## Bivariate visualizations

INTRODUCTION TO DATA VISUALIZATION WITH PLOTLY IN PYTHON



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### What are bivariate visualizations?

Bivariate plots are those which display (and can therefore compare) two variables.

Common bivariate plots include:

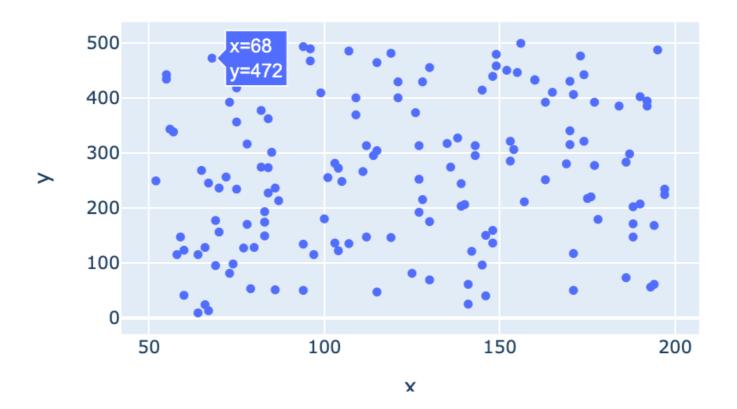
- scatterplots
- Correlation plots
- Line charts



### scatterplot

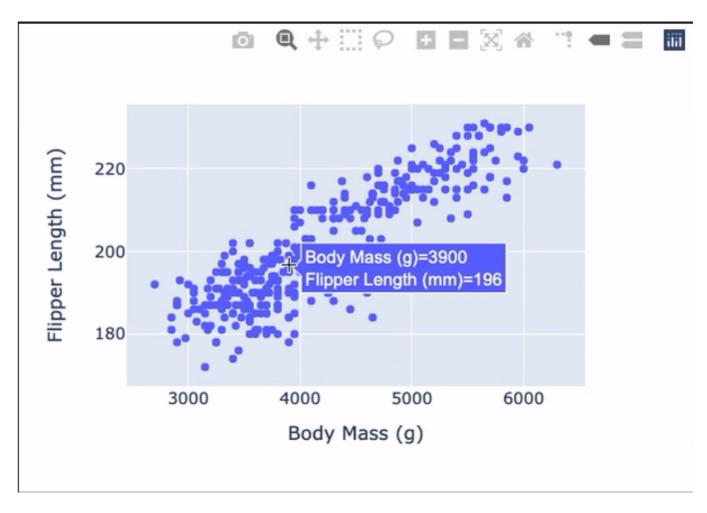
A scatterplot is a plot consisting of:

- A y-axis representing one variable
- An x-axis representing a different variable
- Each point is a dot on the graph, e.g., (68, 472)



### scatterplot with plotly.express

Visualizing Flipper Length and Body Mass with plotly.express:



### More plotly.express arguments

Useful plotly.express scatterplot arguments:

- trendline: Add different types of trend lines
- symbol : Set different symbols for different categories

Check the documentation for more!



### Line charts in plotly.express

A line chart is used to plot some variable (y-axis) over time (x-axis).

Let's visualize Microsoft's stock price.

```
fig = px.line(
  data_frame=msft_stock,
  x='Date',
  y='Open',
  title='MSFT Stock Price (5Y)')
fig.show()
```

#### Here is our simple line chart:

MSFT Stock Price (5Y)



### scatterplots and line plots with graph\_objects

For more customization, graph\_objects uses go.Scatter() for both scatter and line plots.

