



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-

OPEN

OnDemand

OnDemand provides an integrated,

Pinned Apps A featured sub-

DS SIMULIA Abaqus

ABAQUS GUI

System Installed App

ANSYS Workbench

GUI

System Installed App

RStudio Server

System Installed App

MATLAB GUI

System Installed App

Mathematica GUI

System Installed App

VSCode GUI

System Installed App

Desktops

Interactive Desktop

GUIs

ABAQUS GUI

ANSYS Workbench GUI

MATLAB GUI

Mathematica GUI

VSCode GUI

Servers

Jupyter Notebook

RStudio Server

RStudio Server

MATLAB GUI

System Installed App

The University of Arizona

Using RStudio on HPC

The screenshot shows the RStudio interface running on a High-Performance Computing (HPC) system. The top navigation bar includes File, Edit, Code, View, Plots, Session, Build, Debug, Profile, Tools, and Help. The user is currently in the Environment pane, which displays the Global Environment and Data sections. The Data section shows a dataset named 'mydata' with 15 observations and 2 variables. The 'Values' table lists the contents of each variable:

	mydata	15 obs. of 2 variables
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

The Files pane shows the local workspace directory structure:

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

The Plots pane is currently empty.

```
R version 4.0.0 (2020-04-24) -- "Arbor Day"
Copyright (C) 2020 The R Foundation for Statistical Computing
Platform: x86_64-pc-linux-gnu (64-bit)

R is free software and comes with ABSOLUTELY NO WARRANTY.
You are welcome to redistribute it under certain conditions.
Type 'license()' or 'licence()' for distribution details.

R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

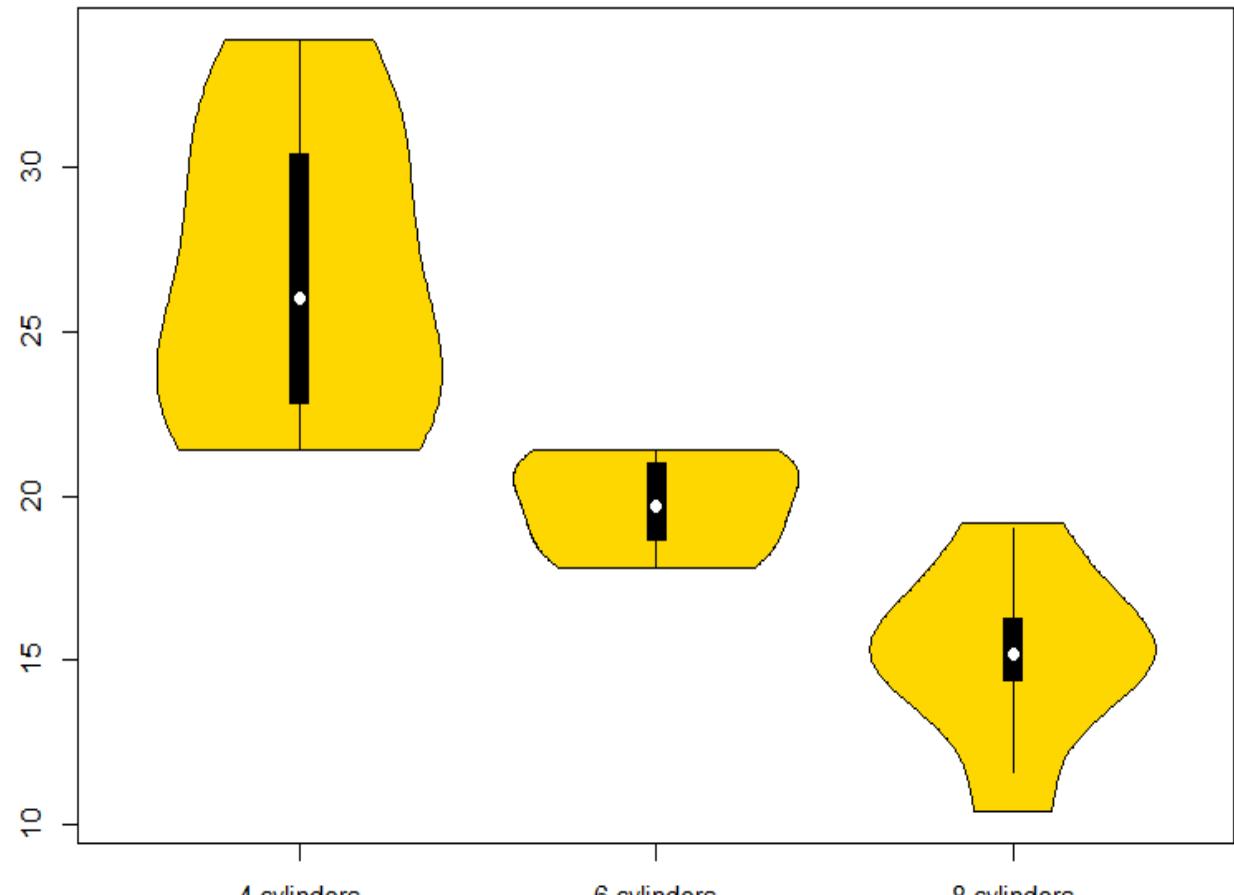
Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

[Workspace loaded from ~/.RData]

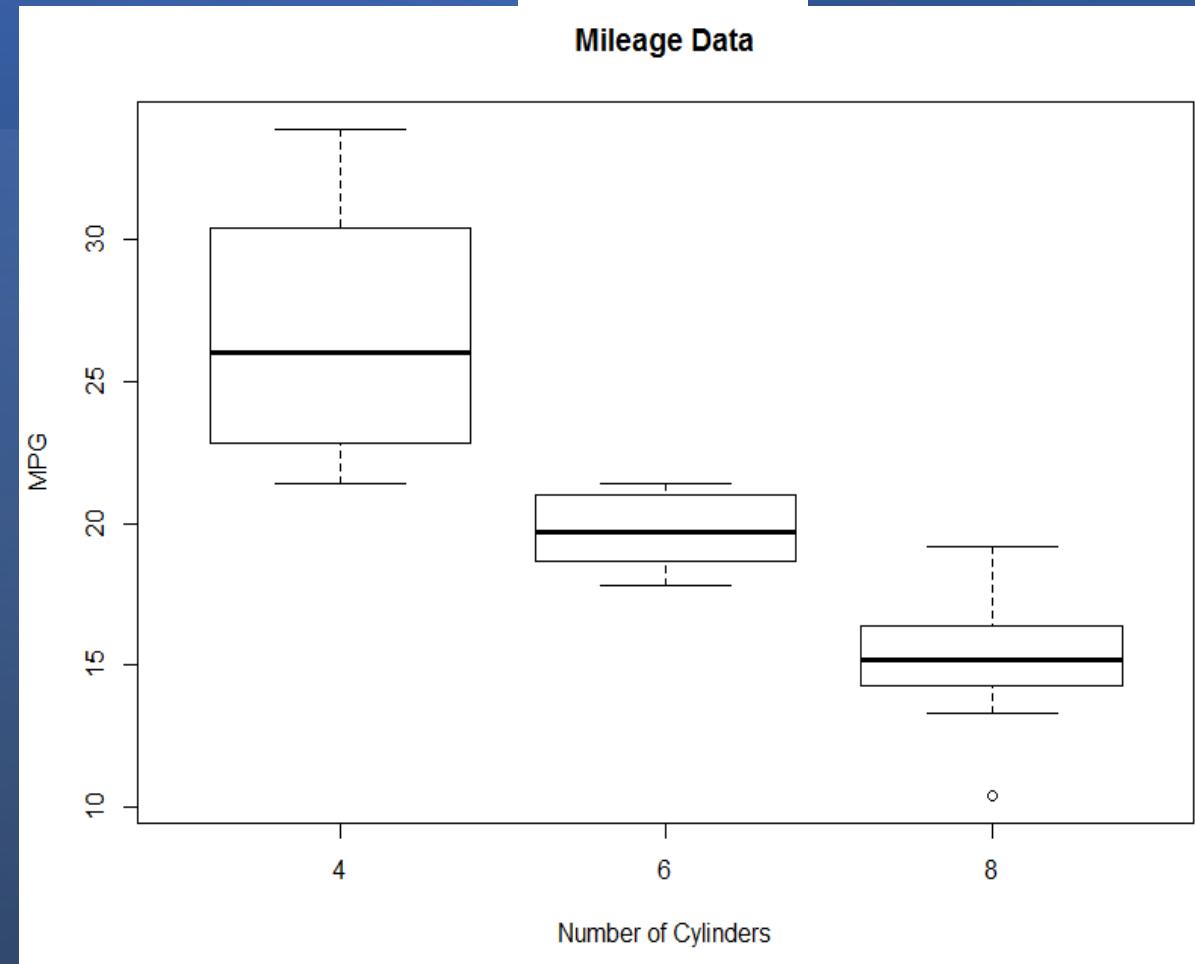
>
```

Using RStudio on HPC

Violin Plot



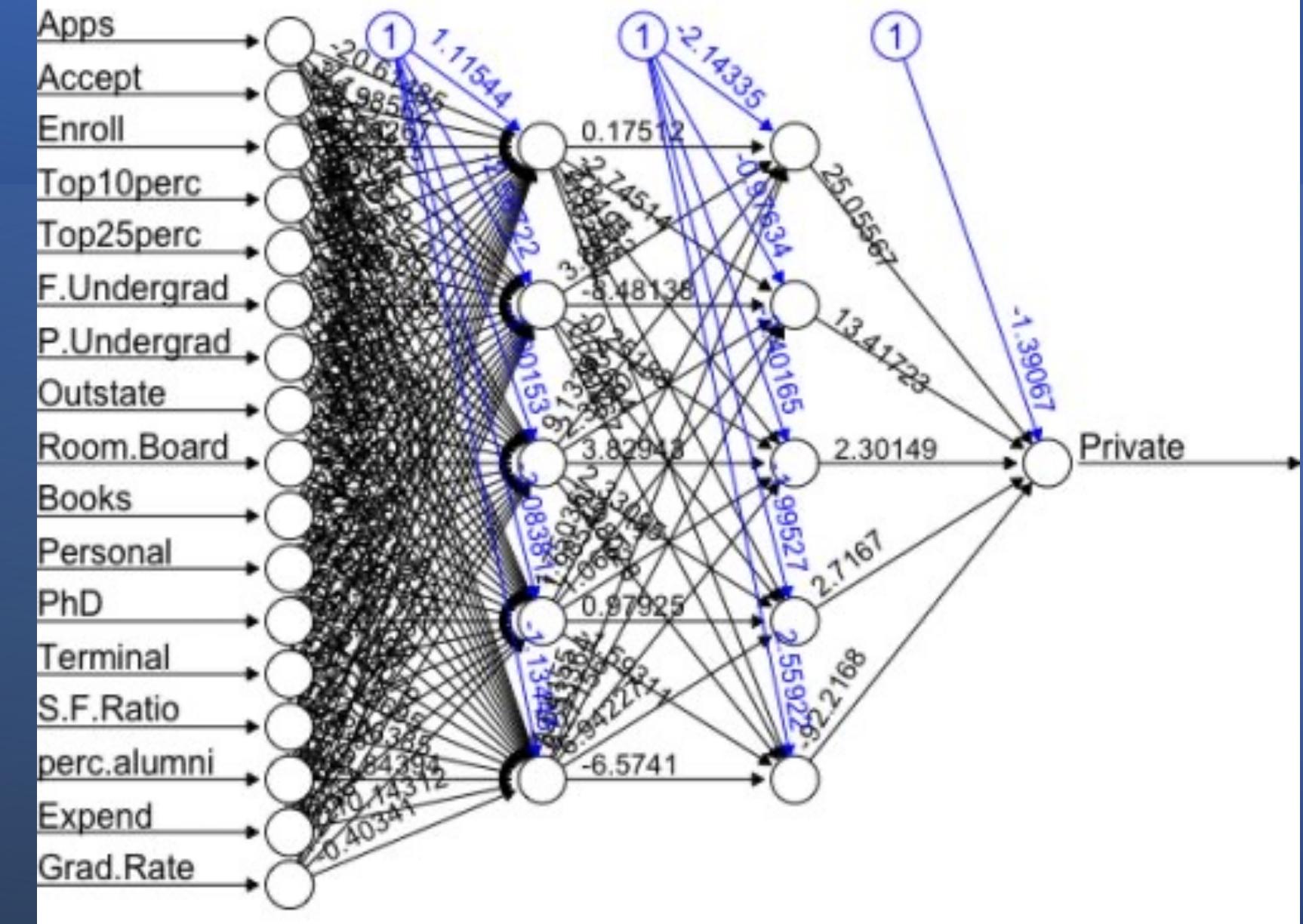
Box Plot



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

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 dat and view the first 5 rows`
`data = pd.read_excel("king_county_house_data.xls")`

