# Visualization for Communication

cs109a



the previous day...

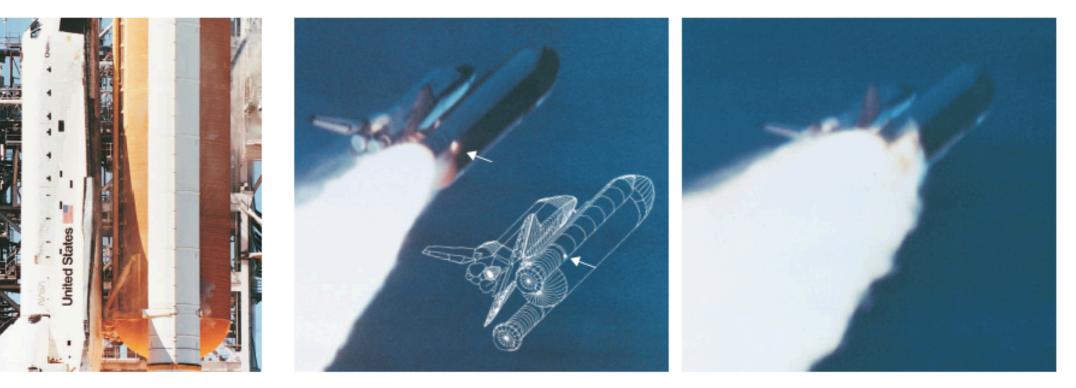
MTI ASSESSMENT OF TEMPERATURE CONCERN ON SRM-25 (51L) LAUNCH CALCULATIONS SHOW THAT SRM-25 O-RINGS WILL BE 20° COLDER THAN SRM-15 O-RINGS 0 0 TEMPERATURE DATA NOT CONCLUSIVE ON PREDICTING PRIMARY O-RING BLOW-BY 0 ENGINEERING ASSESSMENT IS THAT: COLDER O-RINGS WILL HAVE INCREASED EFFECTIVE DUROMETER ("HARDER") 0 "HARDER" O-RINGS WILL TAKE LONGER TO "SEAT" 0 MORE GAS MAY PASS PRIMARY O-RING BEFORE THE PRIMARY SEAL SEATS 0 (RELATIVE TO SRM-15) DEMONSTRATED SEALING THRESHOLD IS 3 TIMES GREATER THAN 0.038" EROSION EXPERIENCED ON SRM-15 0 0 IF THE PRIMARY SEAL DOES NOT SEAT, THE SECONDARY SEAL WILL SEAT 0 PRESSURE WILL GET TO SECONDARY SEAL BEFORE THE METAL PARTS ROTATE O-RING PRESSURE LEAK CHECK PLACES SECONDARY SEAL IN OUTBOARD 0 POSITION WHICH MINIMIZES SEALING TIME 0 MT1 RECOMMENDS STS-51L LAUNCH PROCEED ON 28 JANUARY 1986 SRM-25 WILL NOT BE SIGNIFICANTLY DIFFERENT FROM SRM-15 0 KILMINSTER, VICE PRESIDENT SPACE BOOSTER PROGRAMS

(PCSSCA)

MORTON THIOKOL INC.

Wasatch Division

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Less than 1 second after ignition, a puff of smoke appeared at the aft joint of the right booster, indicating that the O-rings burned through and failed to seal. At this point, all was lost. On the launch pad, the leak lasted only about 2 seconds and then apparently was plugged by putty and insulation as the shuttle rose, flying through rather strong cross-winds. Then 58.788 seconds after ignition, when the Challenger was 6 miles up, a flicker of flame emerged from the leaky joint. Within seconds, the flame grew and engulfed the fuel tank (containing liquid hydrogen and liquid oxygen). That tank ruptured and exploded, destroying the shuttle.



As the shuttle exploded and broke up at approximately 73 seconds after launch, the two booster rockets crisscrossed and continued flying wildly. The right booster, identifiable by its failure plume, is now to the left of its non-defective counterpart.



The flight crew of Challenger 51-L. Front row, left to right: Michael J. Smith, pilot; Francis R. (Dick) Scobee, commander; Ronald E. McNair. Back row: Ellison S. Onizuka, S. Christa McAuliffe, Gregory B. Jarvis, Judith A. Resnik.



171								
30,1981	AFT	SRM No.	( Erosion Depth (in.)	Perimeter Affected (deg)	View Nominal Dia. (in.)	Top Length Of Max Erosion (in.)	Total Heat Affected Length (in.)	Clocking Location (deg)
بر روح م	61A LH Center Field** 61A LH <del>CENTER</del> FIELD** 51C LH Forward Field** 51C RH Center Field (prim)*** 51C RH Center Field (sec)***	22A 22A 15A 15B 15B	None NONE 0.010 0.038 None	None NONE 154.0 130.0 45.0	0.280 0.280 0.280 0.280 0.280 0.280	None NONE 4.25 12.50 None	None NONE 5.25 58.75 29.50	36°66° 338°-18° 163 354 354 354
	41D RH Forward Field 41C LH Aft Field* 418 LH Forward Field	13B 11A 10A	0.028 None 0.040	110.0 None 217.0	0.280 0.280 0.280	3.00 None 3.00	None None 14.50	275 351
7-12	STS-2 RH Aft Field	2B	0.053	116.0	0.280			90

\*Hot gas path detected in putty. Indication of heat on O-ring, but no damage. \*\*Soot behind primary O-ring. \*\*\*Soot behind primary O-ring, heat affected secondary O-ring.

Clocking location of leak check port - 0 deg.

Engineer deck, the previous day...

OTHER SRM-15 FIELD JOINTS HAD NO BLOWHOLES IN PUTTY AND NO SOOT NEAR OR BEYOND THE PRIMARY O-RING.

SRM-22 FORWARD FIELD JOINT HAD PUTTY PATH TO PRIMARY O-RING, BUT NO O-RING EROSION AND NO SOOT BLOWBY, OTHER SRM-22 FIELD JOINTS HAD NO BLOWHOLES IN PUTTY.

BLOW BY HISTORY SRM-15 WORST BLOW-BY	HISTORY OF O-RING TEMPERATURES (DEGREES-F)					
· 2 CASE JOINTS (80°), (110°) ARC	MOTOR	MBT	AMB	O-RING	WIND	
O MUCH WORSE VISUALLY THAN SRM-22	Dm-+	68	36	47	10 mPH	
	Dm-2	76	45	52	10 mpH	
SRM 12 BLOW-BY	QM - 3	72.5	40	48	IO MPH	
· 2 CASE JOINTS (30-40°)	Qm - 4	76	48	51	10 m PH	
	SRM-15	52	64	53	10 mPH	
SRM-13 A, 15, 16A, 18, 23A 24A	5RM-22	77	78	75	IO MPH	
O NOZZLE BLOW-BY	5 RM - 25	55	26	29 27	10 трн 25 трн	

(PCSSCA)

#### CONCLUSIONS :

 TEMPERATURE OF O-RING IS NOT ONLY PARAMETER CONTROLLING BLOW-BY

SRM 15 WITH BLOW- BY HAD AN O-RING TEMP AT 53 F SEM 22 WITH BLOW-BY HAD AN O-RING TEMP AT 75 F FOUR DEVELOPMENT MOTORS WITH NO BLOW-BY WERE TESTED AT O-RING TEMP OF 47 TO 52 F

DEVELOPMENT MOTORS HAD PUTTY PACKING WHICH RESULTED IN BETTER PERFORMANCE

- AT ABOUT 50°F BLOW- BY COULD BE EXPERIENCED IN CASE JOINTS
- O TEMP FOR SRM 25 ON 1-28-86 LAUNCH WILL BE 29°F 9AM 38°F 2PM
- O HAVE NO DATA THAT WOULD INDICATE SRM 25 IS DIFFERENT THAN SRM 15 OTHER THAN TEMP

#### RECOMMENDATIONS :

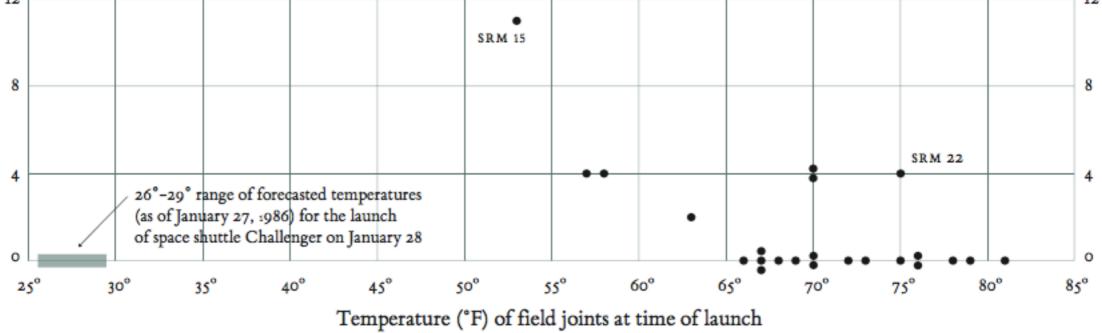
· O-RING TEMP MUST BE ≥ 53 °F AT LAUNCH

DEVELOPMENT MOTORS AT 47° To 52°F WITH PUTTY PACKING HAD NO BLOW-BY SRM 15 (THE BEST SIMULATION) WORKED AT 53°F

· PROJECT AMBIENT CONDITIONS (TEMP & WIND) TO DETERMINE LAUNCH TIME

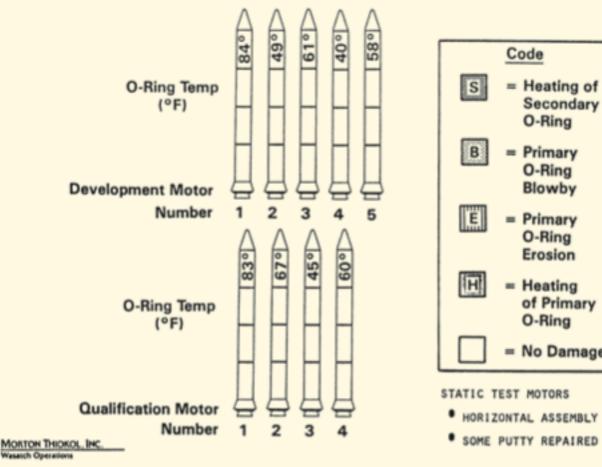
#### (PCSSCA)

Flight	Date	°F	Erosion incidents	Blow-by incidents	Damage index	Comments
51-C	01.24.85	53°	3	2	11	Most erosion any flight; blow-by; back-up rings heated.
41-B	02.03.84	57°	1		4	Deep, extensive erosion.
61-C	01.12.86	58°	1		4	O-ring erosion on launch two weeks before Challenger.
41-C	04.06.84	63°	1		2	O-rings showed signs of heating, but no damage.
1	04.12.81	66°			0	Coolest (66°) launch without O-ring problems.
6	04.04.83	67°			0	
51-A	11.08.84	67°			0	
51-D	04.12.85	67°			0	
5	11.11.82	68°			0	
3	03.22.82	69°			0	
2	11.12.81	70°	1		4	Extent of erosion not fully known.
9	11.28.83	70°			0	
41-D	08.30.84	70°	1		4	
51-G	06.17.85	70°			0	
7	06.18.83	72°			0	
8	08.30.83	73°			0	
51-B	04.29.85	75°			0	
61-A	10.30.85	75°		2	4	No erosion. Soot found behind two primary O-rings.
51-I	08.27.85	76°			0	
61-B	11.26.85	76°			0	
41-G	10.05.84	78°			0	
51-J	10.03.85	79°			0	
	06.27.82	80°			?	O-ring condition unknown; rocket casing lost at sea.
51-F	07.29.85	8				O-ring condition unknown; rocket casing lost at sea. RISK ASSESSMENT?
		O-ri	1g damag	e		
		inde	, each lau	unch		
			.,			



(VST, Tufte)

#### **History of O-Ring Damage in Field Joints**

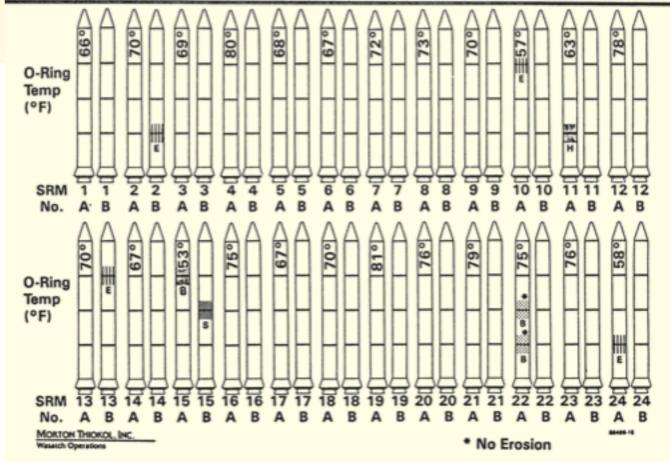


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Secondary of Primary = No Damage

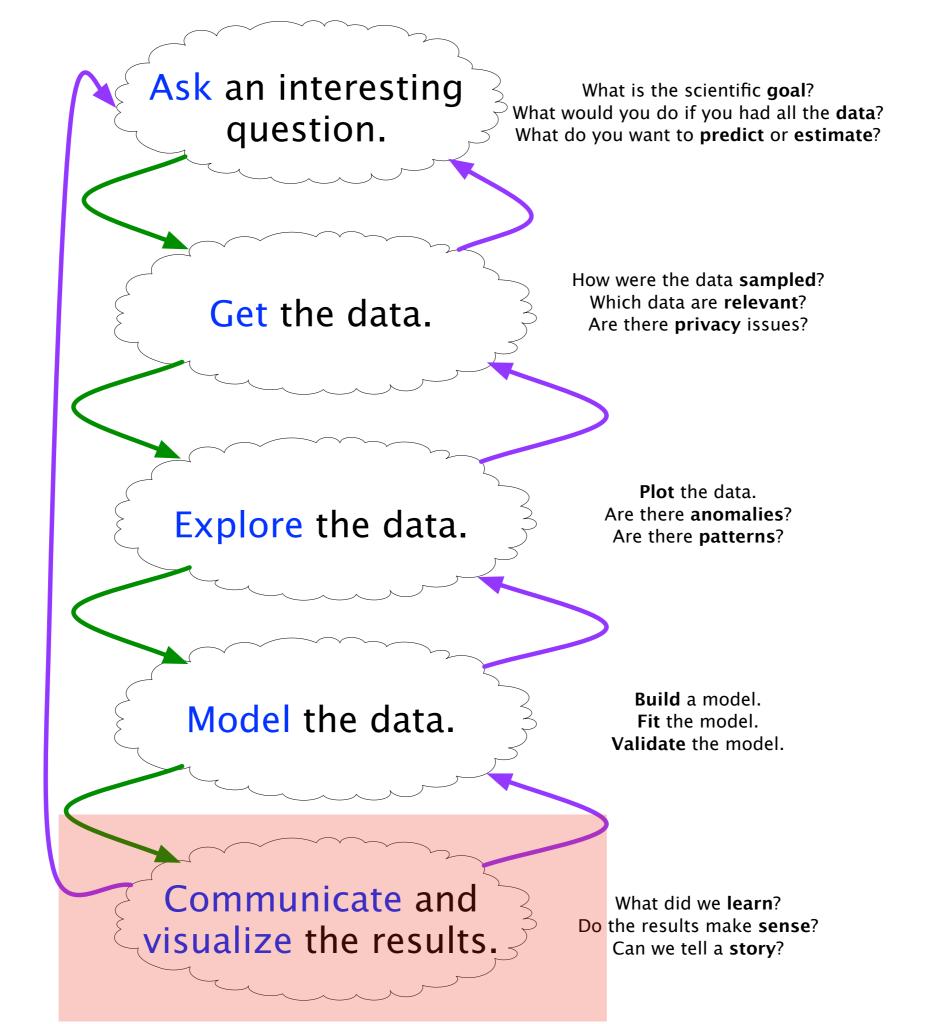
#### Chartjunk at hearings

#### History of O-Ring Damage in Field Joints (Cont)



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(PCSSCA)



