Visualization

Topics

- 1. Matplotlib core framework
- 2. Pandas plot()
- 3. Seaborn statistical visualization
- 4. (not covered) Grammar of graphics (ggplot2 see plotnine)
- 5. (not covered) Interactive plotting

Resources

- 1. Ch 9 in Python for Data Analysis, 2nd Ed, Wes McKinney (UCalgary library and https://github.com/wesm/pydata-book (https://github.com/wesm/pydata-book))
- Ch 4 in Python Data Science Handbook, Jake VanderPlas (Ucalgary library and https://github.com/jakevdp/PythonDataScienceHandbook)
 (https://github.com/jakevdp/PythonDataScienceHandbook)
- Fundamentals of Data Visualization, Claus O. Wilke (Ucalgary library and https://serialmentor.com/dataviz/index.html (https://serialmentor.com/dataviz/index.html (<a
- Overview by Jake VanderPlas https://www.youtube.com/watch?v=FytuB8nFHPQ
 (https://www.youtube.com/watch?v=FytuB8nFHPQ)

Matplotlib

Matplotlib is a Python 2D plotting library which produces publication quality figures in a variety of hardcopy formats and interactive environments across platforms. Matplotlib tries to make easy things easy and hard things possible. For simple plotting the pyplot module provides a MATLAB-like interface

https://matplotlib.org (https://matplotlib.org)

Importing matplotlib looks like this

Two interfaces

There are two ways to interact with Matplot lib: a Matlab style and an object oriented style interface.

See Ch 4 in Python Data Science Handbook, Jake VanderPlas

- Two Interfaces for the Price of One, pp. 222
- Matplotlib Gotchas, pp. 232

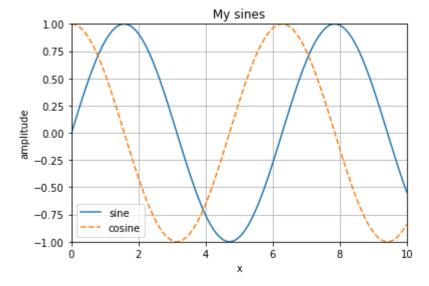
Matlab style interface

```
In [30]:
             \mid x = \text{np.linspace}(0, 10, 100)
                plt.plot(x, np.sin(x), '-')
                plt.plot(x, np.cos(x), '--');
                   1.00
                   0.75
                   0.50
                   0.25
                   0.00
                 -0.25
                 -0.50
                 -0.75
                 -1.00
                                   ż
                                                       6
                                                                8
                         Ò
                                                                          10
```

Adding decorations to the plot is done by repeatatly calling functions on the imported plt module. All calls within the cell will be applied to the current figure and axes.

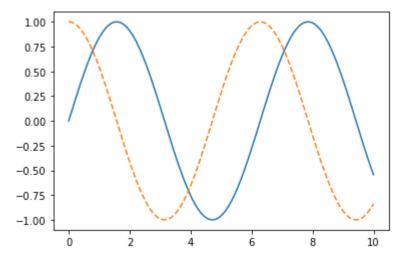
```
In [31]: Plt.plot(x, np.sin(x), '-', label='sine')
plt.plot(x, np.cos(x), '--', label='cosine')

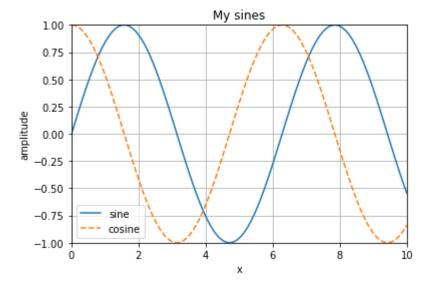
plt.xlim([0, 10])
plt.ylim([-1, 1])
plt.xlabel('x')
plt.ylabel('amplitude')
plt.title('My sines')
plt.grid()
plt.legend();
```



Object oriented interface

With this interface, you first create a figure and an axes object, then call their methods to change the plot.





Save to file

With the figure object at hand, we can save to file

'ls' is not recognized as an internal or external command, operable program or batch file.

Plotting with pandas

We use the standard convention for referencing the matplotlib API ... We provide the basics in pandas to easily create decent looking plots.

