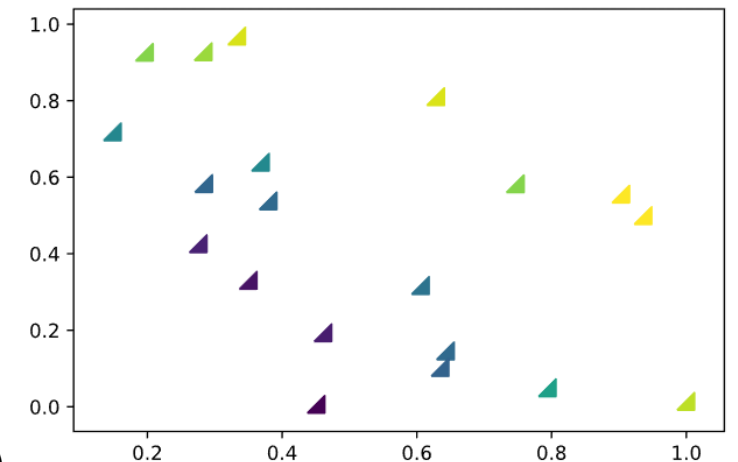
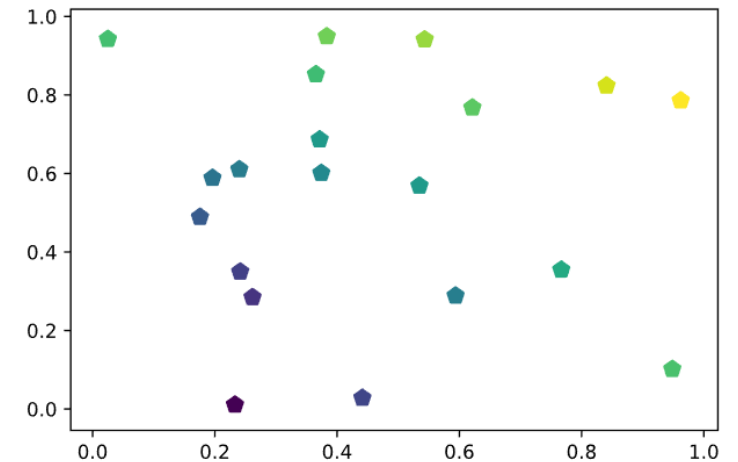
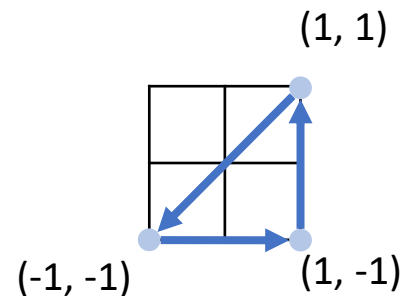
plt.scatter(x, y, s=area, c=colors, alpha=0.5)
```



# Scatter Plot with Polygon Symbols

- Marker = (numsides, style, angle)
  - *numsides*: number of sides
  - *style*: regular polygon(0), star-like symbol(1), asterisk(2), circle(3)
  - *angle*: angle of rotation
- User-defined marker:
  - A list of (x, y) describing a symbol with center = (0, 0)

```
x = np.random.rand(20)
y = np.random.rand(20)
z = np.sqrt(x**2 + y**2)
plt.scatter(x, y, s=80, c=z, marker=(5,0))
v = np.array([[-1,-1],[1,-1],[1,1],[-1,1]])
plt.scatter(x, y, s=80, c=z, marker=v)
```

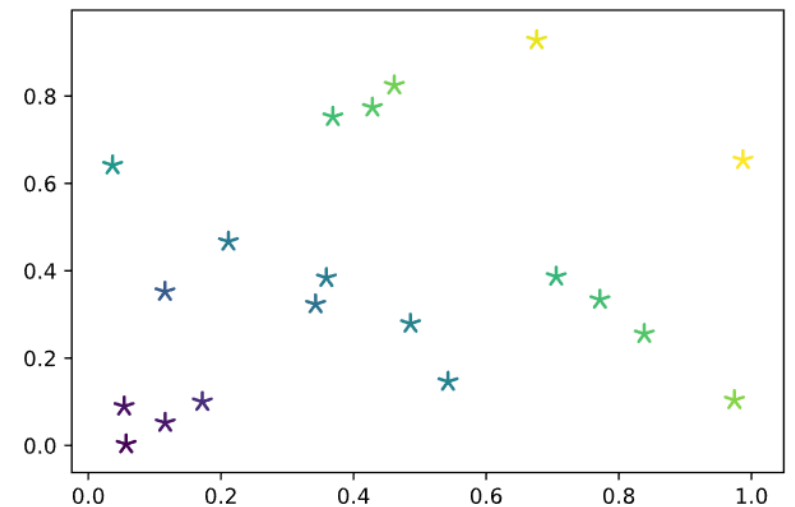
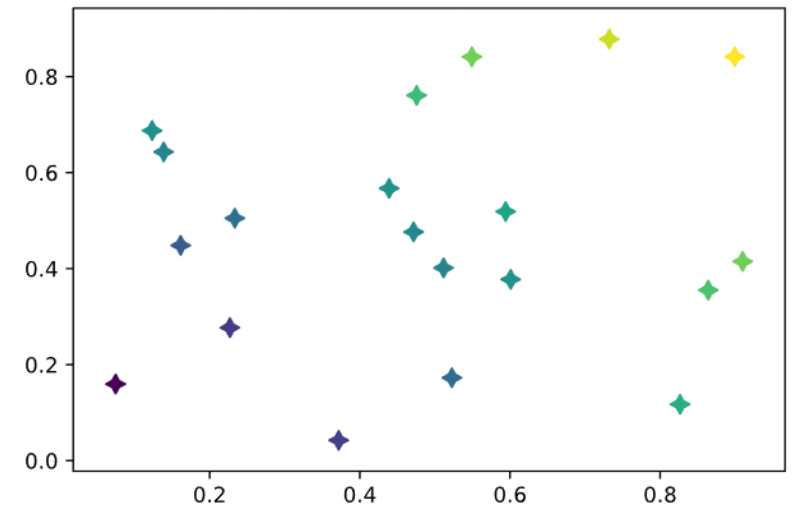


# Scatter Plot with Stars and Asterisks

```
x = np.random.rand(20)
y = np.random.rand(20)
z = np.sqrt(x**2 + y**2)

# with 4-side star
plt.scatter(x, y, s=80, c=z, marker=(4,1))

# with 5-side asterisk
plt.scatter(x, y, s=80, c=z, marker=(5,2))
```





# Default Color Map

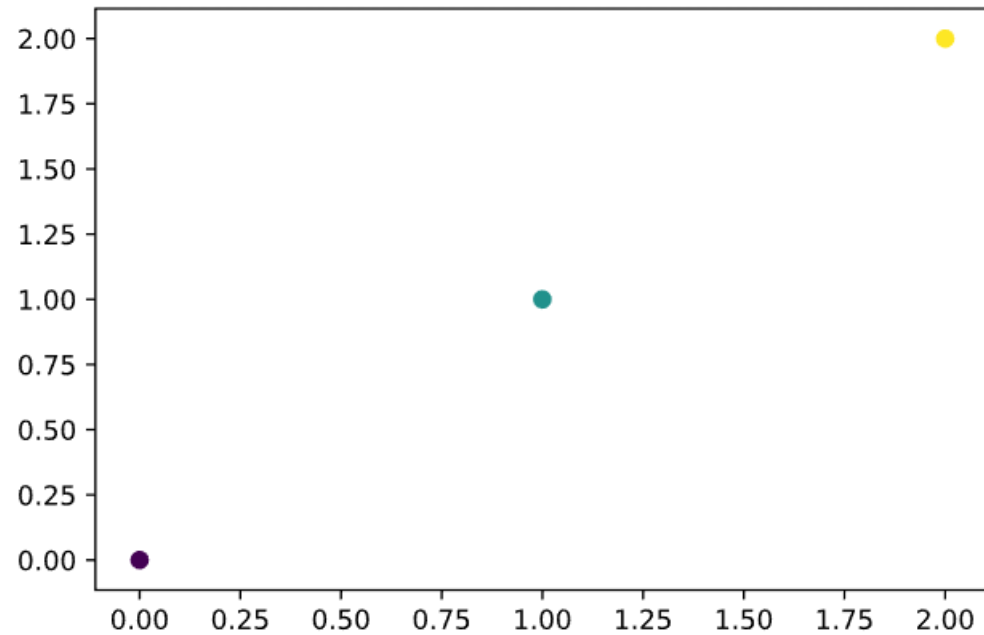
- 'viridis':

Violet

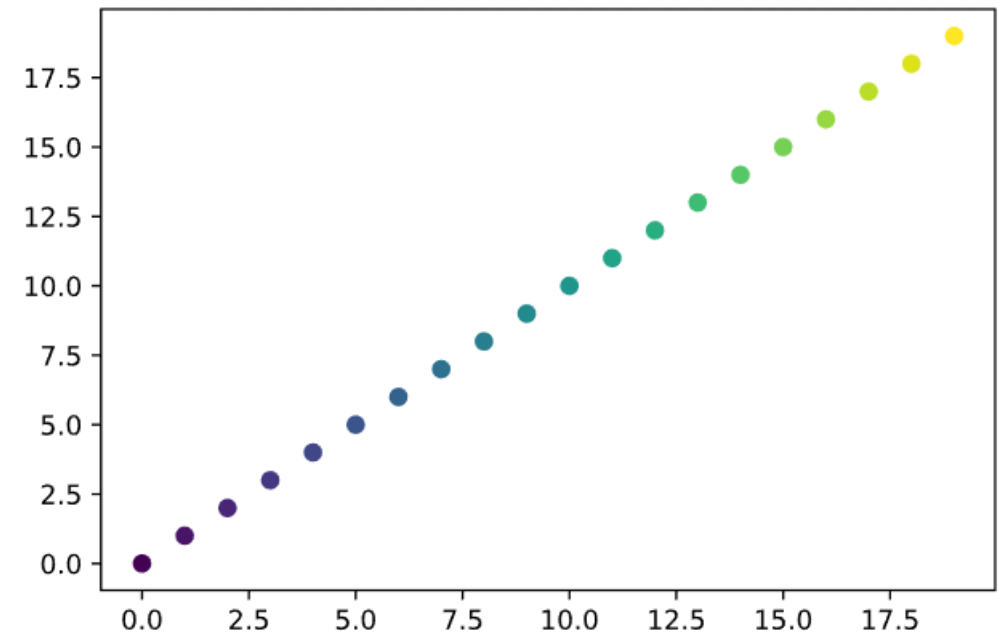
Green

Yellow

```
x = np.arange(3)
plt.scatter(x, x, c=x)
```

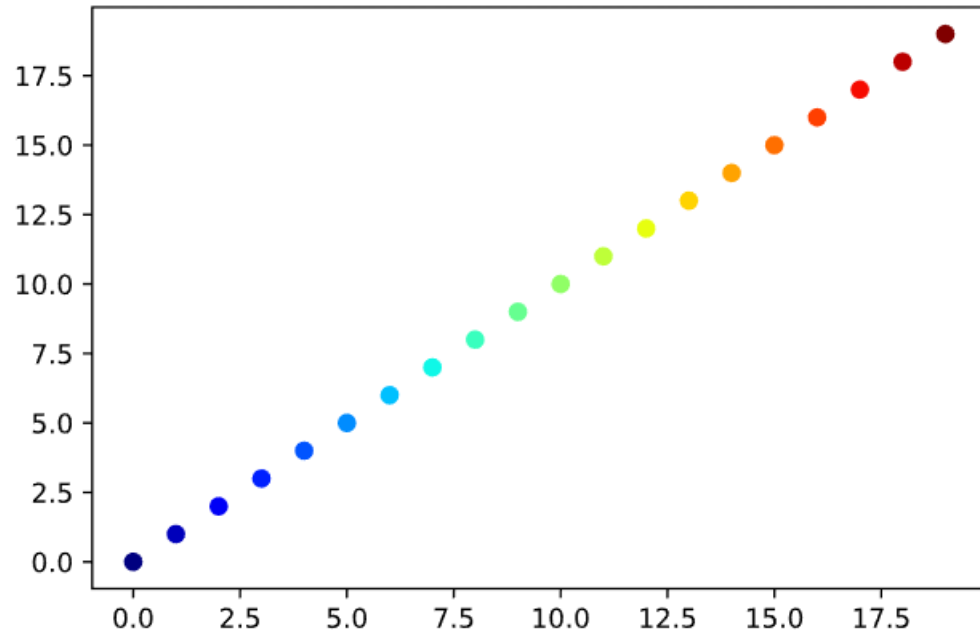