#### Visualization Goals

#### **Communicate (Explanatory)**

Present data and ideas

Explain and inform

Provide evidence and support

Influence and persuade

#### Analyze (Exploratory)

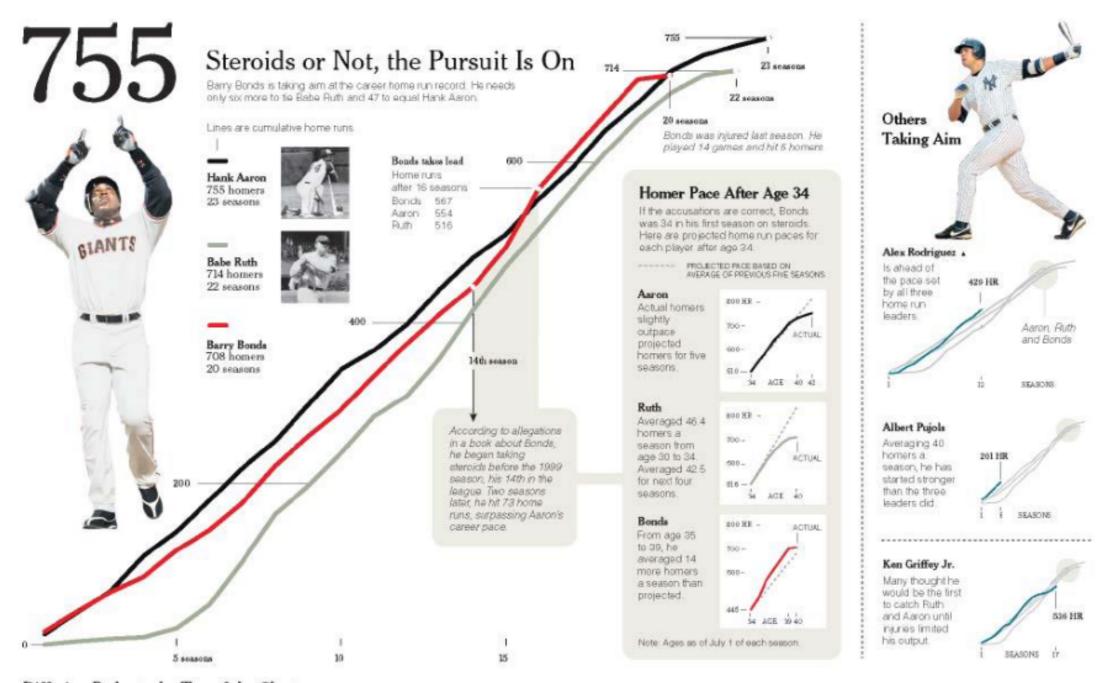
Explore the data

Assess a situation

Determine how to proceed

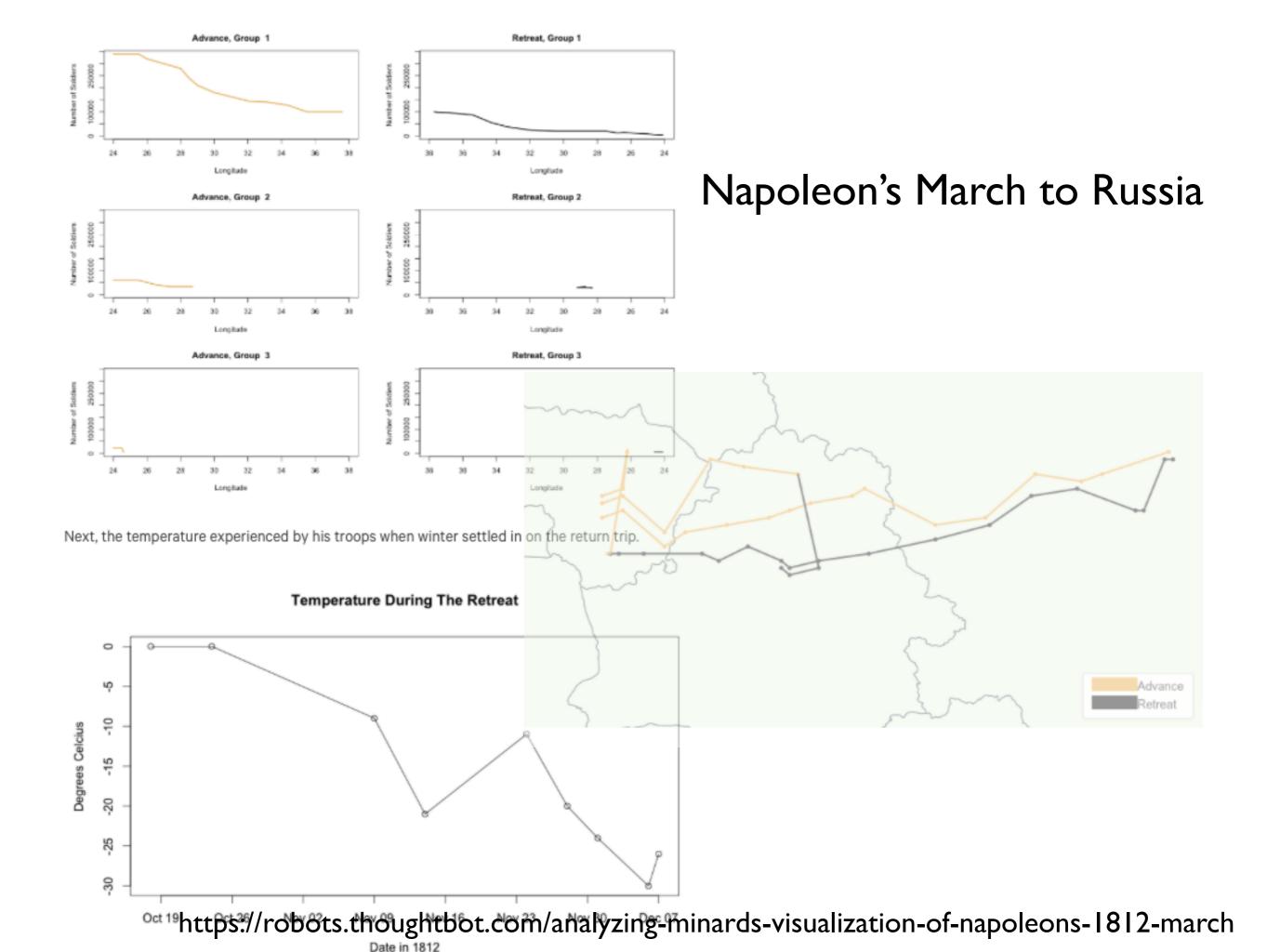
Decide what to do

### Communicate

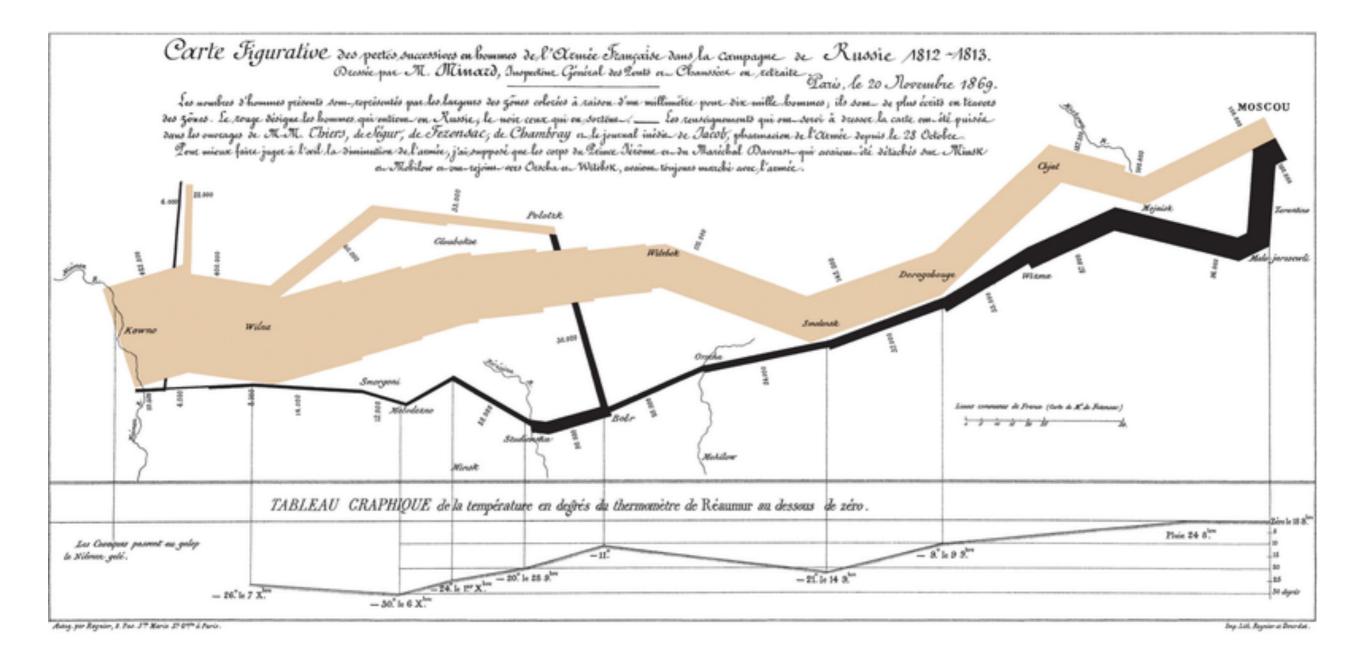


Differing Paths to the Top of the Charts The top seven players on the career home run list, along with a look at Griffey (12th), Rodriguez (37th) and Pujols (tied 257th)

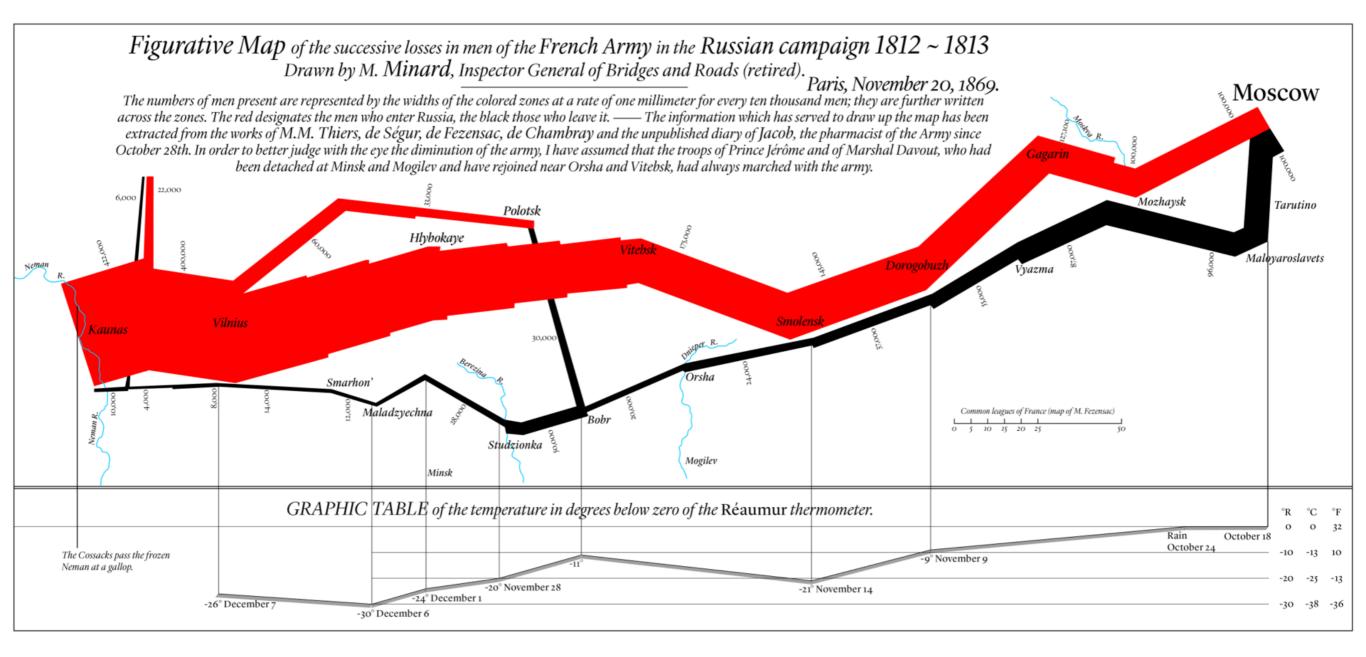




#### Minard's Graphic on Napoleon's Russia Campaign



#### Minard's Graphic on Napoleon's Russia Campaign



(from wikipedia)

# Key Considerations

- Who is your **audience**?
- What **questions** are you answering?
- Why should the audience **care**?
- What are your major **insights** and surprises?
- What **change** to you want to affect?

### Effective Visualizations

- I. Have graphical integrity
- 2. Keep it simple
- 3. Use the right display
- 4. Use color strategically
- 5. Know your audience

#### WRONG

## CITIZENS<sup>Fo</sup>r TRUMP



THE INSIDE STORY OF THE PEOPLE'S MOVEMENT TO TAKE BACK AMERICA

\* \* \* \* \*

**JACK POSOBIEC** 

#### RIGHT

## COUNTIES <sup>Fo</sup>r TRUMP



THE INSIDE STORY of 46% of voters' movement to take back america

\* \* \* \* \*

**JACK POSOBIEC** 

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#### The Persuasive Power of Data Visualization

Anshul Vikram Pandey New York University

Anjali Manivannan New York University

Oded Nov New York University

Margaret L. Satterthwaite NYU School of Law, satterth@exchange.law.nyu.edu

Enrico Bertini New York University After looking into common effects in attitude formation and change we searched for specific mentions to the graphical appearance of charts as a driver for persuasion. Some of the comments we collected seem to back up the findings we found in our results. Some participants explicitly mention the charts as being the main reason for their change: "I already knew that increased incarceration didn't lower crime, but I wasn't sure of the statistics. To see it on the graphs is really eye opening."; "I was influenced by the bar graph showing the reasons why the survey respondents played video games."; "I would not know exact numbers on this issue - the graphs gave a visual and helped identify the numbers"; "Seeing the graphs conflicted with my previous opinion, so I feel like I need to reevaluate my stance in a way."

It is also important to mention that the graphical appearance of charts is not the only factor that has a strong impact on people's attitude. In our collected feedback, we found numerous references to statistics and numbers, suggesting that <u>mere exposure to data does</u> have a persuasive effect – maybe at least partially due to the increased

sense of objectivity evidence supported by numbers carries. We found comments like: "It was concrete data that seemed compelling.; "Seeing numbers is a good indicator of change rather than just reading what someone has to say"; "It showed a large amount of different sources, which made it more credible". More research is needed to disentangle what kind of specific effects each of these components have on persuasion.

http://lsr.nellco.org/cgi/viewcontent.cgi?article=1476&context=nyu\_plltwp

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## Keep it Simple

## Don't Make Them Think!

- Your audience does not want to spend cognitive effort on things you know and can just show them
- Lead them through the major steps of your story
- Point out interesting key facts and insights using captions and annotations



# Don't Bury the Lead

How satisfied have you been with each of these features?

