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# DATA VISUALIZATION AND VISUAL ANALYTICS

# **TAXONOMY OF VISUAL VARIABLES**

# **Cleveland McGill [1984]**







Figure 3. Graphs from position-angle experiment.

# **Cleveland & McGill: graphical encodings**

- Angle
- Area
- Color Hue
- Color Saturation
- Density
- Length
- Position on a common scale
- Position on non aligned scale
- Slope
- Volume

# Angle decoding



- It is difficult to compare angles
  - Underestimation of acute angles
  - Overestimation of obtuse angles
  - Easier if bisectors are aligned
- Area estimation helps

# Angle decoding



- It is difficult to compare angles
  - Underestimation of acute angles
  - Overestimation of obtuse angles
  - Easier if bisectors are aligned

# **Slopes Decoding**



- Same difficulties as angles
- Easier task since one branch is aligned with xaxis

# **Area Decoding**



- Area is not well decoded
  - Different regular shapes
  - Irregular shapes
  - Context influences (thin area within compact thick area)

#### **Length Decoding**



- Straight forward to endoce numerical values
- Difficulties with relative lengths

#### Position on a common scale



 Widely used in statistical charts

# **Position on non-aligned scale**



- Not as bas as common scale
- Still acceptable

# **Designing Effective Visualizations**

- If possible, use graphical encoding that are easily decoded
- Graphical Attributes ordered(Cleveland & McGill):
  - Position along a common scale
  - Position on non aligned scales
  - Length
  - Angle and Slope
  - Area
  - Volume, density, color saturation
  - Color Hue

	1	1				
Experiment 1						
Position (Common)						
Angle				++	<b></b>	
Experiment 2						
Position (Common)	···-+•					
Length		+-	•+			
Experiment 3						
Position (Common)						
Position (Nonaligned)						
Length				+		
Angle						
Slope						
Circle Area					+	••
Blob Area	L			1		
	4	6	8	10	12	14

Error (Deviation from True Percent)



# **PERCEPTION LAWS**

## Weber's Law

- Just-noticeable difference between two stimuli is proportional to their magnitudes
- Case study on length
  - Given two lines with lengths x and x+w
  - If w is small, it is difficult to notice difference between the two lines
  - If w is larger, it is easier to catch the difference
- How large should w be?
  - The probability of detecting the change is proportional to the reltaive value w/x

## Weber's Law

- Given values (90, 92)
- Detect with probability of 2/90
- Given values(90,92)
- Detect with probability of 2/10



## **Stevens' Law**

- Model the relation between a stimulus and its perceived intensity
- Given a stimulus x encoded with a visual attribute
- An observer decode a perceived value p(x)
- Stevens' law states that
  - p(x) = kx<sup>β</sup>
  - where k is constant and
  - β is a constant that depends on the nature of stimulus

# **Stevens' law**

- Better effectiveness when p(x) = kx<sup>β</sup> is linear
- Linearity depends only on β
- Different visual encodings yields typical ranges for β
  - Lengths: 0.9 1.1
  - Area: 0.6 0.9
  - Volume: 0.5 0.8



Underestimation

Length

Area Volume Brightness

Loudness Smell



Sensation

2

#### Weber and Stevens' Laws

- Given two values x<sub>1</sub> and x<sub>2</sub>
- Let the perceived values be p(x<sub>1</sub>) and p(x<sub>2</sub>)

$$\frac{p(x_1)}{p(x_2)} = \left(\frac{x_1}{x_2}\right)^{\beta}$$

#### Weber and Stevens' Laws: areas

- For areas β=0.7
- Let x<sub>1</sub>=2 and x<sub>2</sub>=1
- The perceived difference will be

$$\frac{p(2)}{p(1)} = \left(\frac{2}{1}\right)^{0.7} = 1,6245$$

- For areas β=0.7
- Let x<sub>1</sub>=0,5 and x<sub>2</sub>=1
- The perceived difference will be  $\frac{p(\frac{1}{2})}{p(1)} = \left(\frac{\frac{1}{2}}{1}\right)^{0.7} = 0,6155$

#### Weber and Stevens' Laws: areas vs lengths

- For areas β=0.7
- Let  $x_2 = x_1 + w$
- The perceived difference will be

$$\left(\frac{x+w}{x}\right)^{0.7} \approx 1 + \frac{0.7w}{x}$$

- For lengths β=1
- Let  $x_2 = x_1 + w$
- The perceived difference will be

$$\left(\frac{x+w}{x}\right)^1 = 1 + \frac{w}{x}$$

#### **Takeaway messages**

- Data type for entities and relationships
- Visual variables for representation
- Mapping of types to VVs
- Some VVs are more appropriate for specific data types

# Visual Analytics Dos and Don'ts for visual charts

# **Crash course on effective Charting**

#### THE WALL STREET JOURNAL **GUIDE TO** INFORMATION GRAPHICS THE DOS & DON'TS **OF PRESENTING** DATA, FACTS, AND FIGURES DONA M. WONG

"INVALUABLE." -HOW DESIGN



#### Dona M. Wong Guide to Information Graphics

The Dos and Don'ts of Presenting Data, Facts, and Figures

W. W. Norton & Company

# **Charting Pipeline**



#### **Charting Examples**



May these charts be improved? Why? How?

