



# **Introduction to Data Manipulation & Visualization**

## Visualization Objectives

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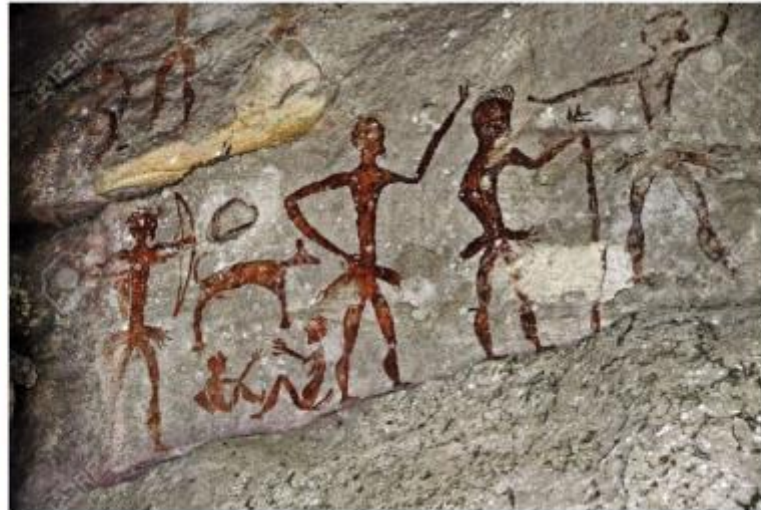
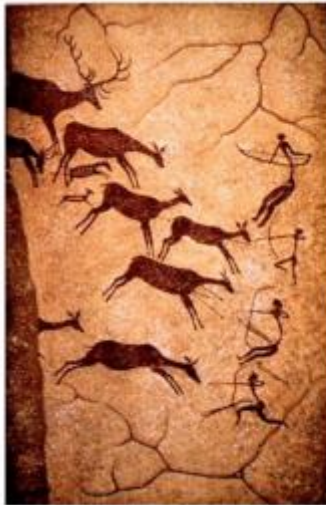
- Record information
- Analyze data to support reasoning
- Confirm hypotheses
- Communicate ideas to others



## Why Visualize?

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### To record information



Why Visualize?

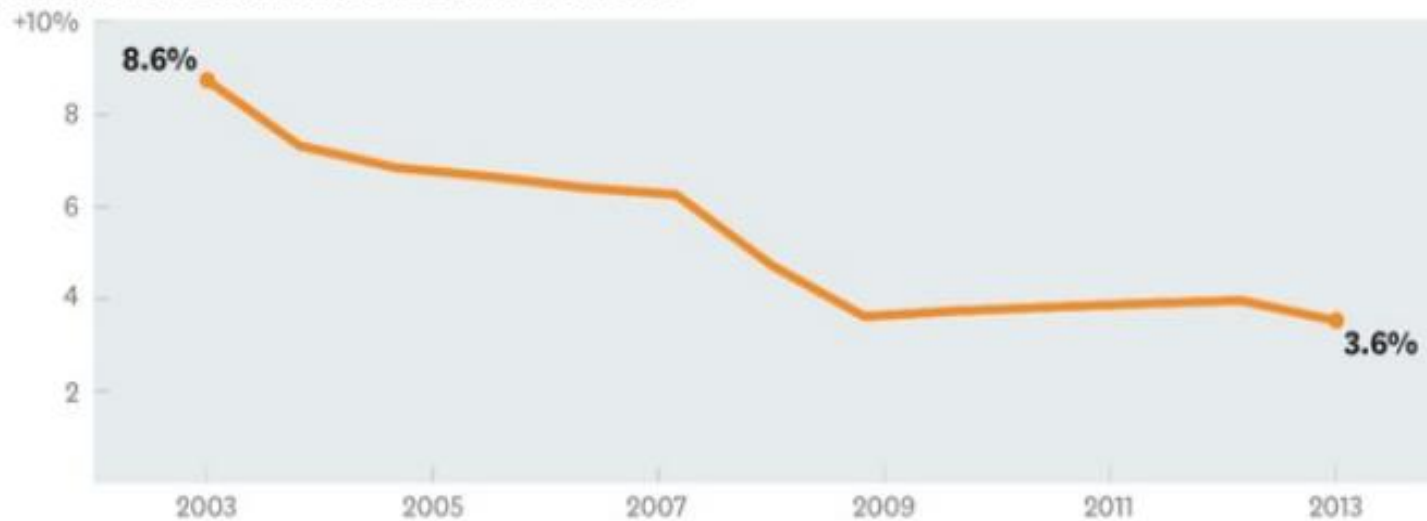
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**To communicate information**



## Annual Growth is Declining

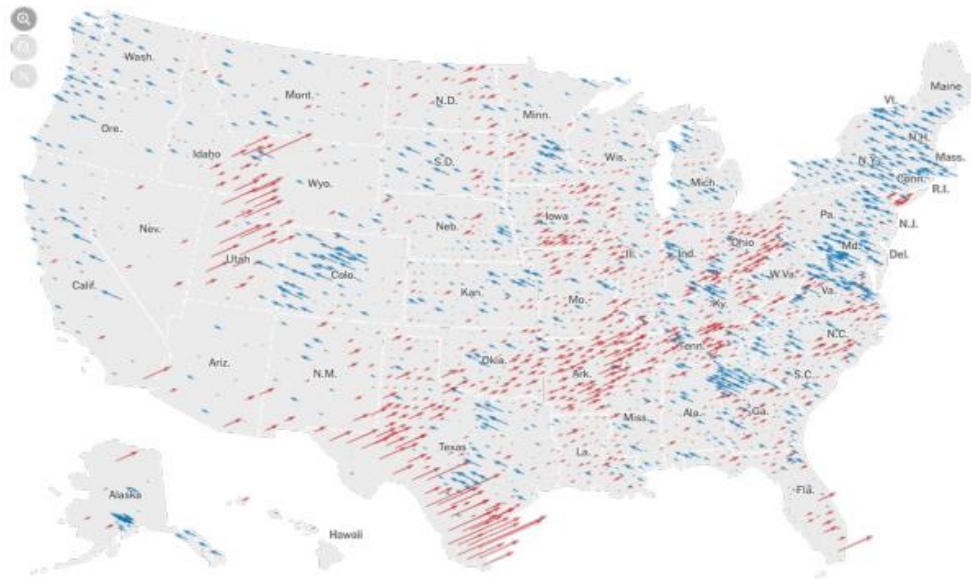
**ANNUAL GROWTH IN HEALTH CARE SPENDING**



Why Visualize?

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**To analyze data**



2020 US Elections (NYTimes)

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## Types of Plots

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- Line plots
- Bar plots
- Scatter plots
- Box plots
- Histograms



## What are line plots?

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Two types of relational plots:

### 1) Scatter plots

- Each plot point is an independent observation

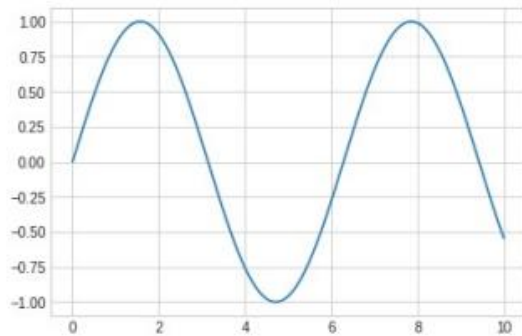
### 2) Line plots

- Each plot point represents the same "thing", typically tracked over time



## Line plot

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- Used for numeric data
- **Used to show trends**
- Compare two or more different variables over time
- Could be used to make predictions



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### **View trends in data over time.**

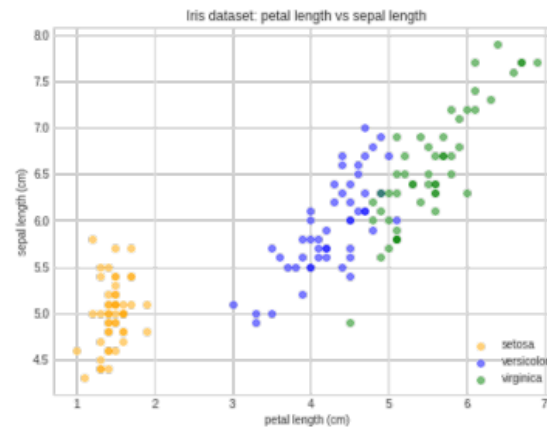
Examples: Stock price change over a five-year period or website page views during a month.