Have not used Not satisfied at all Not very satisfied Somewhat satisfied Very satisfied Completely satisfied

Feature A	11%	40%		£.t	47%	}
Feature B	13%	36%	Summer of		47%	
Feature C	5%	24%	34%		33	%
Feature D	4%	21%	37	%	2	9%: +
Feature E	6%	23%	-1.3	6%		8%
Feature F		<b>5%</b> 20%	11	35%	5 c	25%
Feature G		<b>2%5%</b> 15%	20	3%	33	<b>6</b> 1
Feature H		6 <mark>%</mark> 23%		32%		25%
Feature I		5% 17	7%	27%	. ) - A	27%
Feature J	8% 1	4% 24%	6	27%	1	25%
Feature K		4%	17%	28%	<b>1</b>	21%
Feature L		4%	23%		27%	16%
Feature M		3%	8%	25%	18%	13%
Feature N		9%	14%	24%	17%	10%
Feature O				<b>6%</b> 15%	6 16%	11%

# Don't Bury the Lead

#### User satisfaction varies greatly by feature

#### Product X User Satisfaction: Features

Feature K Feature L Feature M Feature N Feature O

 \* Completely satisfied
 \* Very satisfied
 \* Not very satisfied
 \* Not satisfied at all
 \* Have not used

 Feature A
 0%

 Feature B
 0%

 Feature C
 0%

 Feature D
 0%

 Feature E
 0%

 Feature F
 0%

 Feature G
 1%

 Feature H
 1%

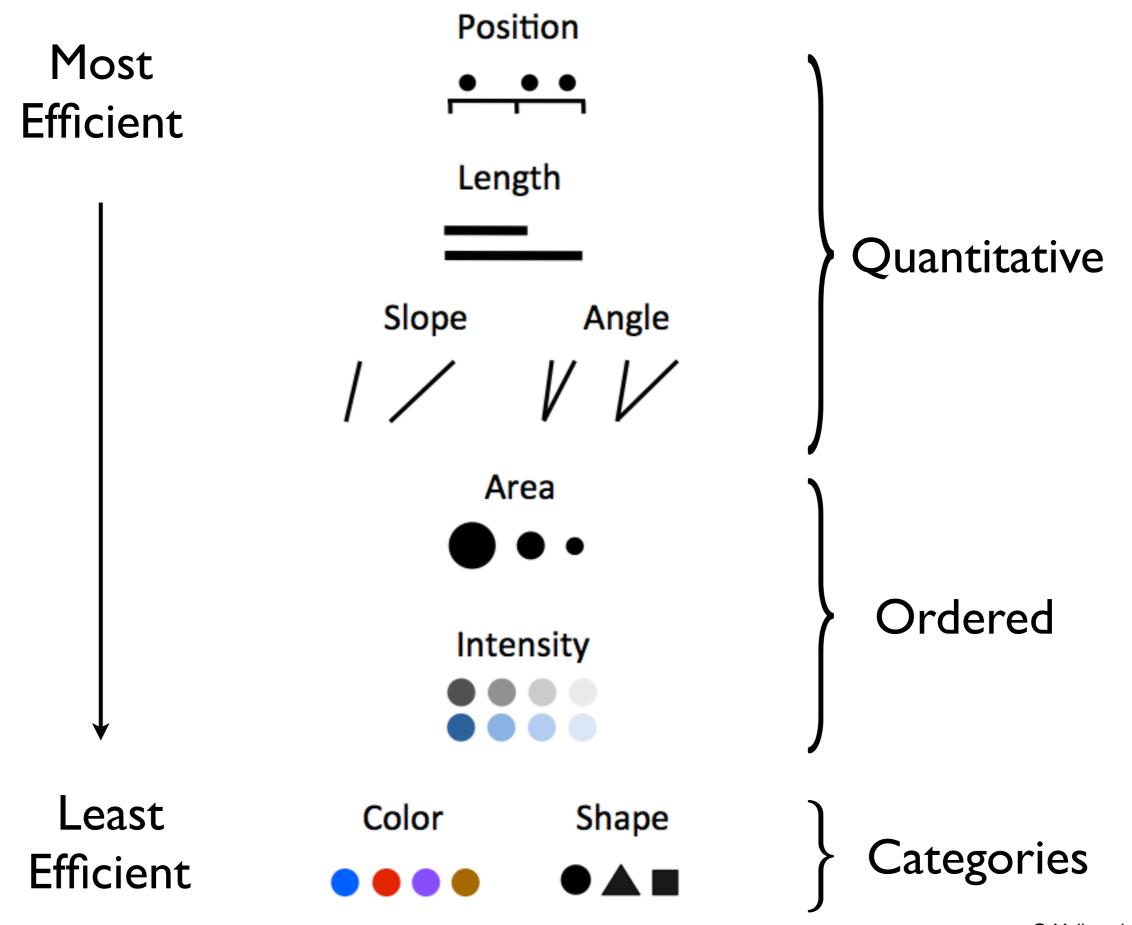
 Feature I
 2%

 Feature J
 2%

Feature O is least-used feature; what steps can we proactively take with existing users to increase use?

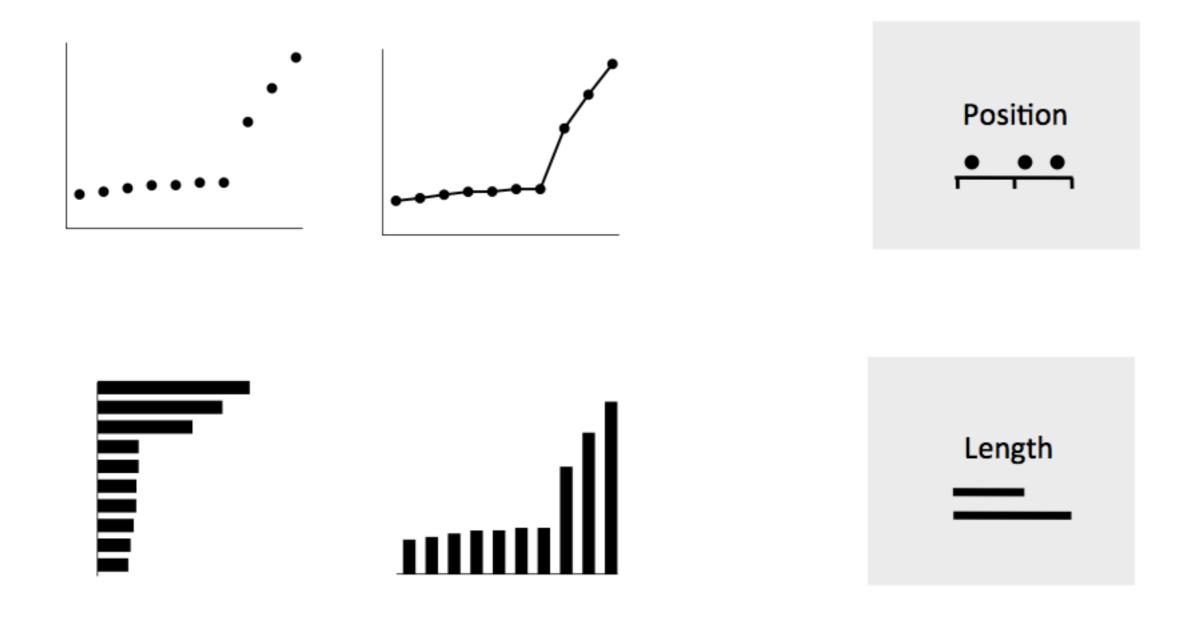
Cole Nussbaumer

## Use the right display



C. Mulbrandon VisualizingEconomics.com

#### Most Effective

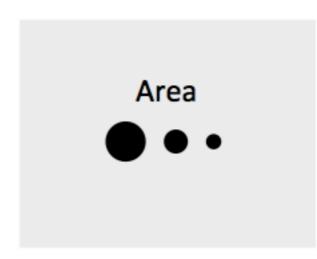


VisualizingEconomics.com

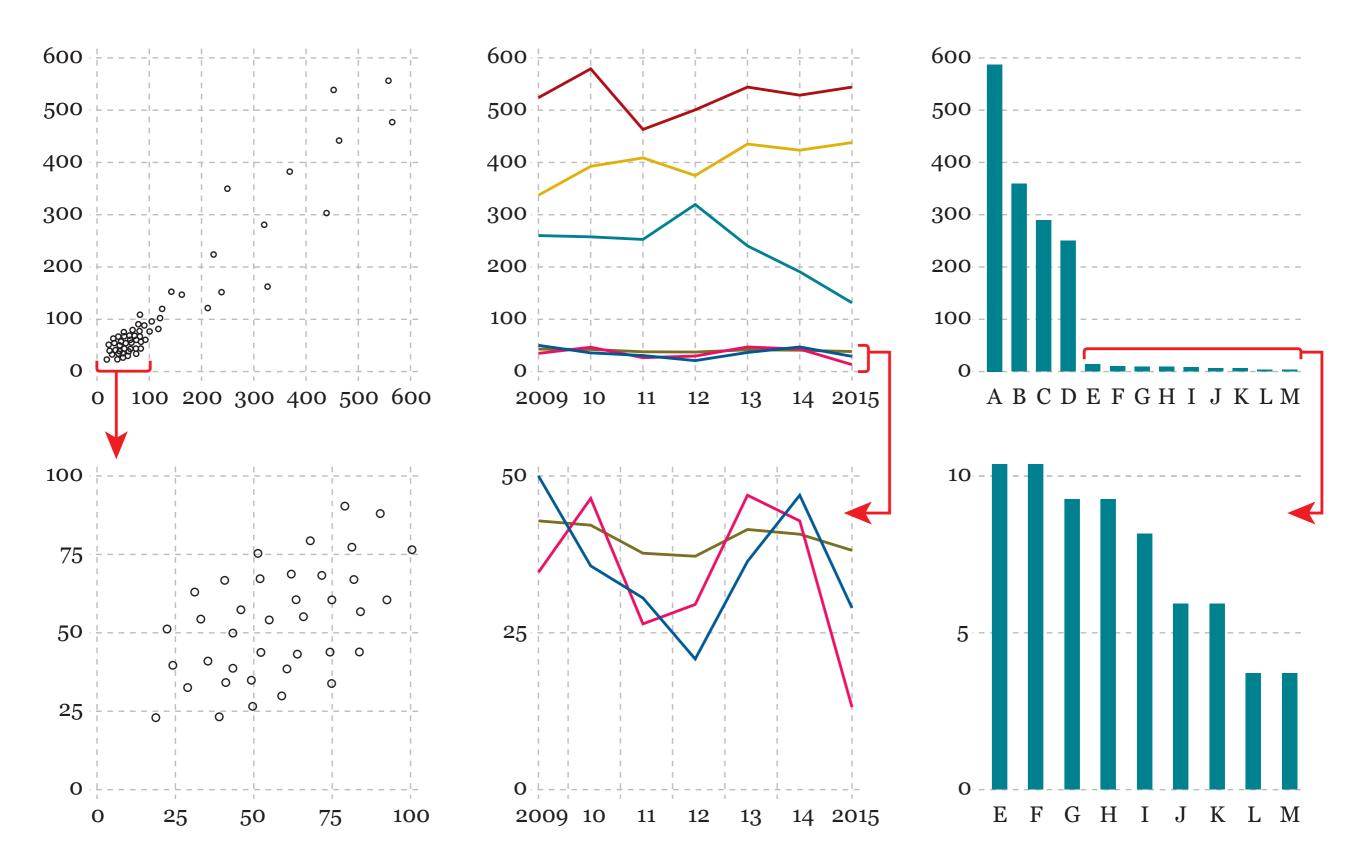
#### Less Effective







VisualizingEconomics.com



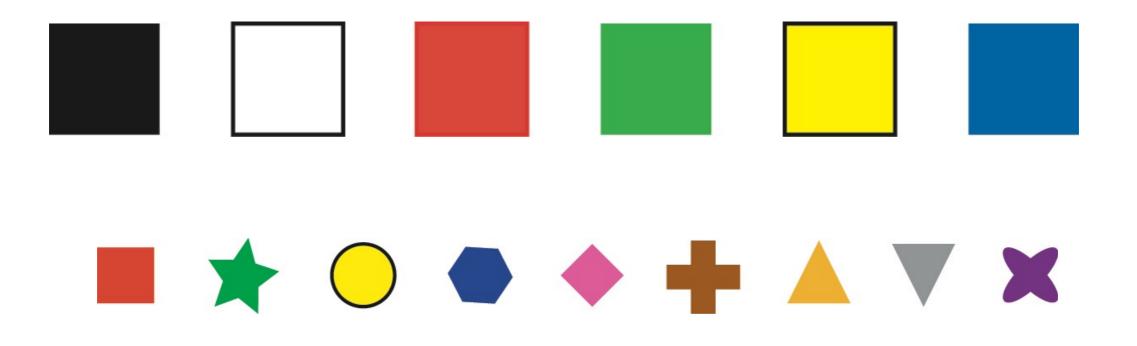
Possible solution to cases when you have data that diverge a lot

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# Use color strategically

## Colors for Categories

#### Do not use more than 5-8 colors at once



Ware, "Information Visualization"

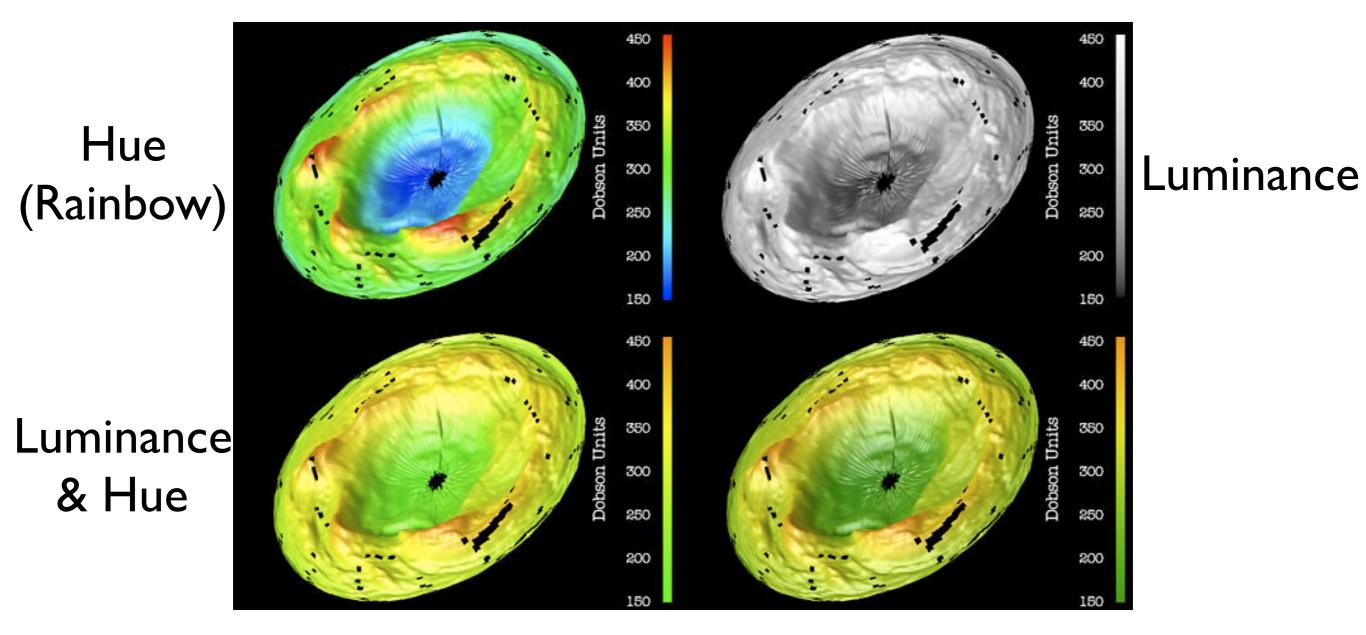
### Colors for Ordinal Data

#### Vary luminance and saturation



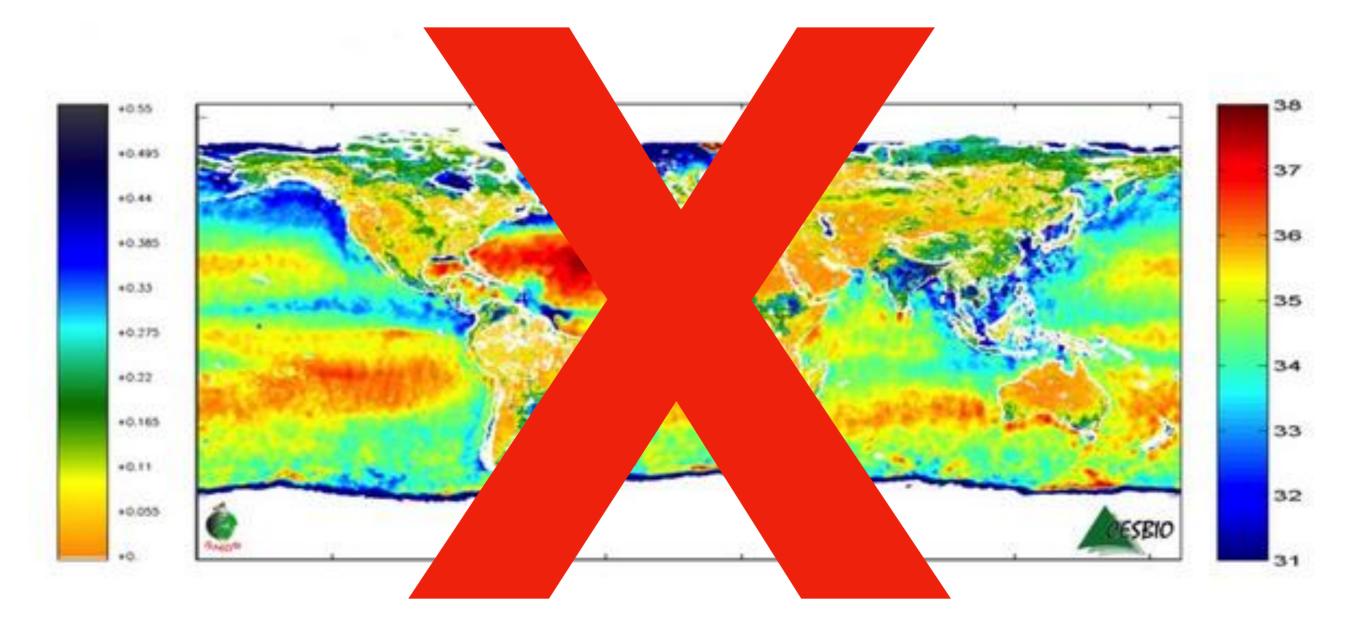
Zeilis et al, 2009, "Escaping RGBland: Selecting Colors for Statistical Graphics"