Here is the code for our penguins scatterplot using graph\_objects

Here is the code for our line chart with graph\_objects

- Remember to set 'mode'
  - And use DataFrame subsets, not column names

```
import plotly.graph_objects as go
fig = go Figure(go Scatter(
```

```
fig = go.Figure(go.Scatter(
    x=penguins['Body Mass (g)'],
    y=penguins['Flipper Length (mm)'],
    mode='markers'))
```

```
fig = go.Figure(go.Scatter(
    x=msft_stock['Date'],
    y=msft_stock['Opening Stock Price'],
    mode='lines'))
```

### graph\_objects vs. plotly.express?

When should we use plotly.express or graph\_objects? Largely, it is about customization - graph\_objects has many more options!

graph_objects	express
plotly.graph_objects. Scatter (arg=None, cliponaxis=None, connectgaps=None, customdata=None, customdatasrc=None, dx=None, dy=None, error_x=None, error_y=None, fill=None, fillcolor=None, groupnorm=None, hoverinfo=None, hoverinfosrc=None, hoverlabel=None, hoveron=None, hovertemplatesrc=None, hovertextsrc=None, ids=None, idssrc=None, legendgroup=None, hovertemplatesrc=None, marker=None, meta=None, metasrc=None, mode=None, name=None, opacity=None, orientation=None, r=None, rsrc=None, selected=None, selected=None, showlegend=None, stackgaps=None, stackgroup=None, stream=None, text=None, text=None, textfont=None, textposition=None, textpositionsrc=None, textsrc=None, textemplatesrc=None, text=None, uid=None, uirevision=None, unselected=None, visible=None, x=None, xaxis=None, xaxis=None, xaxis=None, xperiod=None, yperiod=None, yperiod=None, yperiod0=None, yperiodd=None,	plotly.express. scatter (data_frame=None, x=None, y=None, color=None, symbol=None, size=None, hover_name=None, hover_data=None, custom_data=None, text=None, facet_row=None, facet_col=None, facet_col_wrap=0, facet_row_spacing=None, facet_col_spacing=None, error_x=None, error_x_minus=None, error_y=None, error_y=None, error_y=None, error_y=None, category_orders={}, labels={}, orientation=None, color_discrete_sequence=None, color_discrete_map={}, color_continuous_scale=None, range_color=None, color_continuous_midpoint=None, symbol_sequence=None, symbol_map={}, opacity=None, size_max=None, marginal_x=None, marginal_y=None, trendline=None, trendline_color_override=None, log_x=False, log_y=False, range_x=None, range_y=None, render_mode='auto', title=None, template=None, width=None, height=None)



### **Correlation plot**

A correlation plot is a way to visualize correlations between variables.

The Pearson Correlation Coefficient summarizes this relationship

- Has a value -1 to 1
- 1 is totally positively correlated
  - As x increases, y increases
- 0 is not at all correlated
  - No relationship between x and y
- -1 is totally negatively correlated
  - As x increases, y decreases



### Correlation plot setup

df contains data on bike sharing rental numbers in Korea with various weather variables

pandas provides a method to create the data needed:

```
cr = df.corr(method='pearson')
print(cr)
```