## Scatter plots

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- Investigate relationships between quantitative values.
  - Used to visualize relation between two numeric variables
  - Used to visualize correlation in a large data set



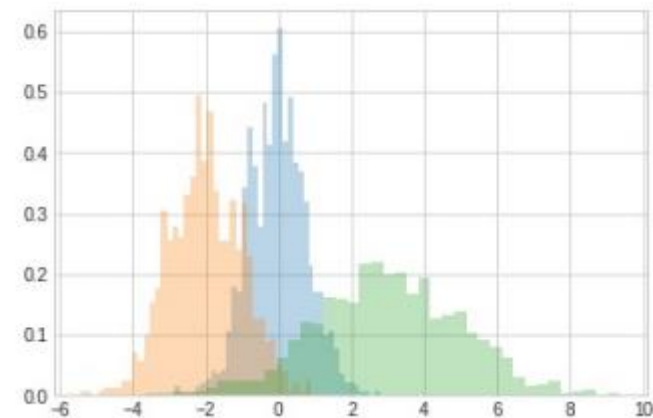
# Histograms

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- Understand the distribution of your data.
  - Displays the frequency distribution (shape)
- Summarize large data sets graphically
- Compare multiple distributions

Examples:

- Number of customers by company size,
- Student performance on an exam,
- Frequency of a product defect.

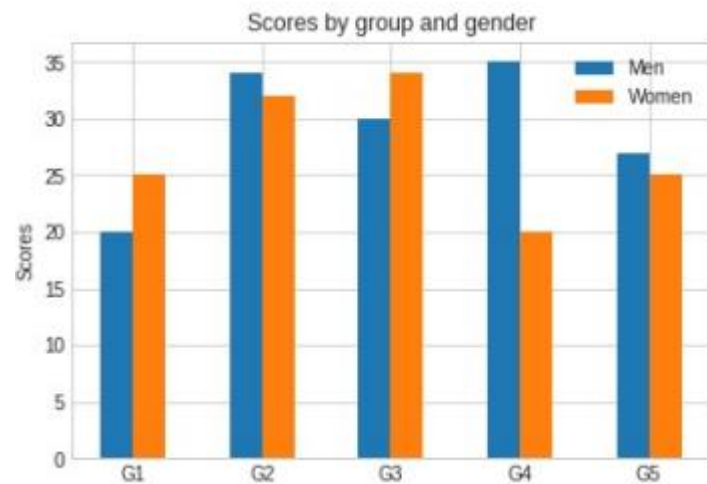


## Bar plots

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**Compare data amongst different categories**

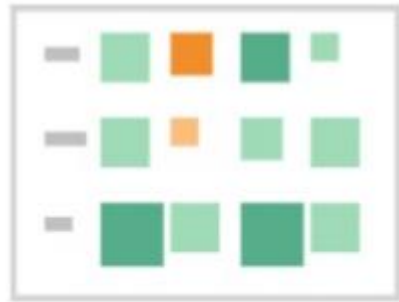


Examples: percent of spending by department.



## Heat Map

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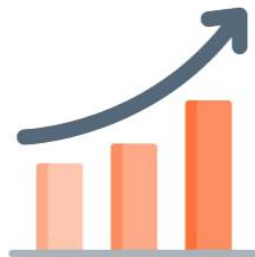


— Show the relationship between at least two factors.

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# Data Visualization

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## Part 1

### Basic Visuals | Matplotlib, Seaborn

Basic Visualization Concepts, Introduction and Comparison b/t Matplotlib and Seaborn Python Libraries in Jupyter Notebook.



## Part 2

### Interactive Visuals | Plotly, Bokeh, Tableau, etc.

Deeper insights into more interactive and fun data visualization functions. Introduction to Plotly, Bokeh and Tableau.

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## Data Visualization-cont'd

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### **What is data visualization?**

**Data visualization** is the graphical representation of information and data.



### **What makes for effective data visualization?**

Visualization **transforms data into images effectively and accurately** represent information about the data.



### **What are the advantages of data visualization?**

Makes for easier **interpretation of patterns and trends** as opposed to looking at data in a tabular/spreadsheet format.

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## About Data Visualization

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What Would You Like to Show?

- Relationships between variables
  - Composition of the data over time
  - Distribution of variable(s) in data
  - Comparison of data with relation to time, variables, categories, etc.
-

## Now, what you'll learn

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- How do you choose an appropriate plot?
- How do you interpret common types of plots?
- What are best practices for drawing plots?





## Continuous and categorical variables

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- Continuous: usually numbers
  - heights, temperatures, revenues
- Categorical: usually text
  - eye colors, countries, industry
- Can be either
  - age is continuous, but age group is categorical
  - time is continuous, month of year is categorical



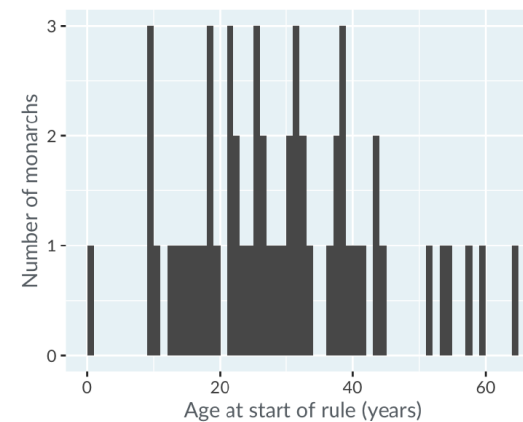
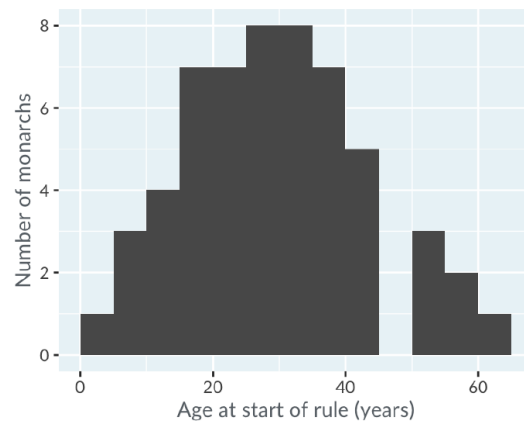
## Histograms

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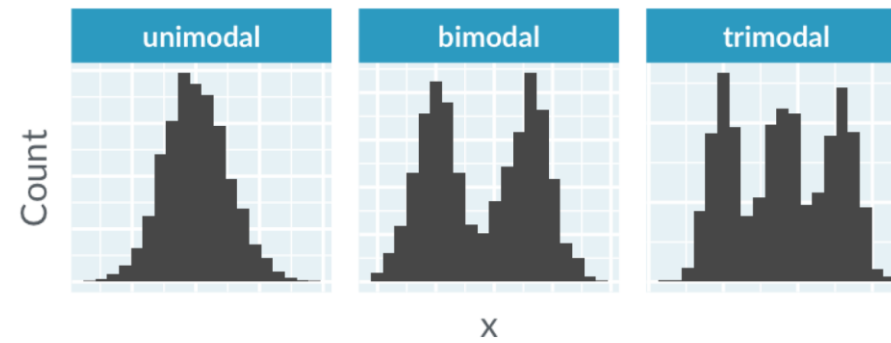
When should you use a histogram?

- 1) If you have continuous variable(s).
- 2) You want to ask questions about the shape of its distribution.

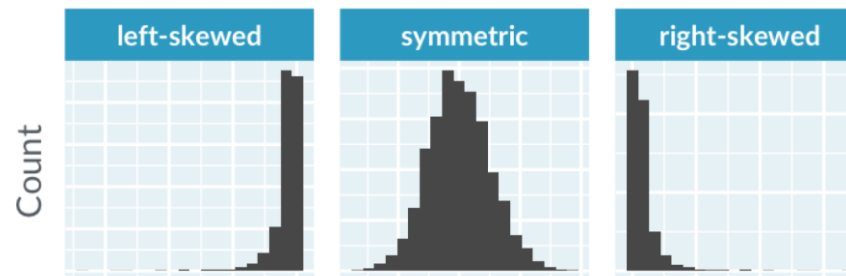


Modality: how many peaks?

