

# Data visualization with Matplotlib and Seaborn

Inteligencia Artificial en los Sistemas de Control Autónomo  
Máster en Ciencia y Tecnología desde el Espacio

Departamento de Automática

## Objectives

1. Motivate the importance of data visualization
2. Avoid some common mistakes in data visualization
3. Choose the proper visualization technique
4. Overview Matplotlib
5. Introduce Seaborn

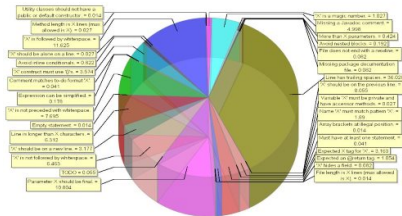
## Bibliography

Jake VanderPlas. *Python Data Science Handbook*. Chapters 4. O'Reilly. (Link).

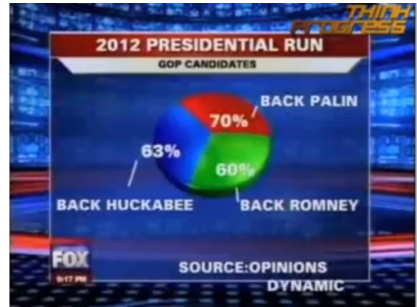
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# Visualization examples (I)

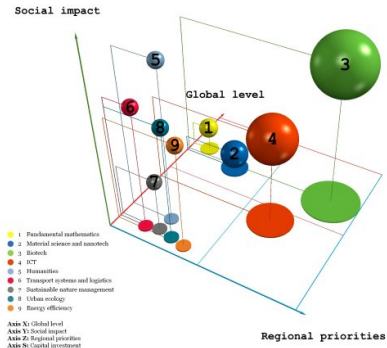


(Source)

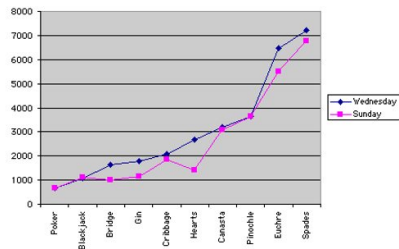


(Source)

# Visualization examples (II)

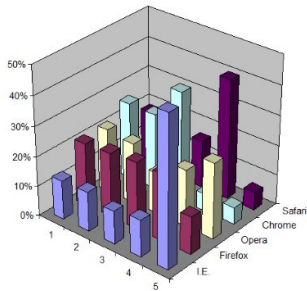


(Source)

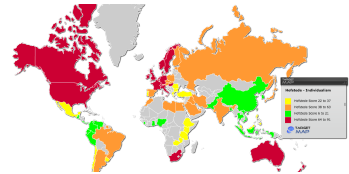


(Source)

# Visualization examples (III)



(Source)

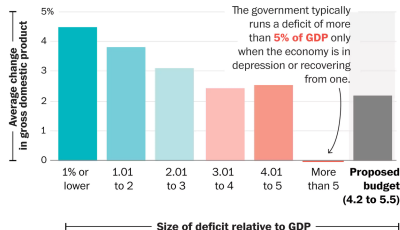


(Source)

# Visualization examples (IV)

## Strange time for a stimulus

What annual **economic growth** averaged under various deficit-to-GDP ratios, since 1967



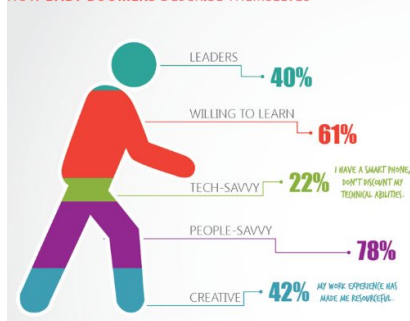
Notes: To capture the environment in which the budget was set, deficit-to-GDP ratios are compared with the economic climate of the prior fiscal year. GDP growth is adjusted for inflation and seasonality. Indicators for the current budget are based on the average of available data in fiscal 2017 and 2018 years. Fiscal years end in September.