```
x = np.arange(20)
plt.scatter(x, x, c=x)
```



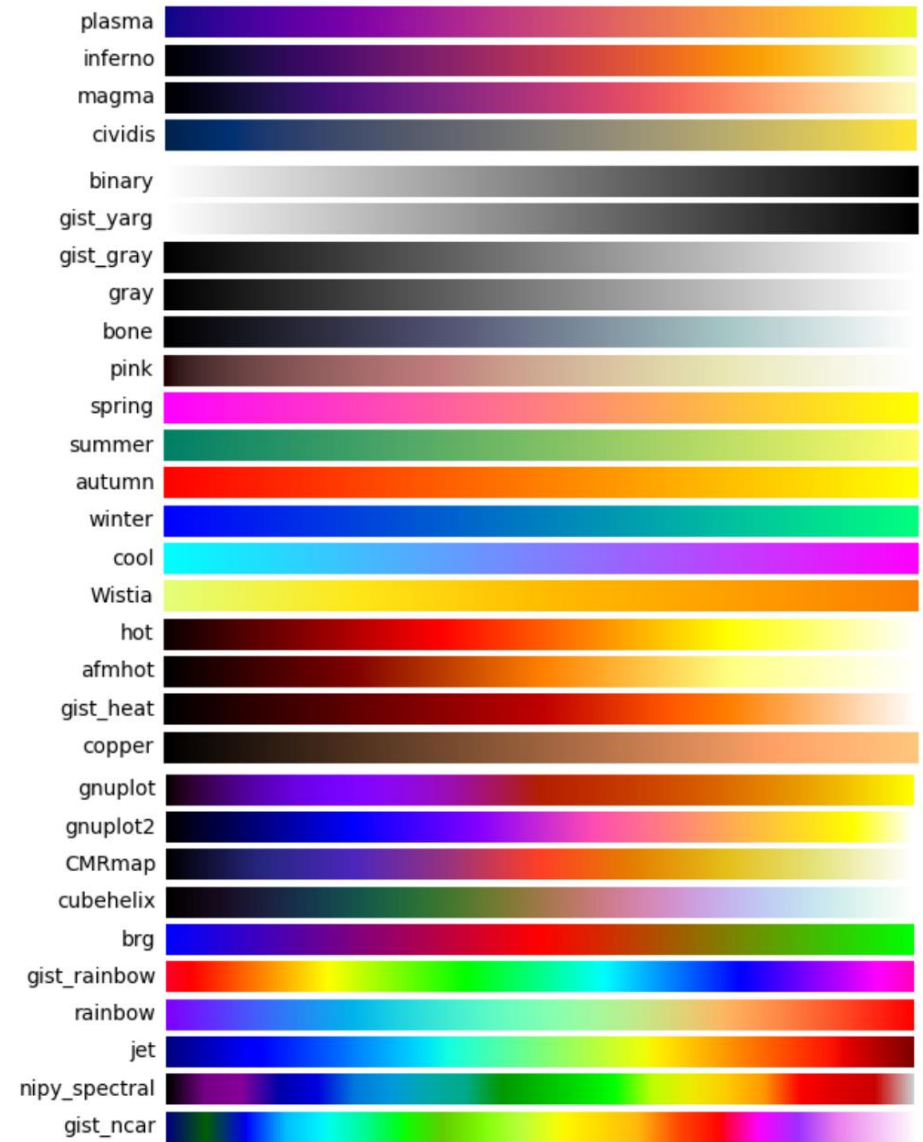
# Changing a Color Map

```
plt.rc('image', cmap='jet')  
x = np.arange(20)  
plt.scatter(x, x, c=x)
```



For other colormaps, visit

<https://matplotlib.org/tutorials/colors/colormaps.html>

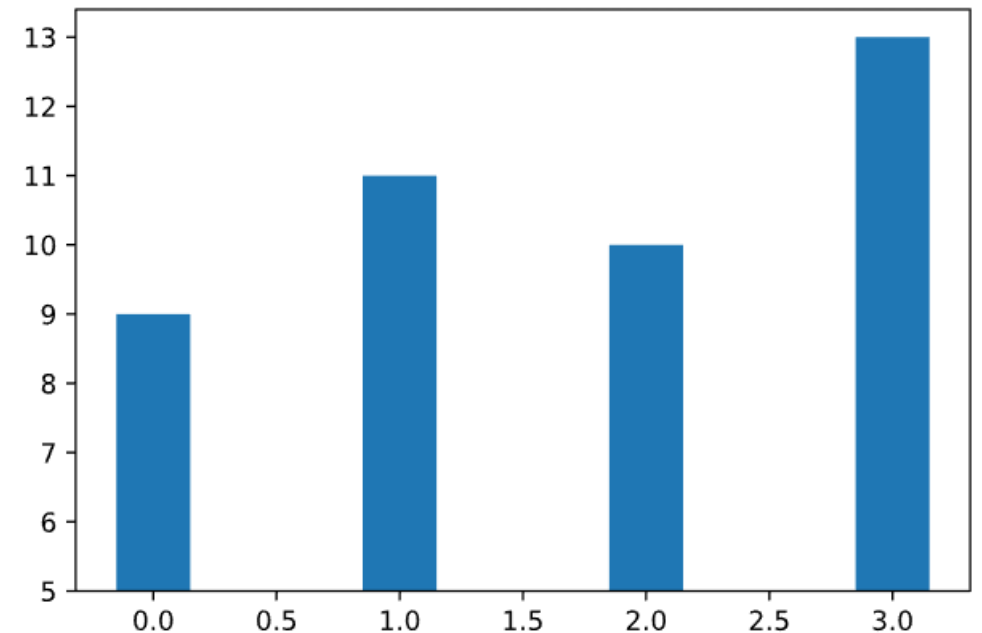


# Bar Chart

# bar()

- `plt.bar(x, height, [width], [bottom], [align], ...)`
  - Make a bar plot
  - `x, height`: data points
  - `width`: the width(s) of the bars (default: 0.8)
  - `bottom`: the y coordinate(s) of the bars bases
  - `align`: alignment of the bars to the x coordiantes -- 'center' (default) or 'edge'

```
x = np.arange(4)
y = [4, 6, 5, 8]
plt.bar(x, y, width=0.3, bottom=5)
```



# Side-by-Side Bar Chart

```
N = 5
```

```
m = (20, 35, 30, 35, 27)
```

```
w = (25, 32, 34, 30, 27)
```

```
x = np.arange(N)
```

```
width=0.35
```

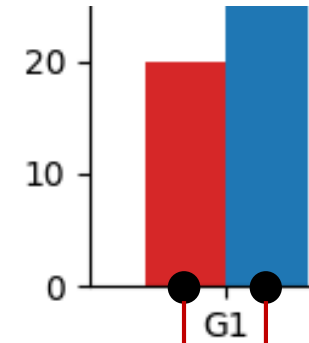
```
plt.bar(x-width/2, m, width, color='#d62728')  
plt.bar(x+width/2, w, width)
```

```
plt.xticks(x, ('G1','G2','G3','G4','G5'))
```

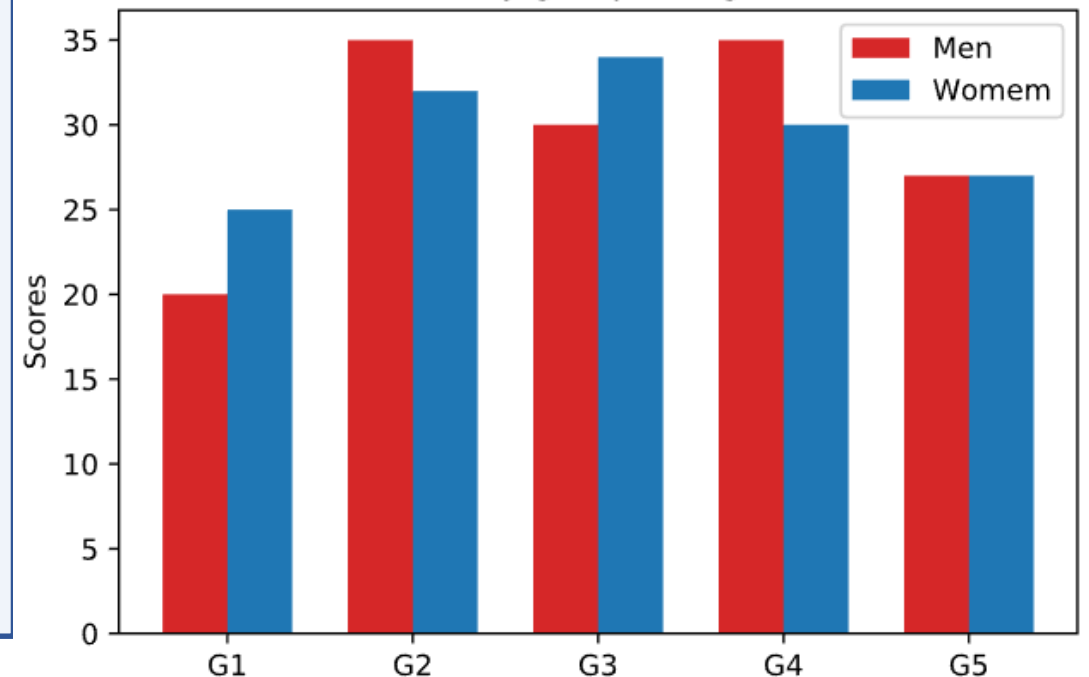
```
plt.ylabel('Scores')
```

```
plt.title('Scores by group and gender')
```

```
plt.legend(('Men', 'Womem'))
```



Scores by group and gender



# Stacked Bar Chart

```
N = 5
```

```
m = (20, 35, 30, 35, 27)
```

```
w = (25, 32, 34, 30, 27)
```

```
x = np.arange(N)
```

```
width=0.35
```

```
plt.bar(x, m, width, color='#d62728')
```

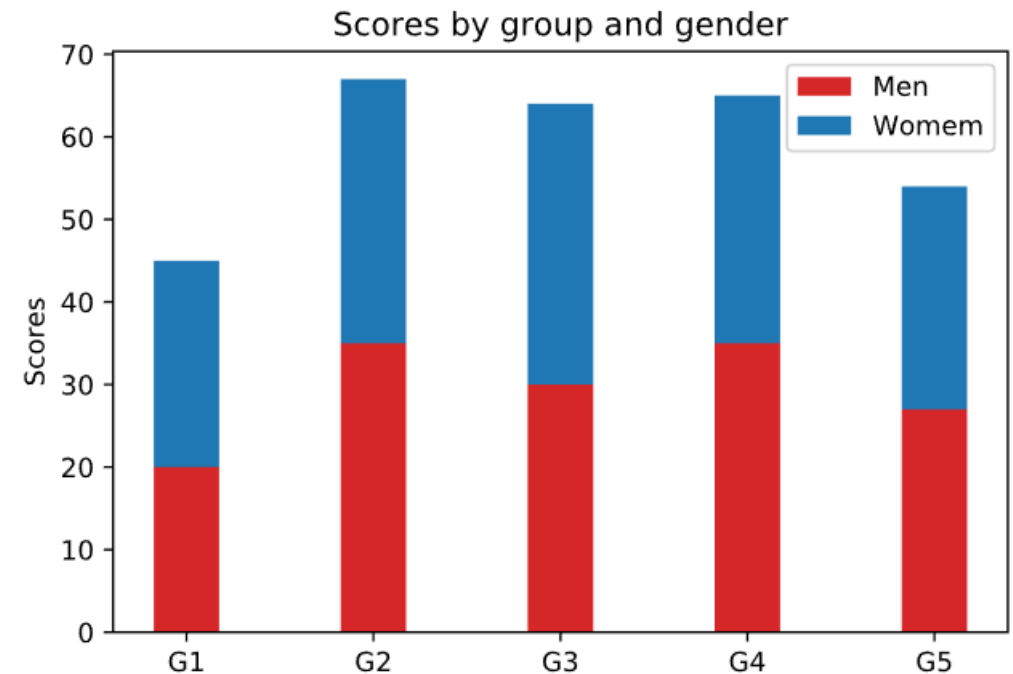
```
plt.bar(x, w, width, bottom=m)
```

```
plt.xticks(x, ('G1', 'G2', 'G3', 'G4', 'G5'))
```

```
plt.ylabel('Scores')
```

```
plt.title('Scores by group and gender')
```

```
plt.legend(('Men', 'Womem'))
```



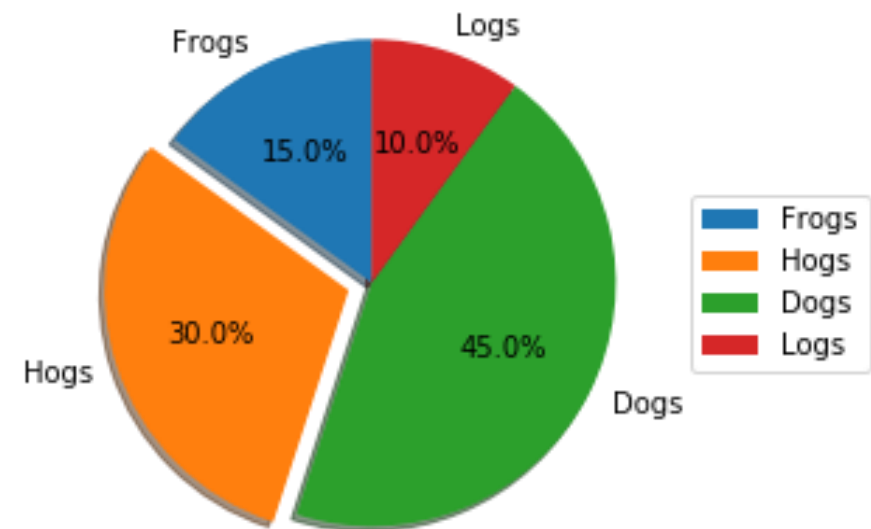
# Pie Chart

# bar()

- `plt.pie(x, [labels], [colors], [explode], [startangle], [autopct], ...)`

- Plot a pie chart
- `x`: the wedge sizes
- `labels`: a sequence of labels
- `colors`: a sequence of colors
- `explode`: the fraction of the radius with which to offset each wedge
- `startangle`: start pie chart with this degree
- `autopct`: label wedges with numeric value

```
labels = ['Frogs', 'Hogs', 'Dogs', 'Logs']
sizes = [15, 30, 45, 10]
ex = (0, 0.1, 0, 0)
plt.pie(sizes, explode=ex, labels=labels,
        autopct='%.1f%%', shadow=True, startangle=90)
plt.legend(loc='center right',
        bbox_to_anchor=(1.4, 0.5))
```



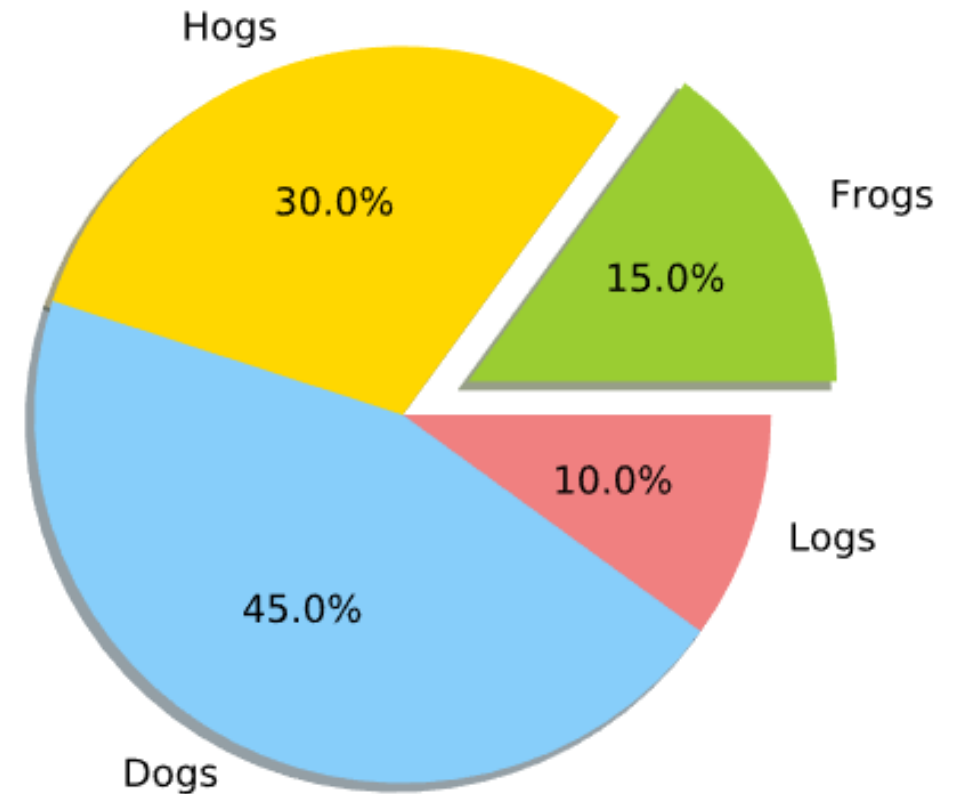


# Using Specific Colors