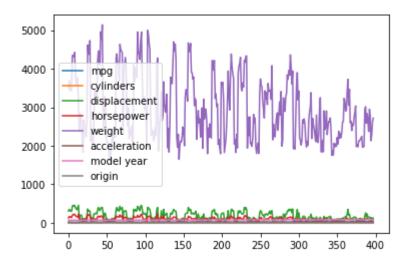
https://pandas.pydata.org/pandas-docs/stable/user_guide/visualization.html (https://pandas.pydata.org/pandas-docs/stable/user_guide/visualization.html)

Let's load the heart attack dataset

Plotting all columns, works, but does not provide a lot of insight.

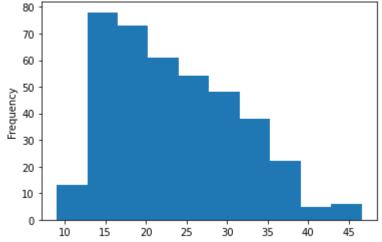
```
In [37]: ► data.plot()
```

Out[37]: <AxesSubplot:>



Let's look at the age distribution (a histogram)



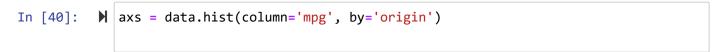


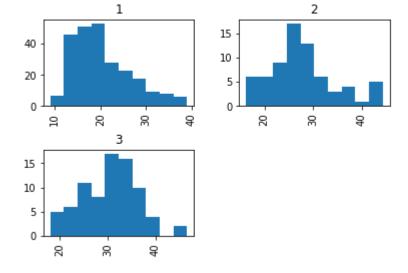
How many male and female samples do we have?

Name: origin, dtype: int64

Notice that we accessed the gender column with dot notation. This can be done whenever the column name is 'nice' enough to be a python variable name.

Do we have similar ages in females and males? Plotting two histograms for each gender side beside directly form the dataframe:

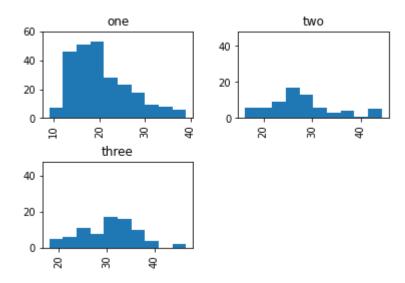




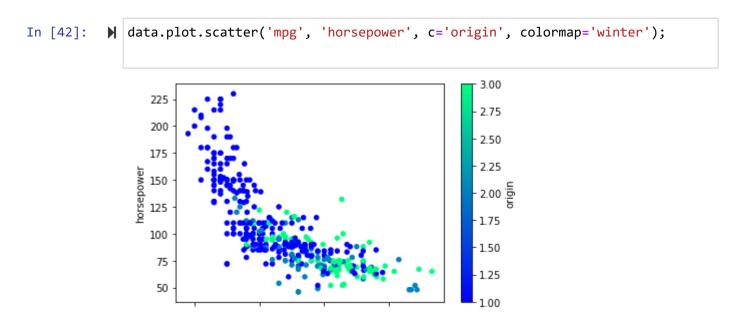
To format this plot, we can work on the axes (array) that is returned by the plot call. We use Matplotlib object oriented interface methods to do this

```
In [41]: N axs = data.hist(column='mpg', by='origin')
axs[0, 0].set(title='one', ylim=[0, 60])
axs[0, 1].set(title='two', ylim=[0, 48])
axs[1, 0].set(title='three', ylim=[0, 48])
```

Out[41]: [Text(0.5, 1.0, 'three'), (0.0, 48.0)]



Is age and blood pressure correlated? Maybe it is different for females and males? Let's have a look with a scatter plot.

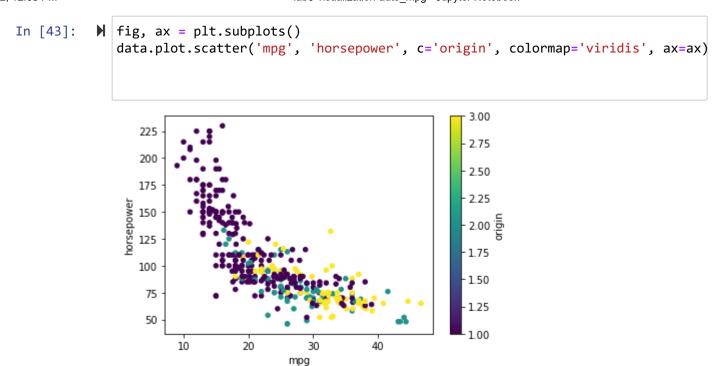


According to:

https://stackoverflow.com/questions/43578976/pandas-missing-x-tick-labels (https://stackoverflow.com/questions/43578976/pandas-missing-x-tick-labels)

the missing x-labels are a pandas bug.

