

Psychology of Data Visualization: Course Overview

Michael Friendly
Psych 6135



<http://euclid.psych.yorku.ca/www/psy6135/>

@datavisFriendly

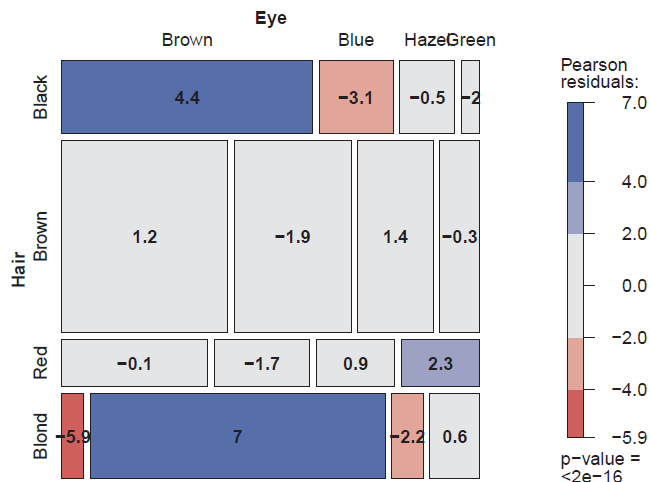
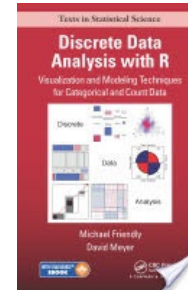
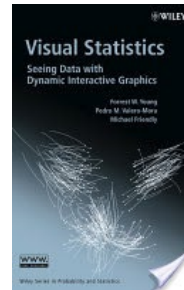
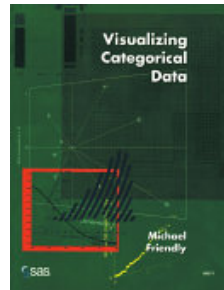
Introducing: me

I wear two hats, both reflected on my license plate:

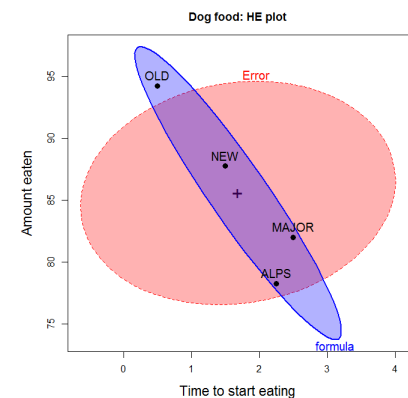
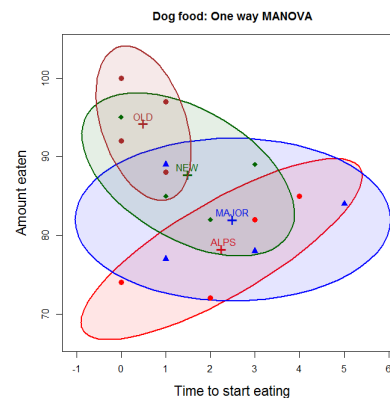
Statistical graphics developer (categorical & multivariate data analysis)



Yours to discover!



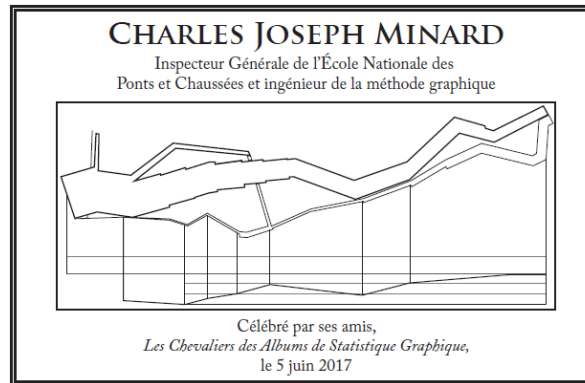
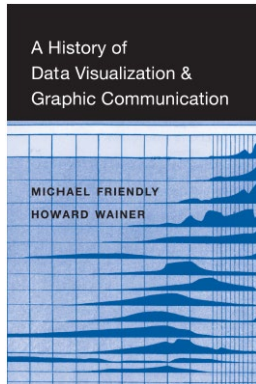
mosaic plots for frequency tables



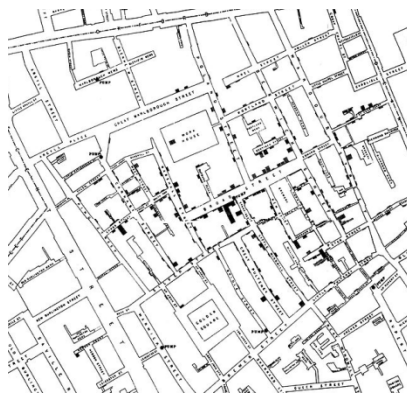
HE plots for MANOVA

Introducing: me

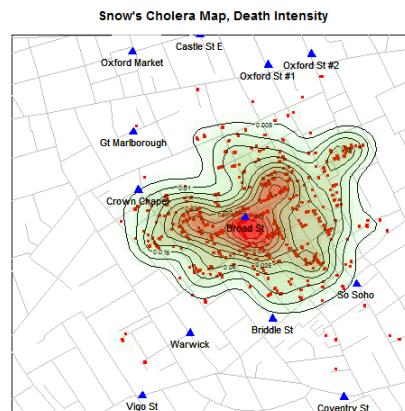
History of data visualization: *Les Chevaliers*; Friendly & Wainer (2021)



John Snow's map of cholera in London, 1854

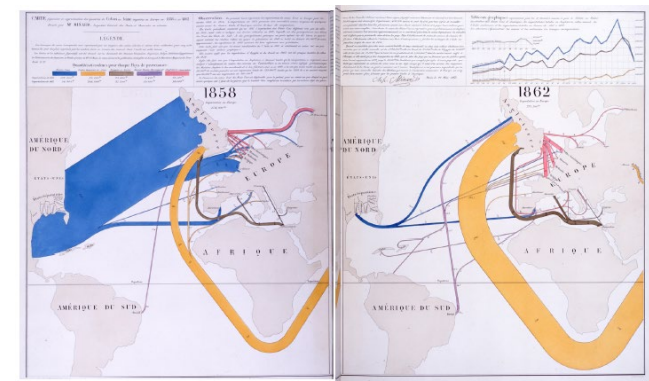


original



modern enhancement

C. J. Minard: Flow maps of cotton trade



Visual explanation: What happened in the US Civil War?

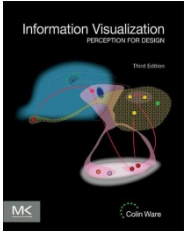
Course Topics

- Varieties of information visualization
 - Goals of visualization
 - Survey of graphic forms
- History of information visualization
- Psychological models, theories and results
 - What can people see, understand and remember from data displays?
 - Perceptual aspects, cognitive aspects
- Software tools for information visualization
- Visualization in statistics: case studies
 - Categorical data; High-D data; Dynamic and interactive methods
- Human factors research: how to tell what works

Your role

- Weekly readings – see the course [web](#) site for updates
- Discussion – no formal grade, but please contribute
- Discussion leader (20%)
 - Each week 1-2 of you will lead a brief discussion on one of the readings or sub-topics (~ 5 min.)
- Class presentation (40%)
 - In the last week, each person will give a ~ 20 min presentation on a topic of research, application or software related to data visualization
- Research proposal (40%)
 - Prepare a brief research proposal on a data visualization topic

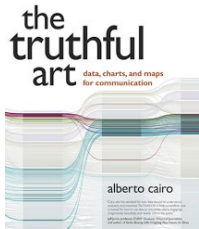
Books & Readings



Colin Ware, *Information Visualization*, 3rd Ed.

What perceptual science has to say about data visualization, from a bottom-up perspective

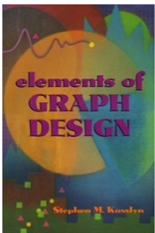
Course notes at: <http://ccom.unh.edu/vislab/VisCourse/index.html>



Alberto Cairo, *The Truthful Art*

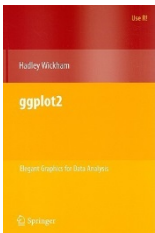
Information graphics from a communication perspective

Blog: <http://www.thefunctionalart.com/>



Steven Kosslyn, *Elements of Graph Design*

A cognitive psychologist looks at graphs and presents some do's and don'ts. There are better books like this today.



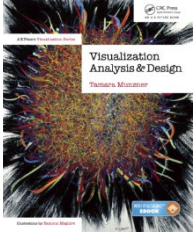
Hadley Wickham, *ggplot2: Elegant graphics for data analysis*, 2nd Ed.

1st Ed: Online, <http://ggplot2.org/book/>

ggplot2 Quick Reference: <http://sape.inf.usi.ch/quick-reference/ggplot2/>

Complete ggplot2 documentation: <http://docs.ggplot2.org/current/>

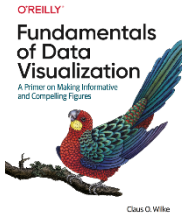
More books I like



Tamara Munzner (2014), *Visualization Analysis & Design*

An attractive new book combining computer science and design perspectives

Web page: <http://www.cs.ubc.ca/~tmm/vadbook/> with lots of illustrations & lectures



Claus Wilke(2019), *Fundamentals of Data Visualization*

A detailed, practically oriented book on data visualization methods

Online: <https://clauswilke.com/dataviz/>

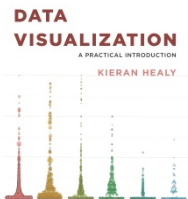
Course notes: <https://wilkelab.org/SDS375/>



Manuel Lima, *The Book of Trees: Visualizing branches of knowledge*

A visual delight; an entire history of tree-type diagrams

Blog: <http://www.visualcomplexity.com/vc/blog/>



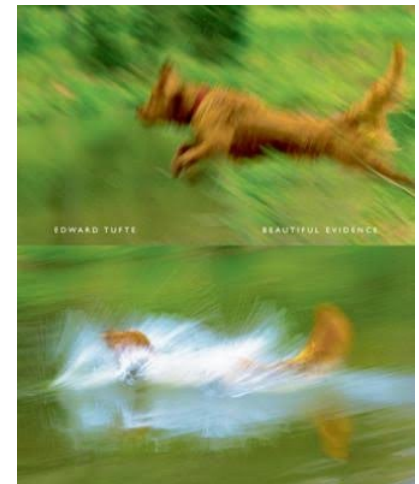
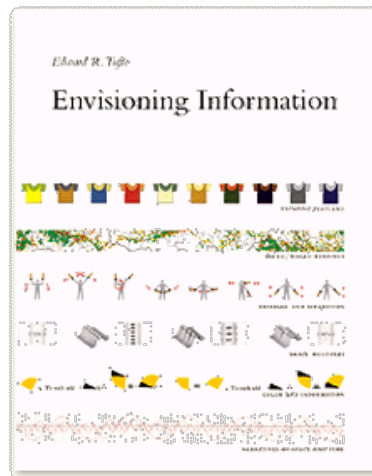
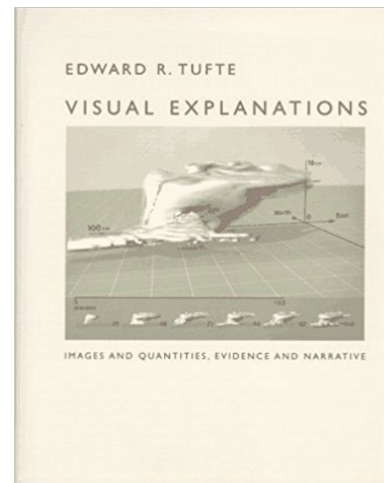
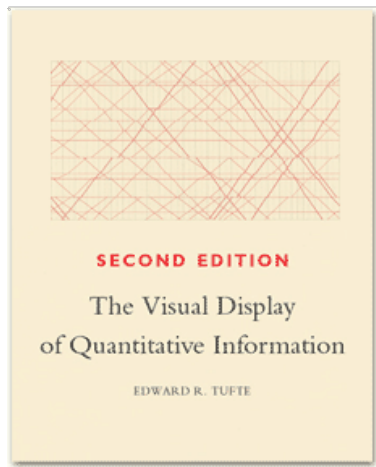
Keiran Healy, *Data Visualization: A Practical Introduction*

An accessible primer on how to create effective graphics from data using ggplot2

Online: <http://socvis.co>

Tufte Stufe

Four books by Edward Tufte largely defined the landscape for data visualization and information design



Concepts introduced:

- chart junk,
- data-ink ratio,
- small multiples,
- substance takes precedence over visual design

Web site:

<https://www.edwardtufte.com>

Blogs & Web resources



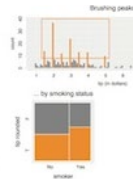
My web site, <http://datavis.ca>. Contains the [Milestone Project](#) on the history of data vis, [Data Visualization gallery](#), links to books, papers and courses.



Kaiser Fung, <http://junkcharts.typepad.com/>. Fung discusses a variety of data displays and discusses how they can be improved.



Nathan Yau's blog, <http://flowingdata.com>. A large number of blog posts illustrating data visualization methods with tutorials on how to do these with R and other software.



<http://visiphilia.org/>. Statisticians Di Cook and Heike Hofmann from Iowa State University blog about data visualization topics, using R



Manuel Lima's blog, <http://www.visualcomplexity.com/vc/blog/>, with hundreds of projects on all types of visualizations

Blogs & Web resources

DATA STORIES

<http://datastori.es/>. A podcast on data visualization with Enrico Bertini and Moritz Stefaner; interviews with over 100 graphic designers & developers.



KANTAR
Information is Beautiful
Awards

Annual awards celebrate excellence and beauty in data visualizations, infographics, interactives & information art.
<https://www.informationisbeautifulawards.com>



<https://www.r-bloggers.com/>. A large collection of posts on R news and tutorials by over 750 R bloggers.

info we trust.



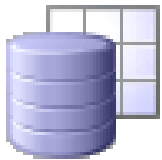
Raymond Andrews, <http://infowetrust.com/>. A visual storyteller delights with graphic stories from the history of data visualization

Psychology facts: Why visualization?

- ~90% of information about the environment is received by a person through the eyes.
- ~50% of our brain neurons are constantly involved in the processing of visual information.
- The presence of pictures increases desire to read the text by ~80%.
- A person remembers 10% of what he/she heard, 20% of what's read, and 80% of what's seen!
- People perceive 70% of the information if there are no illustrations. Add pictures there — the figure will increase up to 95%.

Data, pictures, models & stories

Goal: Tell a credible story about
some real data problem



data

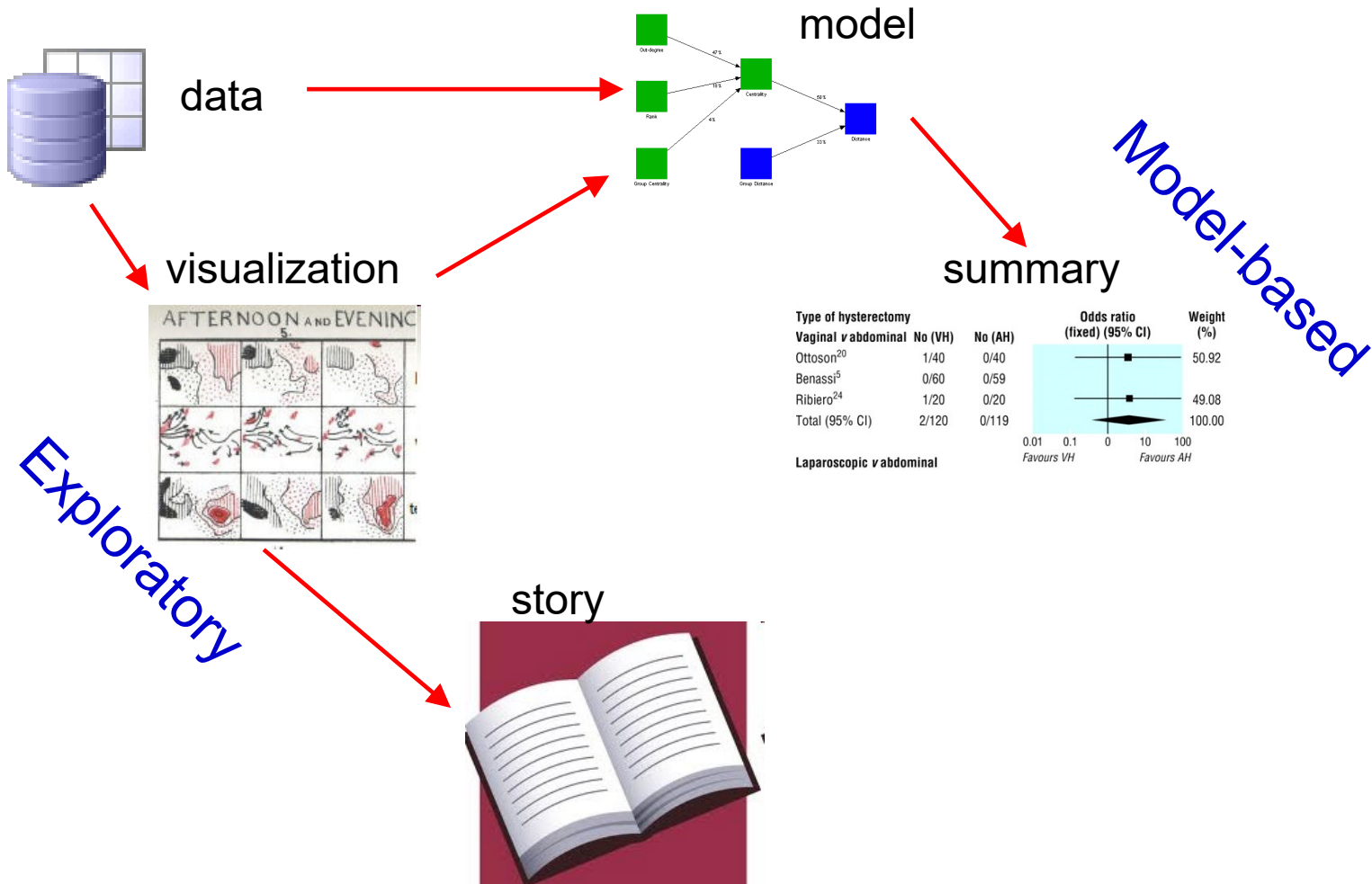
Measles vaccination
Global warming
...

story



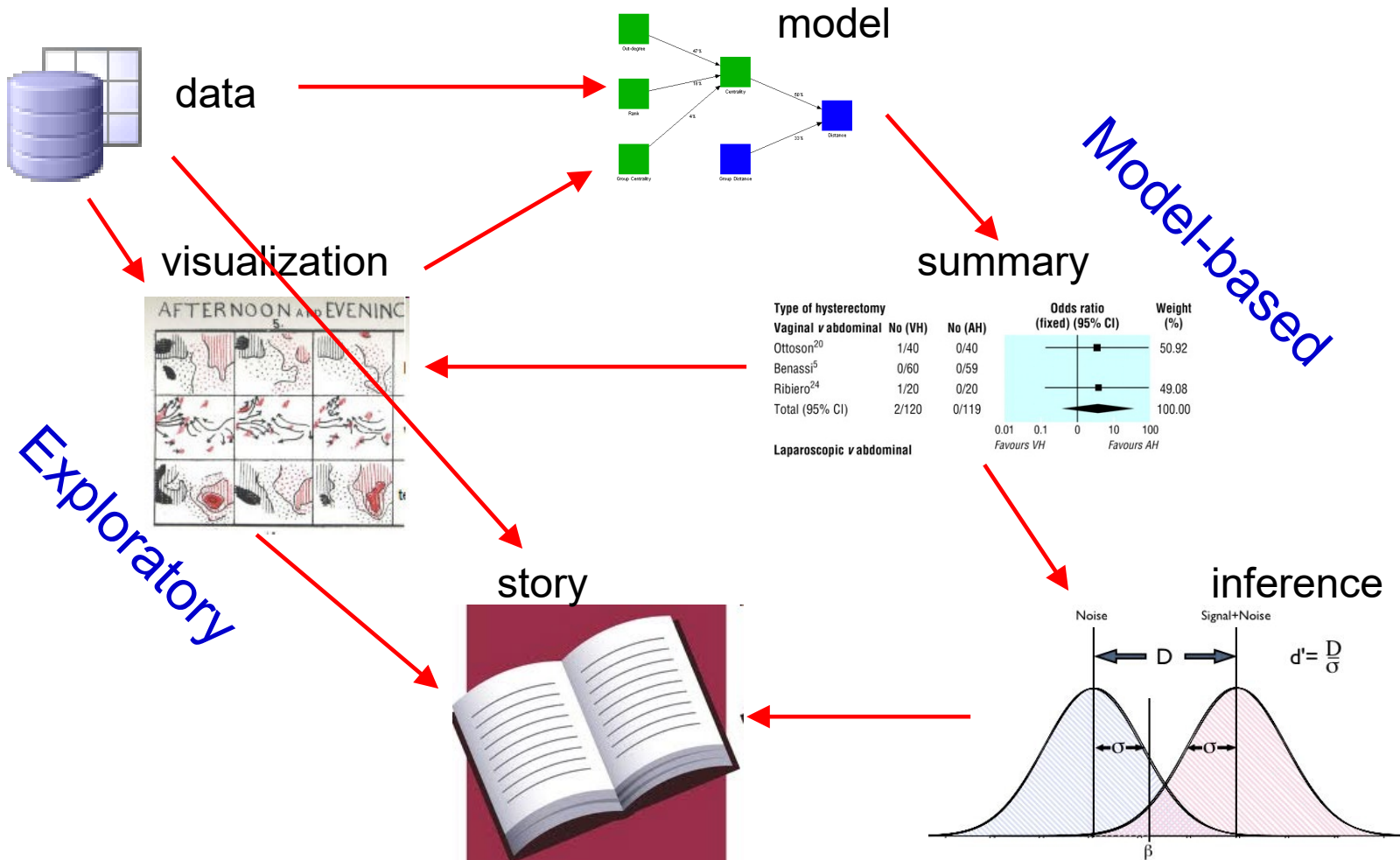
Data, pictures, models & stories

Two paths to enlightenment



Data, pictures, models & stories

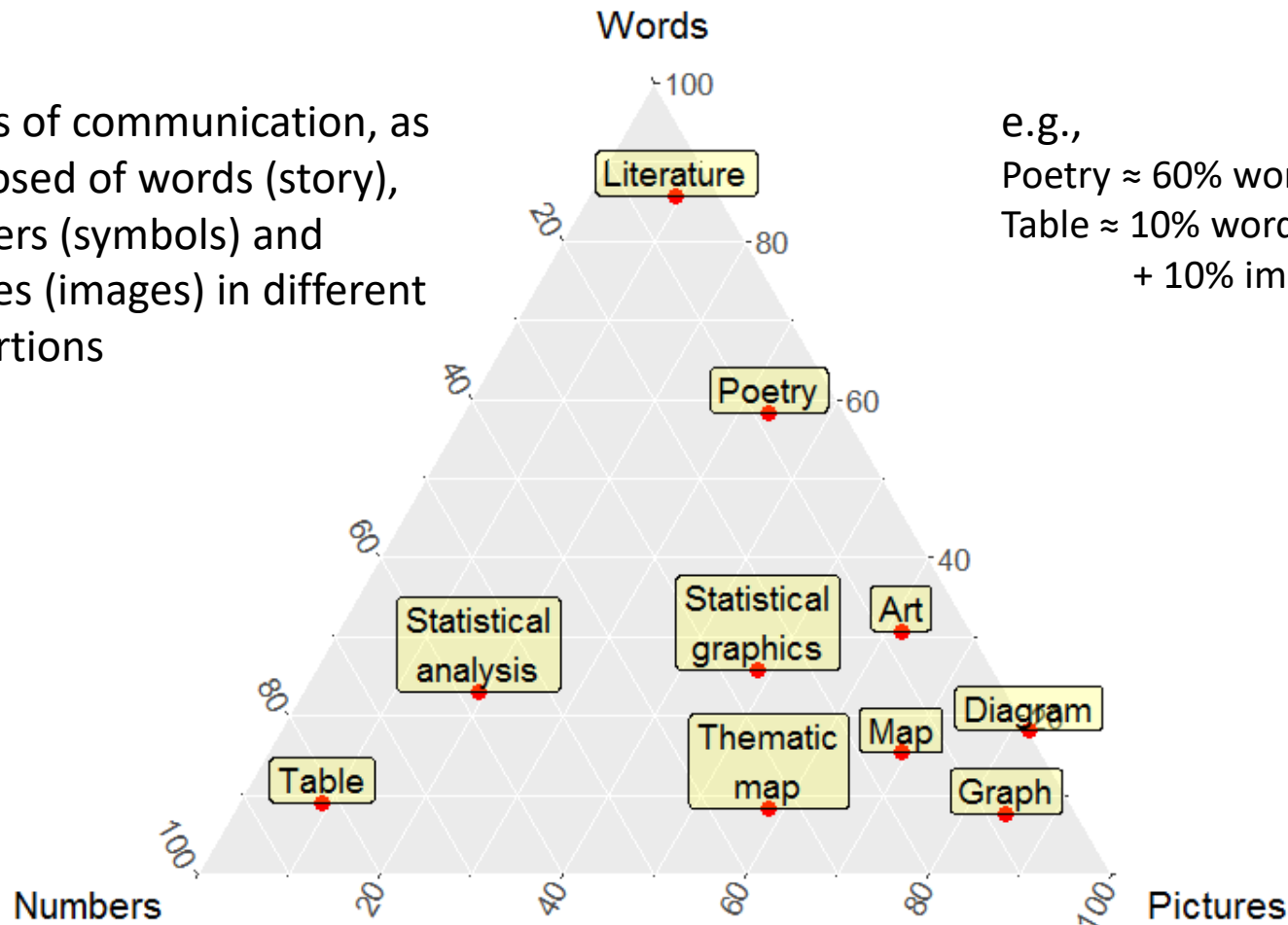
Now, tell the story!