### Colors for Quantitative Data



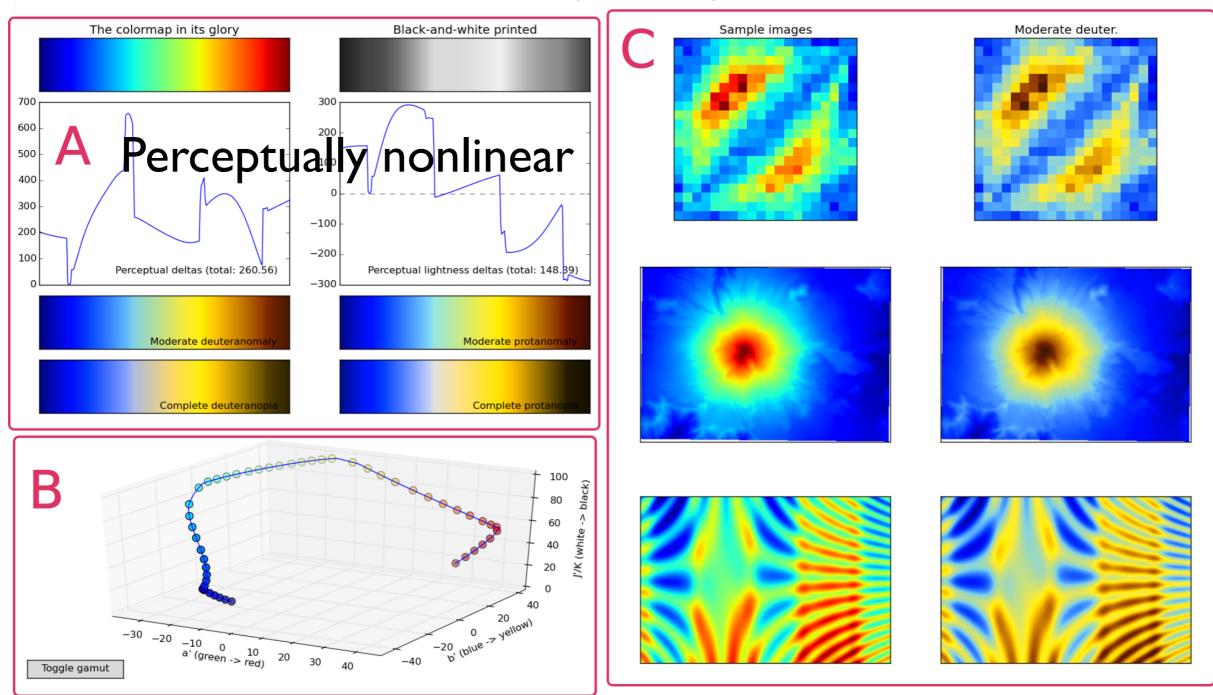
Rogowitz and Treinish, Why should engineers and scientists be worried about color?

### Rainbow Colormap



## Rainbow Colormap

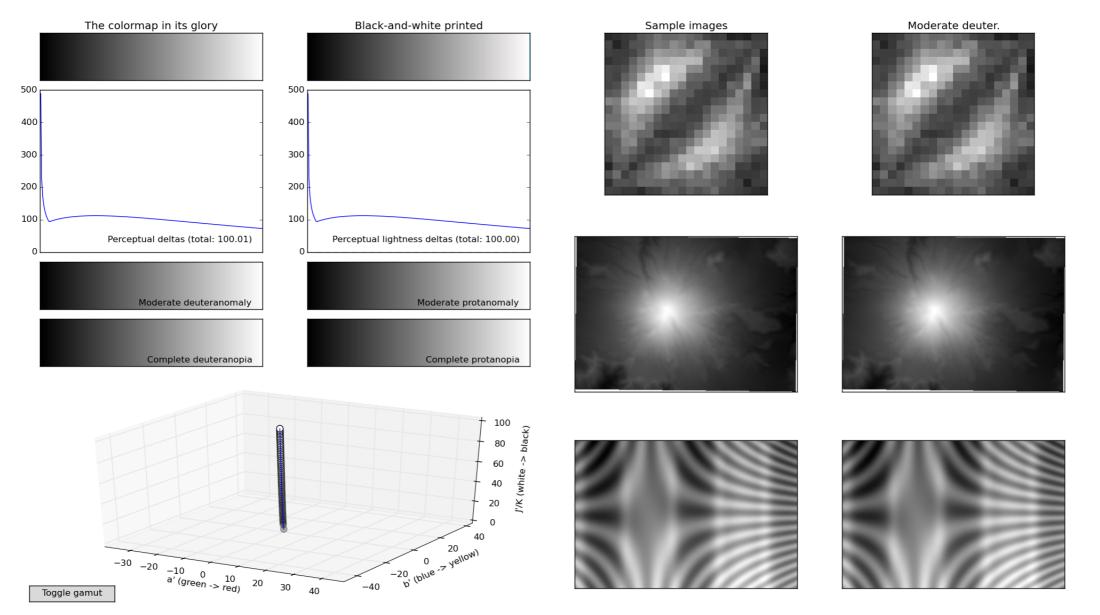
Colormap evaluation: jet



R. Simmon

Gray

#### Colormap evaluation: gray



### Color Blindness

<image/>	<image/> <section-header></section-header>		<image/> <section-header></section-header>
		green encies	Blue / Yellow deficiency

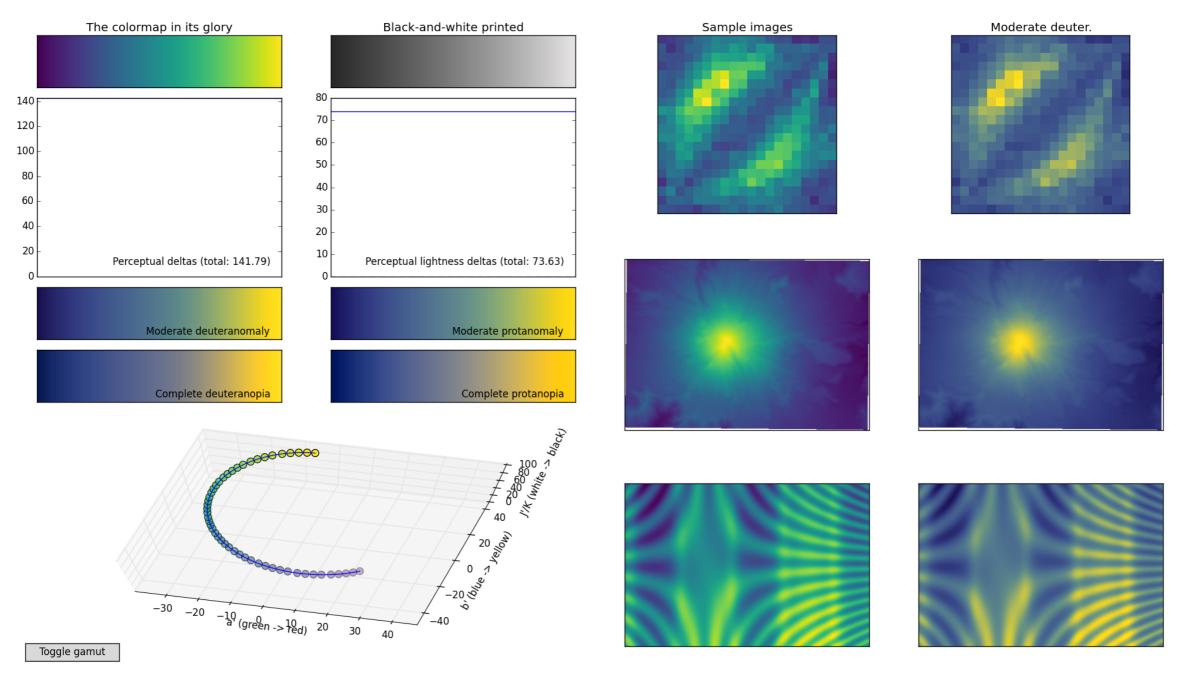
Based on slide from Stone

## Color Blindness

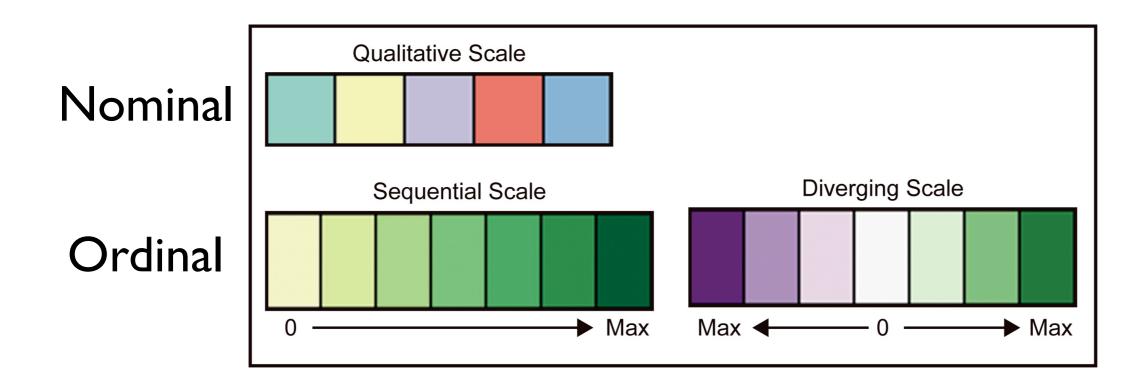
Normal	Protanope	Deuteranope	Lightness

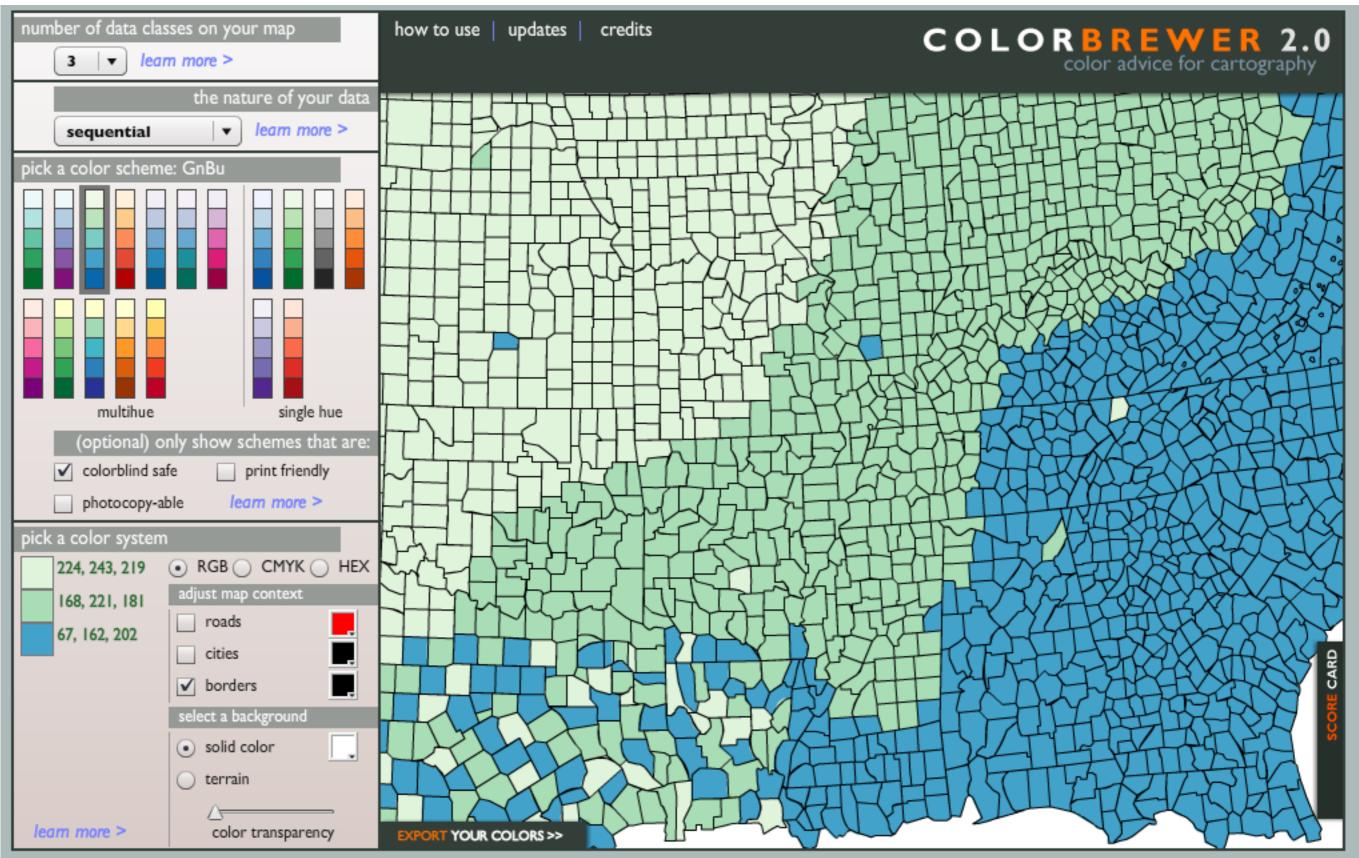
# Viridis

### Colormap evaluation: option\_d.py



# Color Brewer





© Cynthia Brewer, Mark Harrower and The Pennsylvania State University Support Back to ColorBrewer 1.0



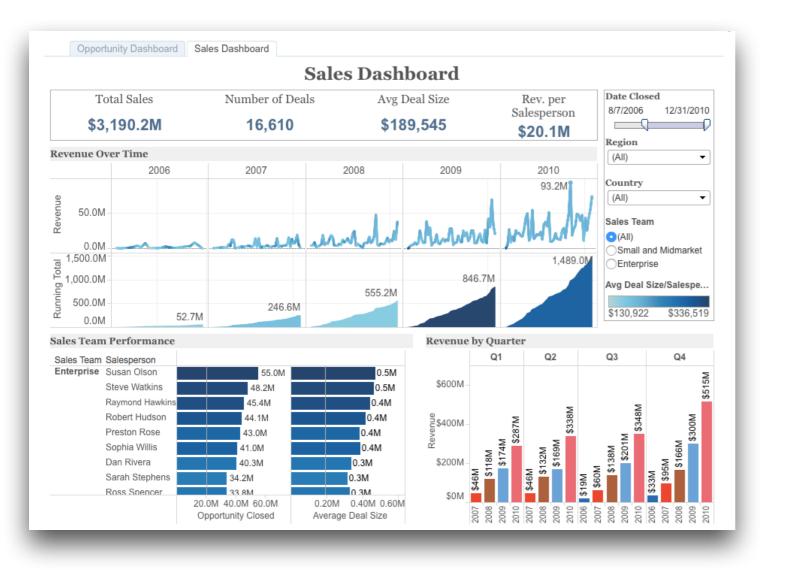
# Know your audience

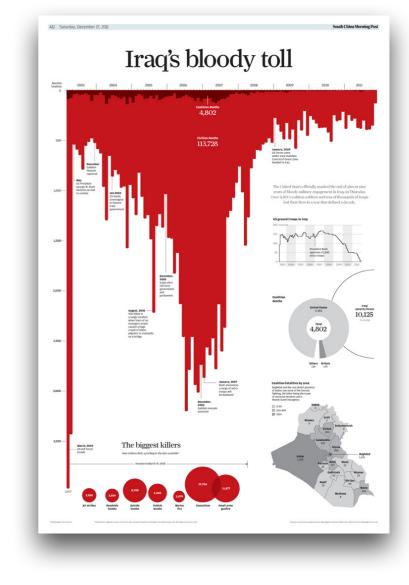
- What do they know?
- What motivates them? What do they desire?
- What experiences do you share? What are common goals?
- What insights can you give them? What tools and "magical gifts"?

