



INMAS

Data Visualization

- *Exploratory Data Analysis*
- *Graphing in Python*

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INMAS Statistical Methods Workshop Fall 2021



Lecture Objectives

- **Apply** quantitative and visual exploratory techniques on data
- **Deduce** and **interpret patterns** revealed during exploratory data analysis (EDA)
- **Create** common statistical graphs using Seaborn.

Exploratory Data Analysis

Definition:

Exploratory Data Analysis (EDA) is a philosophy for the beginning of an analysis that describes a variety of techniques that are *quantitative* and *visual* in nature to look for patterns in data.

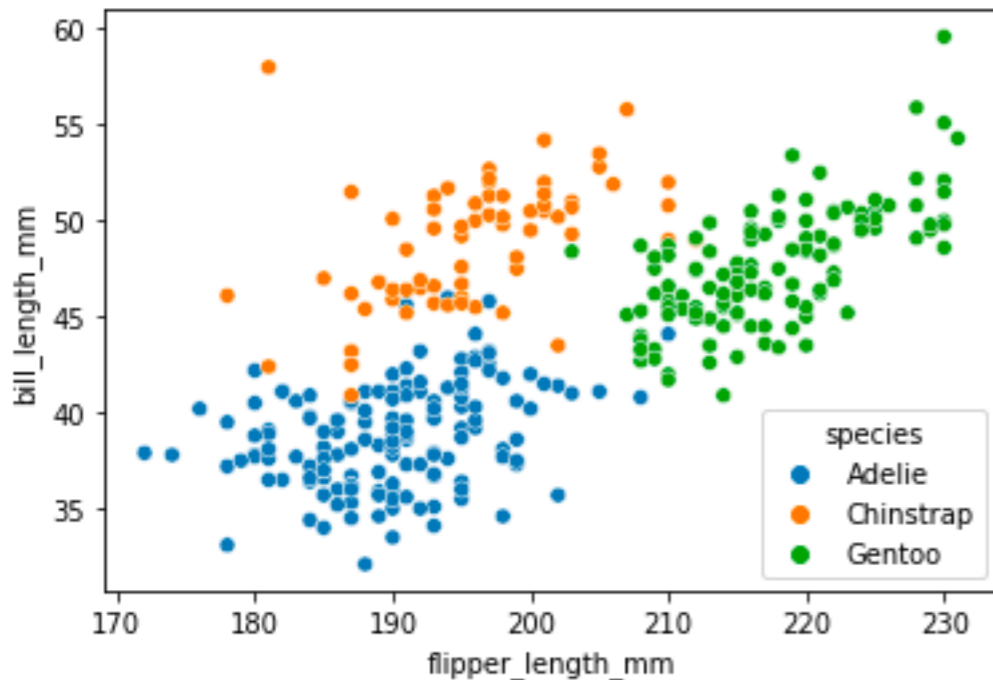


[Source](#)

Types of EDA

Quantitative vs. Visual

	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
count	342.000000	342.000000	342.000000	342.000000
mean	43.921930	17.151170	200.915205	4201.754386
std	5.459584	1.974793	14.061714	801.954536
min	32.100000	13.100000	172.000000	2700.000000
25%	39.225000	15.600000	190.000000	3550.000000
50%	44.450000	17.300000	197.000000	4050.000000
75%	48.500000	18.700000	213.000000	4750.000000
max	59.600000	21.500000	231.000000	6300.000000



Obtain summary information
penguins.describe()

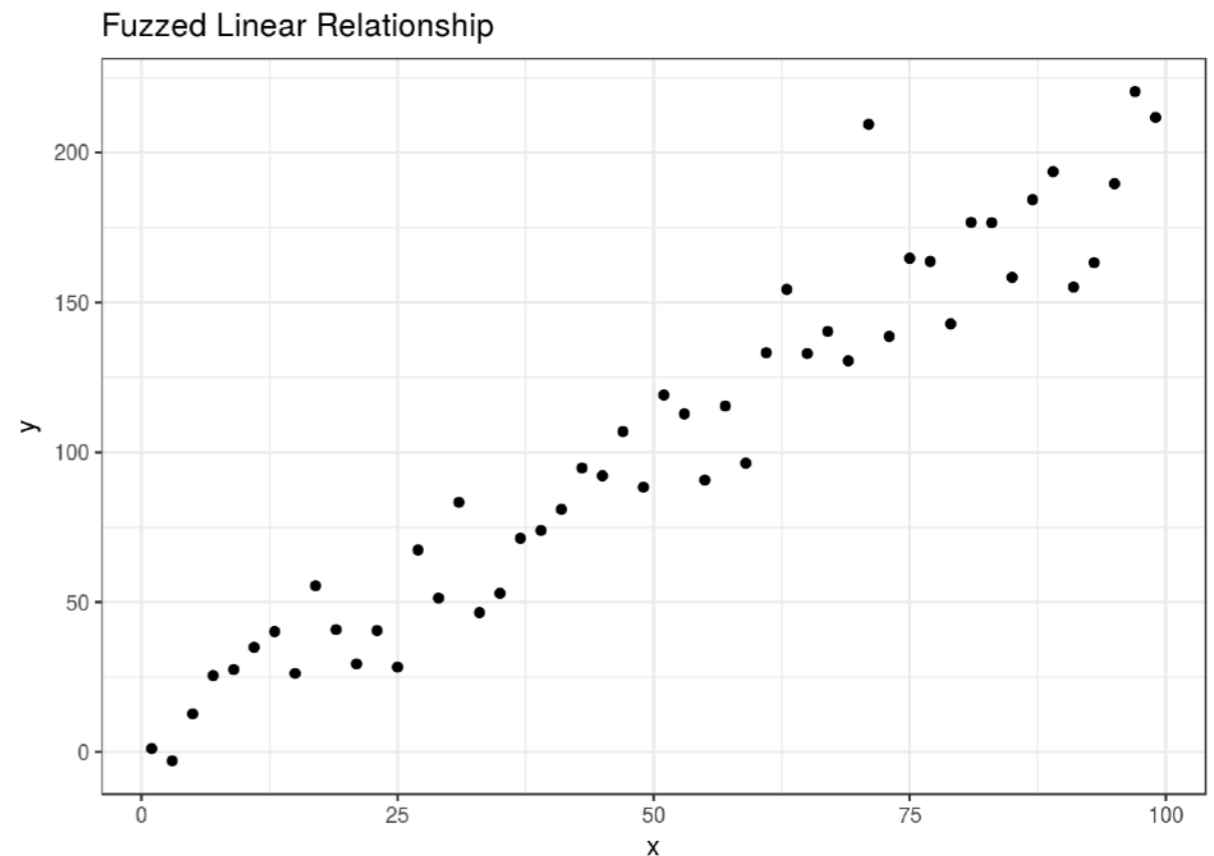
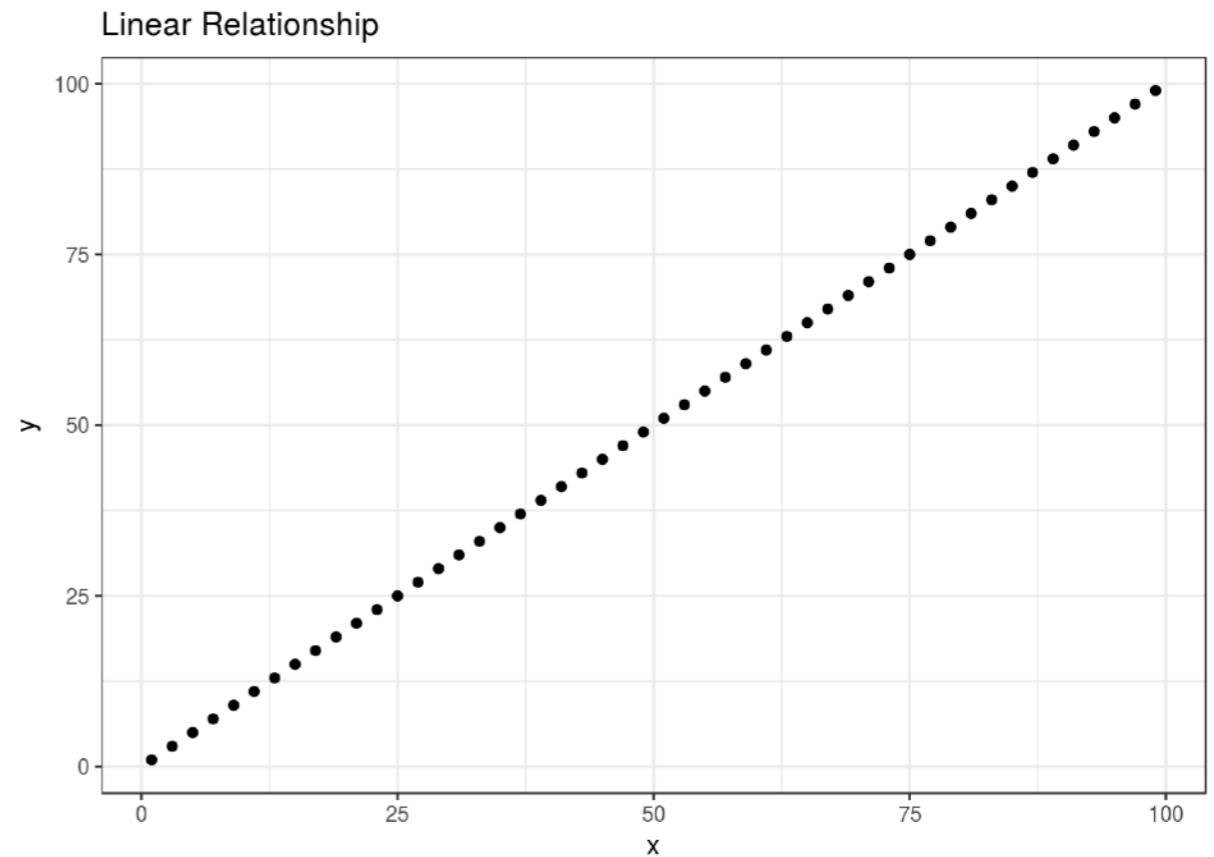
Visualize information with
a scatterplot

```
sns.scatterplot(  
    data=penguins,  
    x="flipper_length_mm",  
    y="bill_length_mm",  
    hue="species");
```