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**Skewness: is it symmetric?**



## Box plots

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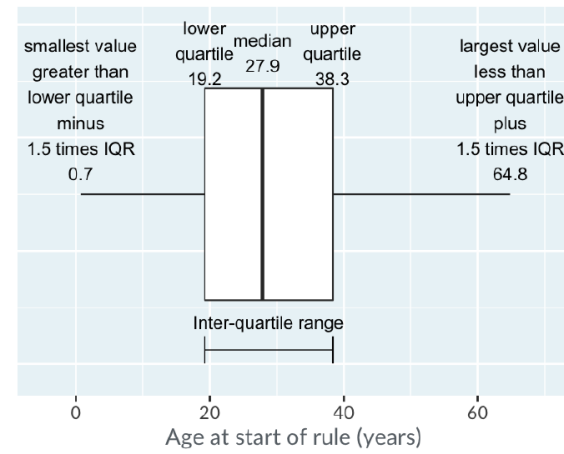
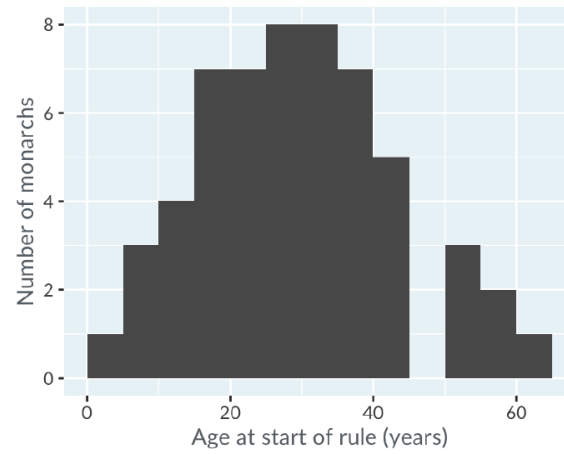


### When should you use a box plot?

- 1) When you have a continuous variable, split by a categorical variable.
- 2) When you want to compare the distributions of the continuous variable for each category.

## Histogram vs. box plot

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## Scatter plots

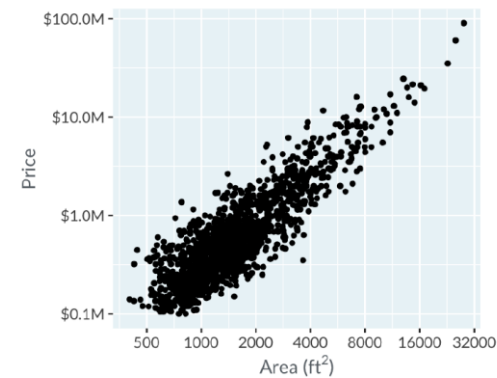
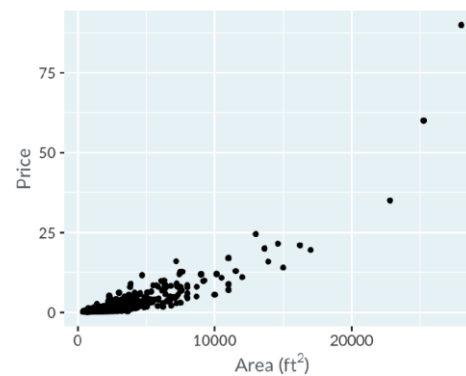
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When should you use a scatter plot?

- 1) You have two continuous variables.
- 2) You want to answer questions about the relationship between the two variables.

### Prices vs. area

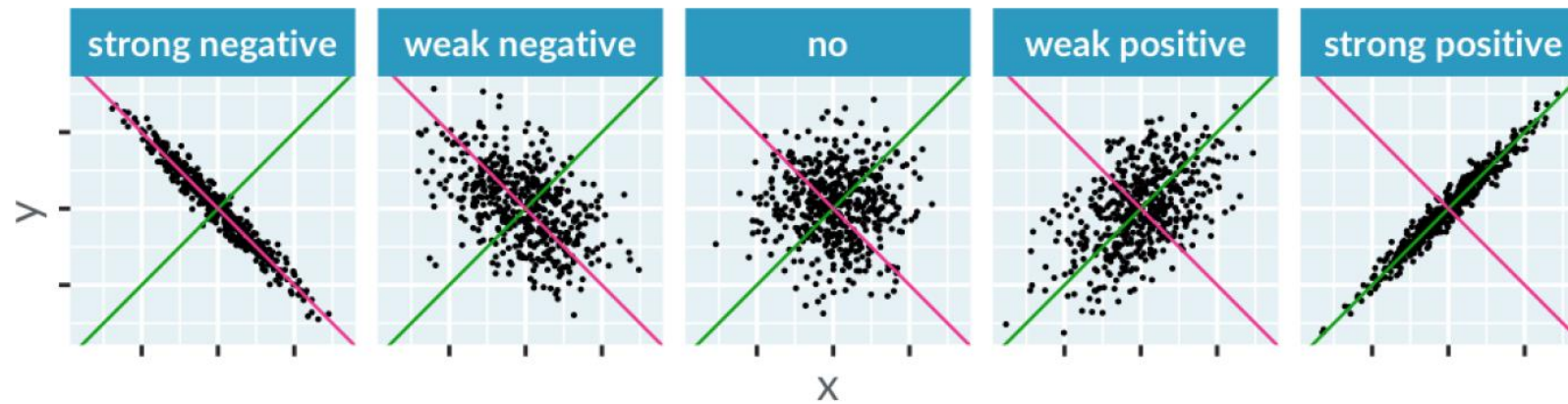


## Correlation

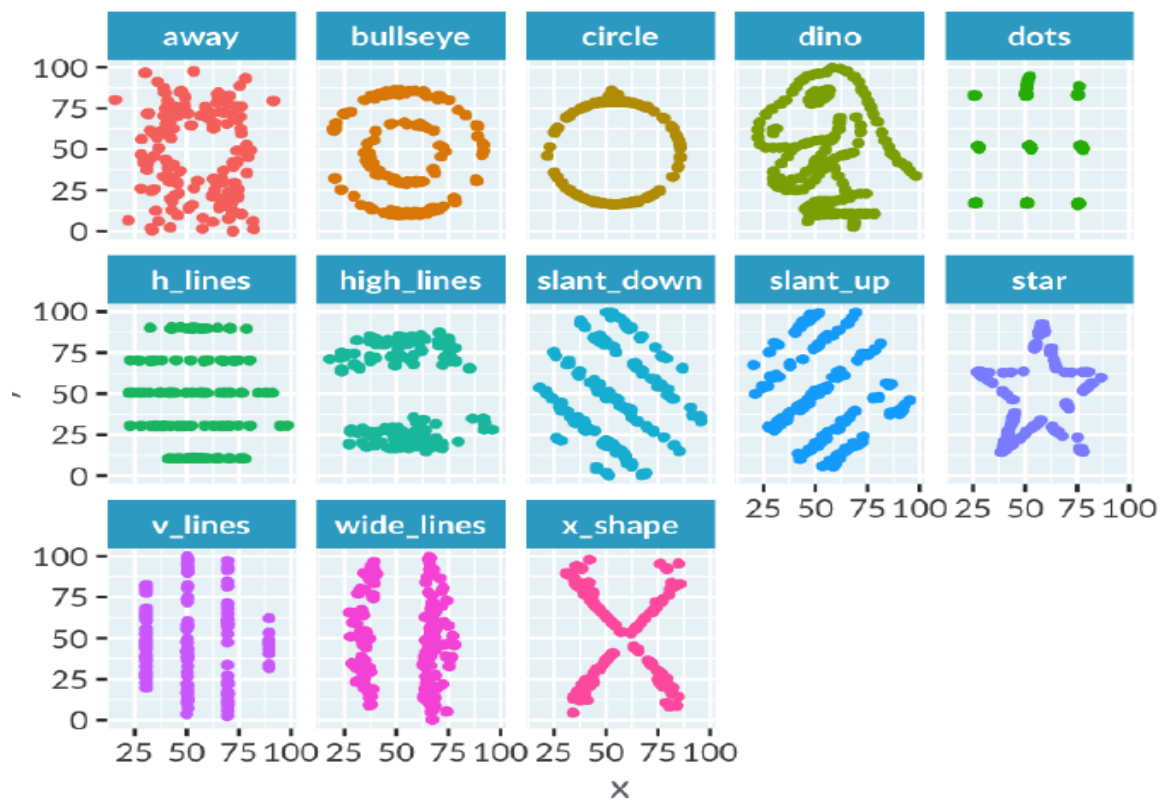
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Straight line through the points?