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

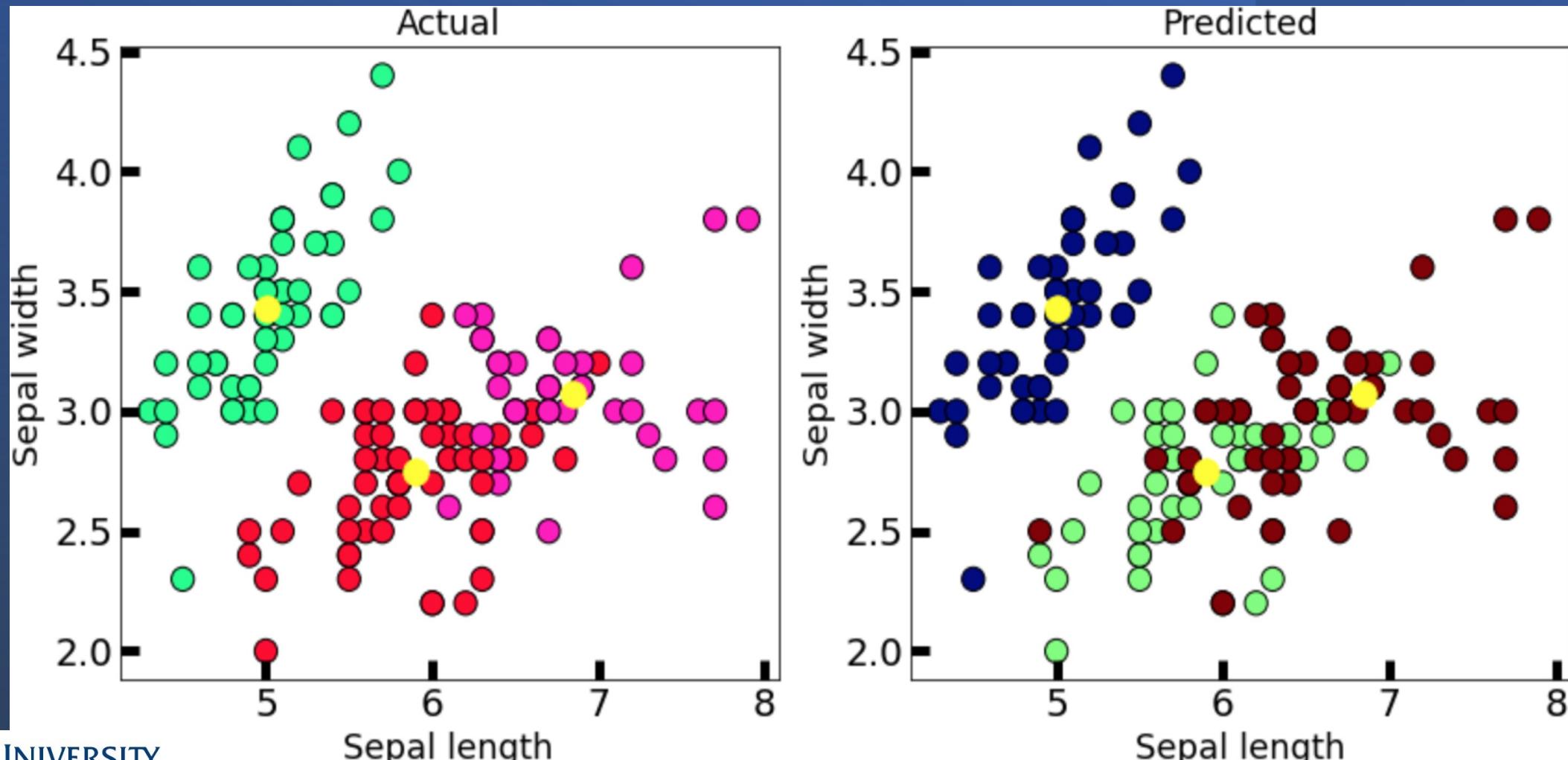
Out[7]:

	id	date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	...	grade	sqf
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

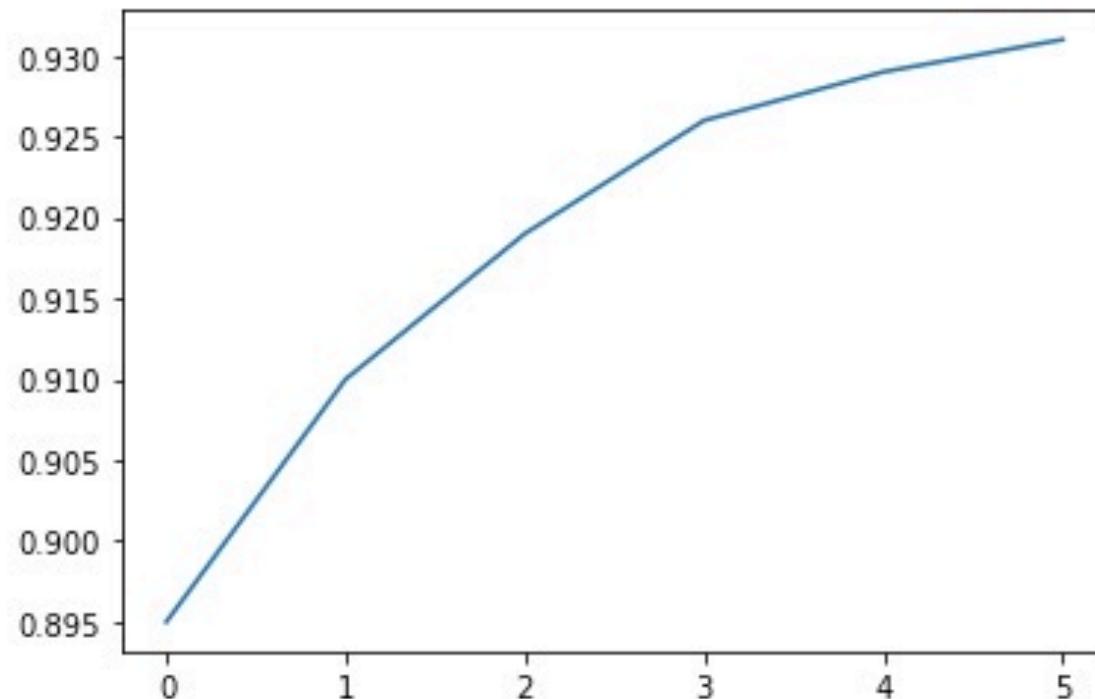
Matplotlib	Seaborn
Used for basic graph plotting like line, bar or pie charts	Mainly used for statistics and performs complex viz with fewer commands
Mainly works with datasets and arrays	Works with entire datasets
Acts productively with data arrays and frames	More organized and functional
Pairs well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

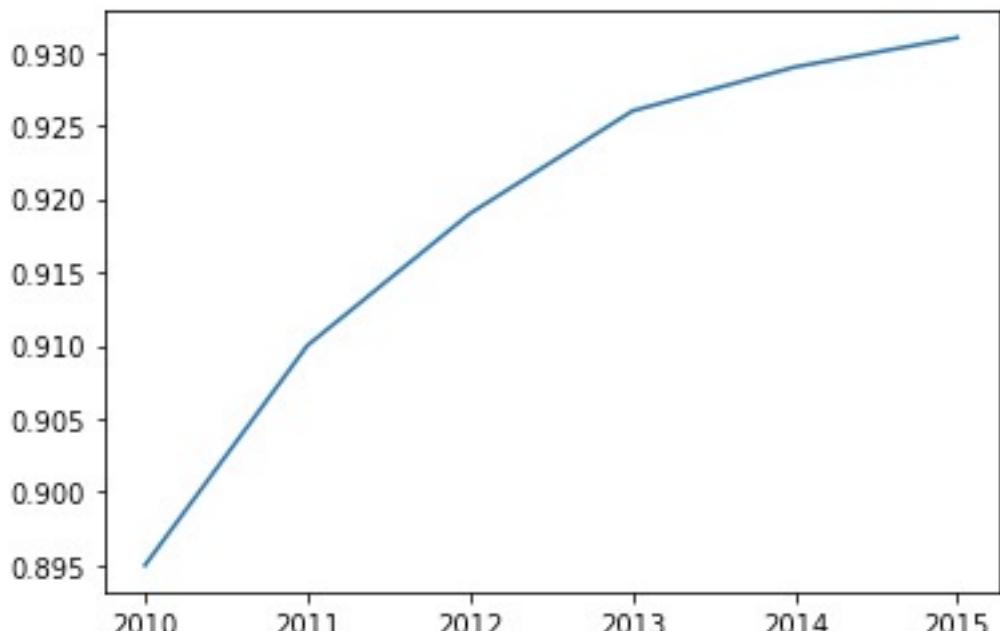
```
Out[2]: [<matplotlib.lines.Line2D at 0x7f0f271e6520>]
```