Patterns

... detecting, analyzing, and communicating ...

1. What kind of **relationship exists** is present?
2. What's the **level of strength** of the relationship?
3. Are there any **confounding variables** that might be behind it?
4. How does the **pattern exist in subsets** of the data?



Definition:

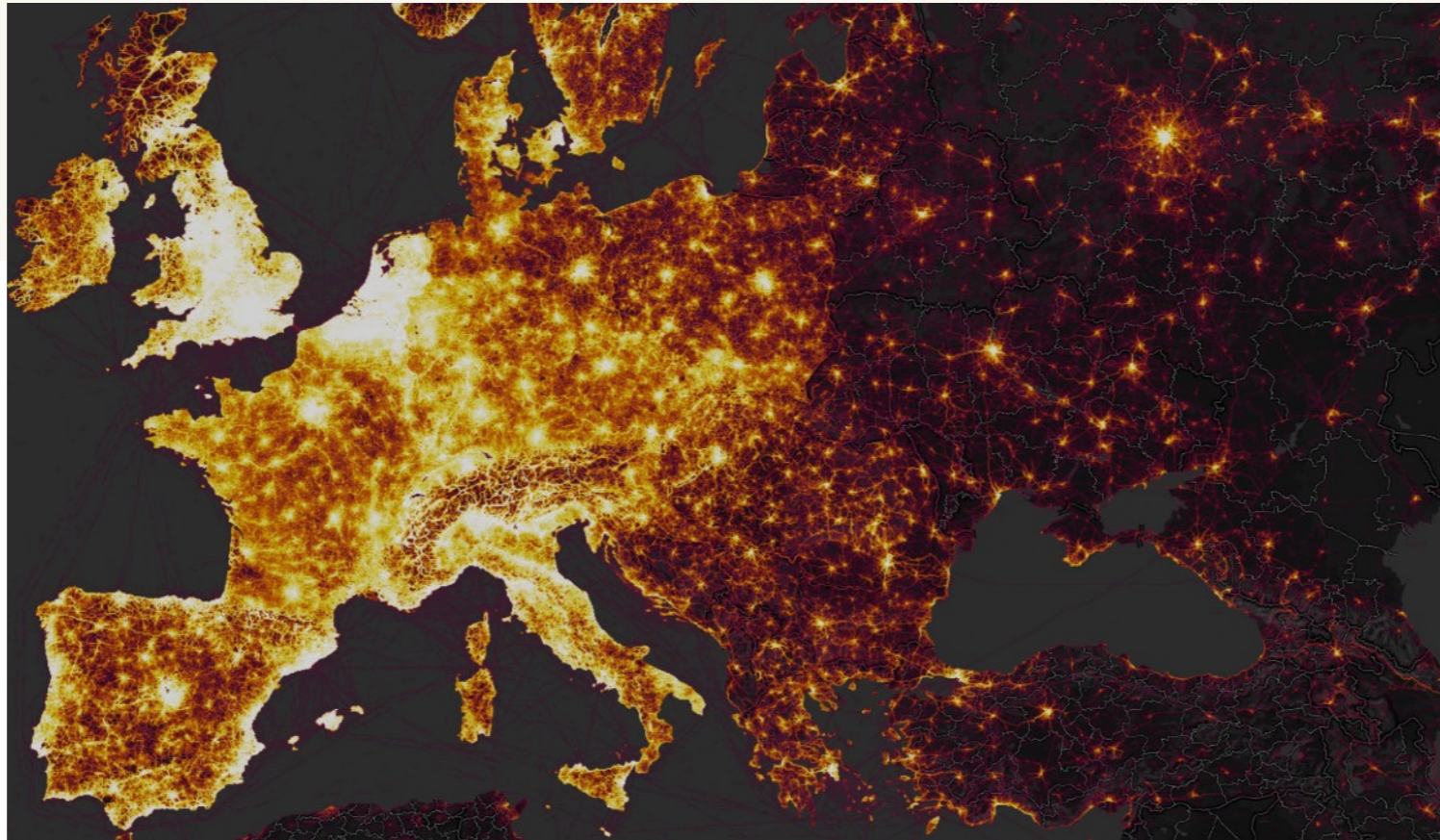
Variation is the difference between observations in **one variable**.



[Source](#)

Definition:

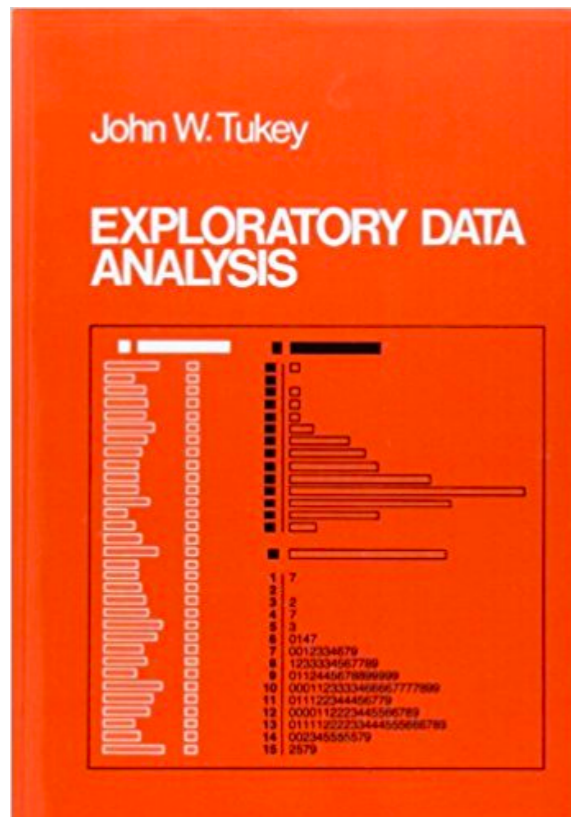
Covariation is the difference among observations between **two or more** variables.



[Source](#)

“...make **both** calculations **and** graphs. Both sorts of output should be studied; each will contribute to understanding.”