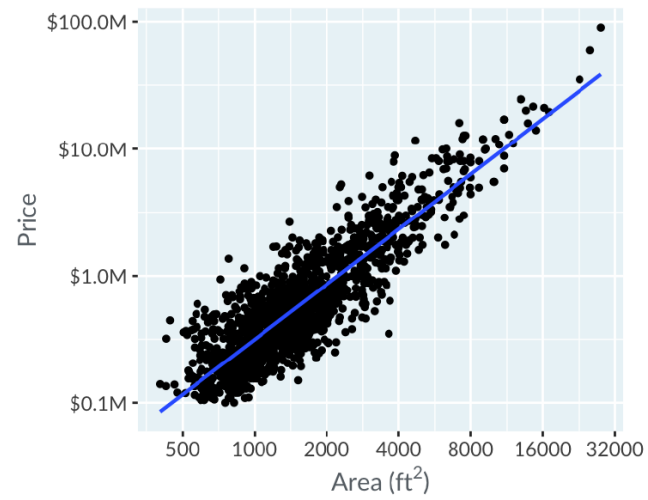
Sometimes correlation isn't helpful





## Adding trend lines

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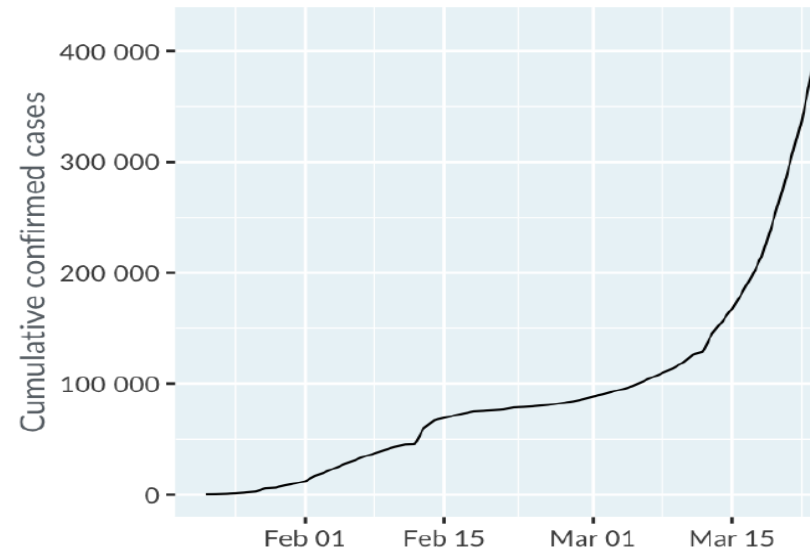
## Line plots

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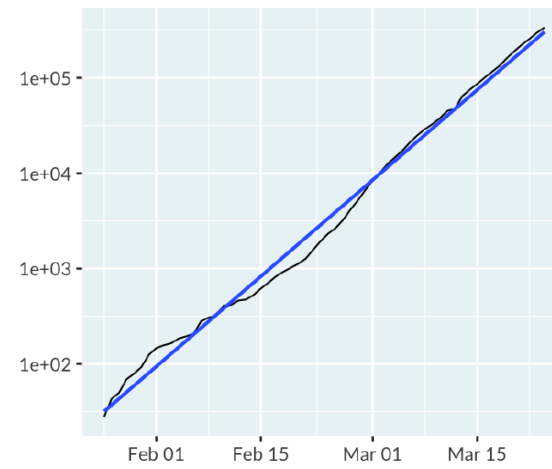
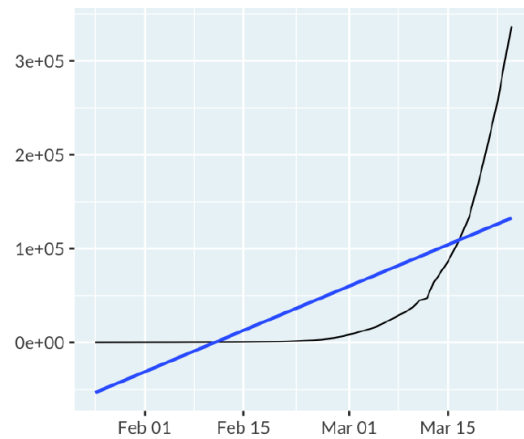
When should you use a line plot?

1. You have two continuous variables.
2. Consecutive observations are connected somehow. Usually, but not always, the x-axis is dates or times.



## Trend lines + log scale

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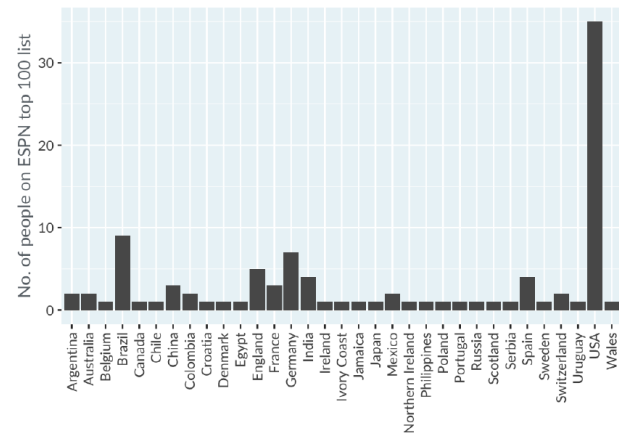
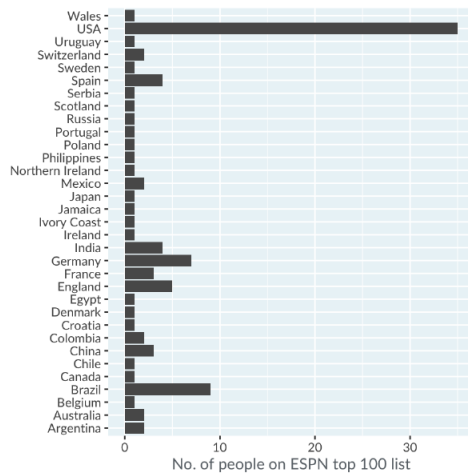


Bar plots:

## When should you use a bar plot?

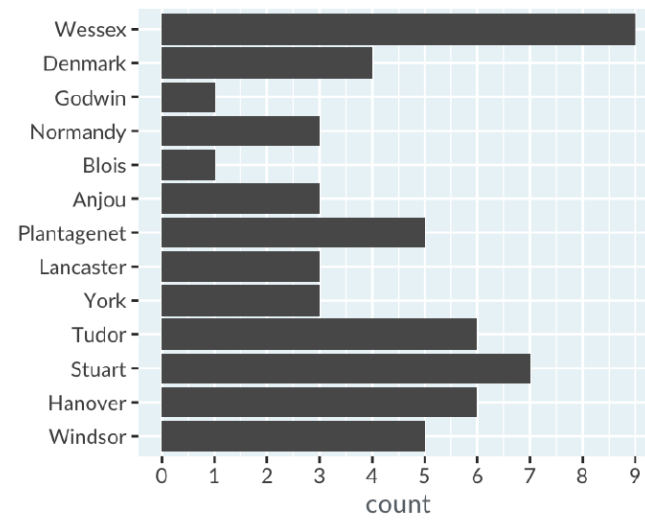
Most common cases:

- 1) You have a categorical variable.
- 2) You want counts or percentages for each category.



