

# Looking For Truth Or At Least Data

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# Important Disclaimers

- All the numbers in this presentation are made up.
- The stories are true.
- I am not a statistician.
- I'm done with the funky transitions now.

# Audience

- System Administrators
- Not statisticians
- Mostly collecting data about machines

- Numbers: good
- Believing appearances: bad
- Making stuff up: ??

# What Am I Talking About?

- An attitude
- A hobby
- Where science, system administration, and security overlap

# Fundamentals

- “That’s interesting. I wonder what I could find out about it?”
- Distinguish between “what appears to be” and “what is”.
- Understand numbers.

# Why Might You Care?

- Planning systems and upgrades
- Troubleshooting
- Being good at security
- Just plain fun
- Not falling for pseudo-science

# Recognizing Data

- Is this data?
- What is it data about?
- What conclusions can we draw from it?

# Is This Data?

- “The CEO says the network is slow.”
- “47 users complained about network slowness yesterday.”
- “Average network latency yesterday was 15 milliseconds.”

# Is This Data?

- “I feel like something might be wrong with a core router.”
- “Brand A’s router has an error rate 200% worse than Brand B.”
- “Sites that use Brand A’s router report slowness more often.”

# Is This Data?

- “We didn’t change anything around the time people started complaining about the network.”
- “We changed the routing just before people started complaining about the network.”
- “People are complaining because you changed the routing.”

