# Data Visualization: Subplots and Combining Visualizations

\$ echo "Data Science Institute"

#### Overview of this slide deck, we will:

- Learn about subplot notation in matplotlib
- Put multiple visualizations on the same axes objects
- Show errors
- Adjust figure layout
- Add images to plots

# **Subplots**

#### Recall: How does matplotlib work?

- A **figure** is like a container that holds a set of **axes**
- The axes is our actual plot or graph
- A figure can hold multiple axes (like subplots)
- Every visual element of our plots colour, legends, axis titles and scales, text is called an **artist** and belongs to an axes (not to a figure)

## Setting up

• Let's start by loading our libraries and generating some new sample data

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import scipy
import PIL
import requests
np.random.seed(613)
x1 = np.arange(50)
y1 = np.random.randint(0, 75,50)
x2 = np.array(["Luffy", "Zoro", "Nami", "Usopp", "Sanji"])
y2 = np.array([110, 180, 240, 99, 220])
```

## Introducing subplots

• In past lessons, to define our figure and its one axes, we would type:

```
fig, ax = plt.subplots(figsize=(5, 3))
```

 Now we want to have two plots next to each other, so we just have to define multiple axes and their relative positions:

## Introducing subplots (cont.)

• Then we can use our new figure, with its two axes, and define the types of viz that we want to see in each

## **Activity: Customizing our plots**

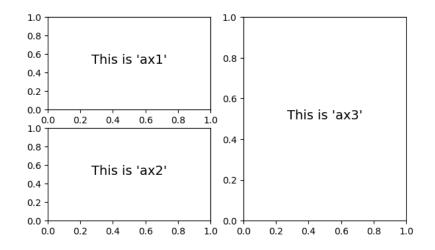
- Refer to past slides and class activities to customize the subplots we just made
- Think about adding titles or annotations, or modifying colour, marker type, and fonts – differently for each subplot
- Share your resulting images in the chat!



## Subplots without a grid arrangement?

- We can also arrange the subplots within our figure by using <a href="mailto:plt.subplot\_mosaic">plt.subplot\_mosaic</a>()
- Each axes in our subplot\_mosaic() function has a 'label'

## Subplots without a grid arrangement?

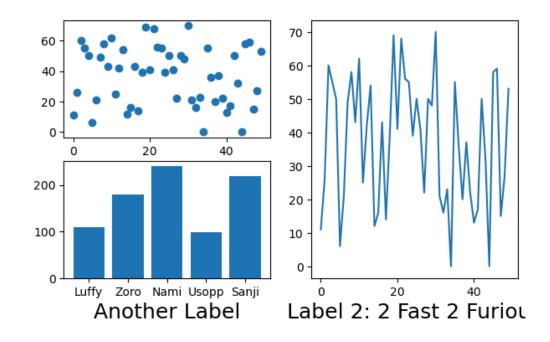


## Subplots without a grid arrangement - Data

• Once we've made our mosaic, we can add data to each of our subplots the same way we did before! Just reference each axes label in our someaxes list:

## Modify figure layout

• Let's try adding some very large x axis titles to our previous plot



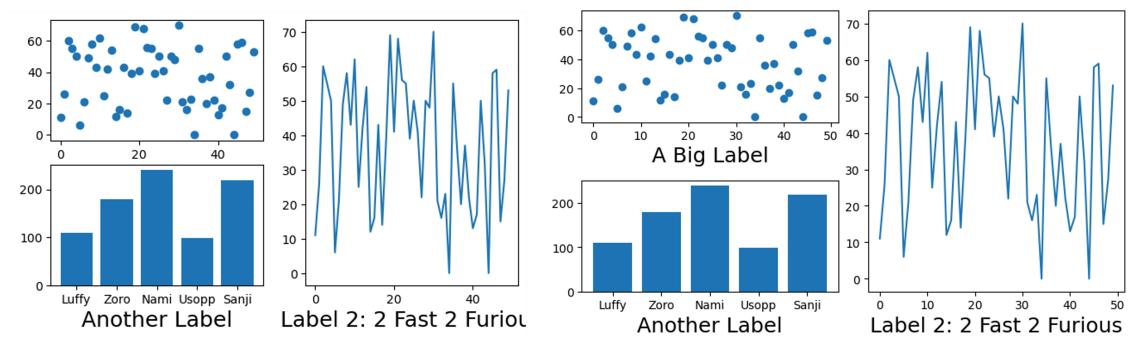
 The axis title for our scatter plot doesn't appear, and the axis title for our line plot is cut off!

- We can use layouts to make sure that our subplots fit neatly in our figure area
- There are two main kinds of layout we consider:
  - Tight layout adjusts subplots so tick labels, axis labels, and titles don't overlap or leave the figure area
  - Constrained layout works similarly except it also fits things like legends or colorbars

• First let's see how a constrained layout changes our plot:

```
fig, someaxes = plt.subplot mosaic([['ax1', 'ax3'],
                                ['ax2', 'ax3']],
                                figsize=(7, 4),
                                layout = "constrained")
someaxes["ax1"].scatter(x1,y1)
someaxes["ax2"].bar(x2,y2)
someaxes["ax3"].plot(x1,y1)
someaxes["ax1"].set xlabel('A Big Label', fontsize=18)
someaxes["ax2"].set xlabel('Another Label', fontsize=18)
someaxes["ax3"].set xlabel('Label 2: 2 Fast 2 Furious', fontsize=18)
plt.show()
```

## **Layouts Comparison**



No layout specified

**Constrained layout** 

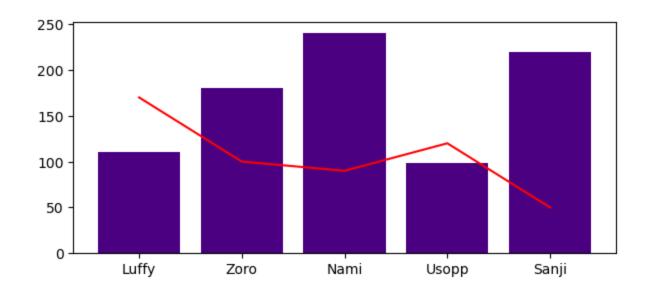
## Multiple viz on one axes

## Multiple viz on one axes object

Super easy: just call multiple plot methods on the same axes

```
# first make our sample data
x = np.array(["Luffy", "Zoro", "Nami", "Usopp", "Sanji"])
y1 = np.array([110, 180, 240, 99, 220])
y2 = np.array([170, 100, 90, 120, 50])
# define our figure and axes (just one this time)
fig, ax = plt.subplots(figsize=(7, 3))
# now call both bar and plot elements to the same axes (ax)
ax.bar(x, y1,
color = "indigo")
ax.plot(x, y2,
color = "red")
```

## Multiple viz on one axes object



Annotations, shapes, etc. can be added as usual!

#### Add error information

• First, calculate standard deviation of our data