```
labels = ['Frogs', 'Hogs', 'Dogs', 'Logs']
sizes = [15, 30, 45, 10]
ex = (0.2, 0, 0, 0)
c = ['yellowgreen', 'gold',
     'lightskyblue', 'lightcoral']

plt.pie(sizes, explode=ex, labels=labels,
        colors=c, autopct='%1f%%',
        shadow=True, startangle=0)

plt.axis('equal')
```

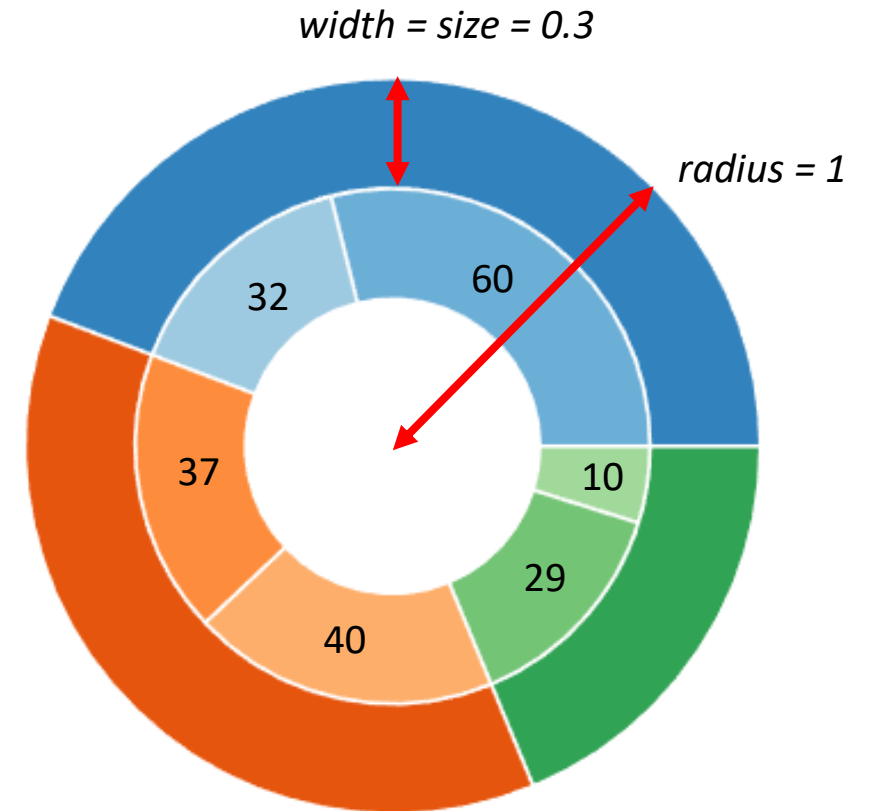


# Nested Pie Chart

```
size = 0.3
vals = np.array([[60.,32.],[37.,40.],[29.,10.]])

cmap = plt.get_cmap('tab20c')
outer_colors = cmap(np.arange(3)*4)
inner_colors = cmap(np.array([1,2,5,6,9,10]))

plt.pie(vals.sum(axis=1), radius=1,
        colors=outer_colors,
        wedgeprops=dict(width=size, edgecolor='w'))
plt.pie(vals.flatten(), radius=1-size,
        colors=inner_colors,
        wedgeprops=dict(width=size, edgecolor='w'))
plt.axis('equal')
```



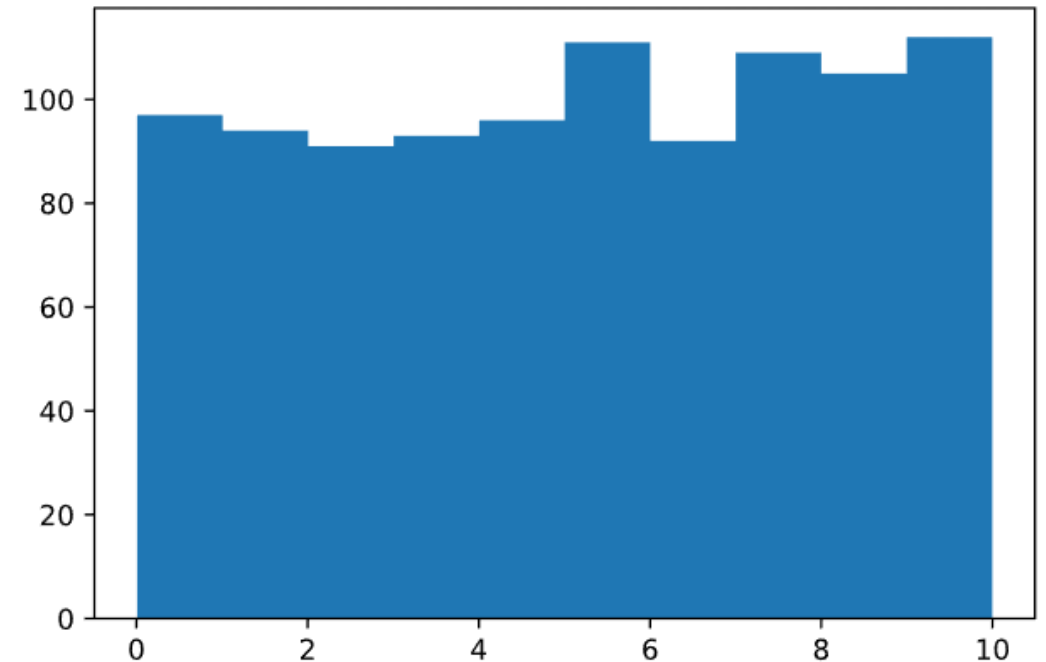
# Histogram

# hist()

- `plt.hist(x, [bins], [range], [cumulative], [bottom], ...)`

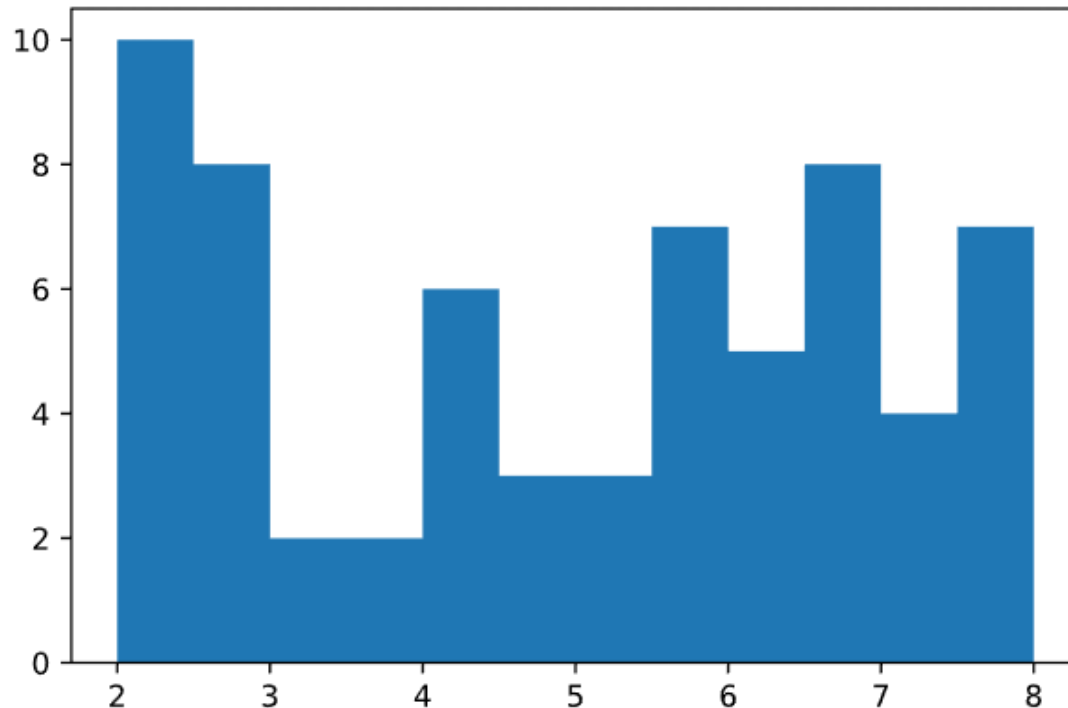
- Plot a histogram
- `x`: input values
- `bins`: number of bins (default: 10)
- `range`: the lower and upper range of the bins
- `cumulative`: if True, shows the cumulative sum
- `bottom`: the baseline of each bin

```
x = np.random.uniform(0., 10., 1000)
plt.hist(x)
```



# Range and Bins

```
x = np.random.uniform(0., 10., 100)
plt.hist(x, range=[2., 8.], bins=12)
```



**Split [2.0, 8.0] into 12 bins:**

bin 0 has the count of  $2. \leq x < 2.5$

bin 1 has the count of  $2.5 \leq x < 3.0$

bin  $i$  has the count of  $2.+0.5i \leq x < 2.5+0.5i$

bin 11 has the count of  $7.5 \leq x < 8.0$

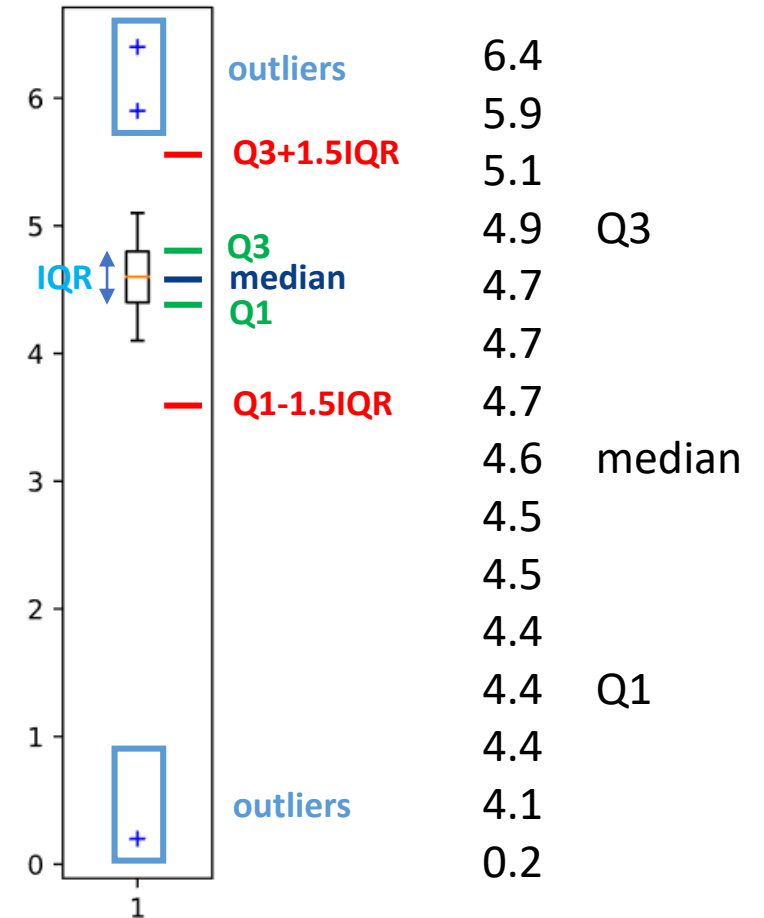
# Box Plot

# boxplot()

- `plt.boxplot(x, [notch], [sym], [vert], [whis], ...)`
  - Make a box and whisker plot
  - `x`: input data
  - `notch`: If True, produce a notched box plot (default: False)
  - `sym`: the symbol for outliers
  - `vert`: If True, make the boxes vertical (default: True)
  - `whis`: the reach of the whiskers (default: 1.5)

(cf.) IQR = InterQuartile Range

```
x = [0.2, 4.1, 4.4, 4.4, 4.4, 4.5, 4.5, 4.6,  
     4.7, 4.7, 4.7, 4.9, 5.1, 5.9, 6.4]  
plt.boxplot(x, sym='b+')
```



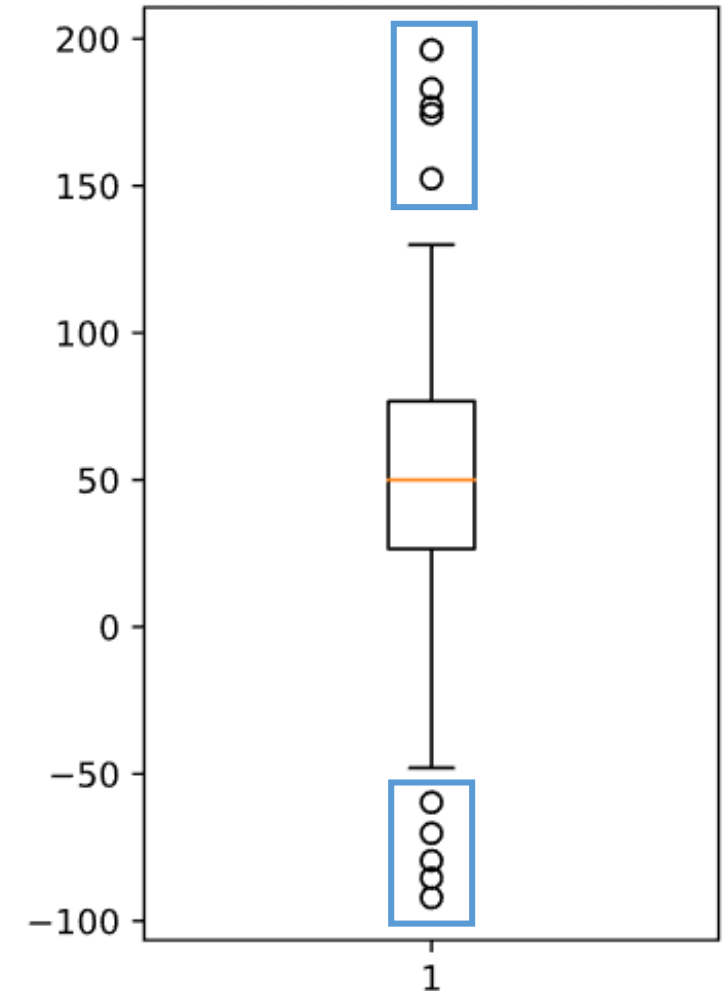
$IQR = Q3 - Q1 = 4.9 - 4.4 = 0.5$   
 $Q1 - 1.5IQR = 4.4 - 1.5 \times 0.5 = 3.65$   
 $Q3 + 1.5IQR = 4.9 + 1.5 \times 0.5 = 5.65$

# Box Plot Example