– F. J. Anscombe, 1973

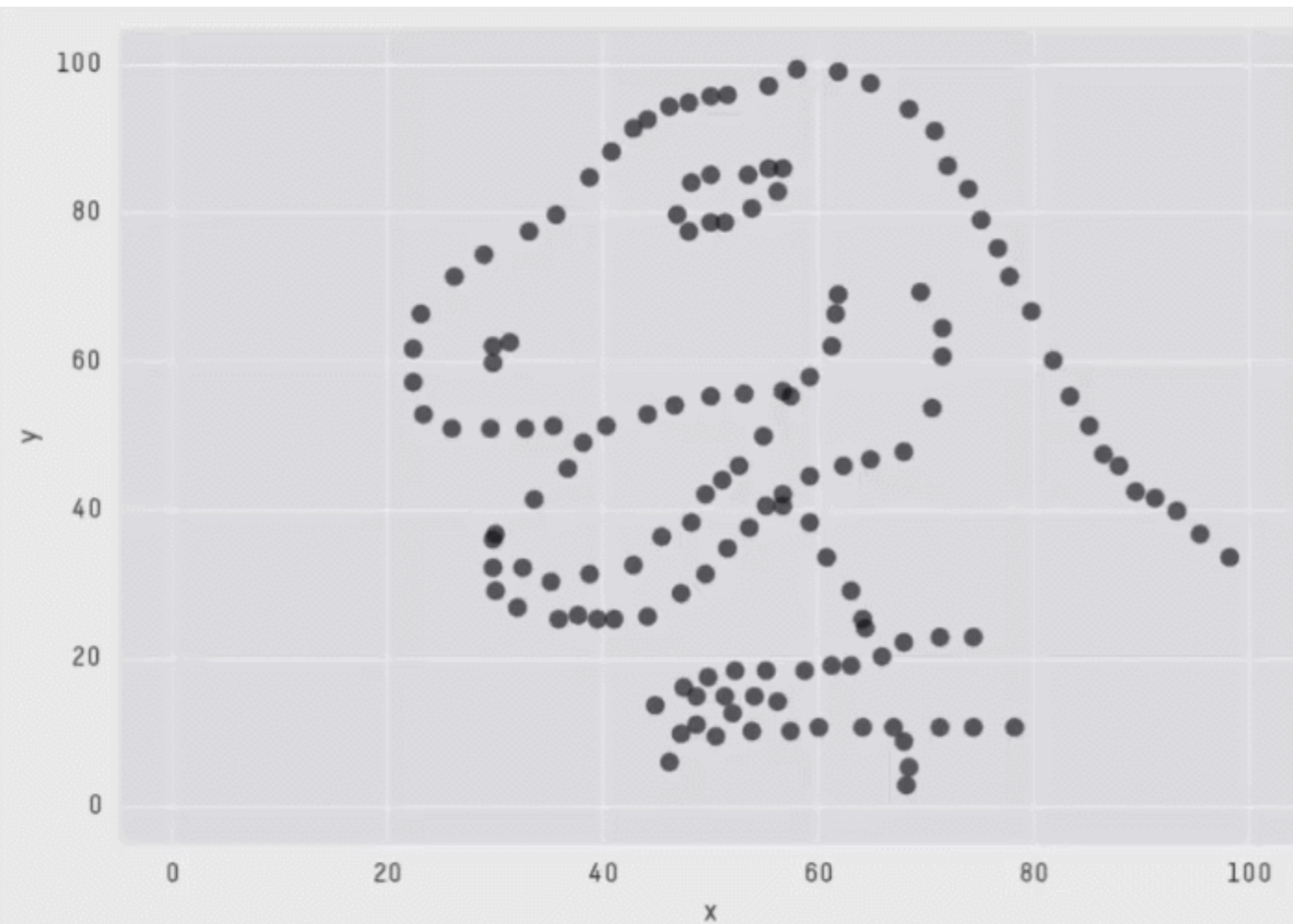


“The greatest value of a picture is when it forces us to notice what we never expected to see.”

–John Turkey in Exploratory Data Analysis (1977)

Datasaurus

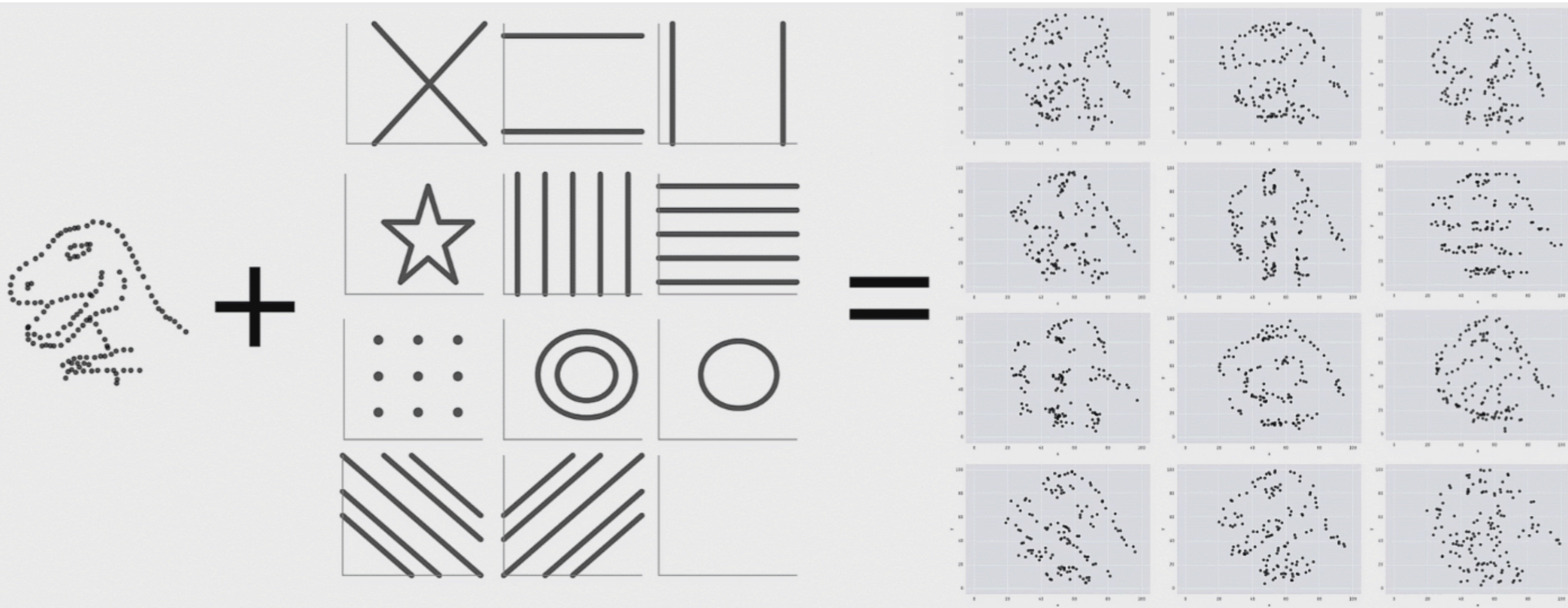
... the new [anscombe's quartet](#) ...



```
X Mean: 54.2659224  
Y Mean: 47.8313999  
X SD : 16.7649829  
Y SD : 26.9342120  
Corr. : -0.0642526
```

Pattern

... the new anscombe's quartet ...

