

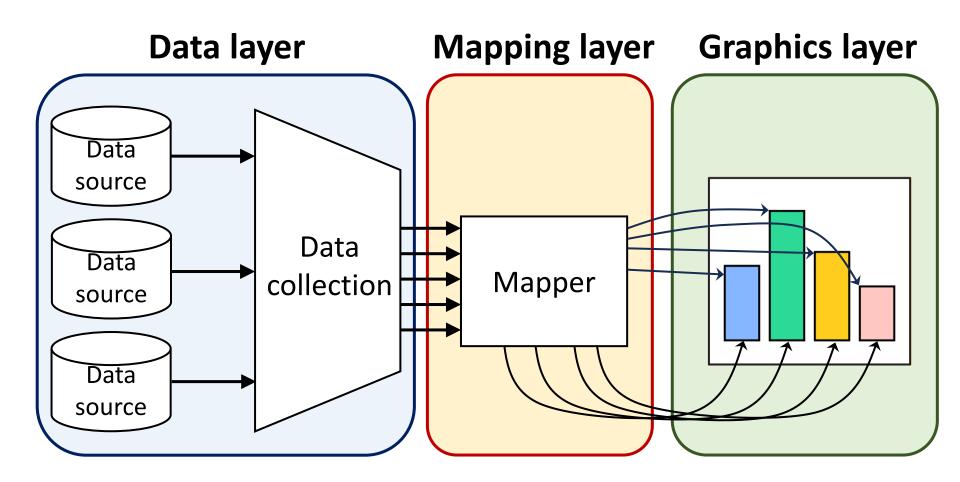
## Data Visualization

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#### Content outline

- Basic charts
- Advanced representations

#### Data visualization framework



#### Data visualization framework

- Locate and obtain data
- Import data in proper format
- Relate data for proper correspondence
- Data analysis and aggregation

- Associate appropriate geometry with corresponding data channels
- Data analysis and algorithms (e.g. contouring)

- Conversion of geometry into displayable image
- Decorations
- Managing interaction

Data layer

Mapping layer

**Graphics layer** 

### Data types

**Discrete** 

(no between values)

**Discrete** 

(values between)

**Ordered** 

(values are comparable)

**Ordinal** 

e.g., sizes: S, M, L, XL

**Quantitative** 

e.g., counts: 1, 2, 3, ...

**Fields** 

e.g., altitude, temperature

**Unordered** 

(values not comparable)

Normal

e.g., shapes:  $\square \bigcirc \triangle$ 

**Categories** 

e.g., nationality

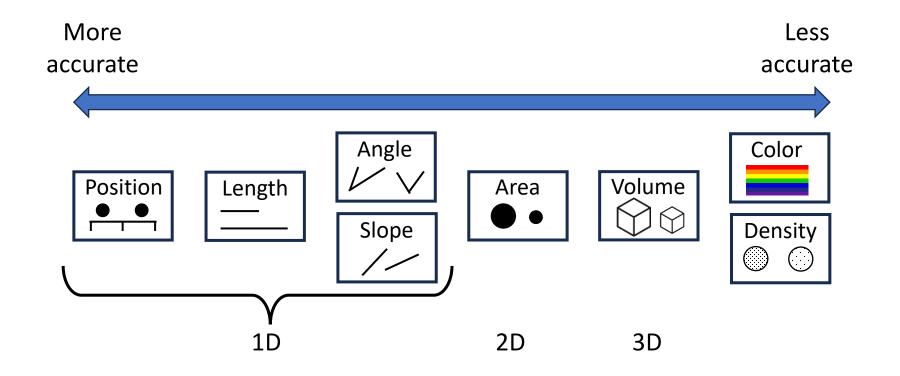
Cyclic values

e.g., directions, hues

### Data as variables

Science	Databases	Data warehouses
Independent variable	Key	Dimension
Dependent variable	Value	Measure

### Ranking of perceptual tasks

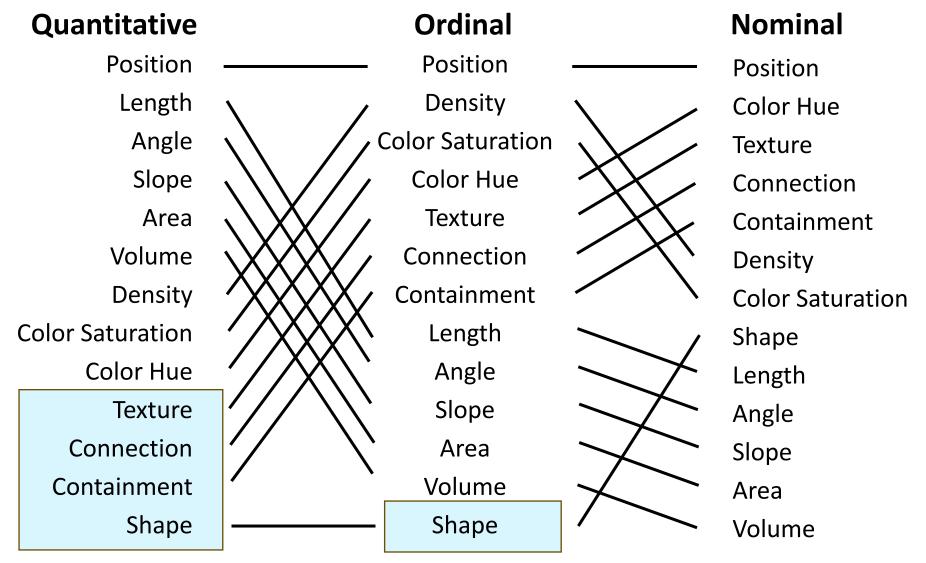


Accuracy ranking of quantitative perceptual tasks.

Higher tasks are accomplished more accurately than lower tasks.

Cleveland and McGill empirically verified the basic properties of this ranking.

Cleveland, W. S., and McGill, R. Graphical perception: Theory, experimentation and application to the development of graphical methods. Journal of the American Statistical Association, 79(387) 1984



Ranking of perceptual tasks.

The tasks shown in the gray boxes are not relevant to these types of data.