```
spread = np.random.rand(50)*100
center = np.ones(25)*50
high = np.random.rand(10)*100+100
low = np.random.rand(10)*(-100)

data = np.concatenate((spread, center, high, low), 0)
plt.figure(figsize=(3, 5)) # fig size in inches
plt.boxplot(data)
```

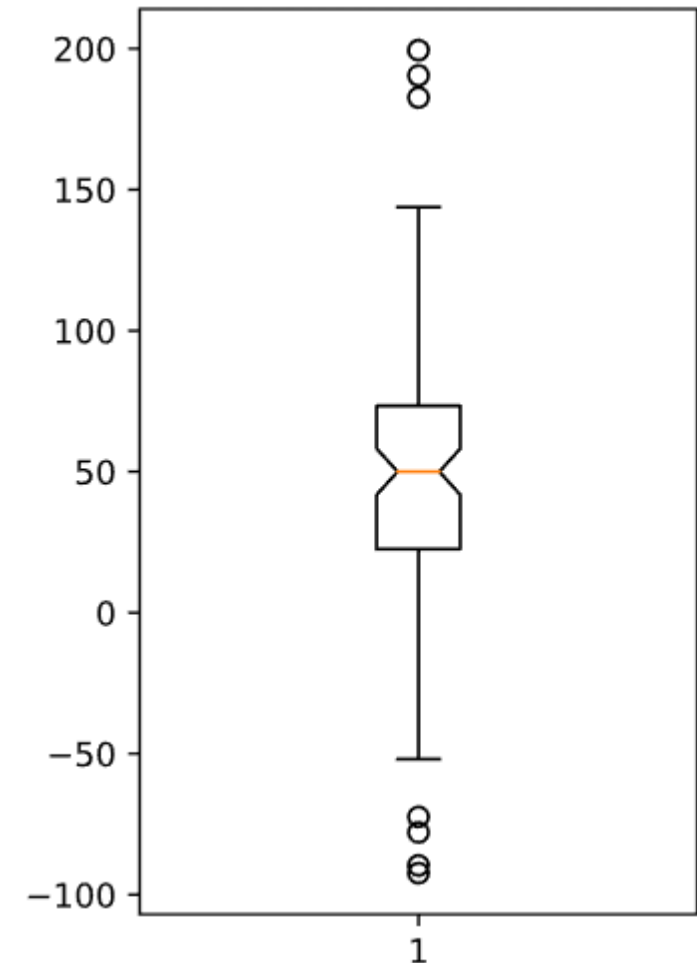
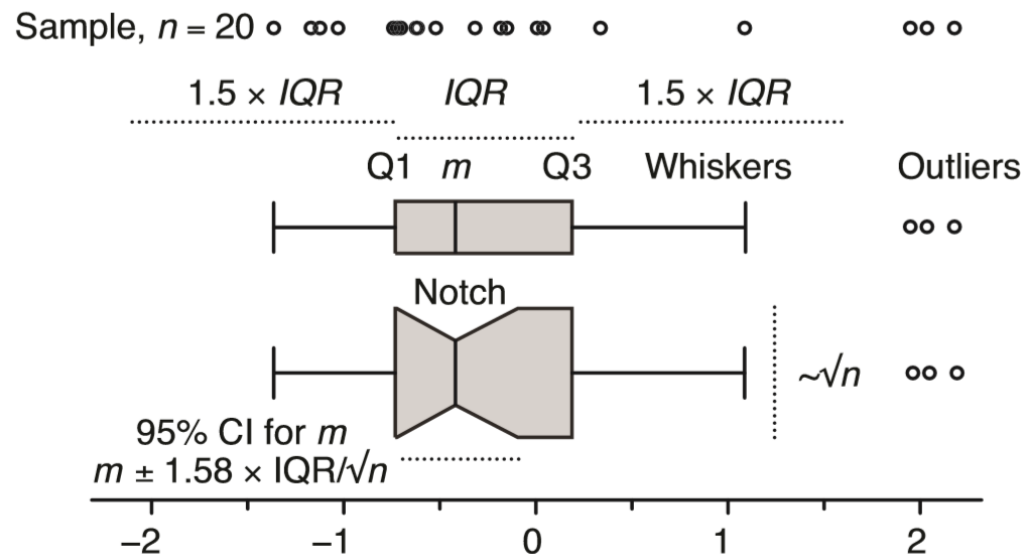
- spread: 50 numbers in  $[0, 100)$
- center: 25 numbers with the value 50.0
- high: 10 numbers in  $[100, 200)$
- low: 10 numbers in  $(-100, 0]$





# Notched Box Plot

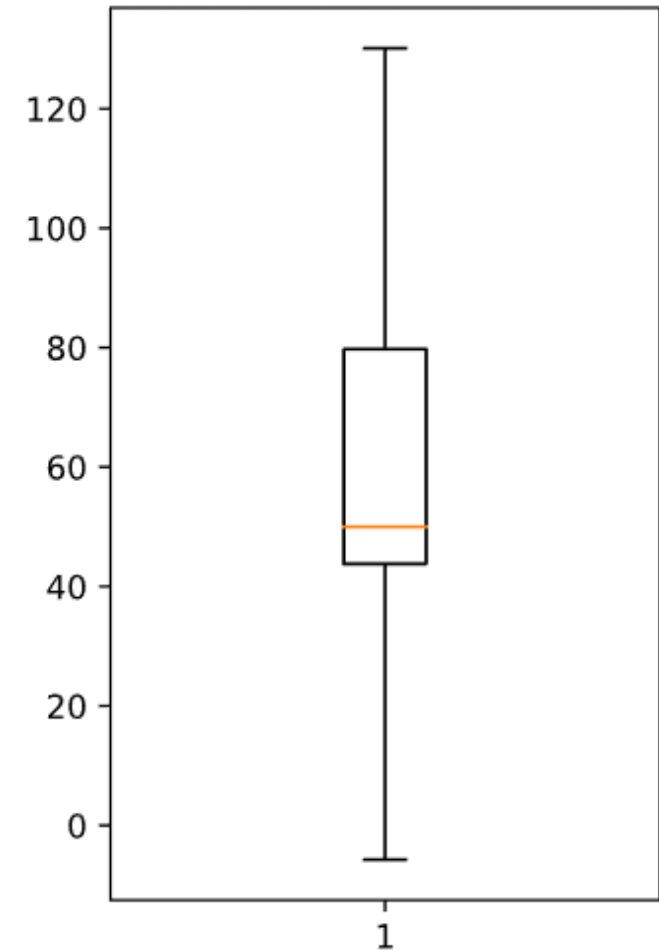
```
spread = np.random.rand(50)*100
center = np.ones(25)*50
high = np.random.rand(10)*100+100
low = np.random.rand(10)*(-100)
data = np.concatenate((spread, center, high, low), 0)
plt.figure(figsize=(3, 5)) # fig size in inches
plt.boxplot(data, notch=True)
```



# Removing Outliers

```
spread = np.random.rand(50)*100
center = np.ones(25)*50
high = np.random.rand(10)*100+100
low = np.random.rand(10)*(-100)
data = np.concatenate((spread, center, high, low), 0)

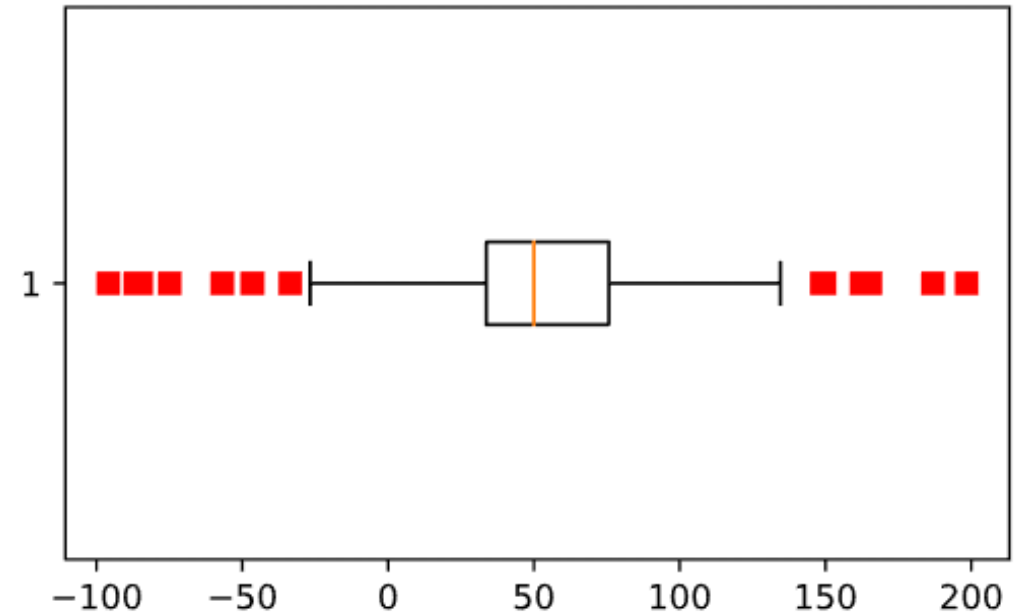
plt.figure(figsize=(3, 5)) # fig size in inches
plt.boxplot(data, sym=' ')
```



# Horizontal Box Plot

```
spread = np.random.rand(50)*100
center = np.ones(25)*50
high = np.random.rand(10)*100+100
low = np.random.rand(10)*(-100)
data = np.concatenate((spread, center,
                        high, low), 0)

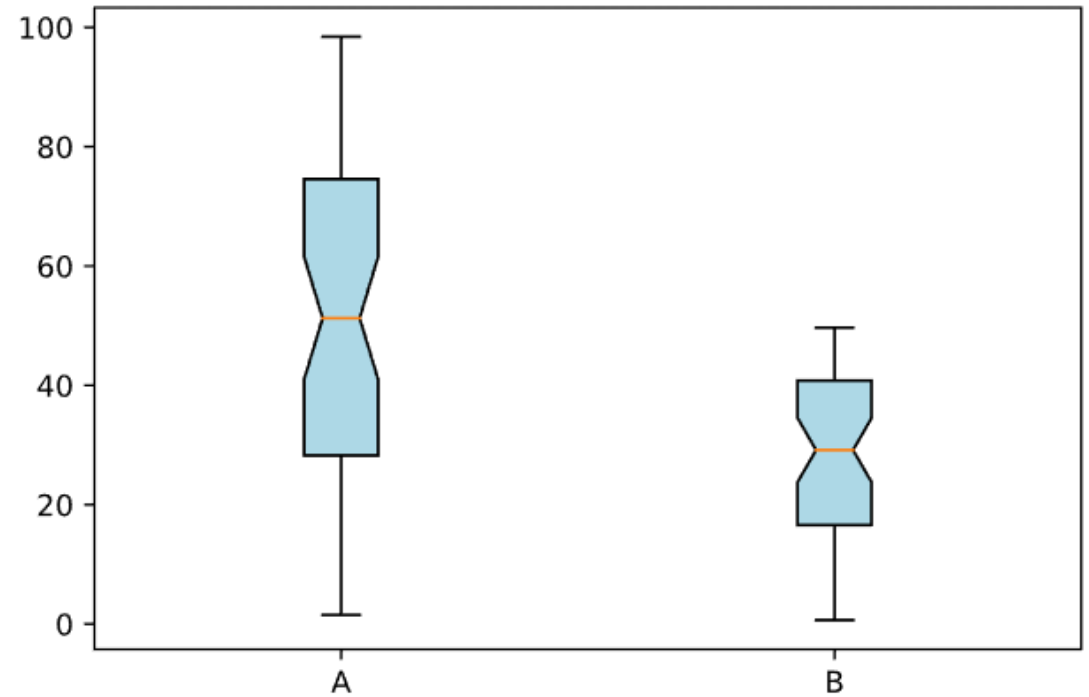
plt.figure(figsize=(5, 3))
plt.boxplot(data, vert=False, sym='rs')
```



# Multiple Box Plots