Words, numbers and pictures

Pictures and images in a wider context

Modes of communication, as composed of words (story), numbers (symbols) and pictures (images) in different proportions



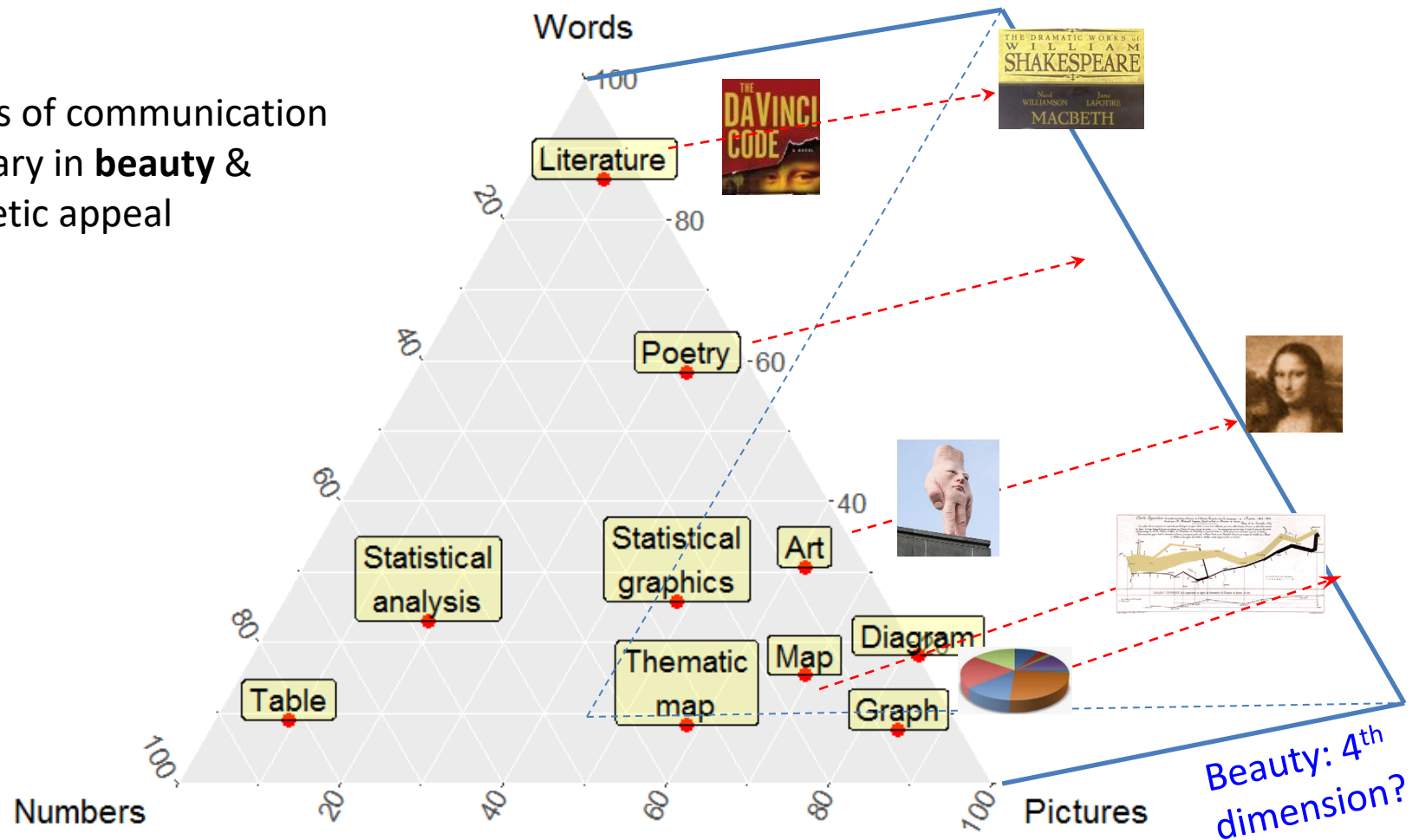
e.g.,

Poetry \approx 60% words + 40% images
Table \approx 10% words + 80% numbers
+ 10% images

Words, numbers and pictures

Beauty: The 4th dimension

Modes of communication
also vary in **beauty** &
aesthetic appeal



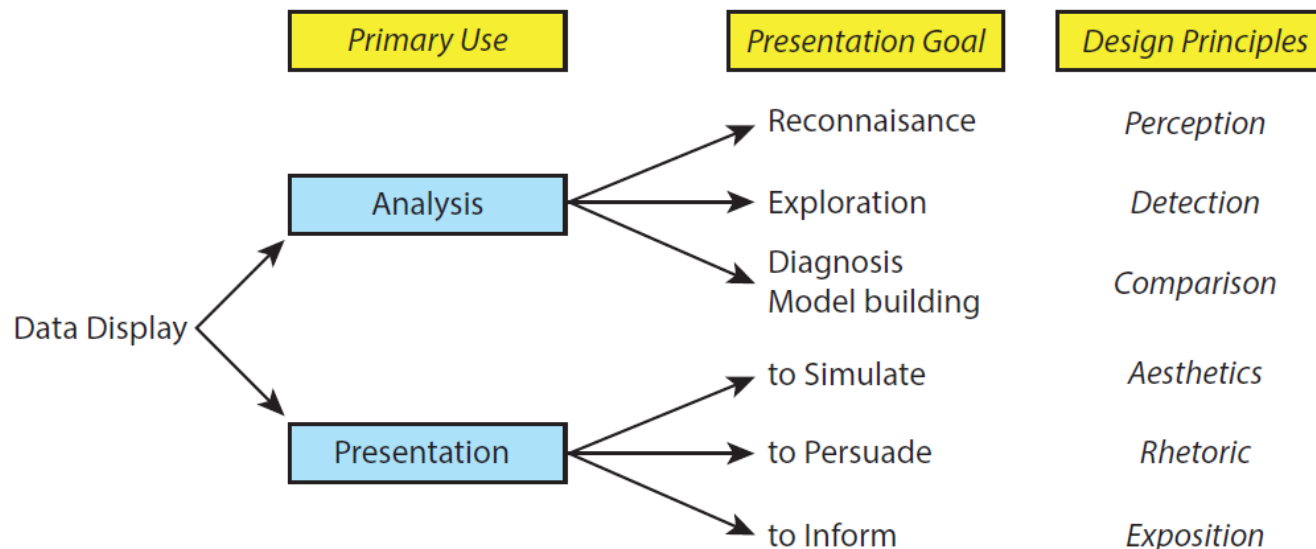
Roles of graphics in communication

- Graphs (& tables) are forms of communication:
 - What is the audience?
 - What is the message?

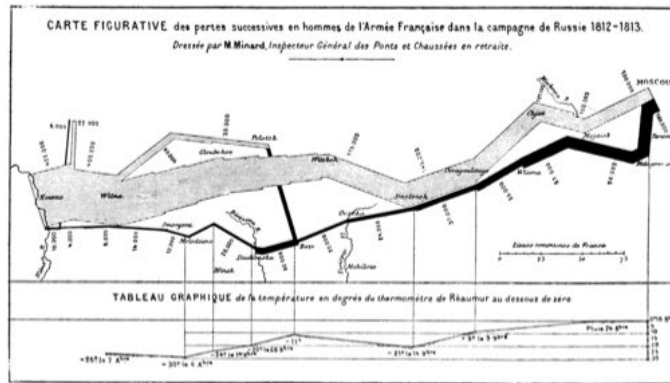
Analysis graphs: design to see patterns, trends, aid the process of data description, interpretation

Presentation graphs: design to attract attention, make a point, illustrate a conclusion

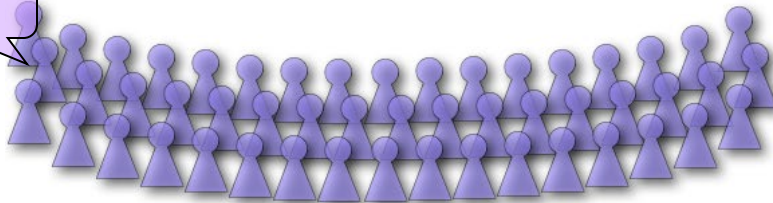
Basic functions of data display



Different graphs for different purposes

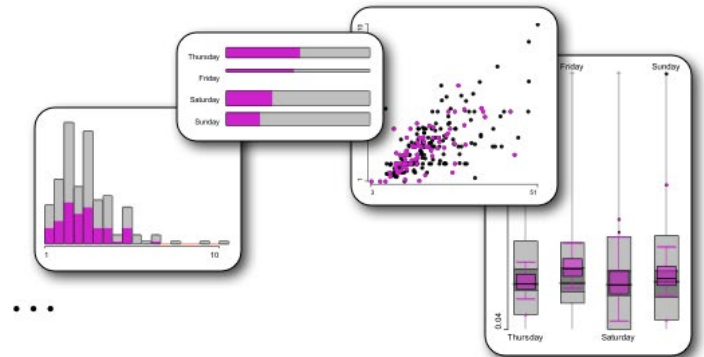


Wow!



Presentation

Goal: the Wow! experience
Single image for a large audience
Tells a clear story!



Ah ha!



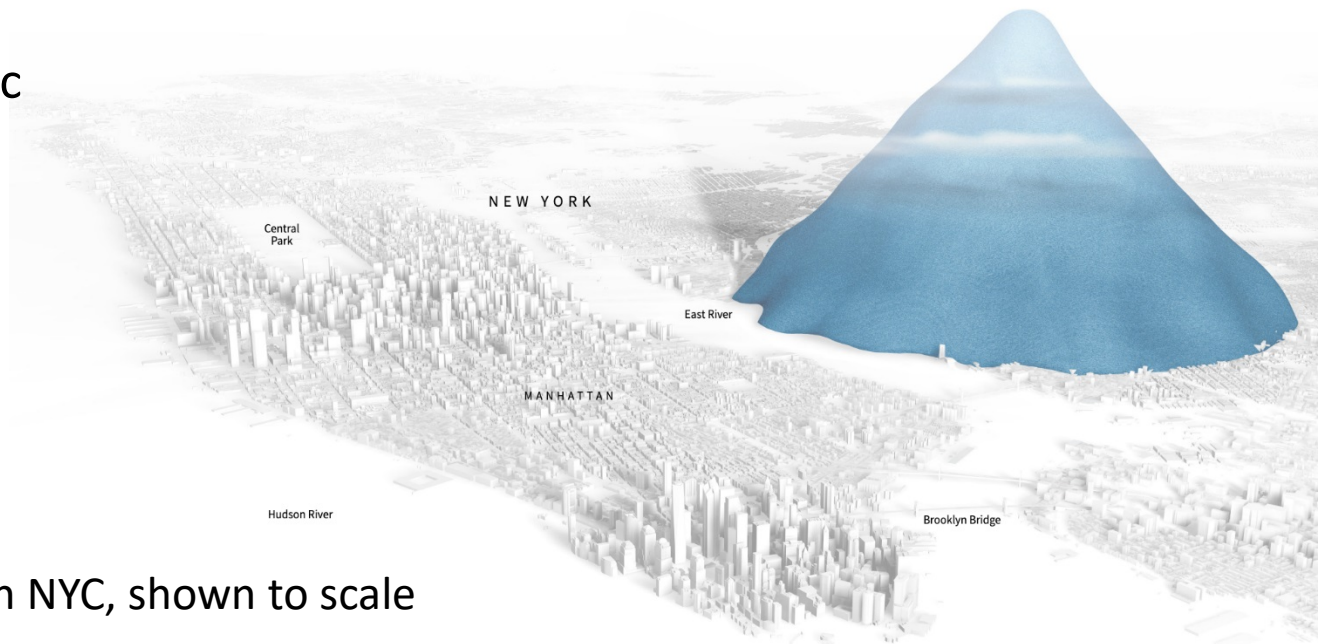
Exploration

Goal: the Ah ha! Experience
Many images, for a narrow audience
(you!), linked to analysis

Infographics

The best infographics tell a story, using numbers, but shown visually

Drowning in plastic



Plastic bottles sold in NYC, shown to scale

From: <https://graphics.reuters.com/ENVIRONMENT-PLASTIC/0100B275155/index.html>

Powerful graphs: Measels and vaccines

Visualizing the impact of health policy interventions

In 2015 Tynan DeBold & Dov Friedman in the *Wall Street Journal* show the effect of the introduction of vaccination programs in the US states on disease incidence, using color-coded heat maps for a variety of diseases

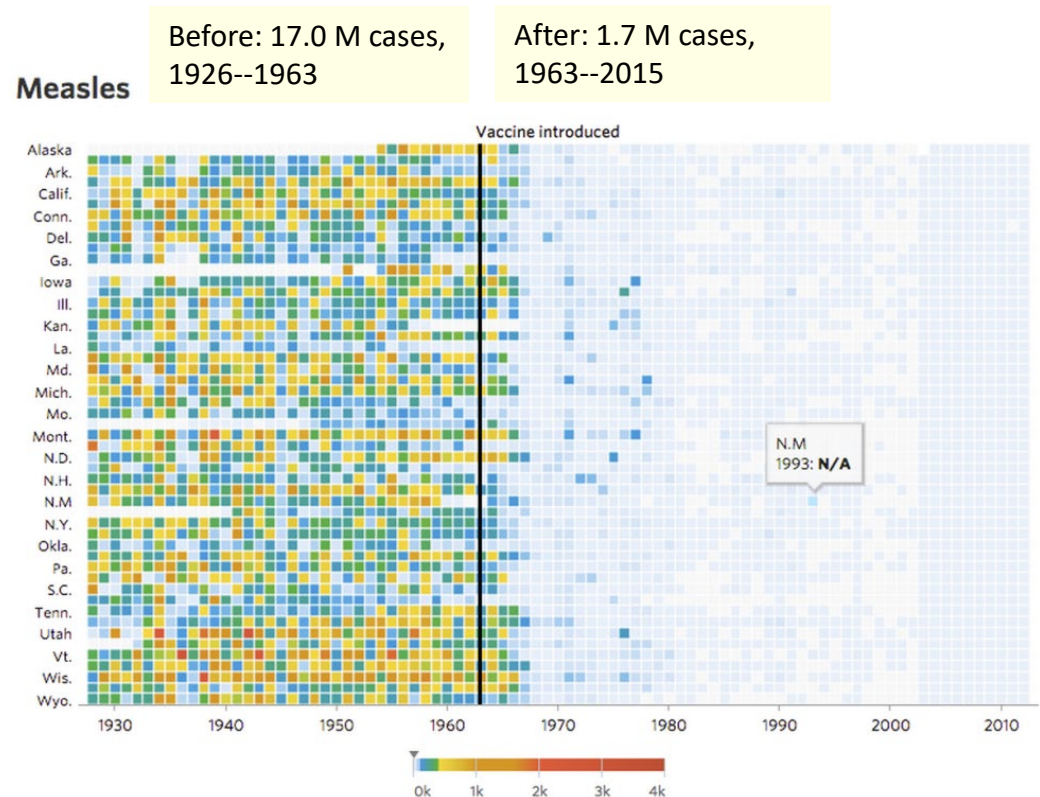
Measles was decimated!

The message hits you between the eyes!

Powerful graphs make comparison easy

In 2014, vaccination rates declined and measles re-emerged in those areas

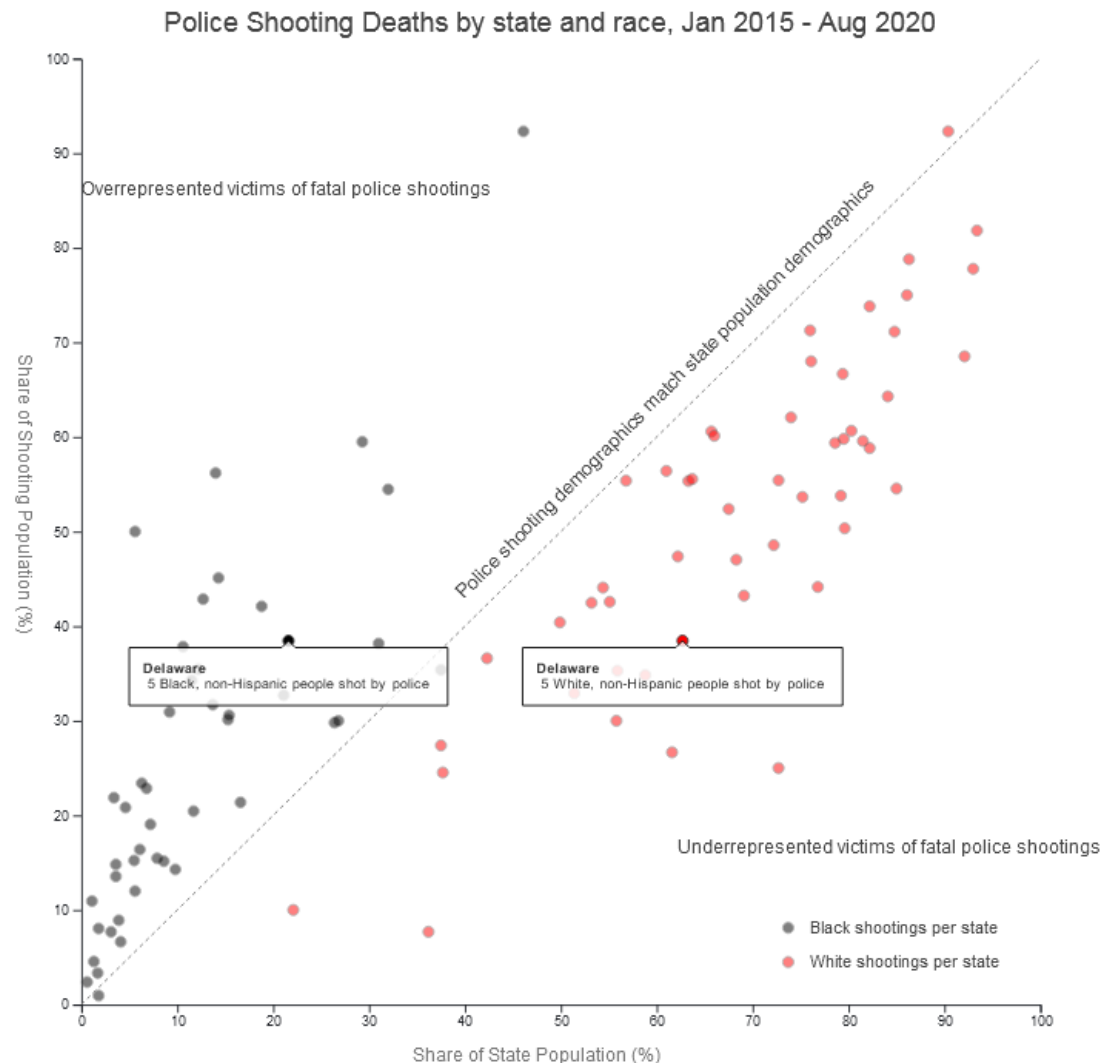
Effective graphs can cure ignorance, but not stupidity.



Police shooting deaths

Analysis of Washington Post database on 5500 police shooting deaths for Blacks vs. **Whites**

Plotting % of shooting vs. % of pop shows a clear & disturbing pattern



Their names: Interactive graphic

As powerful as Yad Vashem & the Washington D.C. Vietnam memorial, this list of 28,000 US fatal encounters with police commands attention.

Each one is linked to a story or description. Classified by {Gender, Age, Cause}

THEIR NAMES					
MISSION STATEMENT					
YEAR	GENDER	AGE	ETHNICITY	LOCATION	CAUSE
OBSOLETE NAMES					
2020					
George 'Big Floyd' Floyd Alexica S. Miloni M. Roosevelt M. Jarvis T. Name withheld by police Dion J. Justin M. Joe L. C. Name withheld by police Gary P. D. John A. V. Memphis I. K. P. Sean T. K. Name withheld by police Stephen G. S. Chad B. Maurice S. G. Bernardo P. Christopher C. Gary P. Name withheld by police Brian M. Willie L. O. S. William J. J. Ryan W. Casey D. Adam S. A. George B. Name withheld by police Ubaldo G. Name withheld by police Name withheld by police Kendrick D. T. Name withheld by police Toby W. Name withheld by police Levi D. M. Keith Y. Wilbon C. W. Name withheld by police Jeffrey H. Tyler H. Wyatt M. Robert W. L. Name withheld by police James H. Antonio W. Tyrone S. Tyrone S. Tre'veon D. Juan M. M. Bernard L. Brian C. Jacob W. B. Name withheld by police Jayden D. T. Tharin W. Name withheld by police Gamaliel S. Wallace D. S. Robert J. J. Trashawn S. Jamarquis B. Robert A. Hunter C. Raymond C. H. Name withheld by police Name withheld by police Darnesha R. John M. S. Name withheld by police Anthony Y. Robert T. Name withheld by police David T. A. Donny L. Buddy D. L. Anthony A. Marie M. Falando M. F. Gregory H. Christopher C. Curtis F. Quinton J. N. Name withheld by police Alexander C. H. LaKendrick O. Daniel R. Name withheld by police Name withheld by police David U. Jada M. D. Jonathan C. Elder M. Ward B. M. I. Yassin M. Jayson T. Damien S. J. William P. Darrin J. P. Adrian M. Sheldon C. F. Lloyd N. J. Ruben E. Name withheld by police McHale R. Finan H. B. Name withheld by police Name withheld by police James P. Christopher P. Daniel G. Dreaseon J. R. Ashlynn L. Name withheld by police Rodney L. B. James J. J. Name withheld by police Name withheld by police Davon W. Justin C. Sandra L. H. Ethan M. Stuart S. W. Daniel Z. Clarah R. Jahsean I. H. Name withheld by police Edward L. A. J. Name withheld by police William C. S. Donald L. A. Janine L. Johnny W. Demontre B. Name withheld by police Ronal Z. Josephina I. C. Marcus C. S. Phillip M. C. Chad R. A. Kortney S. P. Asante S. C. Armando S. Michael F. LeJune B. C. Juwan L. B. Simone A. D. Damian J. S. Michael F. Name withheld by police Tyler W. K. Felipe P. Name withheld by police Name withheld by police Jordan R. Said J. Spencer G. C. Isaac A. R. Rose R. L. Nicholas P. B. Brent M. Name withheld by police William L. D. Timothy G. Daniel H. B. Christopher L. S. Malcolm X. R. W. Artemio M. A. Name withheld by police Jesus C. Shaun L. F. Joshua K. P. Nicole A. S. Malcolm X. R. Jonas J. William E. A. Robert M. Jack D. B. James R. B. Christopher A. Faron M. H. Jamie L. D. William G. De Q. T. Bradley G. P. Jose H. Name withheld by police Name withheld by police Kelvin D. S. Reed F. Name withheld by police Larry D. H. Jacqueline M. Brandon R. N. Michael B. R. Marlon A. B. Timothy D. S. J. Roy J. Christopher C. P. a. Luan L. A. Dalton M. L. Fred B. Mason F. Guy S. Joshua J. Craig A. D. Name withheld by police Christopher L. Name withheld by police Elmer L. M. Chase R. Nicolas C. Name withheld by police Thomas M. M. I. Elijah M. S. Virgil T. Ramon T. V. Richard L. Mason W. Name withheld by police Joel A. Steven D. T. William L. T. Jose A. P. D. Mario H. Austin D. H. Name withheld by police Willie J. H. Name withheld by police Name withheld by police Reggie L. A. Derick L. P. Marcus E. E. Jasman W. Robert N. B. Shane S. Olin H. Salvatore F. J. Goldie B. Randy S. A. Juan A. Tony M. C. Anthony W. Leslie F. J. James H. Thomas A. P. I. Name withheld by police Kwamane M. Napoleon J. M. Javier V. Name withheld by police Marcel P. Evaristo H. J. Ruben D. Justin S. Andre C. Shawn L. Miguel G. Errol K. B. Leah B. Name withheld by police Kanisha N. F. Matthew B. G. Name withheld by police Name withheld by police Justin B. Giuseppe P. Jonathan L. A. Zyon R. W. Todd M. G. Desmond F. Daisean W. Kenneth J. B. Zachary G. Derek P. B. Joshua D. R. Name withheld by police Rick H. Name withheld by police Name withheld by police Idris A. Joshua B. Jacob M. D. Derek T. S. Brandon M. S. Michael S. Dewayne C. L. Yamil A. Kelvin P. Keshawn A. W. Antonio S. W. Rashad D. H. Carl M. Richard L. K. Name withheld by police Michael H. Philip C. I. Joshua T. Name withheld by police Aaron B. Lenear M. B. Jarvis D. B. Name withheld by police Jose S. Maria B. Joseph B. Joseph T. J. Name withheld by police Nathan R. H. David A. X. Joseph S. Z. Joseph Z. Jose M. Austin H. David M. C. Rico R. Name withheld by police Kanadel B. Zoe C. Shane T. K. John H. R. I. Etonne T. T. Anthony P. Valente A. Jessie T. S. Timothy T. Dolmorio R. E. J. Jacob E. M. Shane F. Thomas O. Tyrell F. William P. F. Robert H. S. Name withheld by police Maurice J. J. Maurice D. G. Deanna O. John M. H. J. Matthew M. Glenn A. W. Name withheld by police Timothy W. Michael W. Kathryn R. H. Christopher J. M. Todd G. Jared M. F. Danae S. Mauricio H. Gabriel D. R. Charles E. W. Jeffrey S. M. Carlos D. Derrick U. Gerald J. Darius G. Charles G. L. Bryan P. Israel L. Charles T. E. Name withheld by police Peter C. A. Mycael J. a. M. Kamaal K. E. Alvin L. B. I. Larry M. Jake L. W. Harold S. Charles P. Kyle A. E. Karreim L. H. Jose M. R. Jason R. C. Lebaron B. Allen F. B. Name withheld by police Domingo L. William D. T. S. Marc A. Steven A. Gregory K. B. Debra H. Darwin B. F. Charity T. Juanito O. Name withheld by police Robert W. S. Marvin O. Jordan M. D. S. Darrell W. M. S. Michael B. P. Douglas J. F. Dakota W. Y. Joaquin S. R. Catherine G. Princess J. A. Jesse P. C. J. Chase E. B. Jeramiah M. Rory E. M. Name withheld by police Breonna T. Dung N. Marc N. M. Christopher C. M. Justin M. R. Jorge M. Rosario A. A. Name withheld by police Donnie S. Duncan S. L. Travis J. B. Name withheld by police Edwin L. G. Kenney L. J. Joshua C. N. Mary K. F. Rodney R. Joshua R. K. Pablo E. Jennifer T. Desiree N. G. Jesus B. G. Kevin K. Zachary A. J. Barry G. Law T. S. Jerry K. G. Jacob F. Kenneth M. Devan A. T. Elijah J. B. Gene B. Ajay K. A. Curtis H. David J. B. Dane C. T. Patricia H. T. Name withheld by Tucker A. B. Name withheld by police Jose G. Name withheld by police Damond M. S. Zachary A. H. Terry F. C. Andre S. Cody H. Name withheld by Cedarian C. Devin M. H. Stephen D. Ian A. W. Richard R. Antonio R. V. Kent R. K. Jose L. R. Name withheld by police Desmond H. Twashon K. Ramiro C. Name withheld by police William G. R. David W. I. Matthew F. Kyree D. H. Kenneth L. S. Dylan J. O. Terry H. Name withheld by police Lucas J. A. Name withheld by police Boyce M. T. Desiree V. E. Angela W. T. Brittany F. Adam G. J. Rodrigo I. A. Michael L. C. Jacob R. E. Antonio T. police Name withheld by police Joseph C. J. I. Rolando F. Tab K. W. Roger N. Darline G. Q. John D. L. Luis R. Anthony P. D. Thomas E. Z. Rebeca Shakayla H. Travis M. Jeffrey L. F. Name withheld by police Jeremy D. Daelyn P. Kerry M. B. Thomas C. M. Alisha S. L. Brandon L. Rudy A. TI Mary L. Tyler J. L. Gianna N. A. William K. B. Christopher B. G. Dakota L. C. Jeremy S. F. Brian F. F. Abel L. Name withheld by police Adrian L. police Christopher W. Johnnie R. P. Bobby J. G. B. J. B. Thomas R. C. Jeremy T. B. Edward B. I. Jeremy G. Mignonne R. Michael A. L. Zane G. Timothy D. Douglas H. H. Rolando R. Juston R. Michael J. C. John F. R. Robert L. H. Joshua D. Sean P. C. Bracken M. H. Andrew C. Franklin A. N. Name withheld by police Lionel M. Marc D. N. Douglas A. W. Sindi N. M. Alvin C. Jason G. Chris C. Leonard C. P. J. Donnie R. Ronnel M. Clint S. Brian A. H. Richard D. Abdurahman S. Raeshawn A. J. Devel H. Chad E. N. Carlos H. Edwin L. S. Emily S. R. Elisevan N. Justin L. Jaquyn Aaron T. B. Adelbert K. R. William H. G. J. Kevin A. J. Brittany B. John F. T. J. Anthony Y. Name withheld by police Name withheld by police Lisa V. M. Scott J. W. J. Rebecca A. A. Toby E. D. Michael J. R. Andrew J. S. Deandre L. S. Marquis G. Gage S. S. Sherry L. G. James L. M. J. Cur H. Lorenzo A. H. Pearley M. M. Sok C. S. Anthony J. J. Rory J. B. David R. V. A. Name withheld by police Jonathan D. M. Edward R. G. J. Dustin Dustin A. N. Name withheld by police Kevin L. I. Edward N. Charles S. David C. Maurice J. Samuel D. M. David T. C. Bryce M. Stephen K. Thom C. Charles J. N. J. Name withheld by police Murbarak S. Albert L. H. Renard A. D. Aaron P. Frank C. Jose C. I. Zayne C. C. Curtis A. R. Matthew R. K. Cameron S. G. Jonathan A. F. Joshua D. N. Brad S. Ryan S. Henry I. J. Lee A. D. Thomas A. G. J. Jimmy A. A. Nicholas R. David E. H. Thom					

George 'Big Floyd' Floyd, 46

Male, African-American/Black

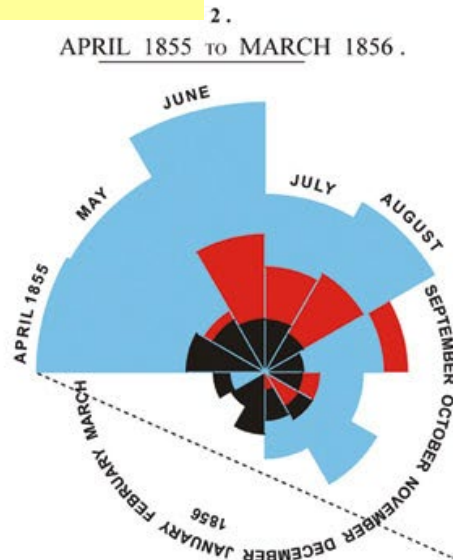
Cause of Death: Asphyxiated/Restrained

"Police responded to a call from a grocery store that claimed George Floyd had used a forged check. When located in his car, police said, he resisted officers. Cell phone video showed Officer Derek Chauvin kneeling on Floyd's neck until he was incapacitated and died. Three other officers were present and didn't stop the assault. The four officers were fired the day after the killing."