## Bar plots vs. box plots

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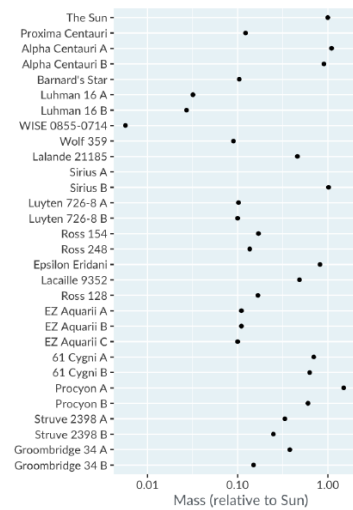


## Dot plots

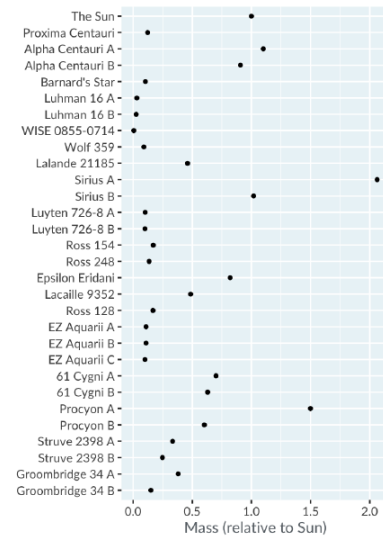
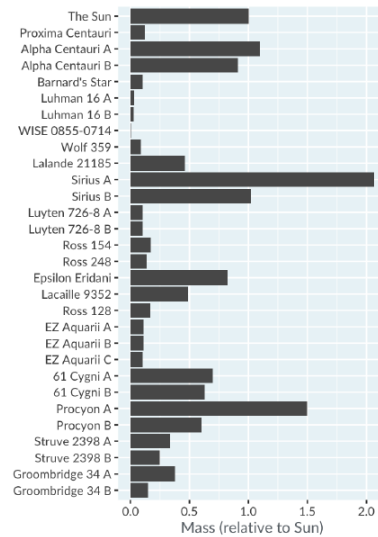
---

When should you use a dot plot?

- 1) You have a categorical variable.
- 2) You want to display numeric scores for each category on a log scale, or
- 3) You want to display multiple numeric scores for each category

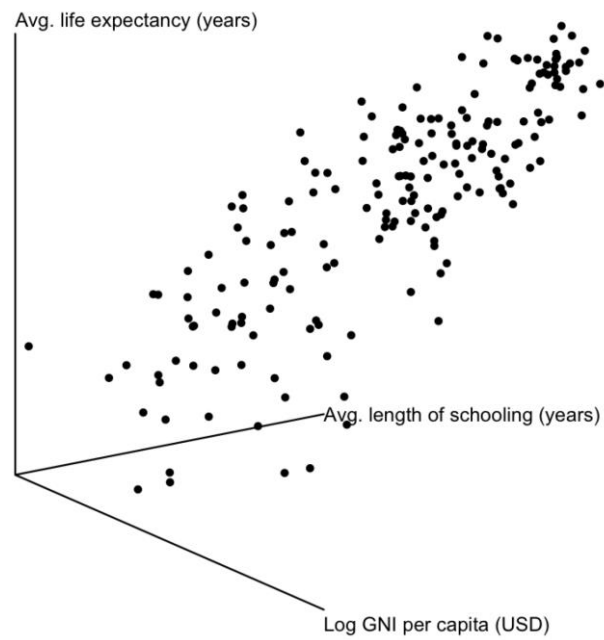


## Bar plot vs. dot plot



## Higher dimensions: 3D

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3D scatter plots

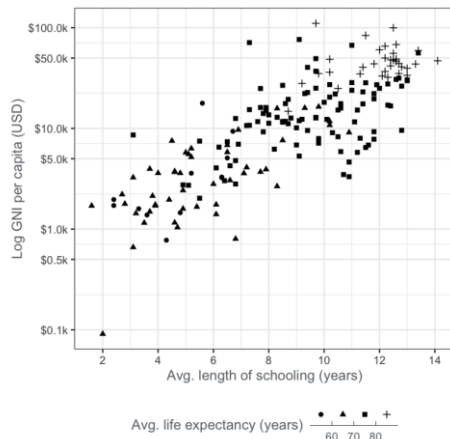




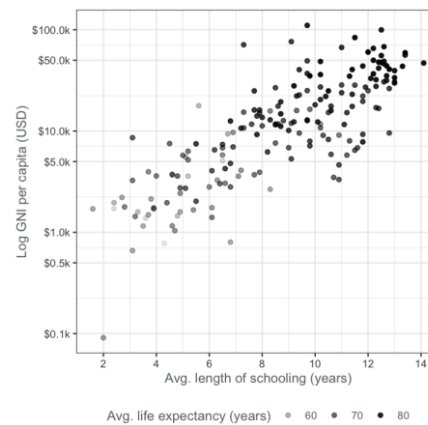
## x and y are not the only dimensions

- Points also have these dimensions
  - color
  - size
  - transparency
  - shape

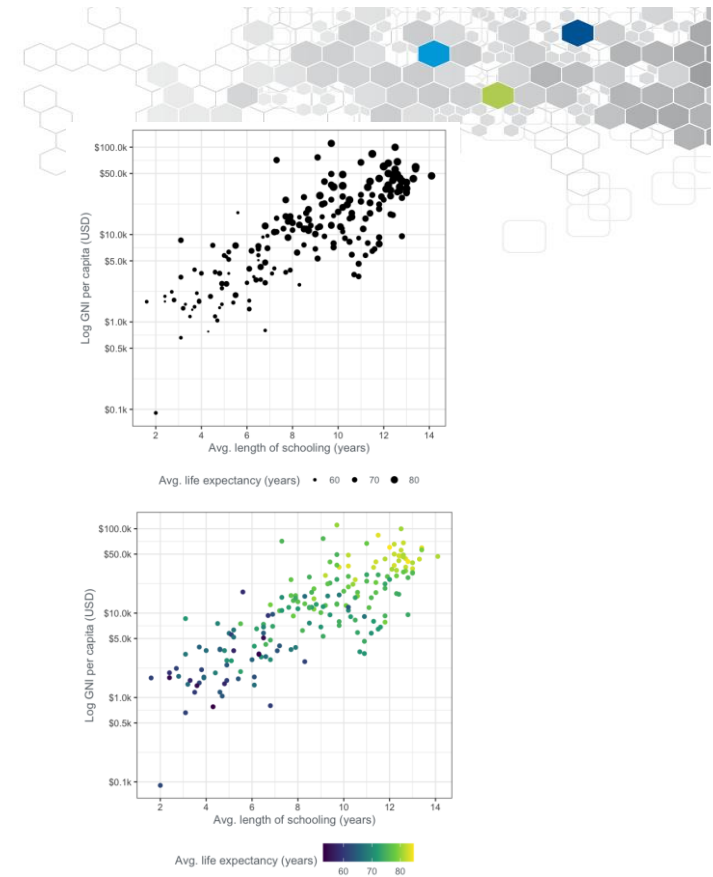
Size



Shape



Transparency

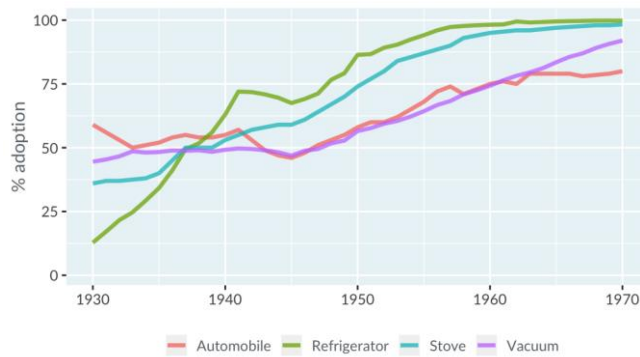


Color

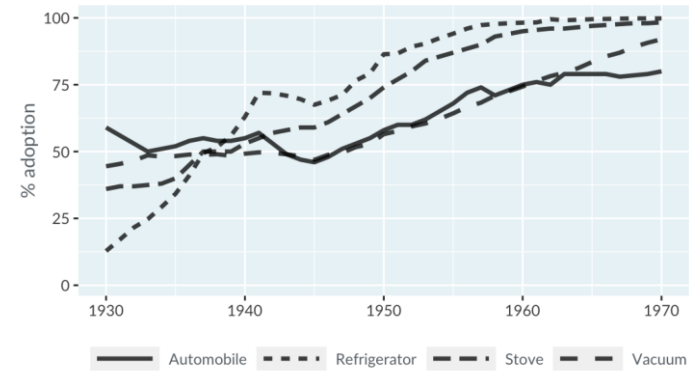
## Other dimensions for line plots

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- color
- thickness
- Transparency
- Line type (solid, dashes, dots)