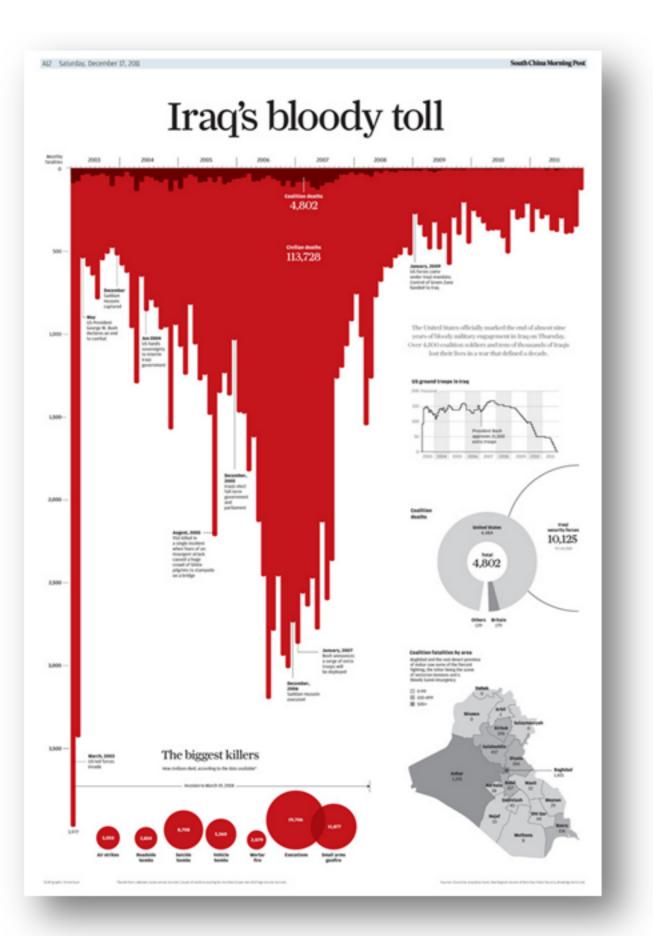
# What is the message?

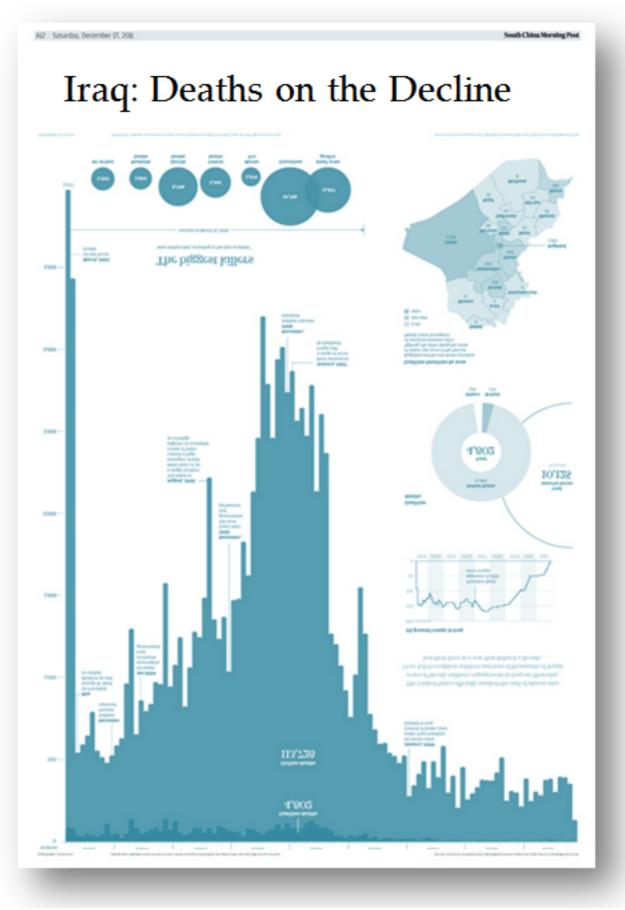
## Exploratory Neutral

Explanatory Opinionated









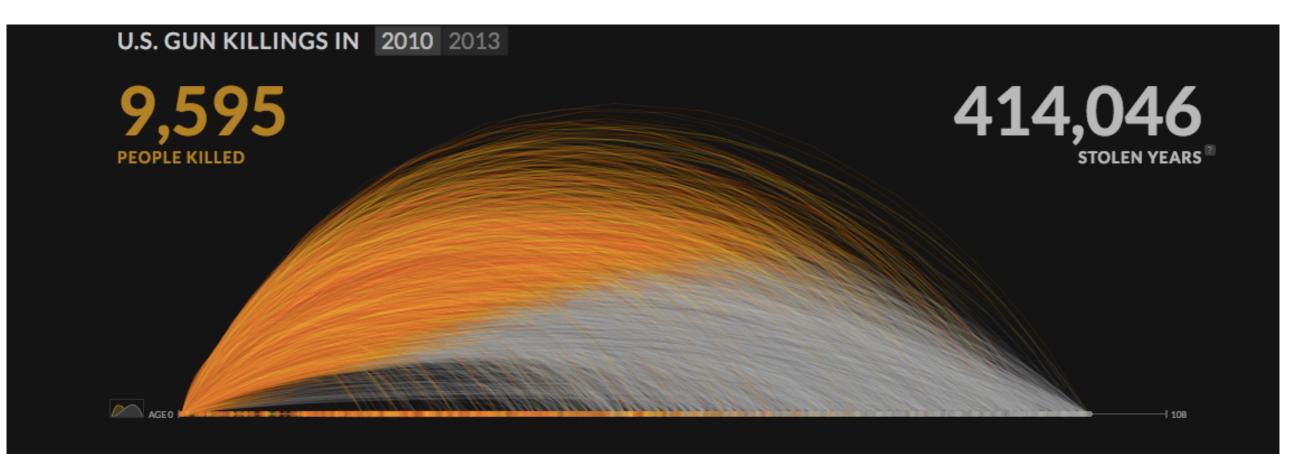
Andy Cotgreave, Tableau

# Framing - Why should I care?

- Tell the audience: "Here is the right way to think about the problem I was trying to solve."
- Catch the audience's attention and frame the story using captions and annotations
- If done well, your insights will seem obvious given this framing. And that's a good thing!



## Gun Deaths in 2010



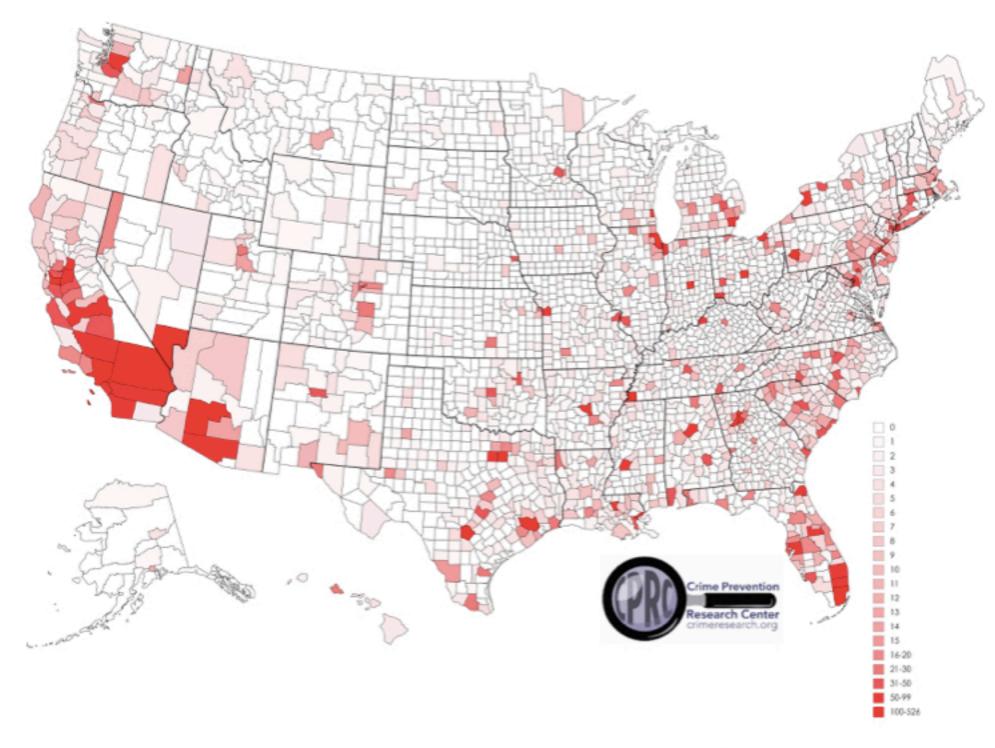
## Tools for interactive graphics

- R/shiny
- plotly/dash
- Tableau
- d3.js
- vega-lite/vega

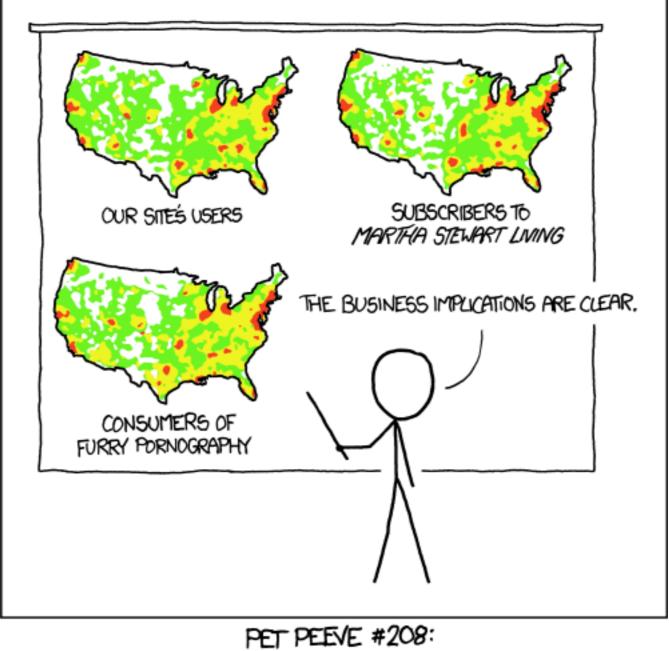
# Is there a story?

Surface it....even if it is incomplete

## 2014 Gun Deaths



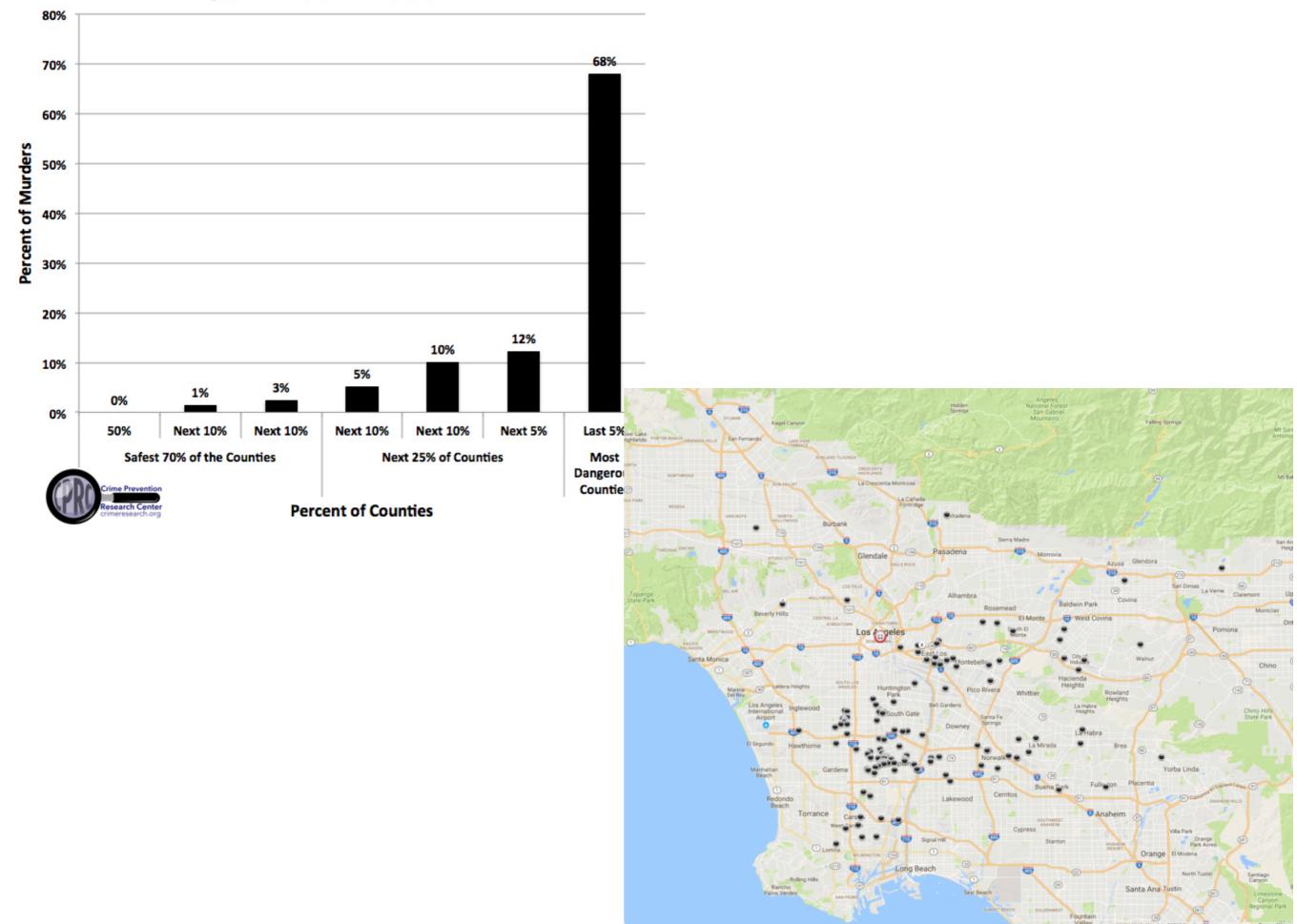
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GEOGRAPHIC PROFILE MAPS WHICH ARE BASICALLY JUST POPULATION MAPS

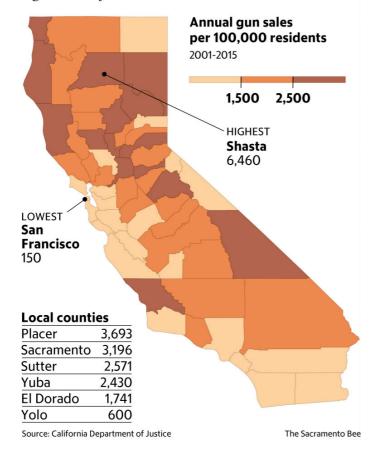
(XKCD)

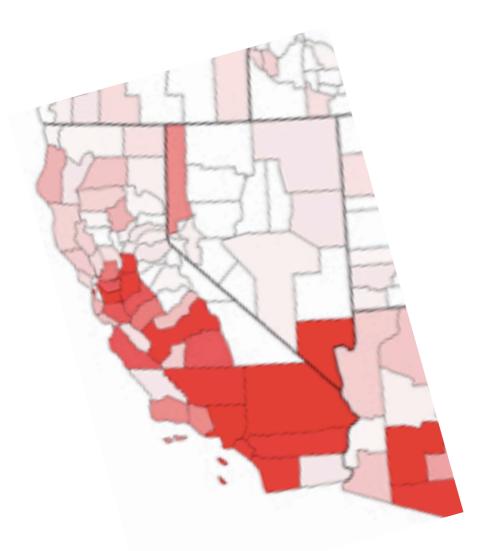
### Figure 1: Percent of murders



### Per capita sales by county

Annual gun sales per 100,000 residents are generally higher in California's rural and northern counties.





Deaths by county, 2014

## (crimeresearch.org)

## Whether Crime Is Up or Down Depends on Data Being Used

#### By TIMOTHY WILLIAMS SEPT. 27, 2016

## 



The news from the <u>F.B.I.</u> crime data was alarming: The murder rate rose sharply last year, driven by jumps in several major cities.

Four urban areas — Baltimore, Chicago, Milwaukee and Washington — accounted for about a fifth of the increase in homicides in 2015. Those cities, however, make up only about 1 percent of the nation's population.

But whether crime is up or down depends on what data is being looked at — and who is doing the looking.

The F.B.I. data showed that violent crime rose about 4 percent last year from 2014, and homicides increased 10.8 percent. Yet crime over all fell in 2015 for the 14th consecutive year.

And the total number of homicides last year was fewer than 20 years ago even as the country's population increased, criminologists said. There were 19,645 homicides in 1996 in a nation of 265 million; in 2015, there were 15,696 in a population of 321 million.

What that data means, criminologists and police officials said, is that the decline in homicides has been so significant in the last quarter century that sudden increases in the number of killings in just a few cities can skew the entire national picture, even as the country has one of its safest periods on record.