```
y2\_sd = np.std(y2)
```

• Then plot our line as before

```
fig, ax = plt.subplots(figsize=(7, 3))
ax.plot(x, y2, color = "red")
```

#### Add error information

- Then use **errorbar()** to add in our error line (standard deviation in this case, but could be whatever value you calculated)
  - yerr specifies that we're plotting vertical error bars
  - fmt makes sure we're not plotting the actual data points, only the error

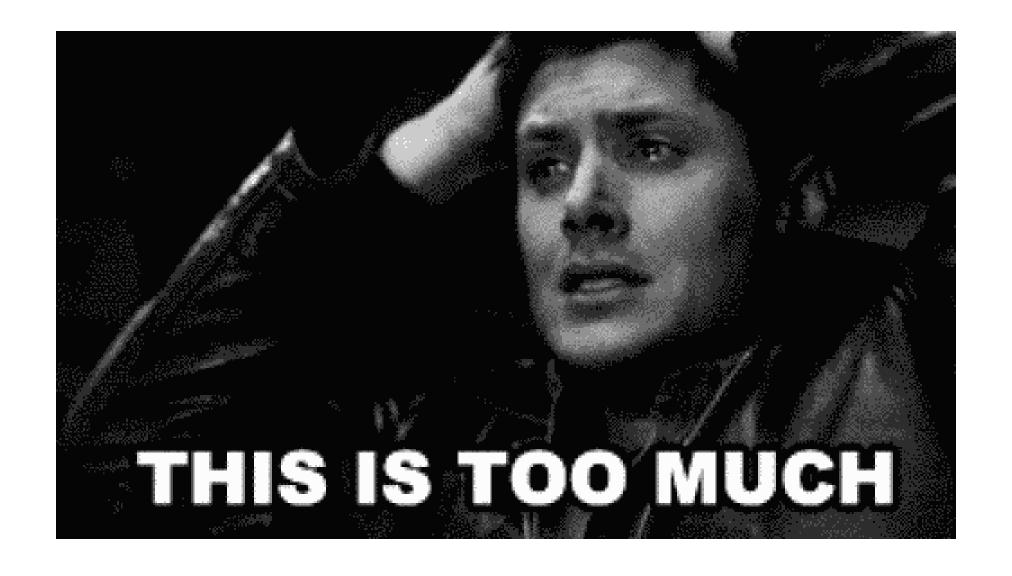
```
ax.errorbar(x, #our x values
    y2, #our y values
    yerr = y2_sd,
    fmt = "none")
```

#### Customizing errorbar appearance

#### **Errorevery**

• If we don't want to see error bars for every single point, we can specify intervals using **errorevery** 

# Add images to plots





## Using images from the internet

• First let's load our libraries

```
from PIL import Image # to open images
import requests # to get images from URLs
from io import BytesIO # to store images
```

Then get our image from the internet

```
response = requests.get('https://upload.wikimedia.org/wikipedia/en/c/cb/Monkey_D_Luffy.png')
image_file = BytesIO(response.content)
image = Image.open(image_file)
```

## Adding an image to a plot

Now make a basic line plot (reusing our data)

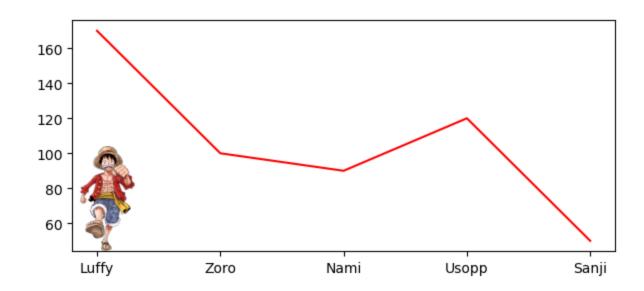
```
fig, ax = plt.subplots(figsize=(7, 3))
ax.plot(x, y2, color = "red")
```

 Then overlay a new axis ('ax\_image') on our figure (on top of 'ax') to act as a container for our image

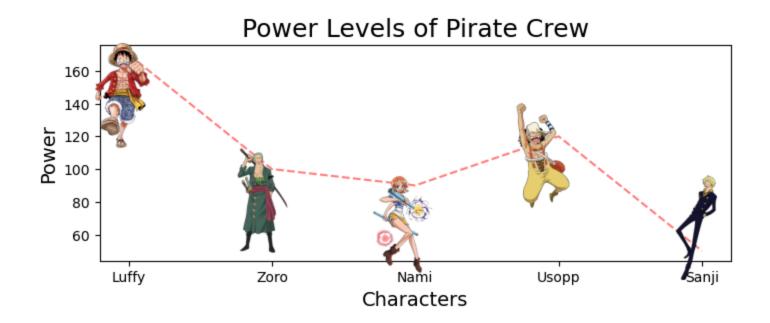
## Adding an image to a plot

• Then add imshow() to add the image we prepared before

# Adding images to our plots



#### What can we make?



## Saving our visualizations

• First, let's define where we want to save our visualization

```
path = 'C:/Users/...'
# can be full path or relative path
filename = '/fig1a.png'
```

Then save the visualization

```
plt.savefig(path+filename, dpi=300)
# note that path shouldn't end with / since filename starts with it
```



# **Assignment 3**

## Feedback!

#### **Next session**

- Data Visualization as Advocacy
- Examples of data visualization used for advocacy, and best practices
- Form, representation, and giving credit as means of incorporating advocacy into our own data visualization practices
- Moving beyond matplotlib other data viz libraries in Python