Python: Line Charts

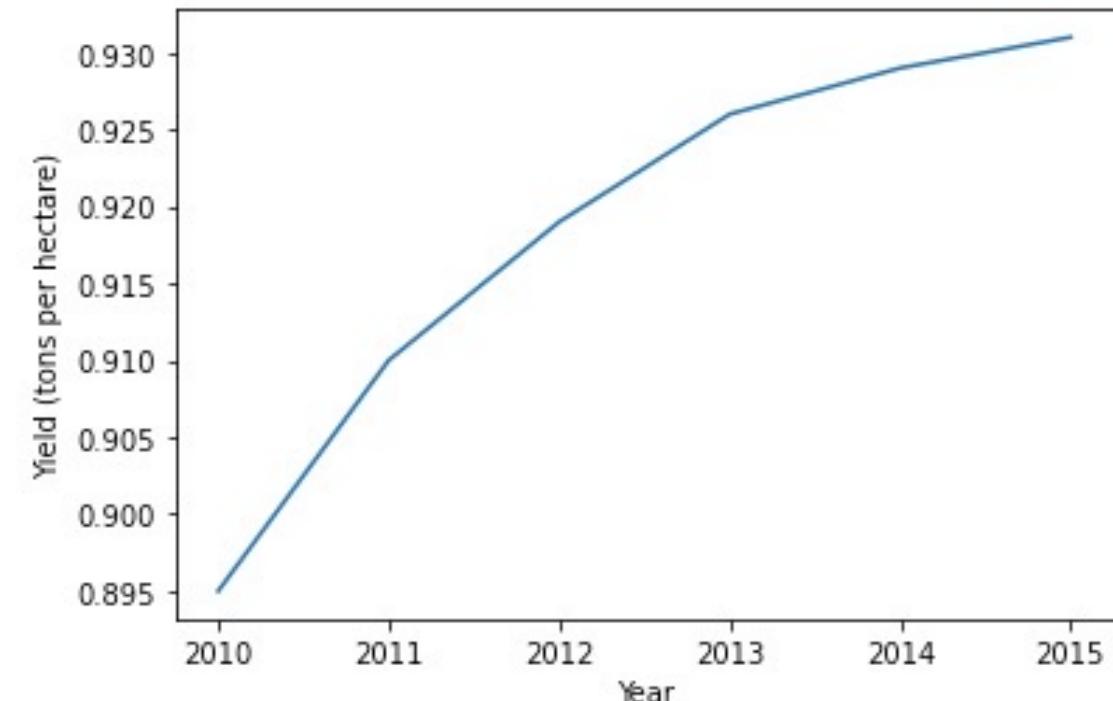
```
years = [2010, 2011, 2012, 2013, 2014, 2015]
yield_apples = [0.895, 0.91, 0.919, 0.926, 0.929, 0.931]
plt.plot(years, yield_apples)
```

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



Add x axis values

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

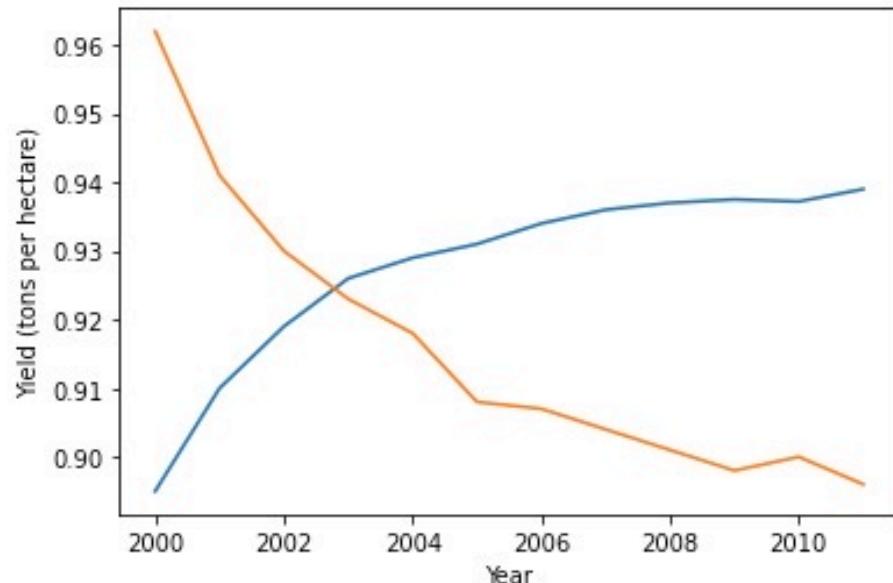


Add labels to the axes

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]
```

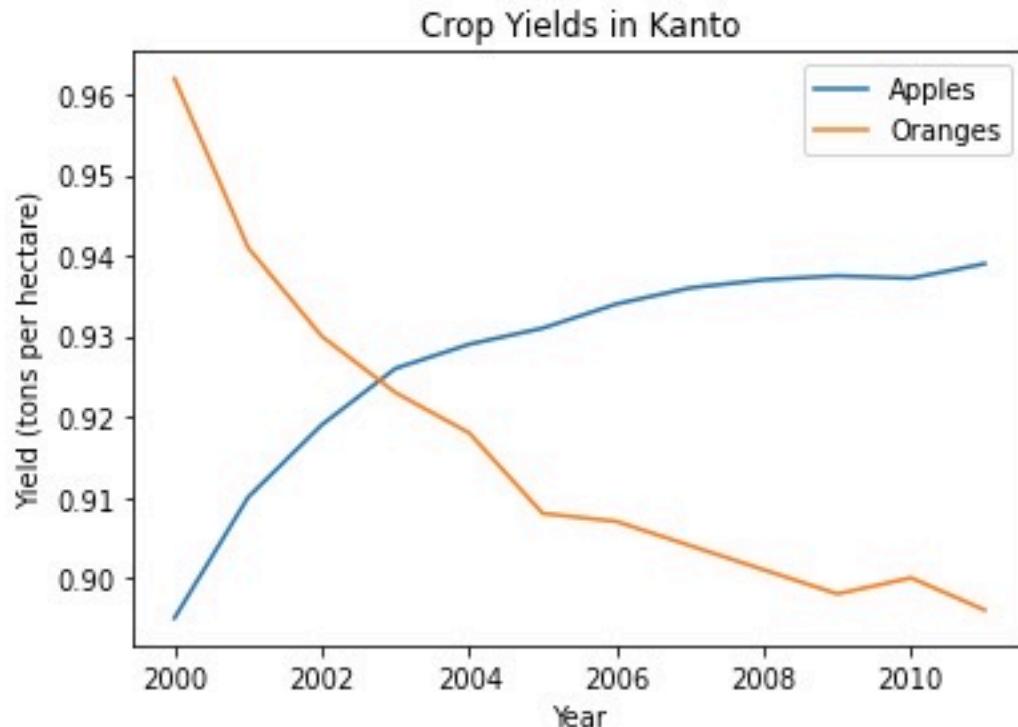
```
In [7]: plt.plot(years, apples)
plt.plot(years, oranges)
plt.xlabel('Year')
plt.ylabel('Yield (tons per hectare)');
```



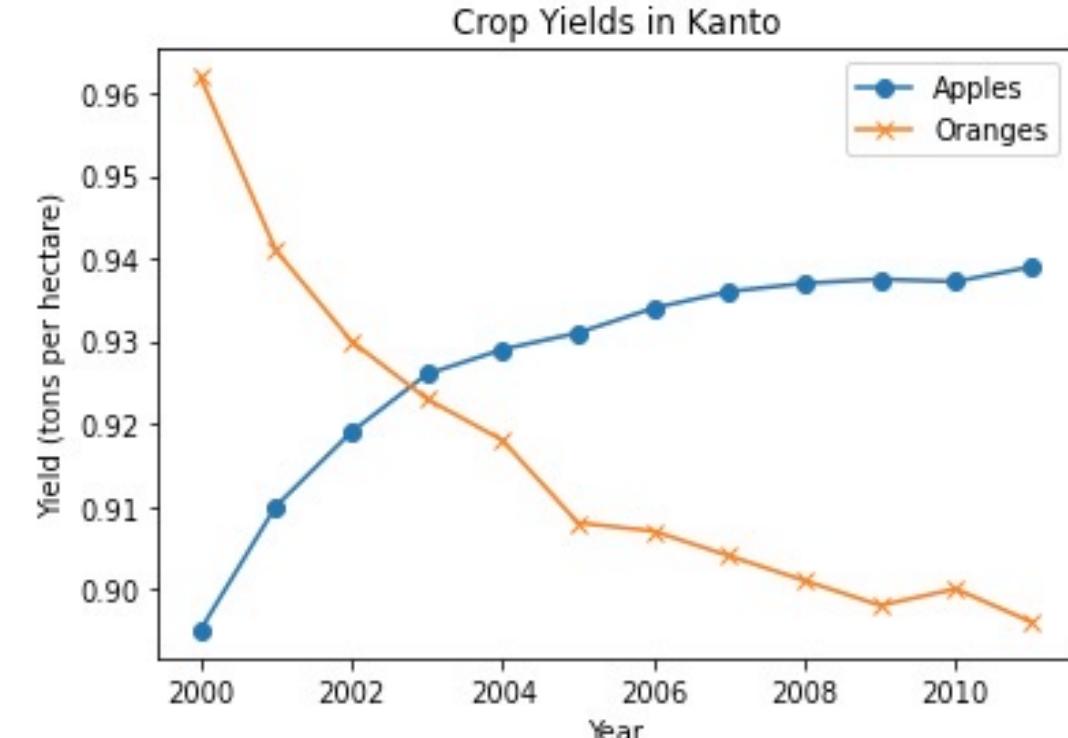
Plotting multiple datasets on the same graph

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']);
```



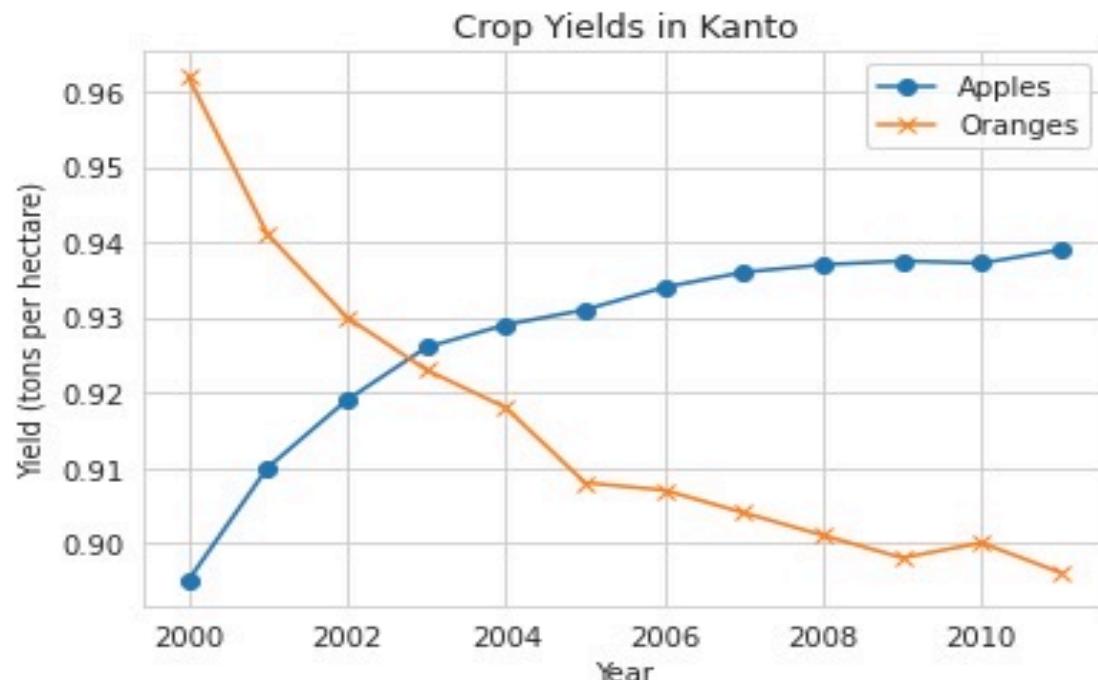
```
plt.plot(years, apples, marker='o')
plt.plot(years, oranges, marker='x')
plt.xlabel('Year')
plt.ylabel('Yield (tons per hectare)')
plt.title("Crop Yields in Kanto")
plt.legend(['Apples', 'Oranges']);
```