Workaraound is to create axes prior to calling plot

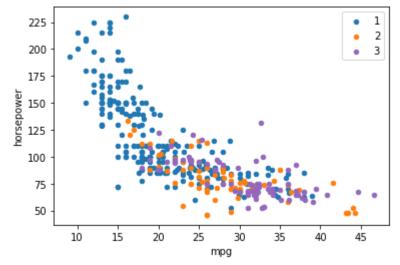


It is a bit annoying that there is a colorbar, we know gender is categorical.

One way to avoid the colorbar is to loop over the categories and assign colors based on the category.

See: https://stackoverflow.com/questions/26139423/plot-different-categorical-levels-using-matplotlib)

```
In [44]: N
colors = {1: 'tab:blue', 2: 'tab:orange', 3:'tab:purple'}
fig, ax = plt.subplots()
for key, group in data.groupby(by='origin'):
    group.plot.scatter('mpg', 'horsepower', c=colors[key], label=key, ax=ax);
```



Seaborn

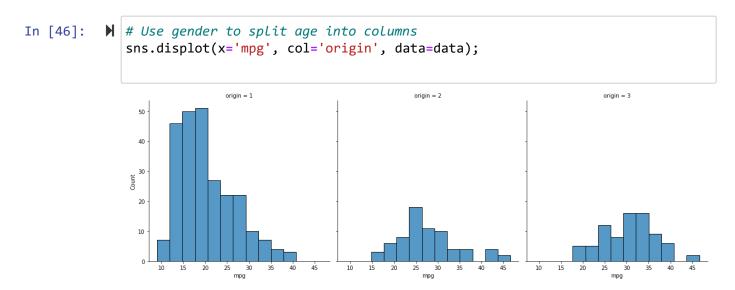
Seaborn is a Python data visualization library based on matplotlib. It provides a high-level interface for drawing attractive and informative statistical graphics.

http://seaborn.pydata.org/index.html (http://seaborn.pydata.org/index.html)

Seaborn is usually imported as sns

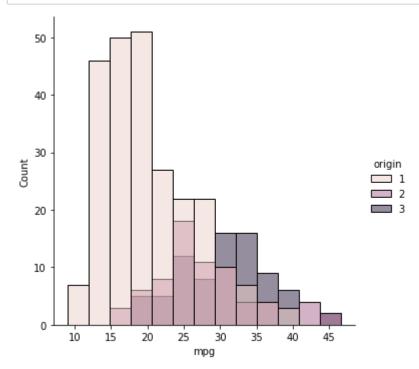
```
In [45]: ▶ import seaborn as sns
```

Let's re-create the histograms by gender with seaborn with the figure level displot() function.



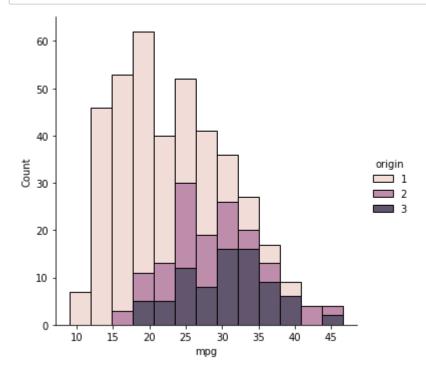
We can display the counts in the same plot, one on top of the other.

```
In [47]: # Use gender to color (hue) in the same plot
sns.displot(x='mpg', hue='origin', data=data);
```

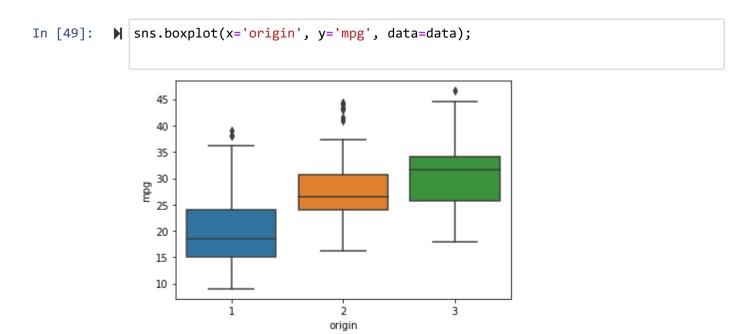


To have an idea of the split between male and female, we can stack the counts, adding up to total.

In [48]: ▶ sns.displot(x='mpg', hue='origin', data=data, multiple='stack');



We can look at the differences in ages with a boxplot too



Let's re-create the scatter plot to see if age and blood pressure are correlated by gender.