Color



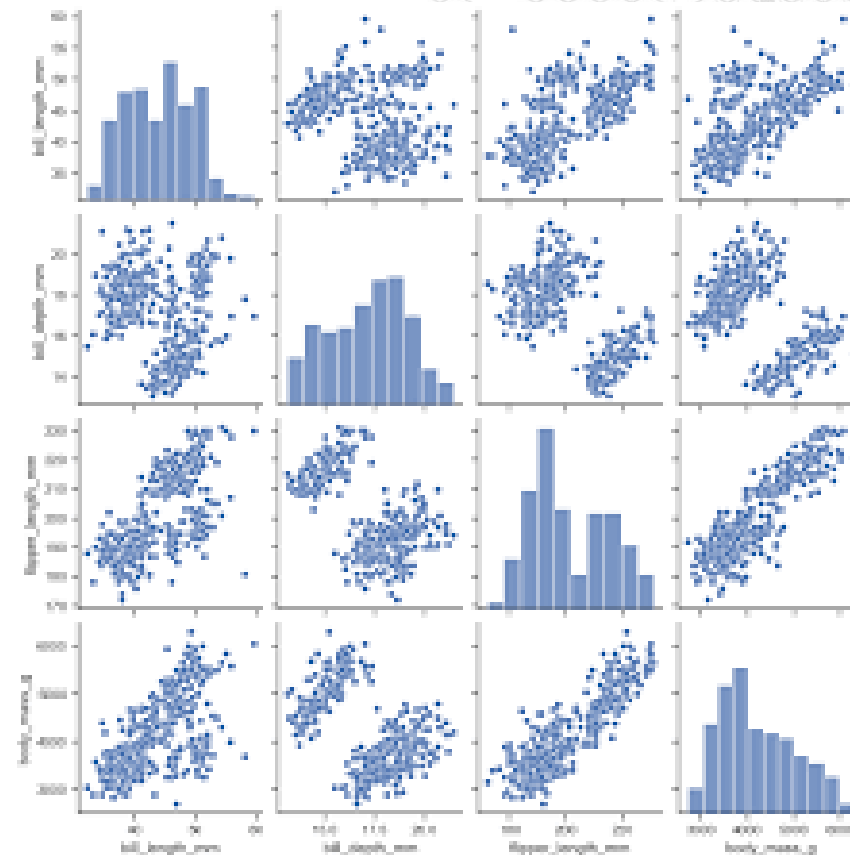
Linetype

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## Plotting many variables at once

When should you use a pair plot?

- You have a set of variables (either continuous, categorical, or a mix).
- You want to see the distribution for each variable.
- You want to see the relationship between each pair of variables.



## Correlation heatmap

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When should you use a correlation heatmap?

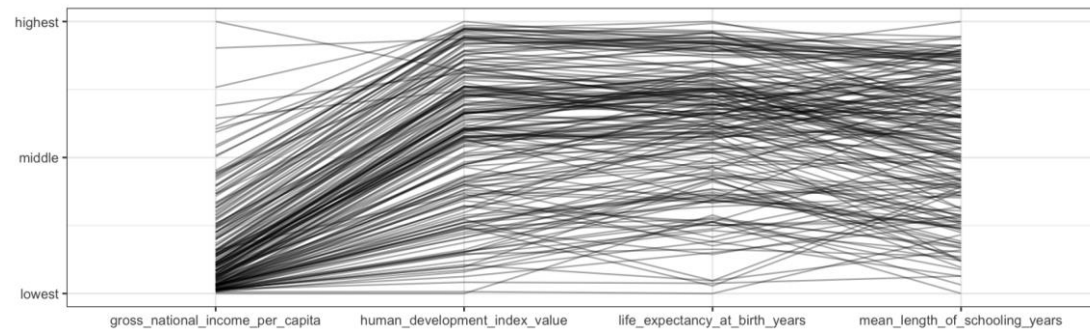
- You have lots of **continuous variables**.
- You want to a simple overview of how each pair of variables is related



## When should you use a parallel coordinates plot?

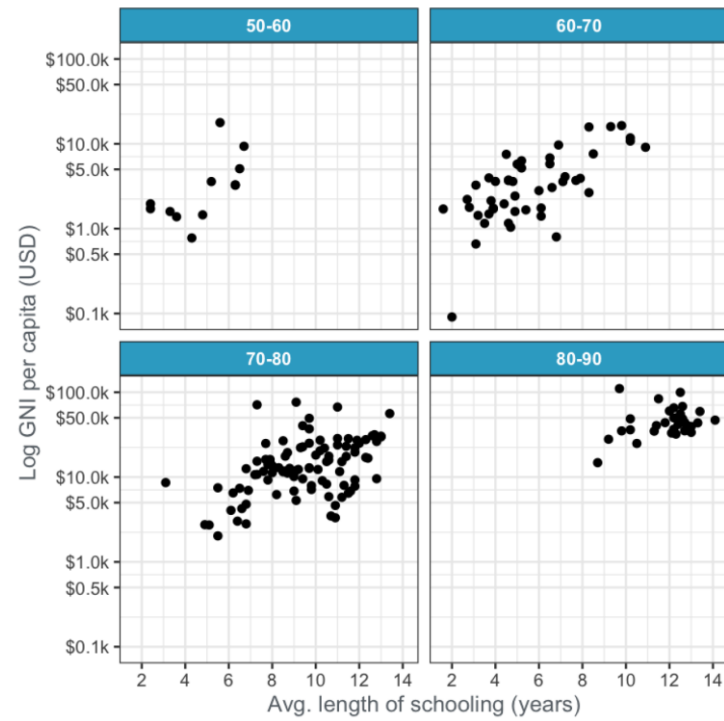
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- You have lots of continuous variables.
- You want to find patterns across these variables, or
- You want to visualize clusters of observations.



Lots of panels

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## Summary

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- Histograms: show a distribution
- Scatter plots: compare two numeric variables
- Line plots: show trends over time
- Bar plots: show counts by category
- Pair plot: compare many variables
- Correlation heatmap: show related variables
- Parallel coordinates plot: find patterns across variables



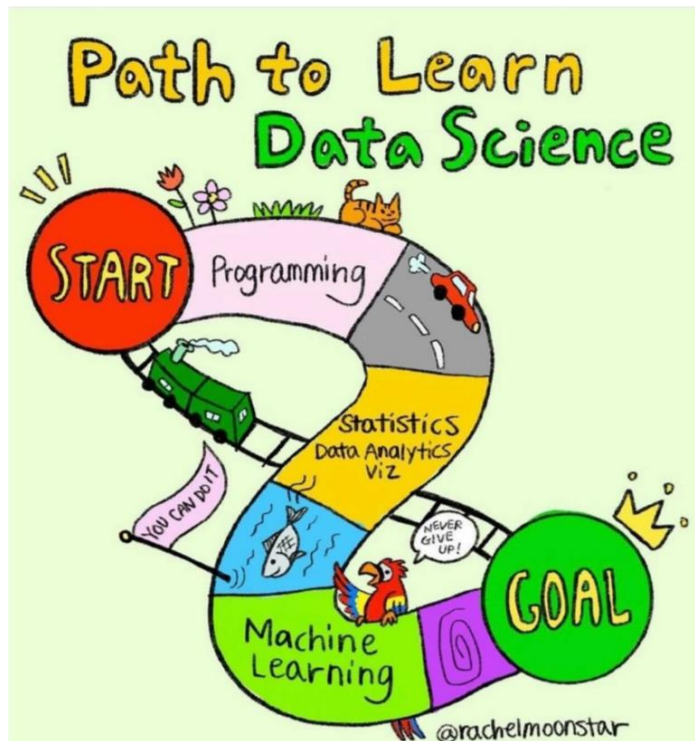


# **DATA MANIPULATION & VISUALIZATION**

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# Data Science



## Jobs!


**50 Best Jobs in America**

This report ranks jobs according to each job's overall Job Score, determined by combining three factors: number of job openings, salary, and overall job satisfaction rating.