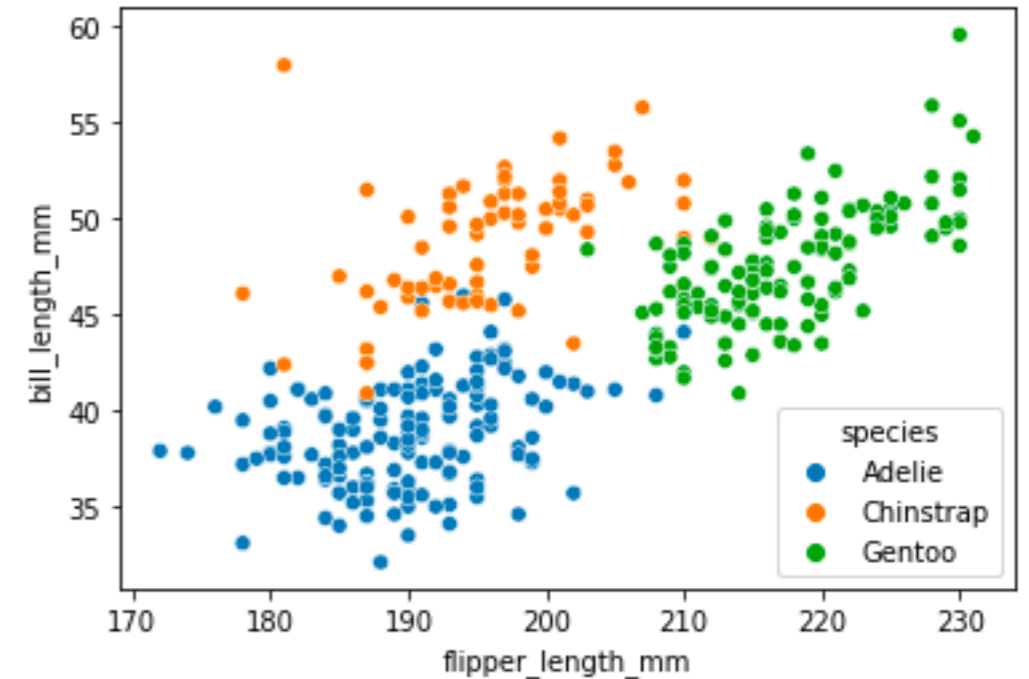
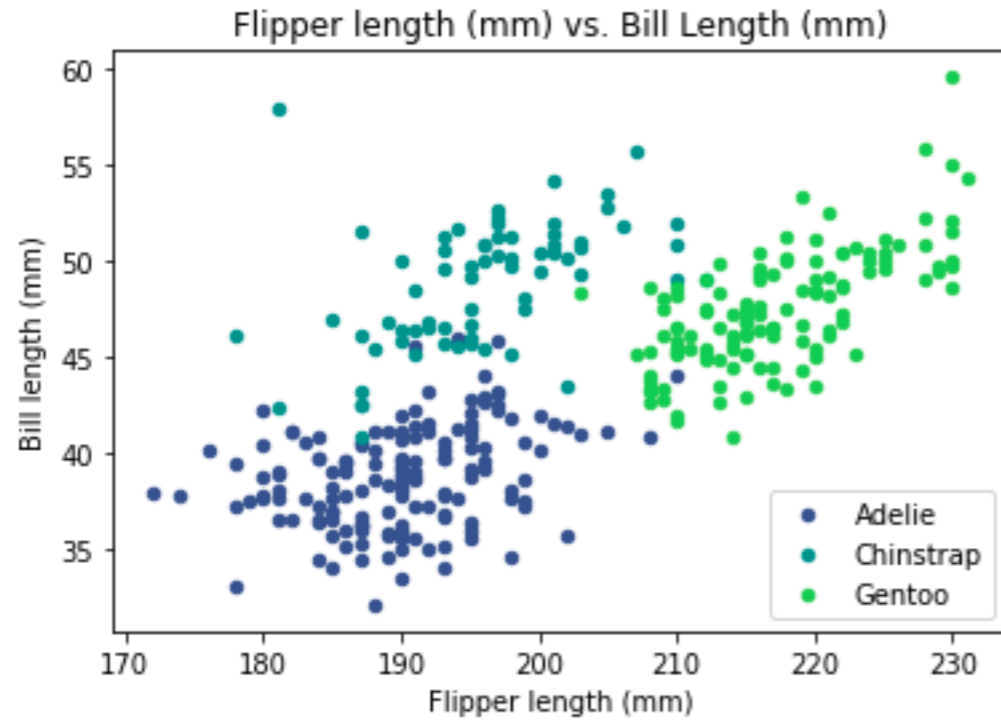


Source

Graphing in Python

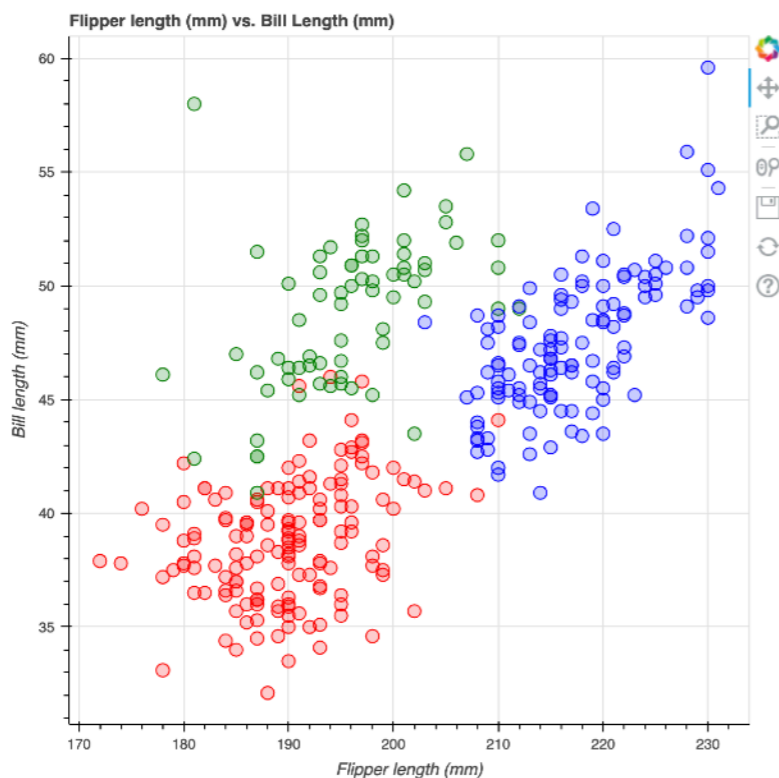
Python Graphing Systems

Comparison of Graphs on similar data

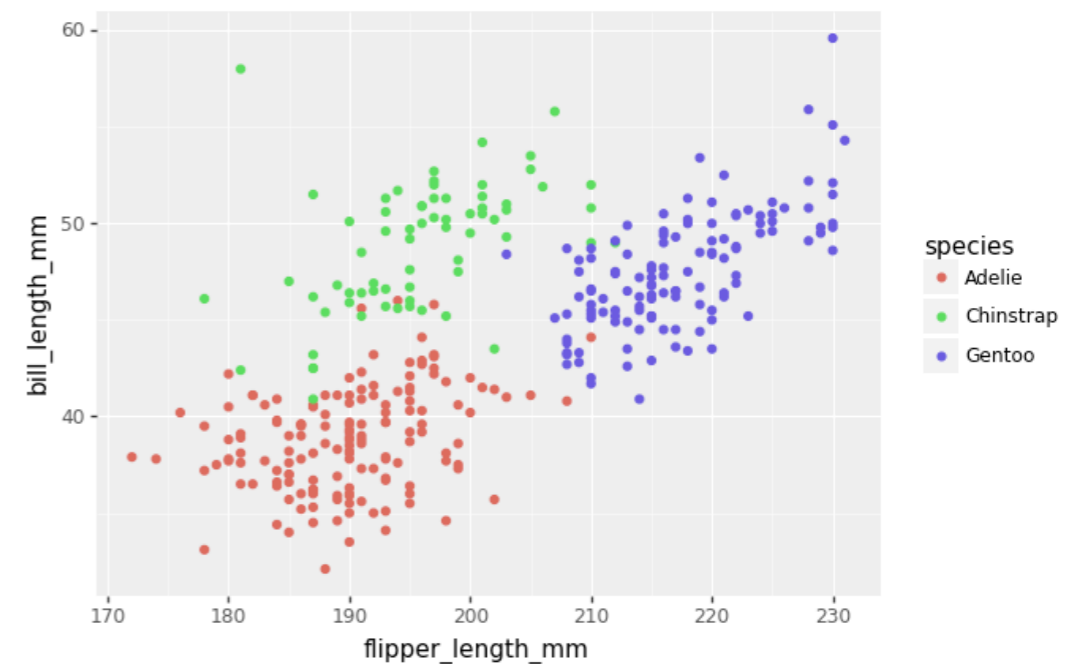


matplotlib

seaborn



Bokeh



plotnine

Goldilocks Scenario

... what graph system to choose ???