Python: Line Charts

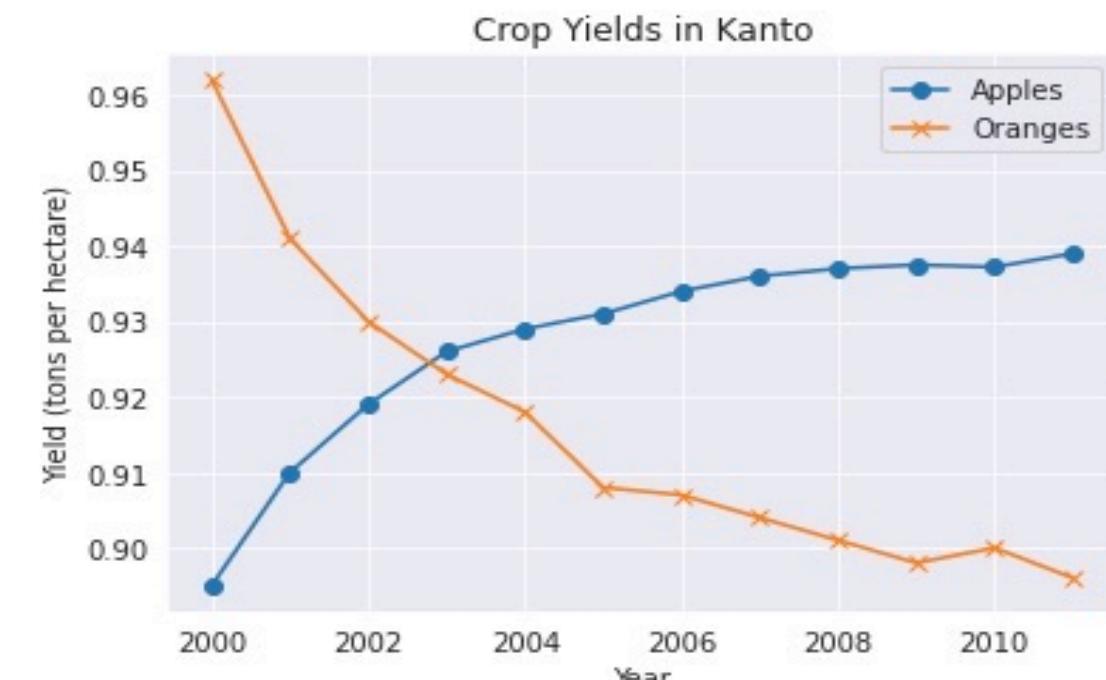
```
sns.set_style("whitegrid")
```

```
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")
```

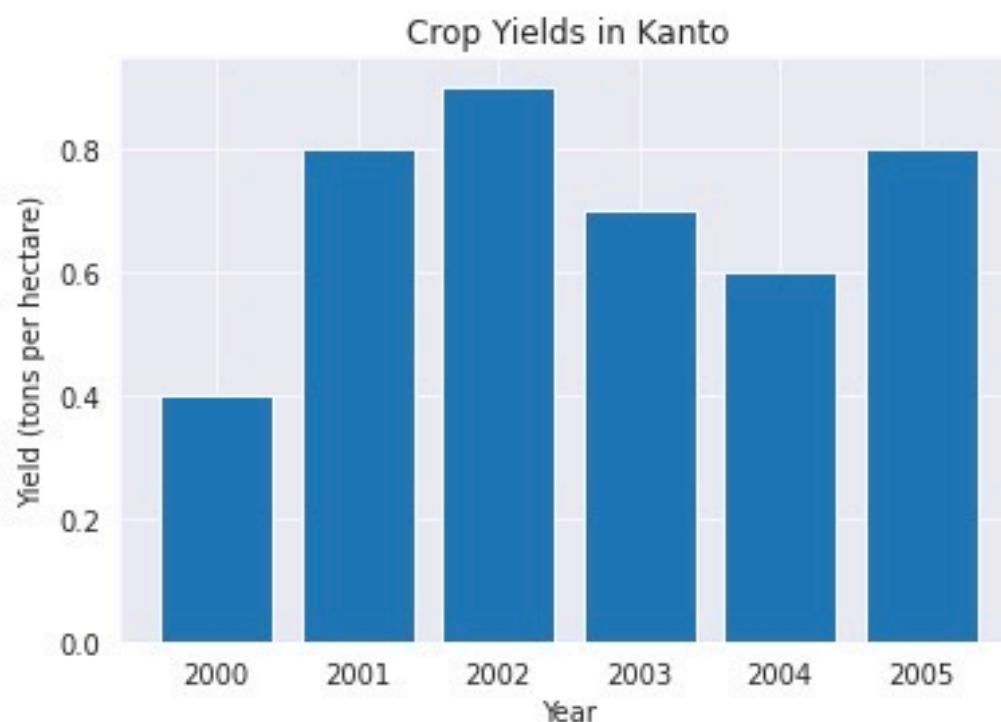
```
plt.plot(years, apples, marker='o')
plt.plot(years, oranges, marker='x')
plt.xlabel('Year')
plt.ylabel('Yield (tons per hectare)')
plt.title("Crop Yields in Kanto")
plt.legend(['Apples', 'Oranges']);
```



```
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')
```

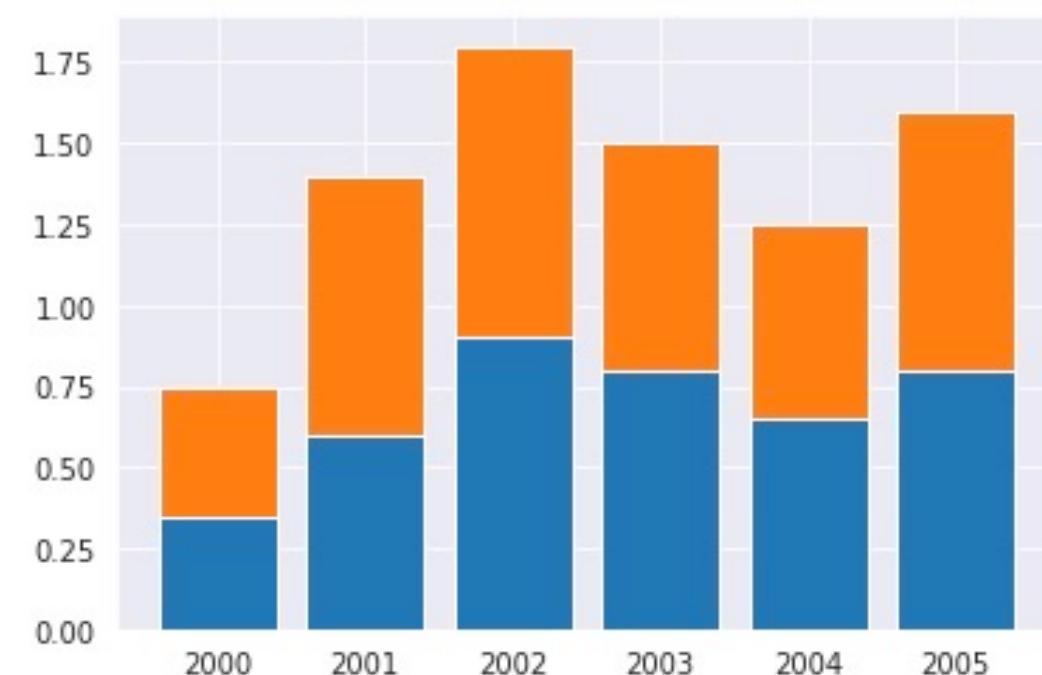


Bar chart

Python: Bar Charts

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

```
<BarContainer object of 6 artists>
```



Stacked Bar chart

Python: Histograms

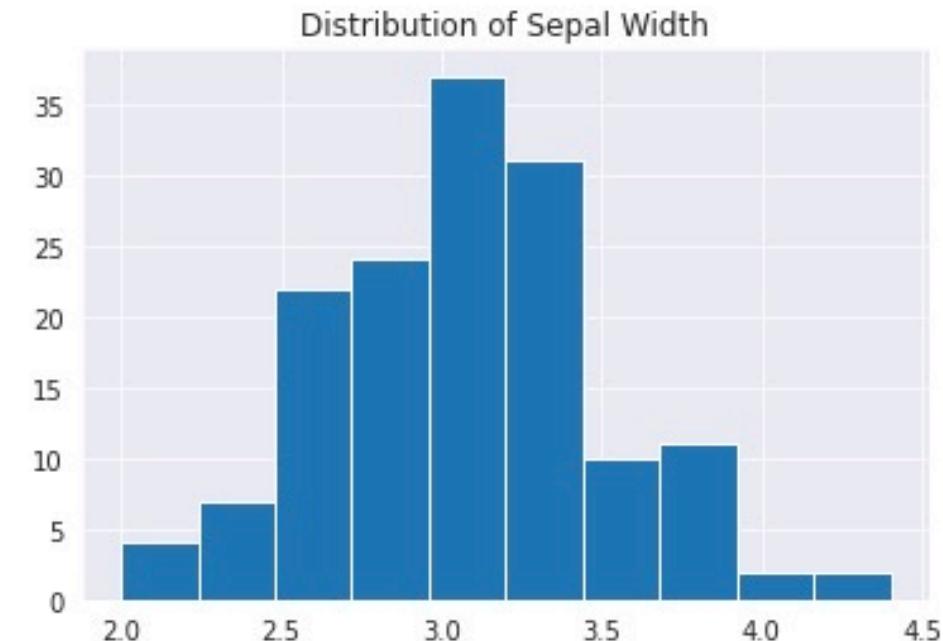
```
flowers_df = sns.load_dataset("iris")
```

```
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:
```

```
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>)
```

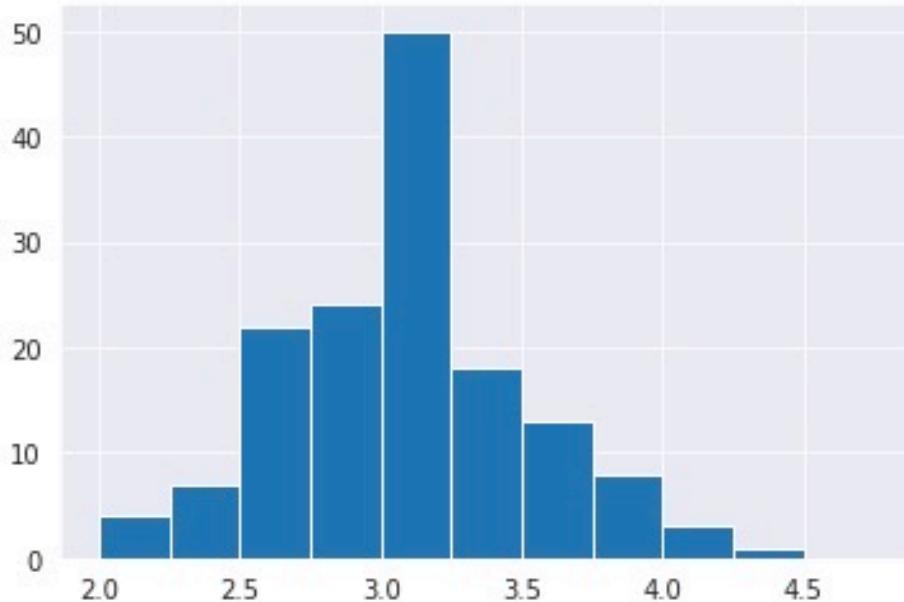


Python: Histograms

```
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>)
```