#### RELATED COVERAGE



U.S. Murders Surged in 2015, F.B.I. Finds

"It isn't a national trend, it's a city trend, and it's not even a city trend, but a problem in certain neighborhoods," said <u>Richard A. Berk</u>, a professor of statistics and criminology at the University of Pennsylvania. "Certainly, people around the country should not be worried. People in Chicago shouldn't be worried. But people in certain neighborhoods might be."

Criminologists and police officials point out that homicides do not usually disrupt entire cities. Instead, they occur in particular neighborhoods — and on the same blocks — leaving much of the rest of the city relatively untouched.

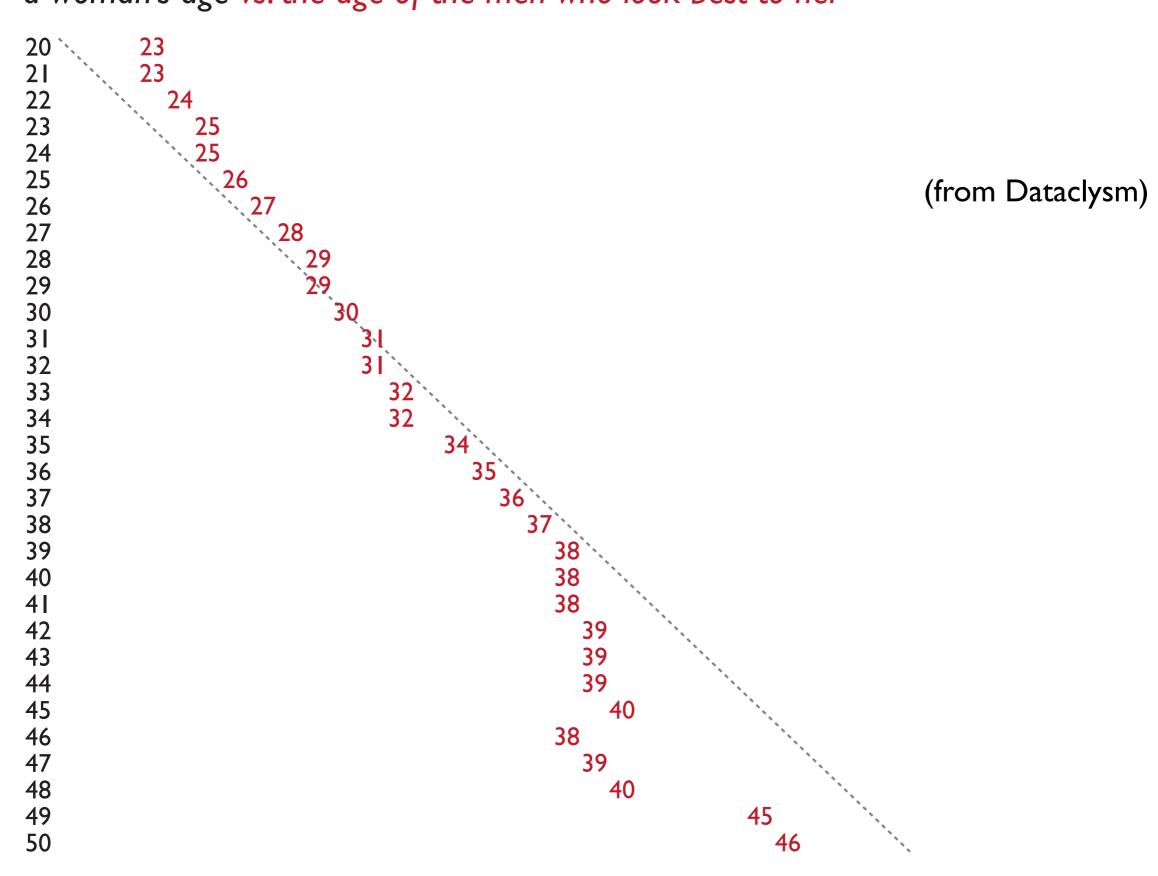
Explanations for the increase in homicides in certain American cities are largely guesswork. Criminologists acknowledge that the required analysis has not been done in the neighborhoods where killings are occurring — or even an agreement of what such a study should include — to arrive at any but the broadest conclusions.

## Careful with amalgamation paradoxes and with outliers

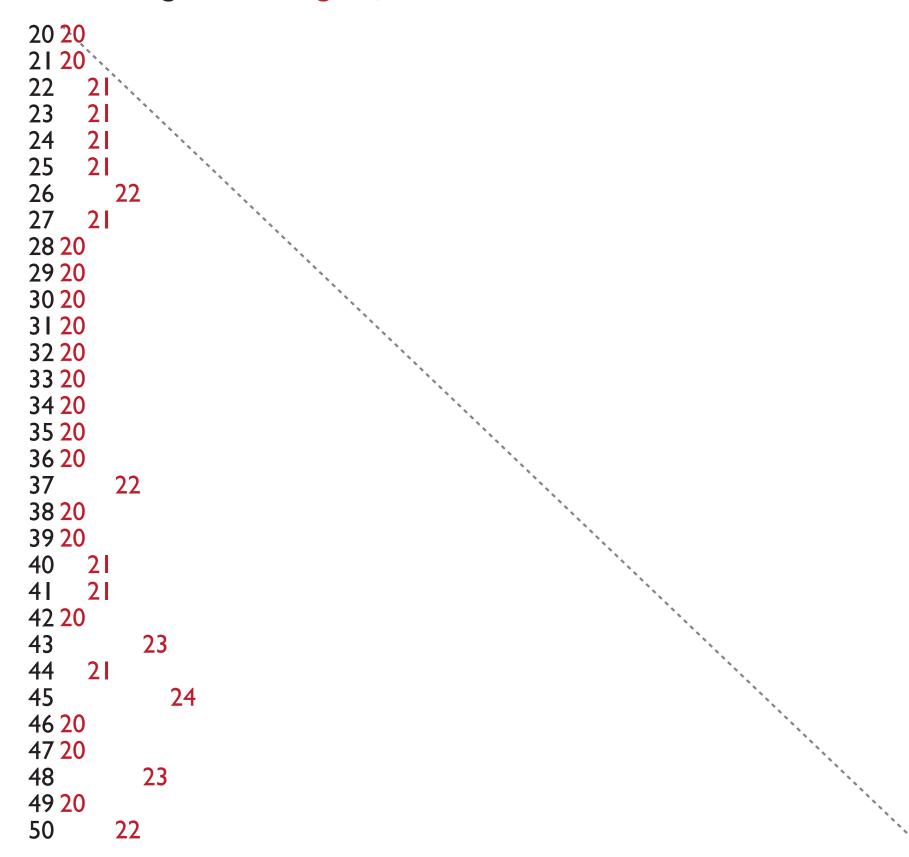
http://journal.frontiersin.org/article/10.3389/fpsyg.2013.00513/full

# Ask Ask Ask

- Is the exact distribution of guns really the important concern?
- did we check the uncertainties?
- Should we be looking at this from a "risk" perspective?
- we tend to believe what we believe and look for confirmation.
- we need to be disciplined about interrogating ourselves
- it is ok (and not against simplicity) to surface our process



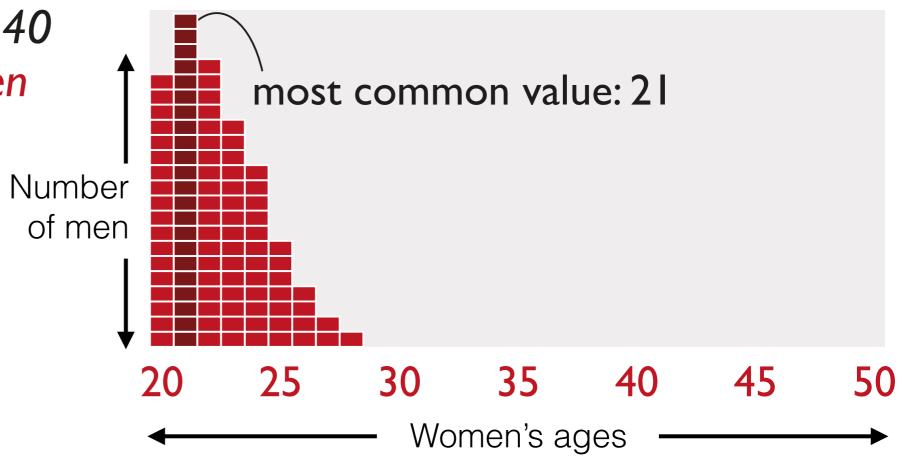
a woman's age vs. the age of the men who look best to her



a man's age vs. the age of the women who look best to him

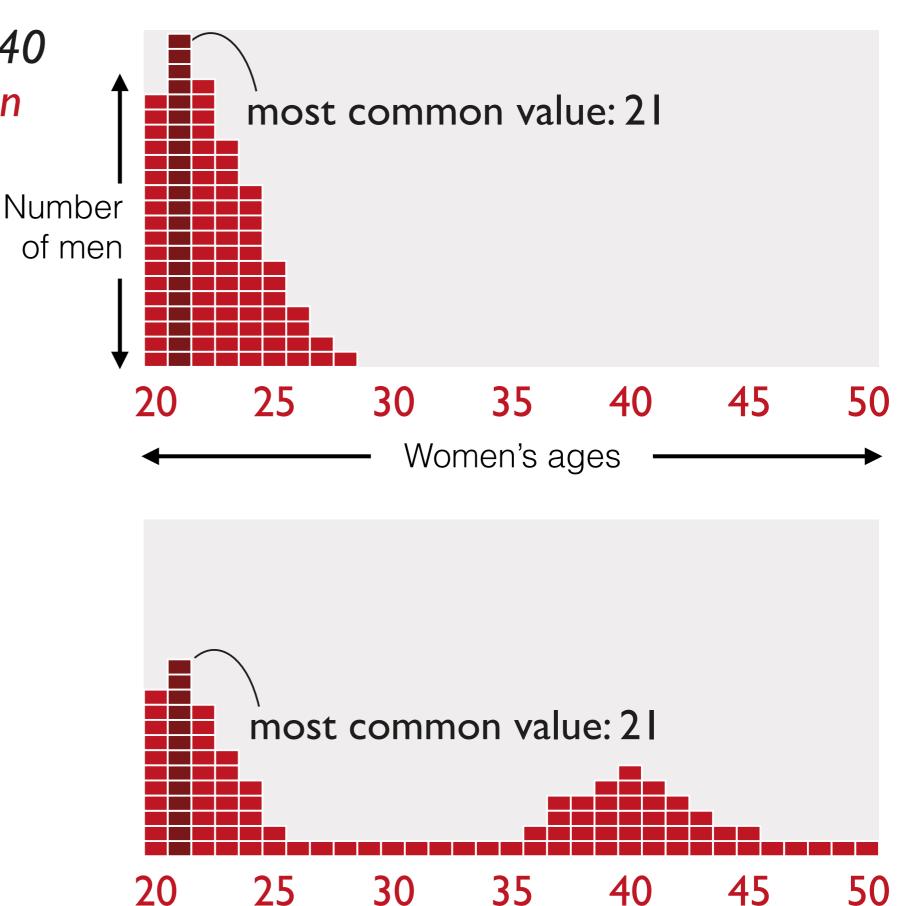
Sample of 100 men of 40 vs. the age of the women who look best to them

= | of men

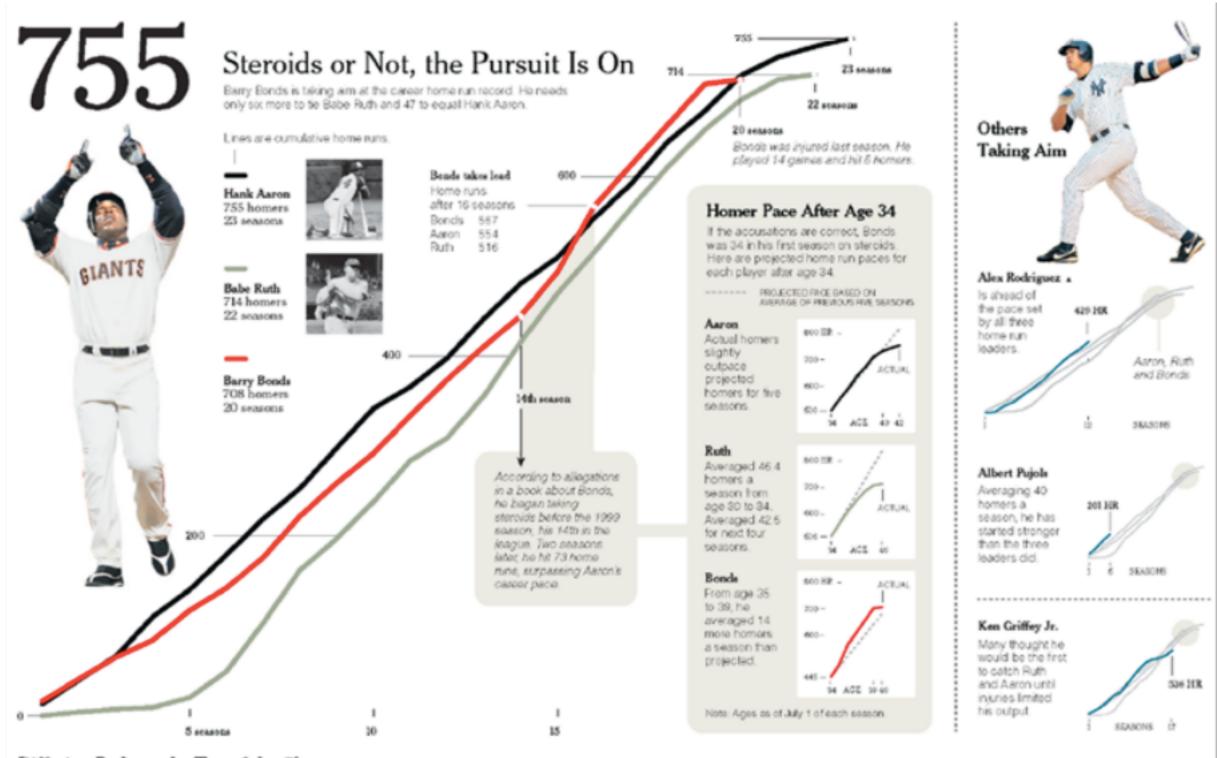


Sample of 100 men of 40 vs. the age of the women who look best to them

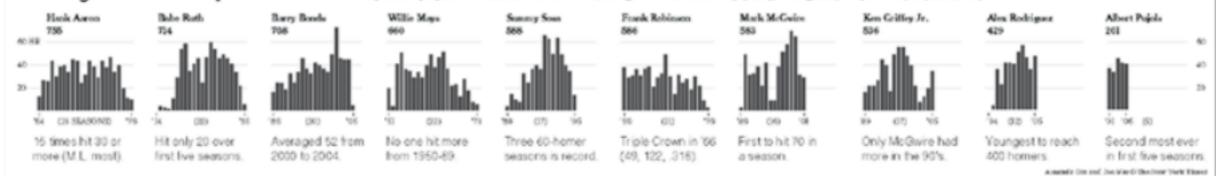
= = I of men



# Structure of communication graphics

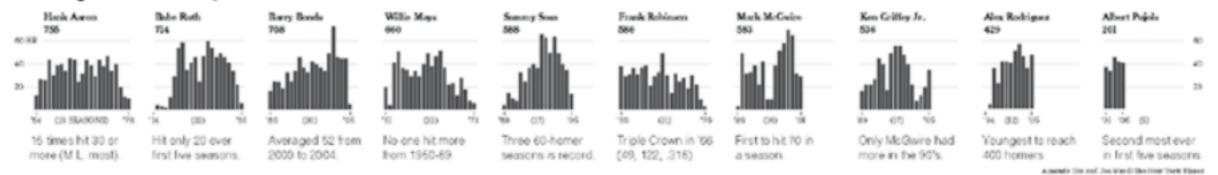


Differing Paths to the Top of the Charts The top seven players on the career home run list, along with a look at Griffey (12th), Rodriguez (37th) and Pujols (Sed 257th).