Sources: Commerce Department (GDP); Congressional Budget Office (historical deficit); Committee for a Responsible Federal Budget (deficit forecasts, budget changes)

THE WASHINGTON POST

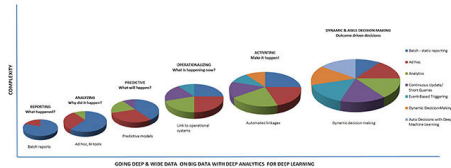
(Source)

## HOW BABY BOOMERS DESCRIBE THEMSELVES

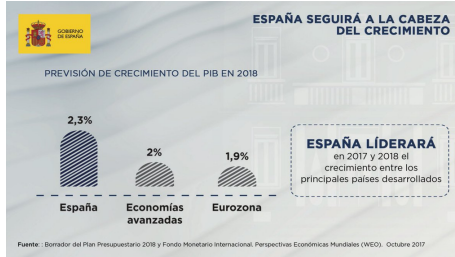


(Source)

# Visualization examples (V)



(Source)



(Source)



# Motivation (I)

## Efficient data visualization tips

- **Define your story**
- The chart must tell the story
- Don't distract from your story (with irrelevant data or visual elements)
- One story, one chart
- Put the story comprehension in first term
- Better several simple charts than one complex chart
- Choose colors wisely (color scale or high contrast)
- Elements order must support the story (legend, bars, etc)
- There is life beyond pies and bars
- Keep it simple, stupid!

# Motivation (II)

Know your data

- Categorical or numerical
- Number of dimensions to represent (1D, 2D, 3D, more dimensions)

Can you use other representation?

- Chart better than table? ...
- ... that depends

What do you want to represent?

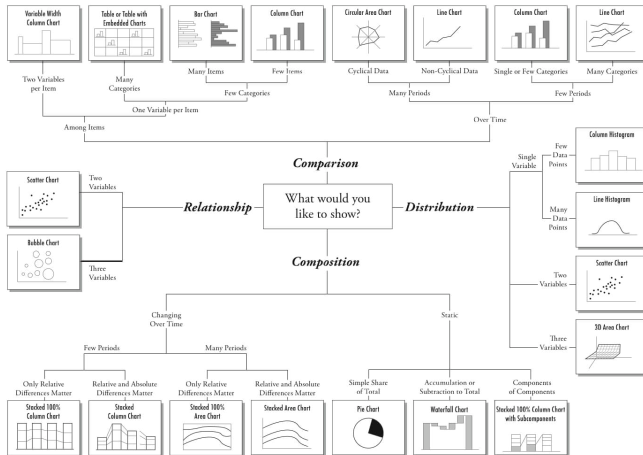
- Distribution, relationship, comparison or composition

Look for templates: (<https://python-graph-gallery.com/>)

# Motivation (III)

## Chart Suggestions—A Thought- Starter

www.ExtremePresentation.com  
© 2009 A. Abela — a.x.abela@gmail.com



(Source)

(Alternative resource)

# Matplotlib (I)

Matplotlib is a Python package

- Based on NumPy
- Imitates Matlab

Three operation modes

- Scripts.  
Must use `plt.show()` to enter event loop. Use it once!
- IPython shell.  
Must use `%matplotlib`
- IPython notebook. Two modes
  - `%matplotlib inline`
  - `%matplotlib notebook`

## Convention

```
import matplotlib as mpl
import matplotlib.pyplot as plt
```

## myplot.py

```
import matplotlib.pyplot as plt
import numpy as np

x = np.linspace(0, 10, 100)

plt.plot(x, np.sin(x))
plt.plot(x, np.cos(x))

plt.show()
```

# Matplotlib (II)

Matplotlib comes with two interfaces

- Matlab-like. Old-fashioned function-oriented API.
- Object-oriented. Object-oriented and more powerful API.