matplotlib

Highly
customizable



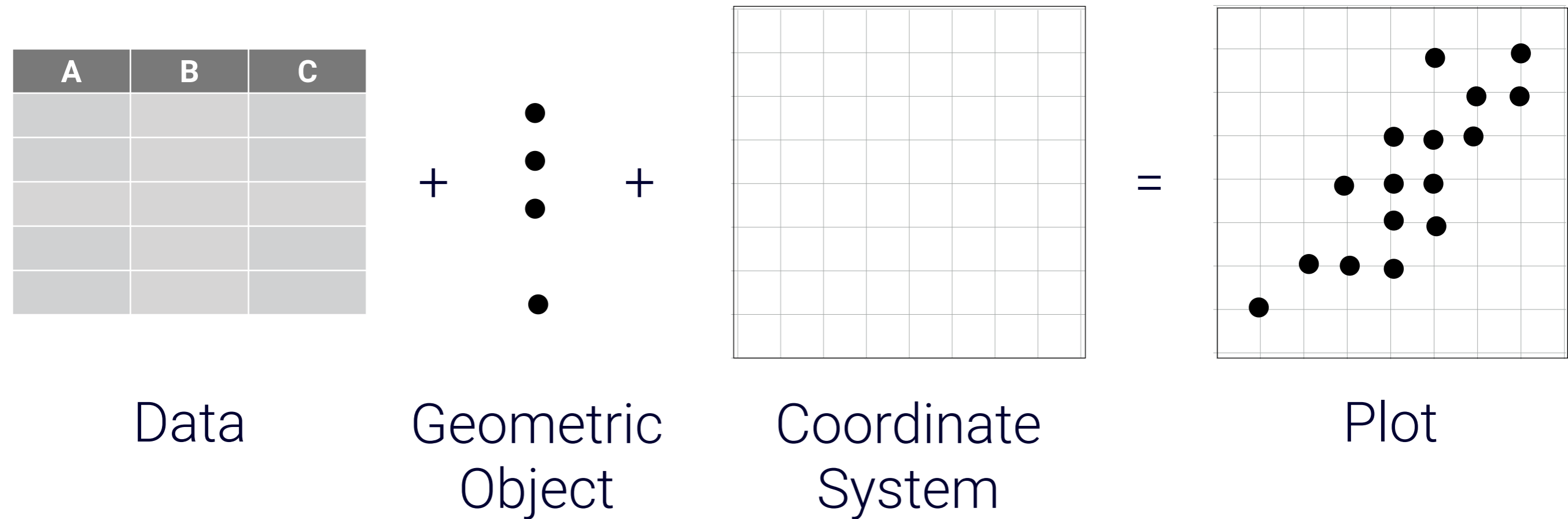
seaborn
bokeh

plotnine

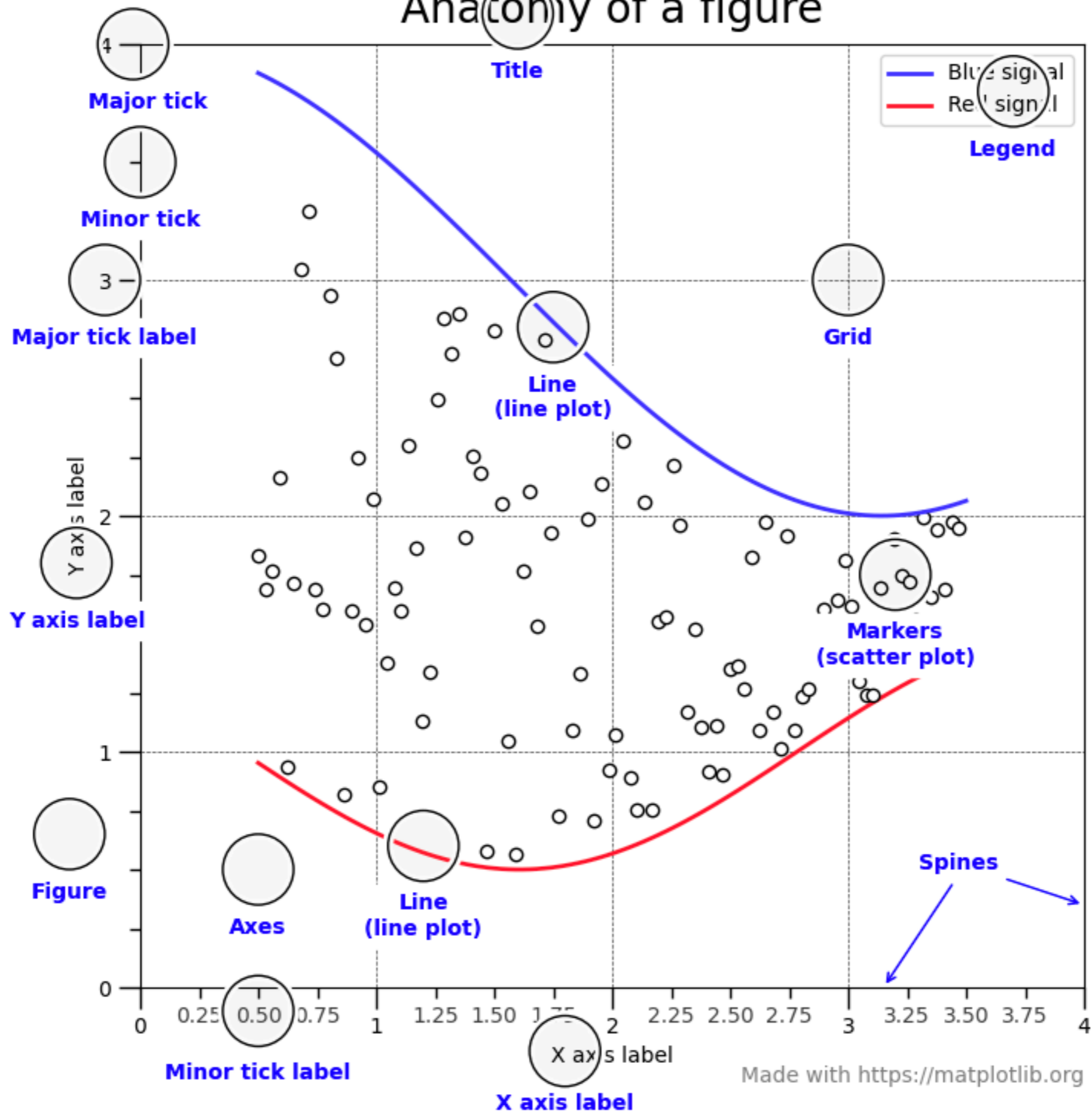


Grammar of a Graph

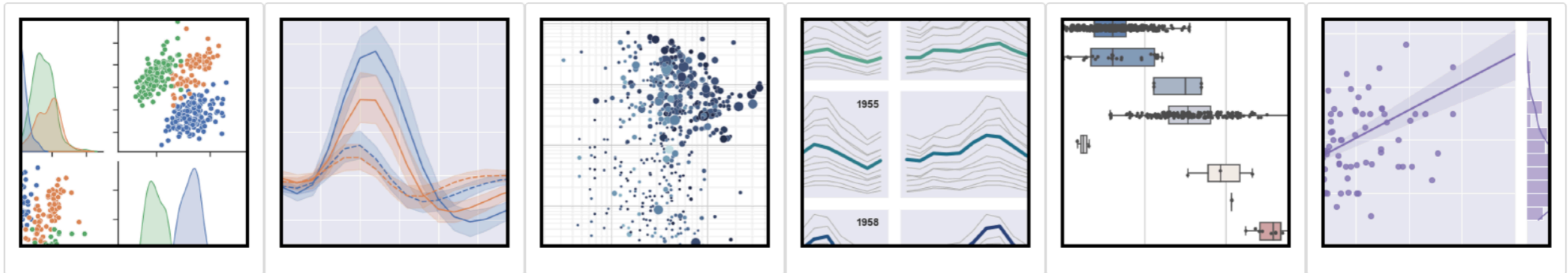
Underlying Structure of Graphs



Anatomy of a figure



seaborn: statistical data visualization



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

For a brief introduction to the ideas behind the library, you can read the [introductory notes](#). Visit the [installation page](#) to see how you can download the package and get started with it. You can browse the [example gallery](#) to see what you can do with seaborn, and then check out the [tutorial](#) and [API reference](#) to find out how.

To see the code or report a bug, please visit the [GitHub repository](#). General support questions are most at home on [stackoverflow](#) or [discourse](#), which have dedicated channels for seaborn.

Contents

- [Introduction](#)
- [Release notes](#)
- [Installing](#)
- [Example gallery](#)
- [Tutorial](#)
- [API reference](#)

Features

- Relational: [API](#) | [Tutorial](#)
- Distribution: [API](#) | [Tutorial](#)
- Categorical: [API](#) | [Tutorial](#)
- Regression: [API](#) | [Tutorial](#)
- Multiples: [API](#) | [Tutorial](#)
- Style: [API](#) | [Tutorial](#)
- Color: [API](#) | [Tutorial](#)

Graphing Outline

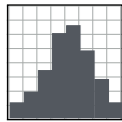
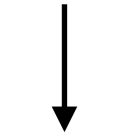
... when to use a graph ...