#### Our Pearson correlation table:

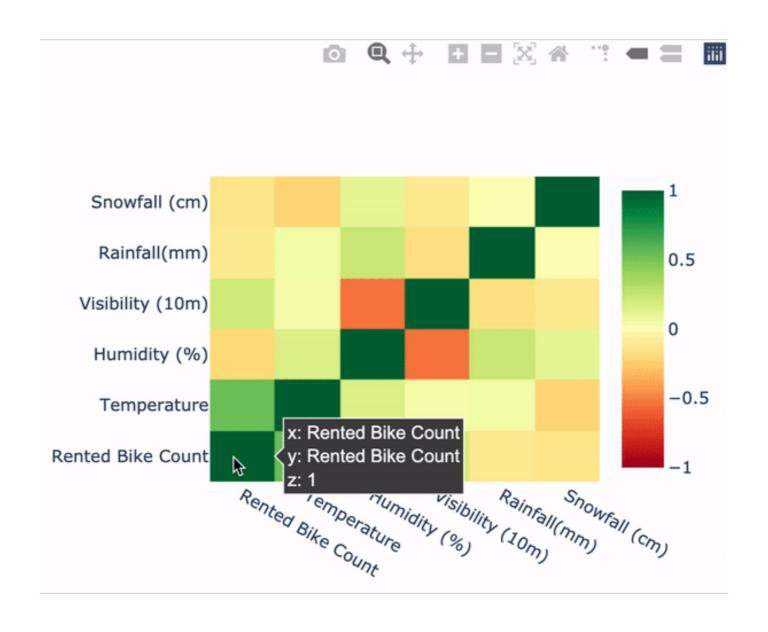
	Rented Bike Count	Temperature	Humidity (%)	Visibility (10m)	Rainfall(mm)	Snowfall (cm)
Rented Bike Count	1.000000	0.538558	-0.199780	0.199280	-0.123074	-0.141804
Temperature	0.538558	1.000000	0.159371	0.034794	0.050282	-0.218405
Humidity (%)	-0.199780	0.159371	1.000000	-0.543090	0.236397	0.108183
Visibility (10m)	0.199280	0.034794	-0.543090	1.000000	-0.167629	-0.121695
Rainfall(mm)	-0.123074	0.050282	0.236397	-0.167629	1.000000	0.008500
Snowfall (cm)	-0.141804	-0.218405	0.108183	-0.121695	0.008500	1.000000

### **Correlation plot with Plotly**

Let's build a correlation plot:

### Our correlation plot

Voila!



### Let's practice!

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# Customizing hover information and legends

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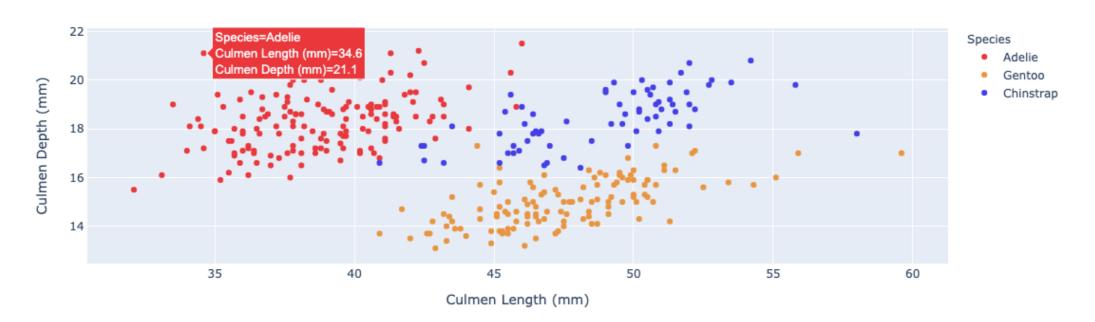


### What do we mean by hover?

**Hover information**: The text and data that appears when your mouse hovers over a data point in a Plotly visualization.

By default, you get some hover information already:



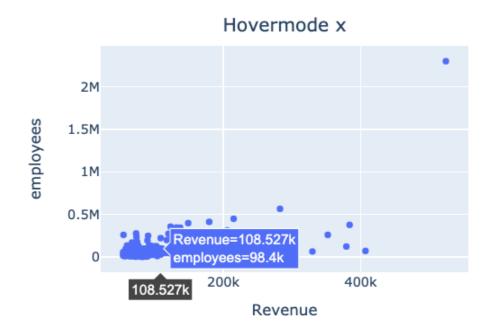


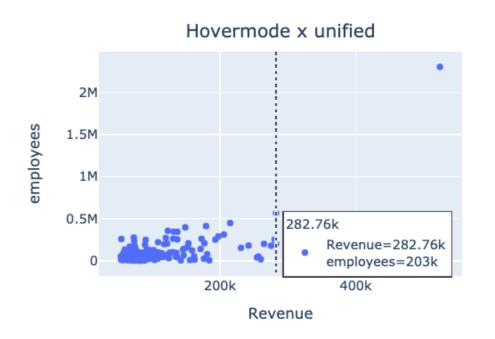


### Other default hover information

The relevant layout argument is hovermode (defaults to closest), which can be set to different values:

- x or y : adds a highlight on the x or y axis
- x unified / y unified : A dotted line appears on the relevant axis (x here) and the hover-box is formatted







### Hover information using plotly.express

Customizing hover data in plotly.express:

- hover\_name = A specified column that will appear in bold at the top of the hover box
- hover\_data = A list of columns to include or a dictionary to include/exclude columns
  - o {column\_name: False} (this will exclude column\_name)

No extensive formatting options

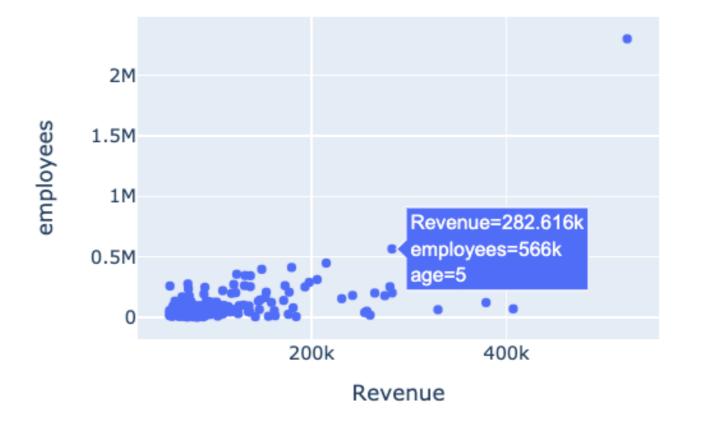


### Variables in hover information

Hover columns don't need to be in the plot!

 E.g.: Revenue vs. company size with age of company displayed on hover

```
We can see age in the hover!
```



### Styling hover information

There are two main ways to style hover information:

- 1. Using the hoverlabel layout element
  - A dictionary of stylistic properties (background colors, borders, font, sizings, etc.)
- 2. Using the hovertemplate layout element
  - An HTML-like string to style the text (beyond this course)

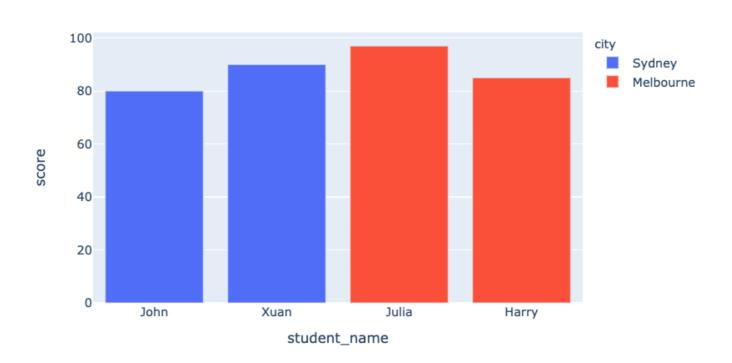


### What is a legend?

A legend is an information box that provides a key to the elements inside the plot, particularly the color or style.

- Legends often automatically appear with plotly.
  - For example, when adding colors to our bar chart





### Creating and styling the legend

You can turn on and style the legend using update\_layout()

- showlegend = True shows the default legend
- Legend = a dictionary specifying styles and positioning of the legend
  - $\circ$  x, y: (0-1) the percentage across x or y axis to position
  - Other stylistic elements such as bgcolor (background color), borderwidth, title, and font