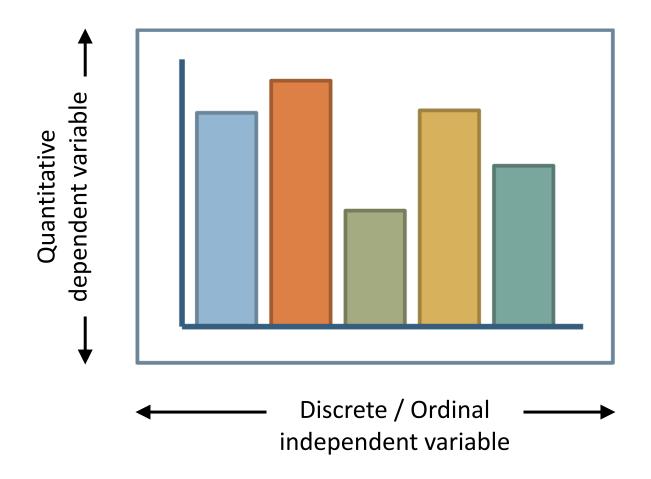
J. Mackinlay, Automating the Design of Graphical Presentations of Relational Information, ACM Transactions on Graphics 5(2), 1986

## Basic charts

#### Bar charts

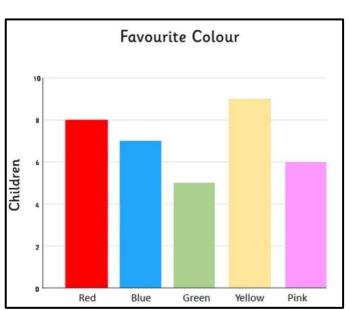
Position Length

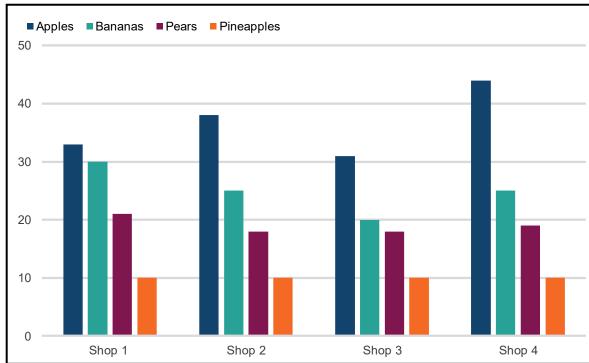
Bar charts benefit from both the position (bar top) and length (bar size).



### Bar charts: Examples





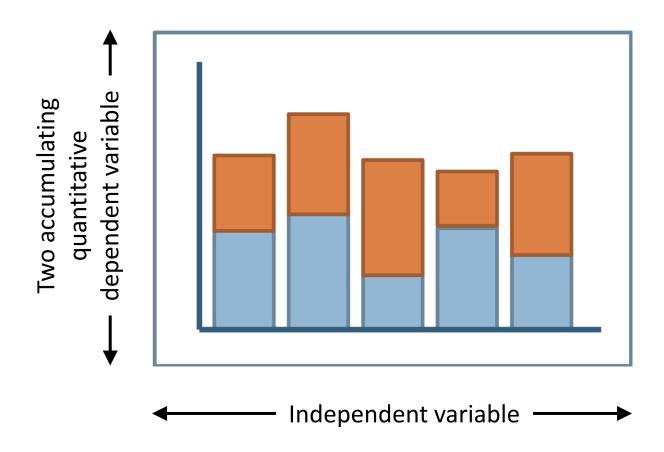


Bar chart for a single attribute (left) and bar chart for contrasting four attributes (right).

#### Stacked bar charts

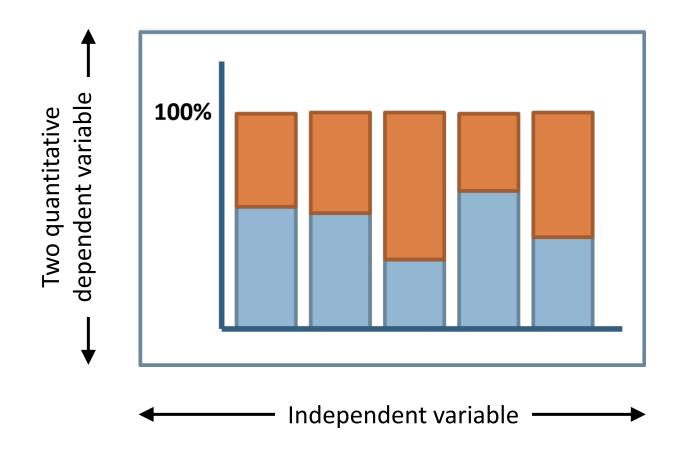
Position Length

Central limit theorem → as more bars are added, sums will vary less.

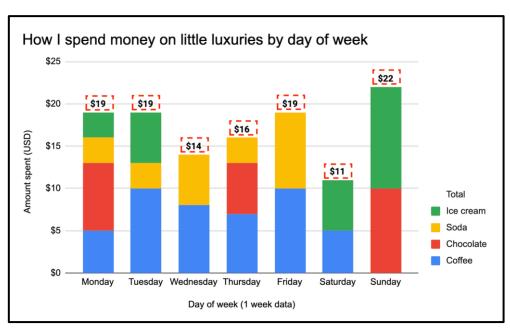


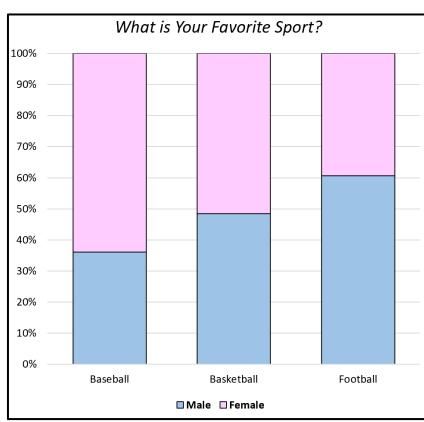
#### Relative stacked bar charts

Position Length



### Stacked bar charts: Examples

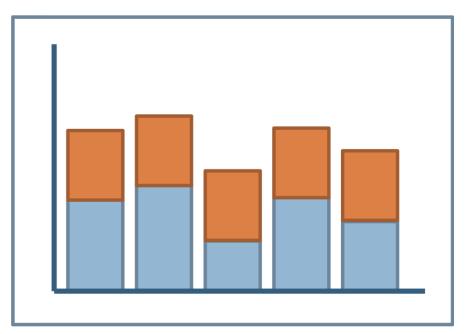


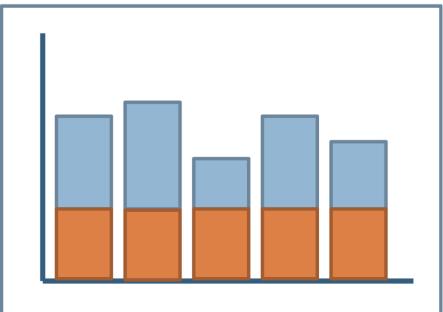


Stacked bar chart (left) and relative stacked bar chart (right)