## Matlab API

```
plt.figure() # create a plot

# create the first of two panels
# and set current axis
plt.subplot(2, 1, 1) # (rows,
# columns, panel number)
plt.plot(x, np.sin(x))

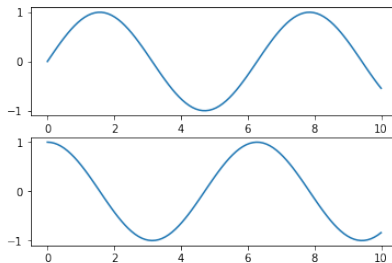
# create the second panel and
# set current axis
plt.subplot(2, 1, 2)
plt.plot(x, np.cos(x));
```

## OO API

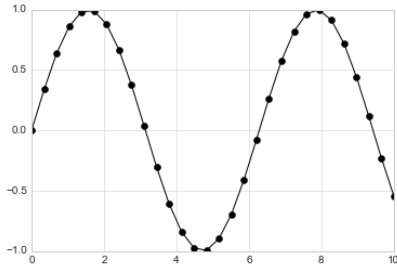
```
# First create a grid of plots
# ax will be an array of two
# Axes objects
fig, ax = plt.subplots(2)

# Call plot() method on the
# appropriate object
ax[0].plot(x, np.sin(x))
ax[1].plot(x, np.cos(x));
```

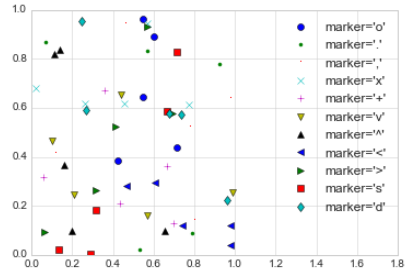
# Matplotlib (III)



# Matplotlib (IV)

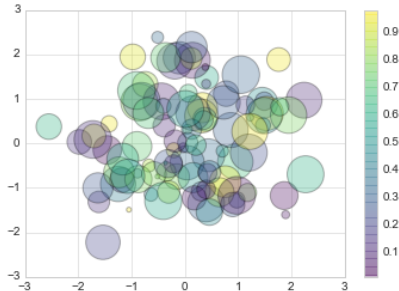


```
plt.plot(x, np.sin(x), '-ok',
         color='black')
```



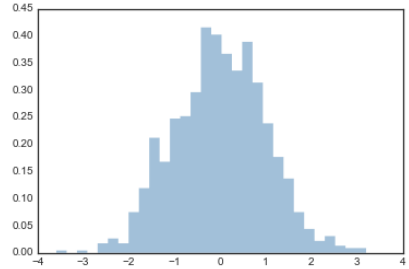
```
for marker in ['o', '.', ',', 'x', '+', 'v', '^', '<', '>', 's', 'd']:
    plt.plot(rng.rand(5), rng.rand(5), marker,
             label="marker = '{0}'".format(marker))
plt.legend(numpoints=1)
plt.xlim(0, 1.8);
```

# Matplotlib (V)



```
rng = np.random.RandomState(0)
x = rng.randn(100)
y = rng.randn(100)
colors = rng.rand(100)
sizes = 1000 * rng.rand(100)

plt.scatter(x, y, c=colors, s=sizes, alpha=0.3,
            cmap='viridis')
plt.colorbar(); # show color scale
```



```
data = np.random.randn(1000)

plt.hist(data, bins=30, normed=True, alpha=0.5,
         histtype='stepfilled', color='steelblue',
         edgecolor='none');
```