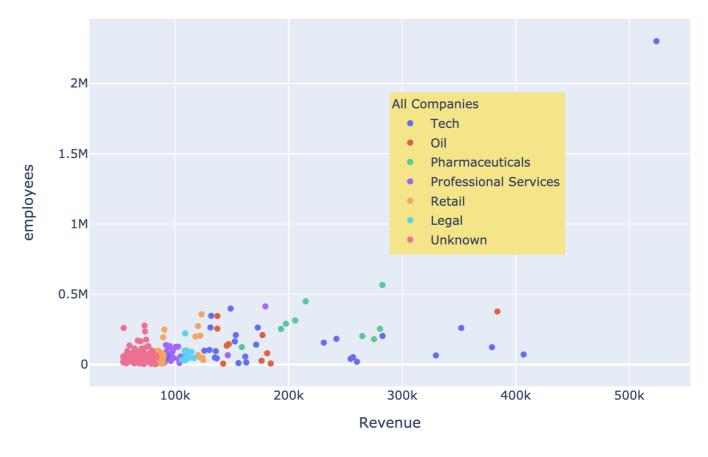
As always - check the documentation (link) for more!

### A styled legend

We can create a styled legend and position it:

```
fig.update_layout({
    'showlegend': True,
    'legend': {
        'title': 'All Companies',
        'x': 0.5, 'y': 0.8,
        'bgcolor': 'rgb(246,228,129)'}
})
```

#### Companies by Revenue Size, Headcount and Industry



### Let's practice!

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### Adding annotations

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#### What are annotations?

Annotations are extra boxes of text and data added to a plot.

Unlike hover information, annotations are always present.

They serve two primary purposes:

- 1. Data-linked annotations (draw attention, add notes) on a particular point
- 2. Add extra notes to a plot,
  - Much like adding a text-box in Microsoft Word

### **Creating annotations**

In Plotly you can add annotations in several ways:

- 1. Using add\_annotation()
  - Adds a single annotation
- 2. Using update\_layout() and the annotations argument
  - A list of annotation objects
  - Useful if adding many annotations

For consistency, we'll stick with update\_layout()



### Important annotation arguments

There are several key elements of an annotation (dictionary) worth highlighting:

- showarrow = True / False
  - Determines whether an arrow will be drawn from the box to the given x / y coordinates
  - You can style the arrow as well!
- text = The actual text to be displayed
  - You can insert variables into this text too
- x and y: coordinates at which to place the annotation

Be careful placing annotations absolutely - if your data changes, things may overlap!



### Positioning annotations

By default, the x and y arguments will be in the units of the plot to link to a data point.

However, you can position absolutely by:

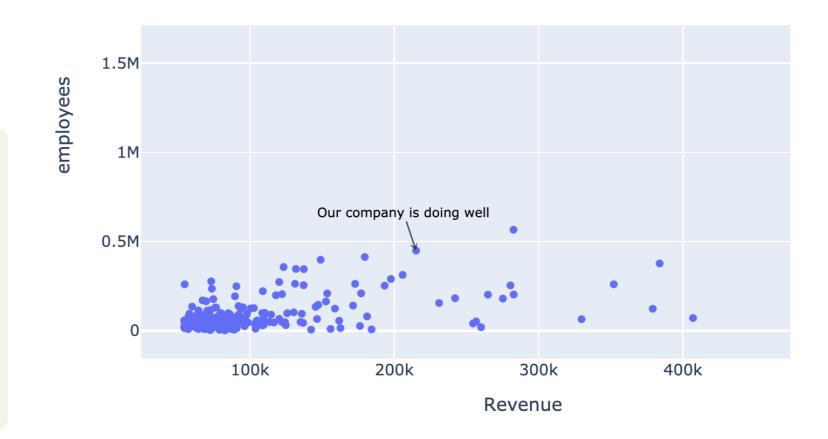
- Setting the arguments xref and yref to paper
  - Now the x and y parameters are 0-1 positions
  - $\circ$  A position of (x=0.5, y=0.5) would be right in the middle of the plot

#### **Data-linked annotations**

Let's annotate **our** company (we know the revenue and employee count) on our previous scatterplot.

```
my_annotation = {
    'x': 215111, 'y': 449000,
    'showarrow': True, 'arrowhead': 3,
    'text': "Our company is doing well",
    'font': {'size': 10, 'color': 'black'}}
fig.update_layout({'annotations': [my_annotation]})
fig.show()
```