Employers: Want to recruit better in 2017? [Find out how.](#)

United States 2017 12k Shares [f](#) [t](#) [in](#) [e](#)

**1 Data Scientist**



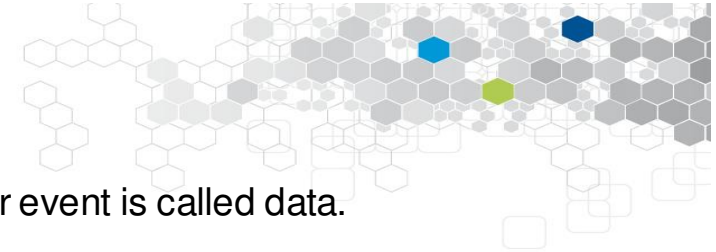
**4.8 / 5** Job Score  
**\$110,000** Median Base Salary  
**4.4 / 5** Job Satisfaction  
**4,184** Job Openings

[View Jobs](#)

**2 DevOps Engineer**

## Data versus Information

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- Data: Any observation collected in respect of any characteristic or event is called data.
  - Information
    - Raw data carry/convey little meaning, when it is considered alone.
    - The data is minimized: processed/analyzed and then presented systematically.
      - It is converted into Information.
  - Data, that is not converted into information is of little value for evaluation and planning and cannot be used by those who are involved in decision making.
-

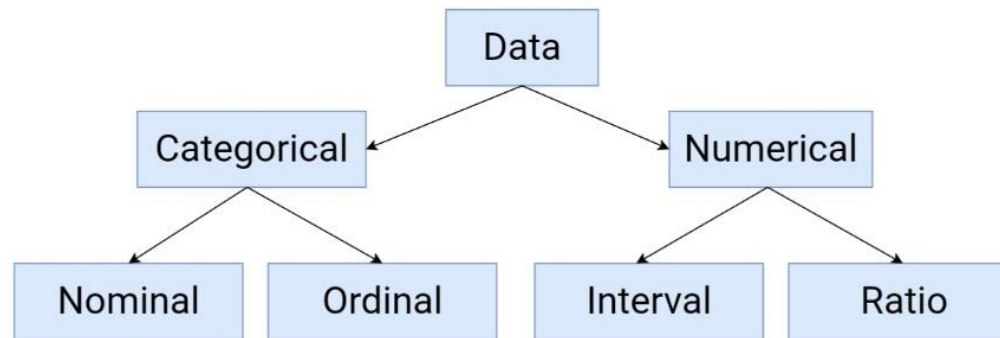
## Data Classification

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Classification data can be divided into two types

- Quantitative data (numerical);
- Qualitative data (descriptive, categorical/frequency count).



## Data Field

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A field, also known as a column, is a single piece of information from a record in a data set.

- Qualitative Field (Dimensions)
  - Describes or Categorizes Data
    - What, when or who
- Quantitative Field (Measures)
  - Numerical Data
  - Provides measurement for qualitative category
  - Can be used in calculations



Dimensions

Measures

## Quantitative Data vs Qualitative Data Field

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Quantitative Data has two types:

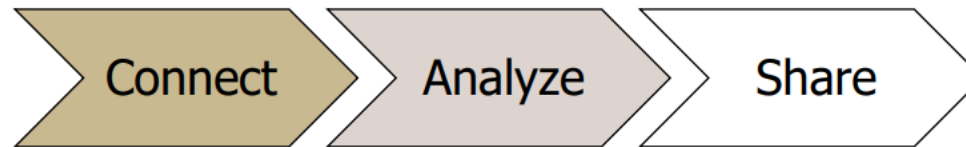
- (a) **Discrete**: Discrete variables can take only certain values.
- (b) **Continuous**: Continuous variables may take any value (typically between certain limits).

Qualitative Data is also called descriptive/ categorical data/ frequency count:

- When the data are arranged in categories on the basis of their quality and there is gap between two values,
  - Qualitative data is initially expressed in non-numerical forms.
-

## Data management

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- **Data source**

- Spreadsheets
  - Excel or csv file
- Relational Databases
  - MySQL or Oracle
- Cloud Data
  - • AWS or Microsoft Azure
- Other Sources

- **Visualize data in Workspace**

- **Dashboard or Story**

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## Data quality

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Data in the real world is dirty

- **Incomplete**: lacking attribute values, lacking certain attributes of interest, or containing only aggregate data,
  - **Noisy**: containing errors or outliers, e.g., Salary="-10"
  - **Inconsistent**: containing discrepancies in codes or names
    - e.g., Age="42" Birthday="03/07/1997"
    - e.g., Was rating "1,2,3", now rating "A, B, C"
    - e.g., discrepancy between duplicate records
-

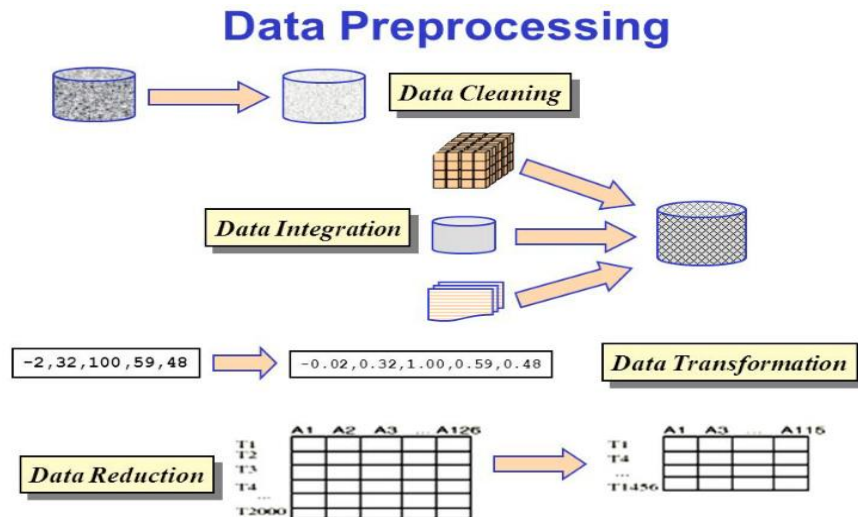
# Why Data Pre-processing?

Data Preprocessing is a technique used to convert the raw data into a clean data set.

- Steps are executed to convert the data into a clean data set.
- Data is gathered from different sources (collected in raw format and it is not feasible for the analysis).
- This technique is performed before the execution of Iterative Analysis.

- Steps of data-preprocessing:

- Data Cleaning
- Data Integration
- Data Transformation
- Data Reduction





## Data Cleaning

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Data quality is a main issue and occurs anywhere in information systems.

These problems can be solved by Data Cleaning:

- is a process used to determine inaccurate, incomplete or unreasonable data
- and then improve the quality through correcting of detected errors
- => reduces errors and improves the data quality.

Data Cleaning can be a time consuming and tedious process, but it cannot be ignored.