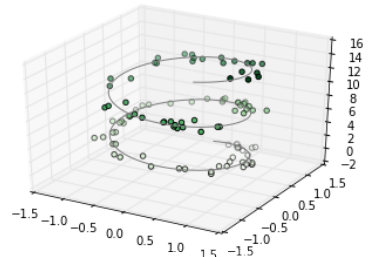


# Matplotlib (VI)

```
ax = plt.axes(projection='3d')

# Data for a three-dimensional line
zline = np.linspace(0, 15, 1000)
xline = np.sin(zline)
yline = np.cos(zline)
ax.plot3D(xline, yline, zline, 'gray')

# Data for three-dimensional scattered points
zdata = 15 * np.random.random(100)
xdata = np.sin(zdata) + 0.1 * np.random.randn(100)
ydata = np.cos(zdata) + 0.1 * np.random.randn(100)
ax.scatter3D(xdata, ydata, zdata, c=zdata, cmap='Greens');
```



# Matplotlib

## Matplotlib notebook

Matplotlib notebook

([Link to notebook](#))

# Seaborn (I)

Seaborn is a modern data-visualization Python package

- Based on matplotlib
- ... it uses matplotlib indeed
- Pandas-aware
- High level
- Advanced visualizations
- Easy to use

Still under development!

Convention

```
import seaborn as sns
```

This documentation is for Seaborn  
0.9 or newer

# Seaborn (II)

## Display initialization

- `plt.show()`
- `%matplotlib`

## Style initialization

- Default Seaborn style `sns.set()`
- By default, same style than matplotlib

## Several functions ...

- ... similar parameters

## Parameters

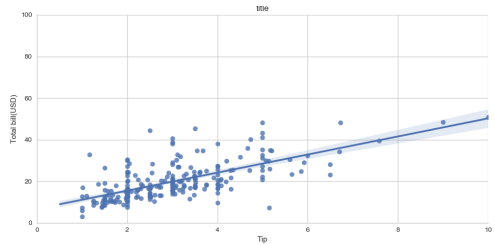
- `x`: Data axis x
- `y`: Data axis Y
- `data`: Dataframe name
- `hue`: Color
- `style`: Style
- `sizes`: Size
- `kind`: Alternate representation

# Seaborn (III)

## Typical Seaborn usage

1. Prepare data
2. Set up aesthetics
3. Plot
4. Customize the plot

```
import matplotlib.pyplot as plt
import seaborn as sns
# Prepare data
tips = sns.load_dataset("tips")
# Set up aesthetics
sns.set_style("whitegrid")
# Plot
g = sns.lmplot(x="tip", y="total_bill", data=tips, aspect=2)
# Plot customization
g = (g.set_axis_labels("Tip", "Total bill (USD)").set(xlim=(0, 10), ylim=(0, 100)))
plt.title("title")
plt.show(g)
```



# Seaborn

## Datasets (I)

Seaborn comes with several dummy datasets

- `sns.load_dataset('name')`

We will use two datasets

- 'iris': The classical iris dataset, numerical
- 'tips': Numeric and categorical variables

### Tips dataset

```
>> tips = sns.load_dataset('tips')
```

```
>> print(tips.head())
```

	total_bill	tip	sex	smoker	day	time	size
0	16.99	1.01	Female	No	Sun	Dinner	2
1	10.34	1.66	Male	No	Sun	Dinner	3
2	21.01	3.50	Male	No	Sun	Dinner	3
3	23.68	3.31	Male	No	Sun	Dinner	2
4	24.59	3.61	Female	No	Sun	Dinner	4

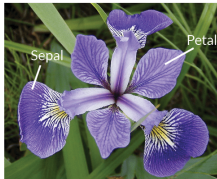
# Seaborn

## Datasets (II)

### Iris dataset

```
»> iris = sns.load_dataset("iris")
»> print(iris.head())
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa



**Iris Versicolor**



**Iris Setosa**

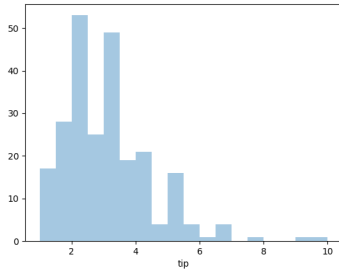


**Iris Virginica**

(Source)

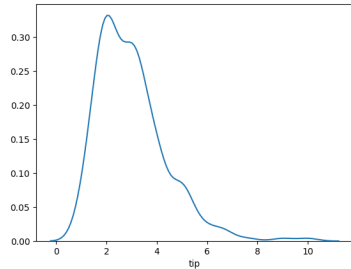
# Seaborn

## Distributions (I)



Histogram