### Stacking order matters

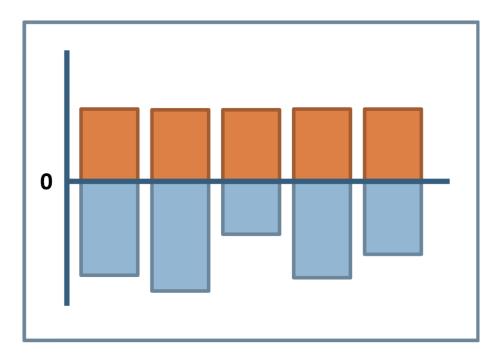
Variance of lower stack elements influences perception of upper elements.

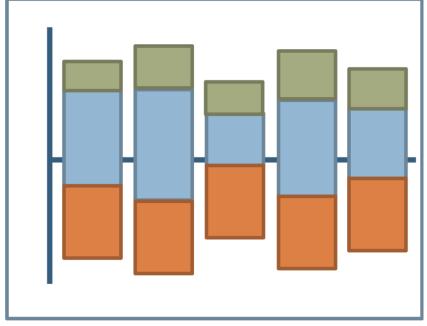




**Position > Length** 

### Diverging stacked bar charts



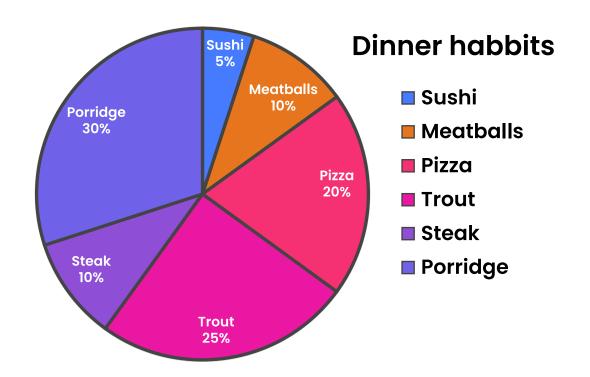


- Benefits from position and length
- Only work for two variables
- Negative connotation for lower bars

- Only indicates length
- Work for many variables
- Bar trends can still be obscured by neighboring bar variance

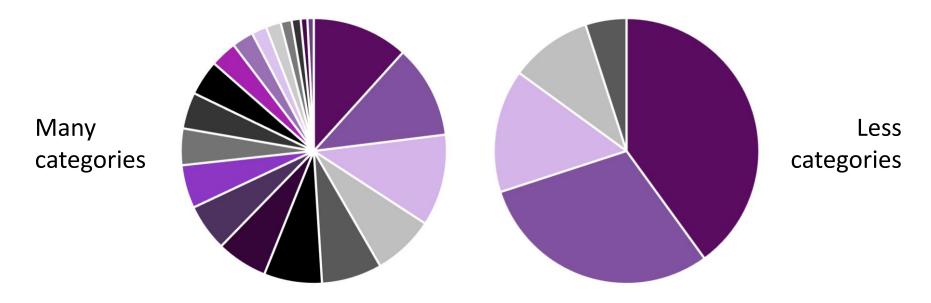
#### Pie charts

 Pie charts indicate relative portions of a quantitative dependent variable of a single dimension via the angles of slices, which sum to 100%.



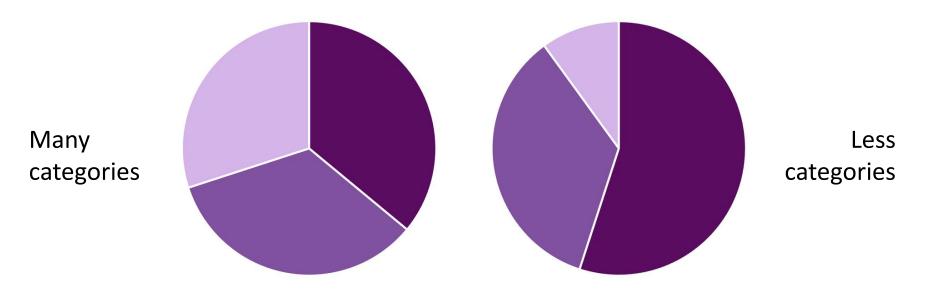
A pie chart showing several options for dinner habbits.

 Drawback: When there are too many categories, the pie chart becomes less intuitive.



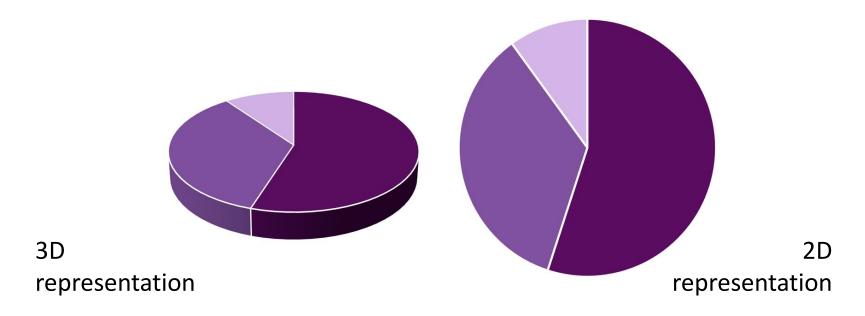
 Fix: group smaller or alike categories to reduce the number of slices or consider bar and column charts as better alternatives

• **Drawback:** It is difficult to distinguish similar-sized slices as angles are harder to interpret than lengths.



