



# Intro to Visualization on HPC

2024



# Using HPC

Need some introductory slides about visualization

# Using RStudio on HPC

ARIZONA Research Technologies HPC Systems

Apps ▾ Files ▾ Jobs ▾ Clusters ▾ Interactive Apps ▾ My Interactive Sessions

Please NOTE: "windfall" jobs will be re-allocated if pre-empted by a "standard" job

**OPEN**  
**OnDemand**

OnDemand provides an integrated, easy-to-use interface for your HPC resources.

**Pinned Apps** A featured subset of apps

- Desktops
  - Interactive Desktop
- GUIs
  - ABAQUS GUI
  - ANSYS Workbench GUI
  - MATLAB GUI
  - Mathematica GUI
  - VSCoDe GUI
- Servers
  - Jupyter Notebook
  - RStudio Server

**System Installed App**

**System Installed App**

**System Installed App**

**System Installed App**

# Using RStudio on HPC

The screenshot displays the RStudio interface with the following components:

- Console:** Shows the R version (4.0.0), copyright information, and a workspace loaded from `~/RData`. The prompt is `>`.
- Environment:** Displays the Global Environment with a data object `mydata` containing 15 observations of 2 variables. The values are as follows:

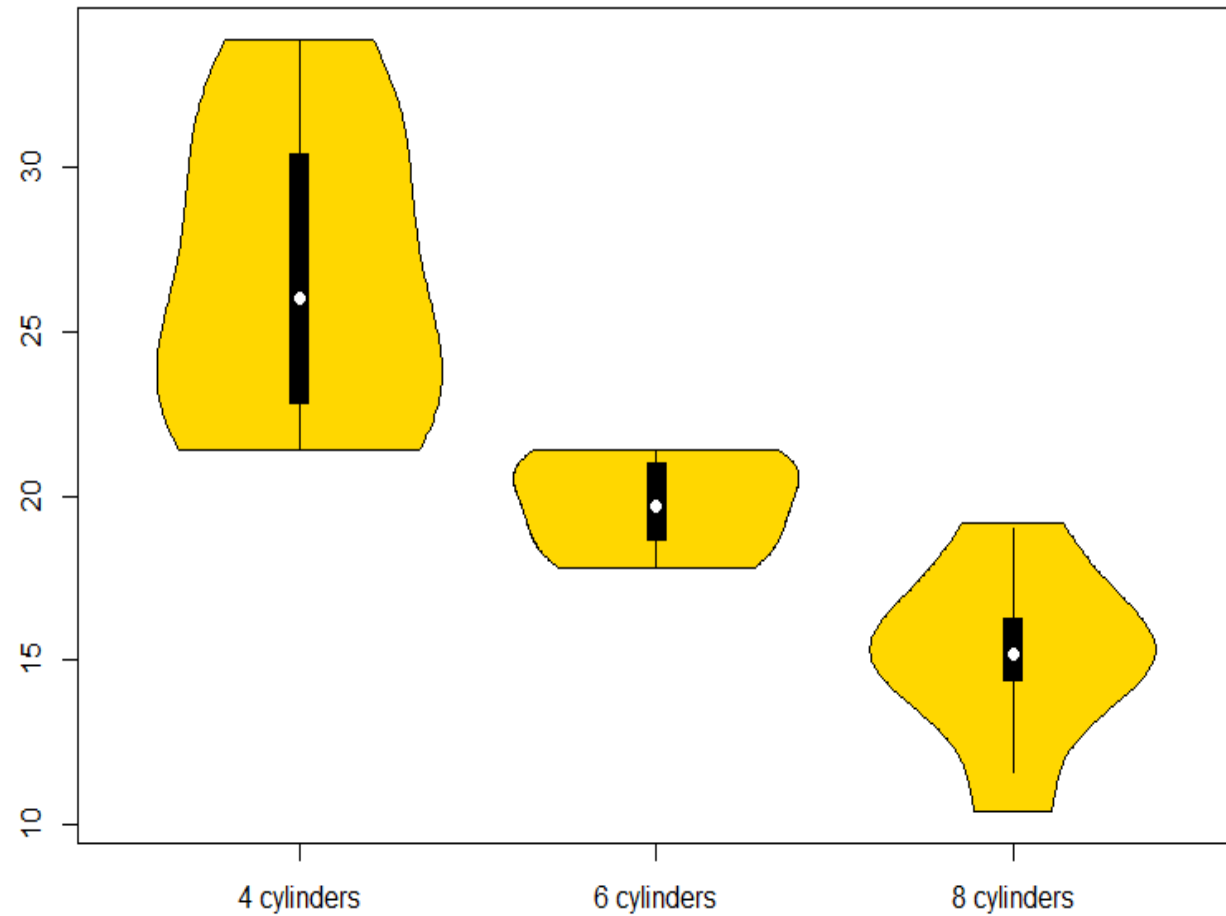
Variable	Values
a	chr [1:5] "aa" "bb" "cc" "dd" "ee"
A	num [1:8] 32311 32624 32908 33219 33499 ...
b	num [1:4] 1 2 4 8
B	num [1:7] 313 284 311 280 322 324 302
C	num [1:10] 27 26 30 34 32 25 31 25 27 32
commute	num [1:10] 27 26 30 34 32 25 31 25 27 32

- Files:** Shows a file explorer with the following files and their sizes and modification dates:

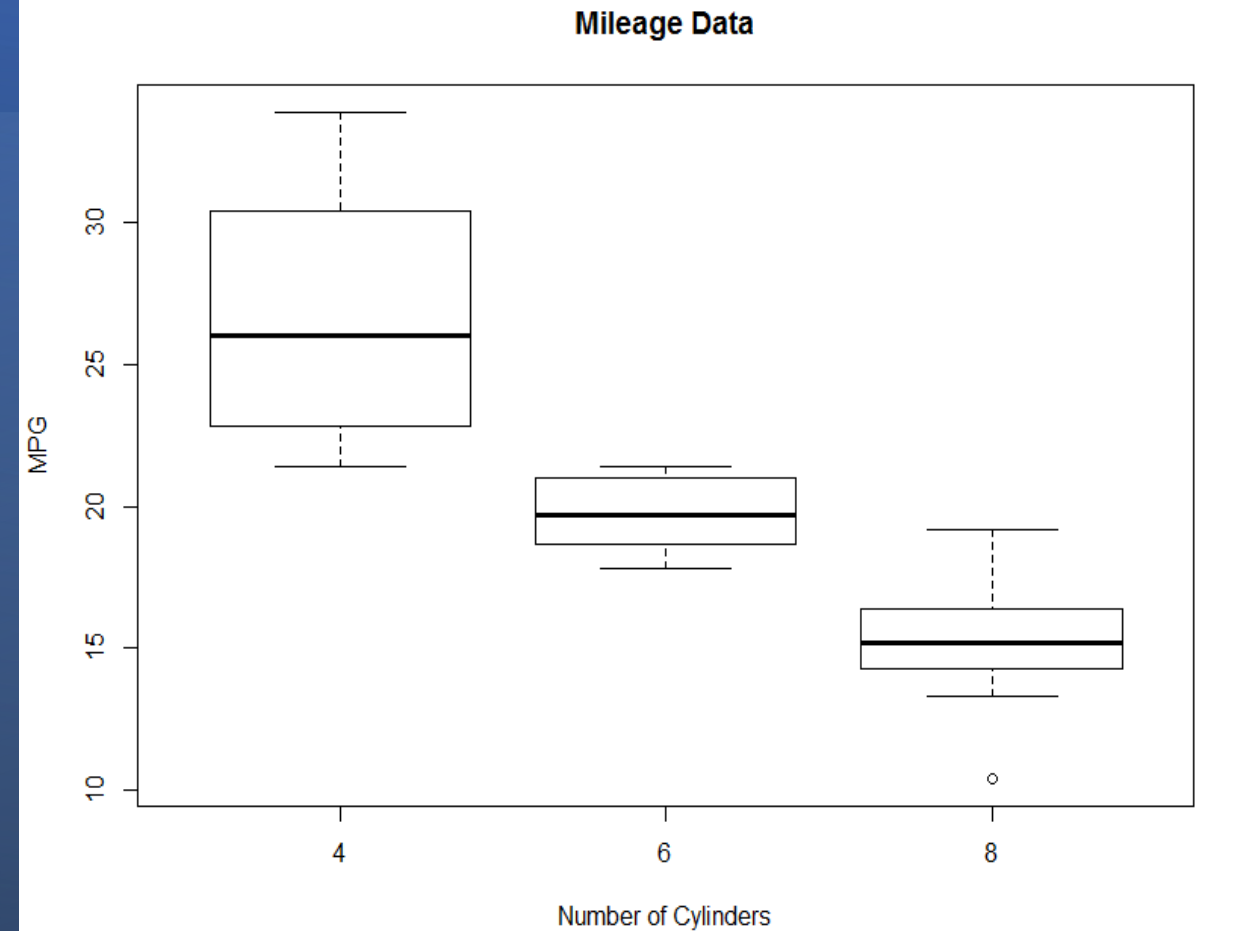
Name	Size	Modified
.RData	941 B	Jul 20, 2019, 8:03 AM
.Rhistory	2.4 KB	Aug 6, 2021, 8:53 AM
__pycache__		
activate.ini	3.3 KB	Mar 21, 2011, 12:05 AM
alpine_latest.sif	2.1 MB	Feb 19, 2022, 4:50 PM
autamus-notes	3.7 KB	Sep 21, 2021, 9:52 AM
conda-bash.sh	3.5 KB	Oct 23, 2019, 4:07 PM

# Using RStudio on HPC

## Violin Plot



## Box Plot

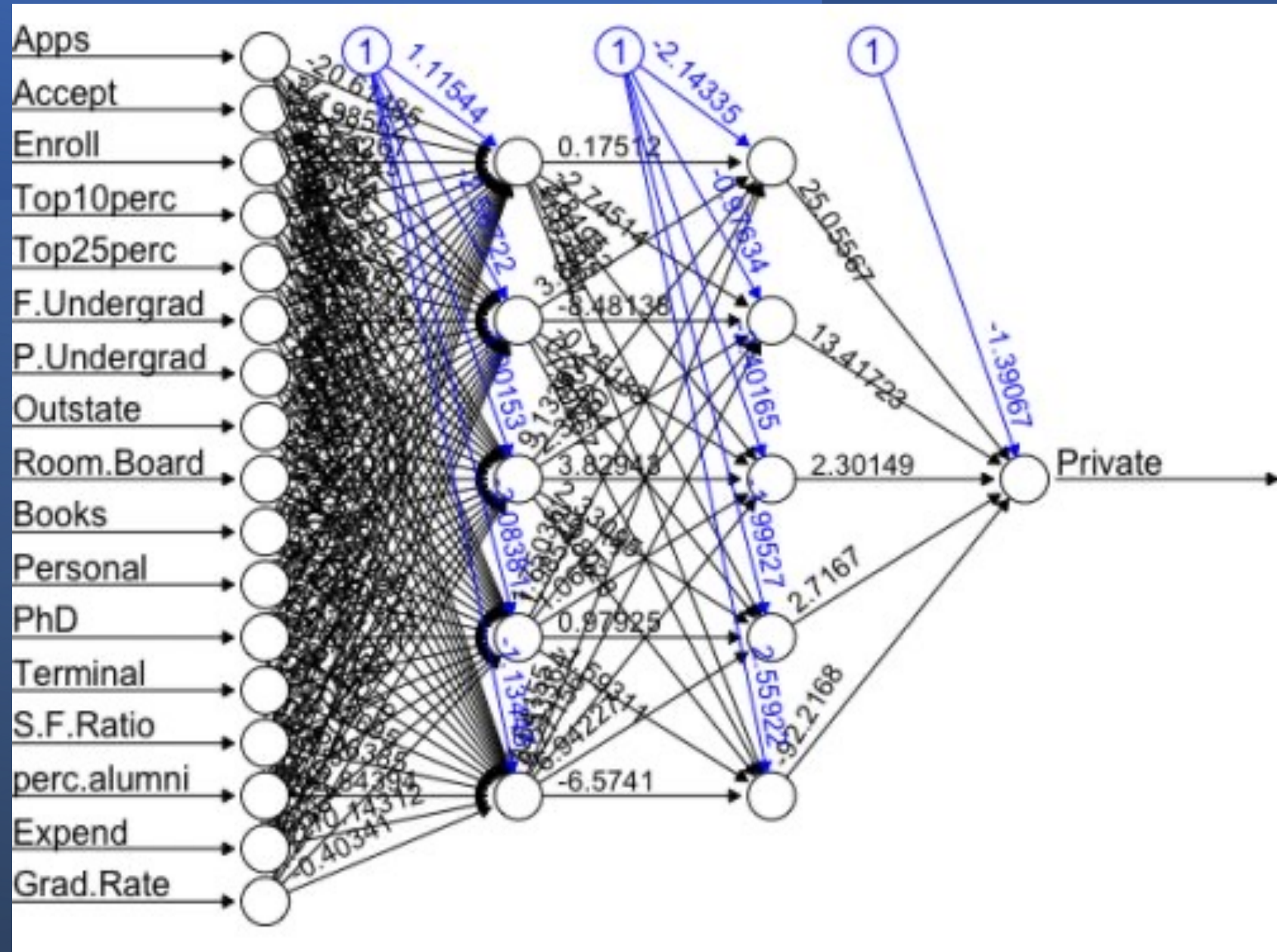




# Using RStudio on HPC

## Neural Network

We can visualize a Neural Network by using the `plot(nn)` command. The black lines represent the weighted vectors between the neurons. The blue line represents the bias added.