```
sns.distplot(tips['tip'],
             kde=False)
```



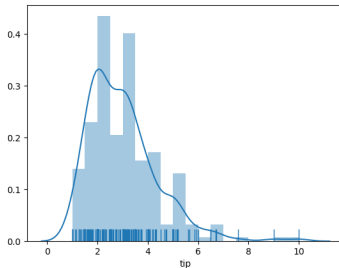
Density plot

```
sns.distplot(tips['tip'],
             hist=False)
```



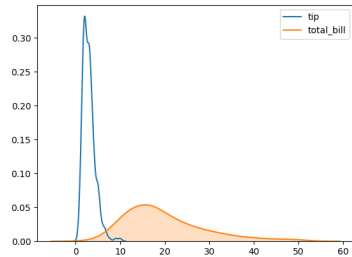
# Seaborn

## Distributions (II)



Histogram + density plot

```
sns.distplot(tips['tip'],
              rug=True)
```



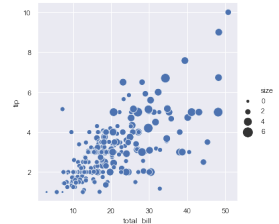
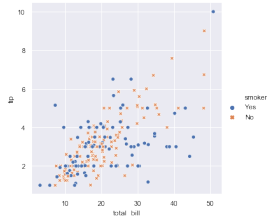
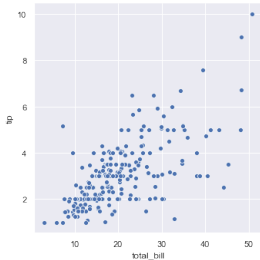
Density plot

```
sns.kdeplot(tips['tip'])
sns.kdeplot(tips['total_bill'],
              shade=True)
```

# Seaborn

## Relationships (I)

### Scatterplots



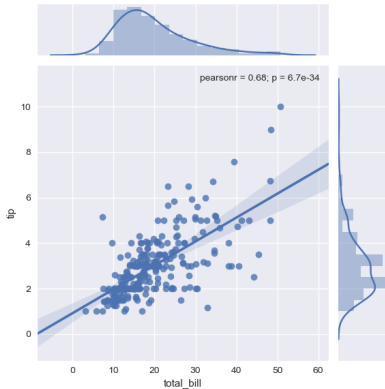
```
sns.relplot(x="total_bill", y="
tip", data=tips)
```

```
sns.relplot(x="total_bill", y="
tip", hue="smoker", style="
smoker", data=tips)
```

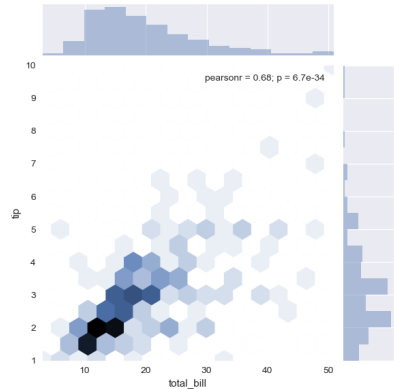
```
sns.relplot(x="total_bill", y="
tip", size="size", sizes
=(15, 200), data=tips);
```

# Seaborn

## Relationships (II)



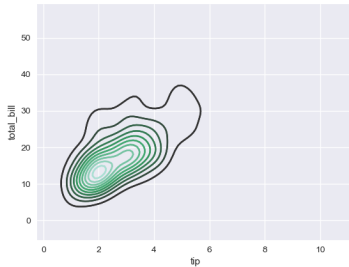
```
sns.jointplot("total_bill", "tip", tips, kind="reg")
```



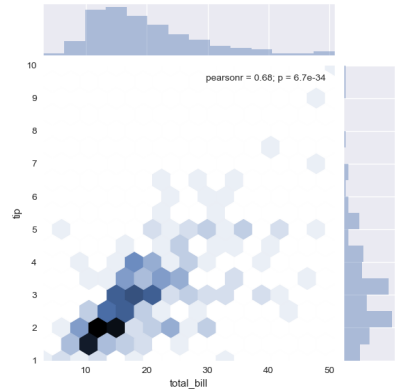
```
sns.jointplot("total_bill", "tip", tips, kind="hex")
```

# Seaborn

## Relationships (III)



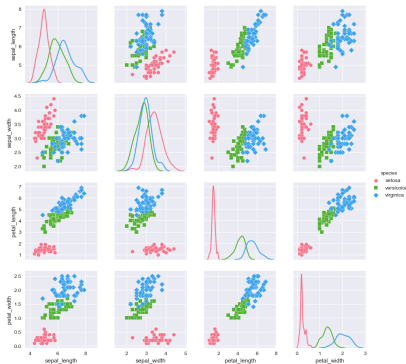
```
sns.kdeplot(tips['tip'], tips['total_bill'])
```