Nice! We can see our company clearly:

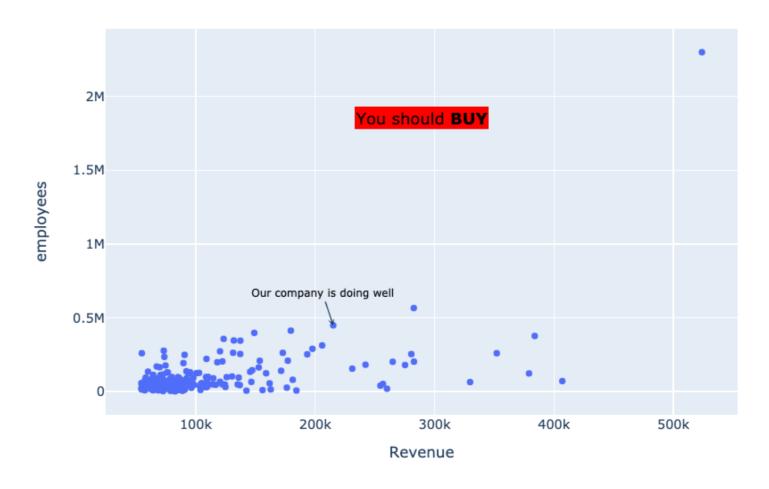


### Floating annotation

We can also have a floating annotation, positioned absolutely.

```
float_annotation = {
    'xref': 'paper', 'yref': 'paper',
    'x': 0.5, 'y': 0.8,
    'showarrow': False,
    'text': "You should <b>BUY</b>",
    'font' : {'size': 15,'color': 'black'},
    'bgcolor': 'rgb(255,0,0)'}
```

We get a strong message!



### Let's practice!

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### Editing plot axes

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### **Our dataset**

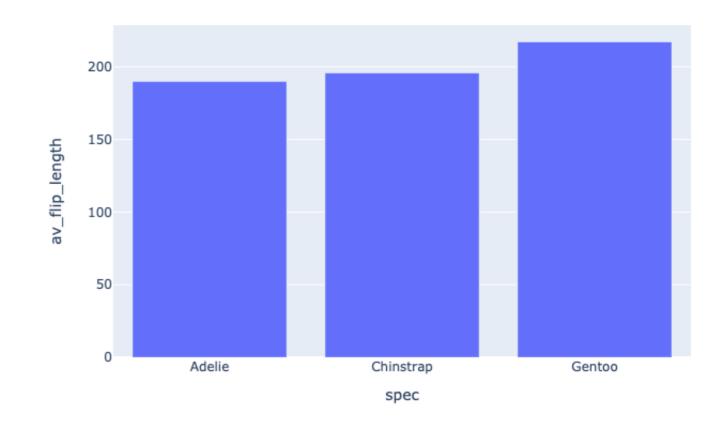
Using the penguins dataset, let's aggregate flipper size by species:

spec	av_flip_length
Adelie	189.953642
Chinstrap	195.823529
Gentoo	217.186992

Those columns aren't labeled well for presentation!

### The default axis titles

Let's create a simple bar chart:



This works, but those axes titles aren't great.