```
p1 = np.random.rand(50)*100
p2 = np.random.rand(50)*50
data = [p1, p2]

plt.boxplot(data, notch=True,
             patch_artist=True,
             boxprops=dict(facecolor='lightblue'),
             labels=['A', 'B'])
plt.boxplot(data)
```



# Subplots

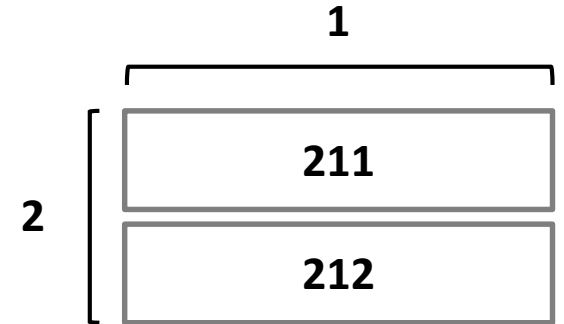
# figure()

- `plt.figure([num], [figsize], [facecolor], [edgecolor], [frameon], ...)`
  - Create a new figure
  - `num`: If not provided, a new figure will be created with the figure number incremented. If provided, a figure with this id is created
  - `figsize`: (width, height) in inches
  - `facecolor`: the background color
  - `edgecolor`: the border color
  - `frameon`: If True, draw the figure frame (default: True)

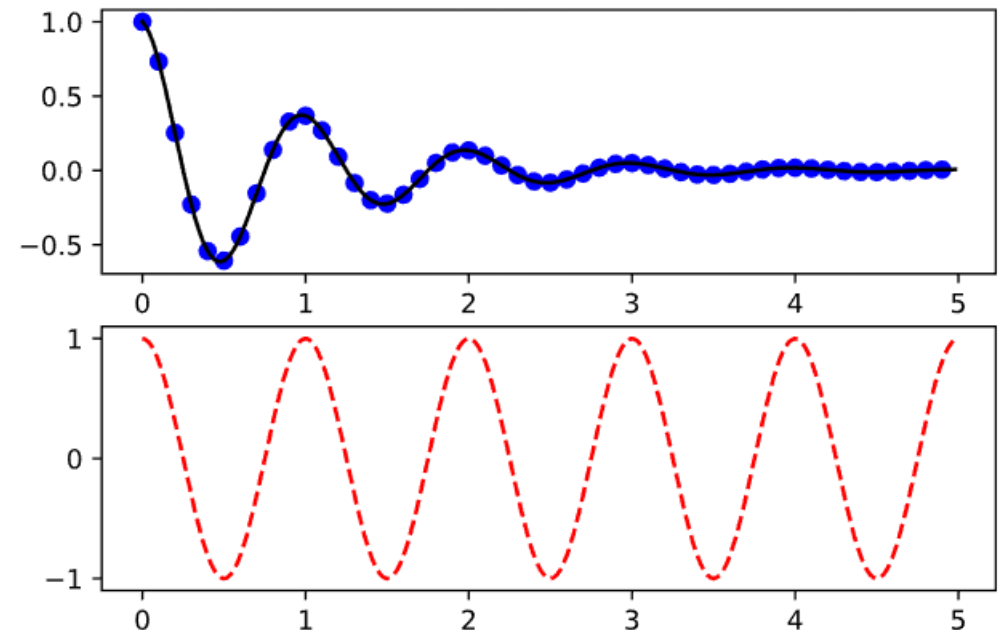
# subplot()

## ■ `plt.subplot([pos], ...)`

- Add a subplot to the current figure
- `pos`: a three digit integer denoting the number of rows, the number of columns, and the index of the subplot
- Returns the axes of the subplot



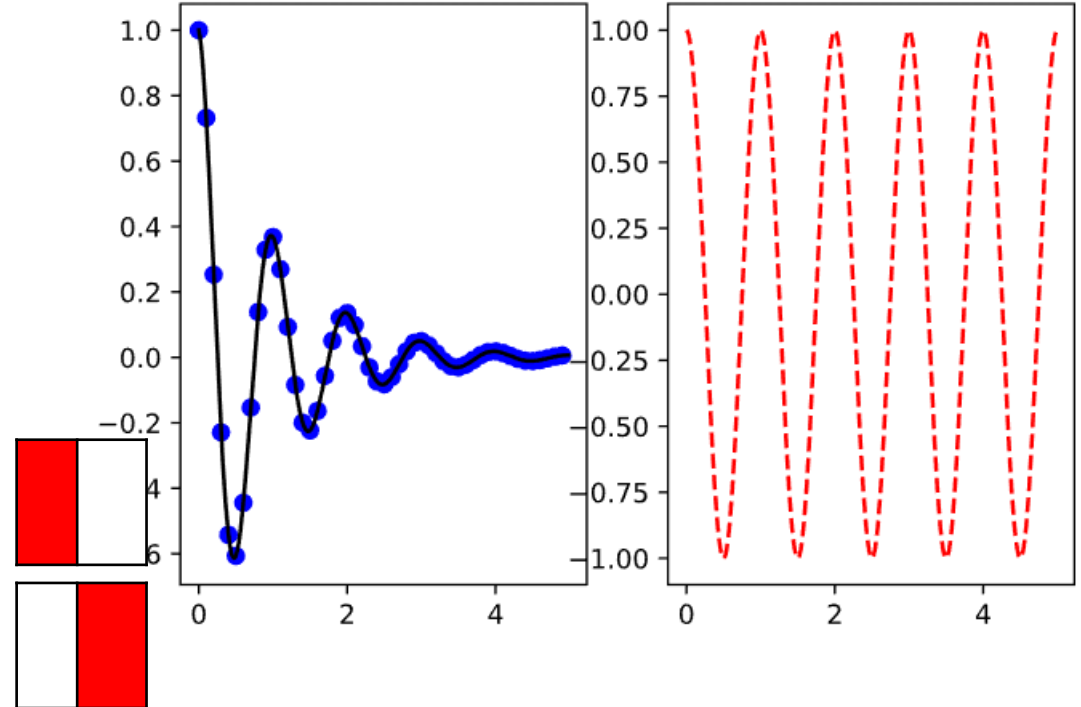
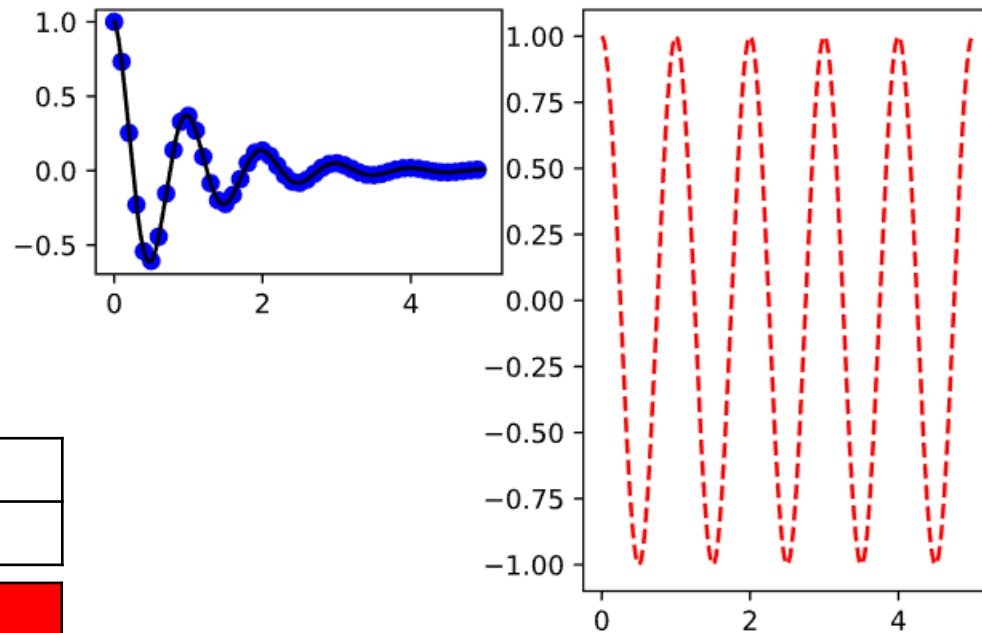
```
def f(t):  
    return np.exp(-t)*np.cos(2*np.pi*t)  
t1 = np.arange(0.0, 5.0, 0.1)  
t2 = np.arange(0.0, 5.0, 0.02)  
plt.figure(1)  
plt.subplot(211)  
plt.plot(t1, f(t1), 'bo', t2, f(t2), 'k')  
plt.subplot(212)  
plt.plot(t2, np.cos(2*np.pi*t2), 'r--')
```



# Subplot Examples

```
plt.subplot(221)  
plt.plot(t1, f(t1), 'bo', t2, f(t2), 'k')  
plt.subplot(122)  
plt.plot(t2, np.cos(2*np.pi*t2), 'r--')
```

```
plt.subplot(121)  
plt.plot(t1, f(t1), 'bo', t2, f(t2), 'k')  
plt.subplot(122)  
plt.plot(t2, np.cos(2*np.pi*t2), 'r--')
```





Seaborn

# What is "Seaborn" Module?

- Python visualization library based on matplotlib
  - More attractive and informative statistical graphics
  - Wrapper of matplotlib which improved the default styles
  - Even if plotting using only matplotlib.pyplot, importing seaborn will make nicer plots
- Dependency
  - Data structures: NumPy, Pandas
  - Statistical routines: SciPy
- Open source
  - <https://seaborn.pydata.org>

```
>>> import seaborn as sns
```

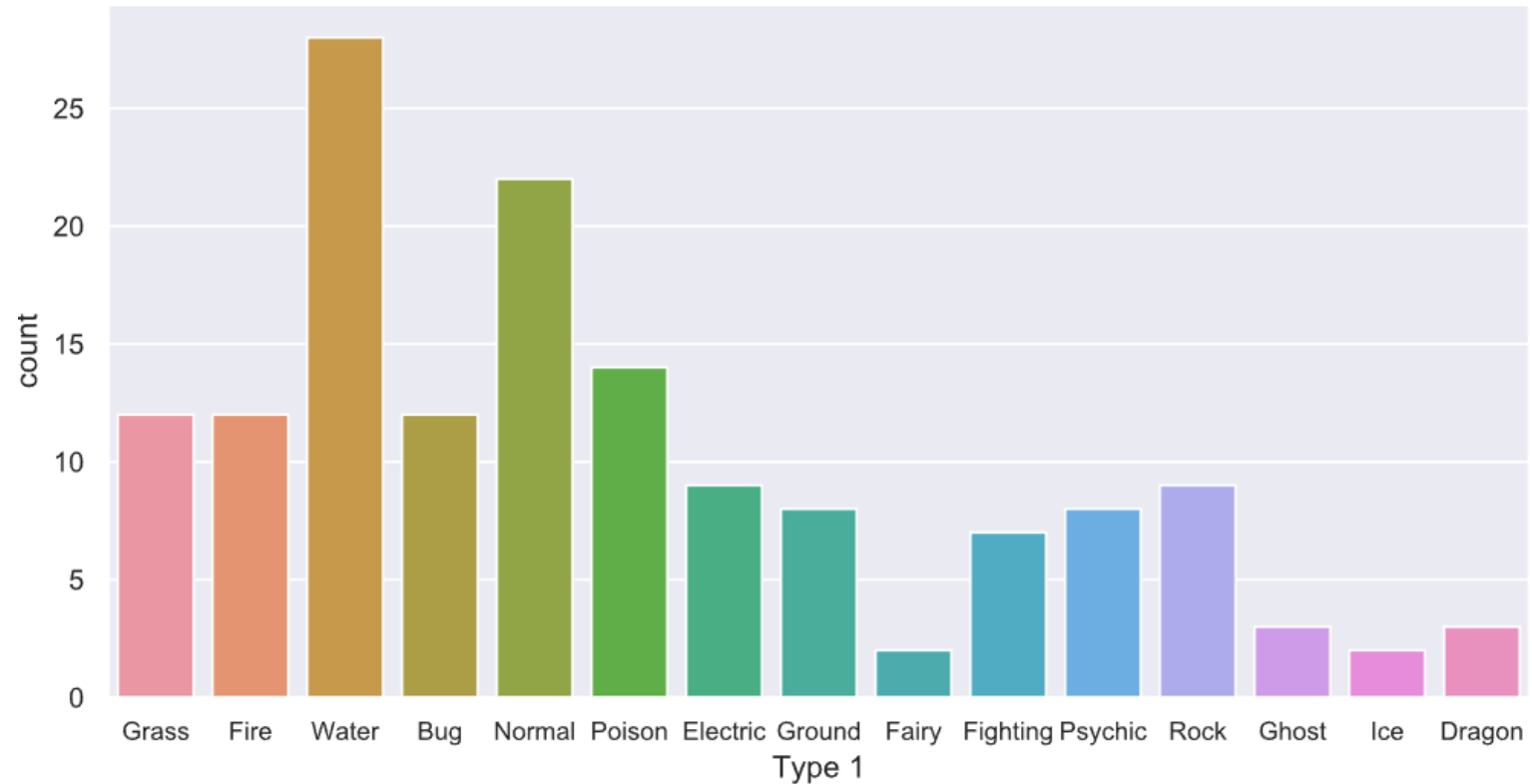
# Pokemon Dataset

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
df = pd.read_csv('pokemon.csv')
df.head()
```

	Name	Type 1	Type 2	Total	HP	Attack	Defense	Sp. Atk	Sp. Def	Speed	Stage	Legendary
#												
1	Bulbasaur	Grass	Poison	318	45	49	49	65	65	45	1	False
2	Ivysaur	Grass	Poison	405	60	62	63	80	80	60	2	False
3	Venusaur	Grass	Poison	525	80	82	83	100	100	80	3	False
4	Charmander	Fire	NaN	309	39	52	43	60	50	65	1	False
5	Charmeleon	Fire	NaN	405	58	64	58	80	65	80	2	False