Source: www.theguardian.com

Presentation graph: Nightingale (1857)

After reform



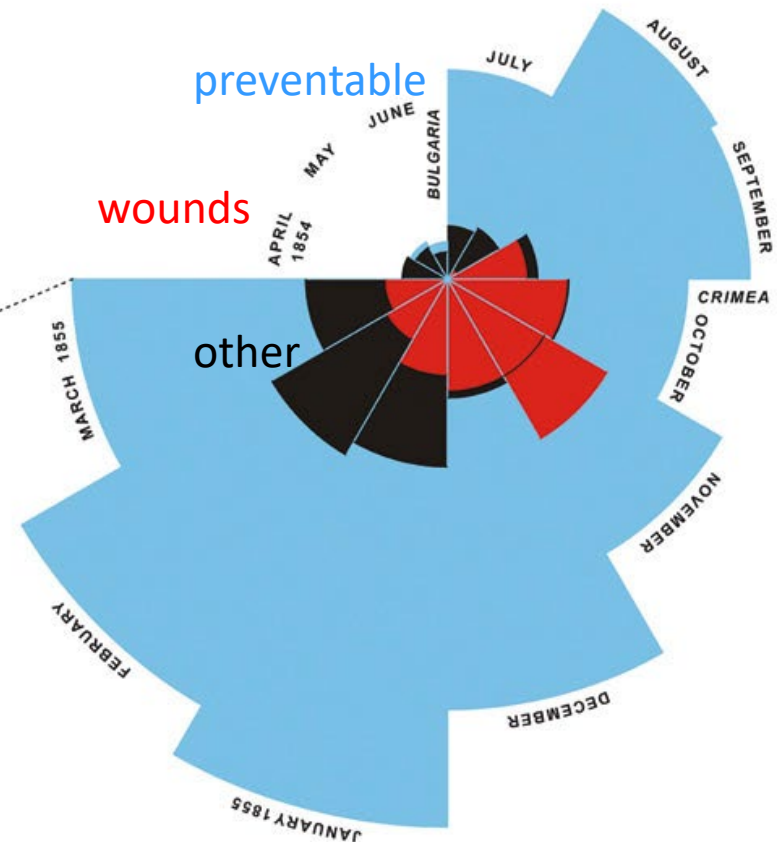
*The Areas of the blue, red, & black wedges are each measured from the centre as the common vertex
The blue wedges measured from the centre of the circle represent area for area the deaths from Preventible or Mitigable Zymotic Diseases, the red wedges measured from the centre the deaths from wounds, & the black wedges measured from the centre the deaths from all other causes*

The best graphs pass the **Interocular Traumatic Test**: the message hits you between the eyes!

DIAGRAM OF THE CAUSES OF MORTALITY IN THE ARMY IN THE EAST.

Before reform

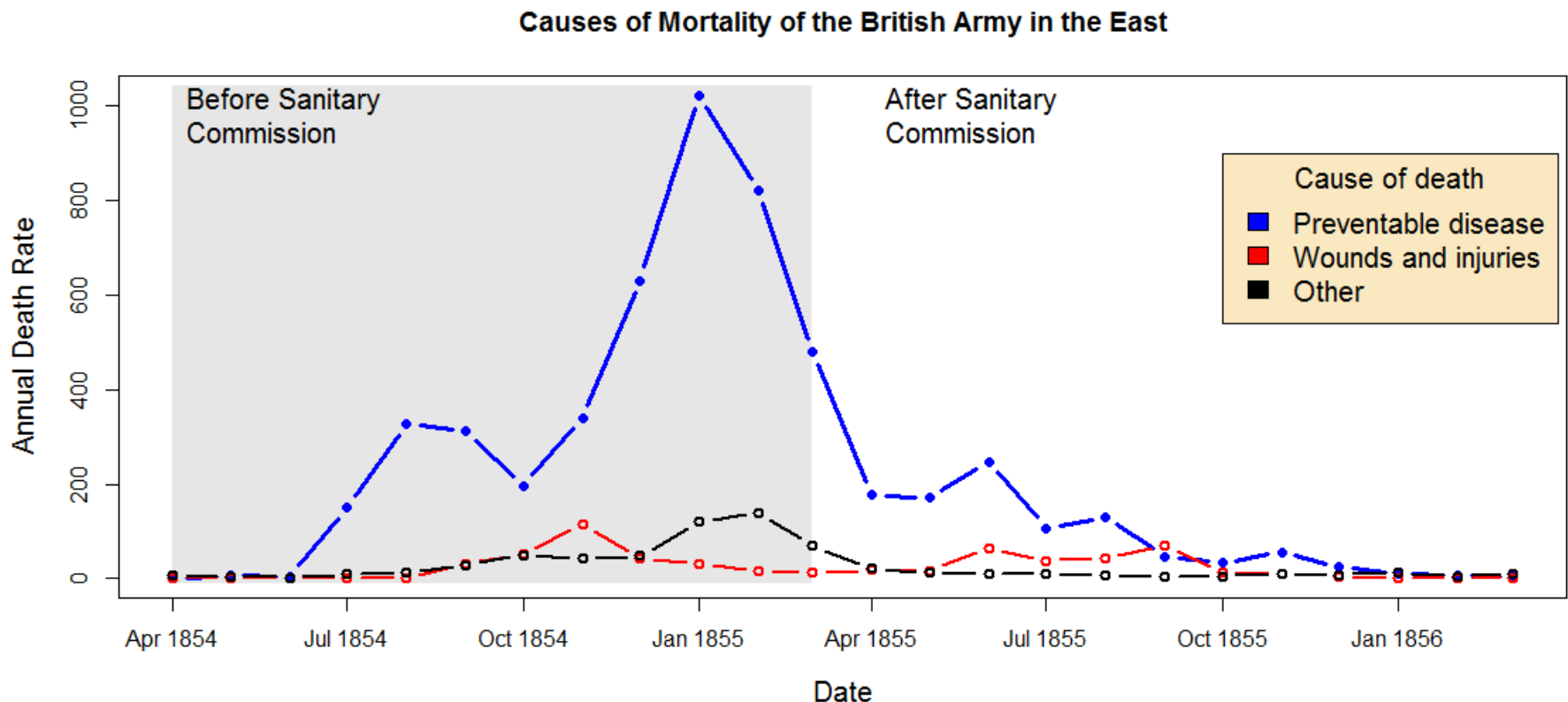
APRIL 1854 TO MARCH 1855.



Data graph: Nightingale (1857)

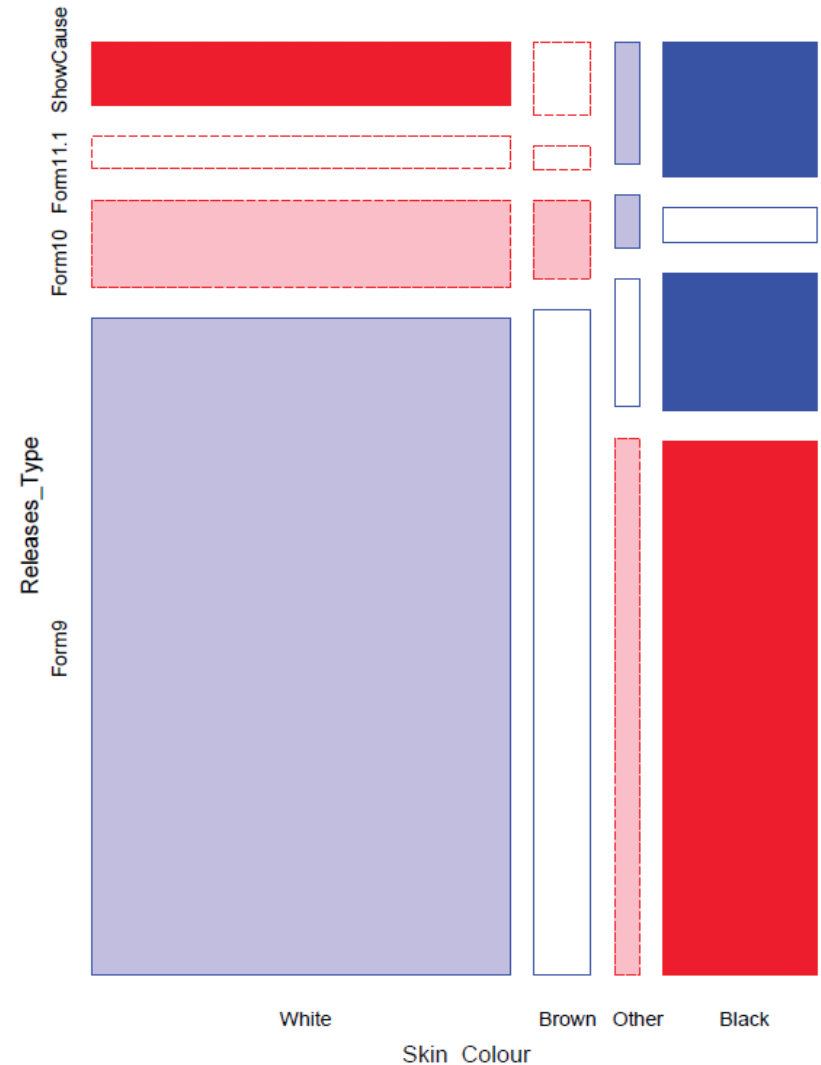
The same, as a data graph, using time-series line plots

Many statisticians might prefer this today, but it doesn't draw attention or interest as Flo's original did.



Racial profiling: Analysis graph

- Toronto Star (2002) study of police actions on a charge of simple possession of marijuana
 - release with a summons (Form 9) vs. hold for bail (Show cause)
 - Evidence for racial bias?
- First graph: mosaic display
 - area \sim frequency
 - shading: \sim residual
 - Obs > Expected in blue
 - Obs < Expected in red



Racial profiling: The process

How to communicate these results most effectively?

- What is the message? What features are directly comprehensible to the audience?

B SECTION > **TORONTO STAR** < WEDNESDAY, DECEMBER 11, 2002 ★ thestar.com

Race and Crime

Graphic designer's
early attempts



My early
attempts

York University professor Michael Friendly's expert statistical analysis provided confirmation for the Toronto Star's series on racial profiling by city police.

Man behind the numbers

Racial profiling: Presentation graphic

Together, we created this **self-explaining** infographic

Title gives the main conclusion

Text description gives details

Bar width ~ charges
Divided by % release

numbers shown in the cells

Legend gives a layman's description of shading levels

Same charge, different treatment

Statistical analysis of single drug possession charges shows that blacks are much less likely to be released at the scene and much more likely to be held in custody for a bail hearing. Darker colours represent a stronger statistical link between skin colour and police treatment.

Degree of likelihood

- Much less** likely to occur
- Much more** likely to occur
- More likely to occur

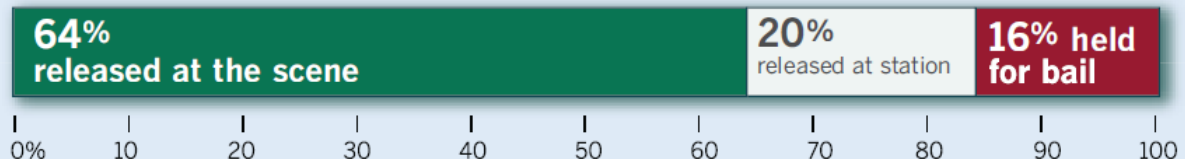
Whites are more likely to be released at the scene

6,662 charges laid



Blacks are much more likely to be held for bail hearings

2,446 charges laid



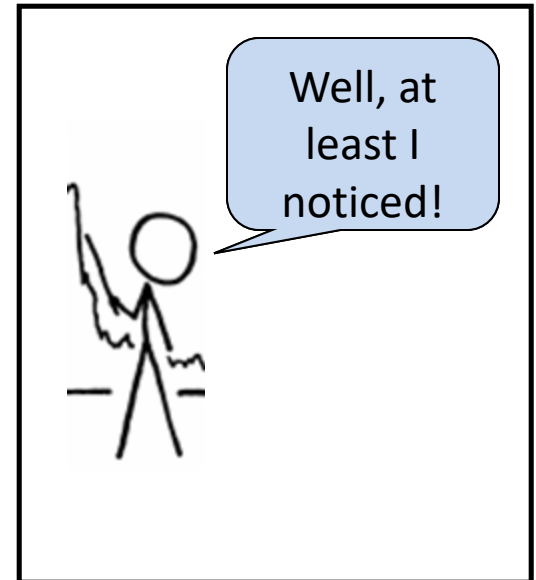
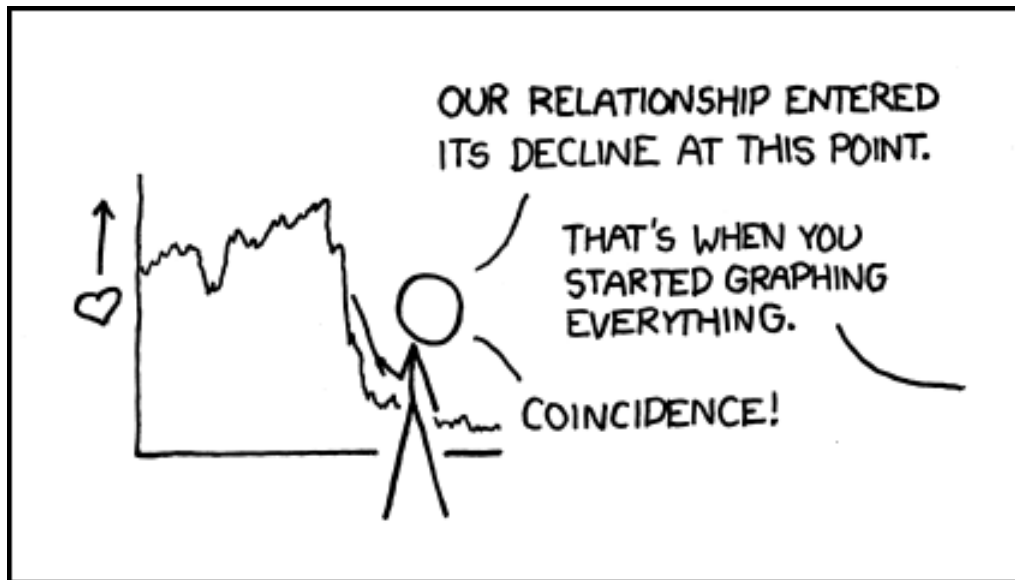
SOURCE: Toronto police arrest records 1996-2002

Why plot your data?

Graphs help us to see

patterns, trends, anomalies and other features

not otherwise easily apparent from numerical summaries.

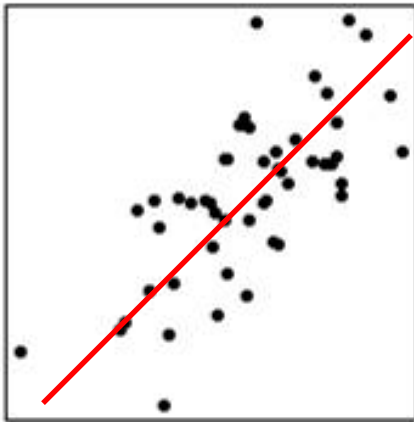


Source: <http://xkcd.com/523/>

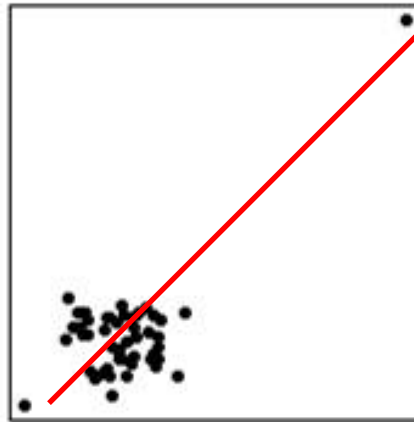
Why plot your data?

Three data sets with exactly the same bivariate summary statistics:

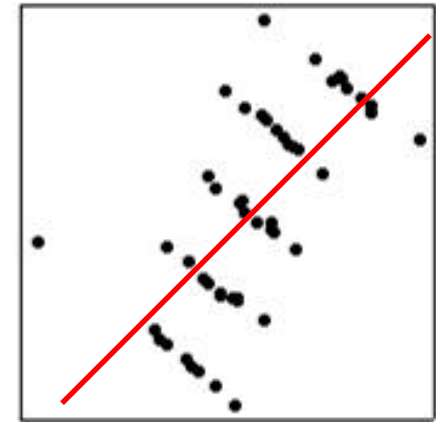
- Same correlations, linear regression lines, etc
- Indistinguishable from standard printed output
- Totally different interpretations!



Standard data



$r=0$ but + 2 outliers

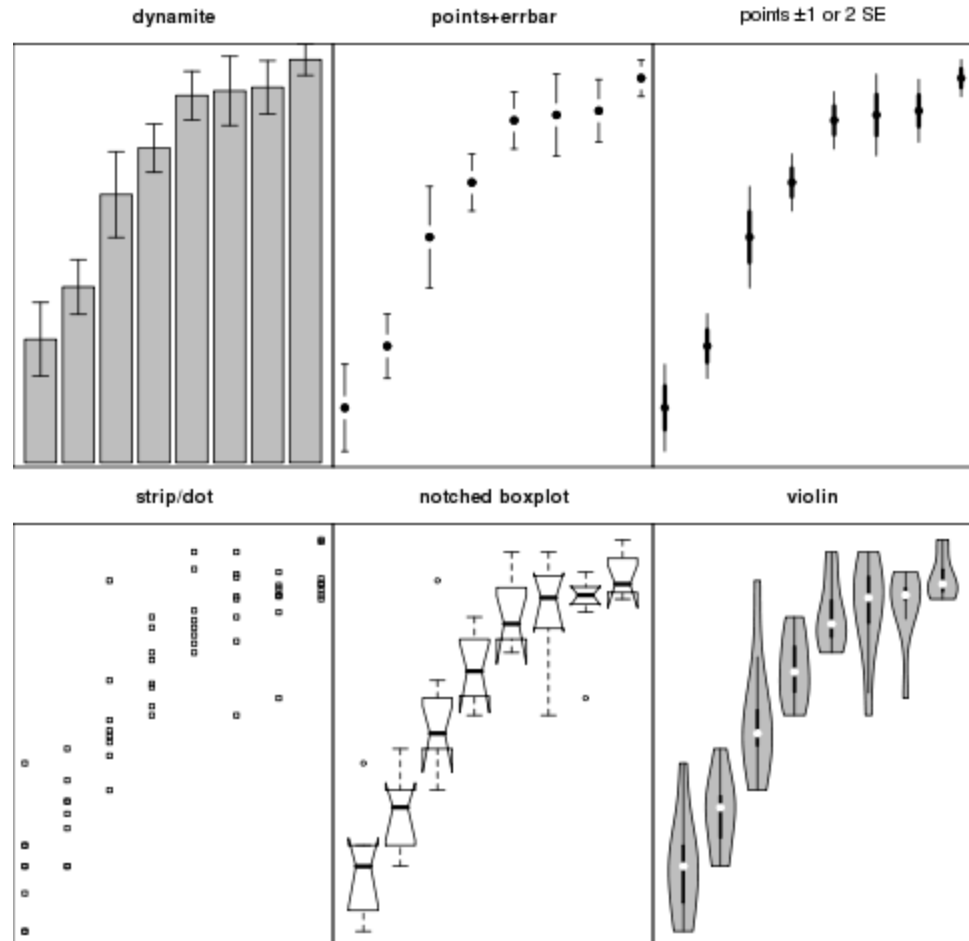


Lurking variable?

Comparing groups: Analysis vs. Presentation graphs

Six different graphs for comparing groups in a one-way design

- which group means differ?
- equal variability?
- distribution shape?
- what do error bars mean?
- unusual observations?



Never use dynamite plots

Always explain what error bars mean

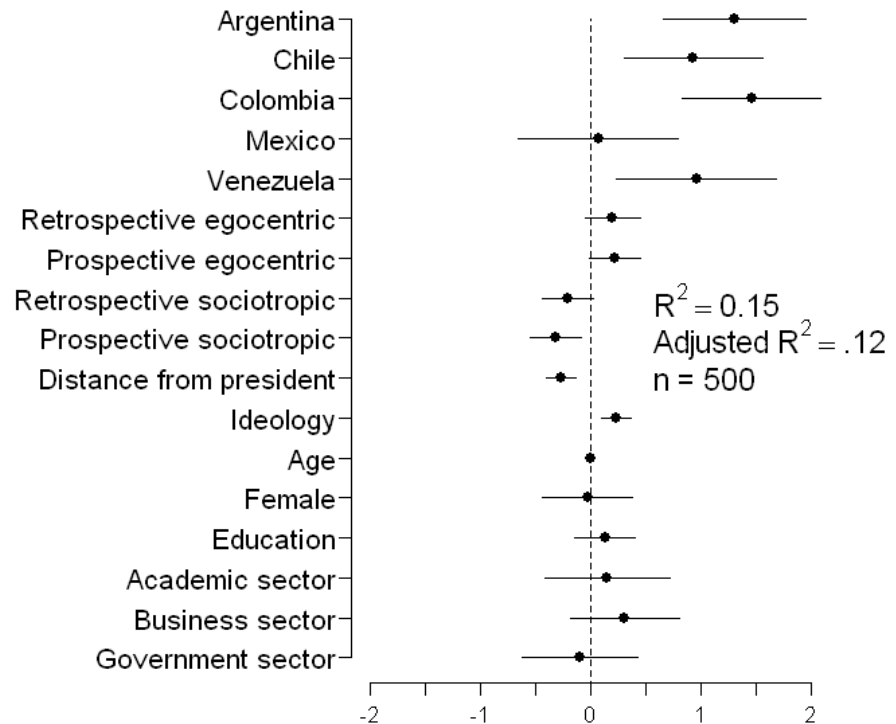
Consider tradeoff between
summarization & exposure

Presentation: Turning tables into graphs

Table 2 from Stevens (2006): Determinants of Authoritarian Aggression

Variable	Coefficient (Standard Error)
Constant	.41 (.93)
Countries	
Argentina	1.31 (.33)### B,M
Chile	.93 (.32)### B,M
Colombia	1.46 (.32)### B,M
Mexico	.07 (.32) ^{A,CH,CO,V}
Venezuela	.96 (.37)### B,M
Threat	
Retrospective egocentric economic perceptions	.20 (.13)
Prospective egocentric economic perceptions	.22 (.12) [#]
Retrospective sociotropic economic perceptions	-.21 (.12) [#]
Prospective sociotropic economic perceptions	-.32 (.12)###
Ideology	
Ideology	.23 (.07)###
Individual Differences	
Age	.00 (.01)
Female	-.03 (.21)
Education	.13 (.14)
Academic Sector	.15 (.29)
Business Sector	.31 (.25)
Government Sector	-.10 (.27)
R ²	.15
Adjusted R ²	.12
n	500
### p < .01, # p < .05, * p < .10 (two-tailed)	
^A Coefficient is significantly different from Argentina's at p < .05;	
^B Coefficient is significantly different from Brazil's at p < .05;	
^{CH} Coefficient is significantly different from Chile's at p < .05;	
^{CO} Coefficient is significantly different from Colombia's at p < .05;	
^M Coefficient is significantly different from Mexico's at p < .05;	
^V Coefficient is significantly different from Venezuela's at p < .05	

Graphs of model coefficients are often clearer than tables



Source: tables2graphs.com

Graphs & Information design

- Graphs & info displays should be viewed in relation to communication goals & audience
- Some criteria for assessing:
 - **comprehensibility**: does it make information as easy to understand as possible?
 - **attention**: does the audience take notice?
 - **aesthetics**: is it visually appealing?
 - **memorability**: will they remember it?
 - **behavior**: does it result in some desired action?