# Using Jupyter Notebooks on HPC

Interactive Apps

Desktops

Interactive Desktop

GUIs

ABAQUS GUI

ANSYS Workbench GUI

MATLAB GUI

Mathematica GUI

Servers

**Jupyter Notebook**

RStudio Server

## Jupyter Notebook

This app will launch a [Jupyter](#) server using [Python](#) on a [UAz cluster](#).

### Cluster

Ocelote Cluster

### Run Time

1

Enter maximum number of wall clock hours the job is allowed to run.

### Core count on a single node

1

Enter the number of cores on a single node that the job is allowed to use.

### Memory per core

6

Enter the number of Gigabytes of RAM needed per core.

### Special Options


Enter node specific requirements, if any.

### PI Group

chrisreidy

Enter an HPC PI group to be charged for time used.

# Using Jupyter Notebooks on HPC

jupyter ML-HPC Last Checkpoint: 2 hours ago (autosaved)  Logout

File Edit View Insert Cell Kernel Widgets Help Trusted Python 3

Run

```
In [1]: import pandas as pd
```

```
In [2]: import numpy as np
```

```
In [3]: import matplotlib.pyplot as plt
```

```
In [4]: from sklearn.linear_model import LinearRegression
```

```
In [5]: from sklearn.model_selection import train_test_split
```

```
In [6]: # Load data and view the first 5 rows
data = pd.read_excel("king_county_house_data.xls")
```

```
In [7]: data.head(5)
```

Out[7]:

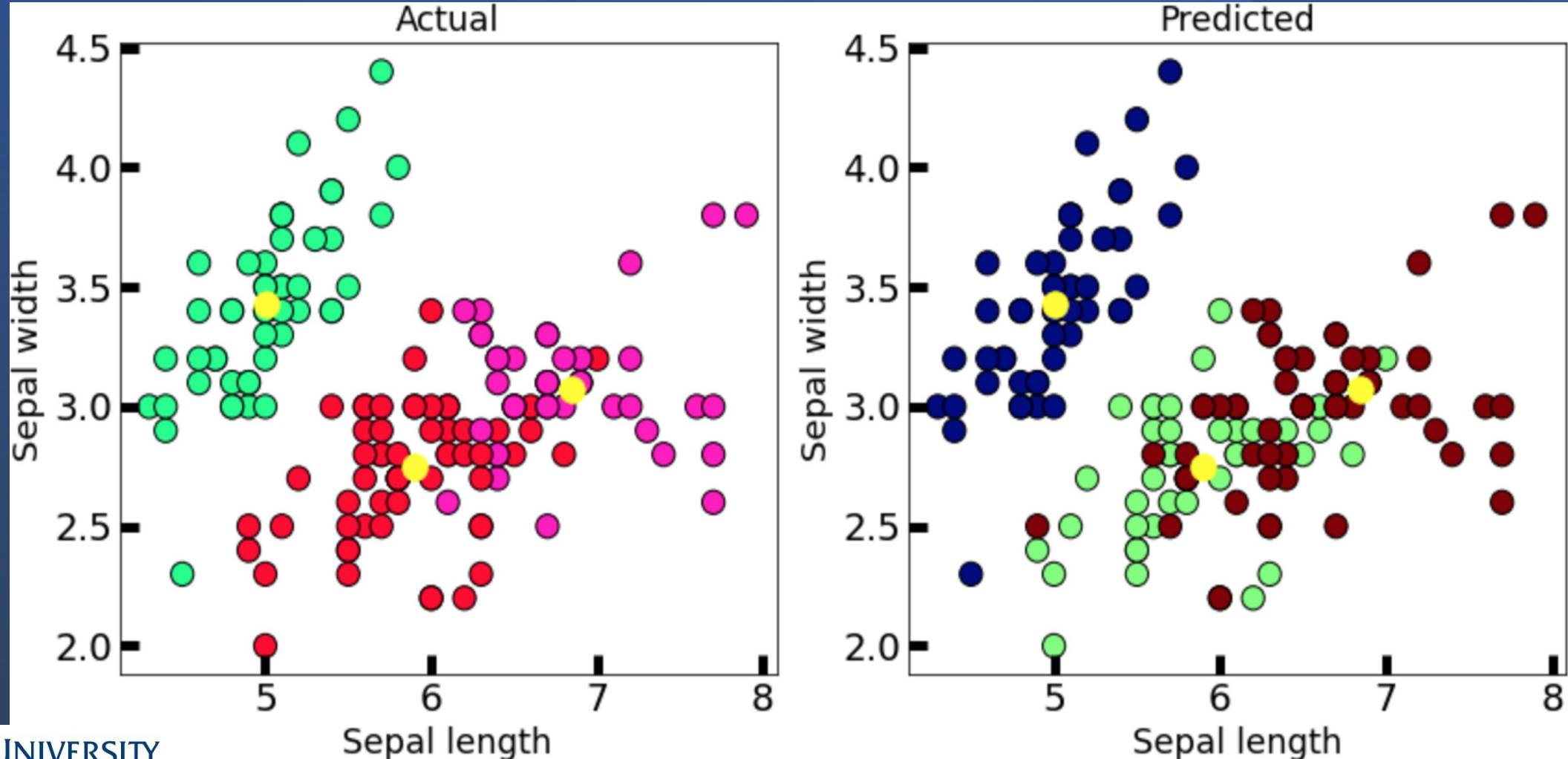
	id	date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	...	grade	sqft
0	7129300520	20141013T000000	221900	3	1.00	1180	5650	1.0	0	0	...	7	
1	6414100192	20141209T000000	538000	3	2.25	2570	7242	2.0	0	0	...	7	
2	5631500400	20150225T000000	180000	2	1.00	770	10000	1.0	0	0	...	6	
3	2487200875	20141209T000000	604000	4	3.00	1960	5000	1.0	0	0	...	7	
4	1954400510	20150218T000000	510000	3	2.00	1680	8080	1.0	0	0	...	8	

5 rows × 21 columns



# Using Jupyter Notebooks on HPC

Visualizing the Iris Database with matplotlib



# Visualization with Python

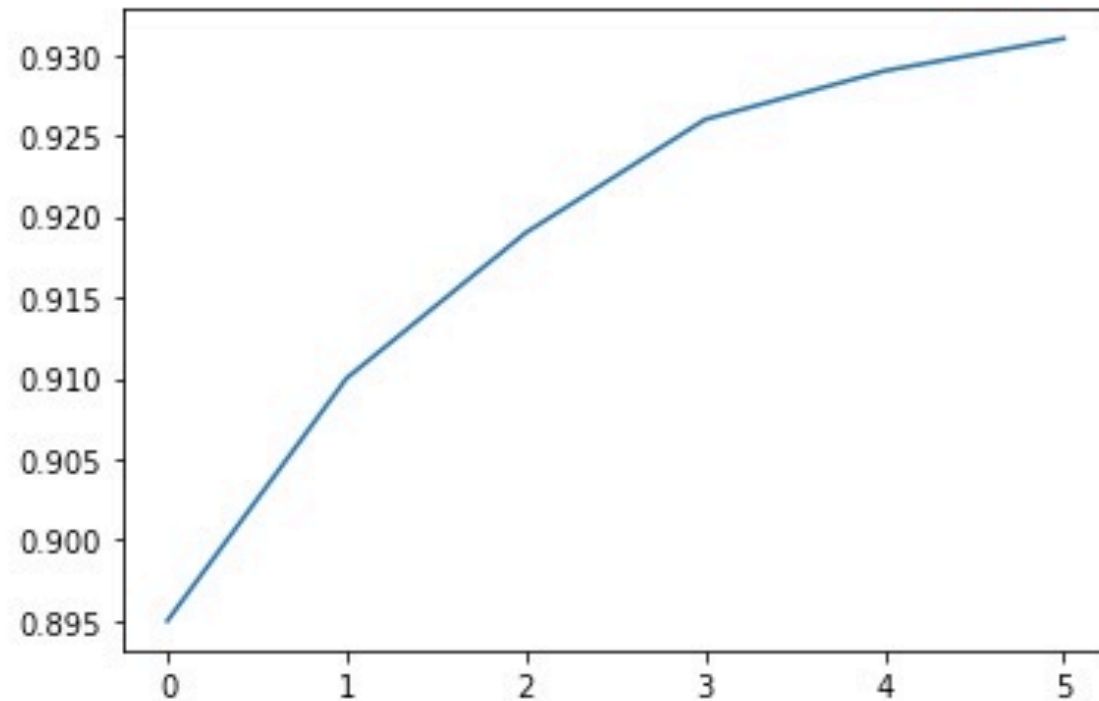
Matplotlib	Seaborn
Used for basic graph plotting like line, bar or pie charts	Mainly used for statistics and performs complex viz with fewer commands
Mainly works with datasets and arrays	Works with entire datasets
Acts productively with data arrays and frames	More organized and functional
Pairs well with Pandas and Numpy	More inbuilt themes

# Python using matplotlib and seaborn

```
In [1]: import matplotlib.pyplot as plt  
import seaborn as sns
```

```
In [2]: yield_apples = [0.895, 0.91, 0.919, 0.926, 0.929, 0.931]  
plt.plot(yield_apples)
```

```
Out[2]: [<matplotlib.lines.Line2D at 0x7f0f271e6520>]
```