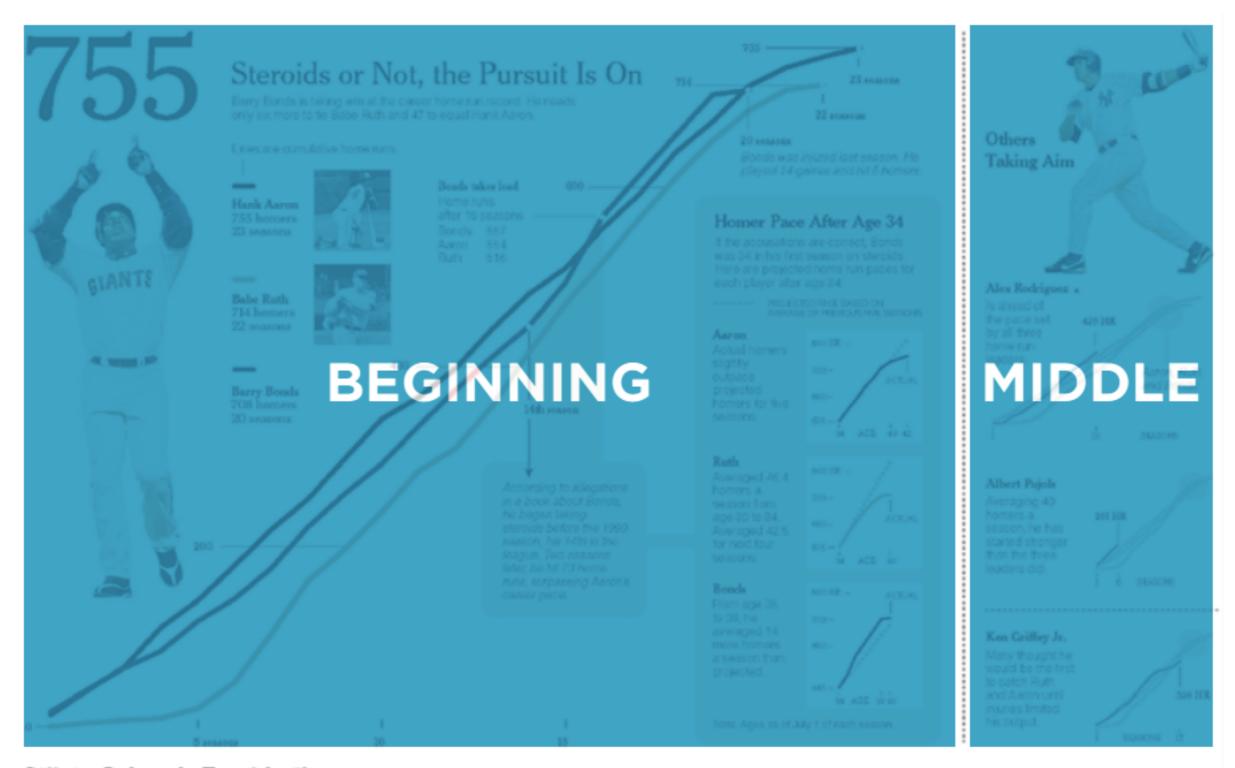




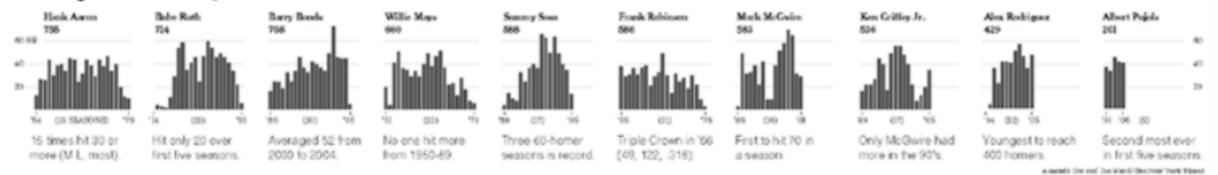
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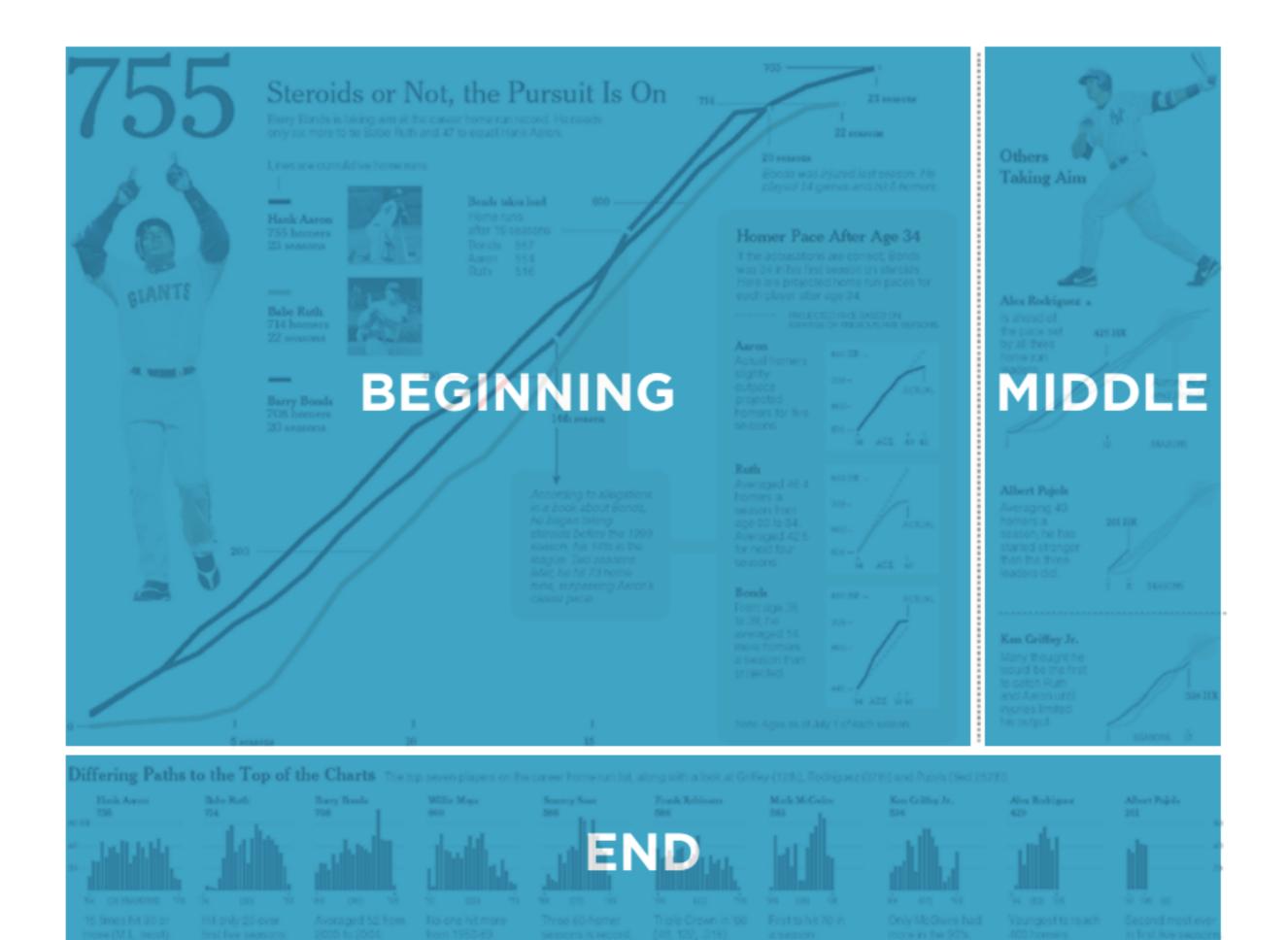


E. Segel

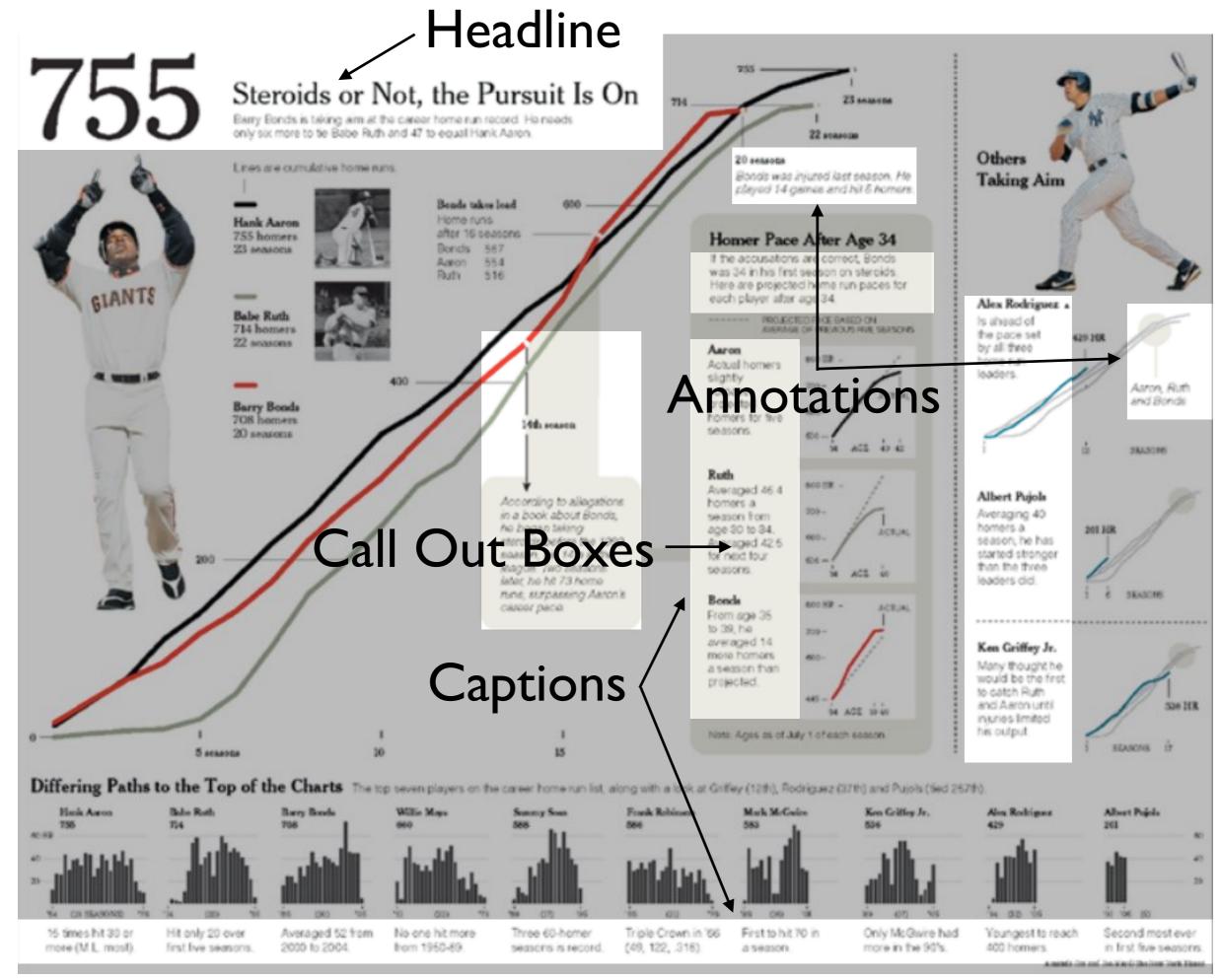


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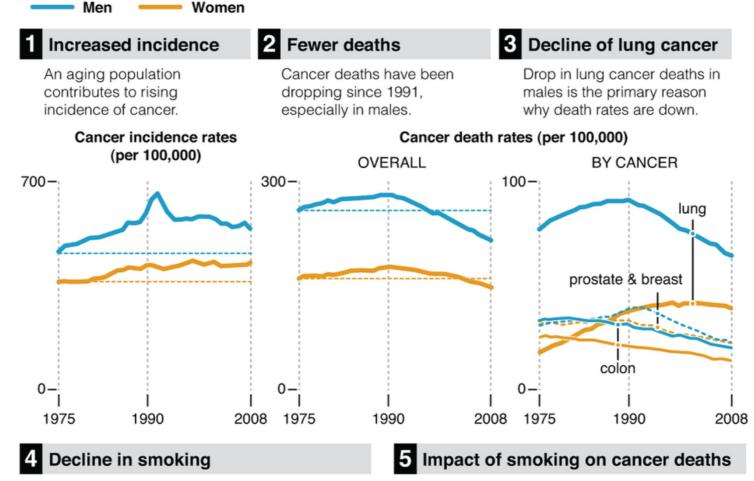


E. Segel

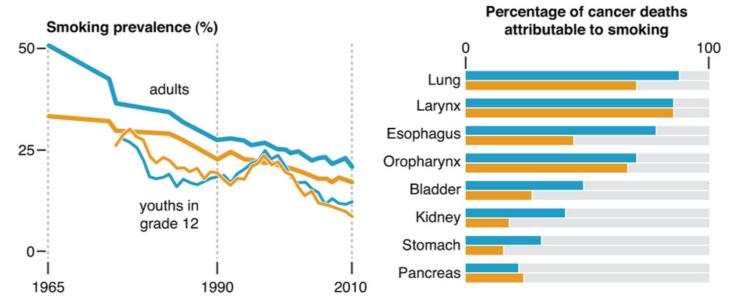


### WHERE THERE'S SMOKE—THERE'S CANCER

Cancer rates are up, but mortality is down. New diagnostics and treatments are responsible for part of this trend. But the greatest single contributing factor is the decline in smoking—rates are at their lowest level in 50 years.



Since the 1964 first Surgeon General's report, smoking rates have been dropping. By 2010, the rate among males was down to 20%, from 50% at its peak. Among youths, rates have been on an even steeper decline since 1997. Smoking is a major risk factor for many types of cancer and significant contributor to cancer-related deaths. It remains the single largest preventable cause of disease and premature death in the US.



M. Krzywinski & A. Cairo

source: American Cancer Society Cancer Statistics 2012; Monitoring the Future (University of Michigan).

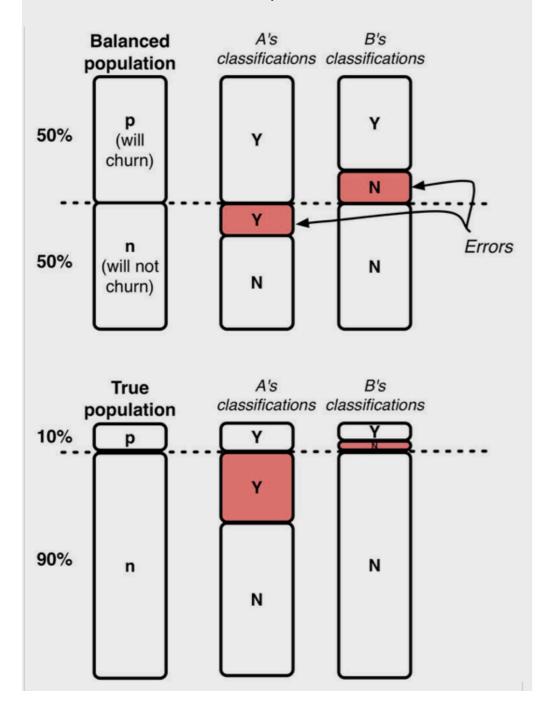
# Application to modeling

# IMAC

- I: inferential goal (scientific question of interest) M: model (all models are wrong, some are useful) A: algorithms
- C: conclusions and checking

The C is crucial: what did we learn? Was the model useful, and how well does it fit? How do we know whether the method is working? Do we understand how it is working? Do we need to iterate and improve the model? What are the limitations and future directions?

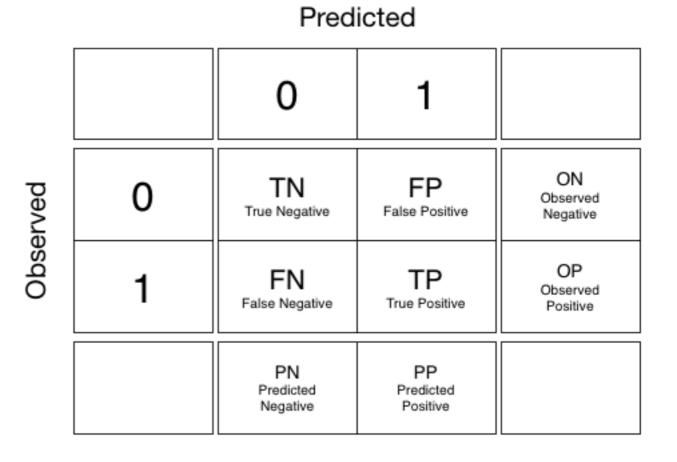
## (from Foster and Fawcett)



Which Model is Better?

## Breast Cancer on a Mammogram

- False positives OK
- False Negatives are disaster
- More people dont have it



# Communicating a model

## **Telecom Churn Problem**

Survey 1000 customers, with an offer with an administrative cost of \$3 and an offer cost of \$100, an incentive for the customer to stay with us.