Variables

One Variable (Variation)

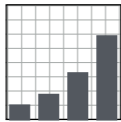
$x = \langle ? \rangle$

Continuous



Histogram

Categorical or Discrete

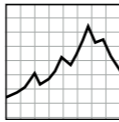


Barplot

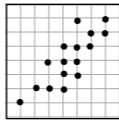
Two Variables (Covariation)

$x = \langle ? \rangle, y = \langle ? \rangle$

Indep: Continuous
Dep: Numerical

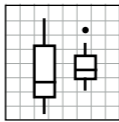


Line Graph

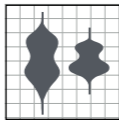


Scatterplot

Indep: Categorical
Dep: Numerical

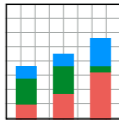


Boxplot

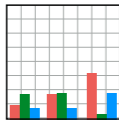


Violin Graph

Indep: Categorical
Dep: Categorical



Stacked Barplot



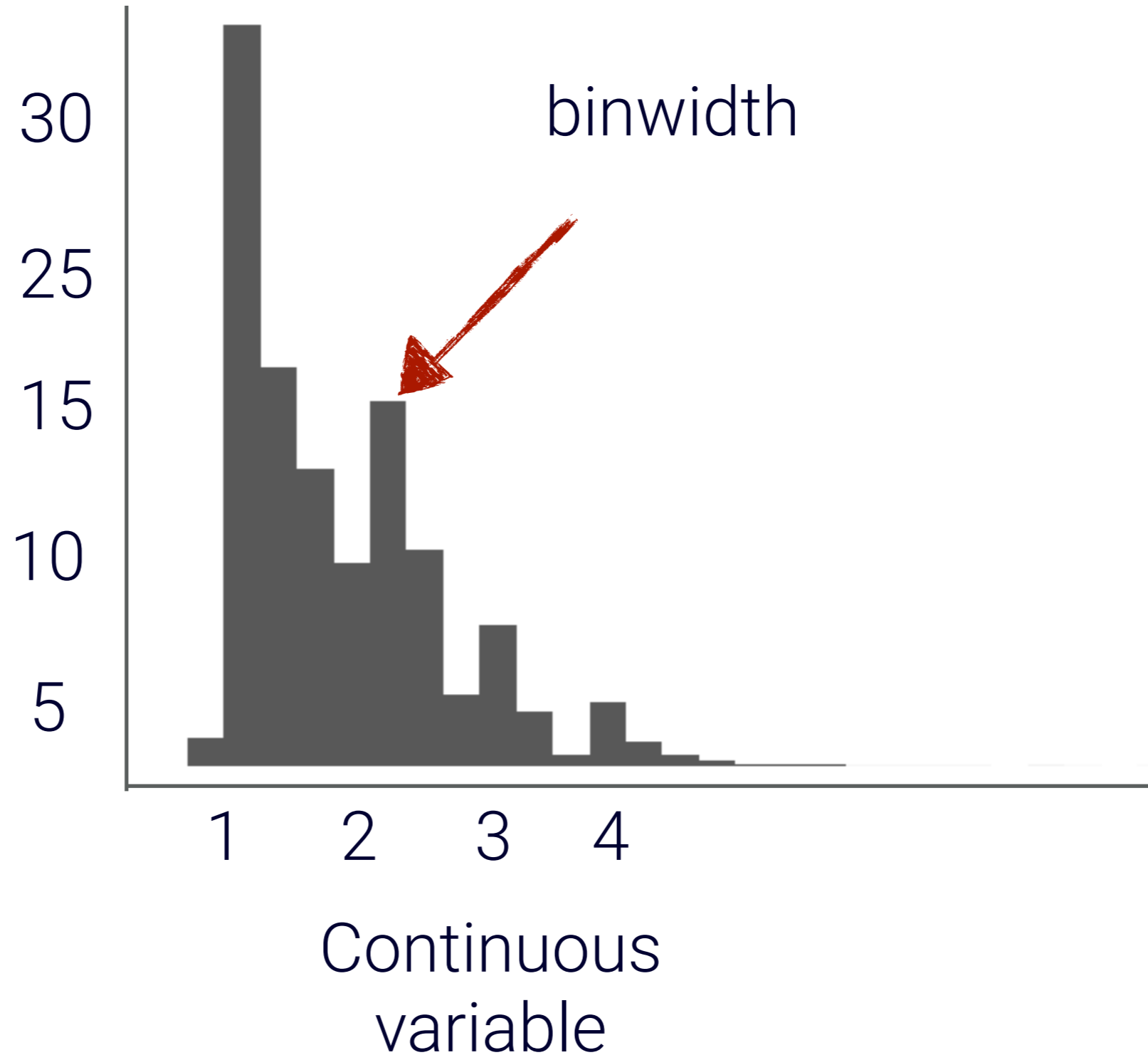
Side-by-side Barplot

Independ
Dependent

Histogram

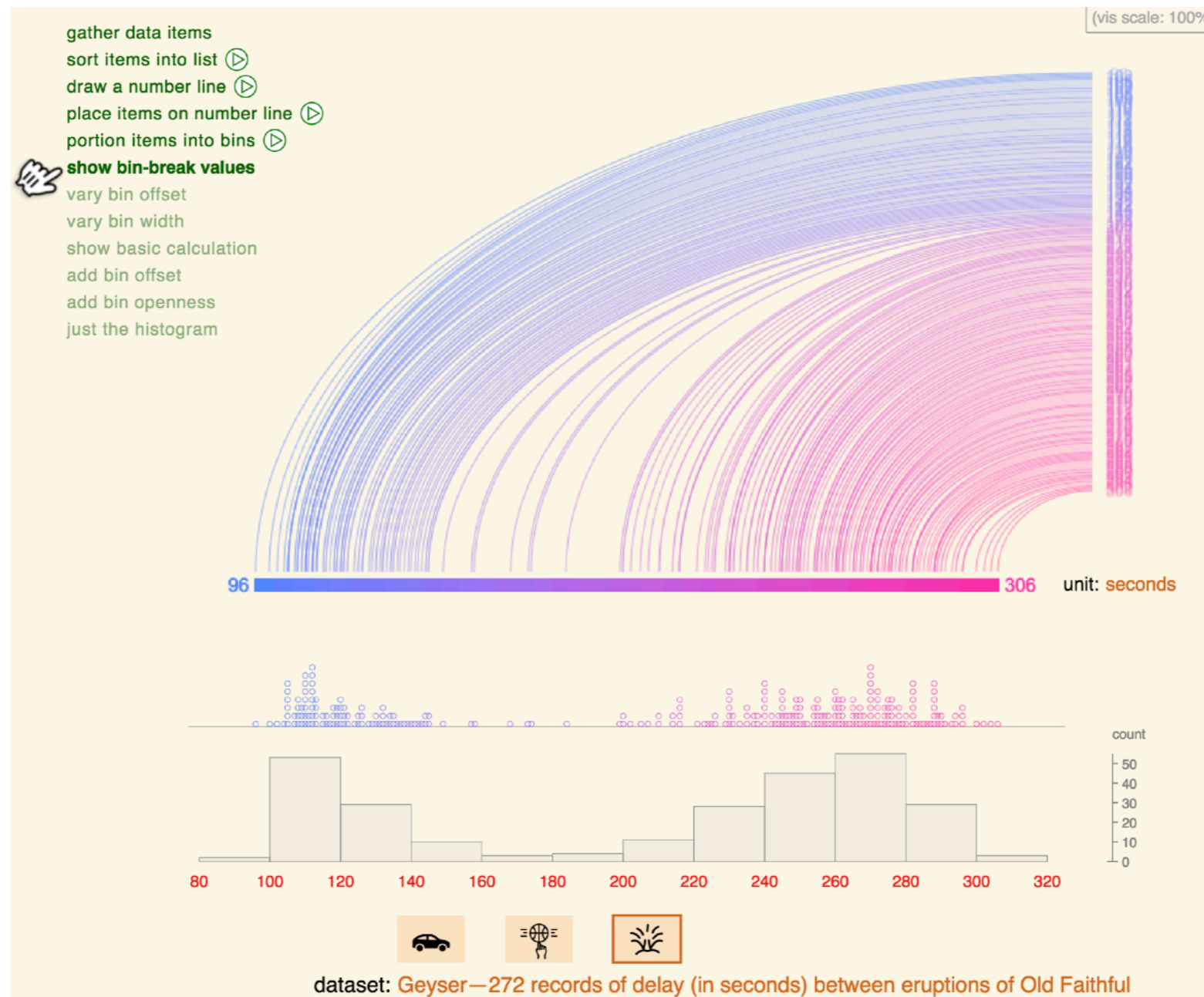
Continuous data for one variable

Frequency of
Observations
that fall into
the bin



Histograms

imagined "bins" for continuous data

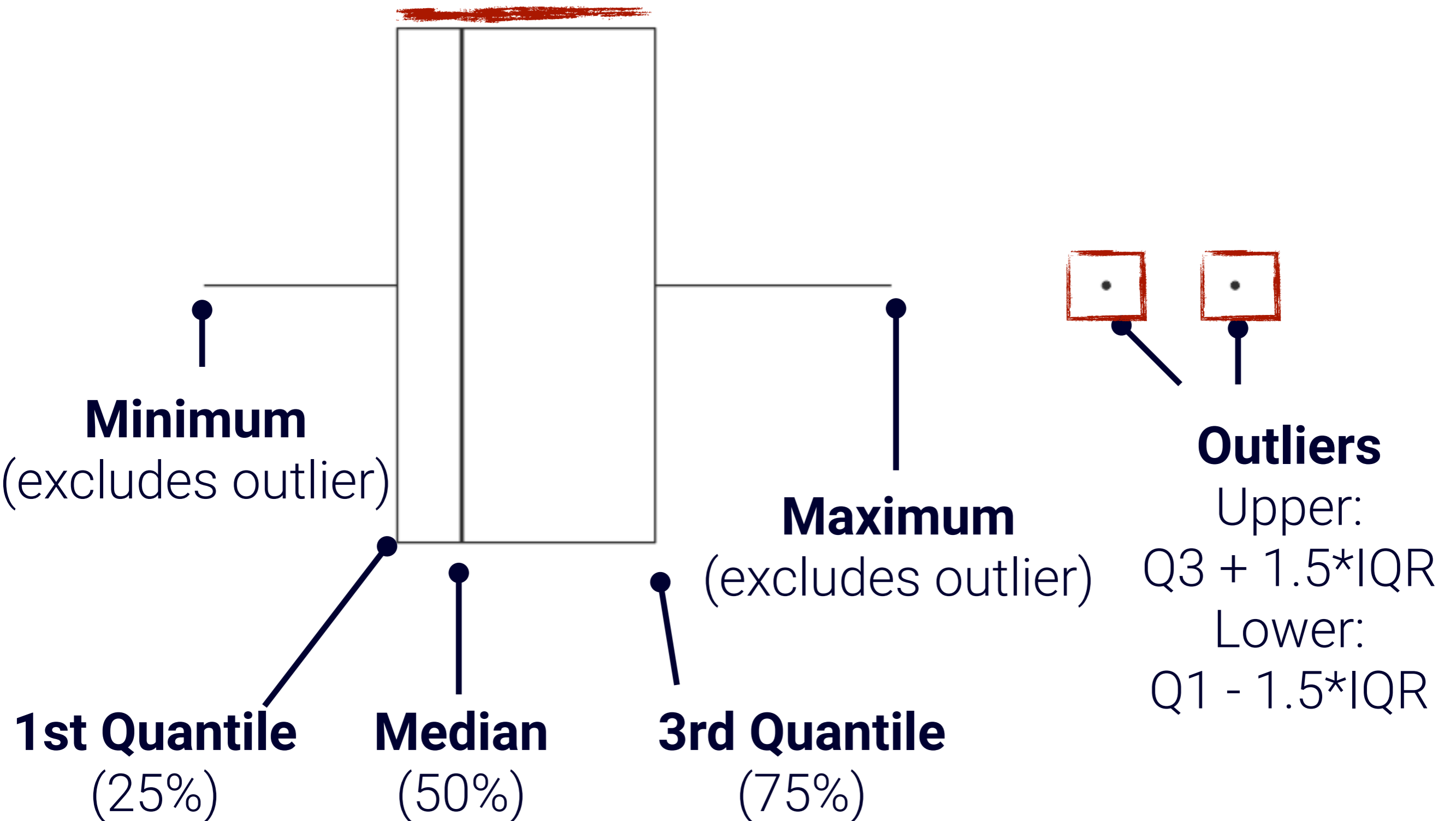