# Count Plot

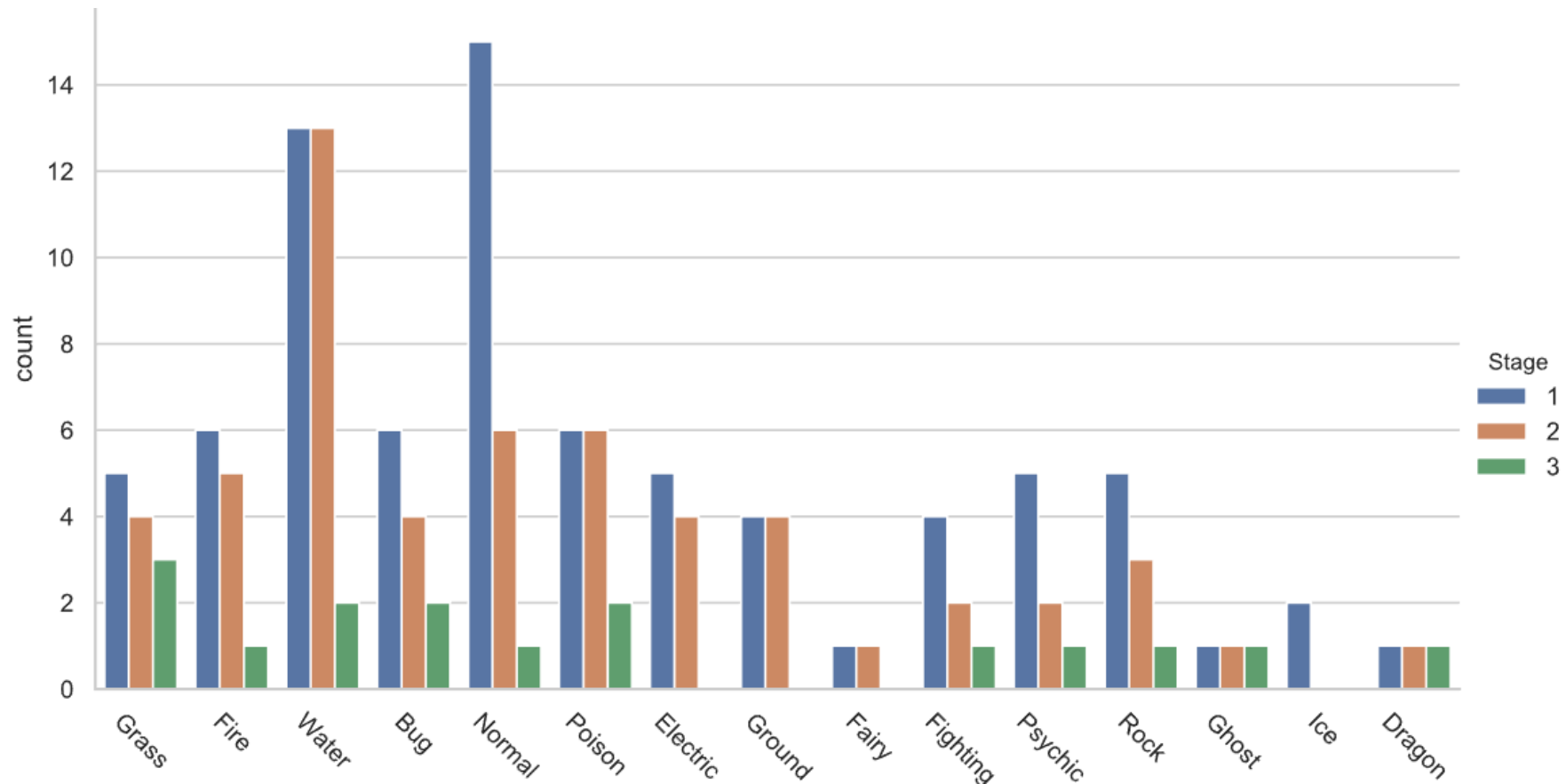
```
sns.set(style='darkgrid')  
plt.figure(figsize=(10, 5))  
plt.xticks(fontsize=10)  
sns.countplot(x='Type 1', data=df)
```



# Categorical Plot

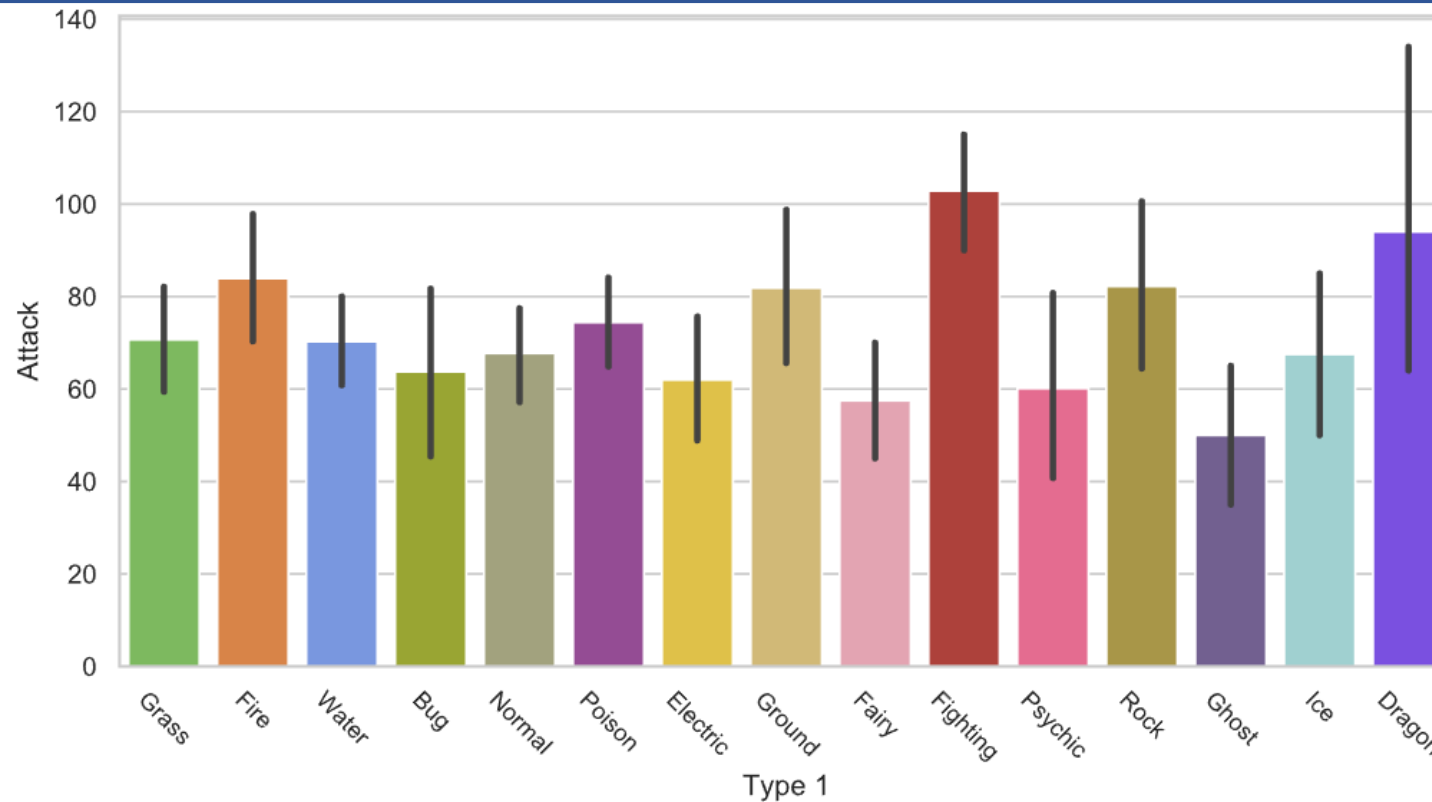
Group data

```
sns.catplot(x='Type 1', kind='count', hue='Stage', data=df, aspect=1.8)  
plt.xticks(rotation=-45)
```



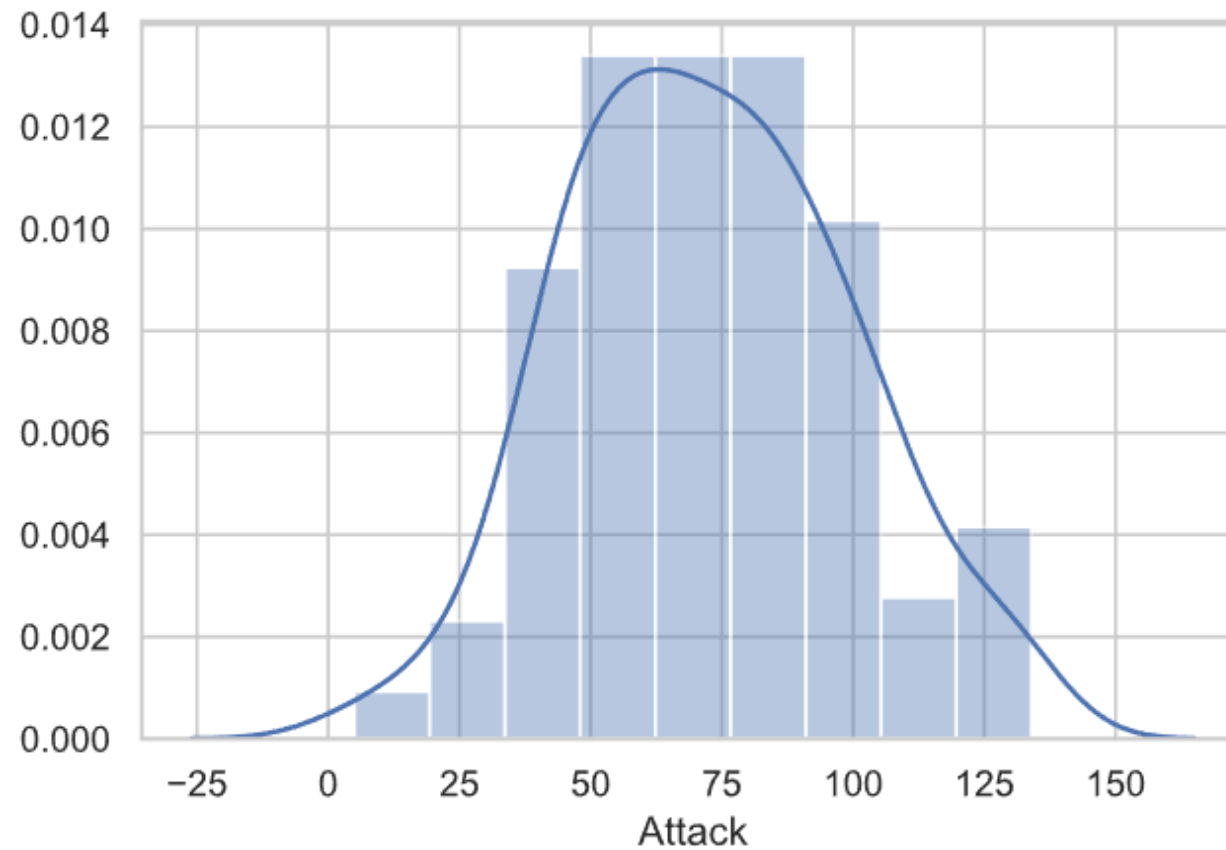
# Bar Plot

```
plt.figure(figsize=(10, 5))  
plt.xticks(fontsize=10, rotation=-45)  
sns.barplot(x='Type 1', y='Attack', data=df, palette=poke_colors)
```



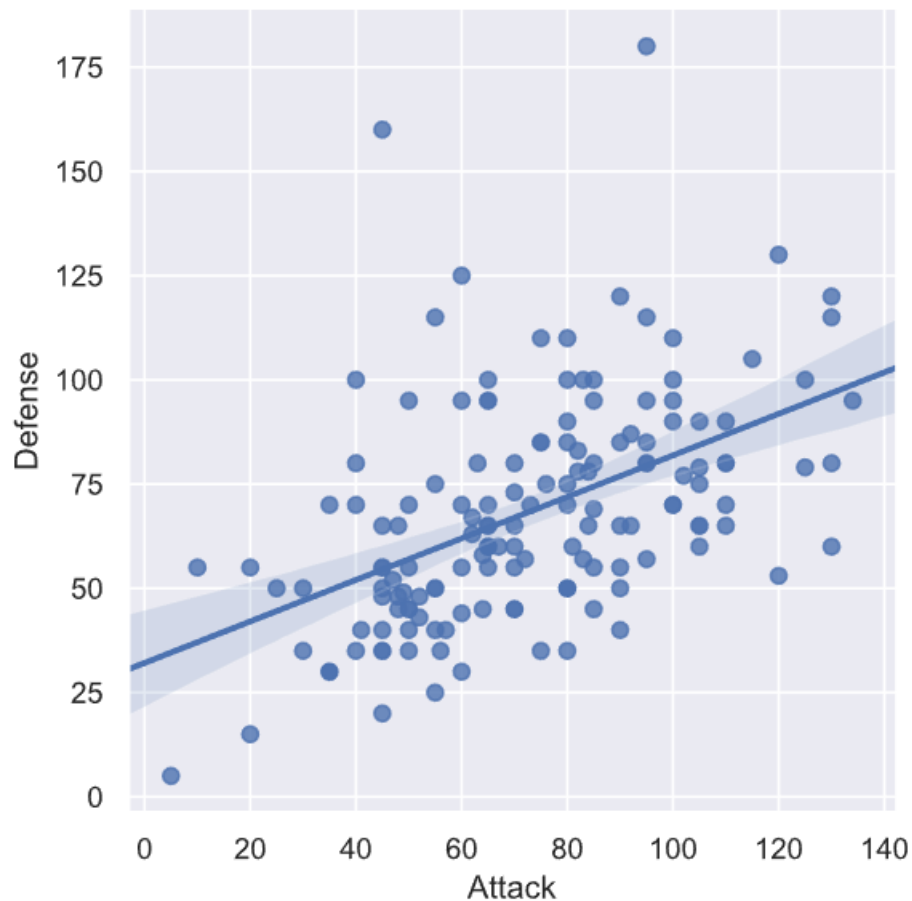
# Histogram

```
sns.distplot(df.Attack)
```