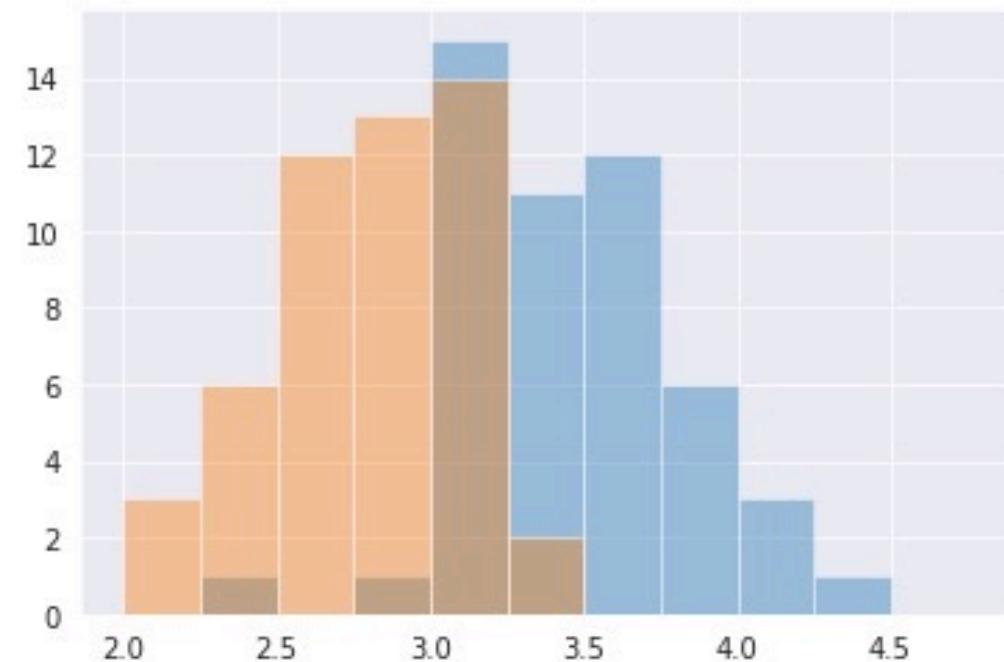


Using numpy to set the bin sizes

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']
```

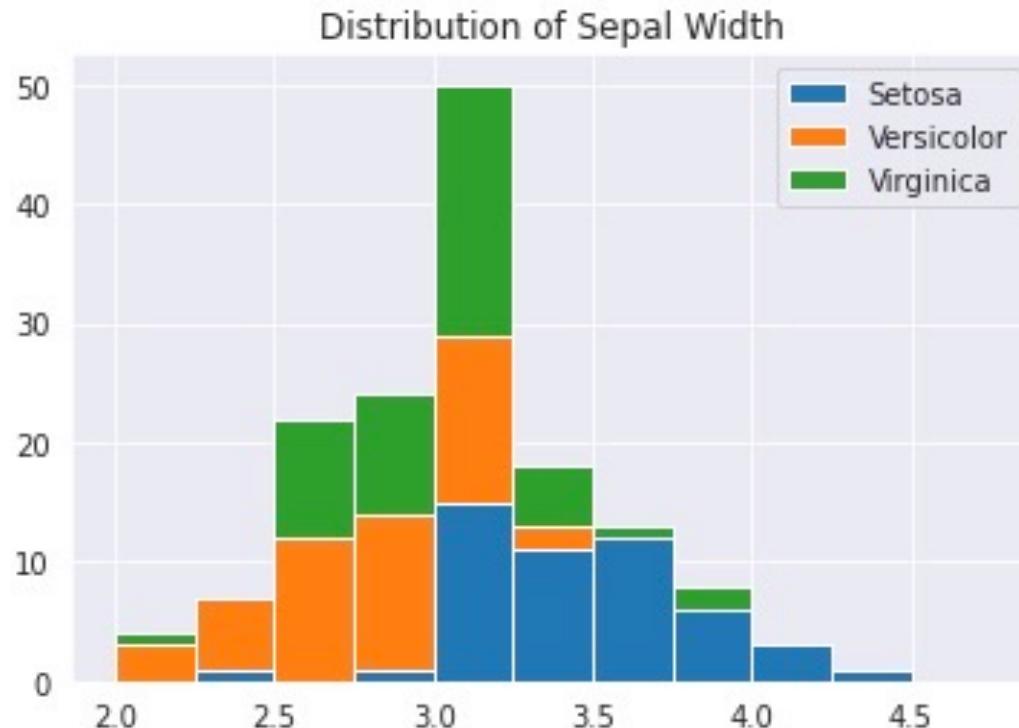
```
plt.hist (setosa_df.sepal_width, alpha=0.4, bins=np.arange(2, 5, 0.25));
plt.hist (versicolor_df.sepal_width, alpha=0.4, bins=np.arange(2, 5, 0.25));
```



Multiple histograms using opacity

Python: Histograms

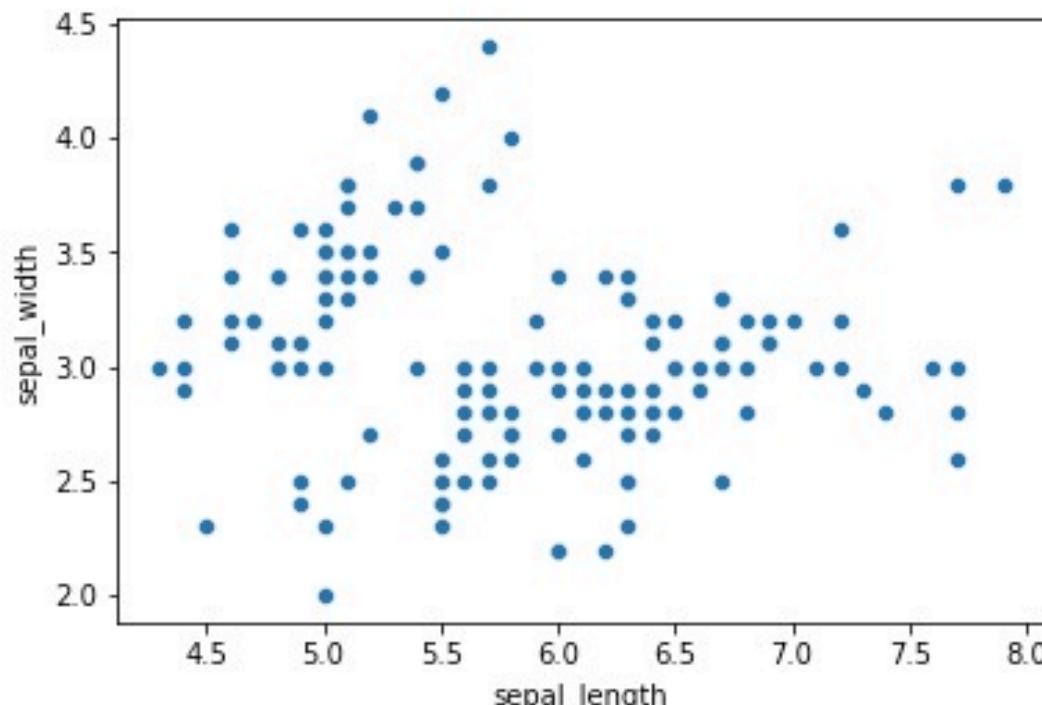
```
plt.title('Distribution of Sepal Width')
plt.hist([setosa_df.sepal_width, versicolor_df.sepal_width,
          virginica_df.sepal_width],
         bins=np.arange(2, 5, 0.25),
         stacked=True);
plt.legend(['Setosa', 'Versicolor', 'Virginica']);
```



Stacked histograms

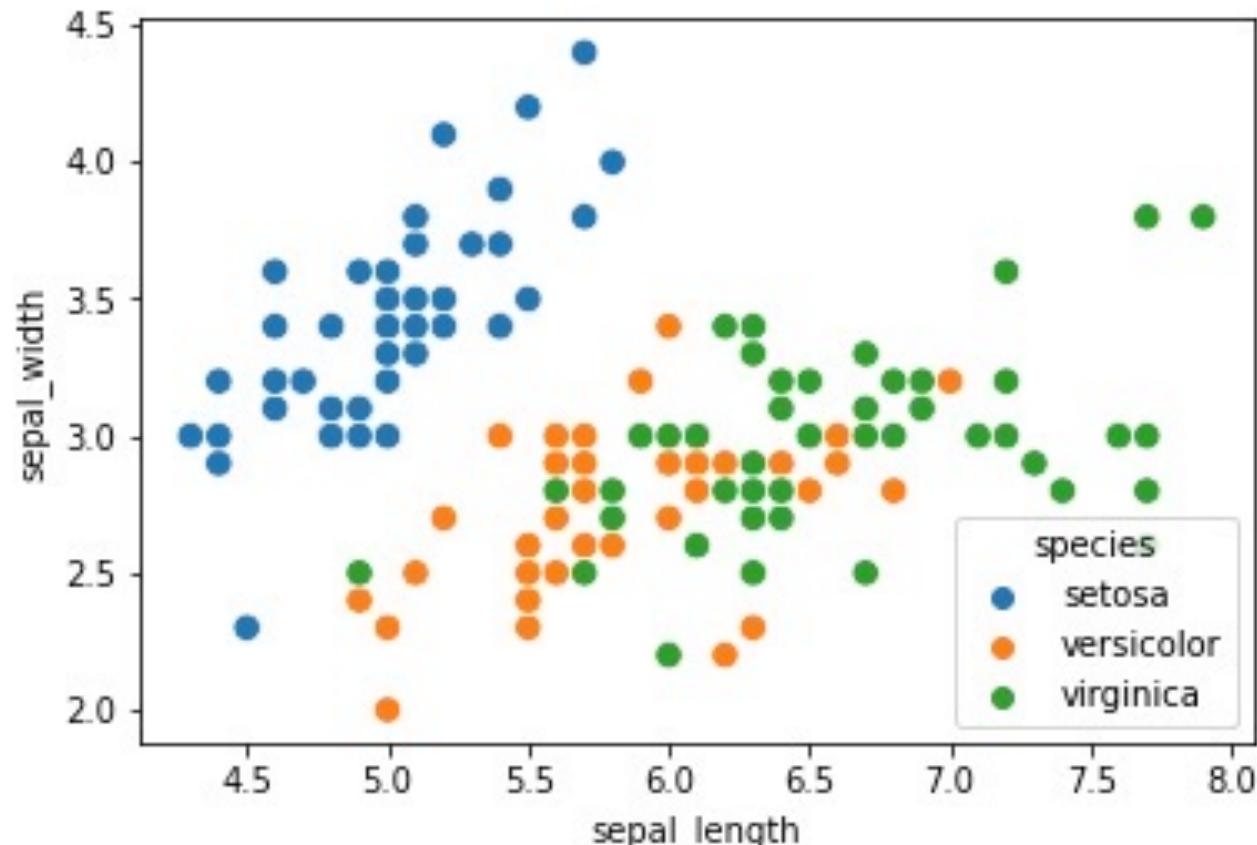
Python: Scatter Plots