```
sns.jointplot("total_bill", "tip", tips, kind="hex")
```

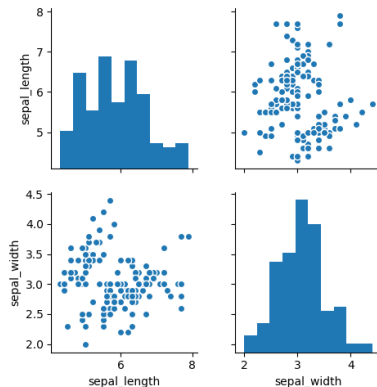
# Seaborn

## Relationships (IV)



### Scatterplot matrix

```
sns.pairplot(iris, hue="species", palette="husl",
             markers=["o", "s", "D"], diag_kind='kde')
```

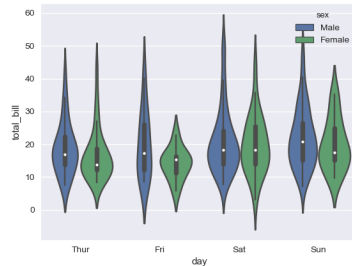
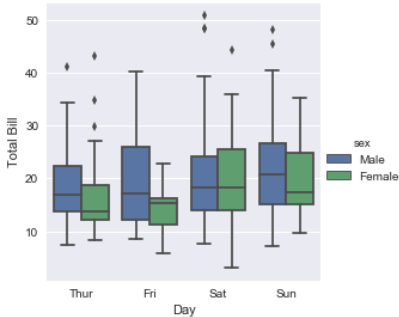


### Scatterplot matrix

```
sns.pairplot(iris, vars=["sepal_length", "sepal_width"])
```

# Seaborn

## Comparisons (I)



### Boxplot

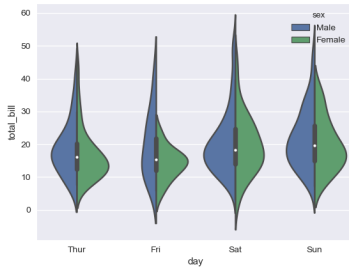
```
with sns.axes_style(style='ticks'):
    g = sns.factorplot("day", "total_bill", "sex",
                      data=tips, kind="box")
    g.set_axis_labels("Day", "Total Bill")
```

### Violin plot