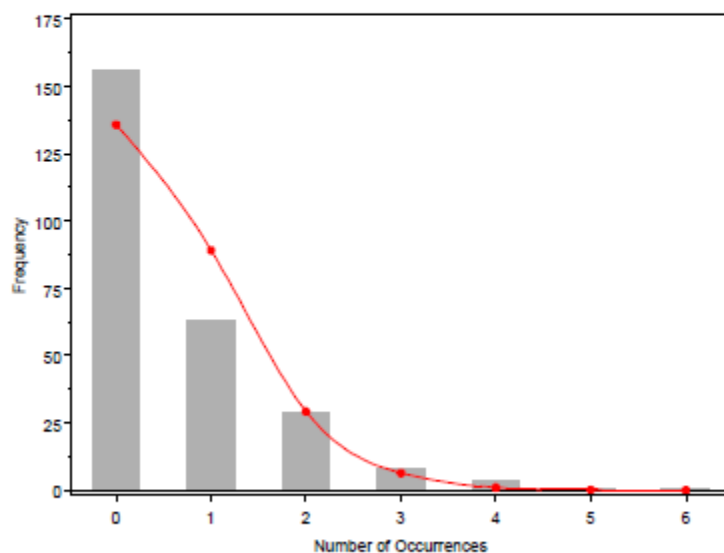
From: Ben Jones, To Optimize or to Satisfice in Data Visualization? <https://dataremixed.com/2016/01/optimize-or-satisfice-in-dataviz/>

Effective data display

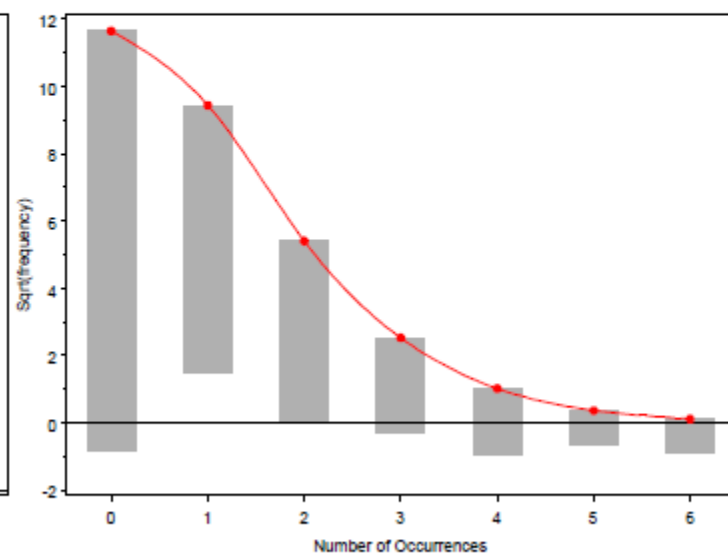
- Make the data stand out
 - Fill the data region (axes, ranges)
 - Use visually distinct symbols (shape, color) for different groups
 - Avoid chart junk, heavy grid lines that detract from the data
- Facilitate comparison
 - Emphasize the important comparisons visually
 - Side-by-side easier than in separate panels
 - “data” vs. a “standard” easier against a horizontal line
 - Show uncertainty where possible
- Effect ordering
 - For variables and unordered factors, arrange them according to the **effects** to be seen

Comparisons— Make visual comparisons easy

- Visual grouping— connect with lines, make key comparisons contiguous
- Baselines— compare *data* to *model* against a line, preferably horizontal
- Frequencies often better plotted on a square-root scale



Standard histogram with fit

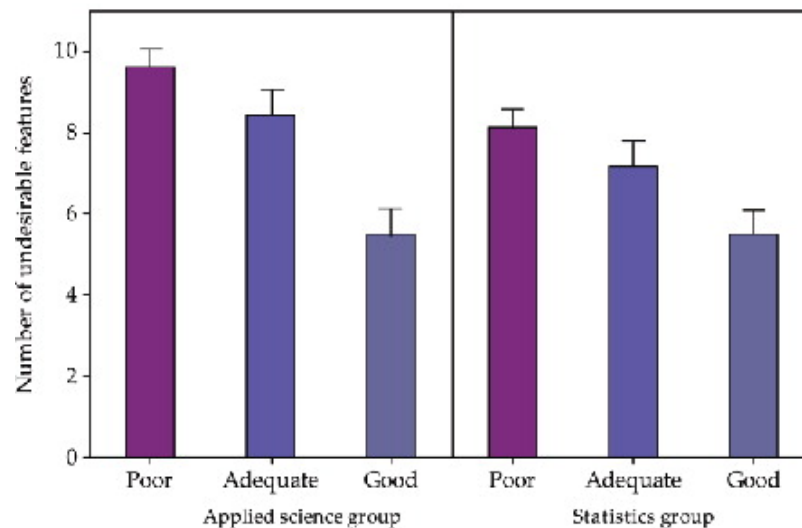


Suspended rootogram

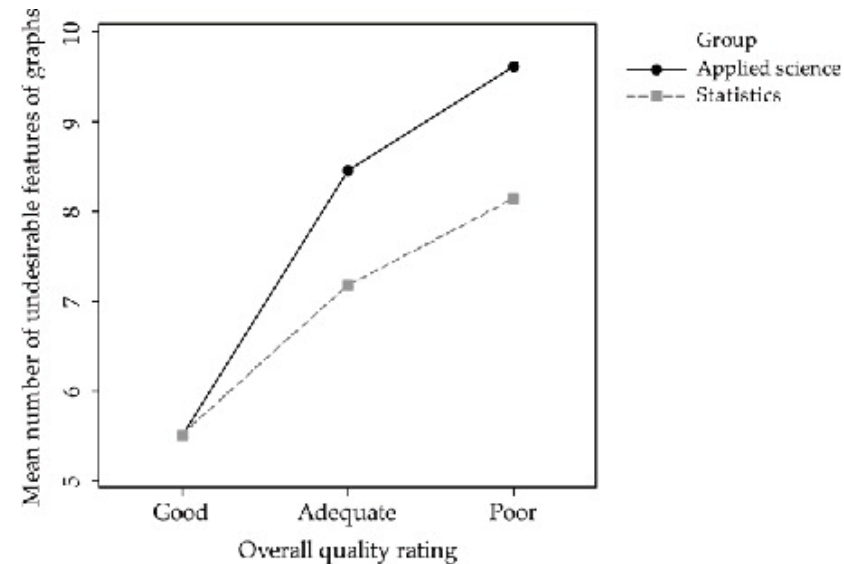
Make comparisons *direct*

- Use points not bars
- Connect similar by lines
- Same panel rather than different panels

Is there evidence of an interaction here?



???

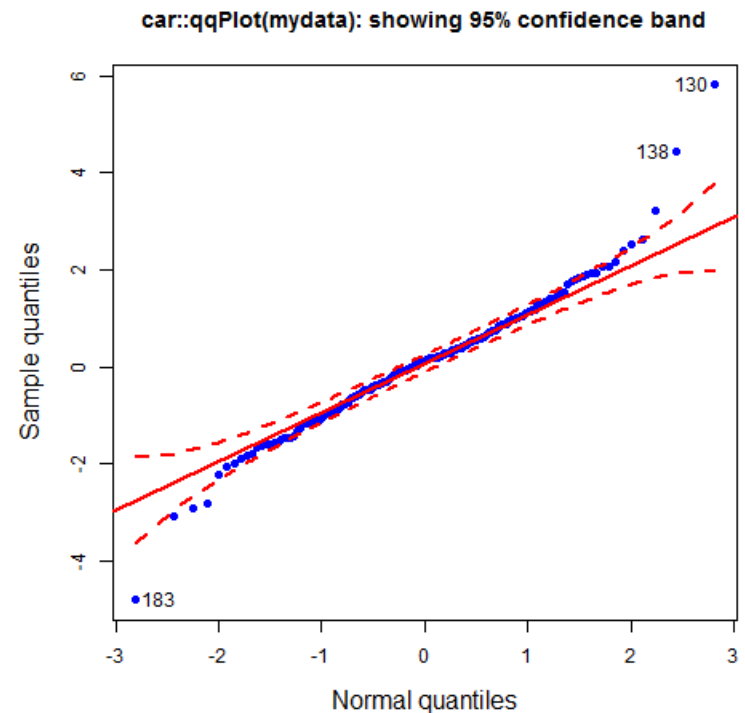
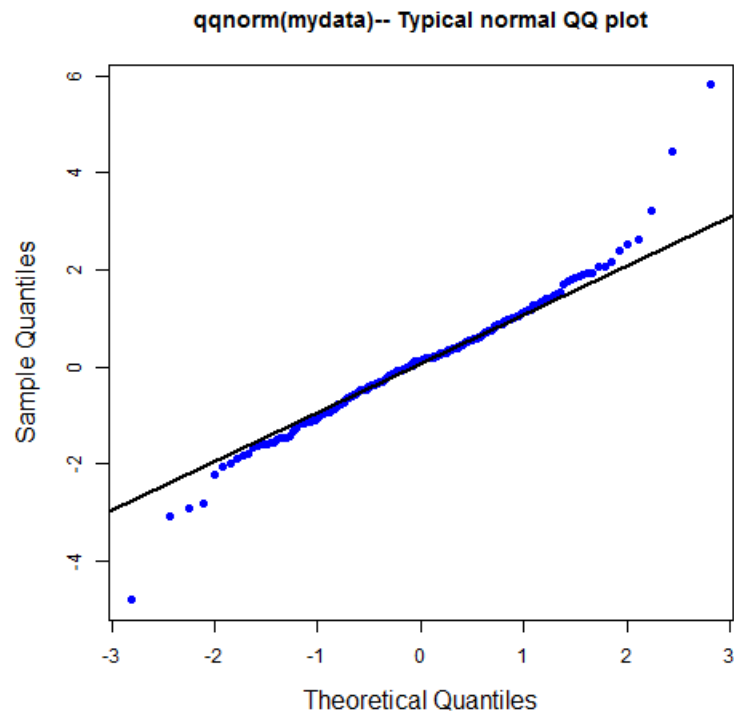


YES

Showing uncertainty

- Standard plots of observed vs. predicted lack a basis for assessment of uncertainty
- Confidence envelopes indicate extent of deviation
- Identify “noteworthy” observations to track them down

Example: Normal QQ plots used to assess normality of data




Effect ordering

- Information presentation is always **ordered**
 - in **time** or sequence (a talk or written paper)
 - in **space** (table or graph)
 - Constraints of time & space are dominant– can conceal or reveal the important message
- Effect ordering for data display
 - Sort the data by the **effects to be seen**
 - Order the data to **facilitate the task** at hand
 - lookup – find a value
 - comparison – which is greater?
 - detection – find patterns, trends, anomalies

Effect order failure: the *Challenger* disaster

- Few events in history provide as compelling illustration of importance of appropriate ordering and display of information
 - On January 28, 1986, the space shuttle Challenger exploded on take-off.
 - The cause was later determined to be that rubber O-rings failed due to cold weather
- Tables and charts presented to NASA by Thiokol engineers showed data from prior launches ordered by **time** (launch number), rather than by **temperature**—the crucial factor.
- The engineers' charts were also remarkable for information obfuscation: “erosion depth” (O-ring damage), “blow-by” (soot on O-rings), ...

HISTORY OF O-RING DAMAGE ON SRM FIELD JOINTS



SRM No.	Cross Sectional View			Top View		Clocking Location (deg)
	Erosion Depth (in.)	Perimeter Affected (deg)	Horizontal Dia. (in.)	Length Of Max Erosion (in.)	Total Heat Affected Length (in.)	
22A	None	None	0.280	None	None	35° - 66°
22A	NONE	NONE	0.280	NONE	NONE	338° - 18°
15A	0.040	154.0	0.280	4.25	5.25	163
15B	0.038	130.0	0.280	12.50	58.75	154
15B	None	45.0	0.280	None	29.50	154
13B	0.028	110.0	0.280	3.00	None	275
11A	None	None	0.280	None	None	--
10A	0.040	217.0	0.280	3.00	14.50	161
2B	0.053	116.0	0.280	--	--	90

Visual explanation: Physics

- NASA appointed members of the Rogers Commission to investigate the cause of the disaster
- the noted physicist Richard Feynman discovered the cause: at low temperature, O-rings became brittle and were subject to failure
- in his testimony, he demonstrated the effect by plunging a rubber O-ring into a cup of ice water

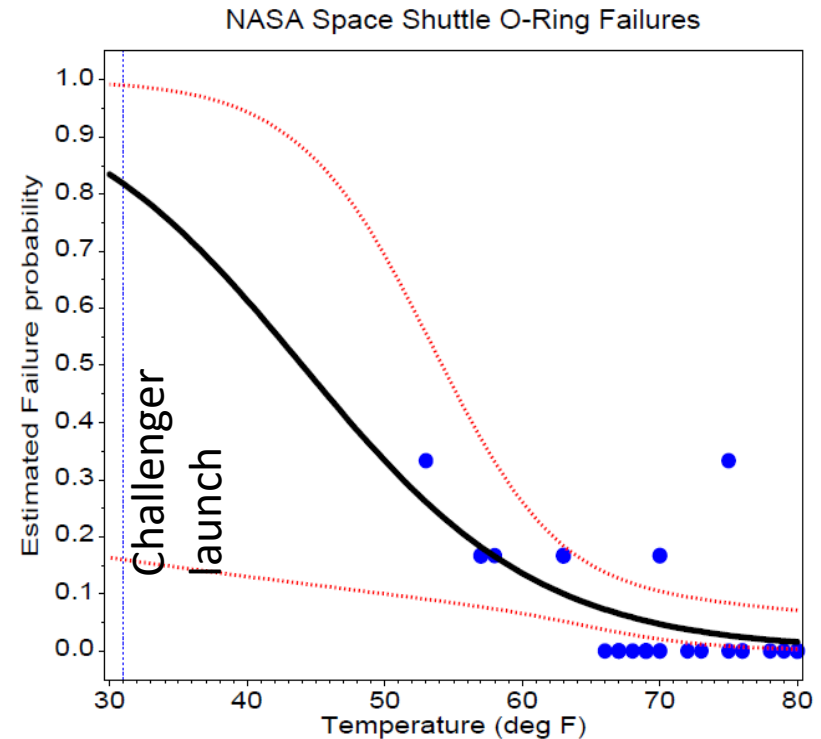


Visual explanation: Graphics

- Subsequent statistical analysis showed the relationship between launch temperature and O-ring failures
- As Tufte (1997) notes: the fatal flaw was in the **ordering** of the data.

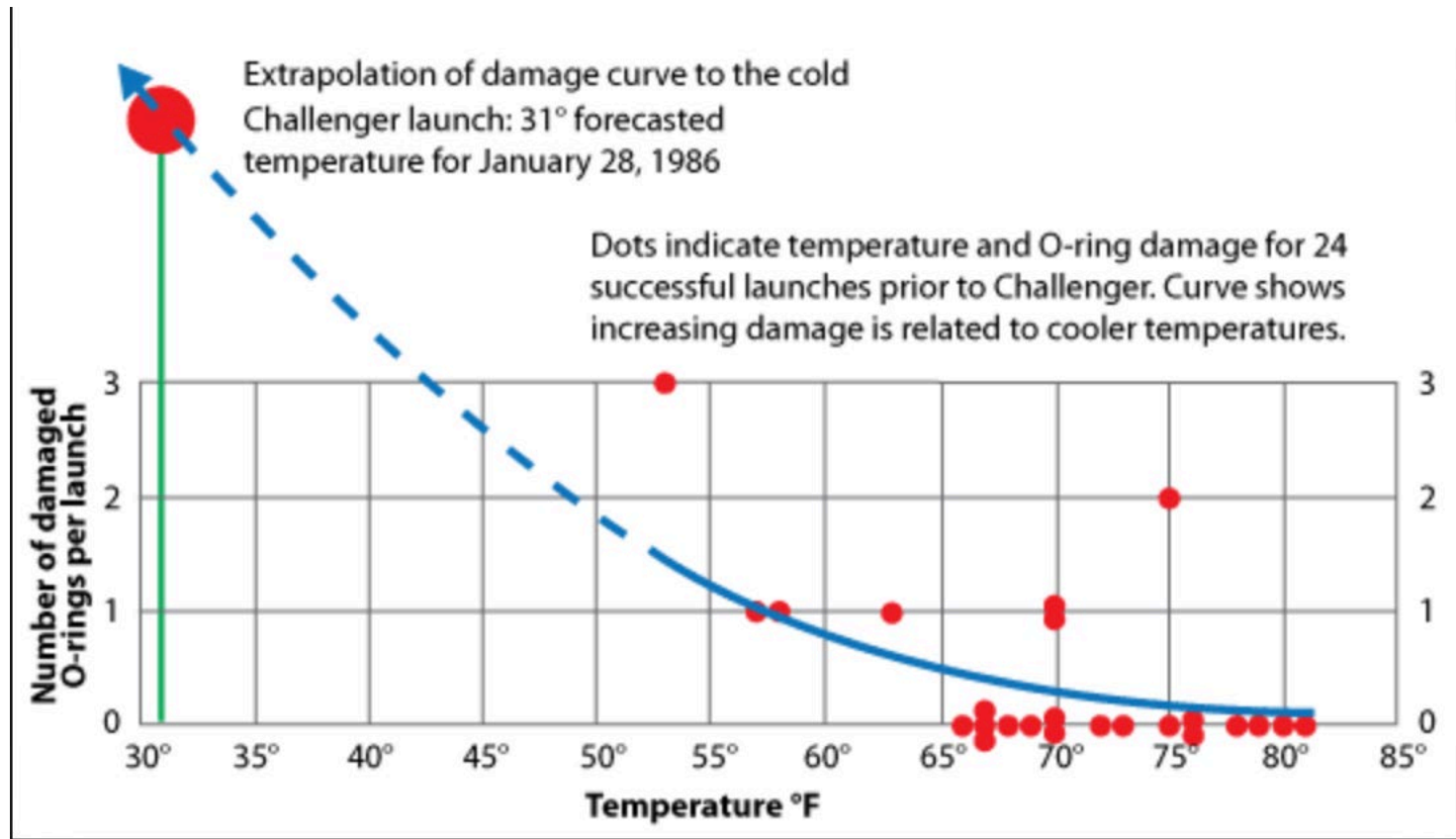
The graph shown here is the result of a statistical model fit to the data

- The **thick** line shows the predicted value of failure vs. temperature
- The **red** dotted lines show uncertainty of the predicted values



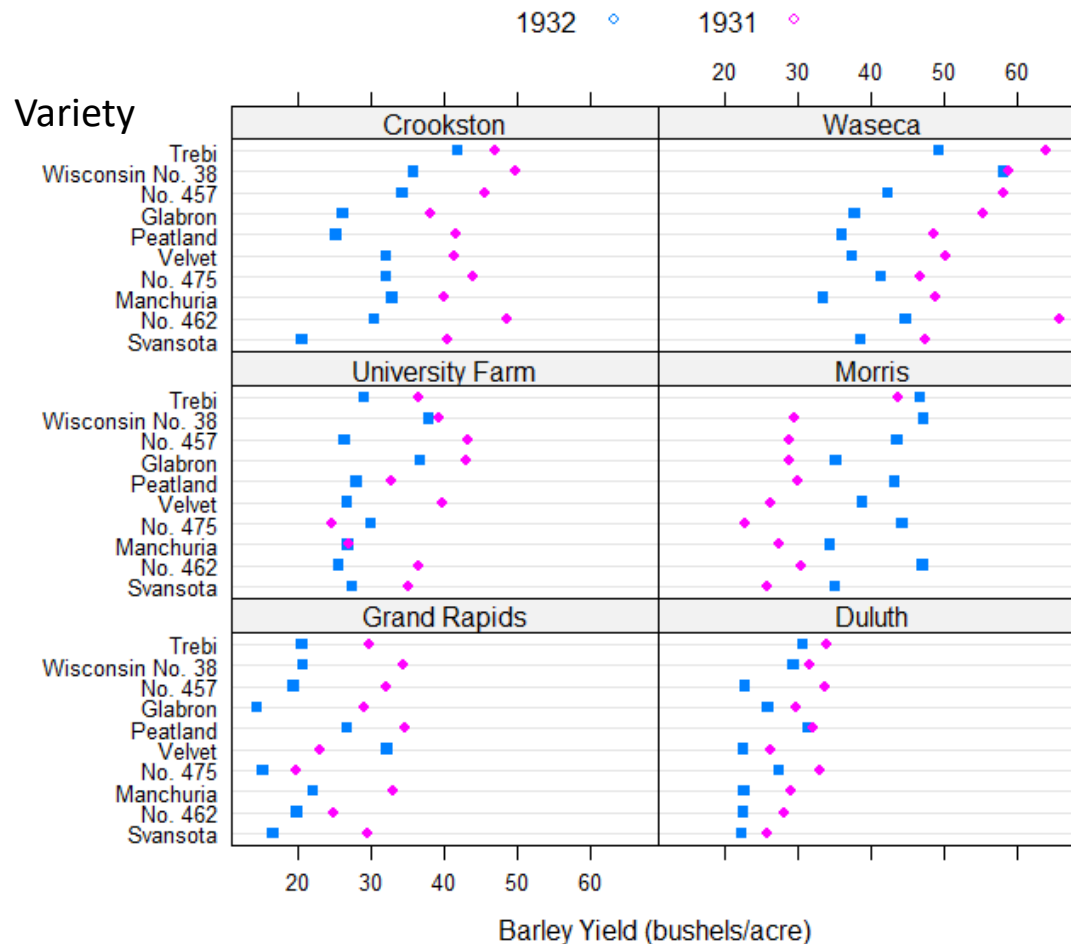
Presentation graphic

A presentation version of the previous graph alters the scales and describes the story in text annotations



Graphic displays: Main effect ordering

- To see trends, patterns, anomalies: **Sort unordered factors by means or medians**



Data on barley yields
10 varieties x 6 sites x 2 years

3 way dot plot, sorted by
main effect means

- Which site has the highest yield?
- Which variety is highest on average?
- Which site stands out in pattern over year?

Tabular displays: Main effect ordering

- Tables are often presented with rows/cols ordered **alphabetically**
 - good for lookup
 - bad for seeing patterns, trends, anomalies

Table 1: Average Barley Yields (rounded), Means by Site and Variety

Variety	Site						<i>Mean</i>
	Crookston	Duluth	Grand Rapids	Morris	University Farm	Waseca	
Glabron	32	28	22	32	40	46	33.3
Manchuria	36	26	28	31	27	41	31.5
No. 457	40	28	26	36	35	50	35.8
No. 462	40	25	22	39	31	55	35.4
No. 475	38	30	17	33	27	44	31.8
Peatland	33	32	31	37	30	42	34.2
Svansota	31	24	23	30	31	43	30.4
Trebi	44	32	25	45	33	57	39.4
Velvet	37	24	28	32	33	44	33.1
Wisconsin No. 38	43	30	28	38	39	58	39.4
<i>Mean</i>	37.4	28.0	24.9	35.4	32.7	48.1	34.4

Tabular displays: Main effect ordering

- Better: sort rows/cols by means/medians
- Shade cells according to residual from additive model

Table 2: Average Barley Yields, sorted by Mean, shaded by residual from the model $\text{Yield} = \text{Variety} + \text{Site}$

Variety	Site						Mean
	Grand Rapids	Duluth	University Farm	Morris	Crookston	Waseca	
Svansota	23	24	31	30	31	43	30.4
Manchuria	28	26	27	31	36	41	31.5
No. 475	17	30	27	33	38	44	31.8
Velvet	28	24	33	32	37	44	33.1
Glabron	22	28	40	32	32	46	33.3
Peatland	31	32	30	37	33	42	34.2
No. 462	22	25	31	39	40	55	35.4
No. 457	26	28	35	36	40	50	35.8
Wisconsin No. 38	28	30	39	38	43	58	39.4
Trebi	25	32	33	45	44	57	39.4
Mean	24.9	28.0	32.7	35.4	37.4	48.1	34.4

Tabular displays: Main effect ordering

Yield difference, $\Delta y_{ij} = 1931 - 1932$ by Variety & Site

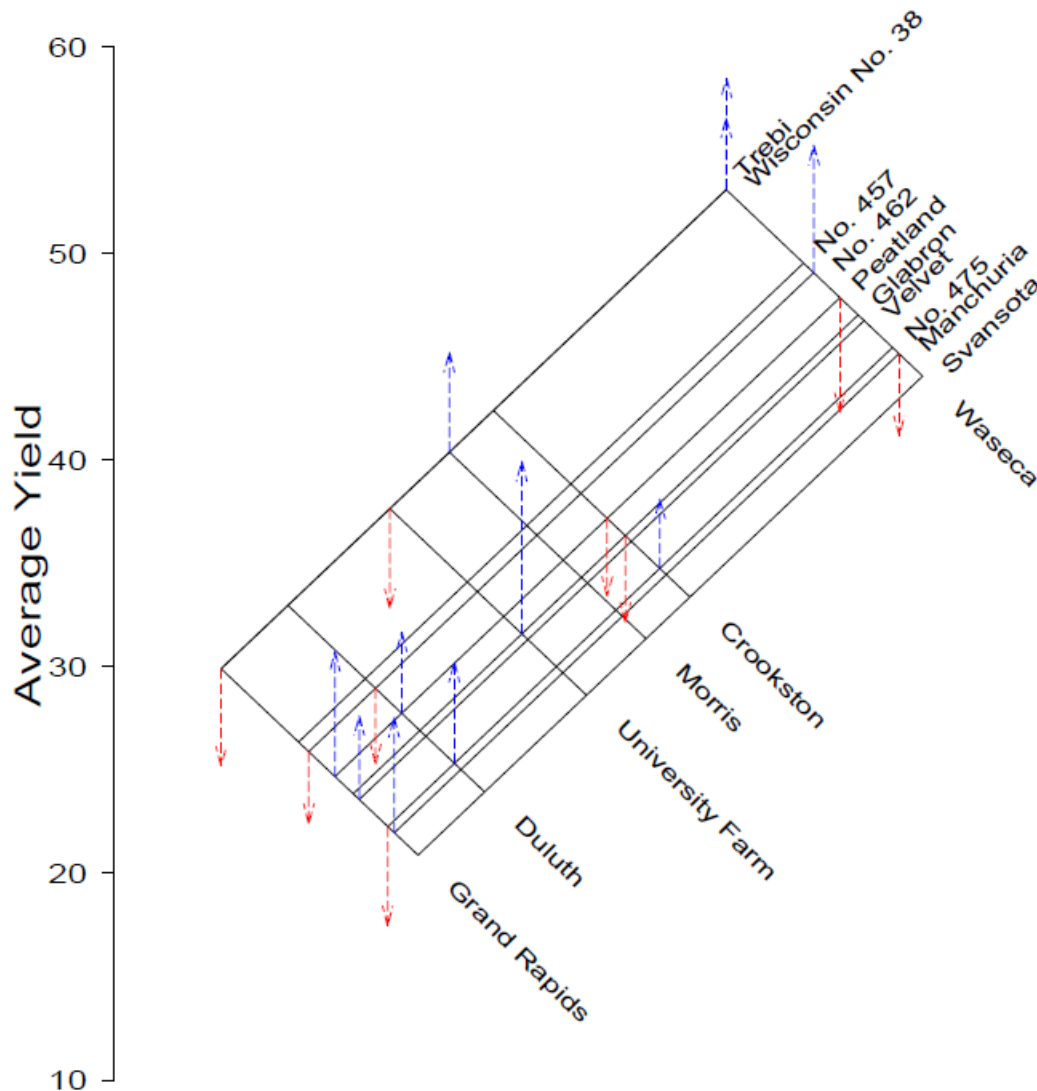
Ordered: by row and column means; **shaded:** by value ($|\Delta y_{ij}| > \{2,3\} \times \sigma(\Delta y_{ij})$)

What features stand out?

