# Data visualization with Matplotlib and Seaborn

Inteligencia Artificial en los Sistemas de Control Autónomo Máster Universitario en Ingeniería Industrial

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

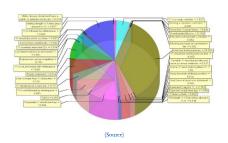
# Bibliography

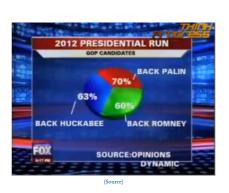
Jake VanderPlas. Python Data Science Handbook. Chapters 4. O'Reilly. (Link).

# Table of Contents

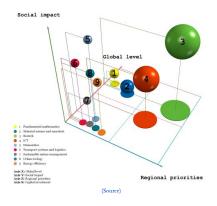
- I. Visualization examples
- 2. Motivation
- 3. Matplotlib
  - Seaborn
  - Seaborn datasets
  - Distributions

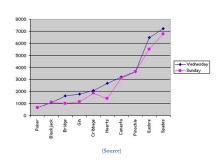
# Visualization examples (I)



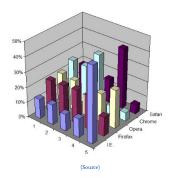


# Visualization examples (II)





# Visualization examples (III)

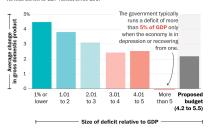




# Visualization examples (IV)

#### Strange time for a stimulus

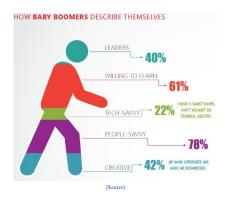
What annual economic growth averaged under various deficit-to-GDP ratios, since 1967



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 Seatember.

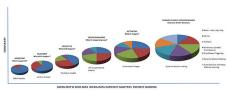
Sources: Commerce Department (GDP); Congressional Budget Office (historical deficit); Committee for a Responsible Federal Budget (deficit forecasts, budget changes) THE WASHINGTON POST

(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 comprension 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 (leyend, 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

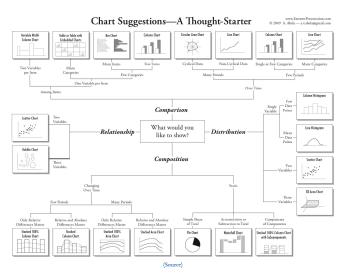
### What do you want to represent?

• Distribution, relationship, comparison or composition

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



# Motivation (III)



# 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(o, 10, 100)

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

plt.show()
```



# Matplotlib (II)

#### Matplotlib comes with two interfaces

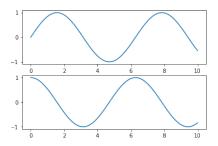
- Matlab-like, Old-fashioned function-oriented API.
- Object-oriented. Object-oriented and more powerfull API.

#### Matlab API

#### OO API

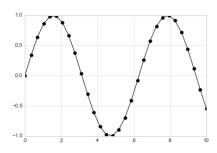


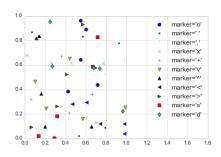
# Matplotlib (III)



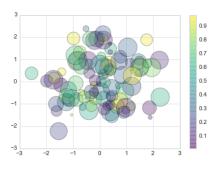


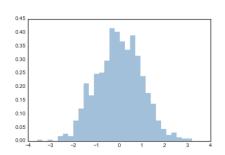
# Matplotlib (IV)





# Matplotlib (V)





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

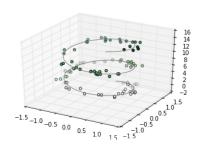
plt.hist(data, bins=30, normed=True, alpha=0.5, histtype='stepfilled', color='steelblue', edgecolor='none');
```