```
sns.violinplot("day", "total_bill", "sex", data=
               tips)
```

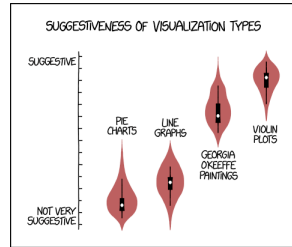
# Seaborn

## Comparisons (II)



### Violin plot

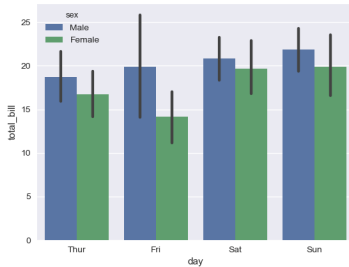
```
sns.violinplot(x="day", y="total_bill", hue="sex",
               data=tips, split=True)
```



(Source)

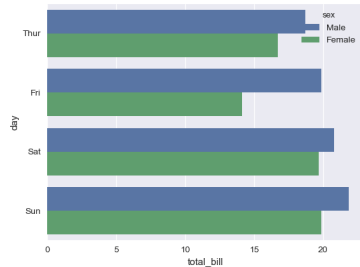
# Seaborn

## Barplots



### Barplot

```
sns.barplot(x="day", y="total_bill", hue="sex",
             data=tips)
```



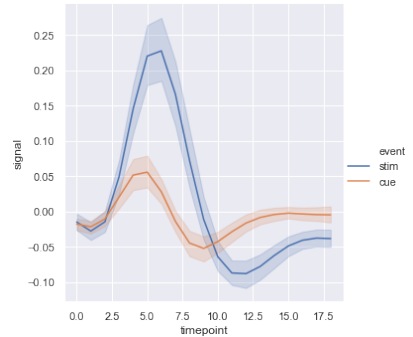
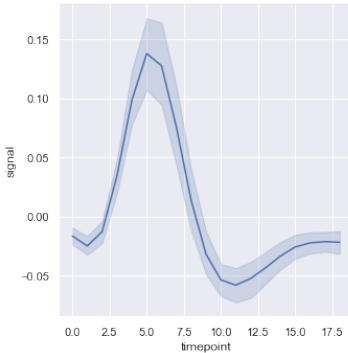
### Barplot

```
sns.barplot(x="total_bill", y="day", hue="sex",
             data=tips, ci=None)
```



# Seaborn

## Continuity

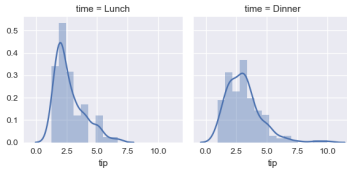


```
fmri = sns.load_dataset("fmri")
sns.relplot(x="timepoint", y="signal", kind="line",
            data=fmri)
```

```
sns.relplot(x="timepoint", y="signal", hue="event",
            kind="line", data=fmri)
```

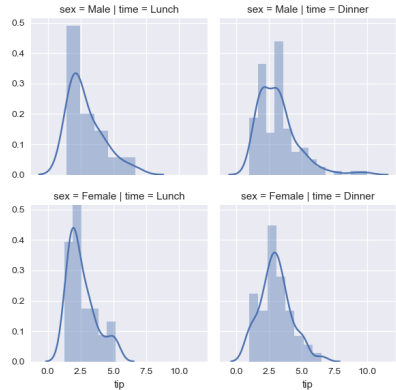
# Seaborn

## FacetGrid

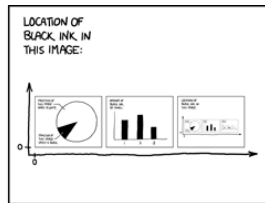
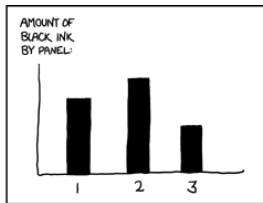
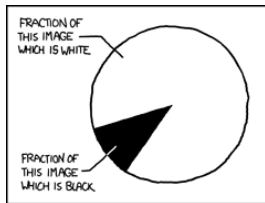


```
fmri = sns.load_dataset("fmri")
sns.relplot(x="timepoint", y="signal", kind="line",
            data=fmri)
```

Seaborn >= 0.9



```
g = sns.FacetGrid(tips, col="time", row="sex")
g.map(sns.distplot, "tip")
```



(Source)

# Seaborn

## Seaborn notebook

Seaborn notebook

[\(Link to notebook\)](#)