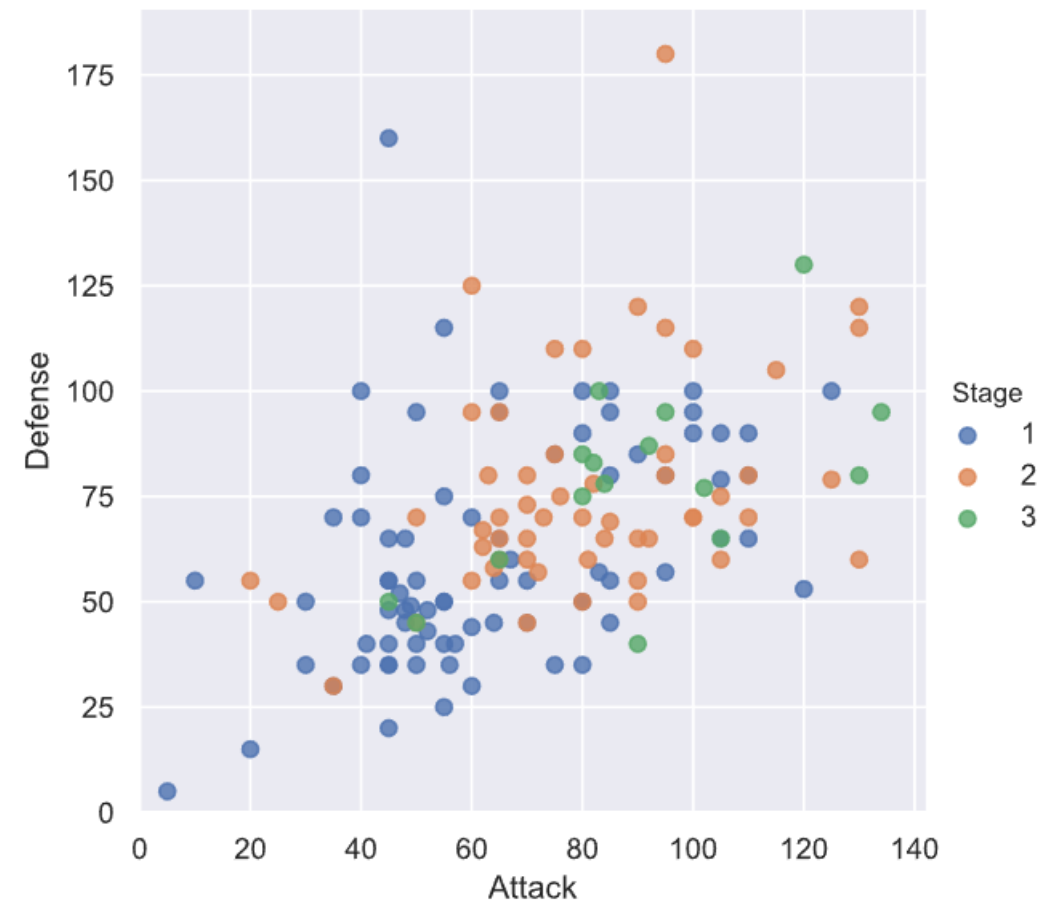


# Scatter Plot

```
sns.lmplot(x='Attack', y='Defense', data=df)
```



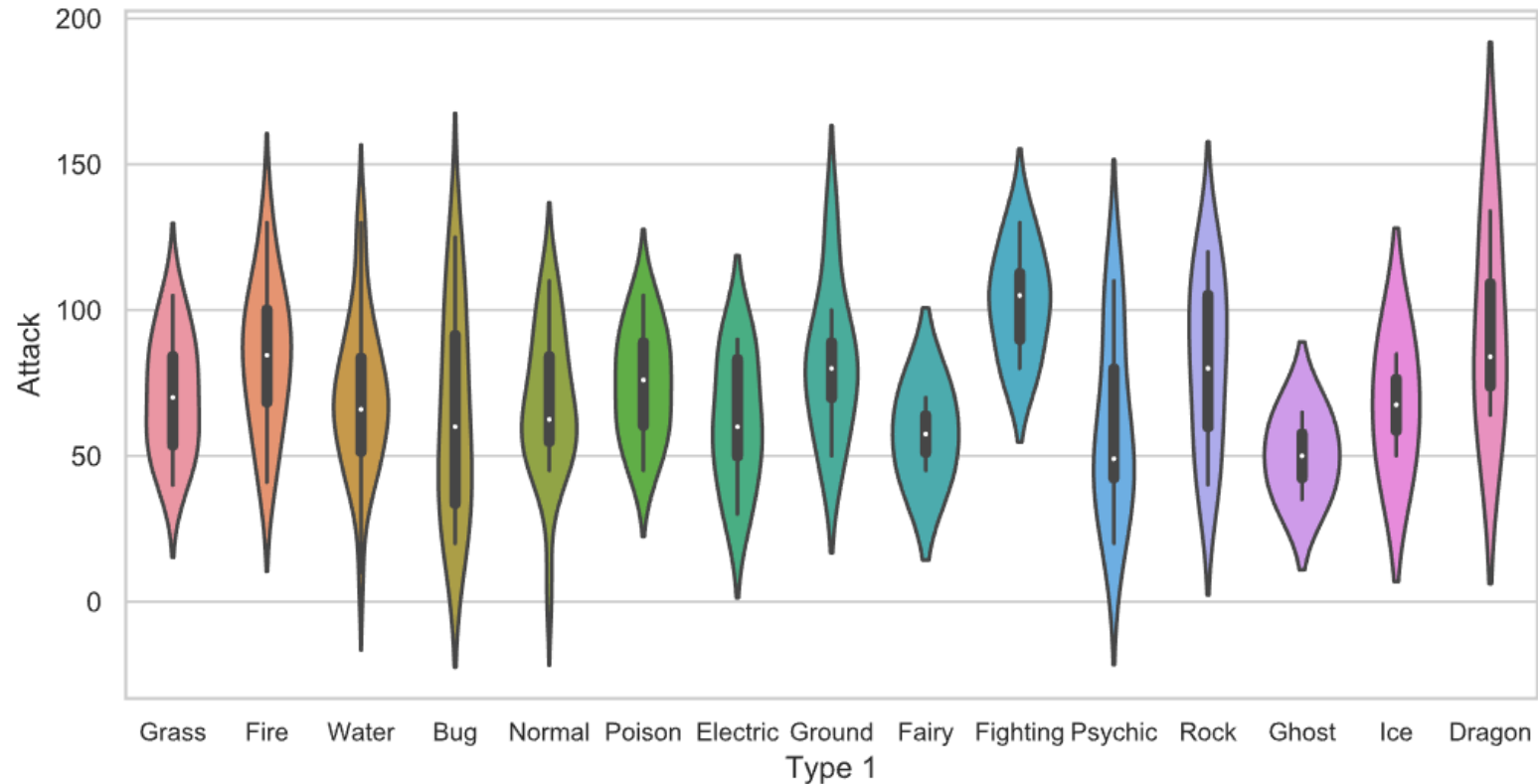
```
sns.lmplot(x='Attack', y='Defense',  
           fit_reg=False, hue='Stage', data=df)
```





# Violin Plot

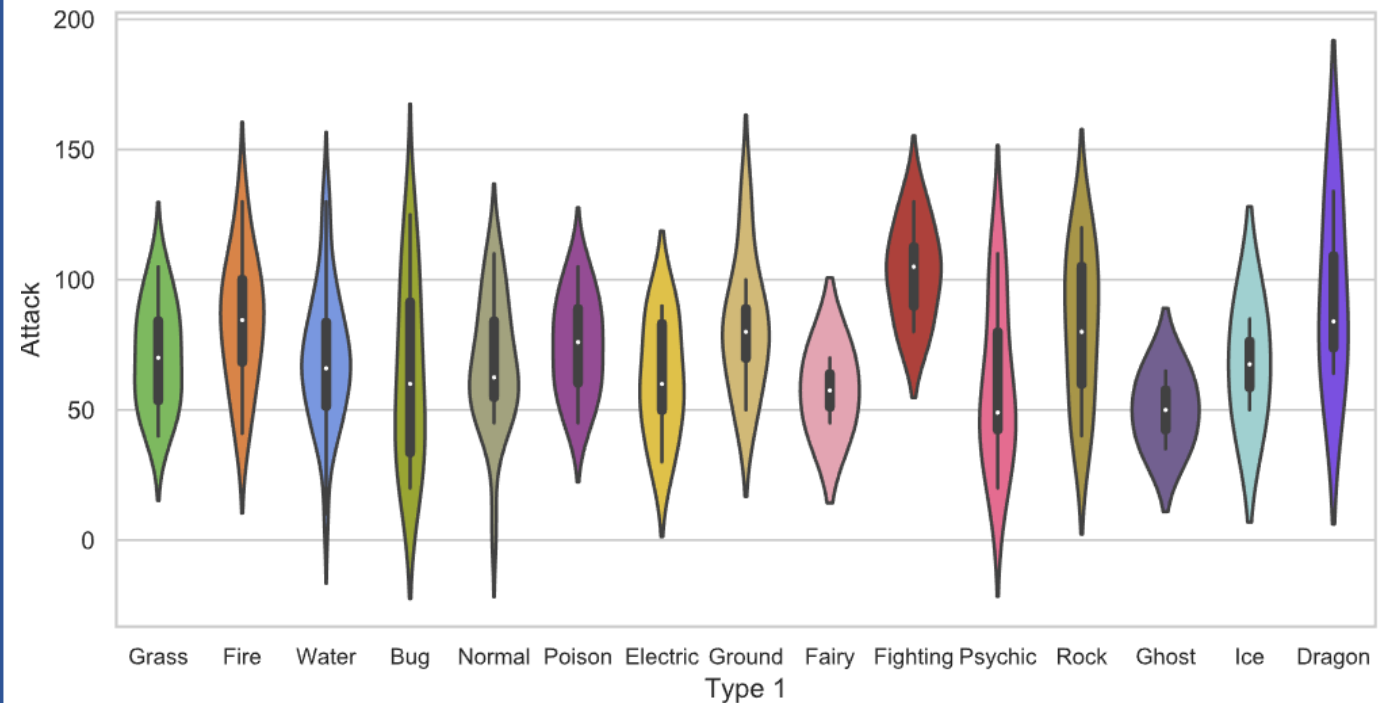
```
sns.set_style('whitegrid')  
plt.figure(figsize=(10, 5))  
plt.xticks(fontsize=10)  
sns.violinplot(x='Type 1', y='Attack', data=df)
```



# Violin Plot (with Customized Colors)

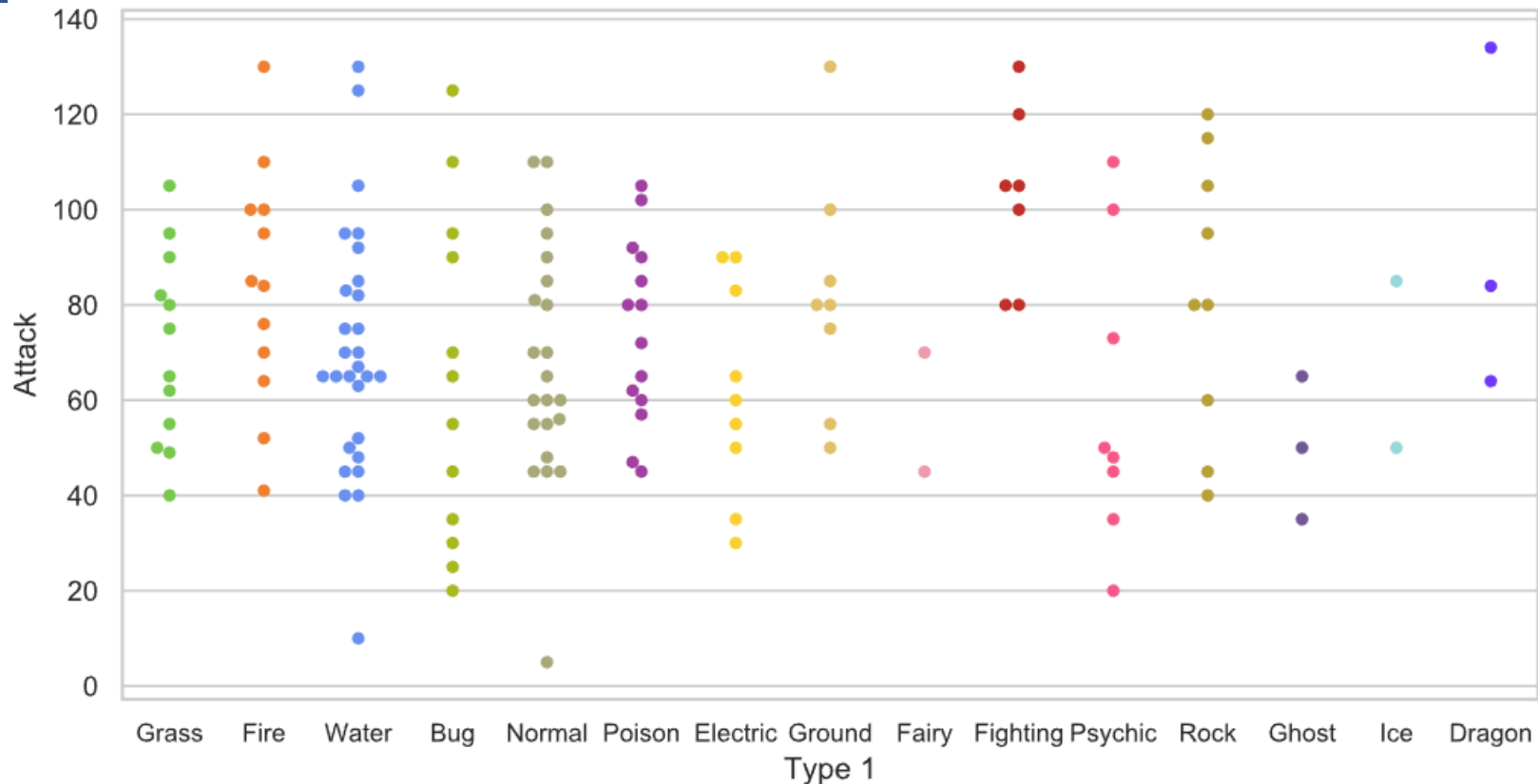
```
poke_colors = ['#78C850', # Grass  
               '#F08030', # Fire  
               '#6890F0', # Water  
               '#A8B820', # Bug  
               '#A8A878', # Normal  
               '#A040A0', # Poison  
               '#F8D030', # Electric  
               '#E0C068', # Ground  
               '#EE99AC', # Fairy  
               '#C03028', # Fighting  
               '#F85888', # Psychic  
               '#B8A038', # Rock  
               '#705898', # Ghost  
               '#98D8D8', # Ice  
               '#7038F8', # Dragon  
               ]
```

```
plt.figure(figsize=(10, 5))  
plt.xticks(fontsize=10)  
sns.violinplot(x='Type 1', y='Attack', data=df,  
               palette=poke_colors)
```