# Matplotlib (VI)

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

# Data for a three—dimensional line
zline = np.inspace(o, 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');
```



# 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! (v. 0.9)

#### Convention

import seaborn as sns

This documentation is for Seaborn

o.g or newer



# Seaborn (II)

### Display initialization

- plt.show()
- %matplotlib

### Style initialization

- Default Seaborn style sns.set()
- $\bullet\;$  By default, same style than matplotlib

#### Several functions ...

• ... similar parameters

### Parameters

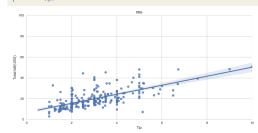
- x: Data axis x
- y: Data axis Y
- data: Dataframe name
- hue: Color
- style: Style
- sizes: Size
- kind: Alternate representation



# Seaborn (III)

# Typical Seaborn usage

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



### 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
О	16.99	1.01	Female	No	Sun	Dinner	2
I	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



### Datasets (II)

>>> 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
I	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 Setosa

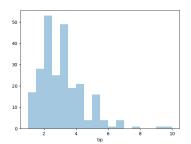


Iris Virginica

(Source)

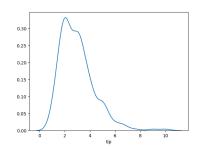


# Distributions (I)



#### Histogram

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

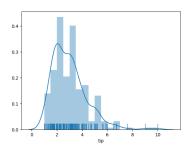


#### Density plot

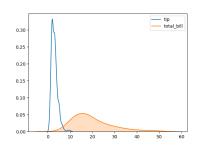
sns . distplot (tips ['tip'],
hist = False)



# Distributions (II)



#### Histogram + density plot

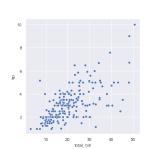


#### Density plot

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



# Relationships (I)



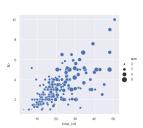
sns.relplot(x="total\_bill", y="
tip", data=tips)

Seaborn >= 0.9

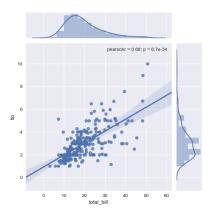
# Scatterplots

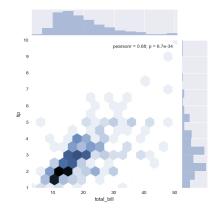


sns.relplot(x="total\_bill", y="
tip", hue="smoker", style="
smoker", data=tips)



# Relationships (II)



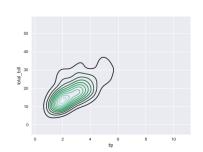


sns.jointplot("total\_bill", "tip", tips, kind="reg"

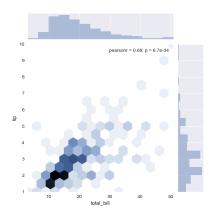
sns.jointplot("total\_bill", "tip", tips , kind="hex
")



# Relationships (III)



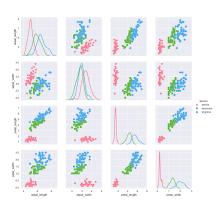
sns.kdeplot(tips['tip'], tips['total\_bill'])



sns.jointplot("total\_bill", "tip", tips , kind="hex

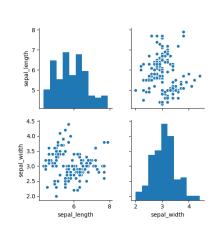


# Relationships (IV)



#### Scatterplot matrix

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

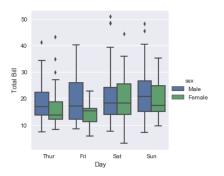


#### Scatterplot matrix

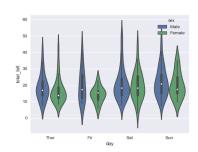
sns.pairplot(iris, vars=["sepal\_length", "sepal\_width"])



# Comparisons (I)

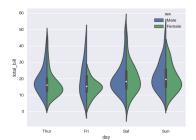


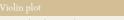


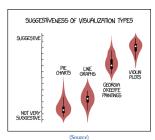


#### Violin plot

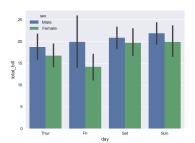
# Comparisons (II)





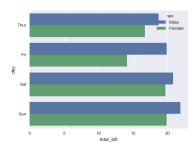


# Barplots



#### Damlat

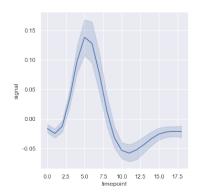
sns.barplot(x="day", y="total\_bill", hue="sex", data=tips)

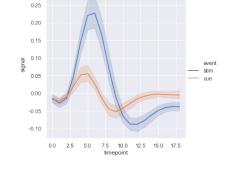


#### D . .... 1 . .

sns.barplot(x="total\_bill", y="day", hue="sex", data=tips, ci=None)

# Continuity

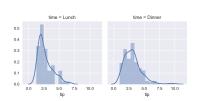




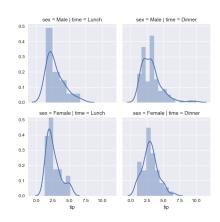
Seaborn >= 0.9



## FacetGrid



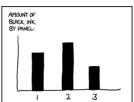
# Seaborn >= 0.9



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







(Source)