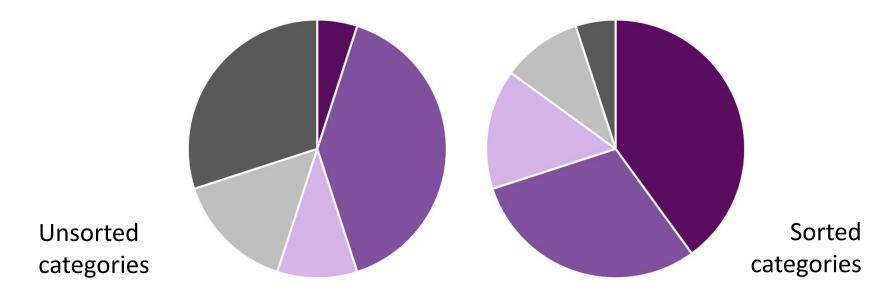
Fix: Bar charts are better for close comparisons between categories.

• **Drawback**: The 3D perspective can distort slice sizes, making some appear larger based on their position.



• Fix: Use 2D pie charts for easier interpretation.

• **Drawback:** Without the logical ordering of categories (e.g., largest to smallest) it becomes difficult to extract meaningful insights from data.



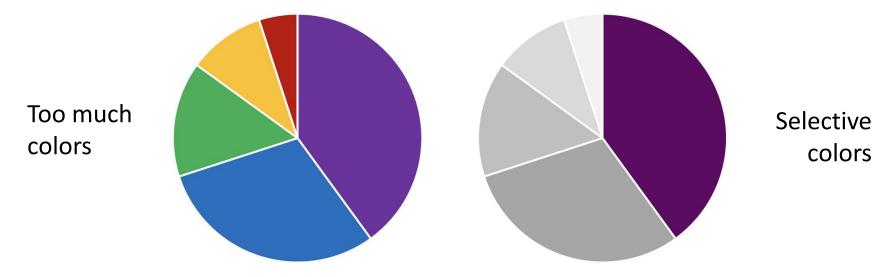
 Fix: Order categories from largest to smallest improves the readability of pie charts.

 Drawbacks: Using pie charts to visualize non-proportional data (i.e., proportions exceeding 100%) often leads to confusion.



Fix: Use alternative data visualizations, such as bar or column charts.

- Too much colour used can detract from the message of the pie chart.
- Additionally, certain colour combinations can be difficult to distinguish, especially for those with colour vision deficiencies.



- Use colour strategically to highlight the most important point in the chart.
- Applying muted tones, such as greyscale, to less relevant data allows the primary colour and key message to stand out.

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#### **Position**

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#### Line charts

Line charts benefit from position but not length.

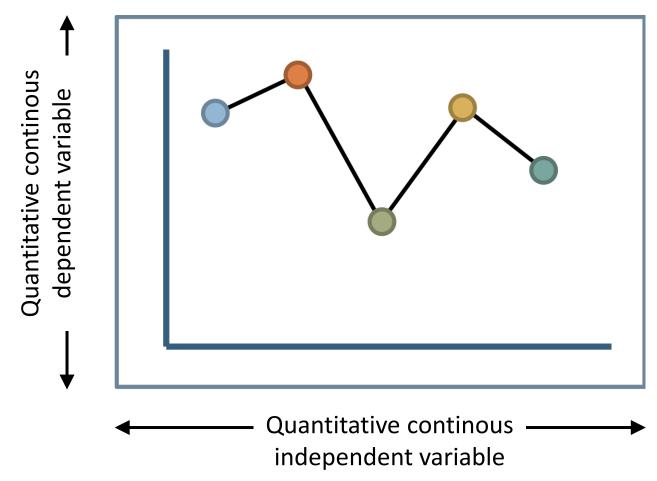
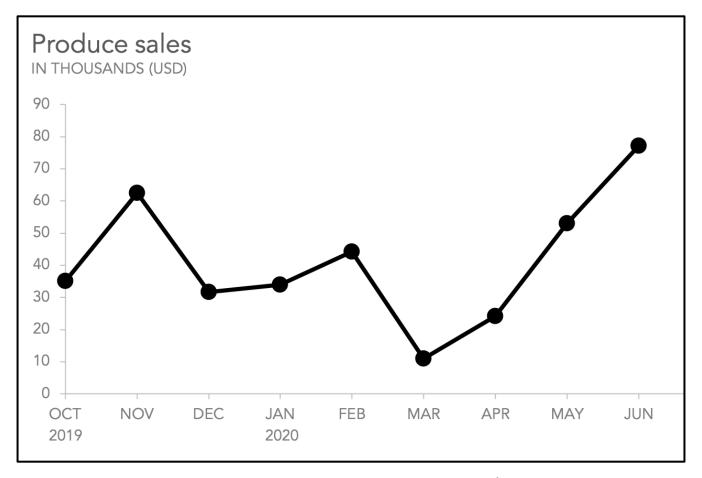


Image credit: Engnovate

### Line charts: Examples



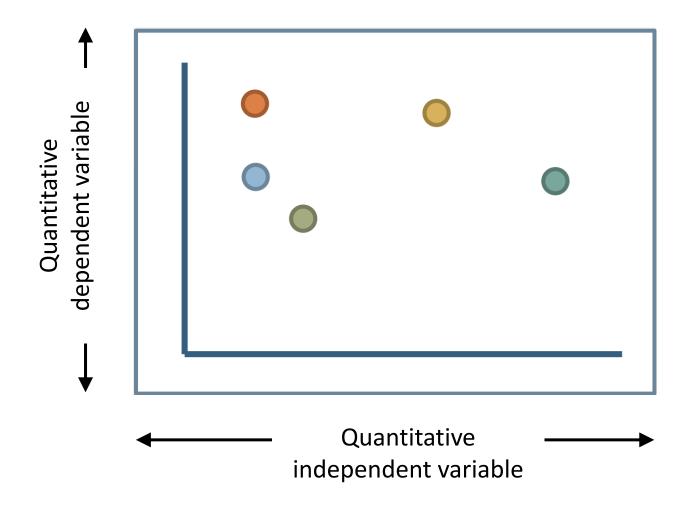
A line chart that show product sales (in thousands USD) for each month from Oct 2019 to Jun 2020.