Want to predict for our 100000 customer base.

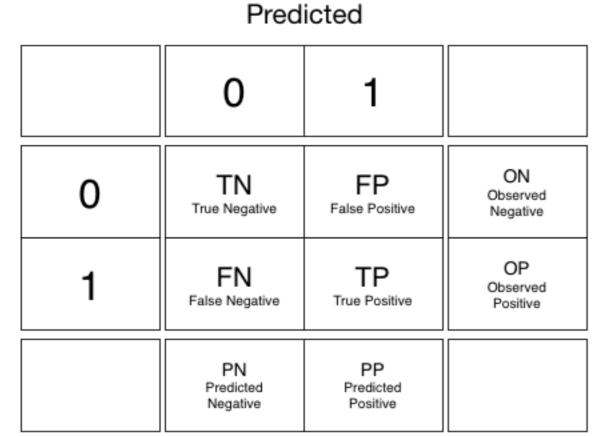
If a customer leaves us, we lose the customer lifetime value, which is some kind of measure of the lost profit from that customer.

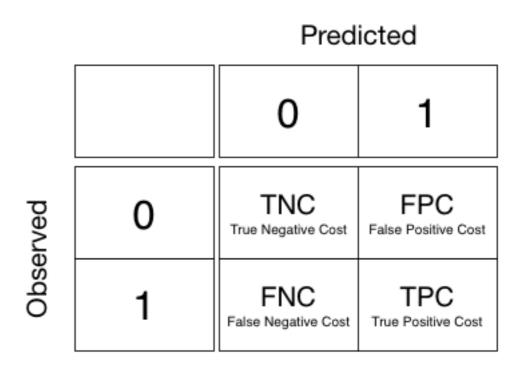
Lets assume this is the average number of months a customer stays with the telecom times the net revenue from the customer per month. We'll assume 3 years and \$30/month margin per user lost, for roughly a \$1000 loss.

admin\_cost=3
offer\_cost=100
clv=1000 # customer lifetime value

- TN=people we predicted not to churn who wont churn.We associate no cost with this as they continue being our customers
- FP=people we predict to churn. Who wont. Lets associate a admin\_cost+offer\_cost cost per customer with this as we will spend some money on getting them not to churn, but we will lose this money.
- FN=people we predict wont churn. And we send them nothing. But they will. This is the big loss, the clv
- TP= people who we predict will churn. And they will. These are the people we can do something with. So we make them an offer. Say a fraction f accept it. Our cost is admin\_cost + f\*offer\_cost + (1-f)\*clv.

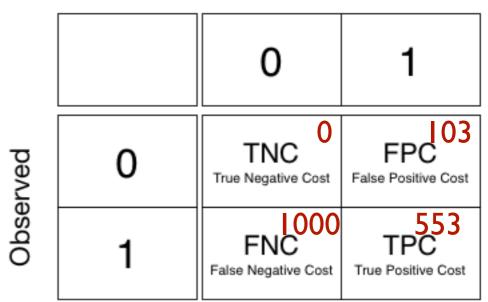
```
f = 0.5
tnc = 0.
fpc = admin_cost+offer_cost
fnc = clv
tpc = admin cost + f*offer cost + (1. - f)*clv
```



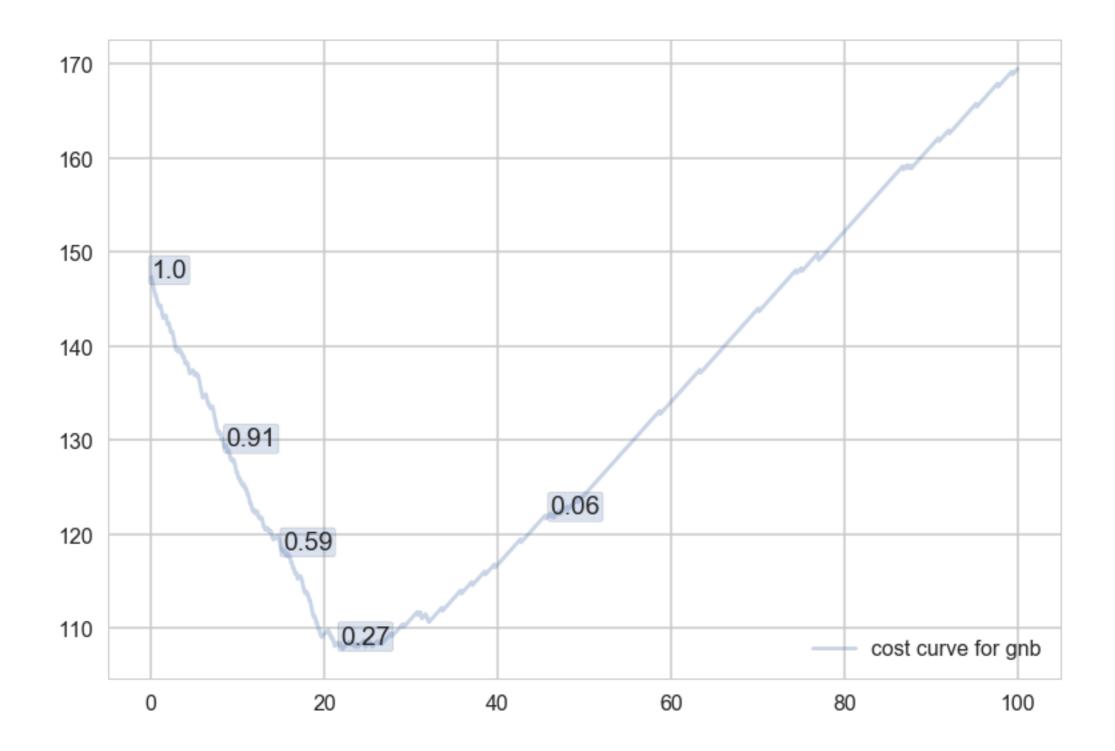


Predicted

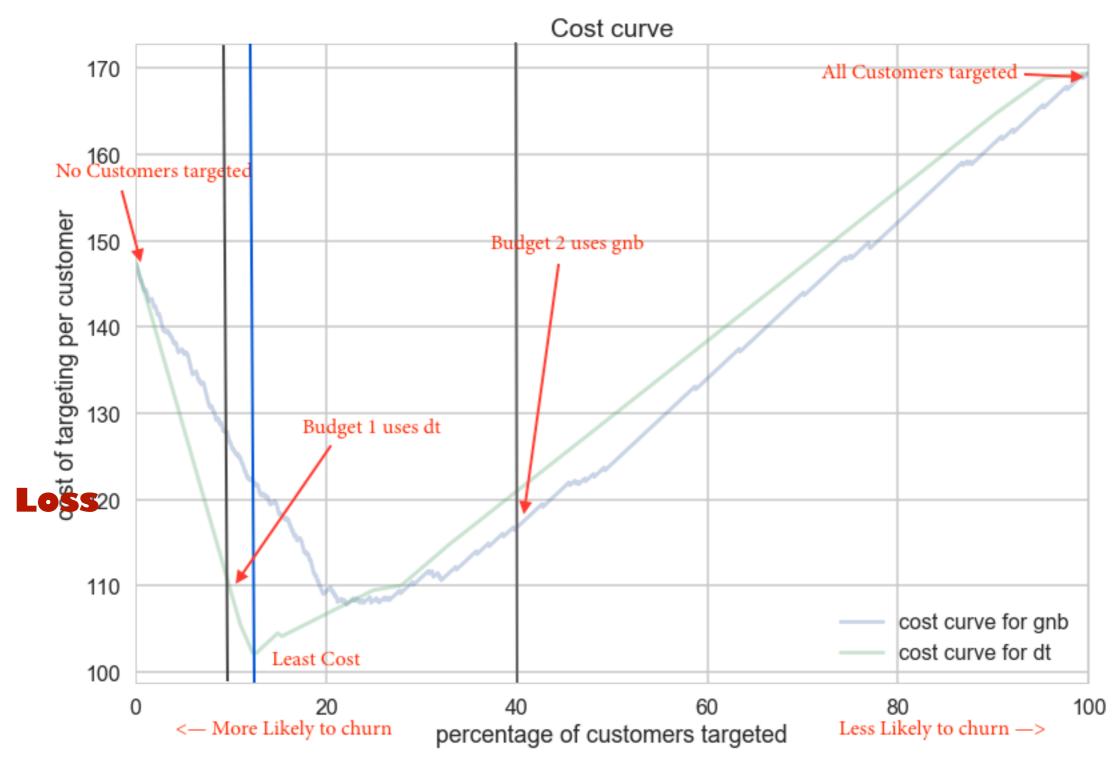
## Average Cost = TN x TNC + TP x TPC + FN x FNC + TP x TPC



Observed

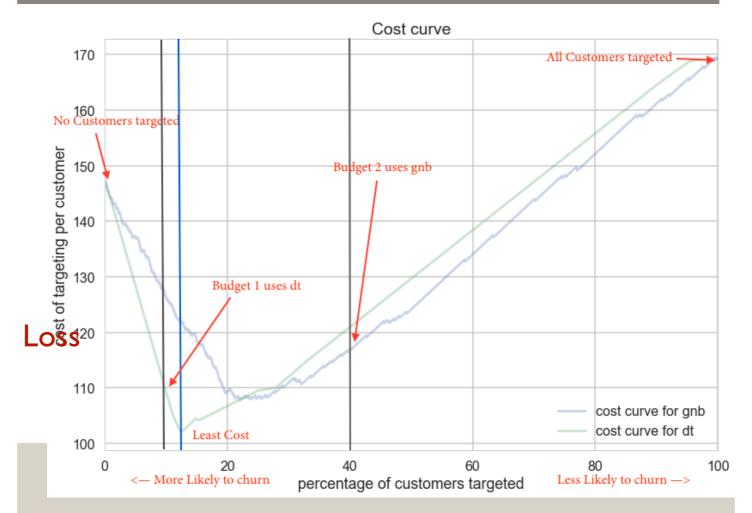


## Annotated Diagram



made with Preview

### Reduce churn and our cost by sending customers an offer



### Making offers within Budget

This study was made on a pilot survey of 1000 customers from our 100000 customer base.

Make an offer with an **administrative cost of \$3** and an **offer cost of \$100**, an incentive for the customer to stay with us.

If a customer leaves us, we lose the customer lifetime value (CLV), a **roughly \$1000 loss**.

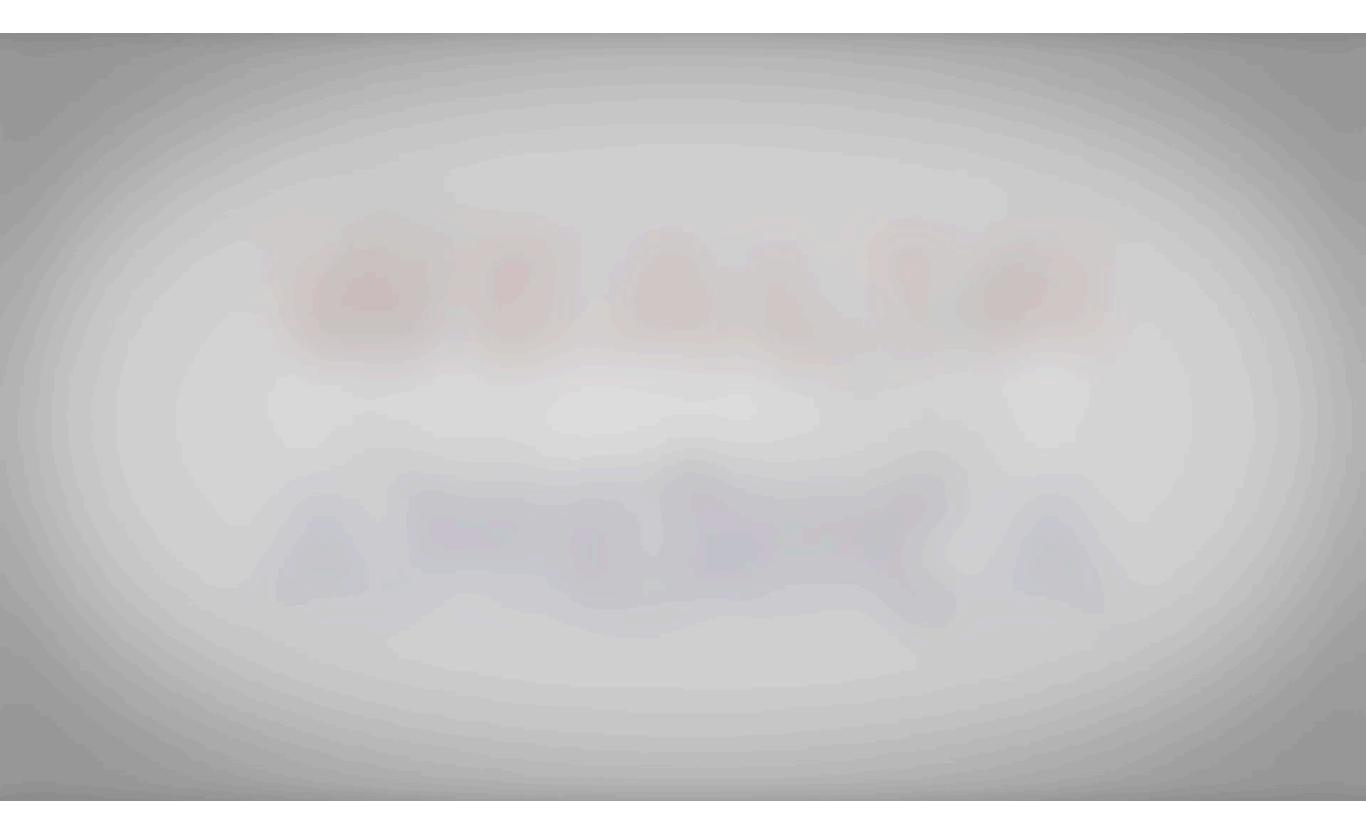
We assume that 50% of those customers targeted will stay with us.

If we do nothing we lose \$150 per customer including CLV

### We **choose which customers to target** according to 2 different models, **dt** and **gnb:**

- Making an offer to 13% of our most likely to leave customers will cut this cost to a lowest value of \$103 per customer according to the **dt** model, for a total cost of \$1.34 million.
- If we only target 10% of the customers (Budget 1) using the **dt** model, we get by in 1.03 million but incur a loss of \$110 per customer including CLV.
- If we target 40% of our customers, we need a budget (Budget 2) of \$4.2 million. Here the **gnb** model performs better and we will choose customers according to it. We incur a loss of \$116 per customer including CLV.

# StoryTelling



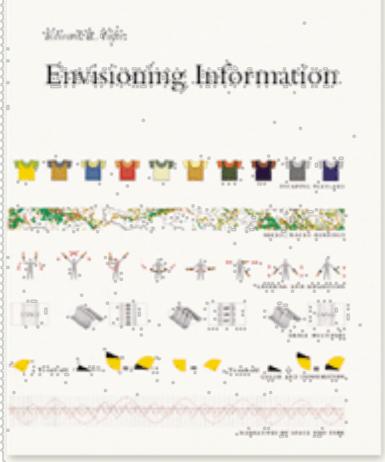
# Edward Tufte

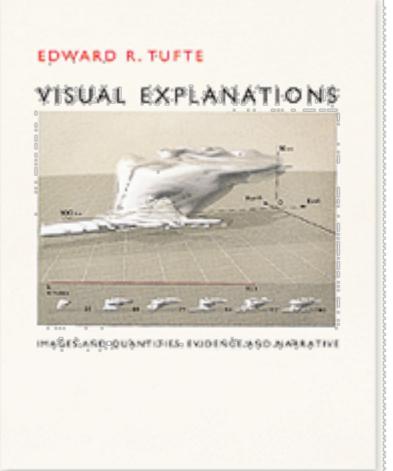




The Visual Display of Quantitative Information

EDWARD R. TUFTE





# Stephen Few





# 2012

an introduction to information graphics and visualization

the truthful art data, charts, and maps for communication

2016

I've always believed in the power of data visualization (the representation of information by means of charts, diagrams, maps, etc.) to enable understanding

Alberto Cairo • University of Miami • <u>www.thefunctionalart.com</u> • Twitter: @albertocairo

alberto cairo

Welcome to Alberto's world. Cairo has done it all in The Functiono Art: theory, practice, examples. And he's done it brilliantly, it is the most comprehensive and sensible book yet on real-world information graphics; we won't need another one for a long time." Nigel Holmes, former graphics director for Time magazine and founder of Explanation Graphics