#### **Charting Examples**



May these charts be improved? Why? How?

# FONTS

#### Fonts



"Metal type". Licensed under Public Domain via Wikimedia Commons - http://commons.wikimedia.org/wiki/ File:Metal\_type.svg#mediaviewer/File:Metal\_type.svg



Typographic parts of a glyph: 1) x-height; **2) ascender line**; 3) apex; **4) baseline**; 5) ascender; 6) crossbar; 7) stem; **8) serif**; 9) leg; 10) bowl; 11) counter; 12) collar; 13) loop; 14) ear; 15) tie; 16) horizontal bar; 17) arm; 18) vertical bar; 19) cap height; **20) descender line.** 

Font size = (1) + (2) + (20) = (19) + (20)

"Typoghaphia" by F I a n k e r (typographic font designed by myself, named Imperator). Licensed under Public Domain via Wikimedia Commons - http:// commons.wikimedia.org/wiki/File:Typoghaphia.svg#mediaviewer/ File:Typoghaphia.svg

#### **Fonts: general rules**

- Leading should be 2 points larger then type size
- Avoid too small or condensed type faces
- Keep style simple: use
  **bold** or *italic* to
  emphasize a word
  (better not *both*)

- Avoid ALL CAPS
- Avoid styled fonts
- Avoid C\*\*\*C Sans Serif
- Reduce type at an angle
- Avoid tracking

#### Fonts are meant to describe, not to adorn

# **Typography in Charts**



Headline of the chart



A brief description that outlines what the data shows

## **Typography in Charts**

Don't

#### Seadline of the chart



Do

#### Headline of the chart



Title of x-axis

Title of x-axis

# **Typography in Charts**

Name	Data	Data	Data		
Company A	0.0	0.0	0.0		
Company B	0.0	0.0	0.0		
Company C	0.0	0.0	0.0		
Company D	0.0	0.0	0.0		
Many elements in					

Name	Data	Data	Data		
Company A	0.0	0.0	0.0		
Company B	0.0	0.0	0.0		
Company C	0.0	0.0	0.0		
Company D	0.0	0.0	0.0		
Give emphasis to relevant results					

Many elements in bold. Which part is highlighted?

Visual Display of Quantitative Data Edward Tufte, 1983

# **DATA-INK RATIO**

#### **Data-ink Ratio**



erie 2 – Serie s

#### **Data-ink Ratio**

Data-Ink Ratio =  $\frac{\text{Data ink}}{\text{Total ink used in graphic}}$ 


#### **Bar Charts**

#### Represent discrete quantities



## **Bar Charts**

#### Avoid non-functional adornation



### **Bar Charts: baseline**



**Chart Title** 

D

**Chart Title** 

## **Bar Charts: baseline**



**Chart Title** 

## **Bar Charts: ordering**







 Pie Charts compares relative sizes and contributions



## **Pie Charts: ordering slices**



### **Charting Examples**



May these charts be improved? Why? How?

### **Takeaway Messages**

- Charts exploit position on scale VV
- Best practice to reduce biases and misinterpretation of charts

## **Visualization Taxonomy**



Visual Tools for Analyzing, Managing, and Communicating

Robert L. Harris

#### Chart Suggestions—A Thought-Starter



#### **Bars vs. Lines**



Line implies trends. Do not use for categorical data

#### **Trend over time**

# WILLIAM PLAYFAIR



#### **Trend over time**

#### Apple Inc. (AAPL) - NasdaqGS



601.10 + 15.53(2.65%) 4:00PM EDT | After Hours: 604.60 + 3.50 (0.58%) 7:15PM EDT - Nasdaq Real Time Price



### **Trend over time**

Published: February 2, 2010

#### Budget Forecasts, Compared With Reality

Just two years ago, surpluses were predicted by 2012. How accurate have past White House budget forecasts been?

#### 1 2 3 4 5 6 NEXT >

#### Latest forecast

Today, with a better understanding of the severity of the economic downturn, the deficit situation is much more dire.