Data quality criteria :

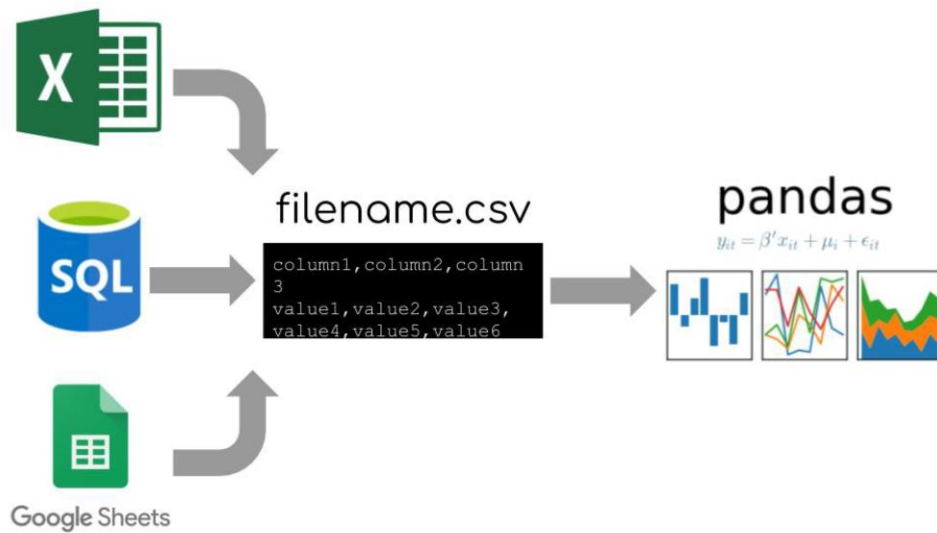
- accuracy, integrity, completeness, validity, consistency, uniqueness.
-

## Data Integration: Loading tabular data from different sources

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### CSV files



# Introduction to Databases

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A database consists of tables

Census				
state	sex	age	pop2000	pop2008
New York	F	0	120355	122194
New York	F	1	118219	119661
New York	F	2	119577	116413

State_Fact		
name	abbreviation	type
New York	NY	state
Washington DC	DC	capitol
Washington	WA	state

## Table consist of columns and rows

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Census				
state	sex	age	pop2000	pop2008
New York	F	0	120355	122194
New York	F	1	118219	119661
New York	F	2	119577	116413

Census				
state	sex	age	pop2000	pop2008
New York	F	0	120355	122194
New York	F	1	118219	119661
New York	F	2	119577	116413



## Tables can be related

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Census				
state	sex	age	pop2000	pop2008
New York	F	0	120355	122194
New York	F	1	118219	119661
New York	F	2	119577	116413

State_Fact		
name	abbreviation	type
New York	NY	state
Washington DC	DC	capitol
Washington	WA	state

## Useful Python Libraries for Data visualization

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NumPy



pandas



matplotlib



seaborn



bokeh

### Matplotlib:

- Provides the **building blocks** for seaborn's and pandas visualizations
- It can also be used on its own to plot data

### Pandas

- It is a foundational library for **analyzing data**
- It also supports **basic plotting** capability

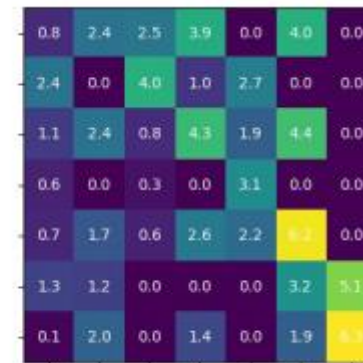
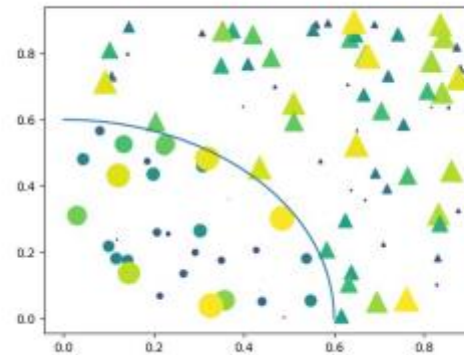
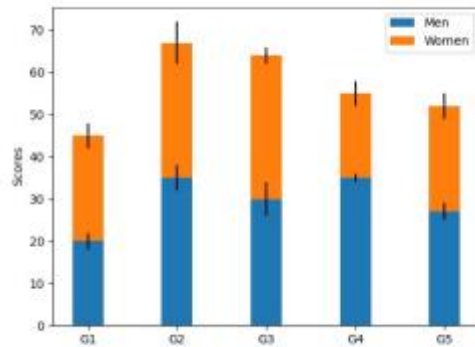
### Seaborn

- Seaborn supports **complex visualizations** of data
  - It is built on matplotlib and works best with pandas' dataframes
-

## Matplotlib

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- Used for basic plotting
- Highly customizable
- Works with NumPy and pandas



## About Matplotlib:

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- Matplotlib is a comprehensive library for creating static visualizations in Python.
- Usage: Matplotlib/Pandas is mostly used for quick plotting of Pandas DataFrames and time series analysis.

### Advantages of Matplotlib:

- Easy to setup and use.
- Very customizable.

### Limitations of Matplotlib:

- Visual presentation tends to be simple compared to other tools.
-



## About Seaborn:

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- Seaborn is a Python data visualization library based on Matplotlib.
- It provides a **high-level interface for drawing attractive and informative statistical graphics**.
- Usage: Those who want to create **amplified data visuals, especially in color**.

### Seaborn's Pros and Cons:

- Pro: Includes higher level interfaces and settings than does Matplotlib
  - Pro: Relatively simple to use, just like Matplotlib.
  - Pro: Easier to use when working with Dataframes.
  - Con: Like Matplotlib, data visualization seems to be simpler than other tools.
-

## Bokeh

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- Bokeh is an interactive visualization Python library.
  - Provides elegant and concise construction of versatile graphics.
  - Usage: Can be used in Jupyter Notebooks and can provide high-performance interactive charts and plots.
-

## Rules for variable names in Python

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### Rules for variable names

- Must start with a letter(usually lowercase)
- After first letter, can use letters/numbers/underscores
- No spaces or special characters
- Case sensitive ( `my_var` is different from `MY_VAR` )

```
# Valid Variables  
bayes_weight  
b  
bayes42
```

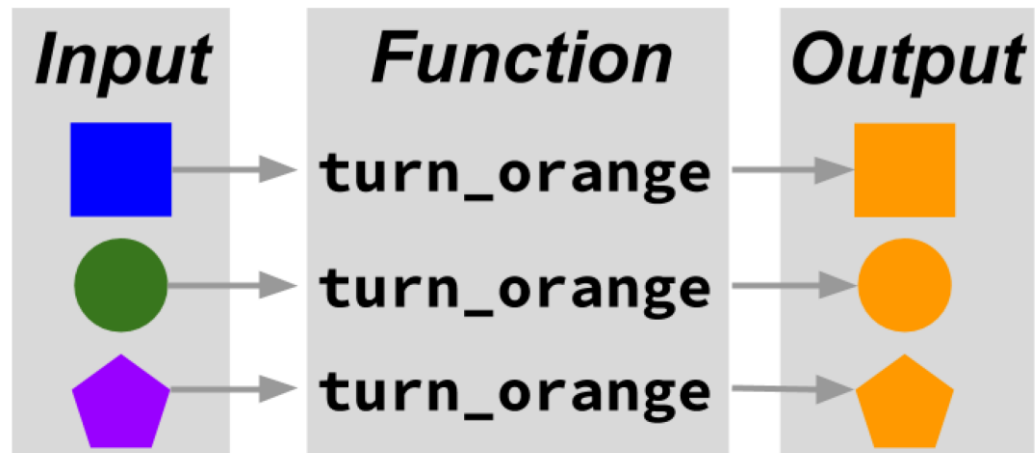
```
# Invalid Variables  
bayes-height  
bayes!  
42bayes
```

## Function

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A function is an action



- Function name is always followed by parentheses ()
-

## NumPy

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- Fundamental package for scientific computing
  - Exceptionally fast – written in C
  - Main data structure:
    - ndarray : n-dimensional arrays of homogeneous data types
  - Data manipulation  $\approx$  NumPy array manipulation
  - Used in other libraries - Matplotlib, pandas, scikit- learn
-

## Pandas

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- **Fundamental tool for handling and analyzing input data**
  - Particularly **suited for tabular data**
  - Implements powerful data operations
    - Easily read datasets from csv, txt, and other types of files
    - Datasets take the form of DataFrame objects
  - Main data structures:
    - DataFrame: A table with rows and columns
    - Series: A single column
-

## pandas Philosophy

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There should be one -- and preferably only one -- obvious way to do it.



What's the point of pandas?

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**pandas is built on NumPy and Matplotlib**