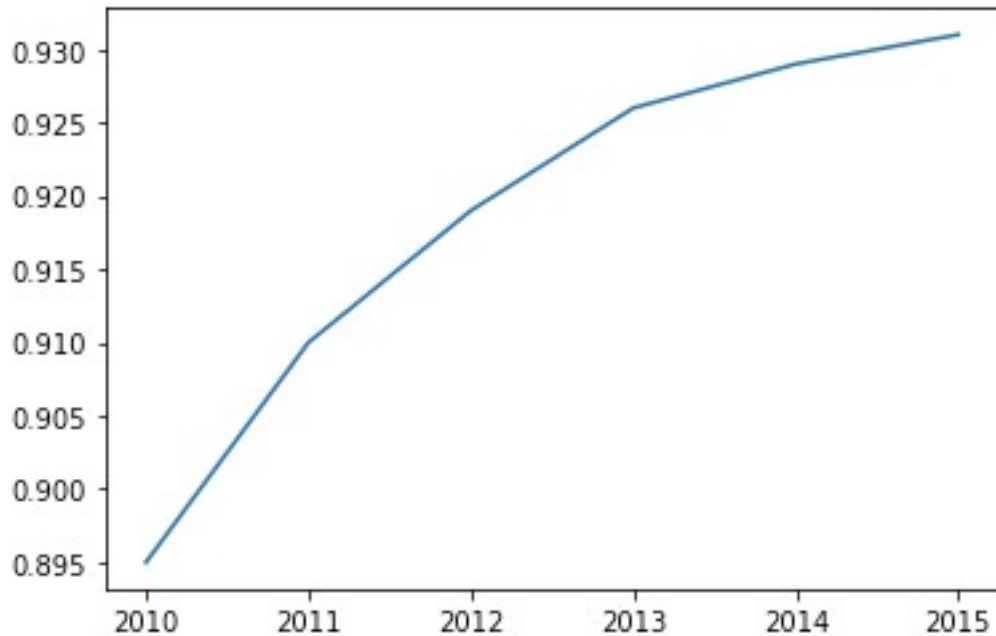
Plotting apple yield



# Python: Line Charts

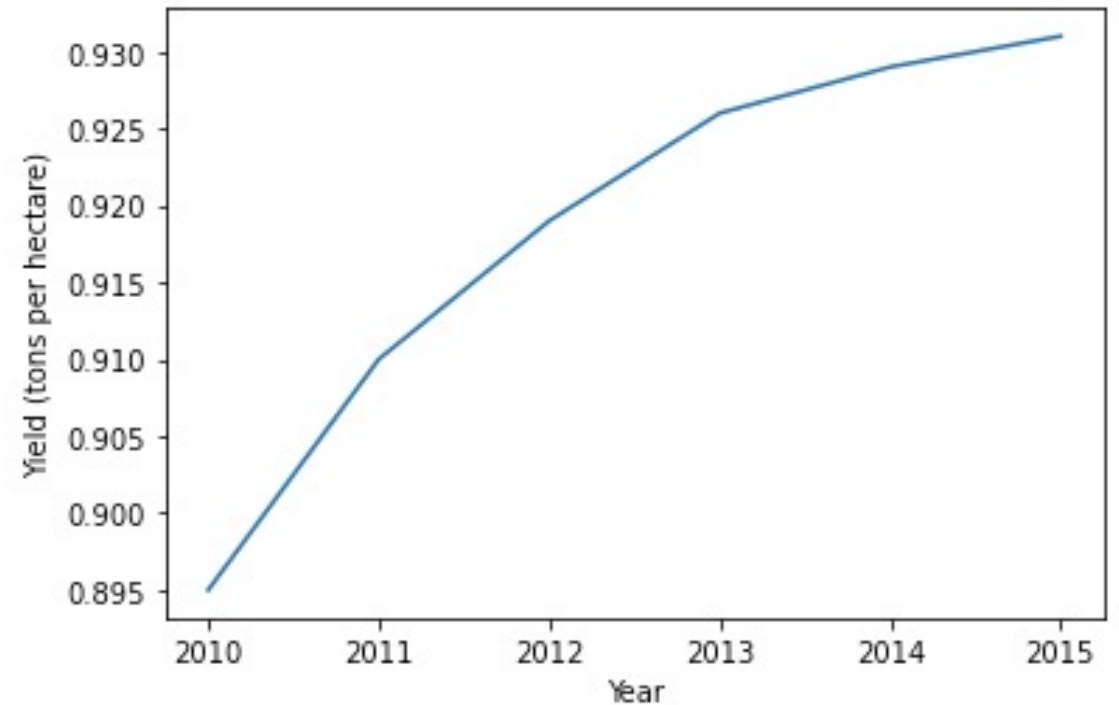
```
years = [2010, 2011, 2012, 2013, 2014, 2015]
yield_apples = [0.895, 0.91, 0.919, 0.926, 0.929, 0.931]
plt.plot(years, yield_apples)
```

[<matplotlib.lines.Line2D at 0x7f0f2506e340>]



Add x axis values

```
plt.plot(years, yield_apples)
plt.xlabel('Year')
plt.ylabel('Yield (tons per hectare)');
```

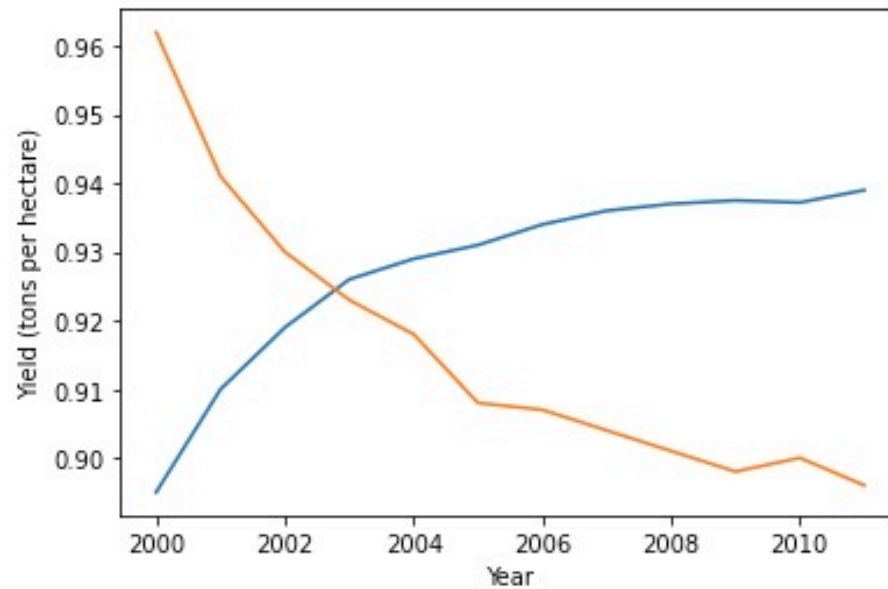


Add labels to the axes

# Python: Line Charts

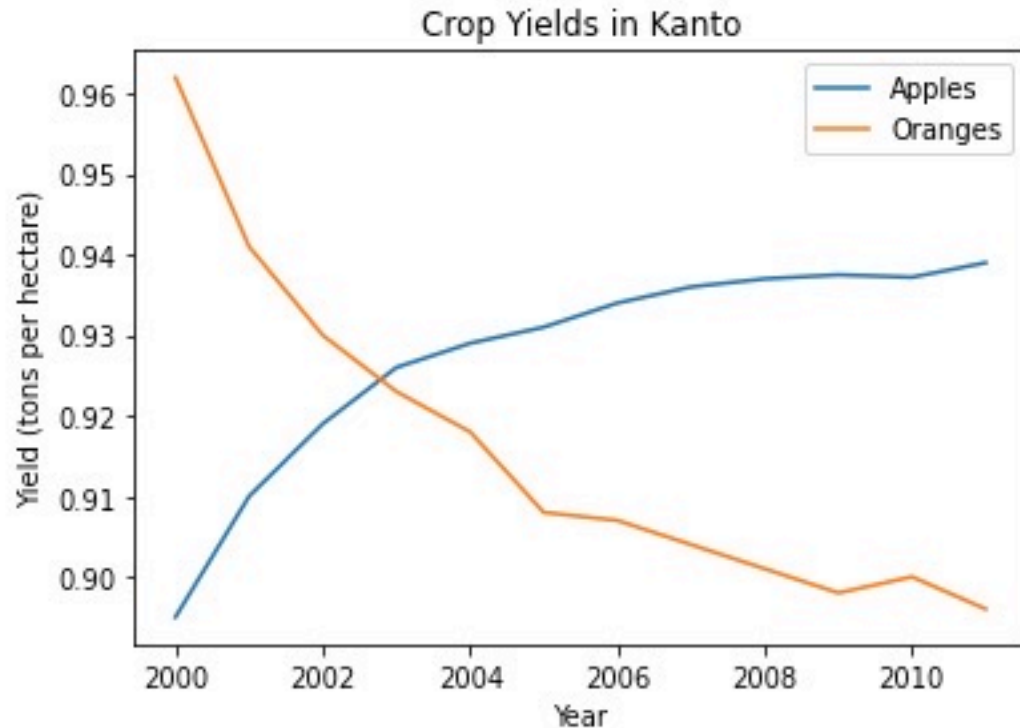
```
In [6]: years = range(2000, 2012)
apples = [0.895, 0.91, 0.919, 0.926, 0.929, 0.931, 0.934, 0.936, 0.937, 0.9375, 0.9372, 0.939]
oranges = [0.962, 0.941, 0.930, 0.923, 0.918, 0.908, 0.907, 0.904, 0.901, 0.898, 0.9, 0.896]
```

```
In [7]: plt.plot(years, apples)
plt.plot(years, oranges)
plt.xlabel('Year')
plt.ylabel('Yield (tons per hectare)');
```

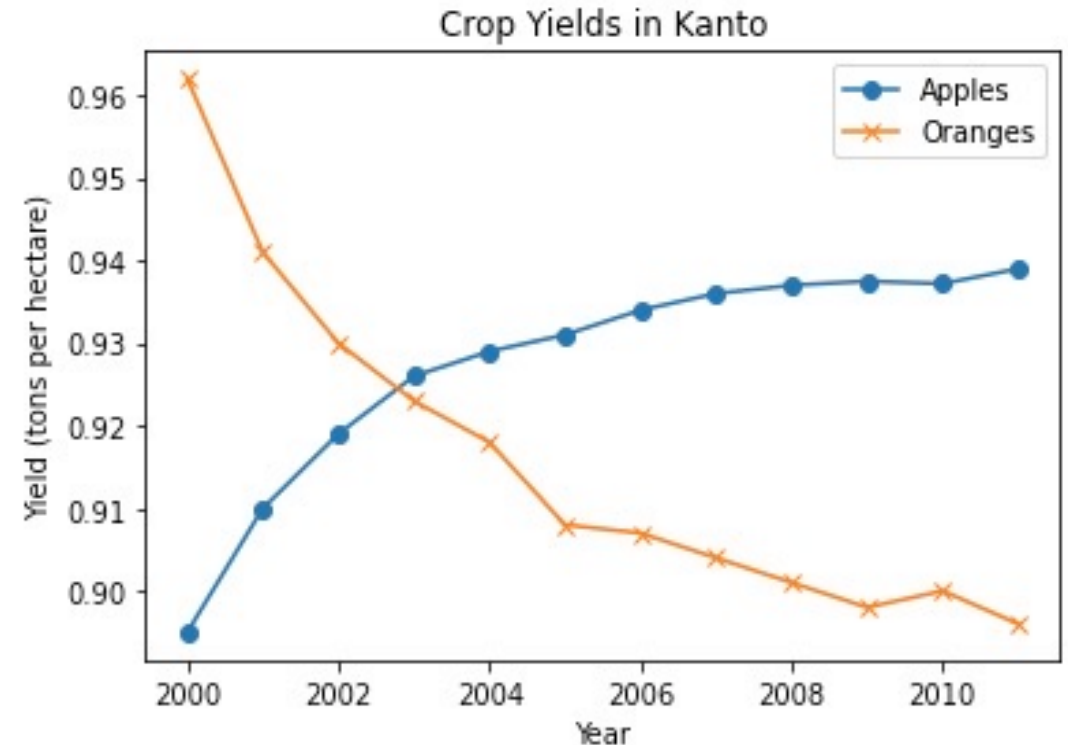


# Python: Line Charts

```
plt.plot(years, apples)
plt.plot(years, oranges)
plt.xlabel('Year')
plt.ylabel('Yield (tons per hectare)')
plt.title("Crop Yields in Kanto")
plt.legend(['Apples', 'Oranges']);
```



```
plt.plot(years, apples, marker='o')
plt.plot(years, oranges, marker='x')
plt.xlabel('Year')
plt.ylabel('Yield (tons per hectare)')
plt.title("Crop Yields in Kanto")
plt.legend(['Apples', 'Oranges']);
```