### Editing axis titles

plotly often has 'shortcut' functions:

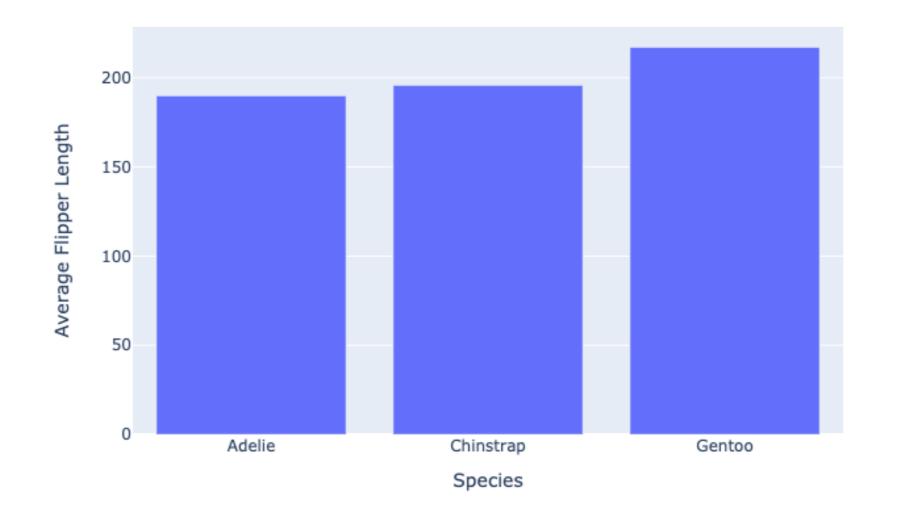
```
fig.update_xaxes(title_text='Species')
fig.update_yaxes(title_text='Average Flipper Length')
```

Or with the more general update\_layout()

We will stick with update\_layout() for consistency

### Cleaning up our plot

Both methods will produce a more presentation-worthy chart.



#### Which method to use?

The shortcut method is helpful to quickly change just that one attribute.

To further style axes, the update\_layout() method allows you to edit:

- Font family, font size
- Text angle
- Text color
- Much more!

See more on the Plotly documentation



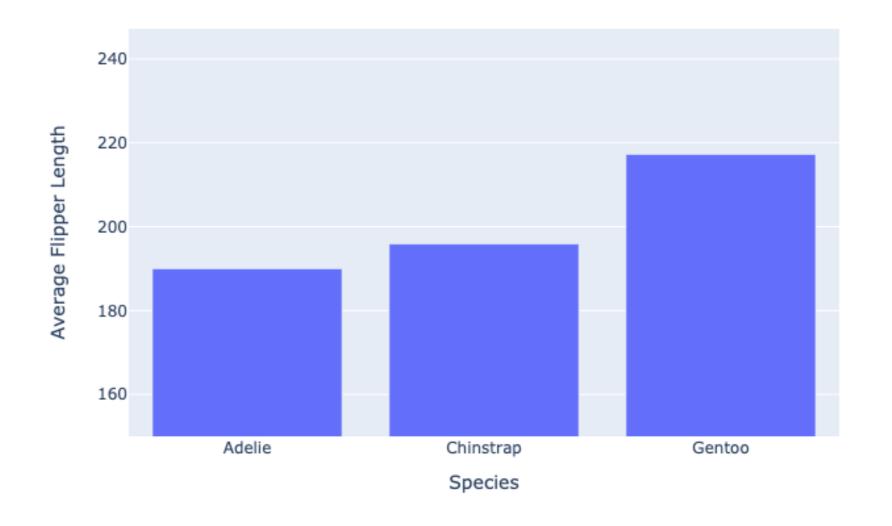
### Editing axes ranges

Plotly automatically calculates axes ranges from your data - this may not be desired!

Let's set the y-axis to start at 150 and go up to a small buffer (30) past the maximum flipper length

### Our new axes ranges

We get specific axes:





### Data scale issues

What happens when some data points are much larger than others?