```
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

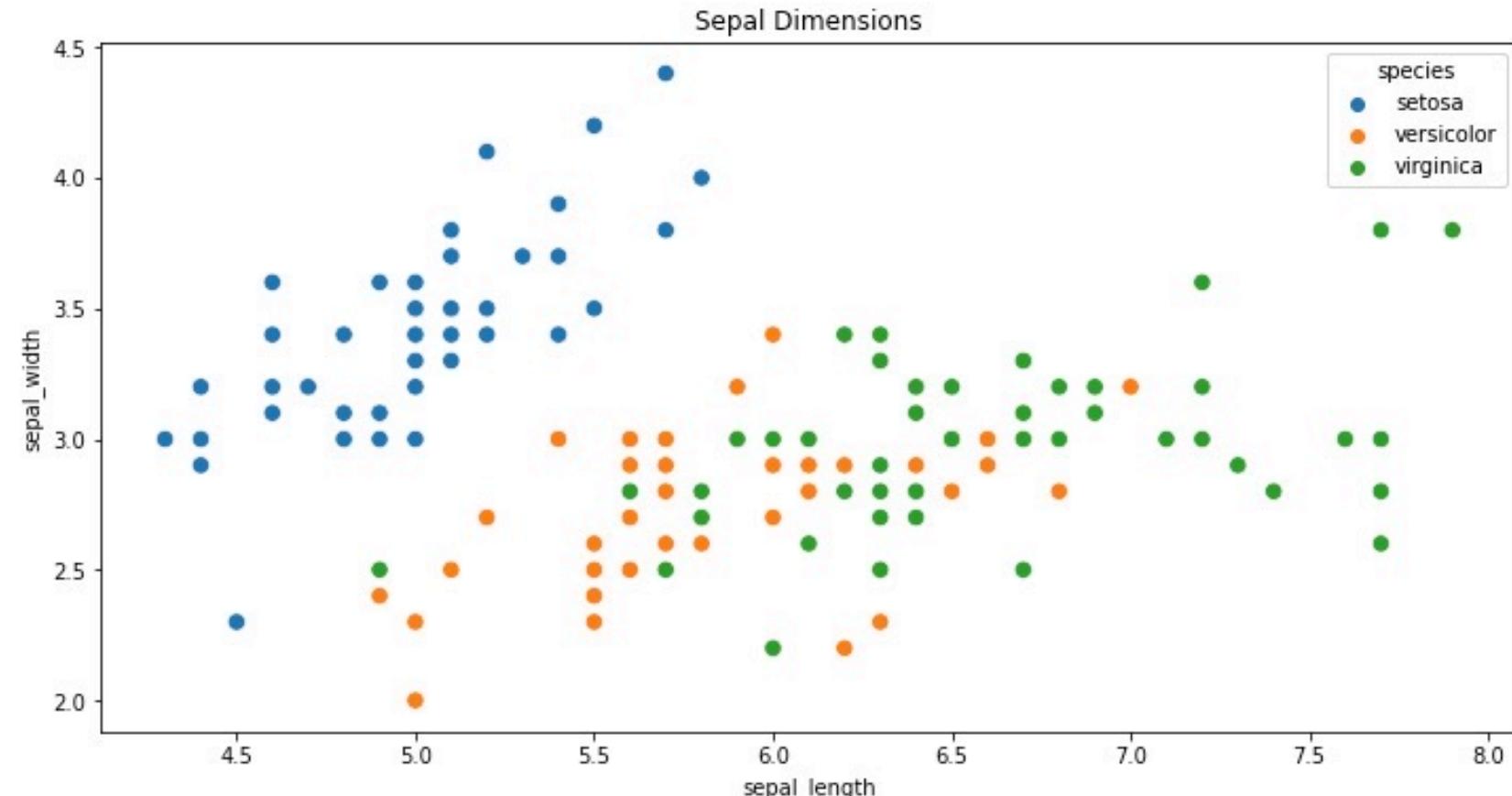
```
sns.scatterplot(x=flowers_df.sepal_length,  
                 y=flowers_df.sepal_width,  
                 hue=flowers_df.species, s=70);
```



Add color to distinguish species

Python: Scatter Plots

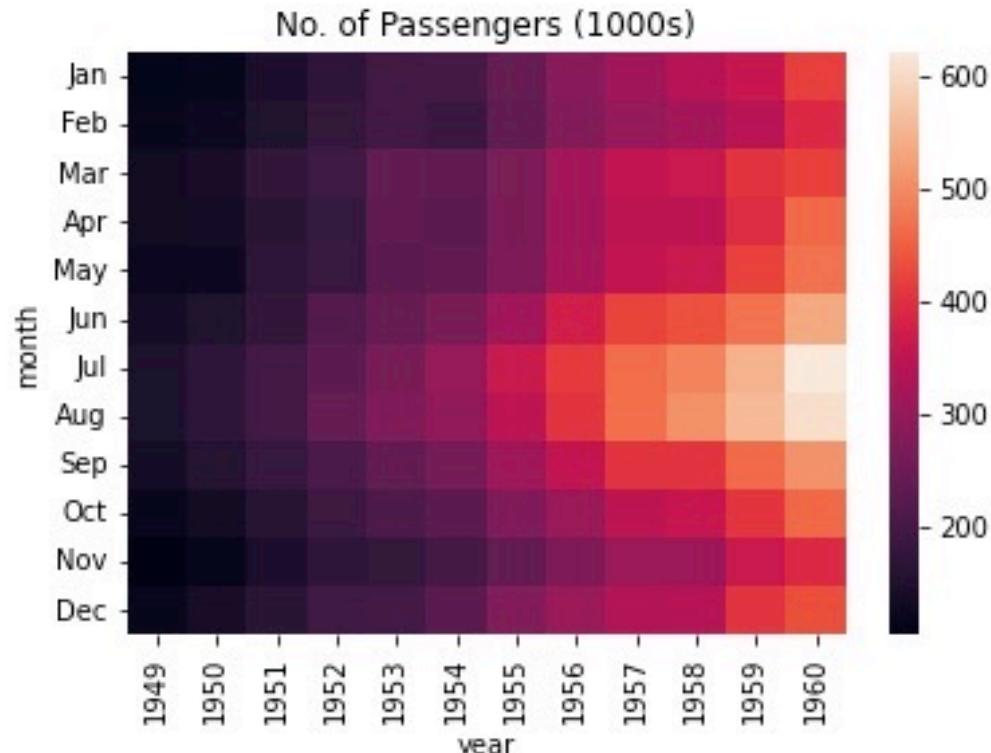
```
plt.figure(figsize=(12, 6))
plt.title('Sepal Dimensions')
sns.scatterplot(x=flowers_df.sepal_length,
                 y=flowers_df.sepal_width,
                 hue=flowers_df.species, s=70);
```



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'>
```



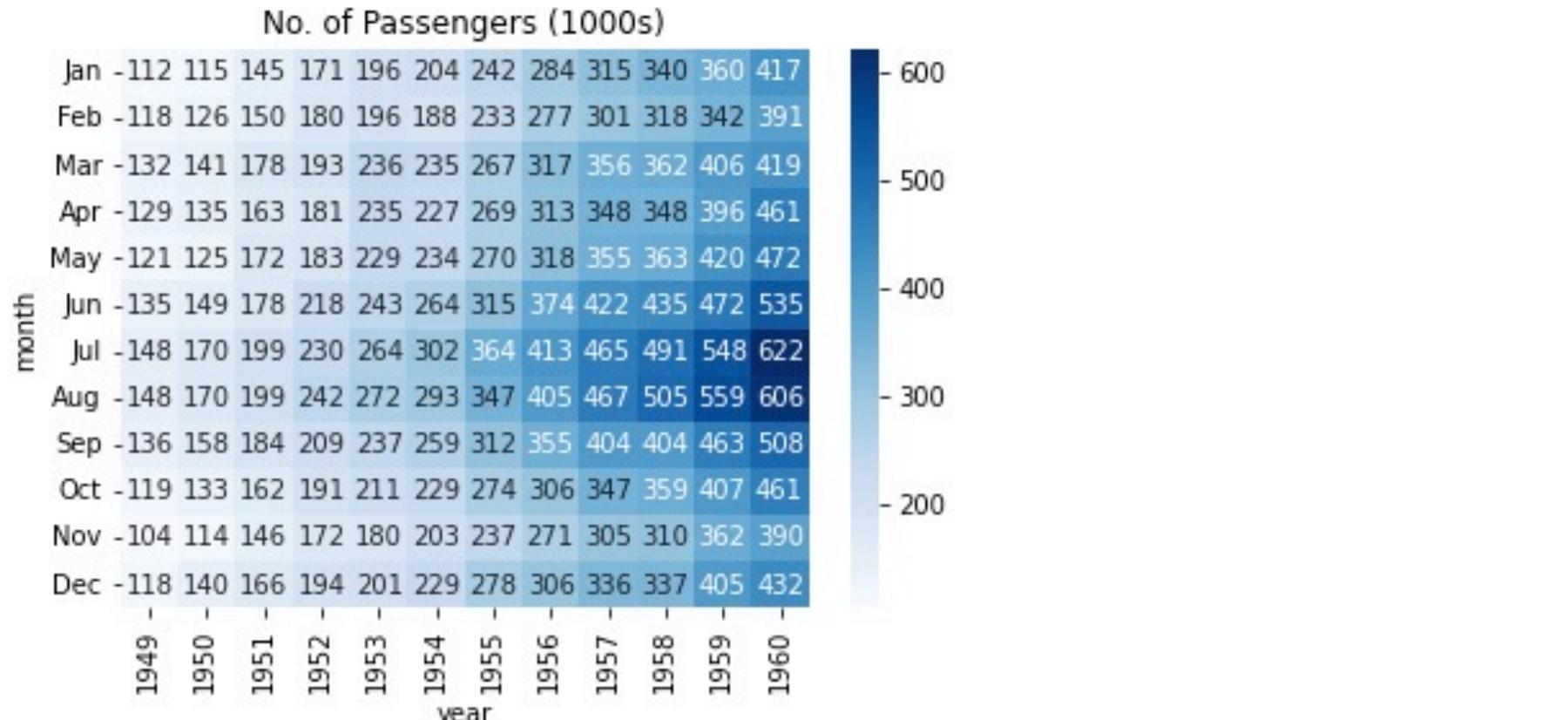
Brighter colors mean more airline traffic

Two conclusions:

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

Python: Heat Maps

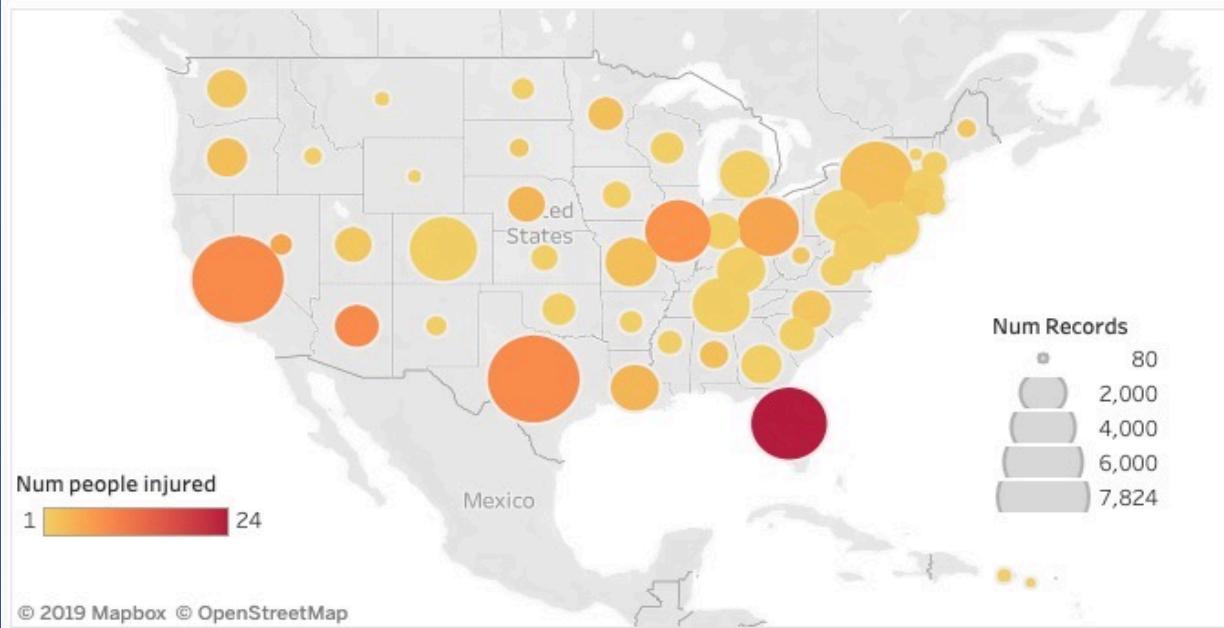
```
plt.title("No. of Passengers (1000s)")  
sns.heatmap(flights_df, fmt="d", annot=True, cmap='Blues')  
  