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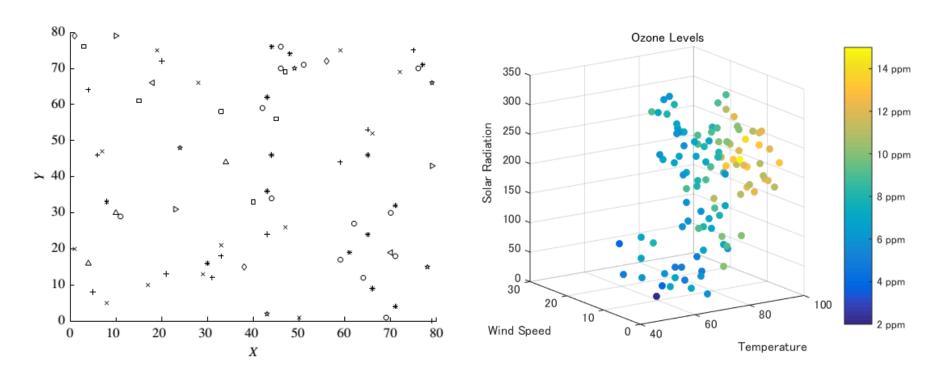
### Scatter plots

Position Density

• A scatter plot relies mostly on position, but clusters also yield density.



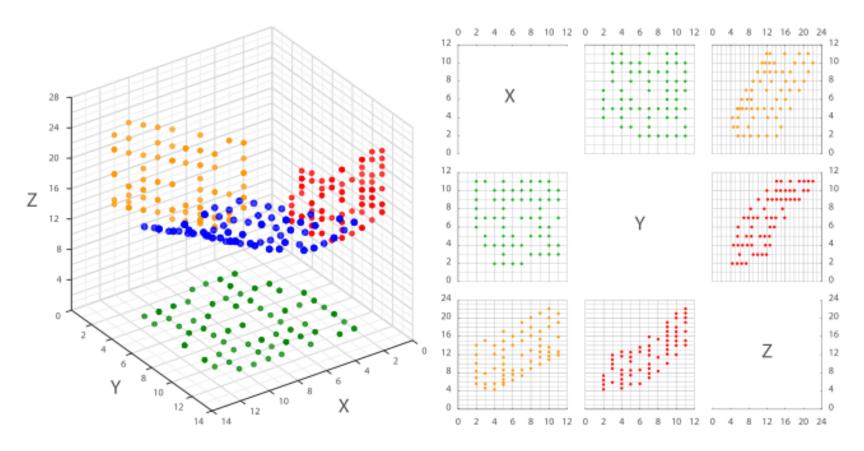
### Scatter plot



2D scatter plot (left) and 3D scatter plot (right)

### Scatter plots: Scatter-plot matrix

• A matrix of 2D scatterplots for all pairs of dimensions from k-dimensions.

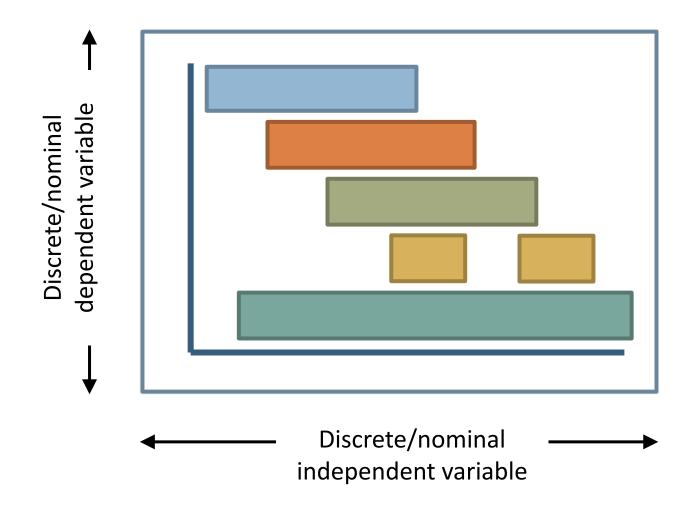


Visualization of 3D data (left) and the correspondent scatterplot matrix (right).

#### Gantt charts

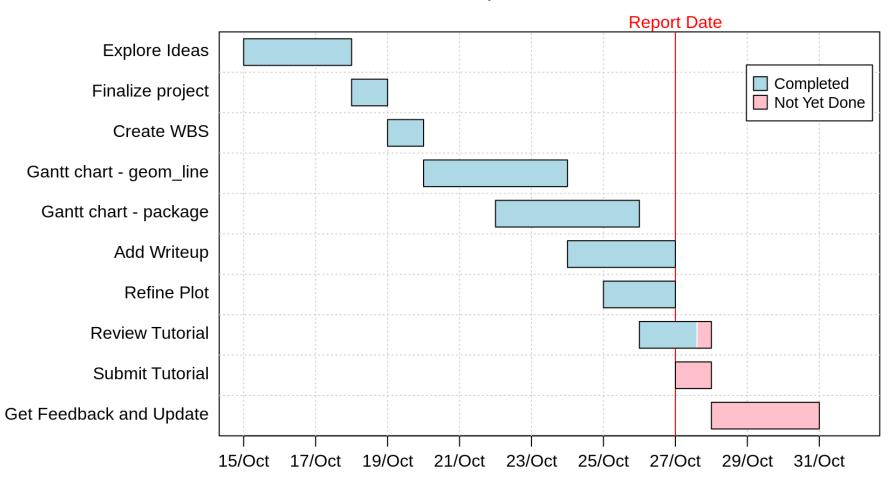
Position Length

A Gantt chart benefits from both position and length.



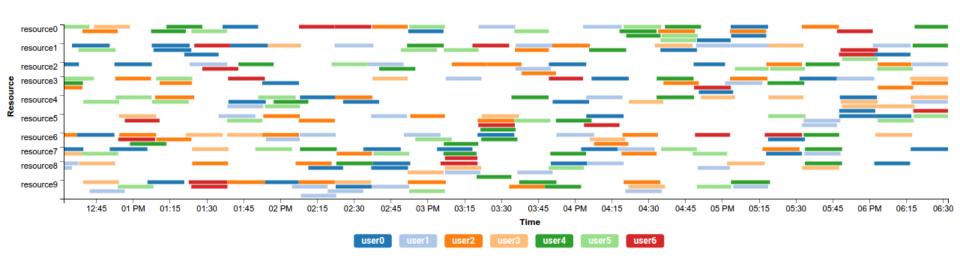
### Gantt charts: Examples

**Community Contribution Gantt Chart** 



A Gantt chart showing the tasks of a project (usually order in time), how long each task takes, and whether a task is fully completed.