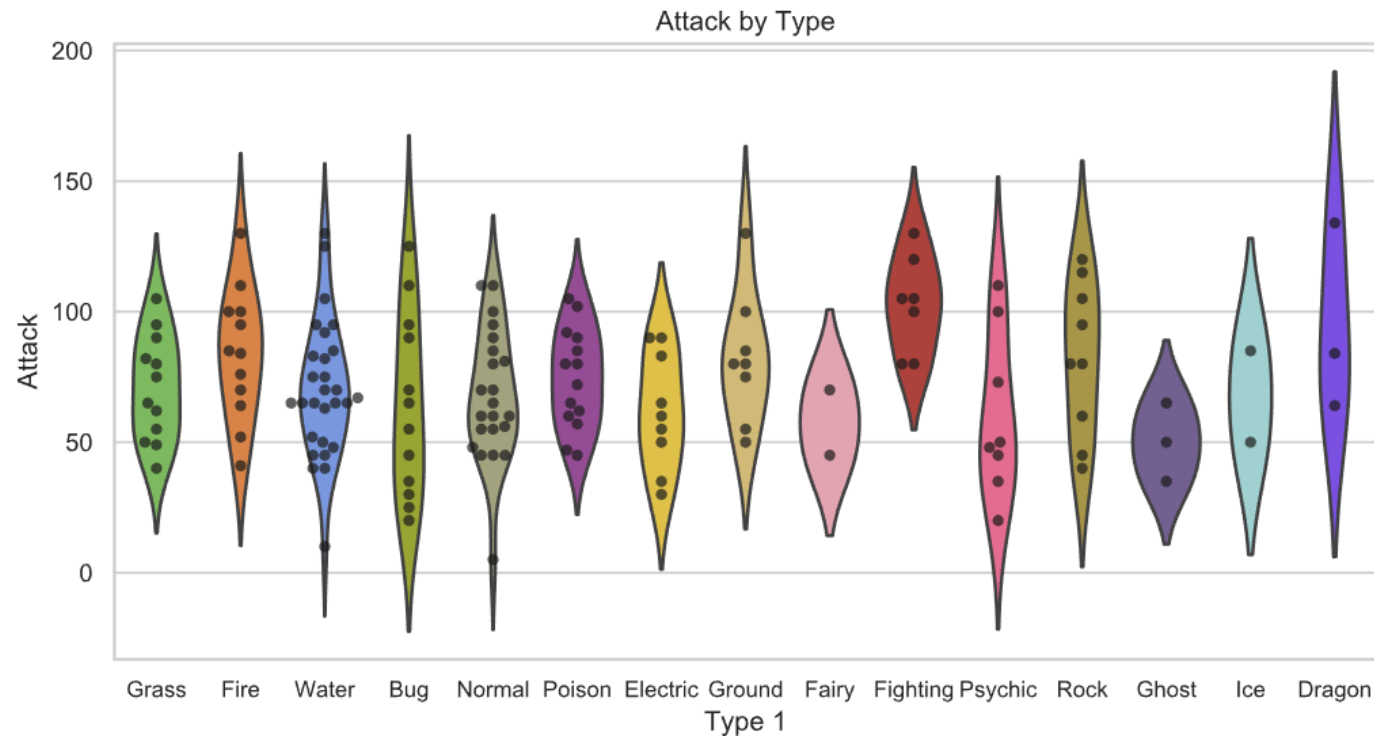
# Swarm Plot

```
plt.figure(figsize=(10, 5))  
plt.xticks(fontsize=10)  
sns.swarmplot(x='Type 1', y='Attack', data=df, palette=poke_colors)
```



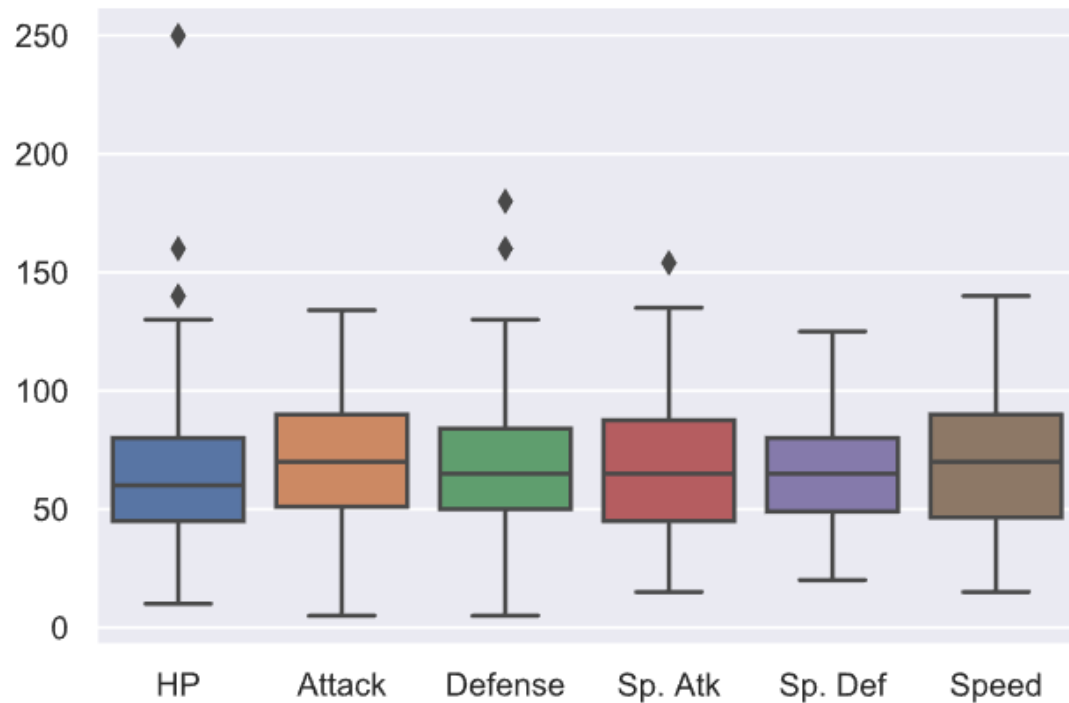
# Violin + Swarm Plot

```
plt.figure(figsize=(10, 5))
plt.xticks(fontsize=10)
sns.violinplot(x='Type 1', y='Attack', data=df, palette=poke_colors, inner=None)
sns.swarmplot(x='Type 1', y='Attack', data=df, color='k', alpha=0.7)
plt.title('Attack by Type')
```

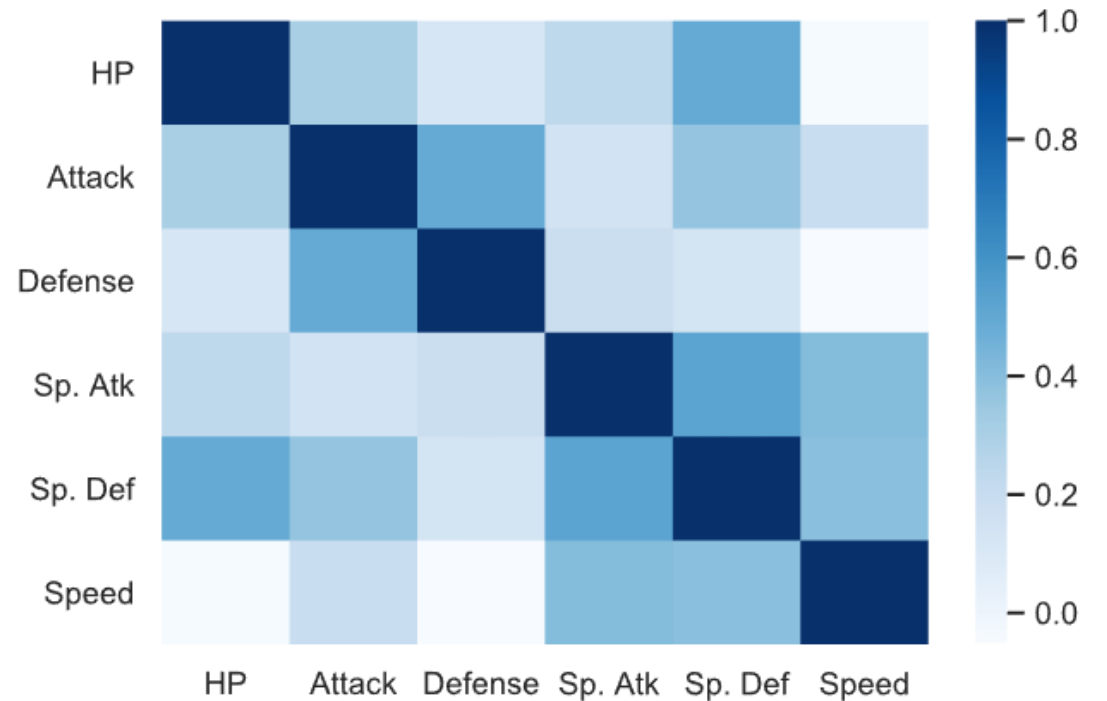


# Box Plot and Heatmap

```
stats_df = df.drop(['Total', 'Stage',  
                    'Legendary'], axis=1)  
sns.boxplot(data=stats_df)
```

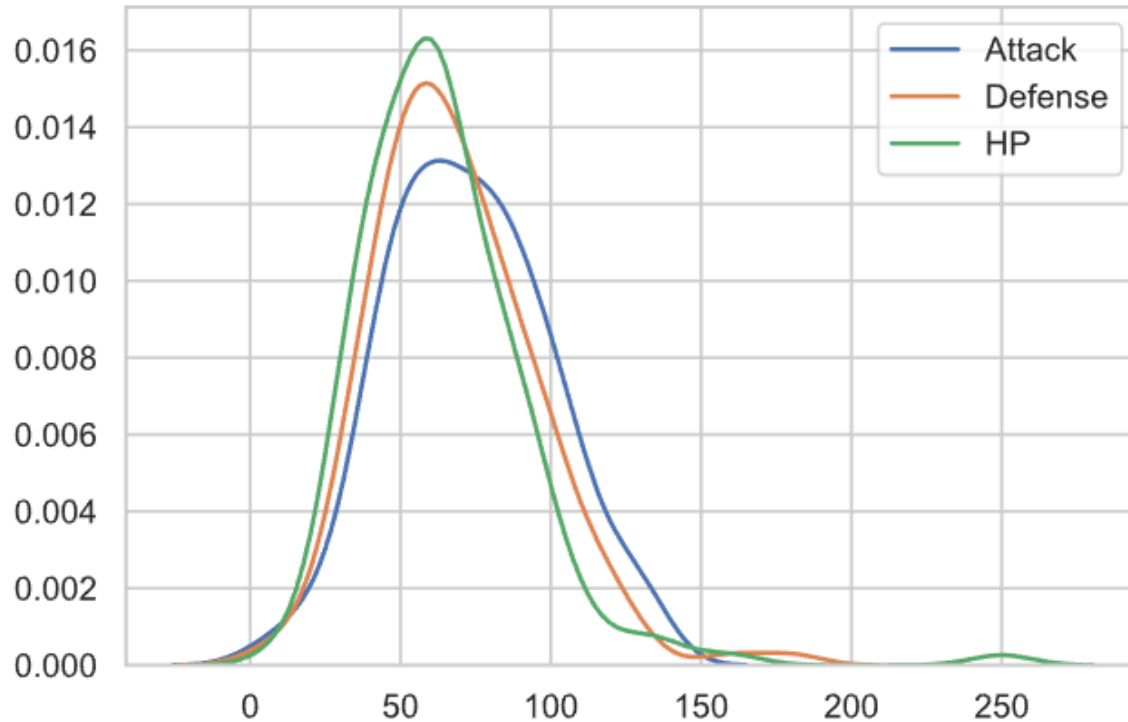


```
corr = stats_df.corr()  
sns.heatmap(corr, cmap='Blues')
```

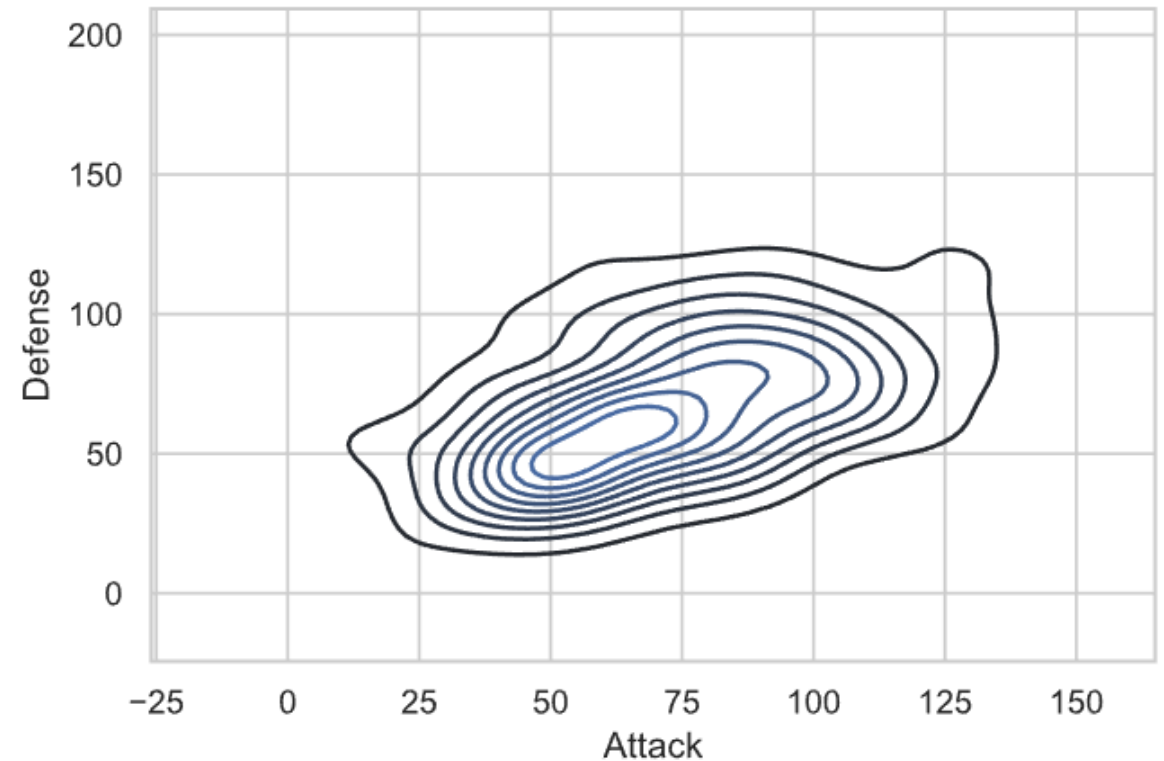


# KDE Plot

```
sns.kdeplot(df.Attack)  
sns.kdeplot(df.Defense)  
sns.kdeplot(df.HP)
```



```
sns.kdeplot(df.Attack, df.Defense)
```



# Pairplot

```
s.pairplot(df[['Type 1',  
              'Attack',  
              'Defense',  
              'HP',  
              'Speed']],  
          hue='Type 1')
```