Table 3: Yield Differences, 1931-1932, sorted by mean difference, and shaded by value

Variety	Site						Mean
	Morris	Duluth	University Farm	Grand Rapids	Waseca	Crookston	
No. 475	-22	6	-5	4	6	12	0.1
Wisconsin No. 38	-18	2	1	14	1	14	2.4
Velvet	-13	4	13	-9	13	9	2.9
Peatland	-13	1	5	8	13	16	4.8
Manchuria	-7	6	0	11	15	7	5.5
Trebi	-3	3	7	9	15	5	6.1
Svansota	-9	3	8	13	9	20	7.3
No. 462	-17	6	11	5	21	18	7.4
Glabron	-6	4	6	15	17	12	8.0
No. 457	-15	11	17	13	16	11	8.8
Mean	-12.2	4.6	6.3	8.2	12.5	12.5	5.3

Graphical display: Two-way tables



Tukey two-way plot of average barley yield

If there is no interaction,

$$y_{ij} = \mu + \alpha_{\text{site}} + \beta_{\text{variety}}$$

Site & variety effects sorted automatically

Effects are spaced by fitted values

More variation among sites than varieties

Waseca best, by a wide margin

Multivariate data: correlation ordering

- Arrange ***variables*** so that:
 - Similar variables are contiguous
 - Ordered to show patterns of relations
- Arrange ***observations*** so that:
 - Similar variables are contiguous
 - Ordered to show patterns of relations

Correlation matrices

Baseball data: Batting, fielding and (log) Salary

Nobody wants to see all those decimals

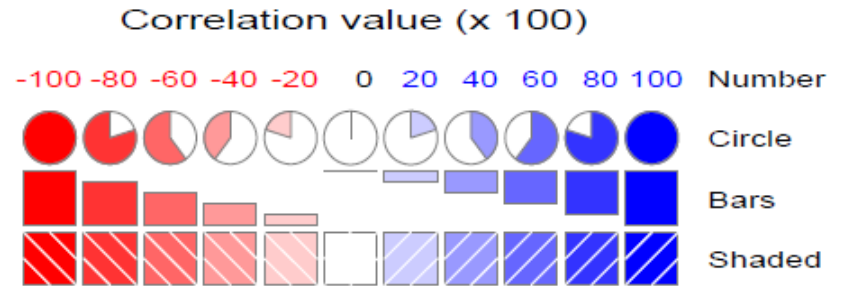
```
> cor(bb)
      Assists Atbat  Errors  Hits  Homer logSal Putouts  RBI  Runs Walks  Years
Assists 1.0000 0.3421 0.70350 0.3040 -0.16160 0.0500 -0.0434 0.0629 0.179 0.1025 -0.0851
Atbat   0.3421 1.0000 0.32558 0.9640 0.55510 0.4149 0.3096 0.7960 0.900 0.6244 0.0127
Errors  0.7035 0.3256 1.00000 0.2799 -0.00974 -0.0208 0.0753 0.1502 0.193 0.0819 -0.1565
Hits    0.3040 0.9640 0.27988 1.0000 0.53063 0.4496 0.2997 0.7885 0.911 0.5873 0.0186
Homer   -0.1616 0.5551 -0.00974 0.5306 1.00000 0.3398 0.2509 0.8491 0.631 0.4405 0.1135
logSal  0.0500 0.4149 -0.02080 0.4496 0.33983 1.0000 0.2245 0.4441 0.426 0.4324 0.5374
Putouts -0.0434 0.3096 0.07531 0.2997 0.25093 0.2245 1.0000 0.3121 0.271 0.2809 -0.0200
RBI     0.0629 0.7960 0.15015 0.7885 0.84911 0.4441 0.3121 1.0000 0.779 0.5695 0.1297
Runs    0.1793 0.8998 0.19261 0.9106 0.63108 0.4256 0.2712 0.7787 1.000 0.6970 -0.0120
Walks   0.1025 0.6244 0.08194 0.5873 0.44045 0.4324 0.2809 0.5695 0.697 1.0000 0.1348
Years   -0.0851 0.0127 -0.15651 0.0186 0.11349 0.5374 -0.0200 0.1297 -0.012 0.1348 1.0000
```

If you are going to present the numbers, round a lot

```
> round(100*cor(bb))
      Assists Atbat  Errors Hits Homer logSal Putouts RBI  Runs Walks Years
Assists    100    34     70   30   -16     5     -4    6   18    10    -9
Atbat      34    100     33   96    56    41     31   80   90    62     1
Errors      70    33    100   28    -1    -2      8   15   19     8   -16
Hits       30    96     28  100    53    45     30   79   91    59     2
Homer     -16    56     -1   53   100    34     25   85   63    44    11
logSal      5    41     -2   45    34   100     22   44   43    43    54
Putouts    -4    31      8   30    25    22    100   31   27    28    -2
RBI         6    80     15   79    85    44     31  100   78    57    13
Runs       18    90     19   91    63    43     27   78  100    70    -1
Walks      10    62      8   59    44    43     28   57   70   100    13
Years      -9     1   -16     2    11    54     -2   13   -1    13   100
```

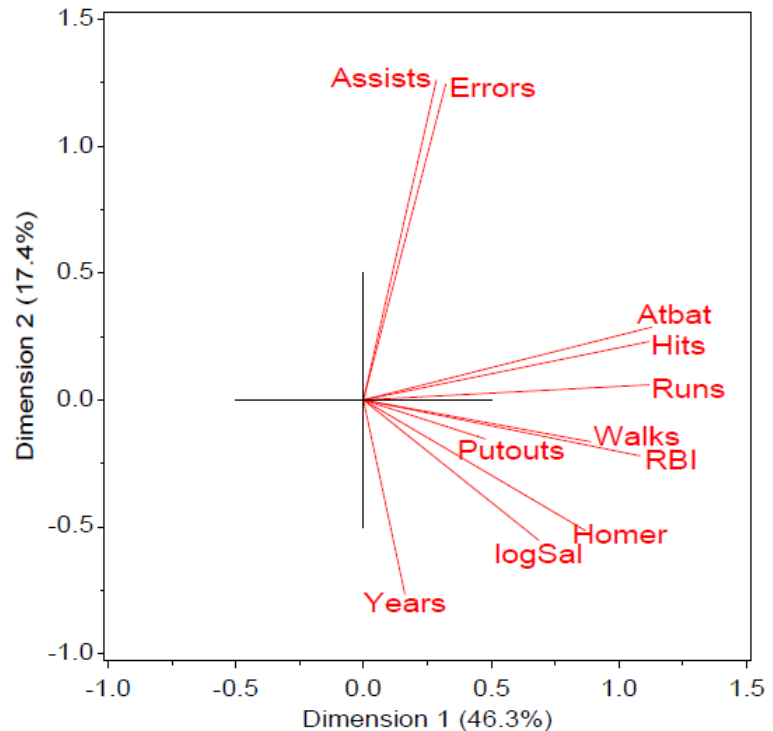
Correlation ordering: corrgrams

Rendering: a correlation value can be displayed in different ways, for different tasks



Correlation ordering:

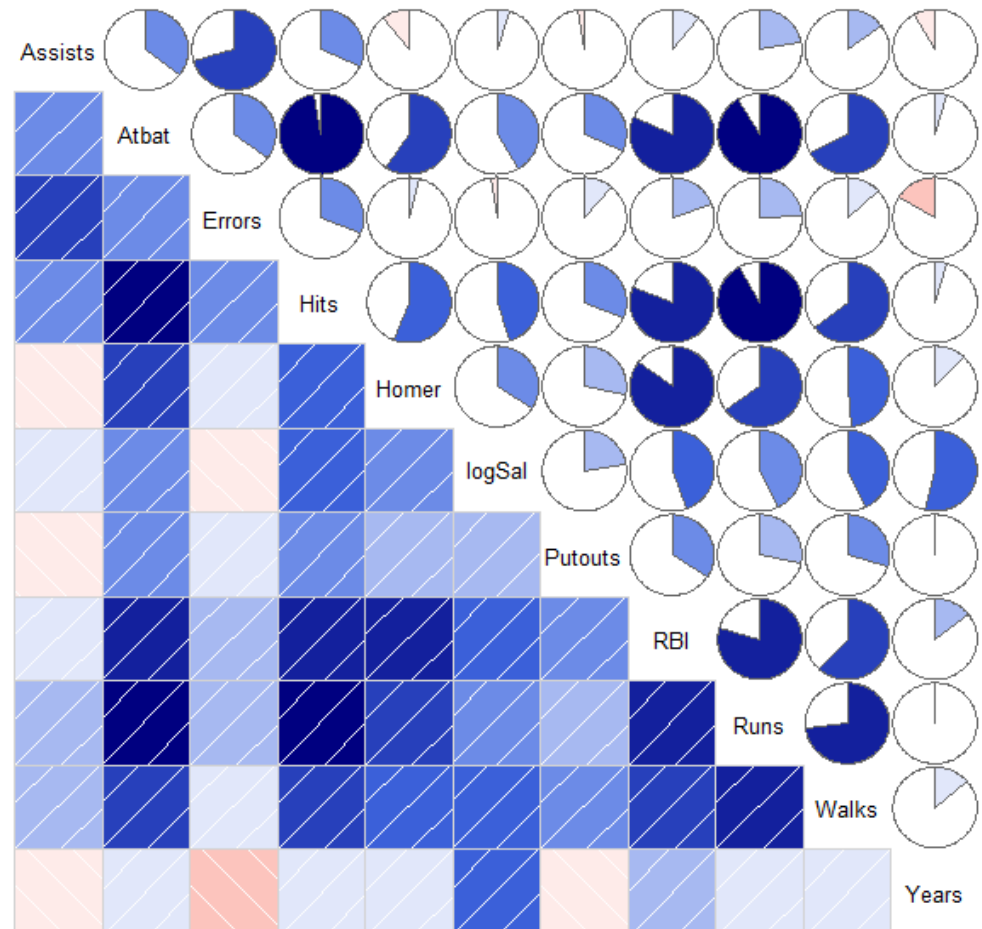
- A PCA finds weighted sums of variable to maximize variance accounted for
- Angles between vectors reflect the correlations
- → Arrange variables in the order of their angles



Baseball data

Baseball data alphabetic order

This is a corrgram display of the correlations among the baseball statistics, with the variables ordered **alphabetically**

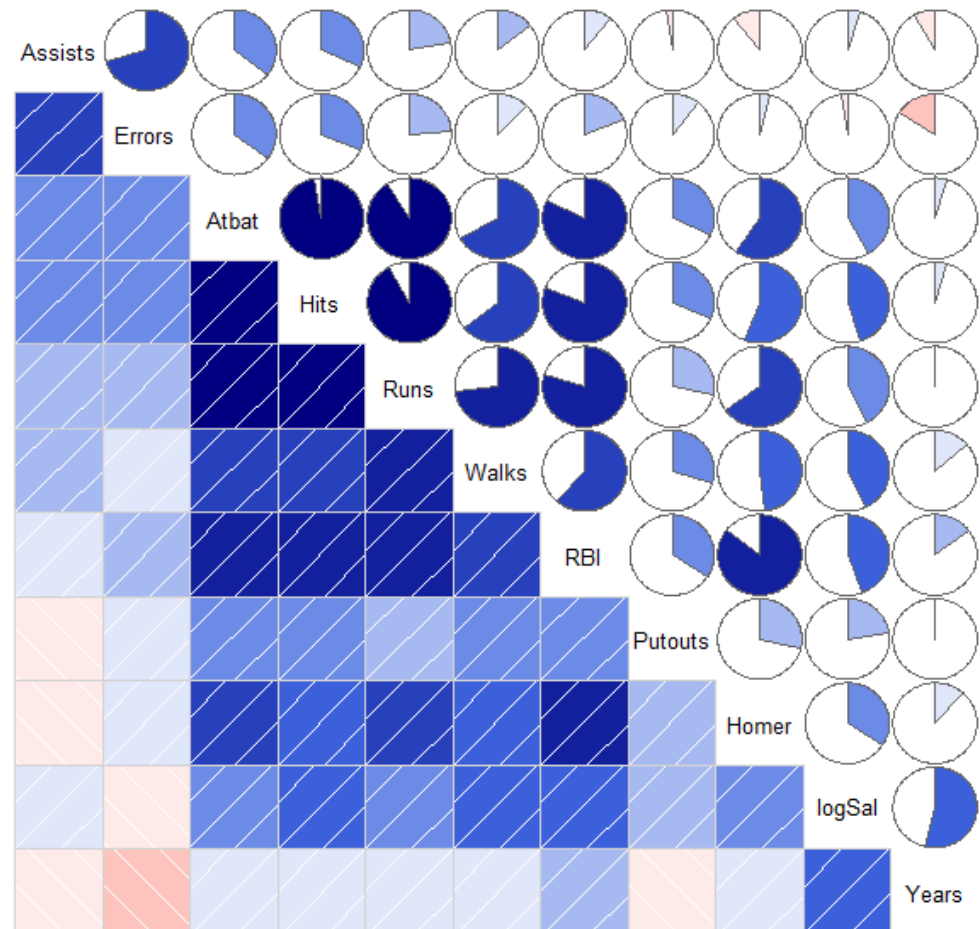


Baseball data

Baseball data PC2/PC1 order

The same display, with the variables sorted according to the angles between vectors in the PCA

Not that dramatic, but it isolates the positive & negative correlations

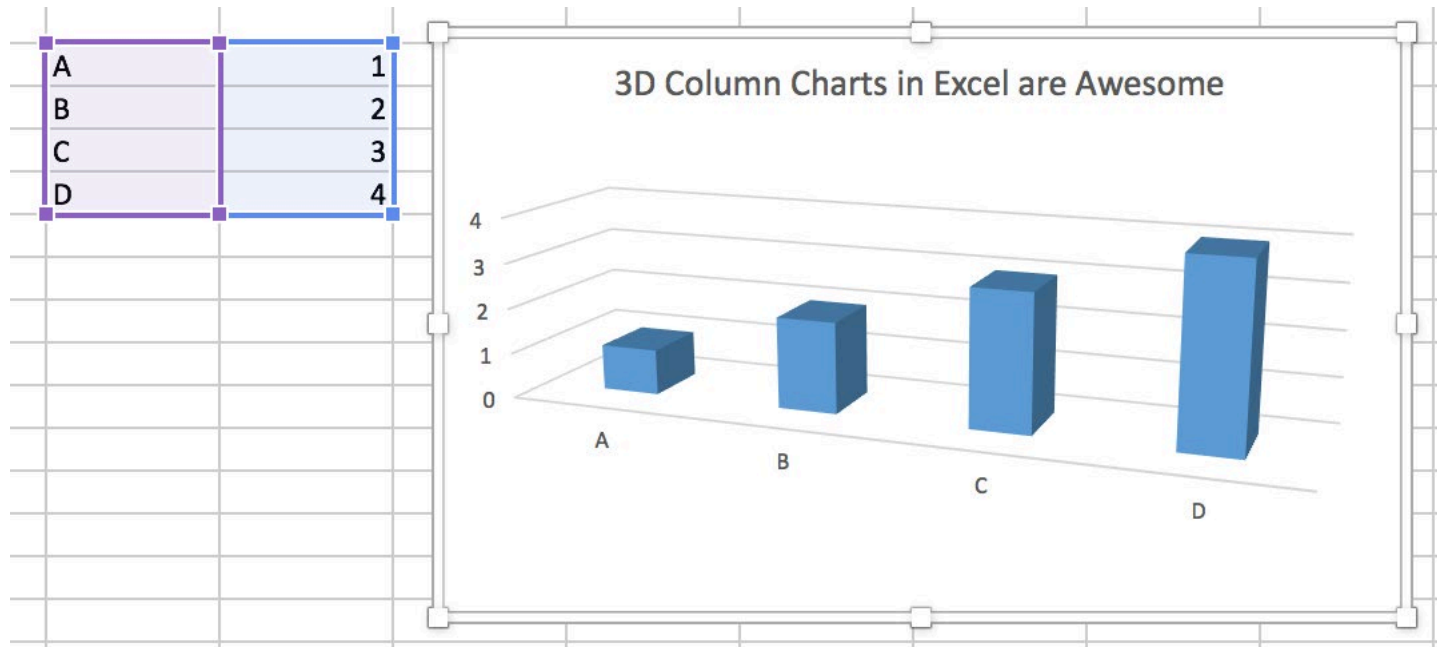


Graphs: Good/Bad, Excellent/Evil

- Like good writing, good graphical displays of data communicate ideas with:
 - clarity,
 - precision, and
 - efficiency— avoids graphic clutter
 - Even better: excellent graphs **make the message obvious**
- Like poor writing, bad graphical displays:
 - distort or obscure the data,
 - make it harder to understand or compare, or
 - thwart the communicative effect the graph should convey.
 - Even worse: **evil graphs distort, or mislead.**

Bad graphs are easy in Excel

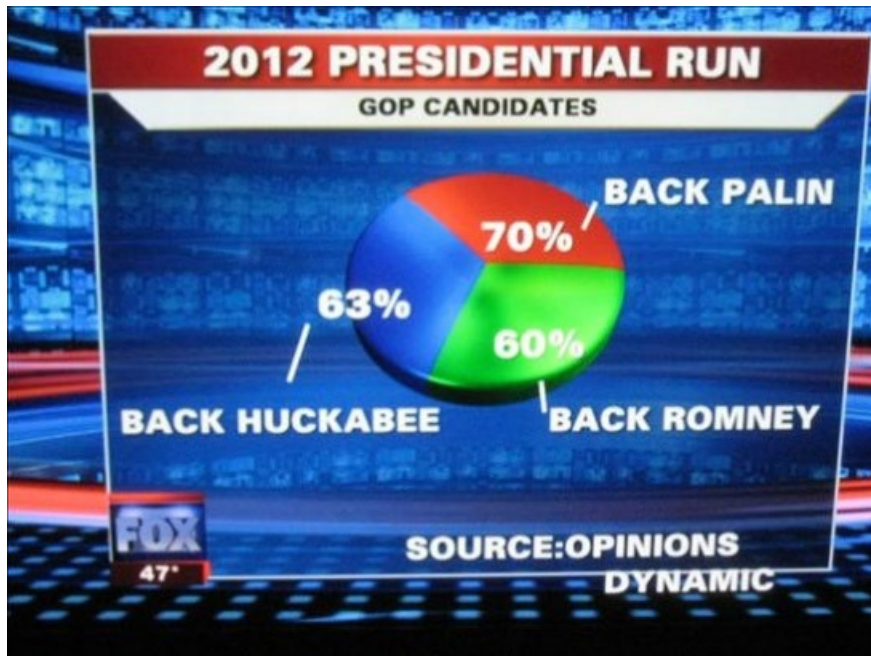
Friends don't let friends use Excel for data visualization or statistics



How many things are wrong with this graph?

Pie charts are easy to abuse

What's wrong with these pictures?



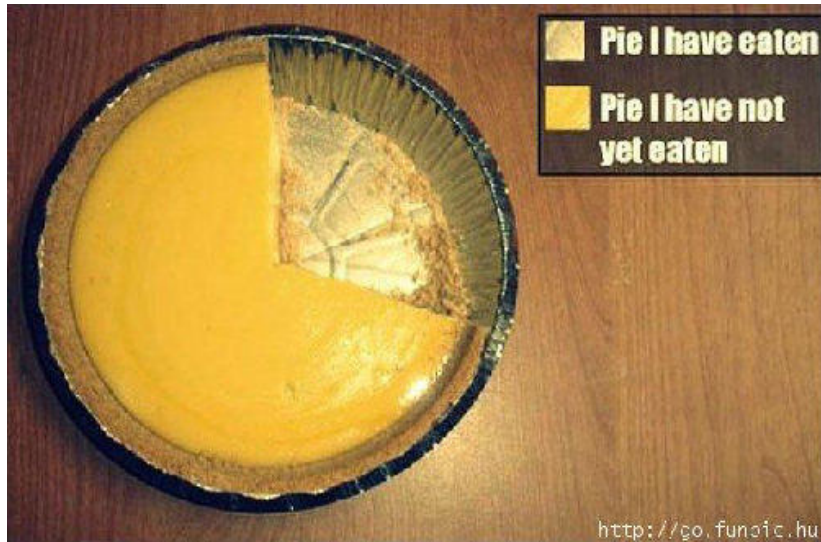
$$1 \pi = 193\%$$



$$1 \pi = 178\%$$

Why do graphic designers so often get this wrong?

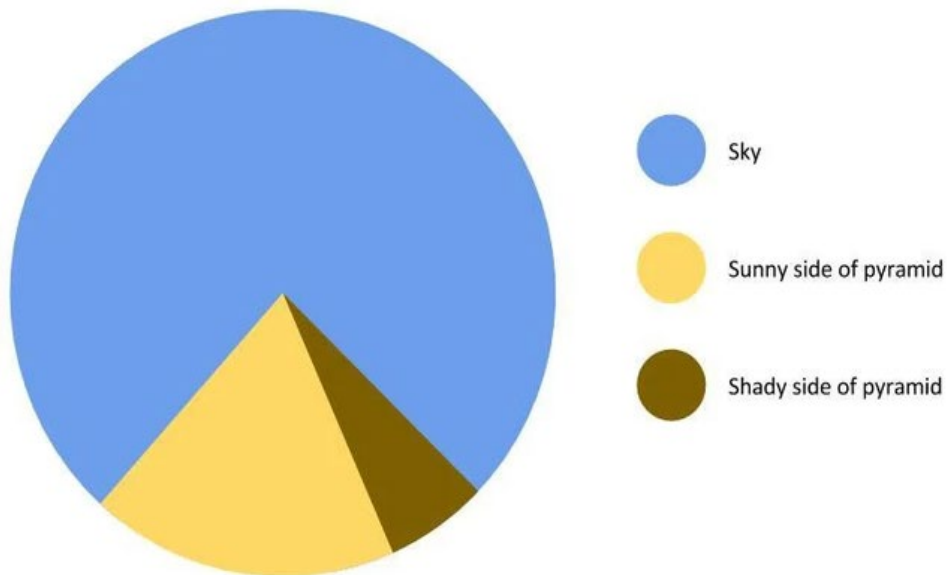
Pie chart merriment



On the other hand, pie charts are a great source of merriment for people interested in graphics

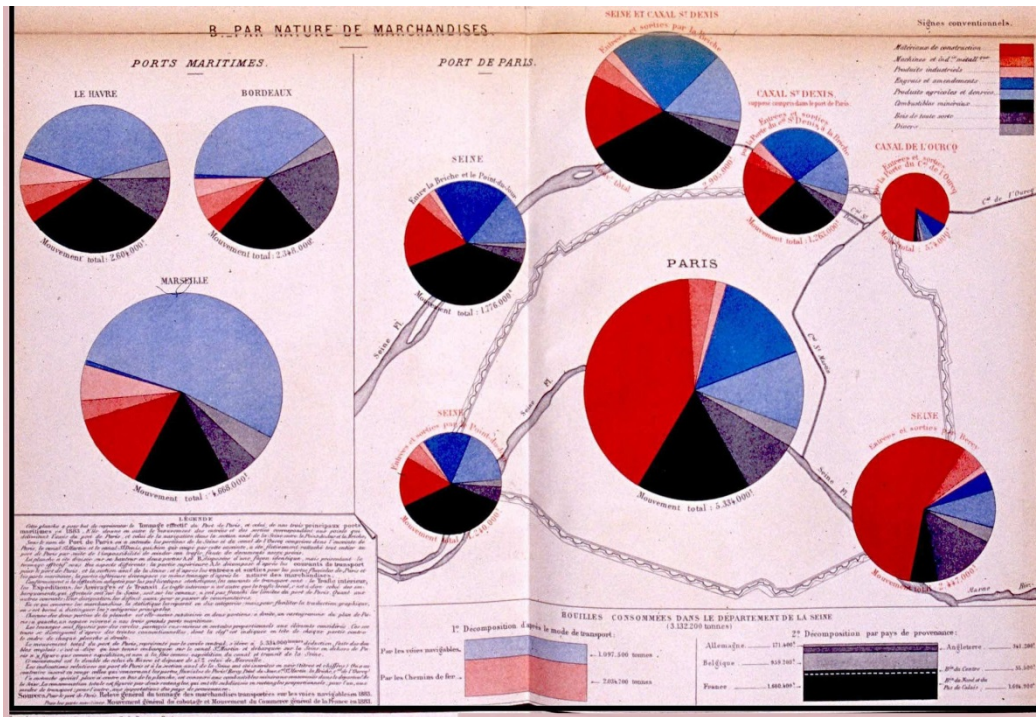
But: how much π have I eaten?

What perceptual ideas make this a great joke and lesson?



Once you see the pyramid, it's hard to see the pie.