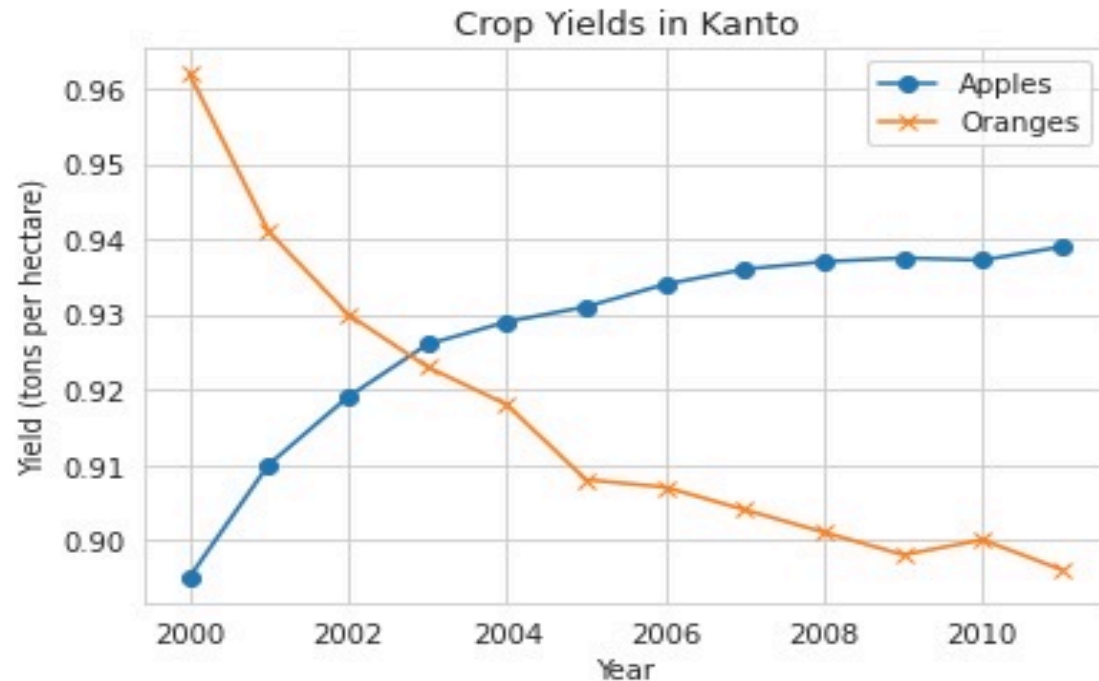




# Python: Line Charts

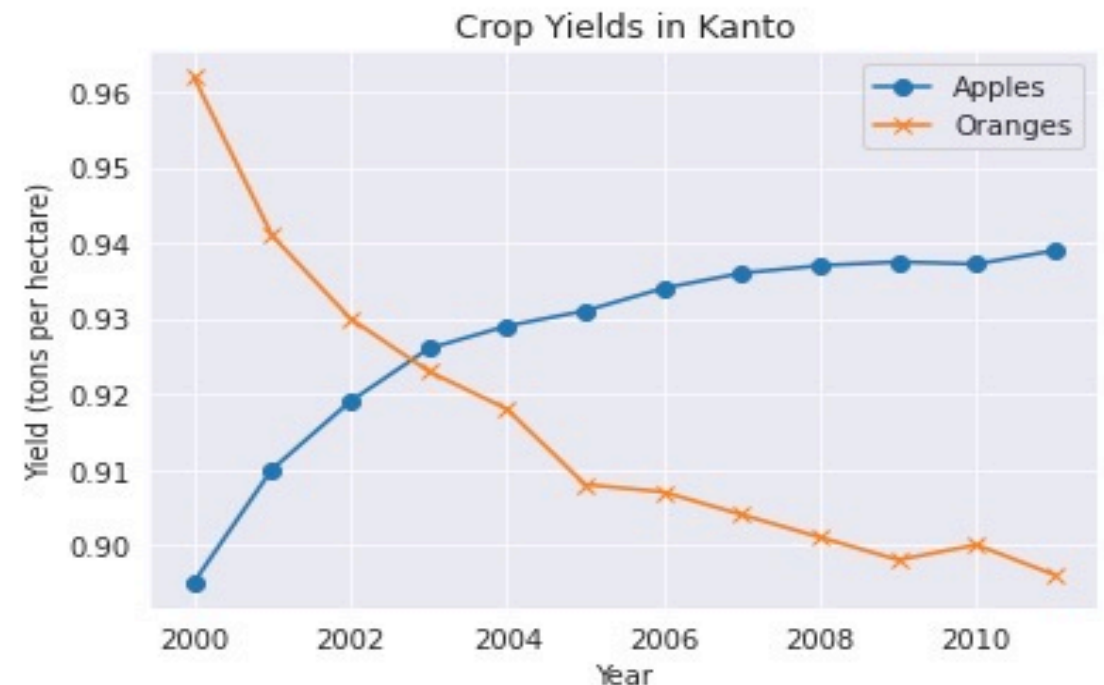
```
sns.set_style("whitegrid")
```

```
plt.plot(years, apples, marker='o')  
plt.plot(years, oranges, marker='x')  
plt.xlabel('Year')  
plt.ylabel('Yield (tons per hectare)')  
plt.title("Crop Yields in Kanto")  
plt.legend(['Apples', 'Oranges']);
```



```
sns.set_style("darkgrid")
```

```
plt.plot(years, apples, marker='o')  
plt.plot(years, oranges, marker='x')  
plt.xlabel('Year')  
plt.ylabel('Yield (tons per hectare)')  
plt.title("Crop Yields in Kanto")  
plt.legend(['Apples', 'Oranges']);
```

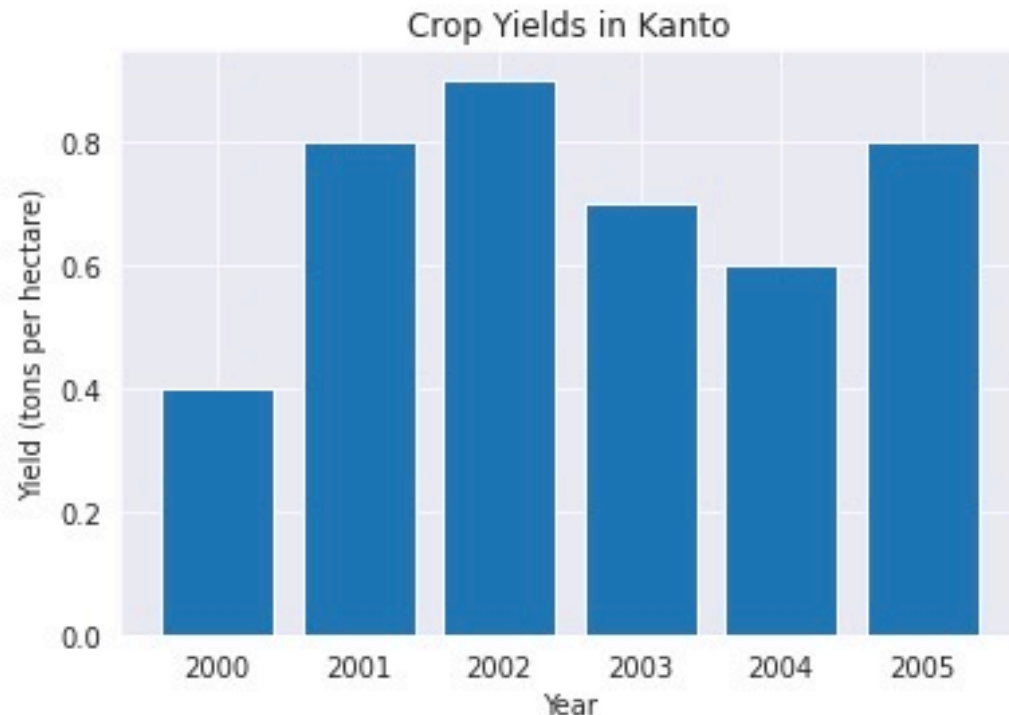


# Python: Bar Charts

```
years = range(2000, 2006)
apples = [0.35, 0.6, 0.9, 0.8, 0.65, 0.8]
oranges = [0.4, 0.8, 0.9, 0.7, 0.6, 0.8]
```

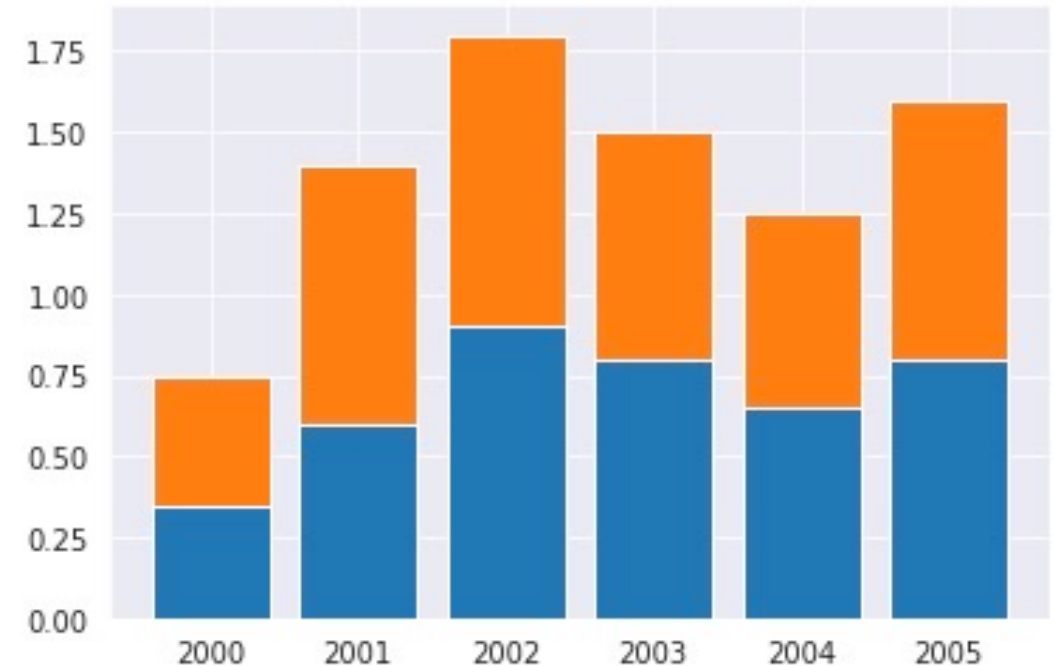
```
plt.bar(years, oranges)
plt.xlabel('Year')
plt.ylabel('Yield (tons per hectare)')
plt.title("Crop Yields in Kanto")
```

```
Text(0.5, 1.0, 'Crop Yields in Kanto')
```



```
plt.bar(years, apples)
plt.bar(years, oranges, bottom=apples)
```

<BarContainer object of 6 artists>



# Python: Histograms

```
flowers_df = sns.load_dataset("iris")
```

```
flowers_df.sepal_width
```

```
0      3.5  
1      3.0  
2      3.2  
3      3.1  
4      3.6
```

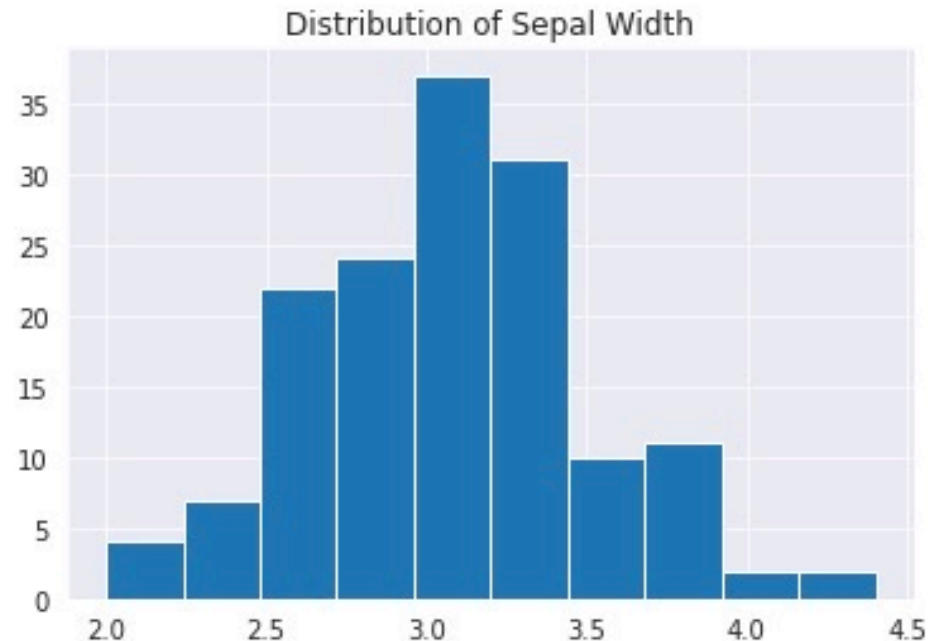
```
...
```

```
145     3.0  
146     2.5  
147     3.0  
148     3.4  
149     3.0
```

```
Name: sepal_width, Length: 150, dtype: float64
```

```
plt.title("Distribution of Sepal Width")  
plt.hist(flowers_df.sepal_width)
```

```
(array([ 4.,  7., 22., 24., 37., 31., 10., 11.,  2.,  2.]),  
array([2.   , 2.24, 2.48, 2.72, 2.96, 3.2  , 3.44, 3.68, 3.92, 4.16, 4.4 ]),  
<BarContainer object of 10 artists>)
```