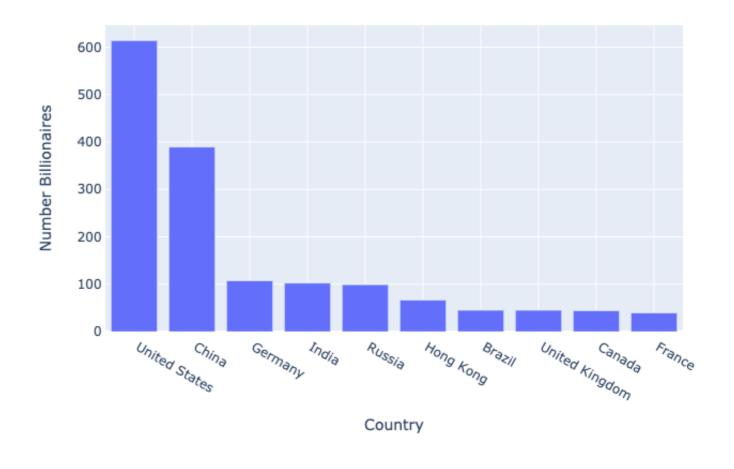
Top 10 countries by number of billionaires

Country	Number	Billionaires
United States		614
China		389
Germany		107
India		102
Russia		99
Hong Kong		66
Brazil		45
United Kingdom		45
Canada		44
France		39



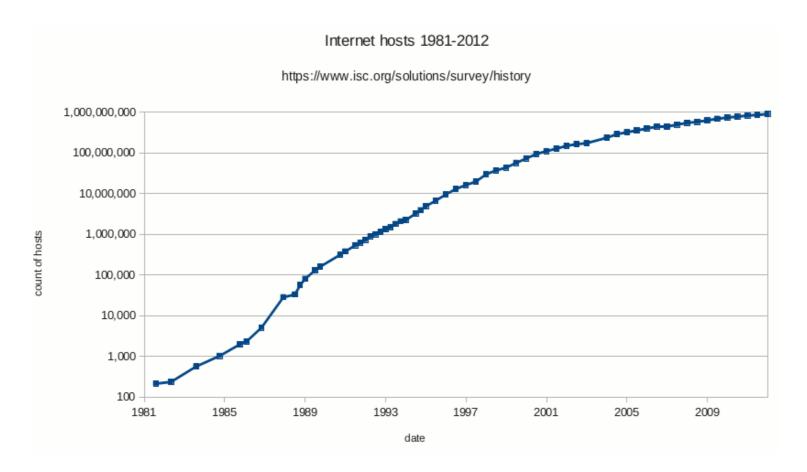
### Our scale problem

Let's plot without any adjustment:



### The log scale

- Common scale used to plot data with large value differences.
- It looks like this:

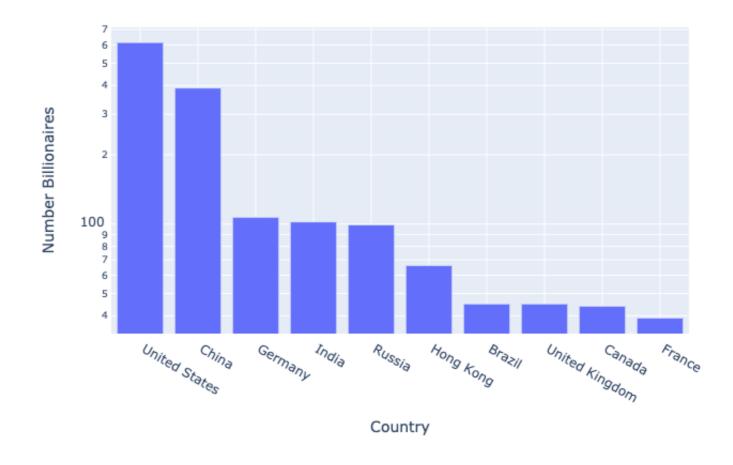


Ticks on our y-axis aren't uniform (10,20, 30, etc.)

Each tick is an *order of magnitude* bigger (10, 100, 1000, etc.)

### Using log with our data

Plotly has log\_y and log\_x arguments



That's better!

### Log scale: a word of warning

When visualizing data, you are telling a *story*.

If your audience doesn't know what a log scale is, there may be miscommunication.

• So remember to keep your audience in mind!



### Let's practice!

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