### Gantt charts: Examples

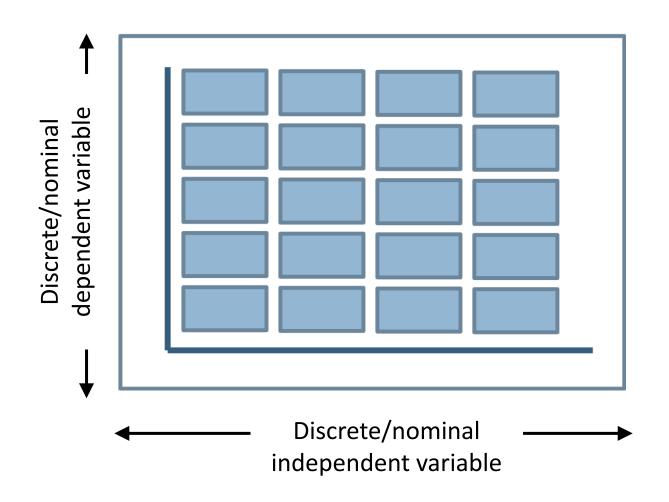


A Gantt chart can be quite complicated with several entities and detailed timeline.

**Position** 

### **Tables**

• A table benefits from position only.



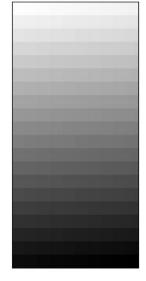
### What chart to use?

Donandant	Quantitative continuous	Bar	Line
Dependent	Quantitative discrete	Bar	Bar
	Quantitative continuous	Gantt	Scatter
Independent	Quantitative discrete	Scatter	Gantt
		Nominal or Quantitative discrete	Quantitative continous
		Independent	

# Advanced charts

#### Pixel-oriented visualization

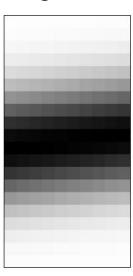
- Consider a dataset of m dimensions and n examples.
- Create a window of n pixels for each dimension
- For each example, the  $i^{th}$  value is mapped to the matching pixel in the  $i^{th}$  window.
- The colors of the pixels reflect the value range.



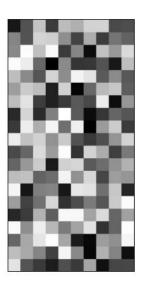
income



credit limit



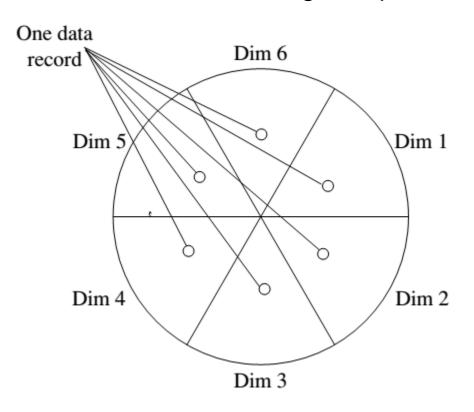
transaction volume

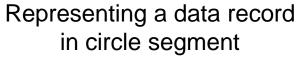


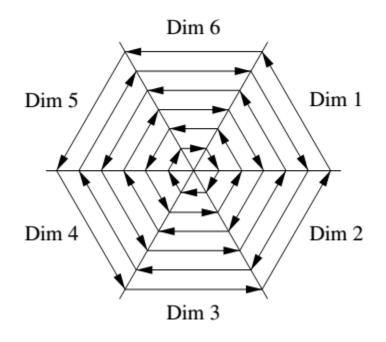
age

### Laying out pixels in circle segments

 Space-filling is often done in a circle segment to save space and show the connections among multiple dimensions.