<AxesSubplot:title={'center':'No. of Passengers (1000s)'}, xlabel='year',
```



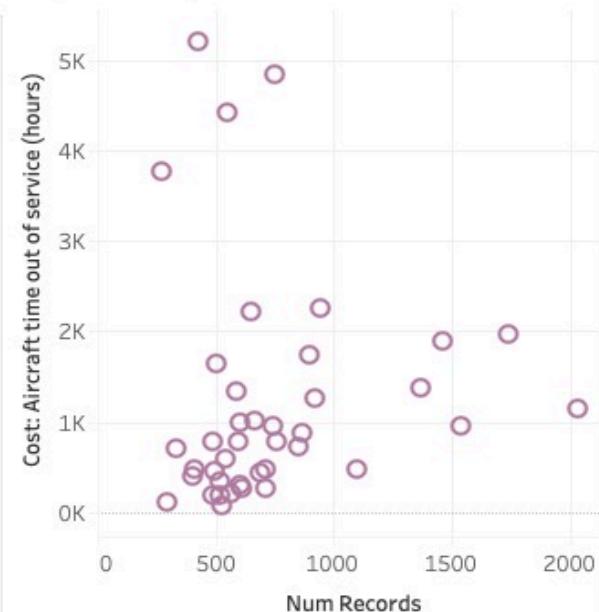


+tableau

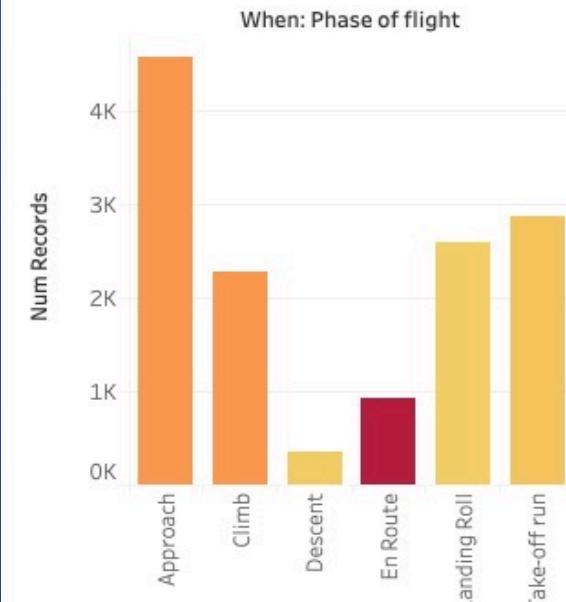
Bird Strikes: 2000 to 2011



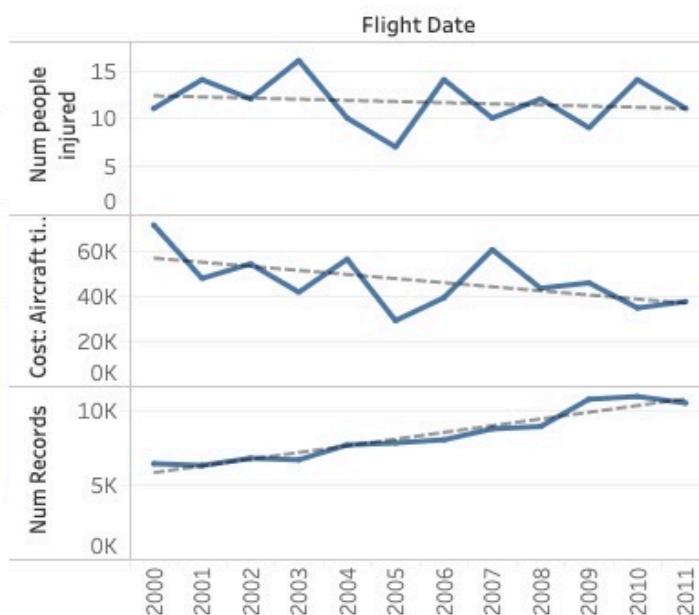
Top 40 Airports



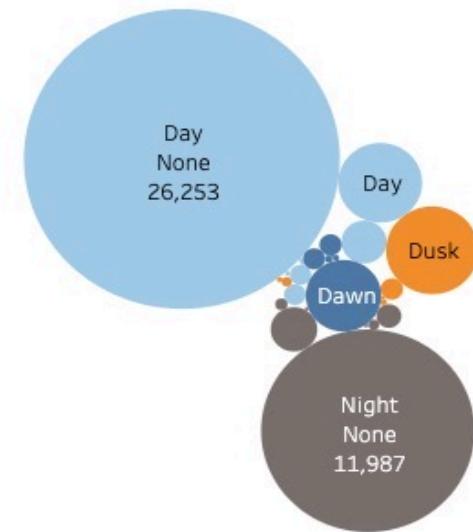
Phase of Flight



Trends



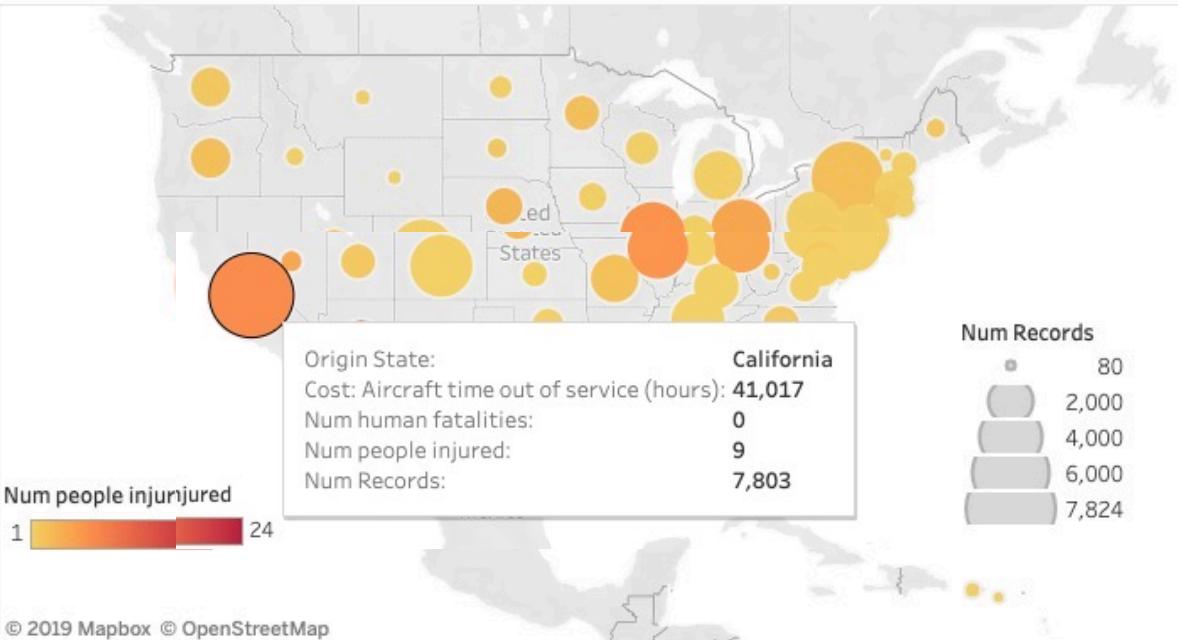
Conditions



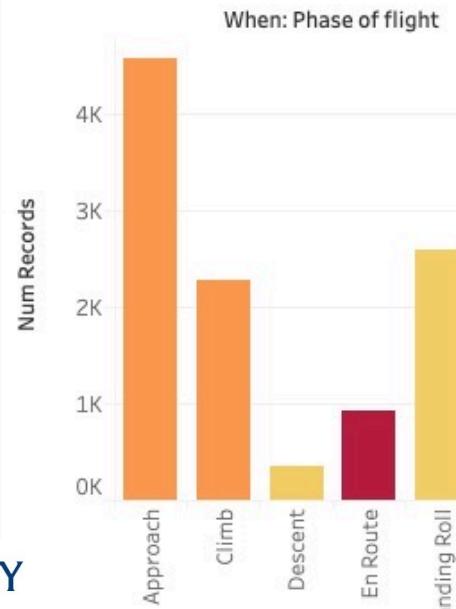
Underlying Details

Hover over data points