But, can be used to great effect

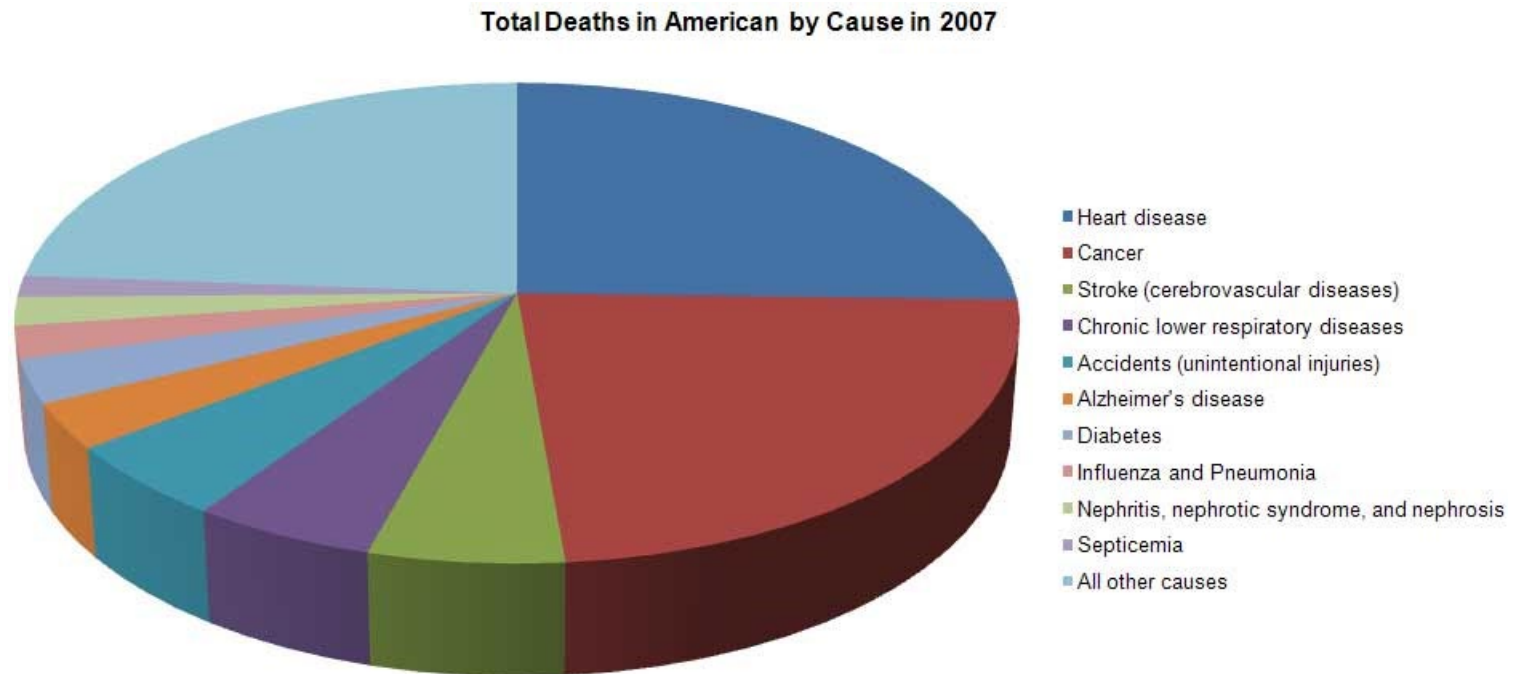


This graphic uses pie charts to show the transport of different kinds of goods to the ports of Paris and the principal maritime ports

- the size of each pie reflects **total**
- the sectors reflect **relative %**
- location places them in **context**

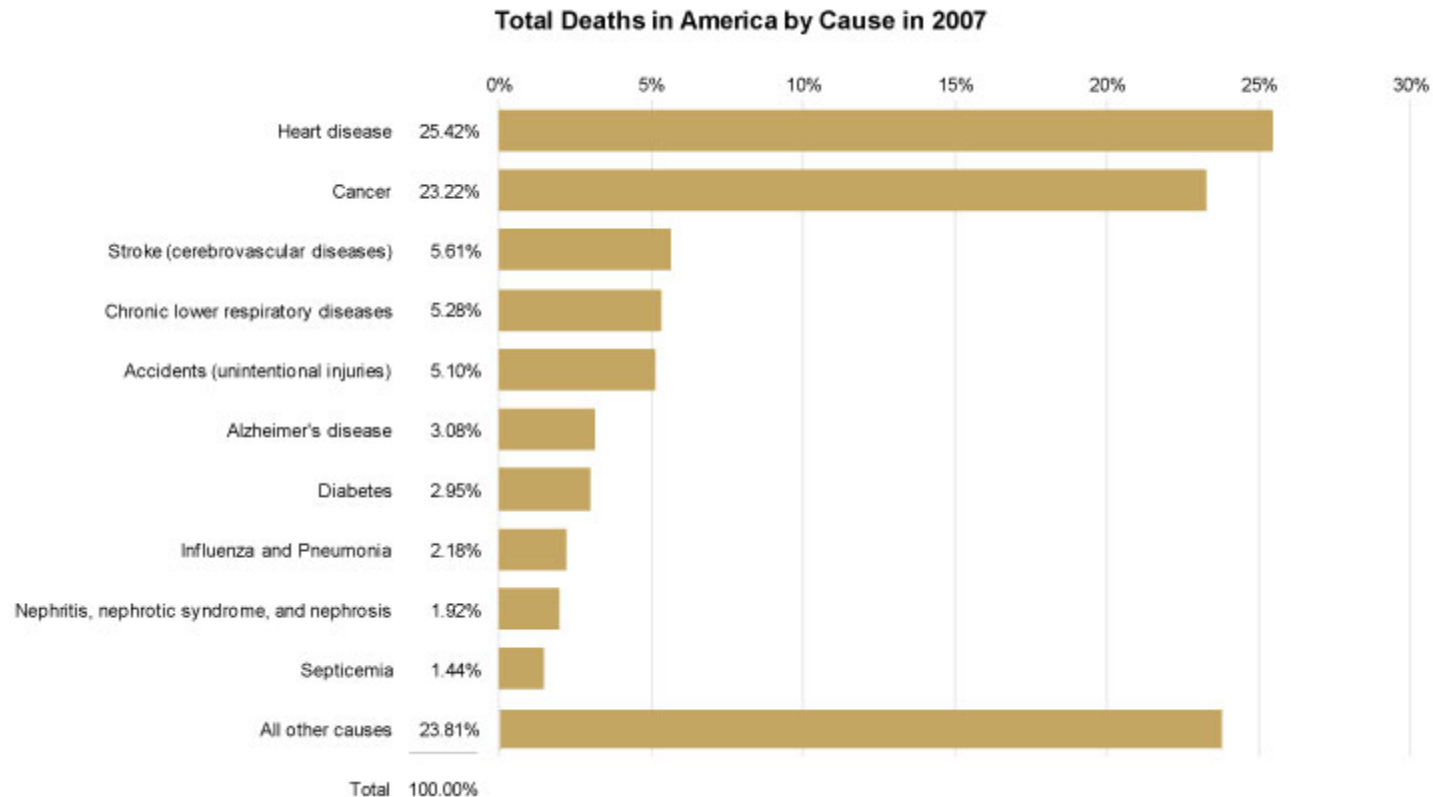
Album de Statistique Graphique, 1885, plate 17.

3D pie charts are usually evil



What was the intent of the designer of this graphic?
Which category led to the greatest total deaths?
What was the proportion of deaths due to strokes?
Did more people die from strokes vs. accidents?

Simple re-design makes it clearer

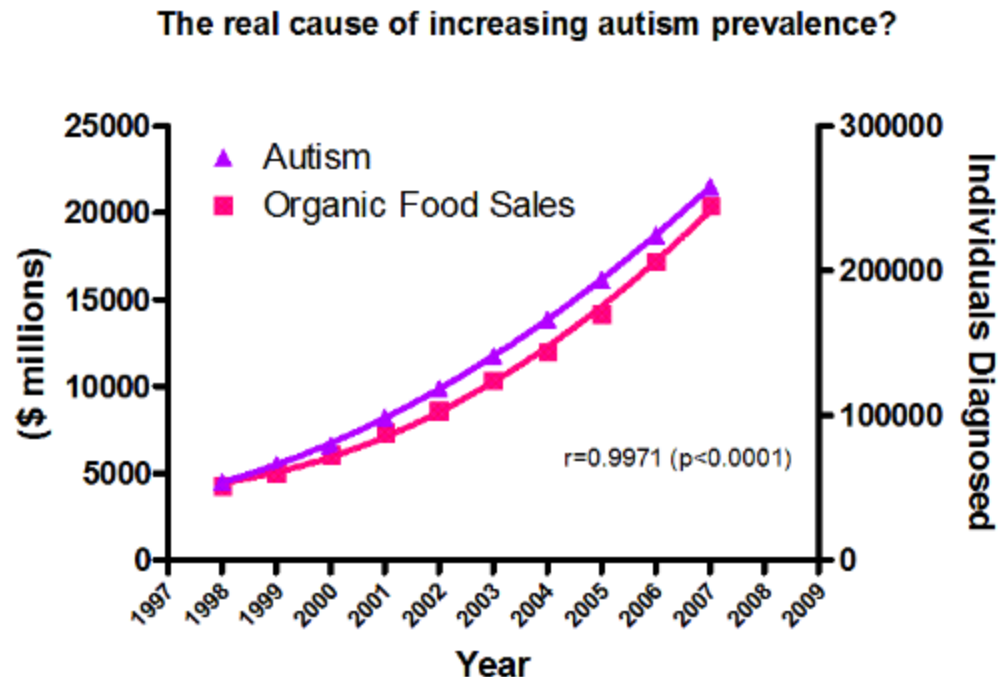


Double Y-axis: Really evil graphs

After pie charts, double Y-axis graphs have caused more trouble than almost any other

OMG, **autism** has been increasing directly with sales of **organic** food!

BAN ORGANIC FOOD!



Sources: Organic Trade Association, 2011 Organic Industry Survey; U.S. Department of Education, Office of Special Education Programs, Data Analysis System (DANS), OMB# 1820-0043; "Children with Disabilities Receiving Special Education Under Part B of the Individuals with Disabilities Education Act"

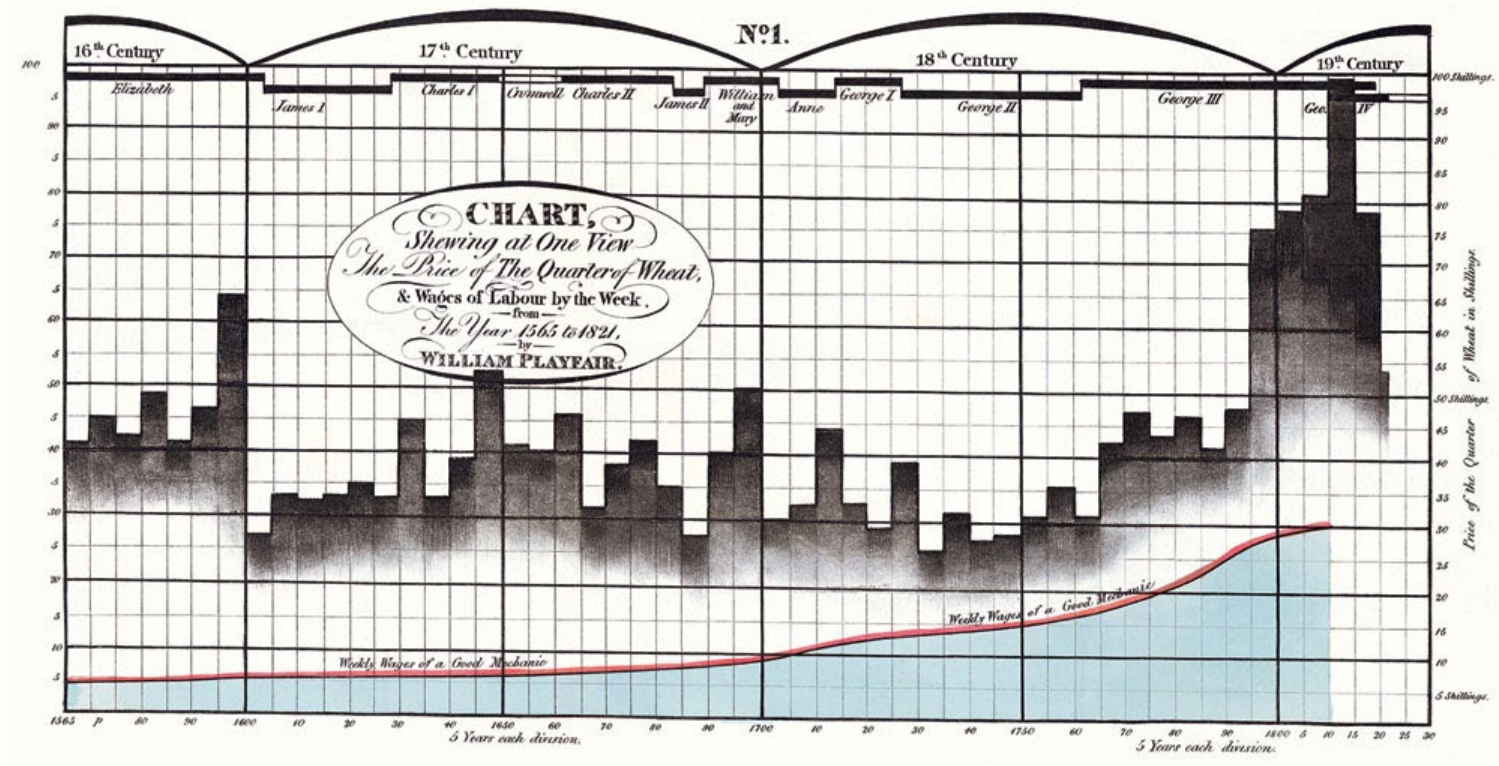
But, can be used to great effect

William Playfair invented the pie chart, line chart and bar chart.

In this figure, he shows 3 parallel time series over a 250-year period, 1560--1810

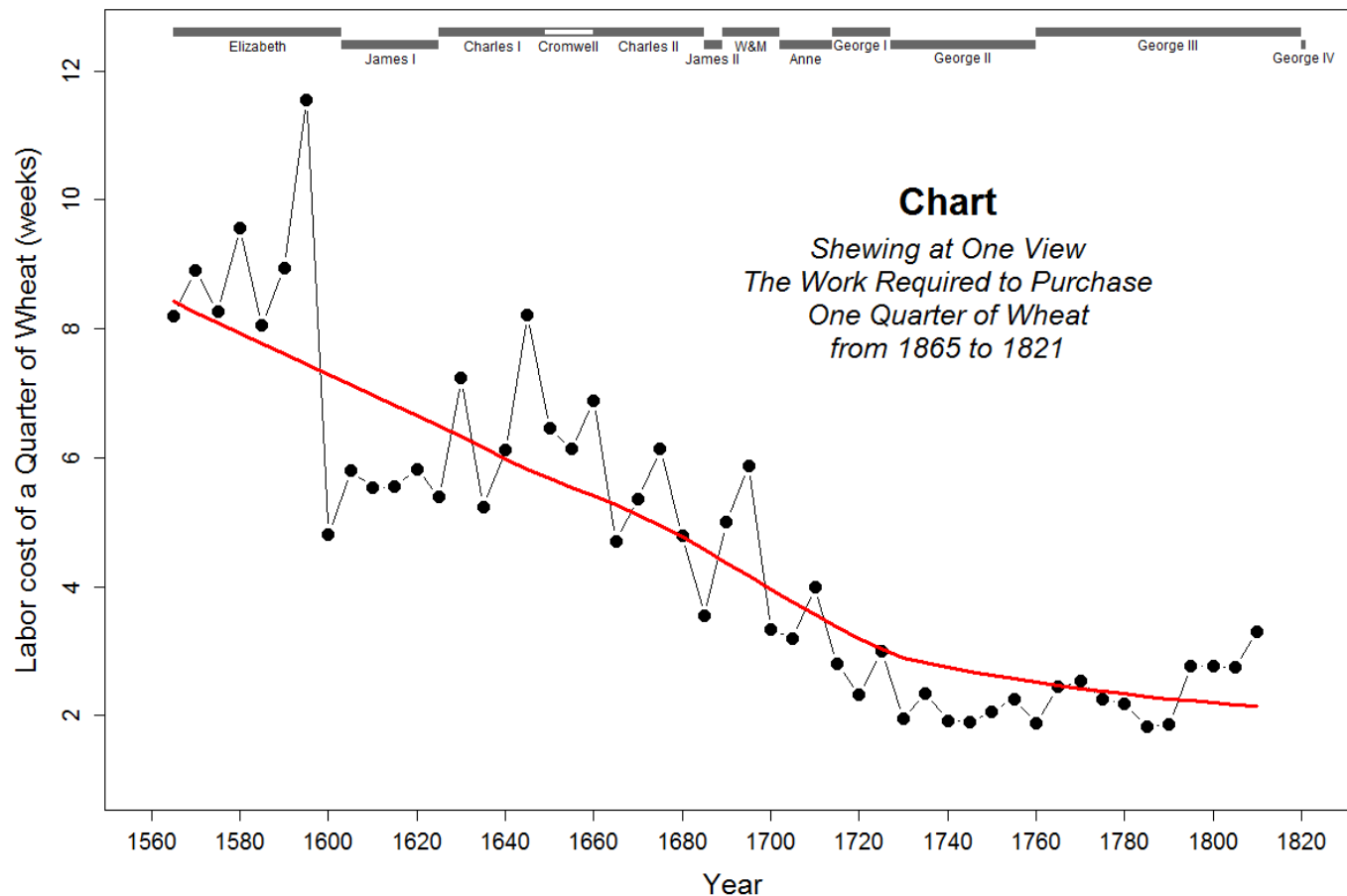
- weekly wages of a good mechanic
- price of wheat
- reigning monarch

Goal: show that workers were better off most recently (1810) than in the past



Or, another graph would have been better

A modern re-vision plots the ratio of price of wheat to wages directly



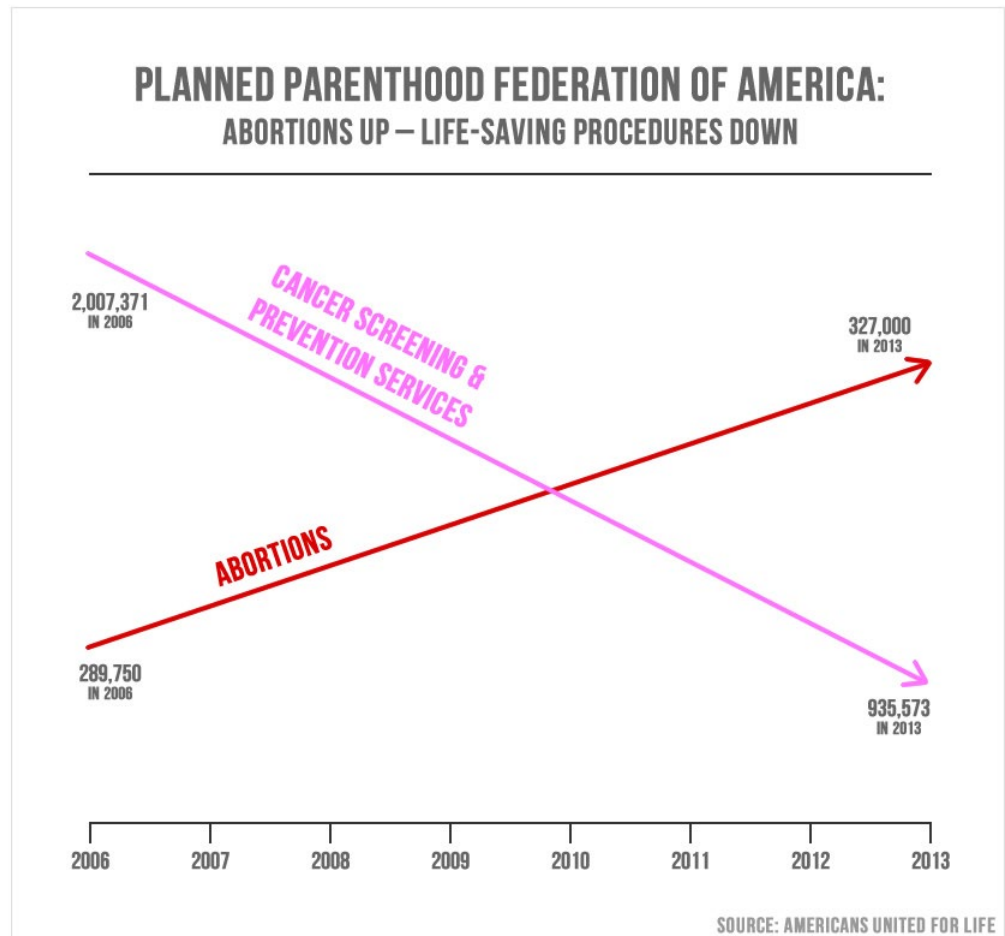
Even more evil: No scales, no data

Rep. Jason Chaffetz, R-Utah, sparred with Planned Parenthood president Cecile Richards during a high-profile hearing on Sept. 29, 2015 and presented this graph.

"In pink, that's the reduction in the breast exams, and the red is the increase in the abortions. That's what's going on in your organization."

Created by an anti-abortion group it is a deliberate attempt to mislead.

Can you see why?



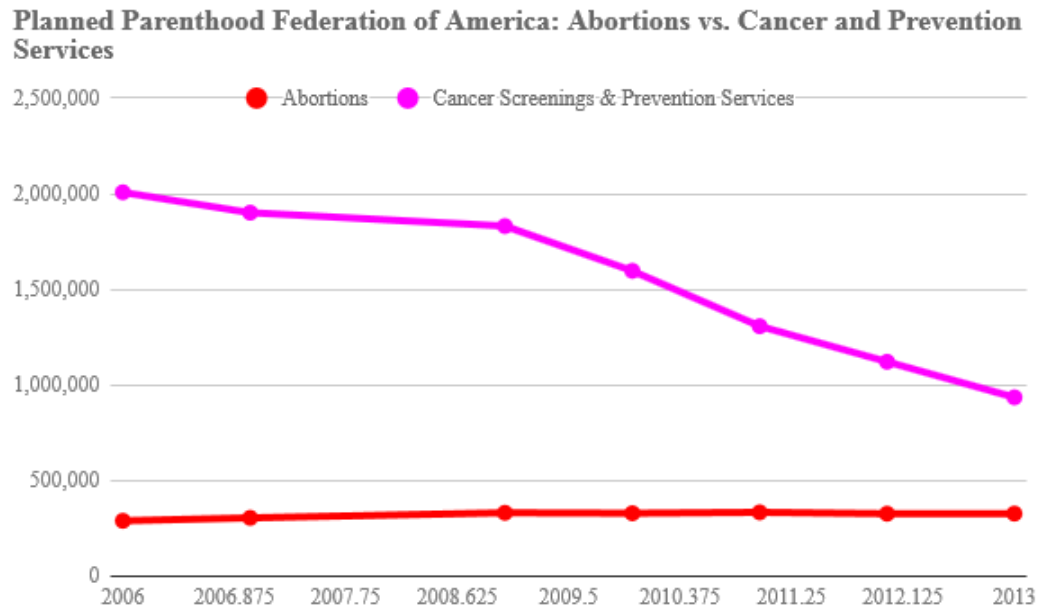
See: <http://www.politifact.com/truth-o-meter/statements/2015/oct/01/jason-chaffetz/chart-shown-planned-parenthood-hearing-misleading-/>

Corrected graph

This graph shows the actual data from the Planned Parenthood reports used by Americans United for Life

The number of abortions was relatively steady.

Some services like pap smears, dropped due to changing medical standards about who should be screened and how often.



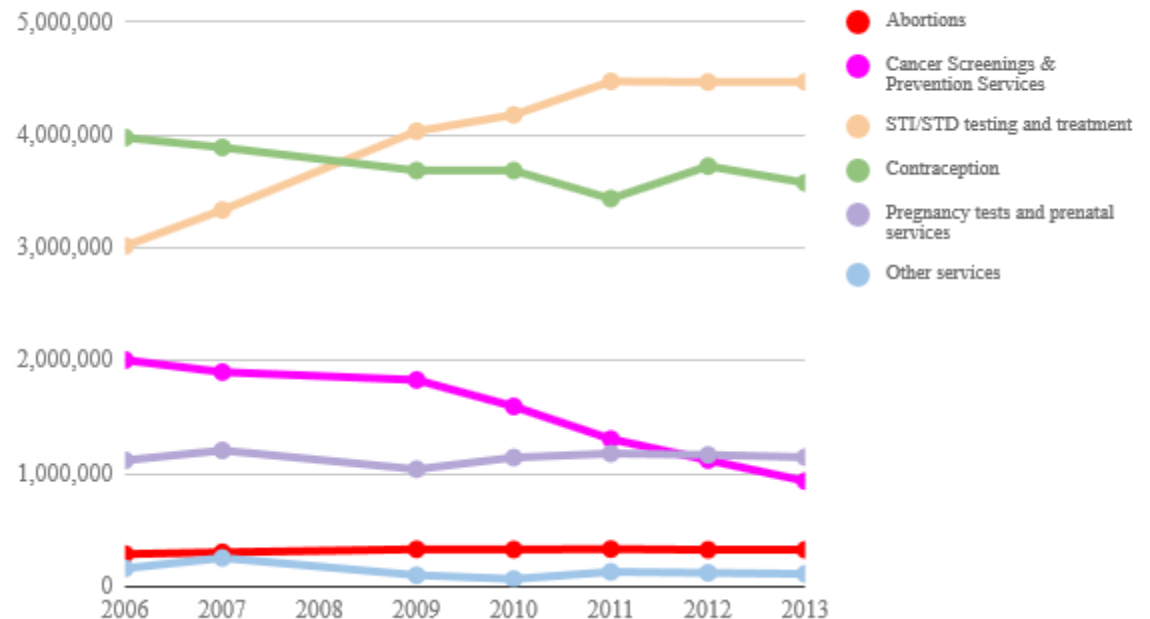
What are a few improvements that could be made to this graph?