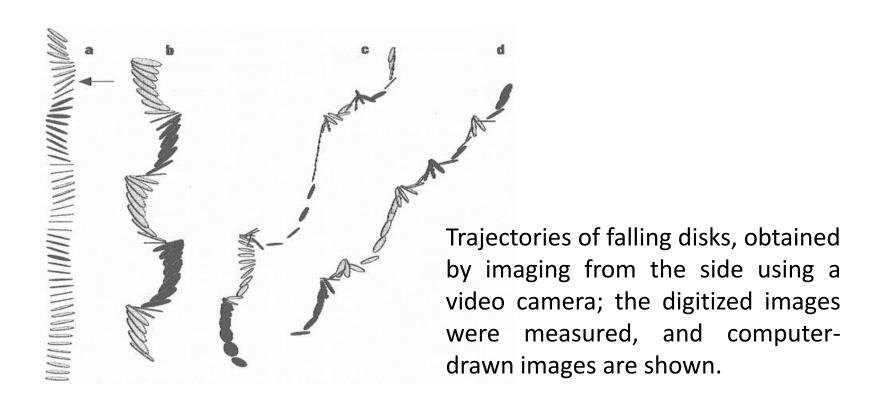






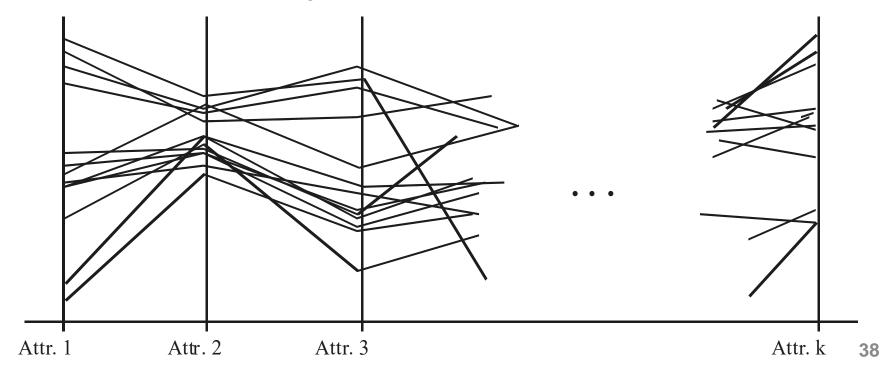
Laying out pixels in circle segment

## Direct visualization



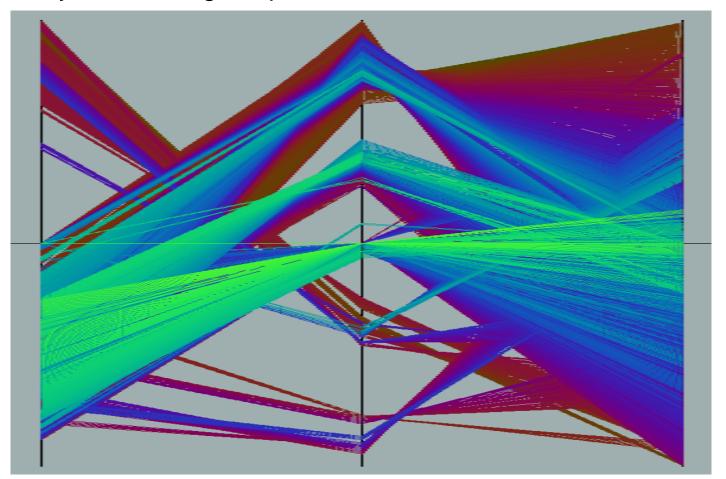
### Parallel coordinates

- n equidistant axes, one for each dimension, parallel to one of the display axes
  - Each axis is scaled to the domain of the corresponding attribute
  - A data record is represented by a polygonal line that intersects each axis at the point corresponding to the associated dimension value.



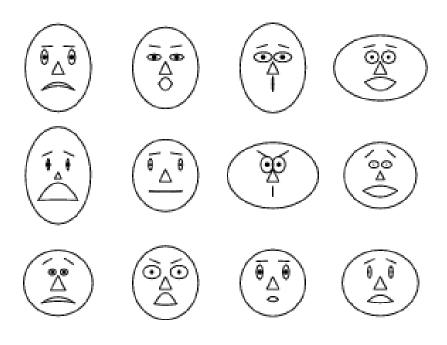
# Parallel coordinates

• Parallel coordinates suffer visual clutter and overlap, often reducing the readability and making the patterns hard to find.



### Icon-based visualization

• Chernoff Face: display variables on a two-dimensional surface, e.g., let *x* be eyebrow slant, *y* be eye size, *z* be nose length, etc.

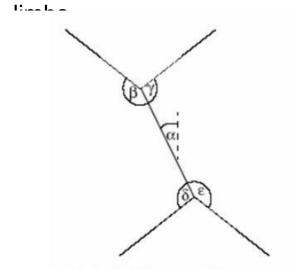


Faces produced using 10 characteristics-head eccentricity, eye size, eye spacing, eye eccentricity, pupil size, eyebrow slant, nose size, mouth shape, mouth size, and mouth opening): Each assigned one of 10 possible values, generated using Mathematica (S. Dickson)

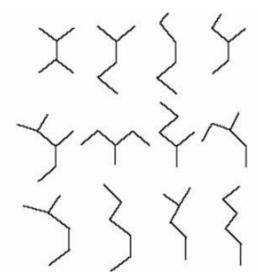
Reference: Gonick, L. and Smith, W. <u>The Cartoon Guide to Statistics.</u> New York: Harper Perennial, p. 212, 1993

### Icon-based visualization

- Stick figures: Each example is a five-piece stick figure icon with one body and four limbs.
  - Two variables are mapped to X-Y axes
  - The remaining attributes are mapped to the angles and/or length of the

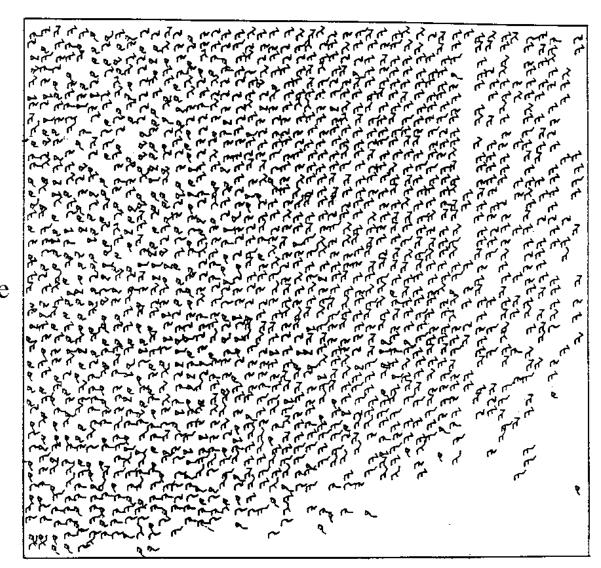


A stick figure icon



A family of stick figures

## Icon-based visualization

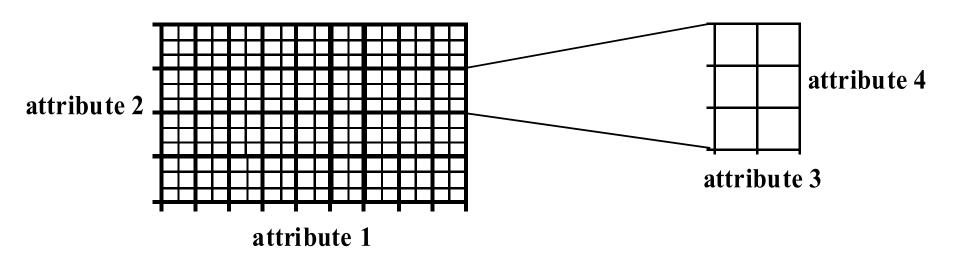


US census data showing age, income, gender, education, etc.