<https://tinlizzie.org/histograms/>

Boxplot

Categorical paired with **numerical** to observe *covariation*

Interquartile range (**IQR** = $Q3 - Q1$)



data-to-viz.com

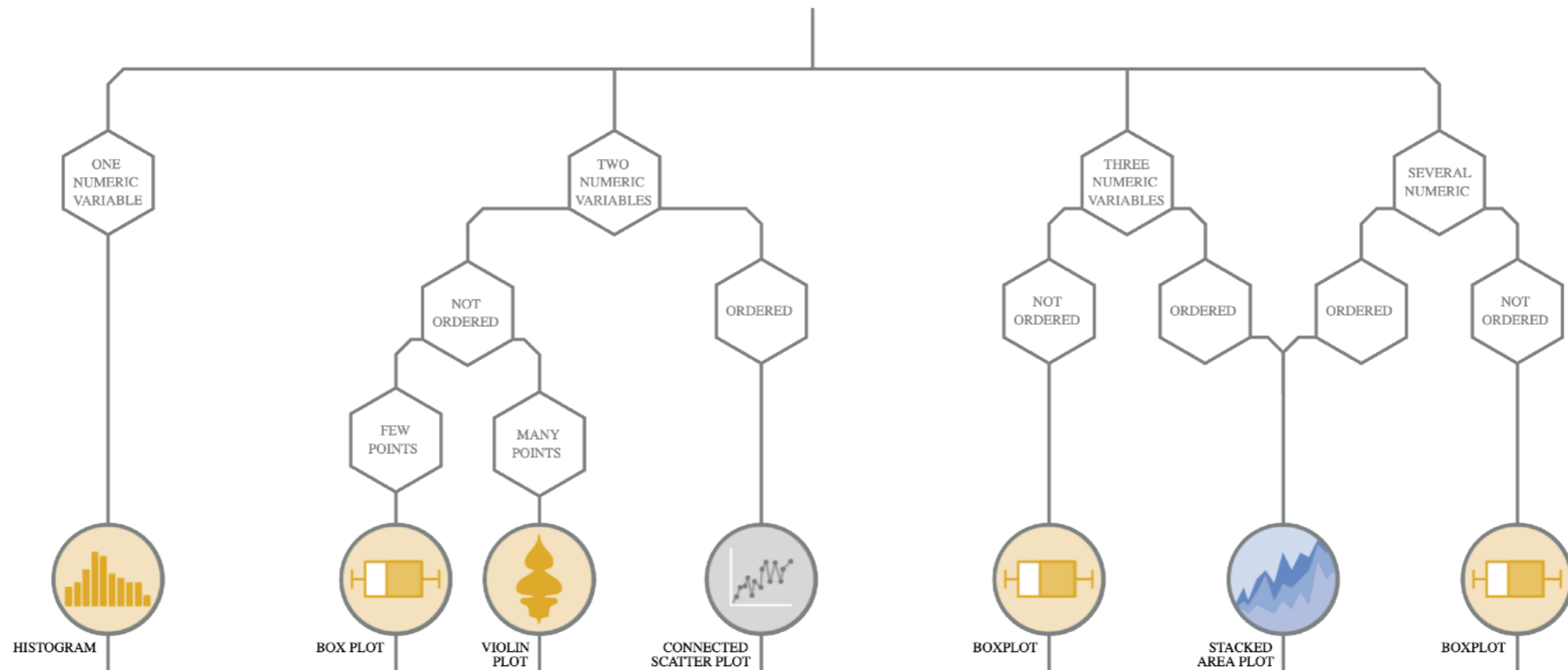
Flowchart for picking the best visualization given data

from Data to Viz

[EXPLORE](#) [STORY](#) [ALL](#) [CAVEATS](#) [POSTER](#) [ABOUT](#) [CONTACT](#)

What kind of data do you have? Pick the main type using the buttons below. Then let the decision tree guide you toward your graphic possibilities.

Numeric [Categoric](#) [Num & Cat](#) [Maps](#) [Network](#) [Time series](#)



Code Examples

VIOLIN

An alternative to *boxplot* to compare the distribution of several groups.

About

Violinplots allow to visualize the distribution of a numeric variable for one or several groups. It is really close from a *boxplot*, but allows a deeper understanding of the distribution.

Violins are particularly adapted when the amount of data is huge and showing individual observations gets impossible.