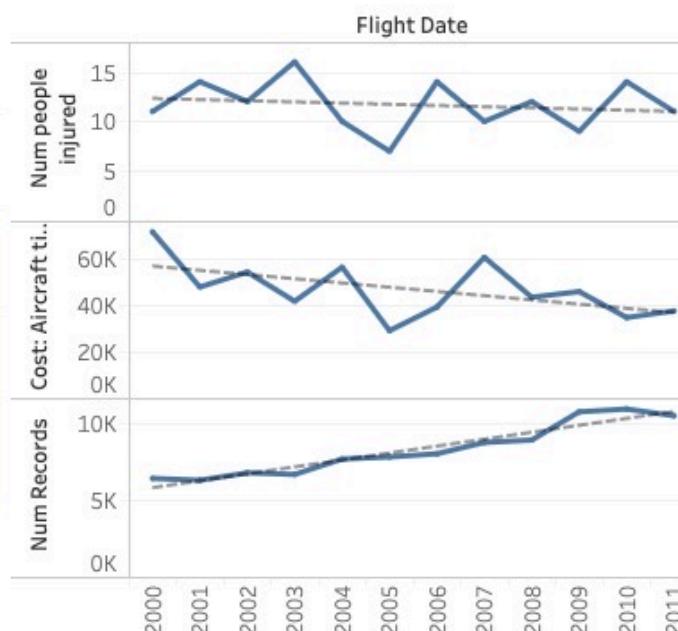
Bird Strikes: 2000 to 2011



Phase of Flight



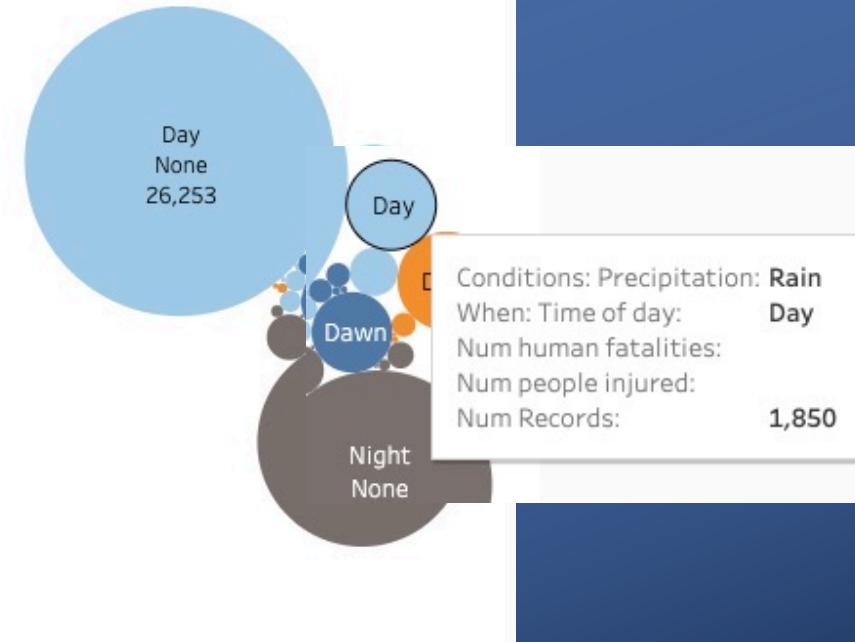
Trends



Top 40 Airports



Conditions





+tableau

Details and Statistics

Hover over a data point to see more information

Trends

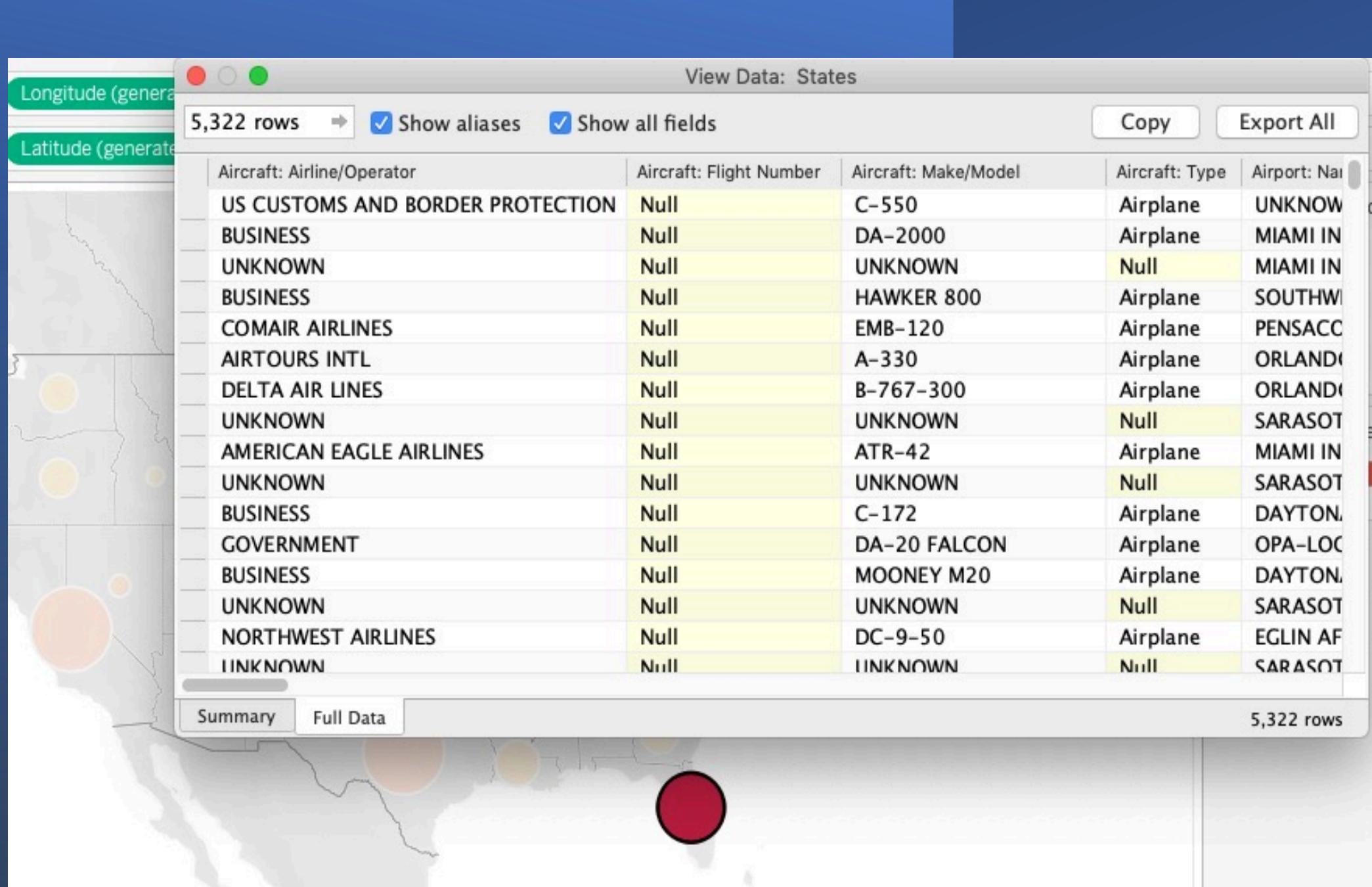




+tableau

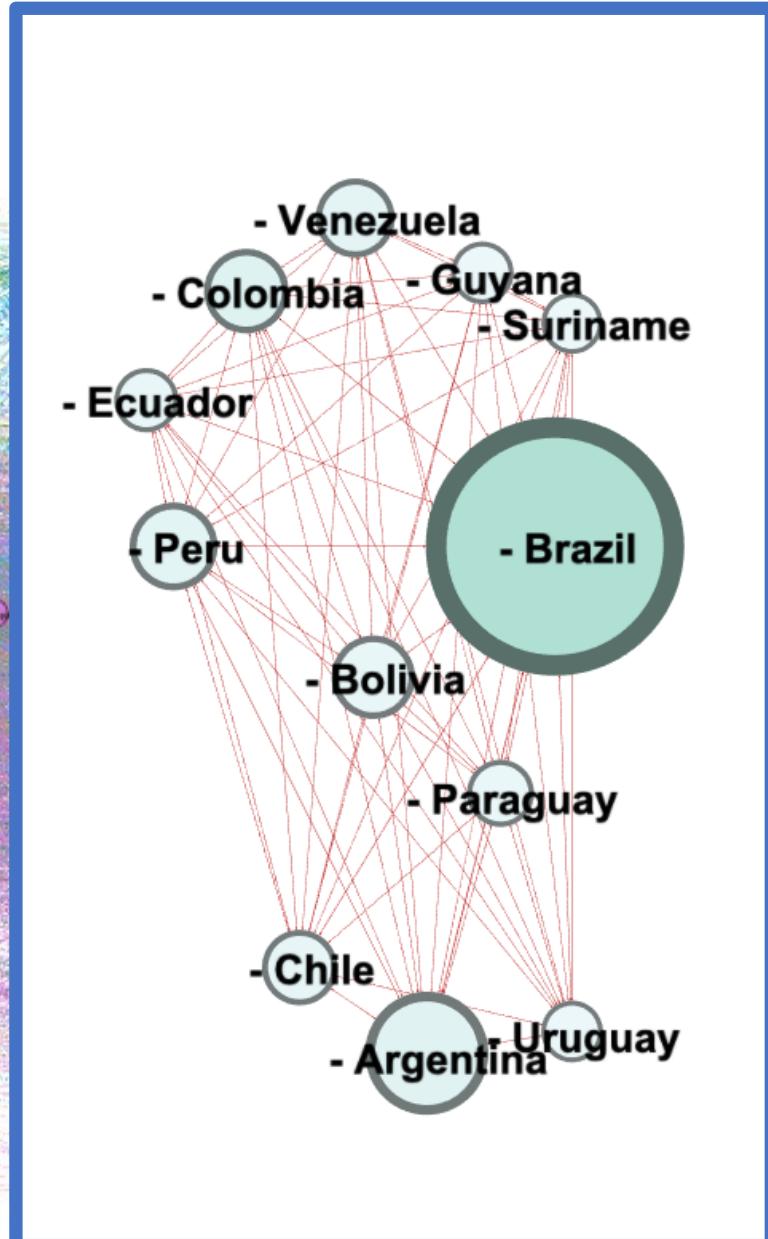
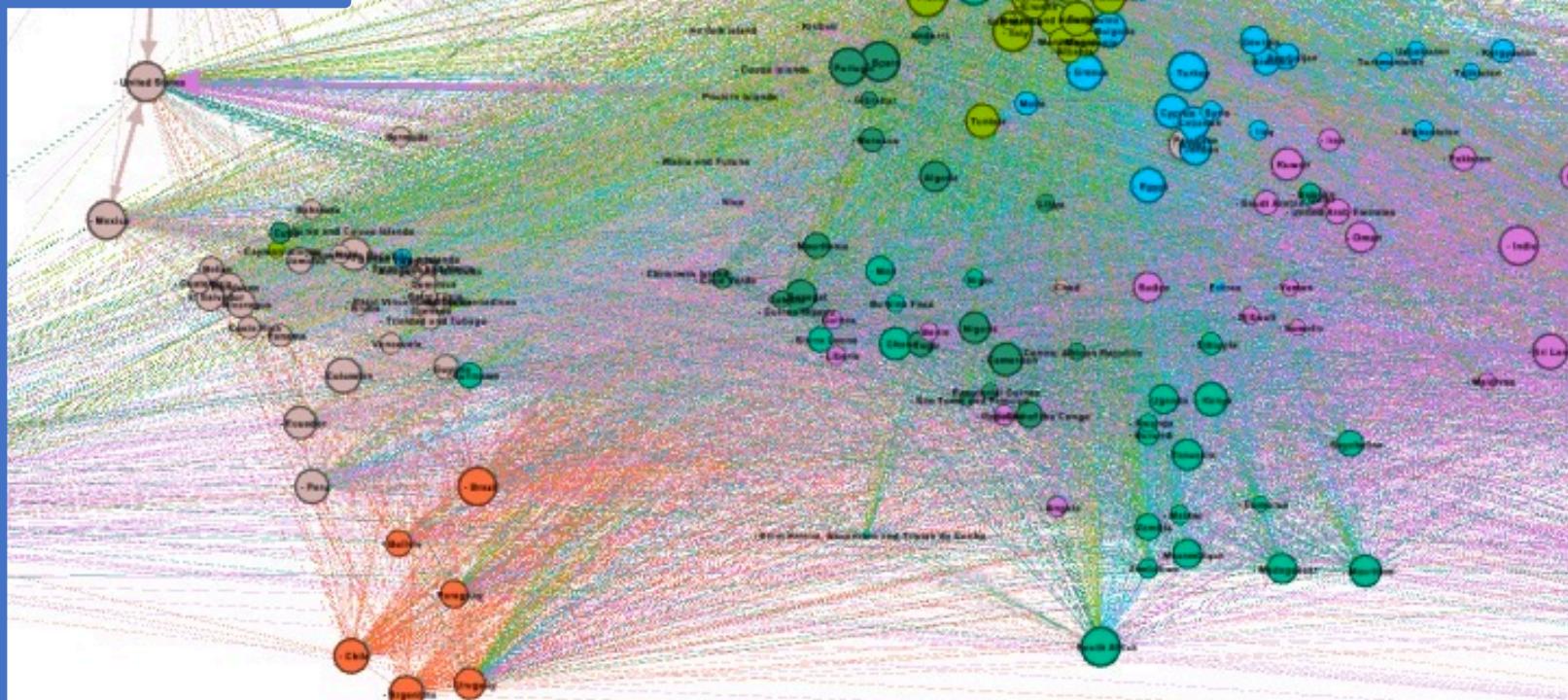
Show Every Record for Florida

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