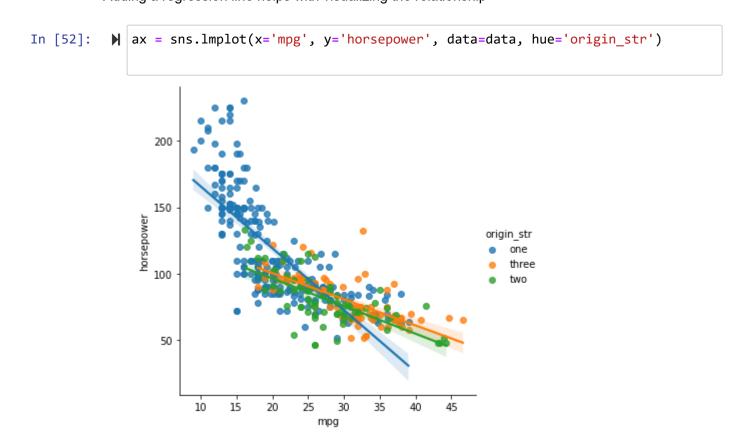
To make the legend show strings we will create a gender string column with female and male strings rather than 0 and 1.

```
In [50]:
            M data['origin_str'] = data['origin'].replace([1, 2, 3], ['one', 'two', 'three']

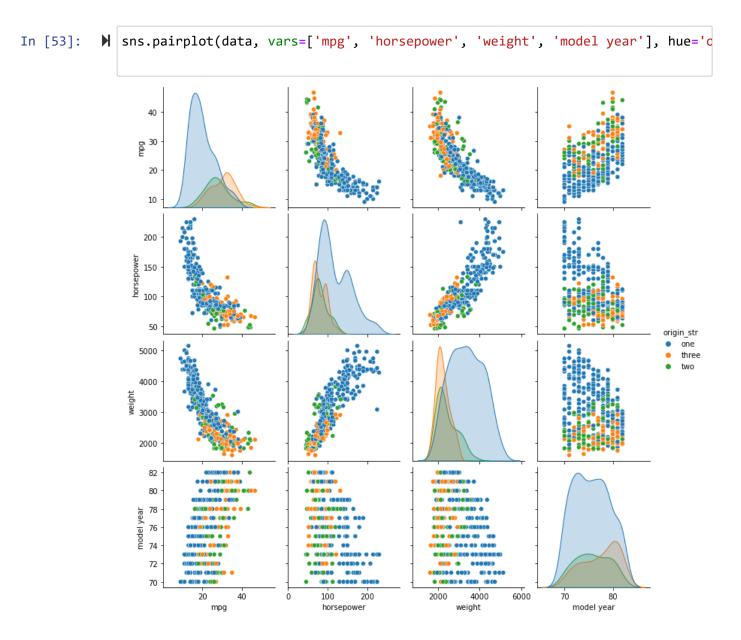
    | ax = sns.scatterplot(x='mpg', y='horsepower', data=data, hue='origin_str')

In [51]:
                   225
                                                                  origin_str
                                                                      one
                   200
                                                                      three
                                                                      two
                   175
                 horsepower
                   150
                   125
                   100
                    75
                    50
                          10
                                15
                                       20
                                                    30
                                                          35
                                             25
                                                mpg
```

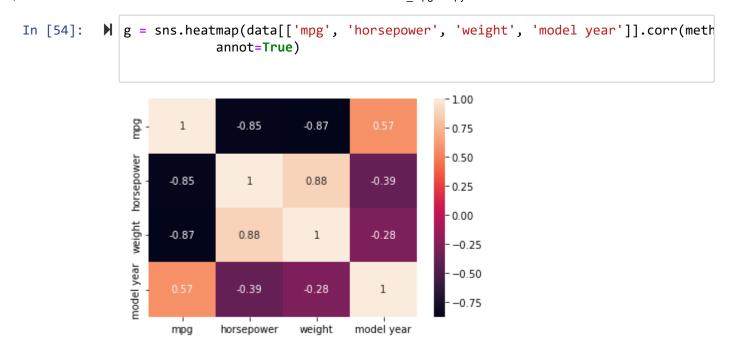
Adding a regression line helps with visualizing the relationship



Maybe there are other correlations in the data set. Pairplot is a great way to get an overview



As an alternative, we can visualize the correlation matrix as a heatmap



There are nice tutorials on the Seaborn website, be sure to check these out.

```
In []: M
```