Corrected graph, in context

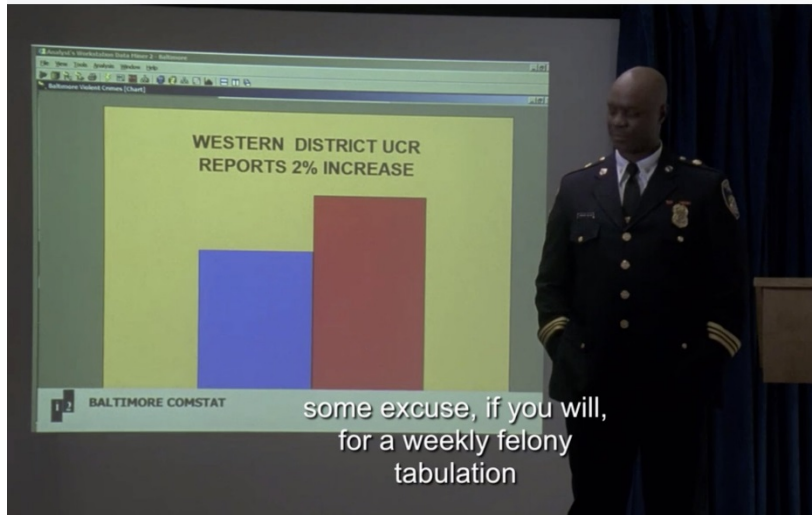
Showing a wider range of PP activities puts these data in context

PP activities were far higher for contraception and STD testing

Services Provided by Planned Parenthood



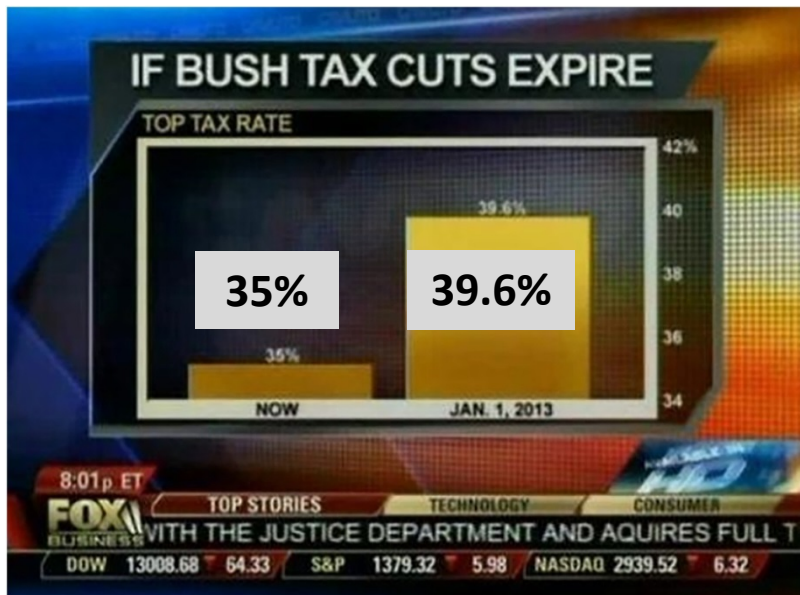
Evil Bars



You can say anything you want if you don't show a scale for the vertical axis

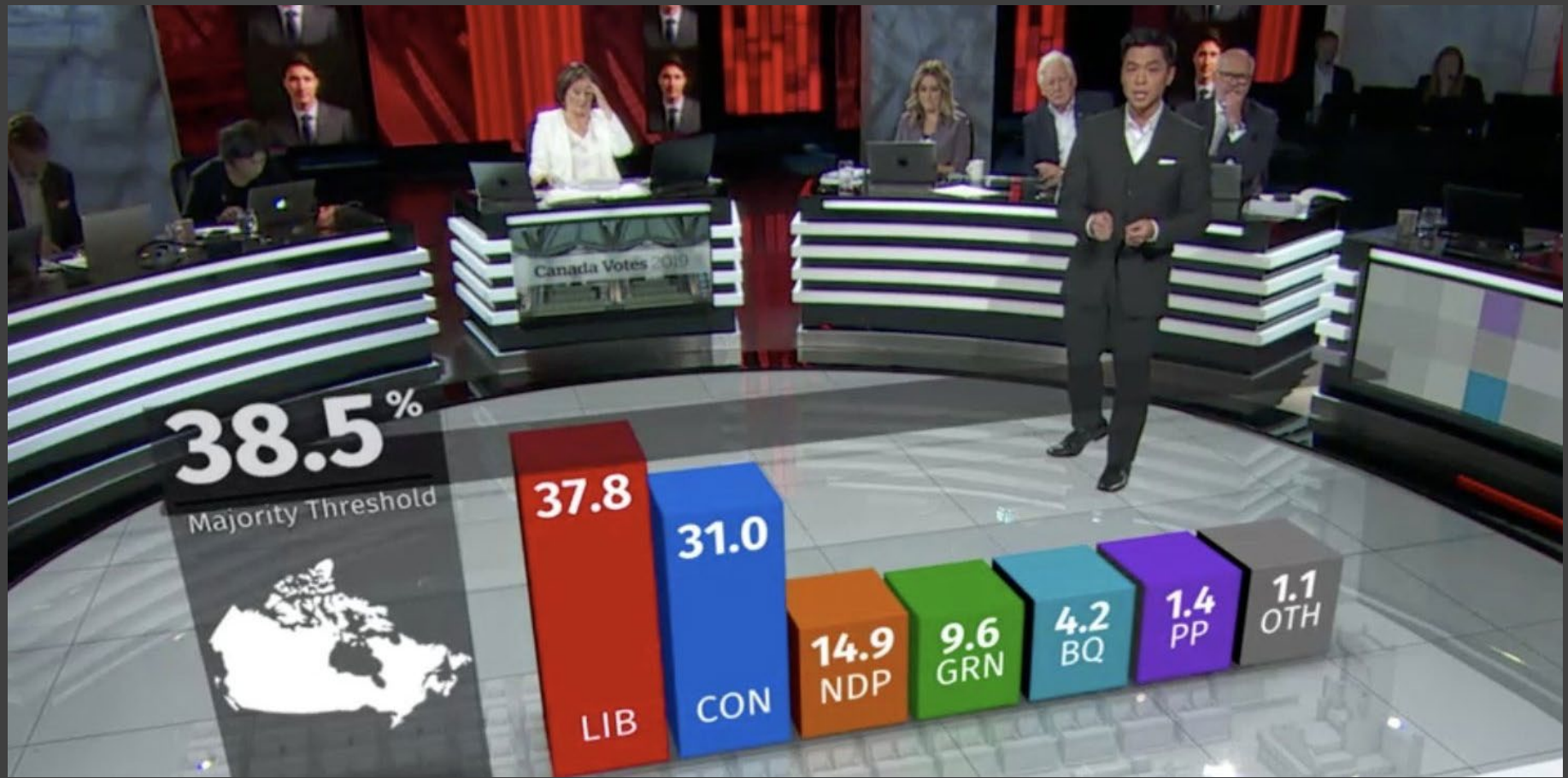
Q: Do people judge the difference in heights or the ratio of heights?

A: It depends on the question



You can greatly distort the perception of difference or ratio by truncating the Y axis.

Y-axis truncation is/was the default in Excel!



More evil: 3D bars

- CBC found it irresistible to make 3D bars to show the 2015 election projection.
- Why do you think the smallest 5 bars are all the same height?

Graphical Excellence: Tables

A study by Abigail Friendly (2017) wanted to show the use of benefits afforded to Toronto developers for their contributions of different types over time

Figure 9: Section 37 benefits by type (1998–2015)

	1998– 2002	2003– 2005	2006– 2009	2010– 2013	2014– 2016	Scale	
Roads, streetscapes	30	35	54	83	15	0 - 10	
Culture, community, recreation	26	50	59	47	16	11 - 20	
Parks	27	41	41	52	20	21 - 30	
Affordable housing	17	26	38	56	11	31 - 40	
Public art	26	25	41	32	4	41 - 50	
Heritage	16	13	26	18	3	51 - 60	
Transit	11	7	10	20	3	61 - 70	
Libraries	6	2	5	11	1	71 - 80	
Other	3	6	7	8	3	81 - 90	

Color background scale from light to dark highlights the largest values

Most frequent benefits appear at the top

Can see overall trends and anomalies

What happened in 2014-2016?

Graphical failure

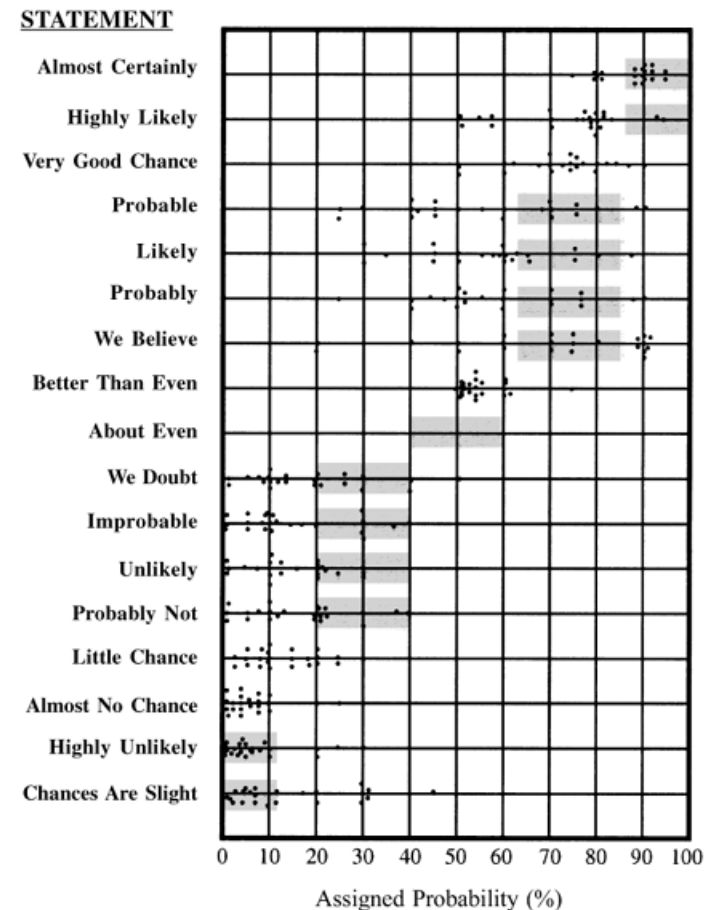
This graph reports the results of a survey by Sherman Kent for the CIA with the question:

What [probability/number] would you assign to the phrase "[phrase]"

The goal was to contribute to an understanding of how intelligence analysts use these terms

Why can this be considered a graphical failure?

Figure 18: Measuring Perceptions of Uncertainty



Graphical excellence

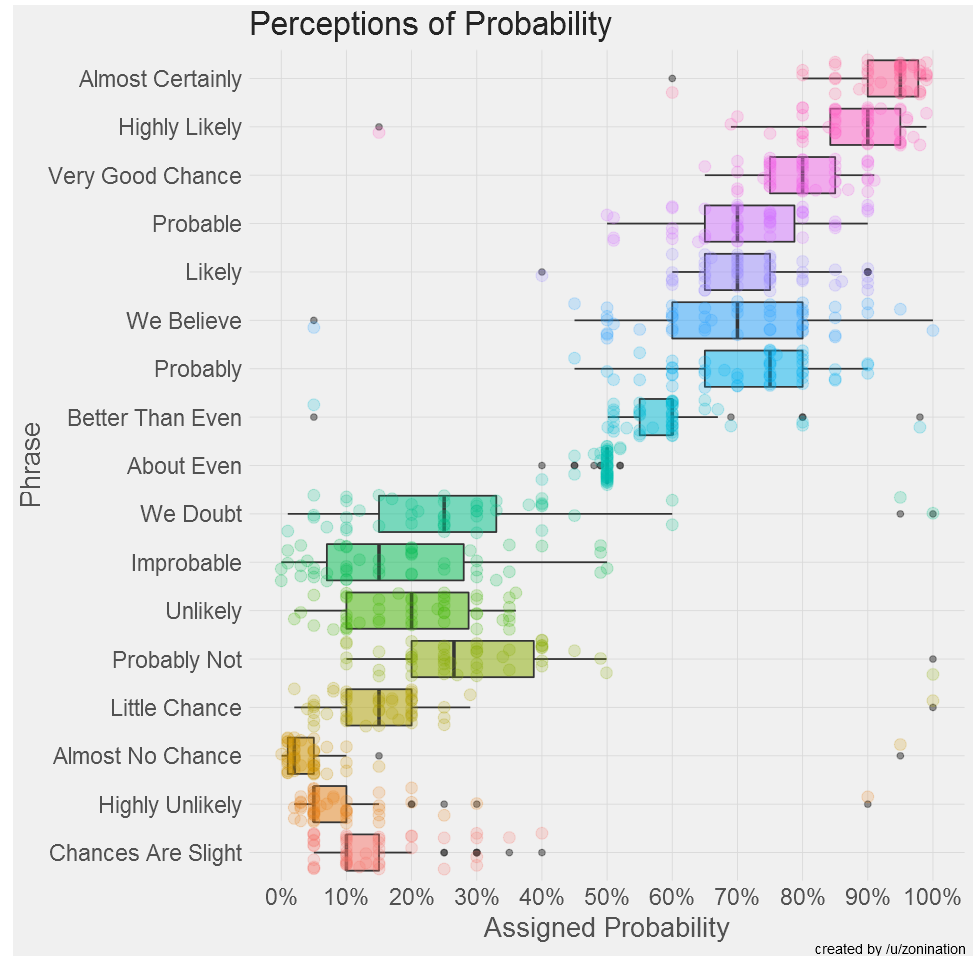
This graph shows the same data, as both dotplots & boxplots

We can see a lot more:

- “about even” has very low variability
- the last 3 categories are listed out of order
- the extreme outliers stand out
- skewness is – for high probability, + for low probability

Technical notes:

- software: ggplot2
- design: faint grid lines
- color: points use transparent color & jittering; outliers also shown in black



Graphical excellence

This graph uses “ridgeline” plots to show the same data

Each one is a small version of a density plot showing a smoothed version of the distribution

Stacking them in this way allows center, variability, shape and other features to be readily compared.

Color & transparency are used effectively

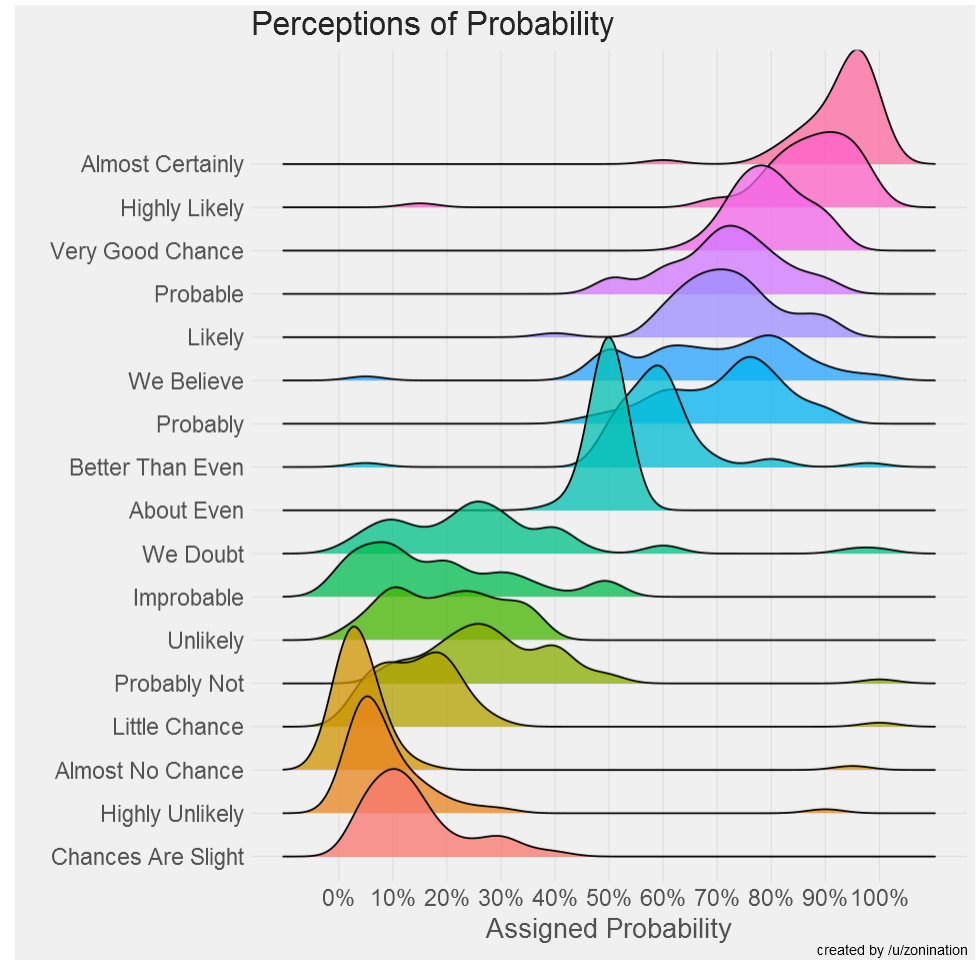


Chart junk or effective info vis?

Charts can be offensive and/or effective

What is the message?
Who is the audience?

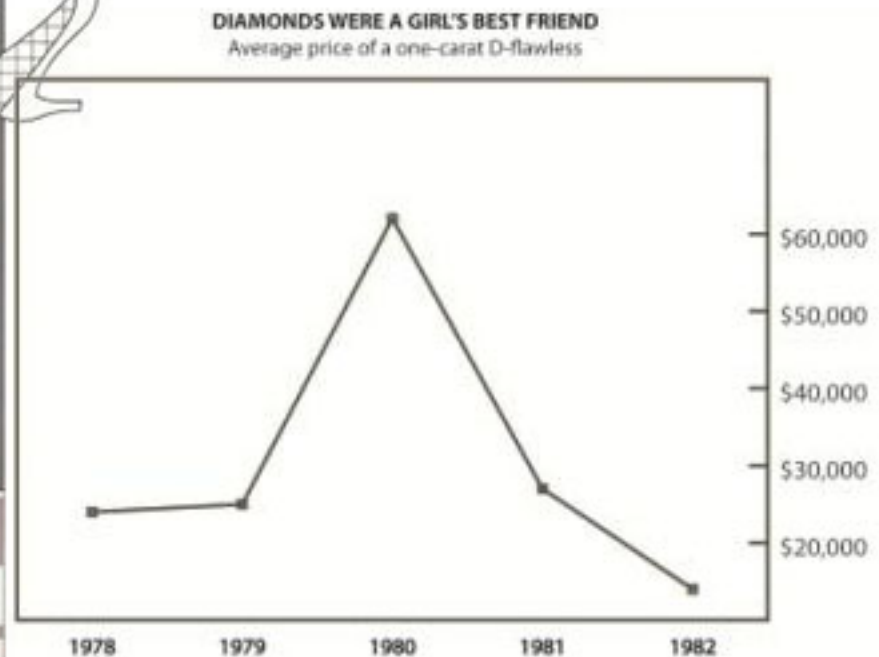
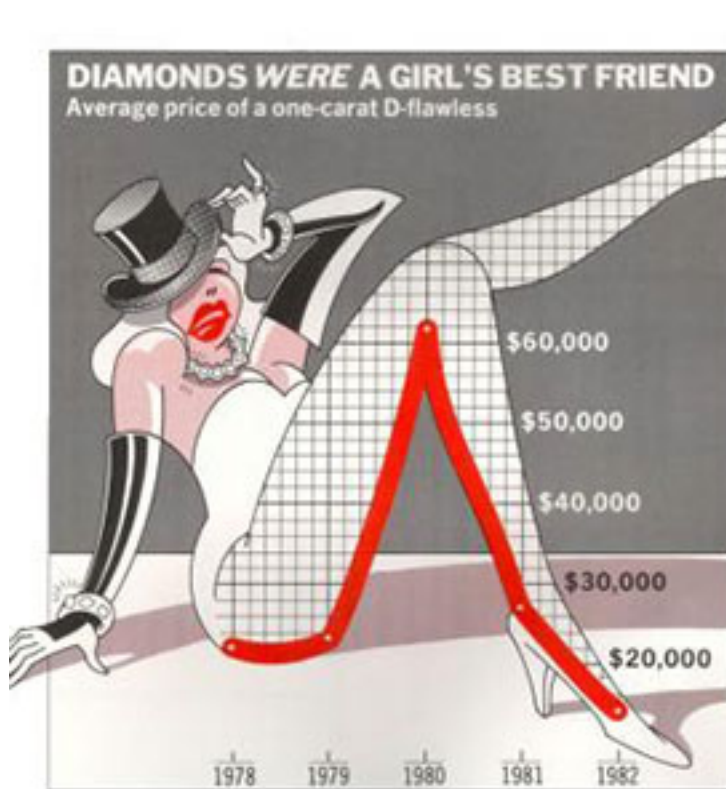


Chart junk or effective info vis?

Suzana Herculano-Houzel has a new method for determining counts of cortical neurons across different species. How to present this effectively?

Goal: compare mammal species brain size and cortical neuron count

Neuron count is shown both as numbers and bars

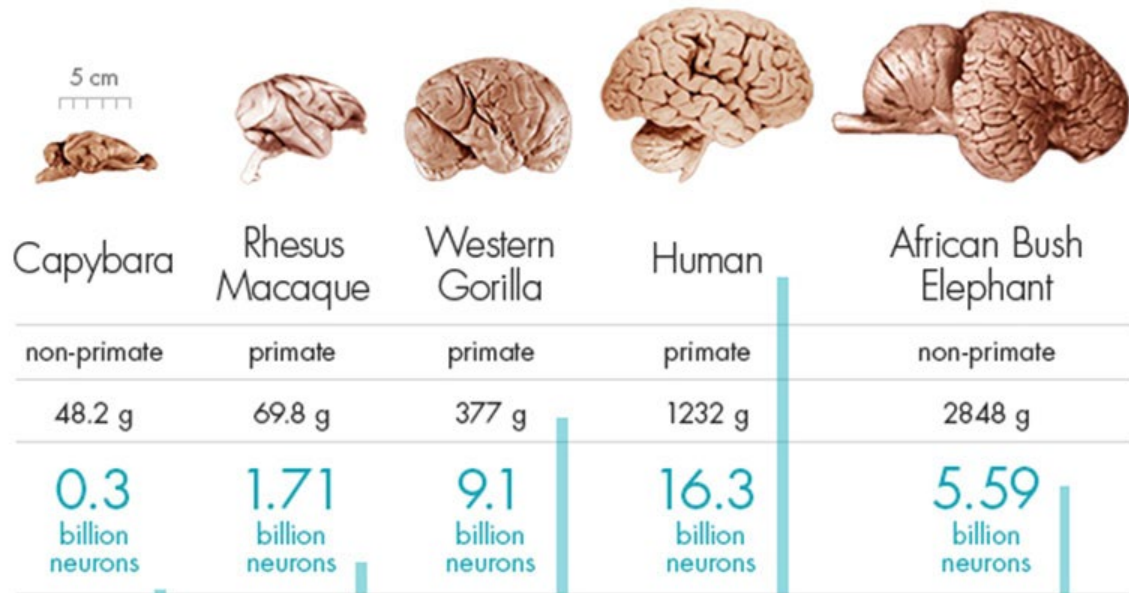
Claim: Human brain is ~linear of primate brains

What do you think?

How could this be made better?

BRAIN SIZE AND NEURON COUNT

Cerebral cortex mass and neuron count for various mammals.



From: Herculano-Houzel, "The human brain in numbers: a linearly scaled-up primate brain "

<https://www.frontiersin.org/articles/10.3389/neuro.09.031.2009/full>

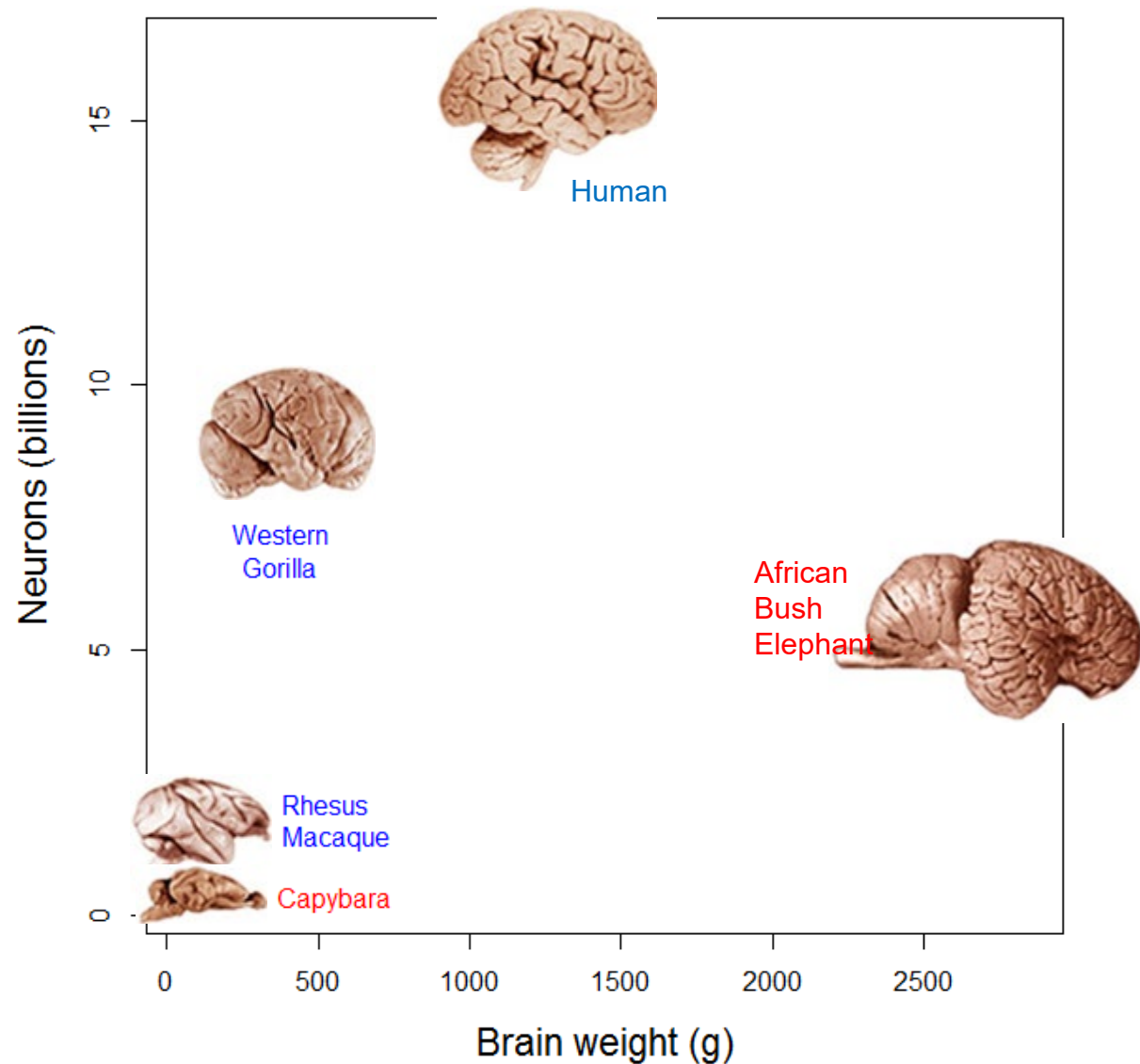
As a scatterplot

A scatterplot makes clear how humans differ from other species

- Using scaled images as point symbols also conveys brain size
- **Primates** are distinguished from **non-primates** by text color

This is arguably a more effective display.

What do you think?

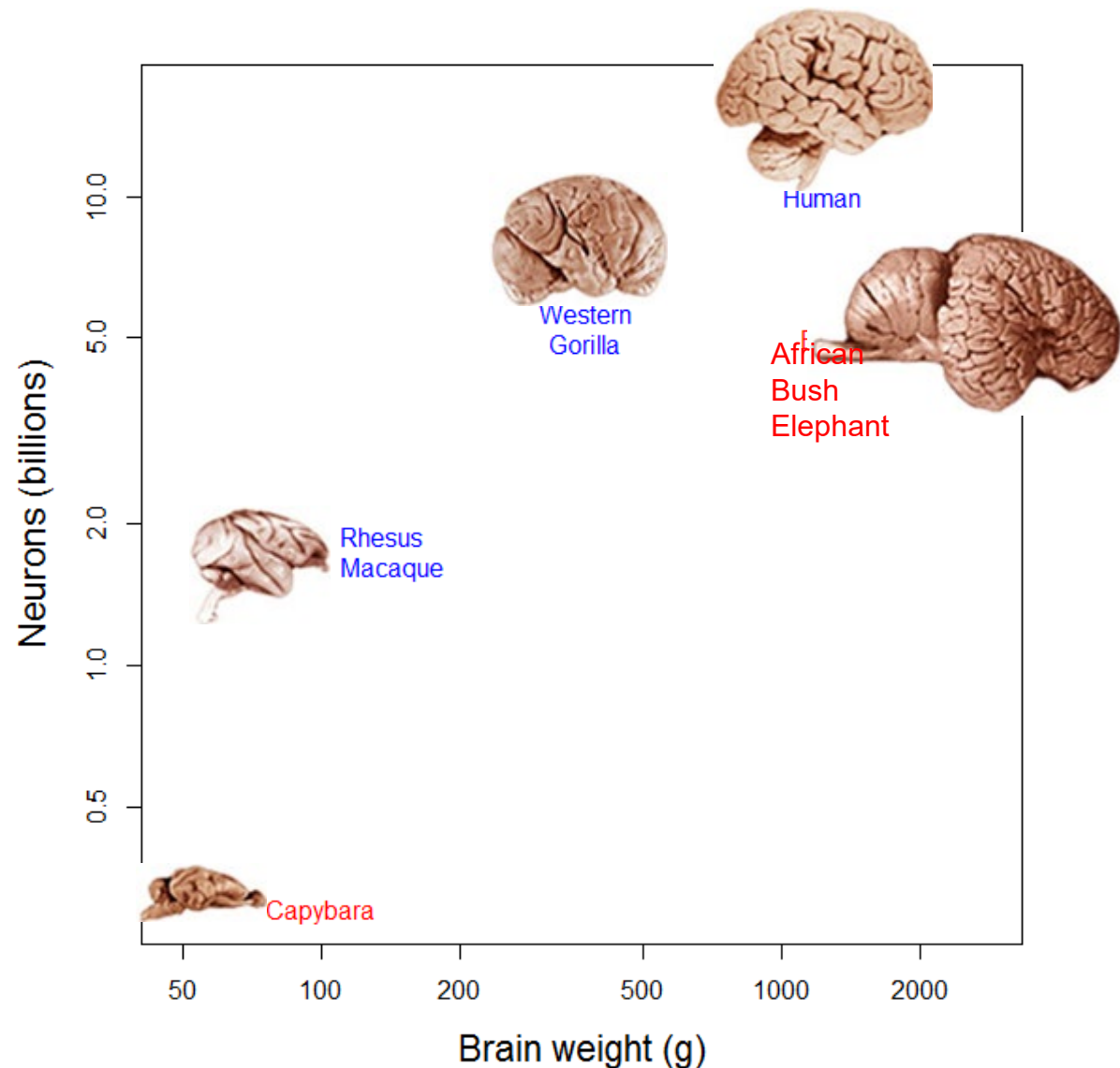


As a scatterplot – log scale

Perhaps even better is to make the plot using log scales for both axes

The relationship is now easier to see, but only approx. linear

The argument for neurons \sim brain weight needs more work.