Common Mistakes

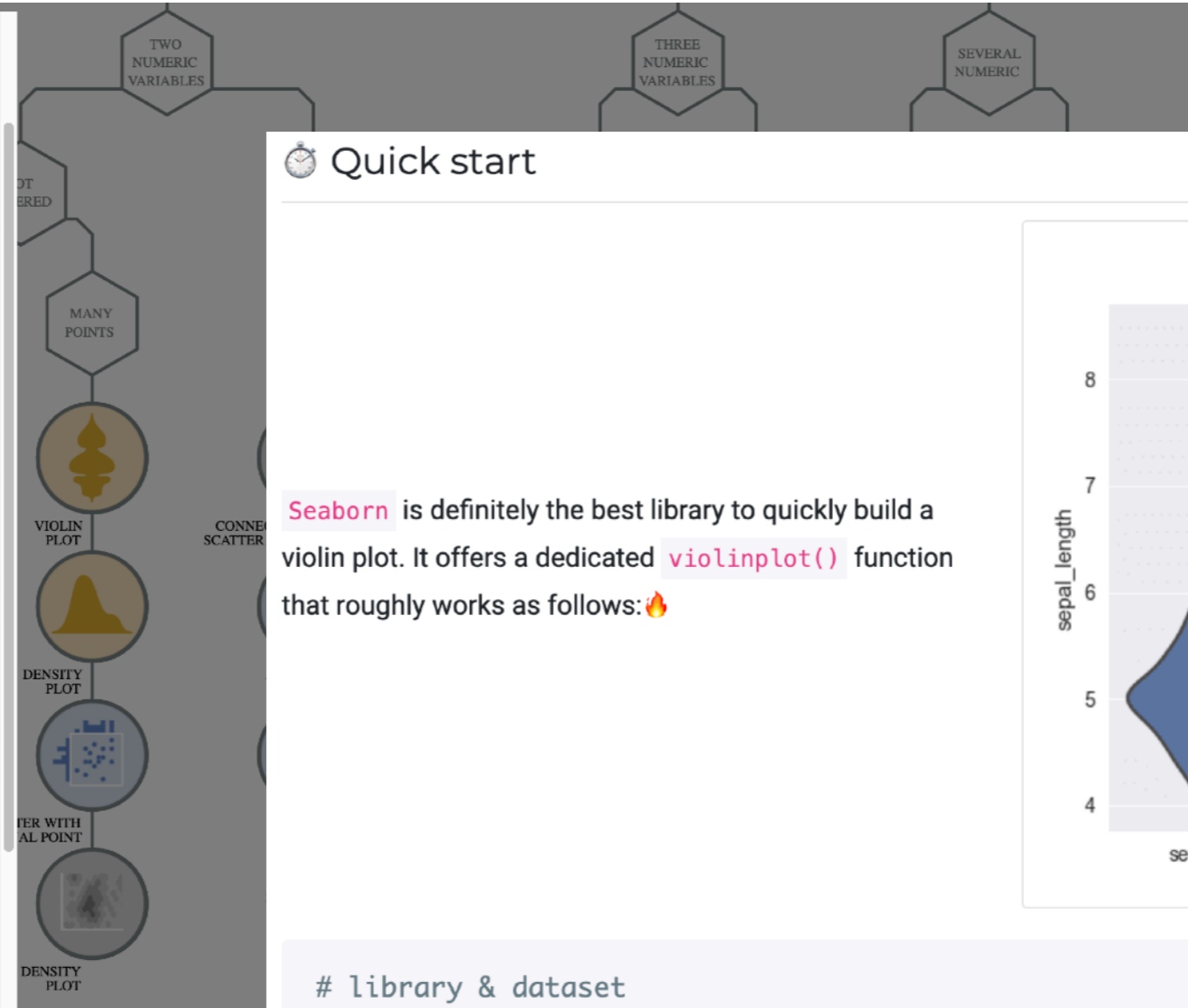
- If you have just a few groups, you are probably interested by *ridgeline charts*.
- If you compare groups with very different sample size, *show it*.
- *Ordering groups* by median value makes the chart more insightful.

Code

[R graph gallery](#)

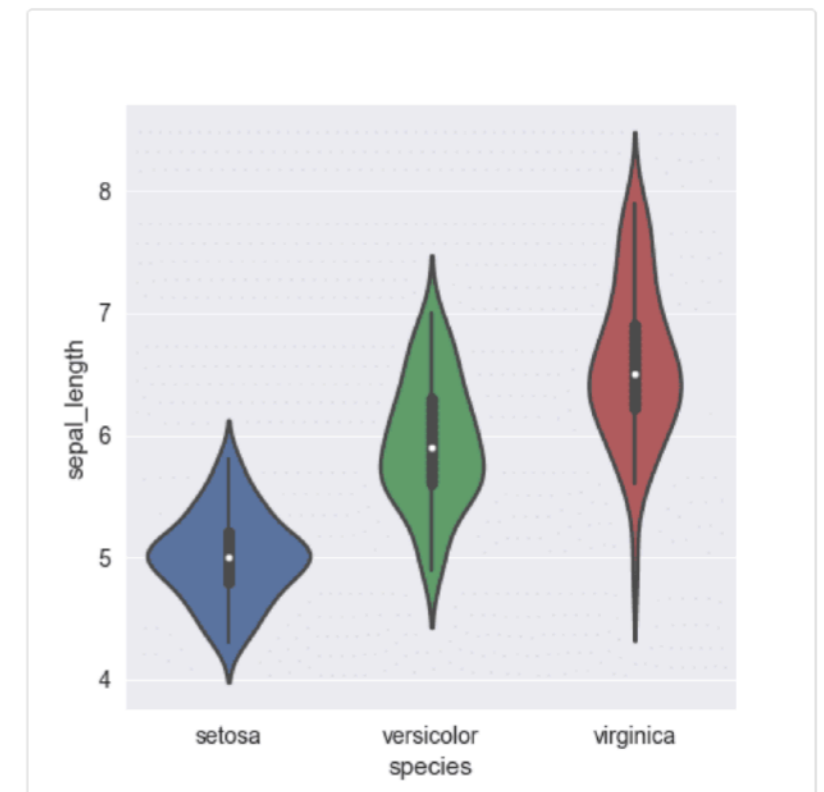
[Python gallery](#)

[D3.js gallery](#)



Quick start

Seaborn is definitely the best library to quickly build a violin plot. It offers a dedicated `violinplot()` function that roughly works as follows:🔥



```
# library & dataset
import seaborn as sns
df = sns.load_dataset('iris')

# plot
sns.violinplot(x=df["species"], y=df["sepal_length"])
```

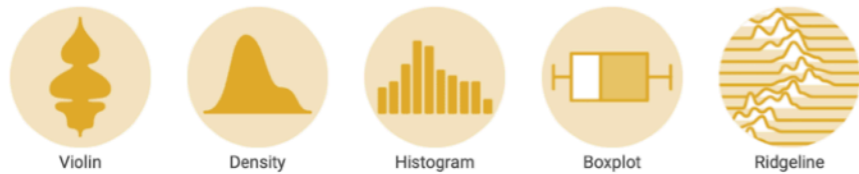

Sample Implementations

Code to Generate Various Graphs in Python and R

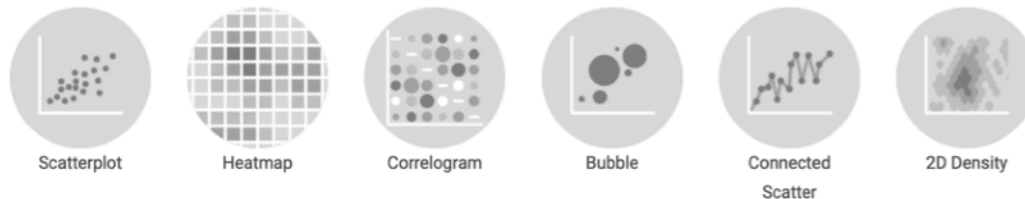
← Python Graph Gallery

CHART TYPES TOOLS ▾ ALL RELATED ▾ ABOUT 🔍

Distribution



Correlation



Ranking



[Source](#)



HOME GGLOT2 ALL GRAPHS BLOG ABOUT PYTHON



Welcome to the [R Graph Gallery](#). Looking for inspiration or help concerning data visualisation? Here, you will find hundreds of distinctive graphics made with the [R programming language](#), always with the reproducible code snippet

available. Charts are displayed in several sections represented by the icons below. The gallery dedicates a special section to tricks you can use with the [ggplot2 library](#). If you are looking to browse for inspiration, the [all graph](#)

[page](#) displays all the charts of the gallery in a row. Feel free to propose a chart or report a bug; any feedback is highly welcome. Stay in touch with the gallery by following it on [Twitter](#) or [Facebook](#), or by subscribing to the blog.

Distribution



[Source](#)

Summary

- Emphasized the benefits of graphing data.
- Discussed different approaches to graphing data with one variable vs. two variables.

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