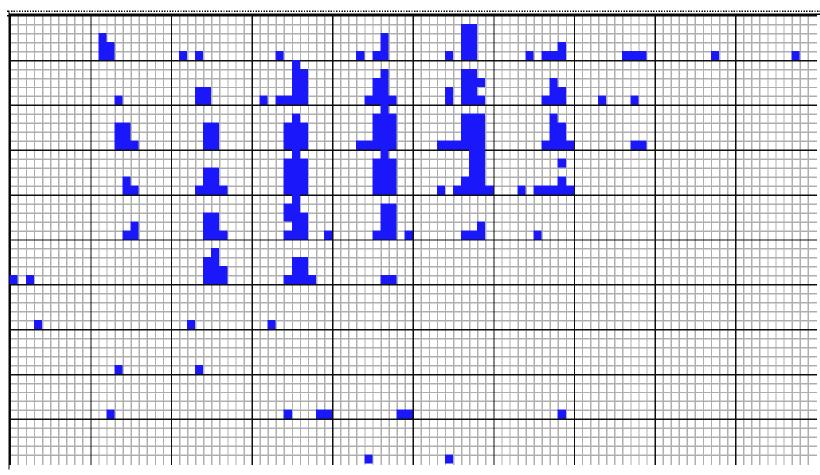
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# Dimensional stacking

- Partition the n-dimensional attribute space in 2-D subspaces, which are 'stacked' into each other
  - The important attributes should be used on the outer levels.
  - Adequate for data with ordinal attributes of low cardinality, yet difficult to display more than nine dimensions

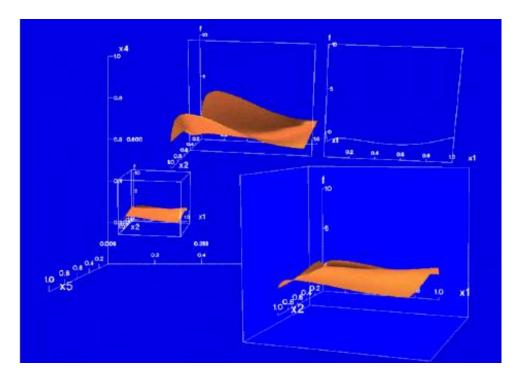


# Dimensional stacking: An example



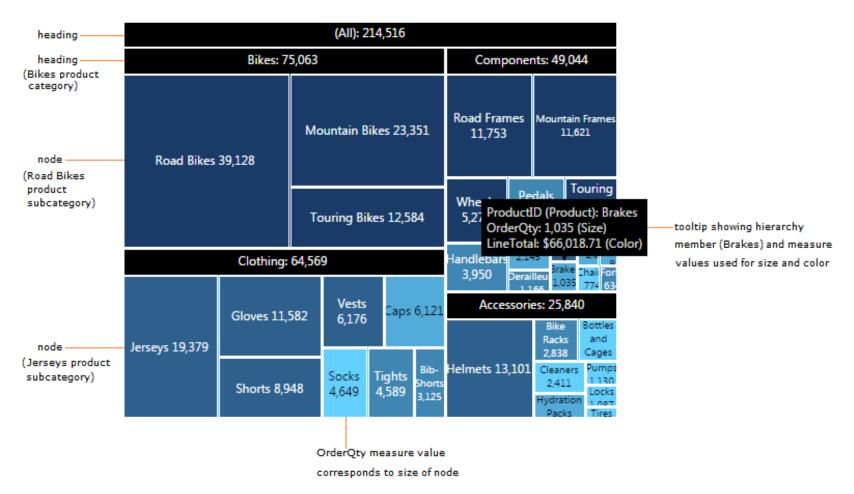
### Worlds-within-Worlds

- Show an attribute *F* changes with respect to the other dimensions in a multi-dimensional dataset.
  - Consider a 6D dataset, where the dimensions are F, X<sub>1</sub>, ..., X<sub>5</sub>, observe how F changes with respect to the other dimensions
  - Fix the values of dimensions  $X_3, X_4, X_5$  to some values, say,  $c_3, c_4, c_5$ .
  - Then visualize F,  $X_1$ ,  $X_2$  using a 3D plot whose origin locates at the point  $(c_3, c_4, c_5)$



# Treemap

Hierarchically partition the screen into regions depending on the attribute values

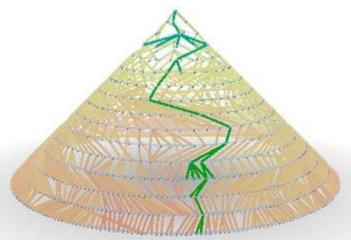


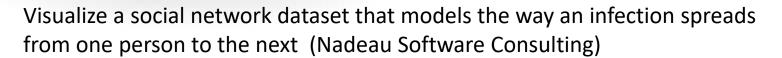
### 3D Cone Tree

 Typically contain more information and hierarchical data than tree diagrams due to its 3D nature

Work well for up to a thousand nodes or so, yet suffering overlaps when

projected to 2D





# Tag cloud

 Visualize statistics of user-generated tags, whose levels of importance are indicated by font sizes or colors

Nap playing wife Watching to kids Talking Bible dogs food meditation

Staying home Walking helped relax nothing resting friends

projects Exercise crafts music Taking Reading baking

family movies Sleep walks outside TV outdoors time gardening

relaxed sex Prayer quiet working people spending time cleaning

Watching nature alcohol work home Video games YouTube Wine going walks

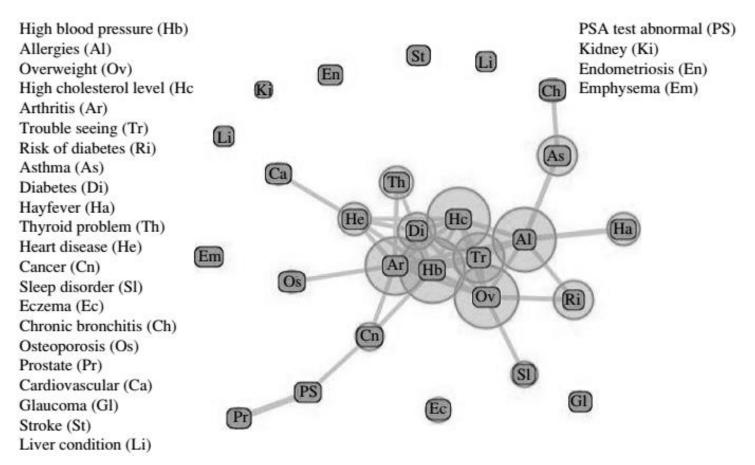
listening music watching news Netflix phone writing hobbies

Social distancing study: How are people spending their time?

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# Influence graph

:Visualize the correlations between objects



Disease influence graph of people at least 20 years old in the NHANES dataset

#### References

- Jiawei Han, Micheline Kamber, and Jian Pei, 2011. Data Mining: Concepts and Techniques (3rd ed.). Morgan Kaufmann Publishers Inc. Chapter 2.
- John C. Hart. Data Visualization. University of Illinois at Urbana-Champaign
- Common Pie Chart Misuses (and How to Fix Them) (<u>link</u>)
- Images are obtained from the above materials and Google

...the end.