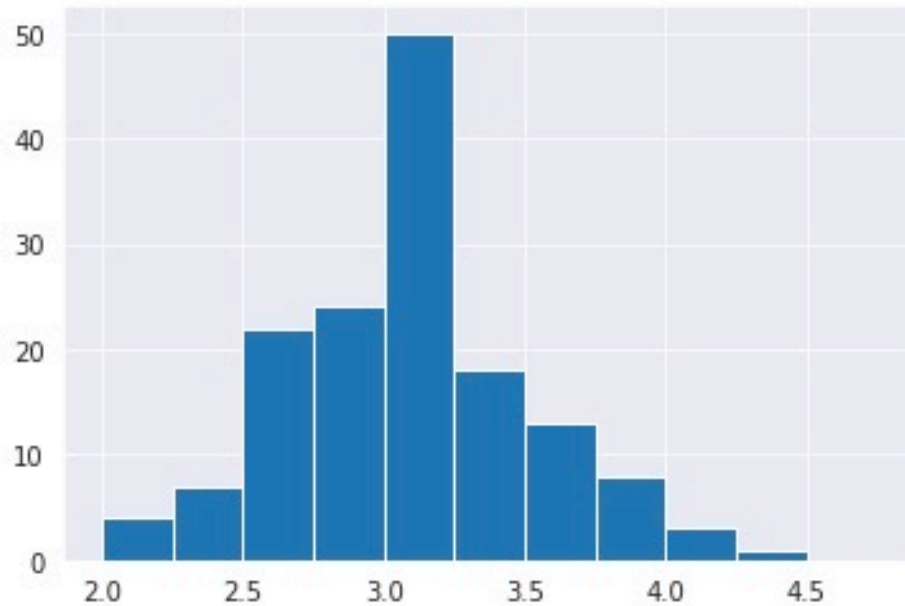


# Python: Histograms

```
import numpy as np
```

```
plt.hist(flowers_df.sepal_width, bins=np.arange(2, 5, 0.25))
```

```
(array([ 4.,  7., 22., 24., 50., 18., 13.,  8.,  3.,  1.,  0.]),  
 array([2. , 2.25, 2.5 , 2.75, 3. , 3.25, 3.5 , 3.75, 4. , 4.25, 4.5 ,  
        4.75]),  
 <BarContainer object of 11 artists>)
```

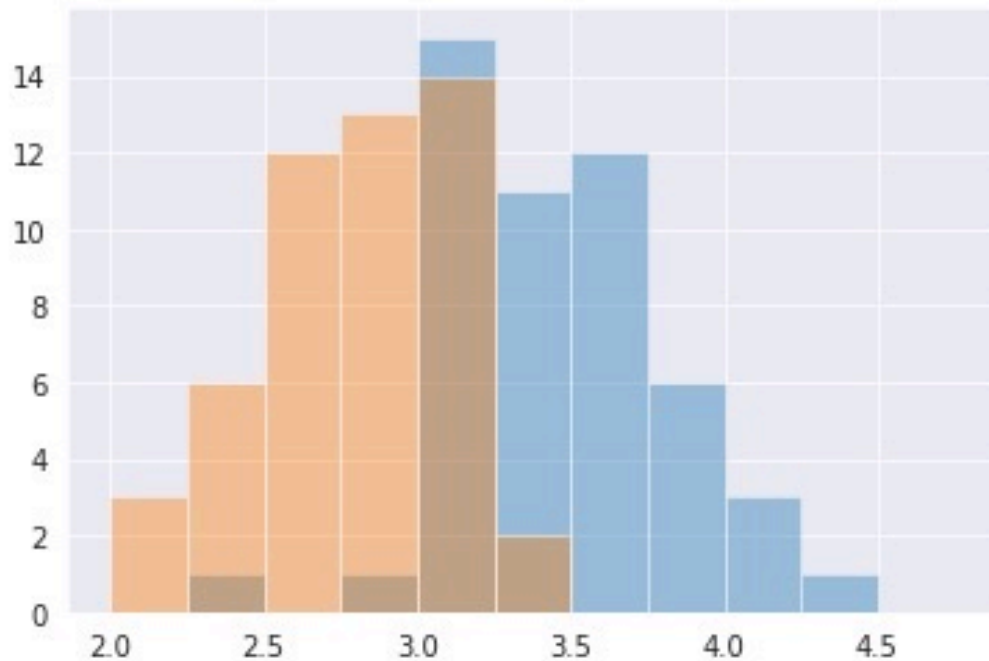


Using numpy to set the bin sizes

# Python: Histograms

```
setosa_df = flowers_df[flowers_df.species == 'setosa']  
versicolor_df = flowers_df[flowers_df.species == 'versicolor']  
virginica_df = flowers_df[flowers_df.species == 'virginica']
```

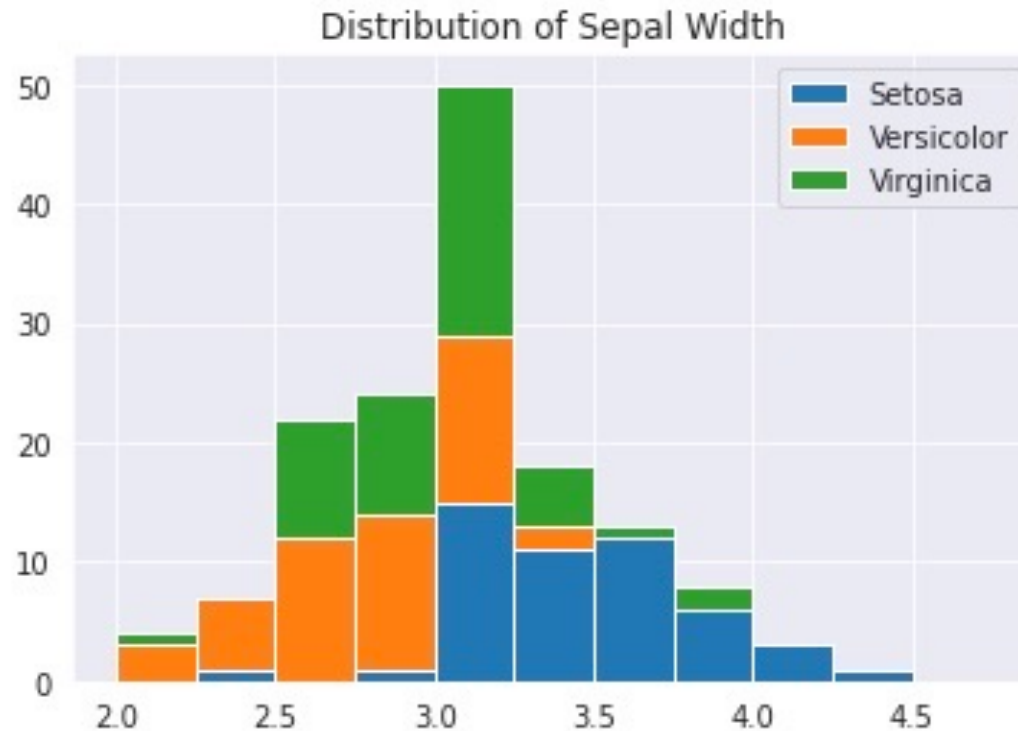
```
plt.hist (setosa_df.sepal_width, alpha=0.4, bins=np.arange(2, 5, 0.25));  
plt.hist (versicolor_df.sepal_width, alpha=0.4, bins=np.arange(2, 5, 0.25));
```



Multiple histograms using opacity

# Python: Histograms

```
plt.title('Distribution of Sepal Width')
plt.hist([setosa_df.sepal_width, versicolor_df.sepal_width,
         virginica_df.sepal_width],
         bins=np.arange(2, 5, 0.25),
         stacked=True);
plt.legend(['Setosa', 'Versicolor', 'Virginica']);
```



Stacked histograms



# Python: Scatter Plots

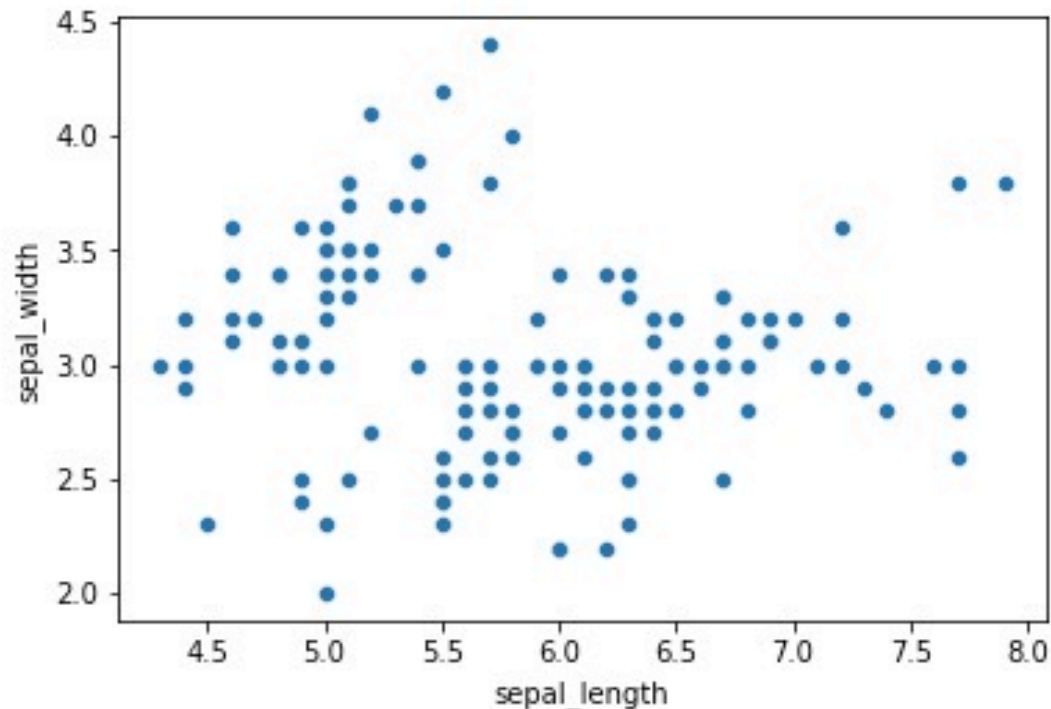
```
flowers_df = sns.load_dataset("iris")
```

```
flowers_df.species.unique()
```

```
array(['setosa', 'versicolor', 'virginica'], dtype=object)
```

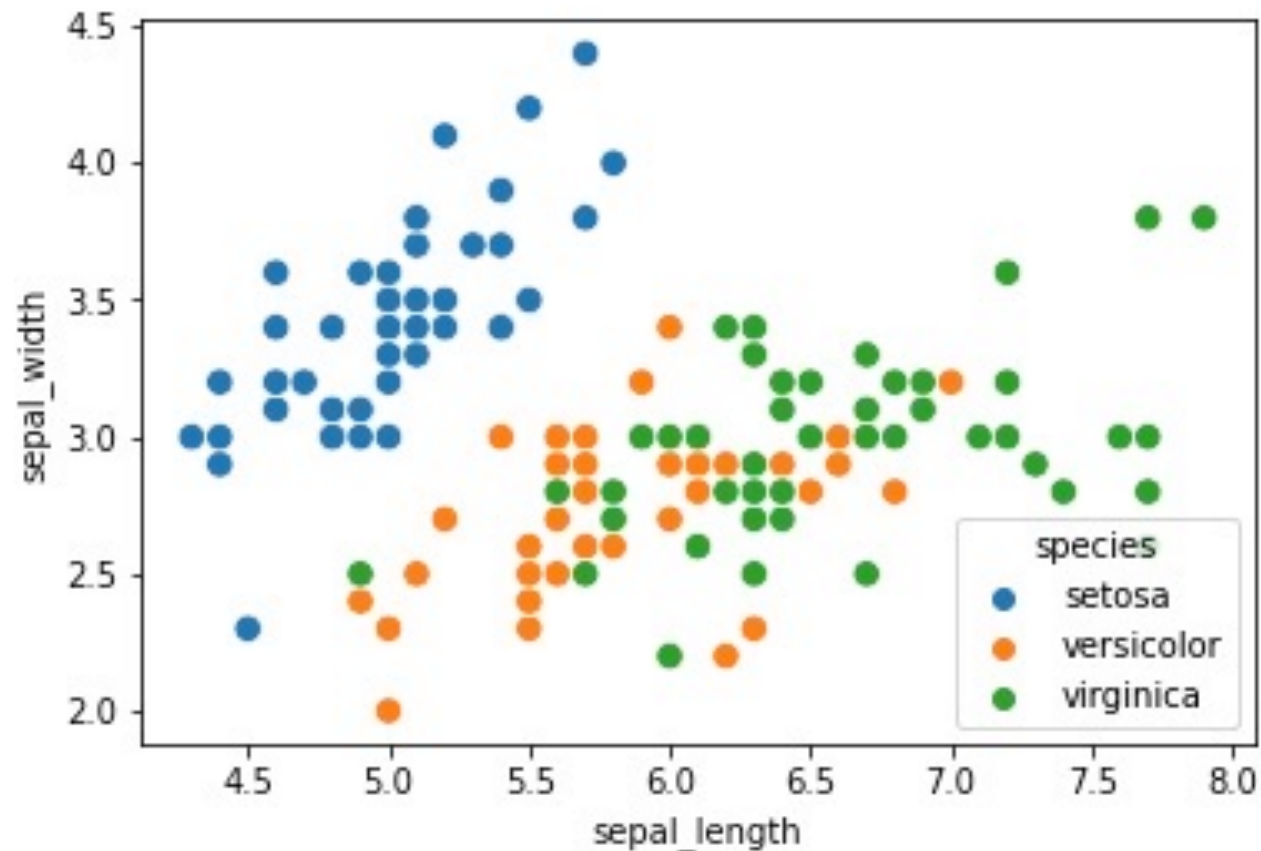
```
sns.scatterplot(x=flowers_df.sepal_length, y=flowers_df.sepal_width)
```

```
<AxesSubplot:xlabel='sepal_length', ylabel='sepal_width'>
```