Why graphs matter: Climate change

In the movie, *An Inconvenient Truth* (2006), Al Gore used the now-famous “hockey stick” graph to show that human activities had greatly increased the degree of global warming over the recent past

The goal was to raise public awareness and call for action to curb environmental effects: CO₂ emissions as the main agent.

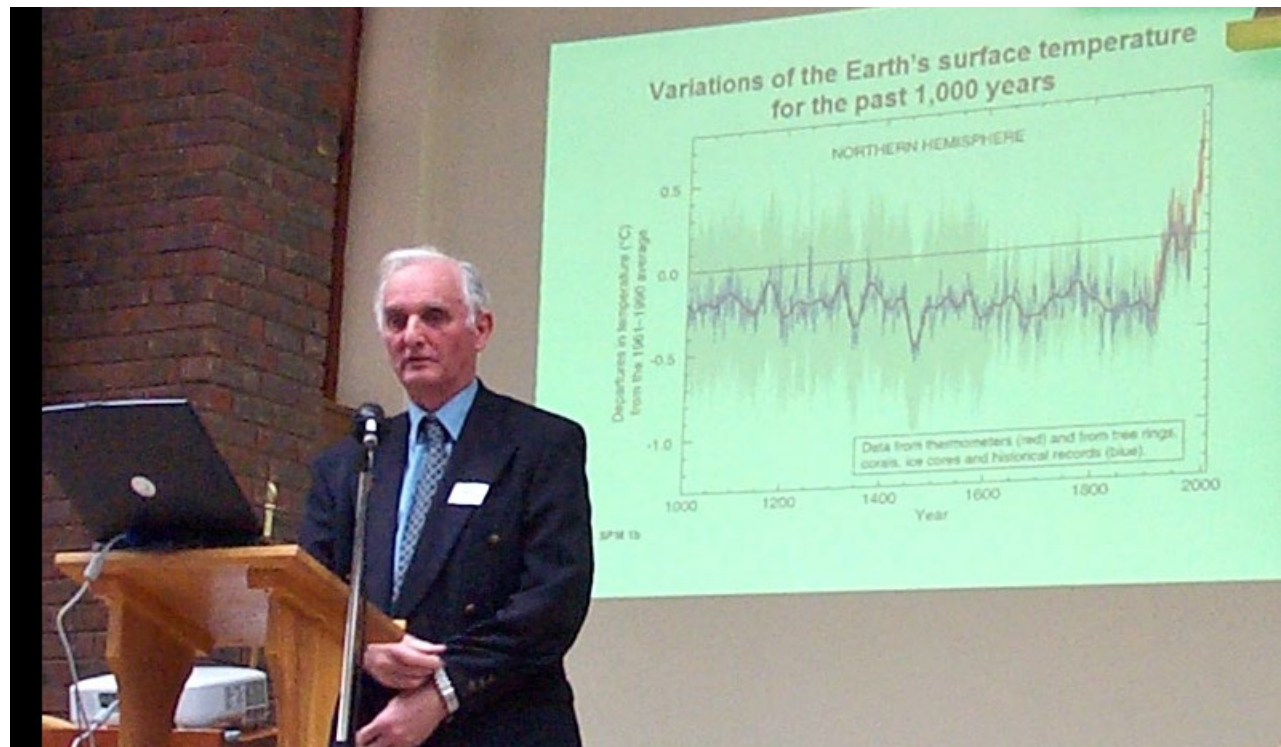


Movie: <https://www.youtube.com/watch?v=8ZUoYGAI5i0>; <http://www.imdb.com/title/tt0497116/>

Climate change: Original graph

Sir John Houghton presents the original Northern Hemisphere hockey stick graph to the [Intergovernmental Panel on Climate Change](#) (IPCC) in 2005.

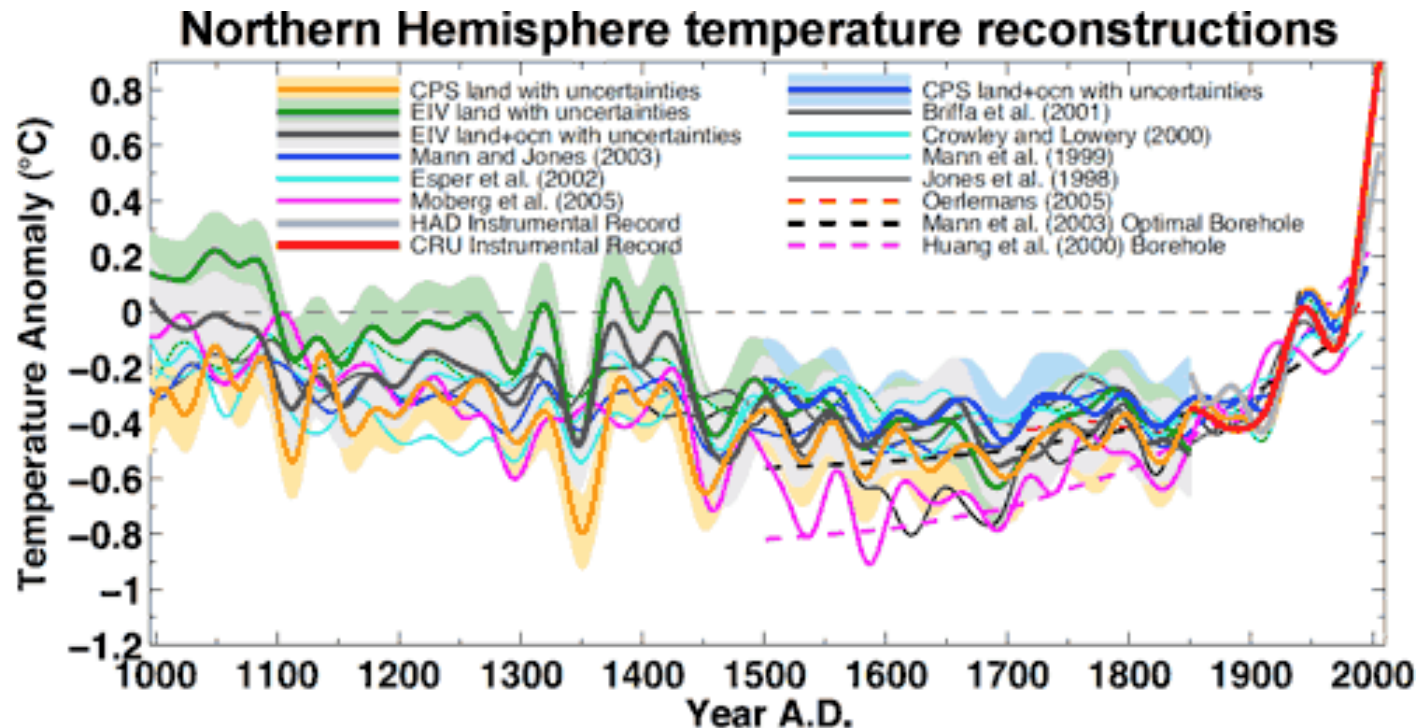
It is based on an analysis by Mann, Bradley & Hughes (1990), with a smoothed curve and uncertainty intervals.



Climate change: data sources

The MBH (1999) paper had used a wide variety of data sources. They were combined using a novel statistical technique, the first eigenvector-based climate field reconstruction (CFR).

Climate scientists understood this; the sceptics did not.



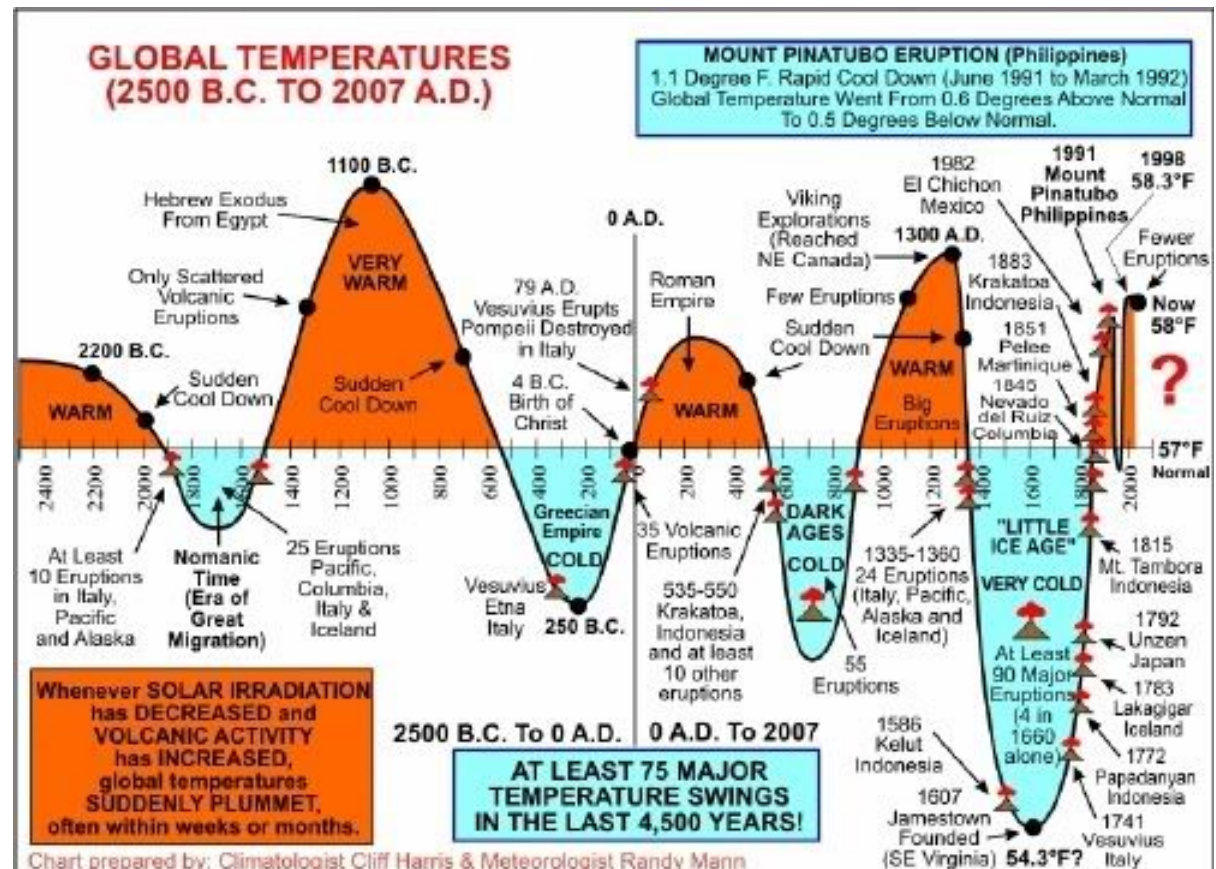
See: https://en.wikipedia.org/wiki/Hockey_stick_controversy for details

Countering climate change

Taking a longer view, and adding a lot of extraneous historical details, climate sceptics were easily able to mount alternative explanations: Solar irradiation & volcanoes

How to mislead:

- Show no temp. scale
- Draw smoothed curve
- Suggest that all is due to “swings” in temp.
- Compress recent history into the end of the time scale

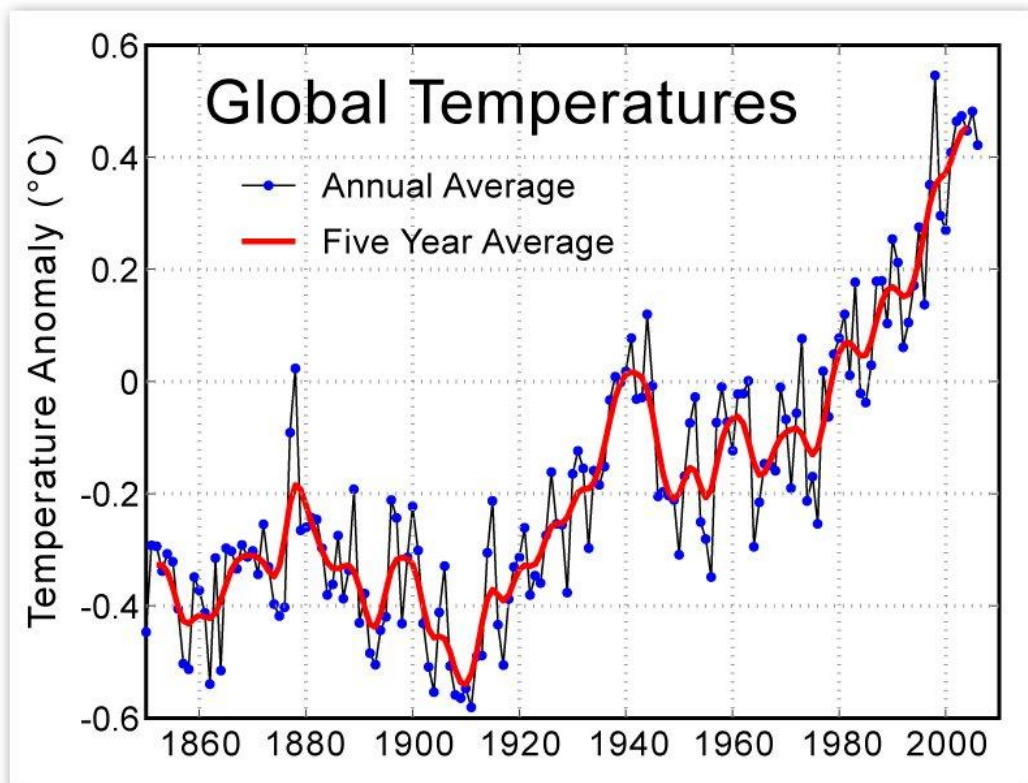


Time scale

Perhaps one fault with the original graphs was trying to show noisy data, from many sources, over too wide a time span.

What could you do to make this graph even more convincing?

- De-emphasize the annual data
- Add an overall smooth curve



Climate change: Infographic

A politically-incorrect graphic shows very clearly the effect of global warming on panty size



Source: <http://www.politically-incorrect-humor.com/2010/03/positive-proof-of-global-warming>

Climate change: other explanations

This infographic attempts to relate global warming to the decrease in pirates

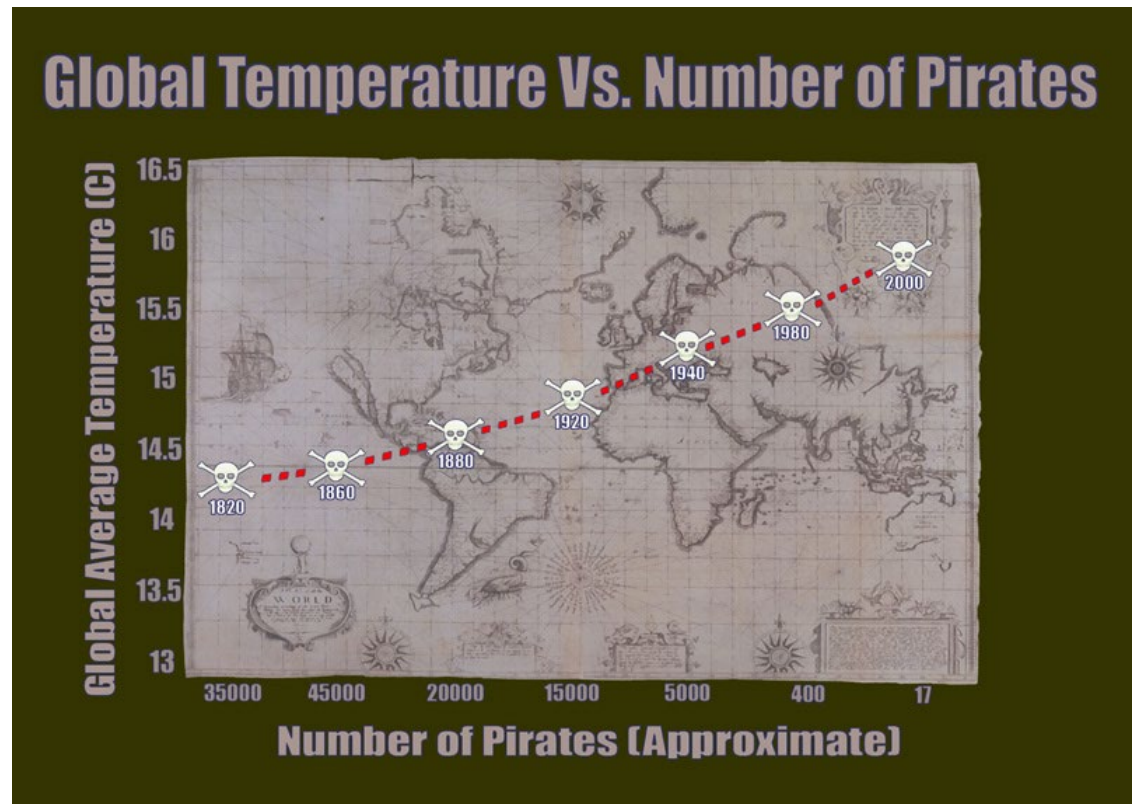
Aside from the substance, how many things are wrong about this graphic?

Simple explanation:

Lack of pirates causes
global warming!

Conclusion:

To stop global warming,
become a pirate!

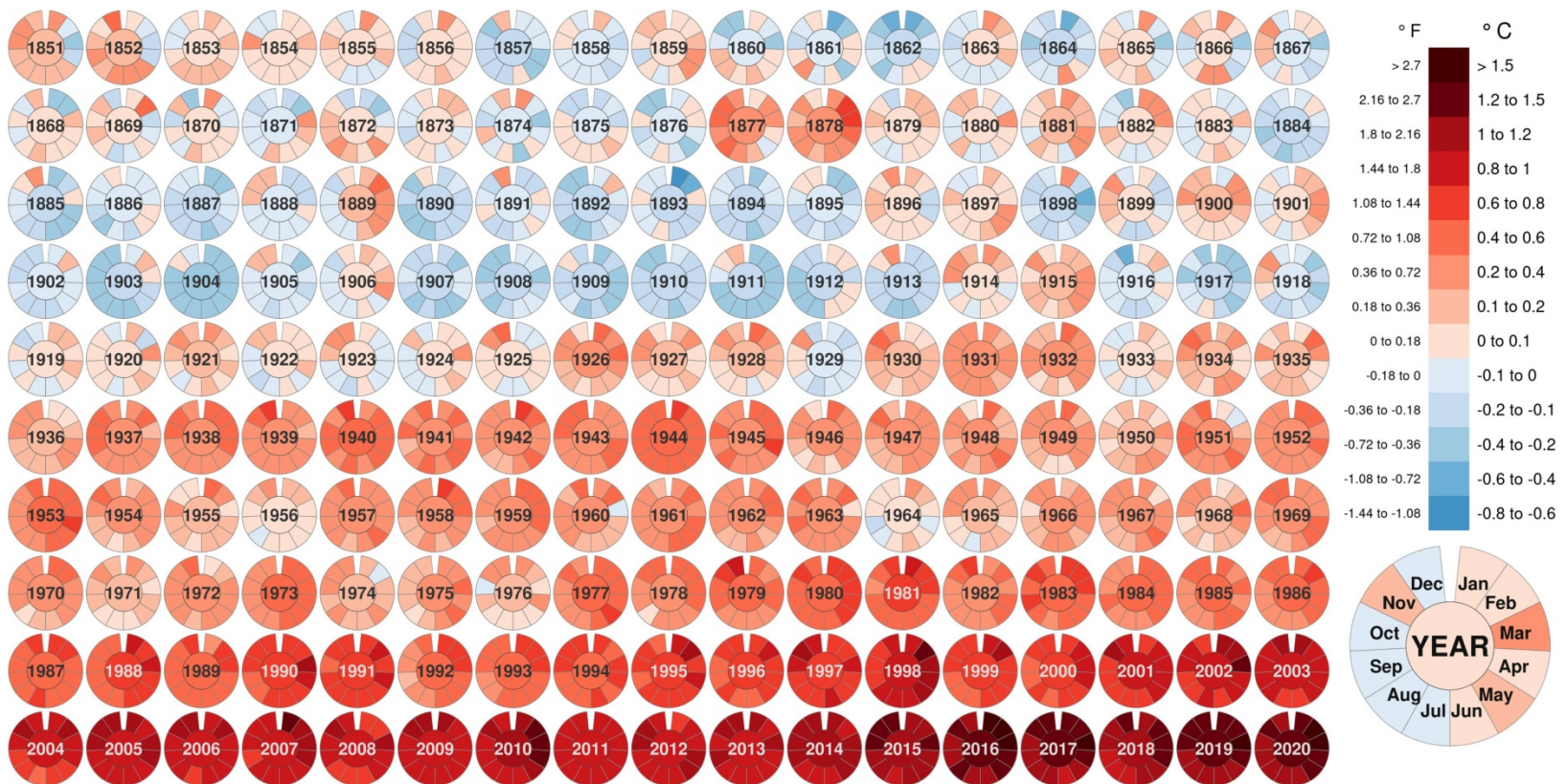


Source: <http://www.forbes.com/sites/erikaandersen/2012/03/23/true-fact-the-lack-of-pirates-is-causing-global-warming>

Circle graphs

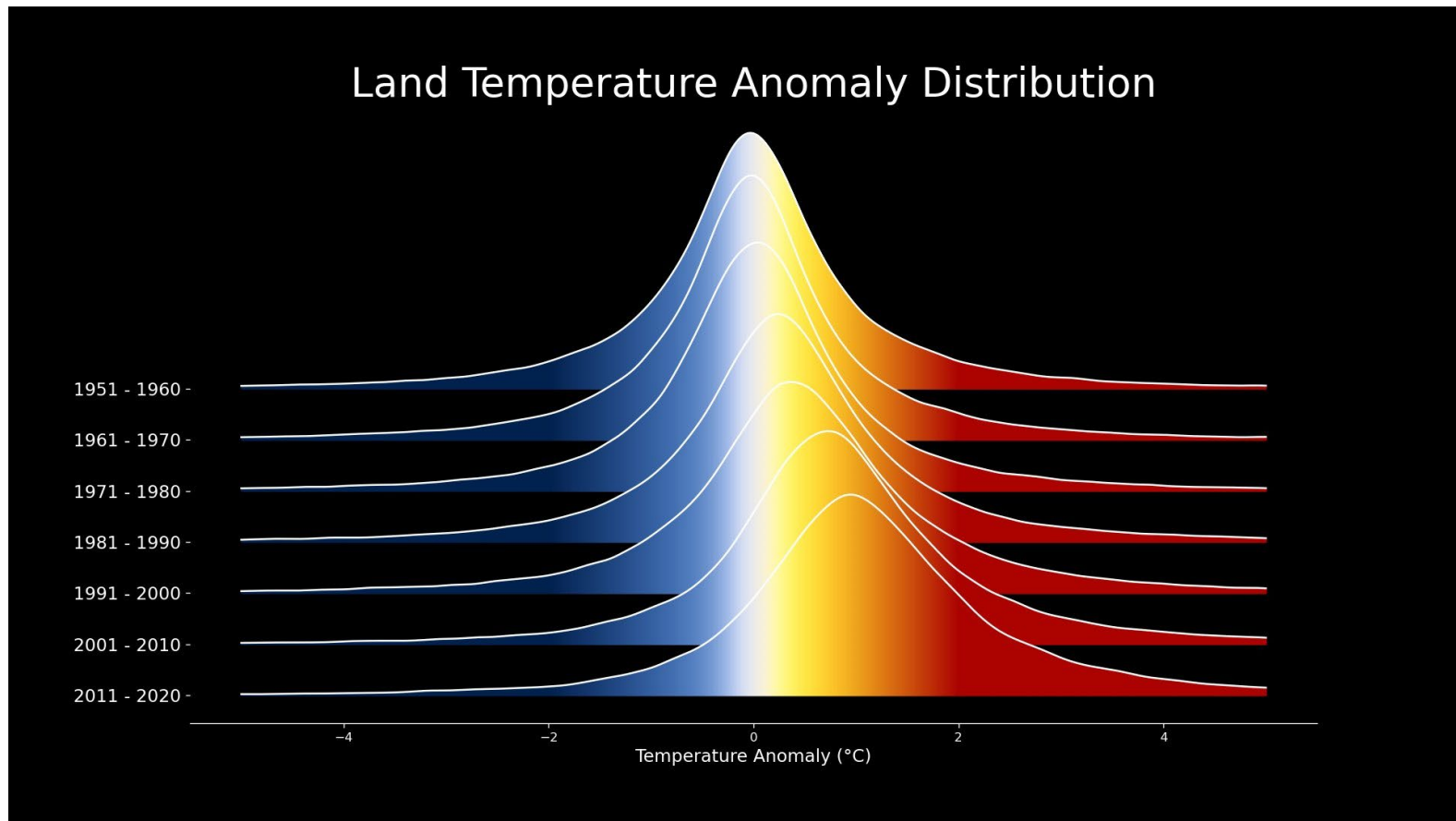
What features makes this graph effective?

Monthly global mean temperature 1851 to 2020 (compared to 1850-1900 averages)



Data: HadCRUT5 - Created by: @neilrkaye

Ridgeline plot



Some graphic rules

- Bars
 - Don't cut off their feet
 - Don't add dynamite fuses
- Pies
 - Generally, best preserved for dessert
 - Well used for part-whole relations with a small number of categories
 - Better used as a graphic form in larger displays
- Axes
 - Avoid double Y-axes
 - Don't truncate without considerable thought
- 3D
 - Avoid for useless "glitz"

Summary

- Graphs as a form of communication
 - Data (numbers), words, images → Stories
- Analysis graphs vs. presentation graphs
- Some principles of effective data display
 - Make the data stand out
 - Facilitate comparisons
 - Effect ordering