#### Make clear distinction between data and prediction

## **Streamgraphs**



#### Vision Statement

IDER WATCH

#### Six Ways to Find Value in Twitter's Noise

hed by Bodd Berleates, data and encentration by Juff Clark

it's easy to dismiss l'witter as jabber, but smart marketers will recognize it as a stream of free consumer data to be mined in near-real time. Online visualization tools can belp physiat what consumers are reaching and sharing, elucidate memes in the chatter, and snearth trends. To show exerketers how they can gain insight from Twitter, we captured more than a half million. tweets containing the word "iPad" that were broadcast during the product's launch weekend in April. We then mapped key words that appeased in those tweets on the graph below. MER Reported Property

#### The iPad Launch by the Numbers

TOTAL TWEETS USED NO PRODUCT'S LAUNCH WORKEND 13% heattraceators presidential 18% Incorrector 8-PERCENTER IN THE PERCENT



.) Big desper into the stream graphegive an everall impression of which people are senting about, it's important to know what other words are being used in relation in Roser in the stream, 4 this ribbers of running containing the same harars but from gamers raresting about the certail of Autidom doll a for the inst. as though a lot of seat-ers neer calling the Pad AMAZINE, but in fast, mind of this cost is save from warrs estimated angle julier from convedian construction. Separated at all of a Constitutions have a second state which assume that a second have a second state have a second state of here a seco factoria del seguitoria del la constanta de

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#### fast of strange Ð TYPING

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printer Product testing

and then singling out the most common words in

those tweets (see above)

gives you level and how

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increages and product development.

#### search why regaring any words are carried any

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Survey Partnering of

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#### product functionality. Cultorine service-could Learn abears conversa-tion dominators, wheth be adjusted to address But suitlenly downaie the tweet stream mean screening has happened BUD worth has my shoul. The turn AAN BUDAN select township to the state is hadness' one of that stores and to an efforts not authorized by Apple. Twodors were straing links

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494,085

How to Read This Graph This stream graph shows tweet volume over time, lack sholet illustrates the

proportion of Machinesta containing a given word, such as MELP, Oxfor is assed only in distinguish dealers.

Committee of

As Harord Burness Booter American

Ann and Hirord Burnes Series \$5.

#### **Pie vs Bar charts**

#### 65% of the market is controlled by companies B and C





#### **Pies vs Bar charts**



Furthermore, we present the distribution of attacks towards employees in detail in Fig. 10 right. The blue employees are secretaries, the green ones are administrators and the red ones are scientific employees. The number following the name is the number of times that person was attacked. All of the names are pseudonyms for real people. The person that suffered the most attacks is Monja a secretary with overall 8 attacks. In contrast, all other victims suffered between 1 and 3 attacks.



## Pies vs Bar charts (improved)



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#### **Episodes per person**

## Pies vs Bar charts (improved)



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#### **Episodes per person**

## **Showing changes**





#### Market Share

Platform	Jun 2011	Dec 2011	Change
Windows (Desktop)	79.5%	78.0%	(1.5%)
Mac (Desktop)	13.3%	13.2%	(0.1%)
iOS	3.6%	4.9%	1.3%
Android	2.4%	2.7%	0.2%
Other	1.3%	1.3%	0.1%

## **Showing Changes**





### **Density Plot**



### **2D Density Plots**



#### **Box Plots**



### Scatterplot



## **Cluttering, Overplotting**





#### alpha=1/100



# A FEW EXAMPLES AND CASE STUDIES



The Office for National Statistics (ONS) said gross domestic product (GDP) expanded by 0.7 per cent in the fourth quarter - an increase from the 0.6 per cent calculated on the watchdog's first look at the economy

Source: http://www.dailymail.co.uk/news/article-4248690/Economy-grew-0-7-final-three-months-2016.html



Source: Facebook Analytics



Source: http://weekend.gazeta.pl/weekend/1,152121,20528386,wierza-mocniej-niz-miastowi-za-to-zarabiaja-mniej-i-calym.html



Source: https://lensvid.com/gear/lensvid-exclusive-happened-photography-industry-2016/

Borkin MA, VoAA, Bylinskii Z, Isola P, Sunkavalli S, OlivaA, Pfister H. What Makes a Visualization Memorable? IEEETransactions on Visualization and Computer Graphics (InfoVis 2013).

http://vcg.seas.harvard.edu/publications/ what-makes-visualization-memorable

# **VISUALIZATION TAXONOMY**
## Area

















Table			Text Based
	Table		Phrase Net
ABC	1234	X45	erspiciatis unde fugitjiste quae voluptatem natus accusantium doloremque vitae dicta eaque error sunt rem aperiam
Category	543,2109	7%	
Group	45.67	45%	
Unit	9876	98%	
Class	123.78	12%	Word Cloud
	Text Chart		word word word
•Sed dignissim v •Nisl quis congue •Sed vitae rhonc	ehicula sus	uding 1 um turpis at tellus crit. Ut nec magna dipiscing dolor eget ut commodo lacus nordise *	word TEXT word word word word
odio •Integer at odio		- Lorem	Word Tree

Heading 3

√Chart 1 √Chart 2

√Chart 3

√Chart 4

Heading 2

Acnean tincidunt sem vel massa cursus non tempus quam auctor. In nisi mi, commodo sit.

Amet rutrum vitoe, fringilla non urna. Quisque sagittis ultrices sapien, quis posuere massa interdum quis.

#### Word Tree



# **Visual Taxonomy**

### The Data Visualisation Catalogue

About · Suggest · Shop · Resources



### http://www.datavizcatalogue.com/



 Appropriate chart type for specific data type and visualization task