# Python: Scatter Plots

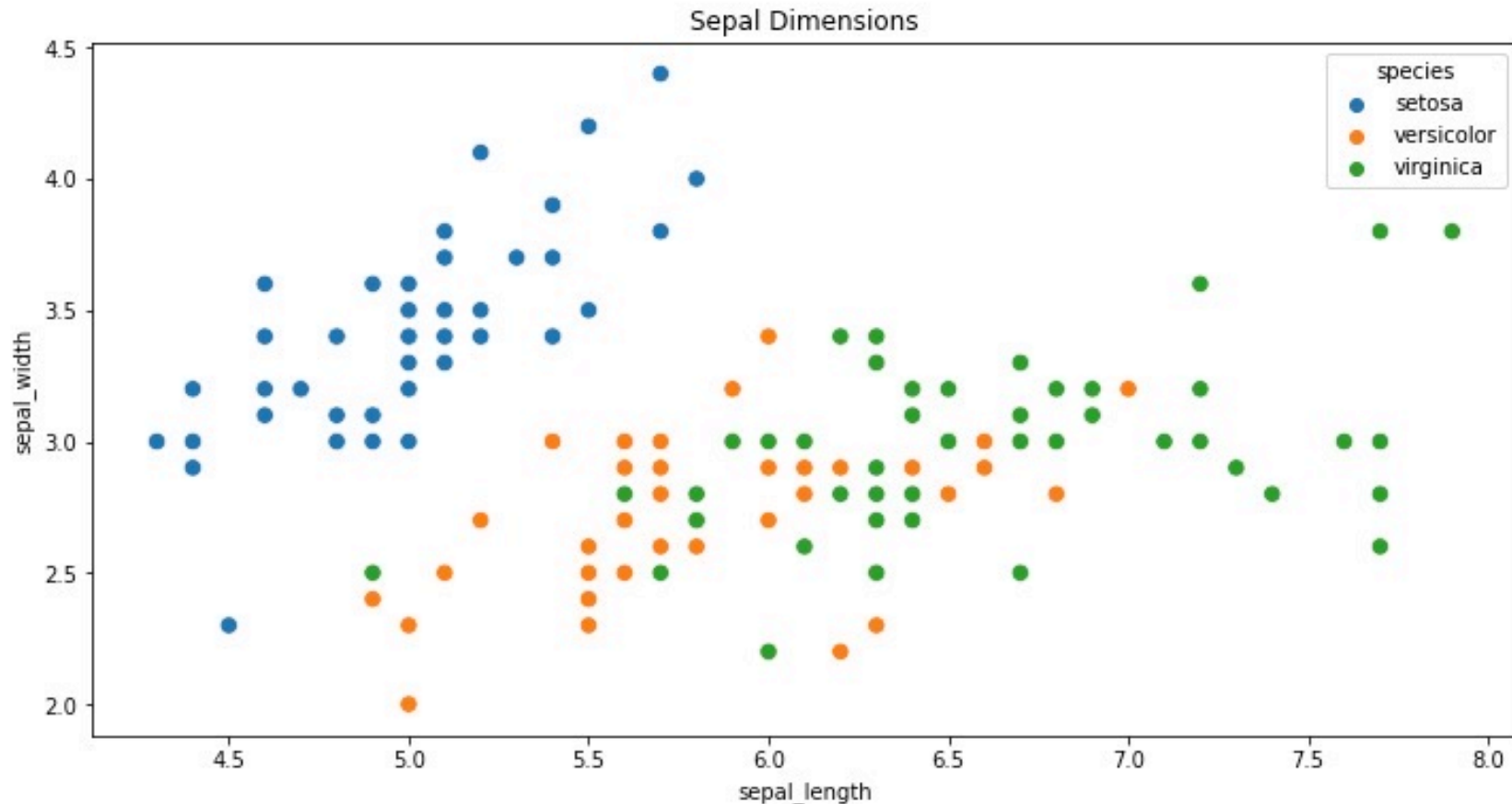
```
sns.scatterplot(x=flowers_df.sepal_length,  
                y=flowers_df.sepal_width,  
                hue=flowers_df.species, s=70);
```



Add color to distinguish species

# Python: Scatter Plots

```
plt.figure(figsize=(12, 6))  
plt.title('Sepal Dimensions')  
sns.scatterplot(x=flowers_df.sepal_length,  
                y=flowers_df.sepal_width,  
                hue=flowers_df.species, s=70);
```



Do the same with seaborn

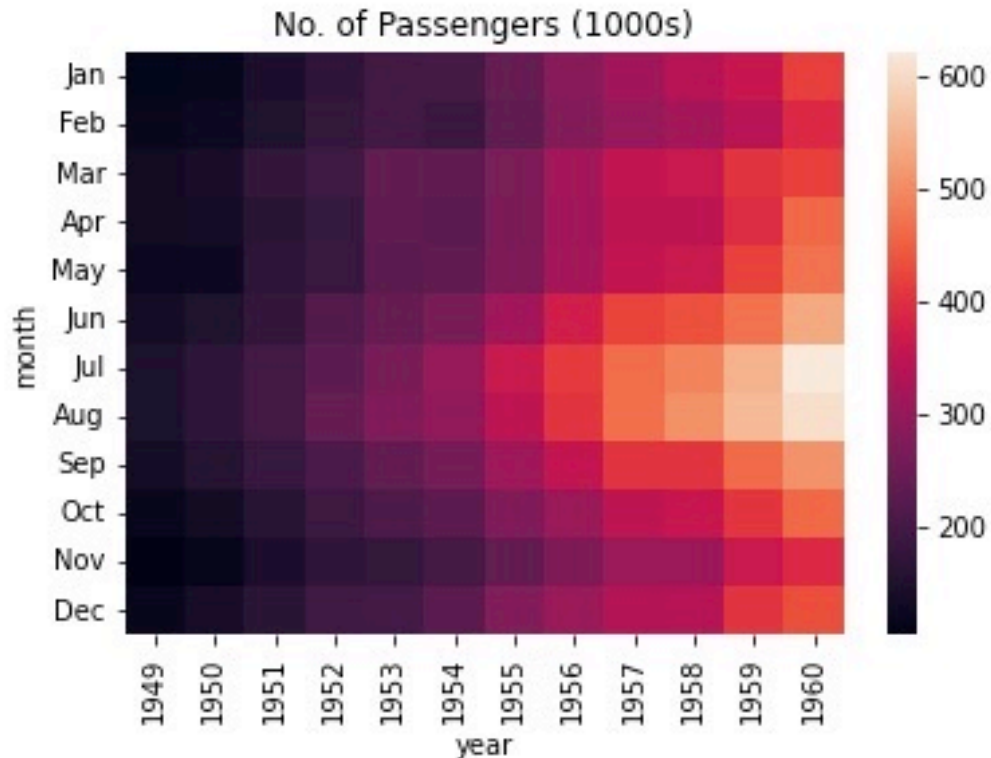


# Python: Heat Maps

```
flights_df = sns.load_dataset("flights").pivot("month", "year", "passengers")
```

```
plt.title("No. of Passengers (1000s)")  
sns.heatmap(flights_df)
```

```
<AxesSubplot:title={'center':'No. of Passengers (1000s)'}, xlabel='year', ylabel='month'>
```



Brighter colors mean more airline traffic

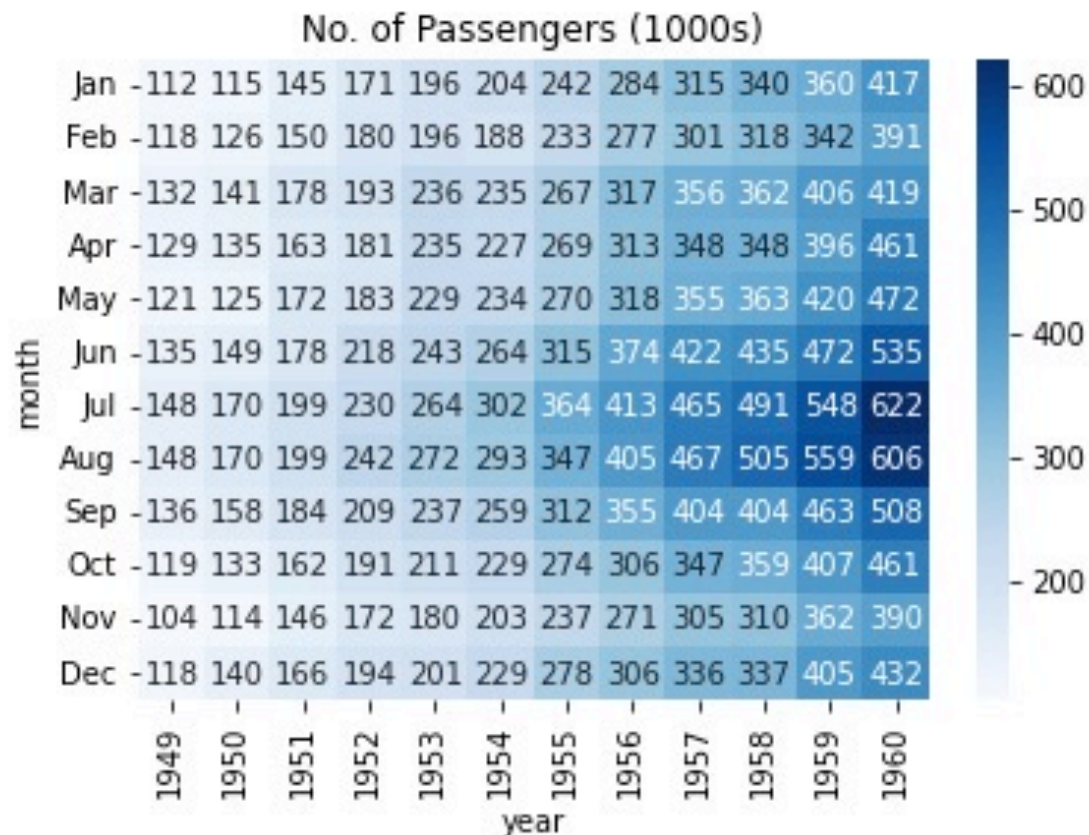
Two conclusions:

1. There is more traffic in July and August
2. Traffic is generally growing each year

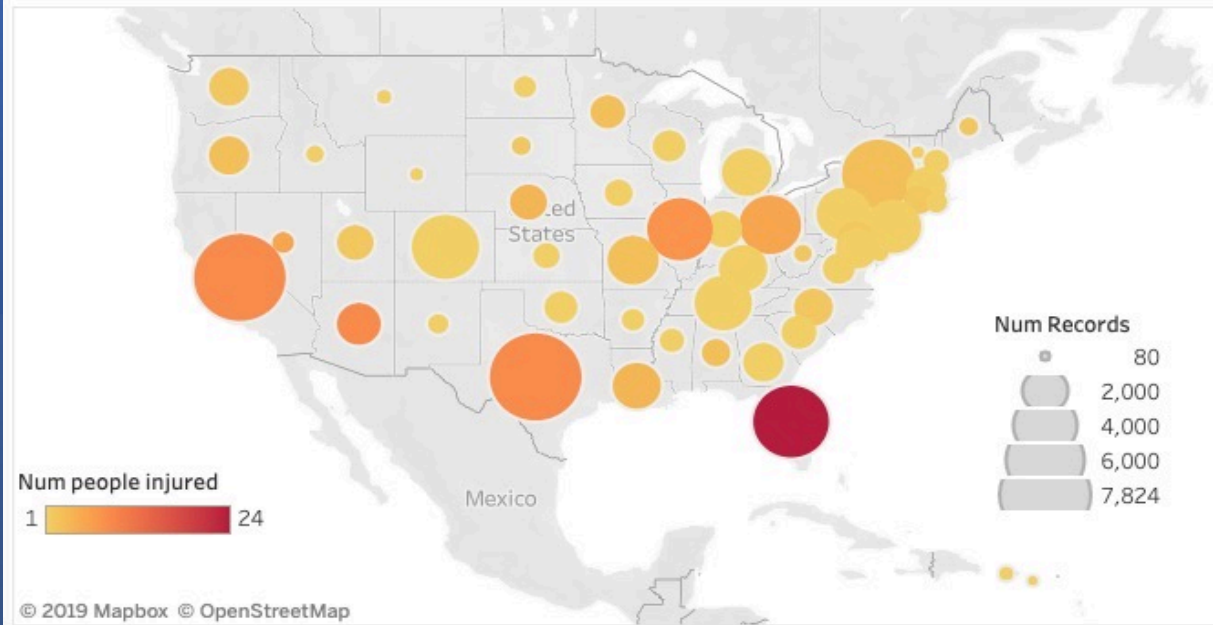
# Python: Heat Maps

```
plt.title("No. of Passengers (1000s)")  
sns.heatmap(flights_df, fmt="d", annot=True, cmap='Blues')
```

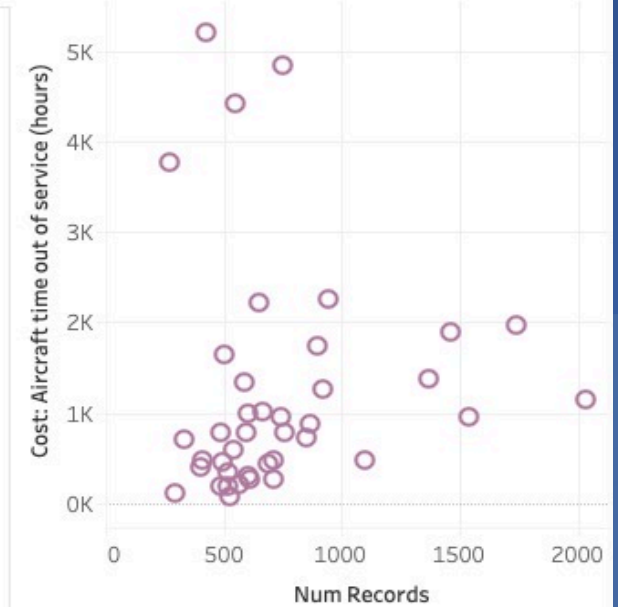
```
<AxesSubplot:title={'center':'No. of Passengers (1000s)'}, xlabel='year',
```



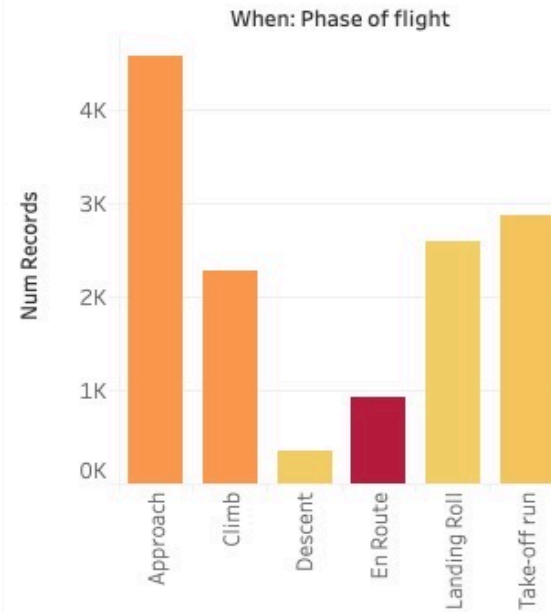
## Bird Strikes: 2000 to 2011



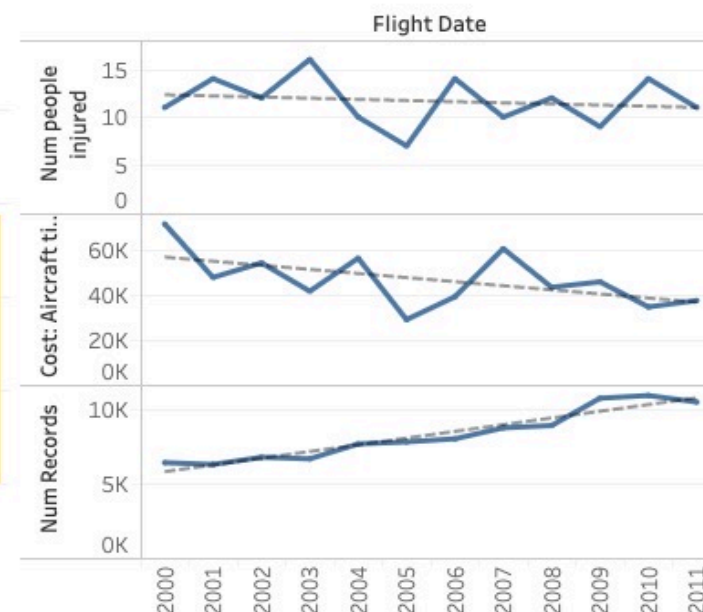
## Top 40 Airports



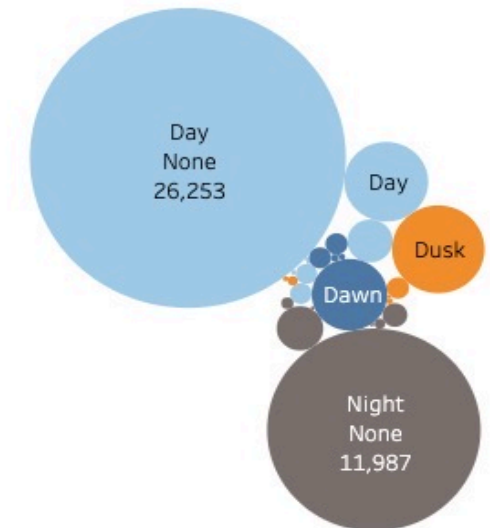
## Phase of Flight



## Trends

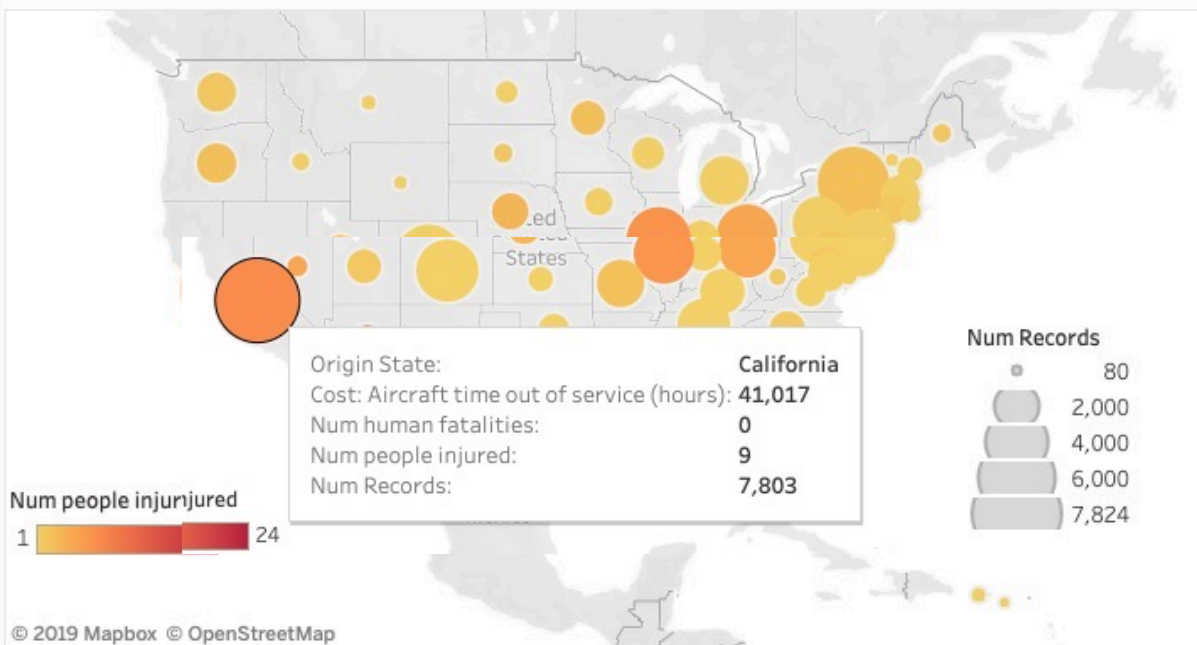


## Conditions

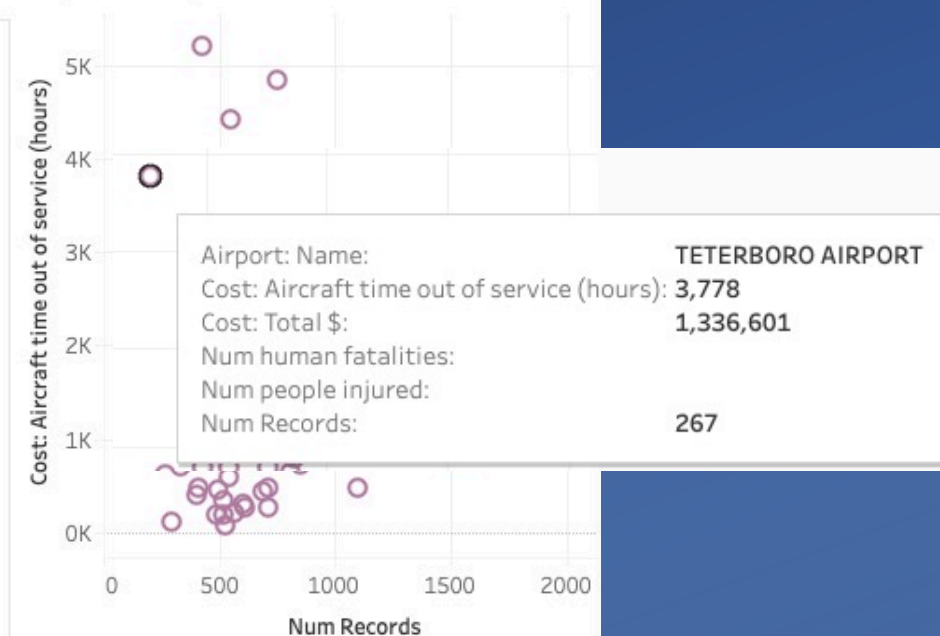




## Bird Strikes: 2000 to 2011

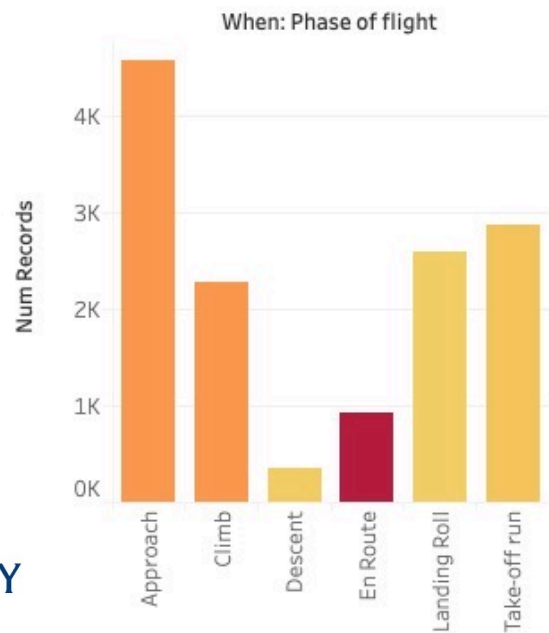


## Top 40 Airports

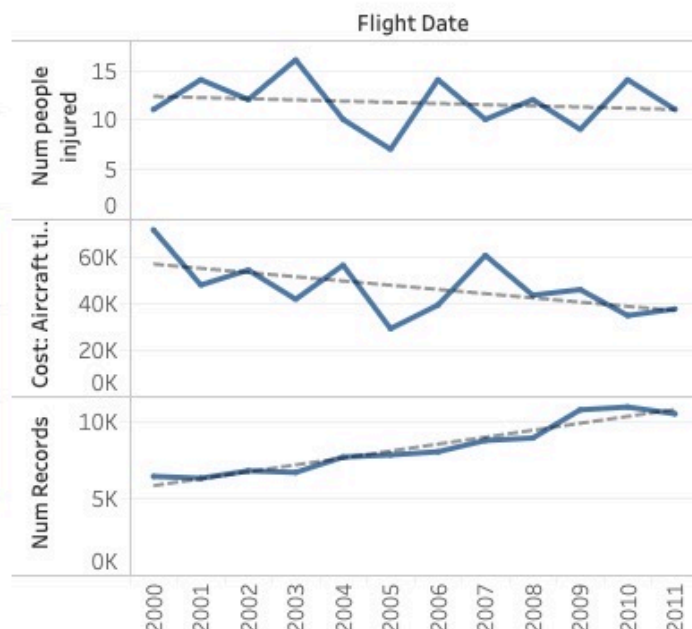


Underlying  
Details  
Hover over  
data points

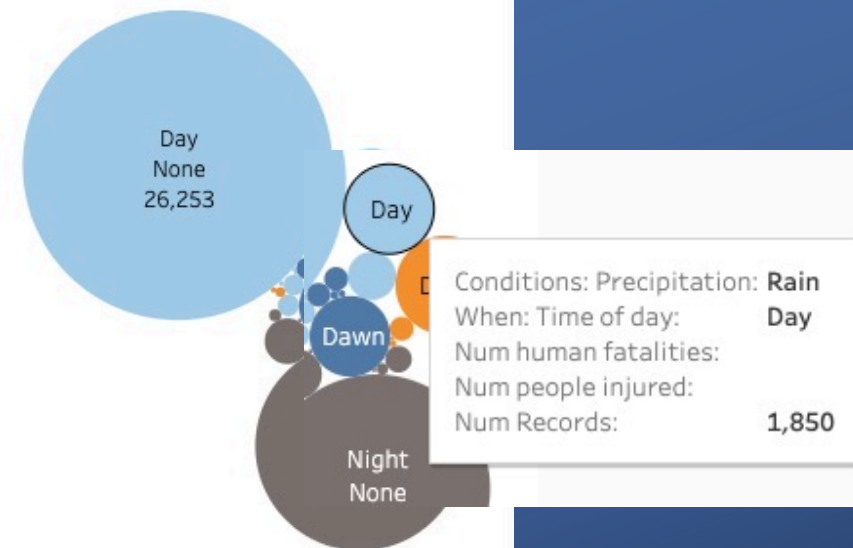
## Phase of Flight



## Trends



## Conditions





Details and Statistics

Hover over a data point to see more information

Trends

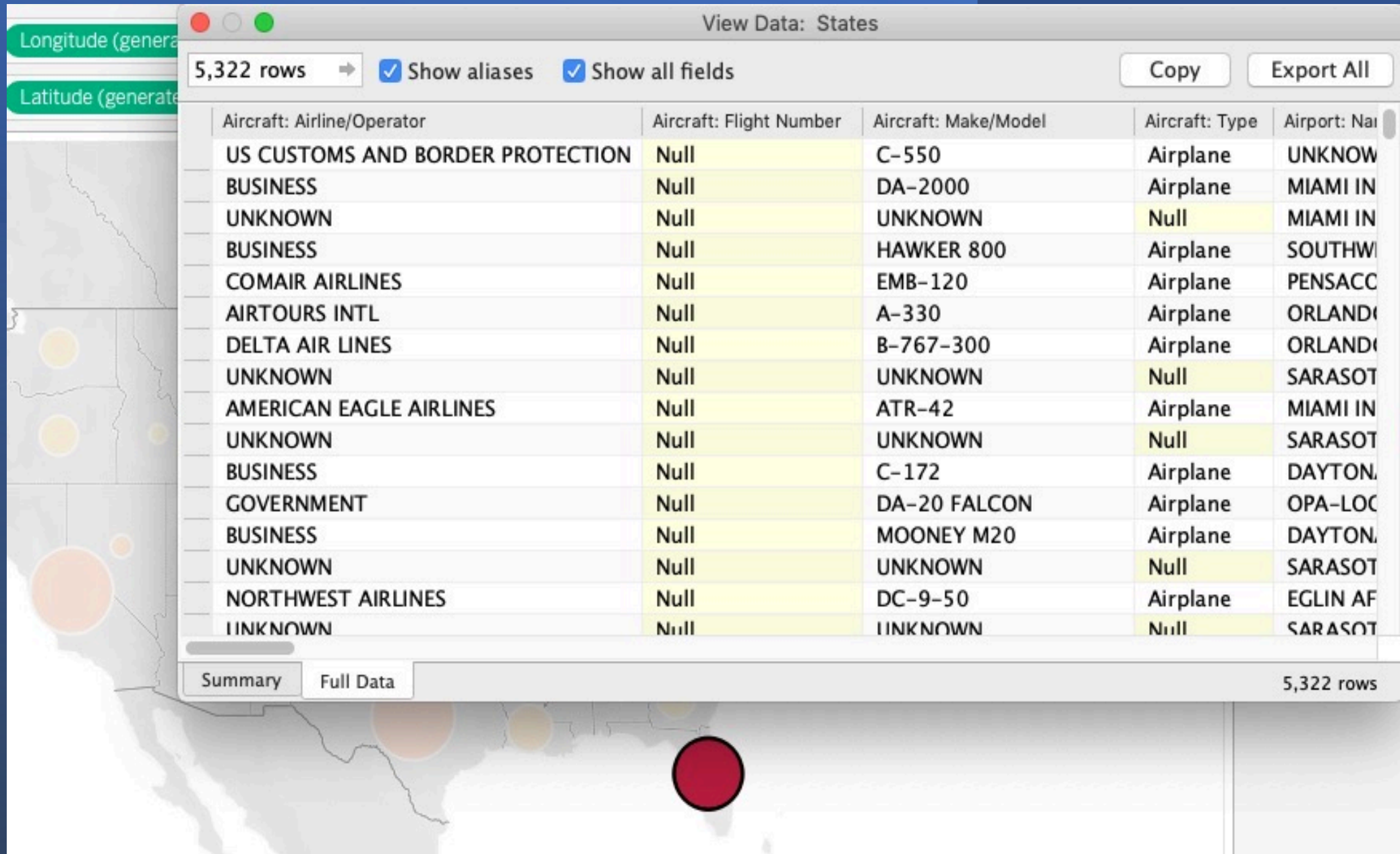




tableau

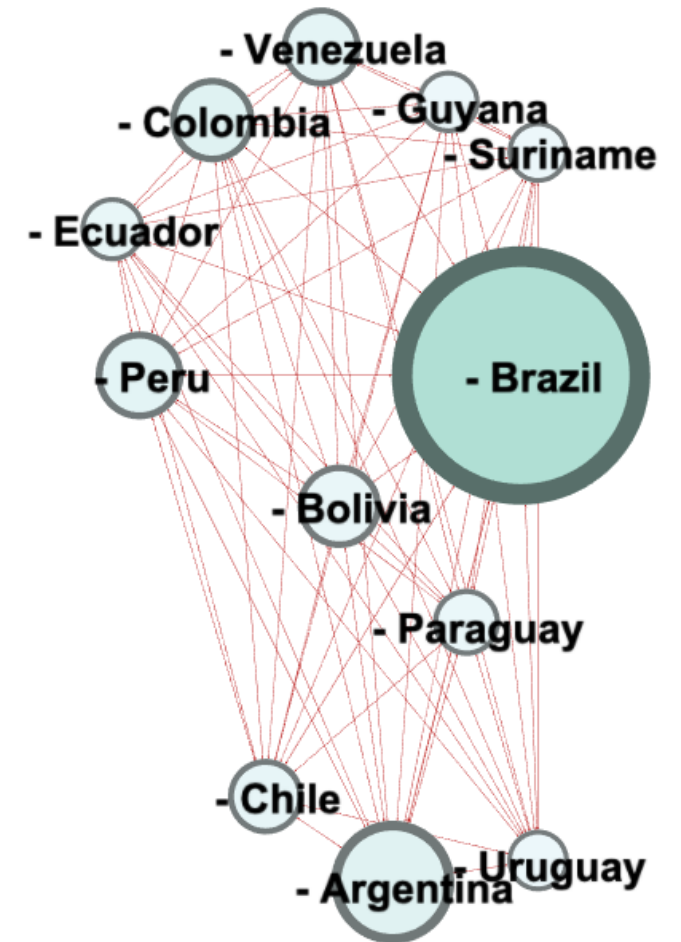
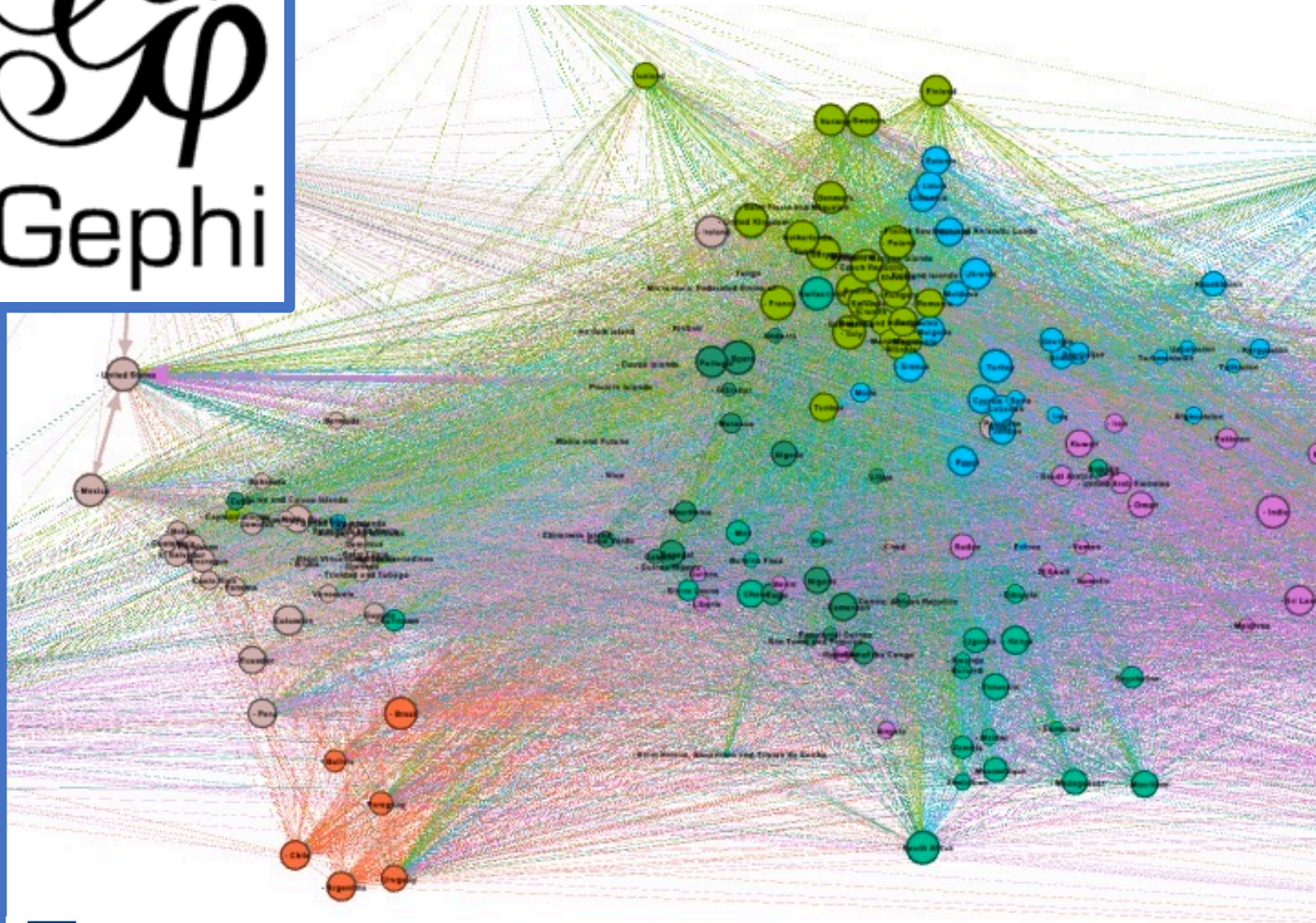
## Show Every Record for Florida

Click on data point and choose View Data option





# World Trade with South America Insert





# World Trade with Insert of South America to the Rest of the World