# Not Data

- Hearsay
- Numbers without context
- Conclusions

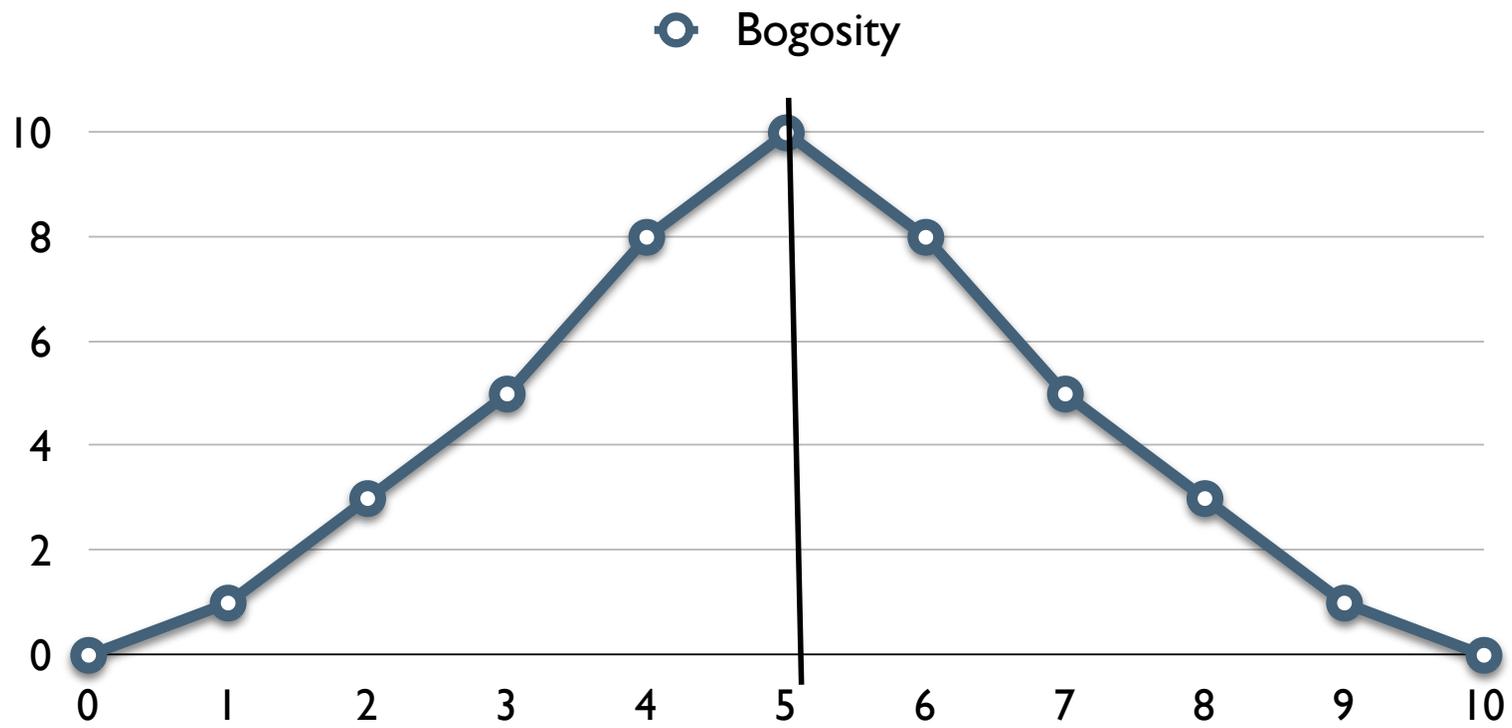
# Data

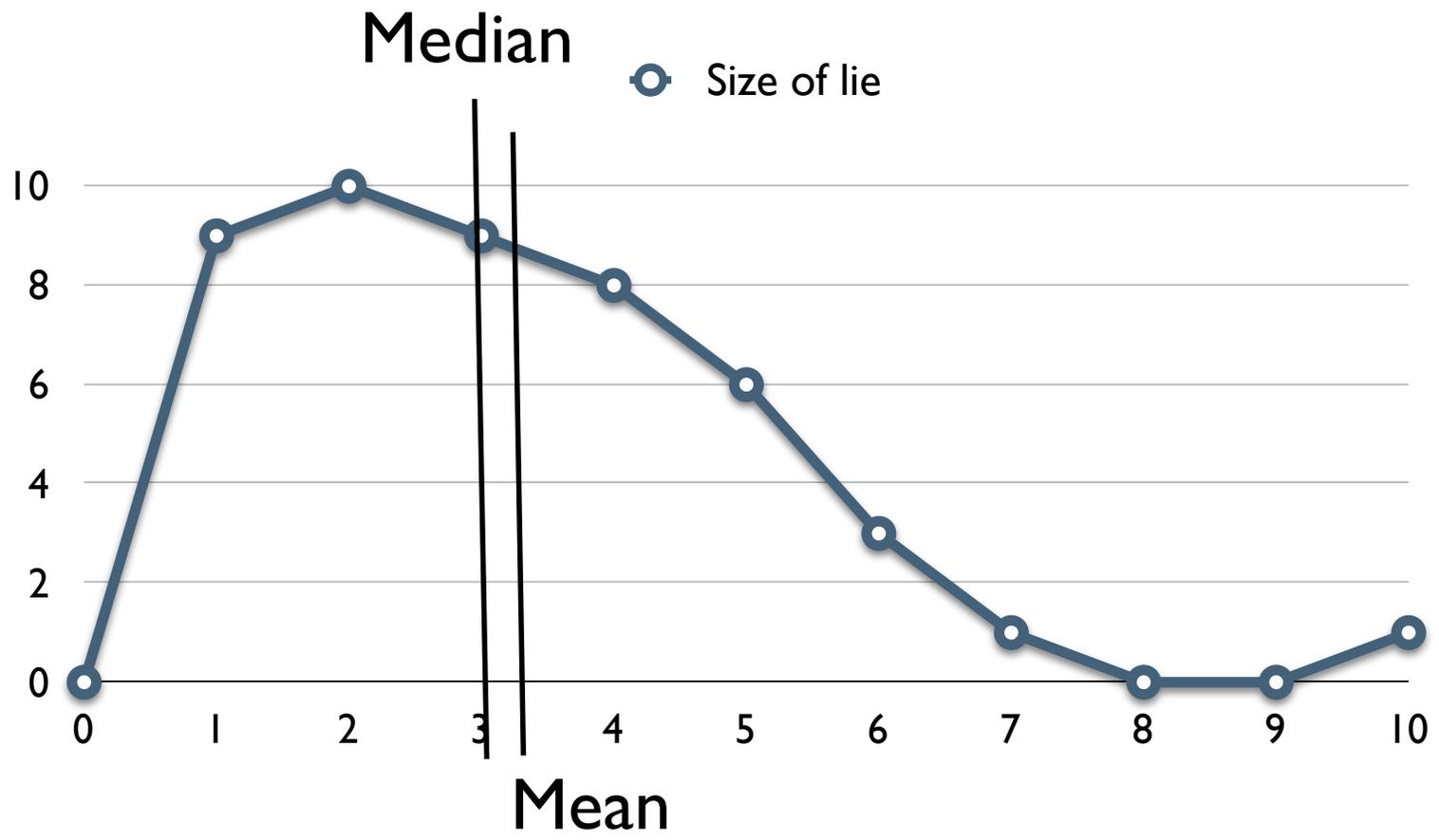
- Observations
- Self-report
- Numbers in context

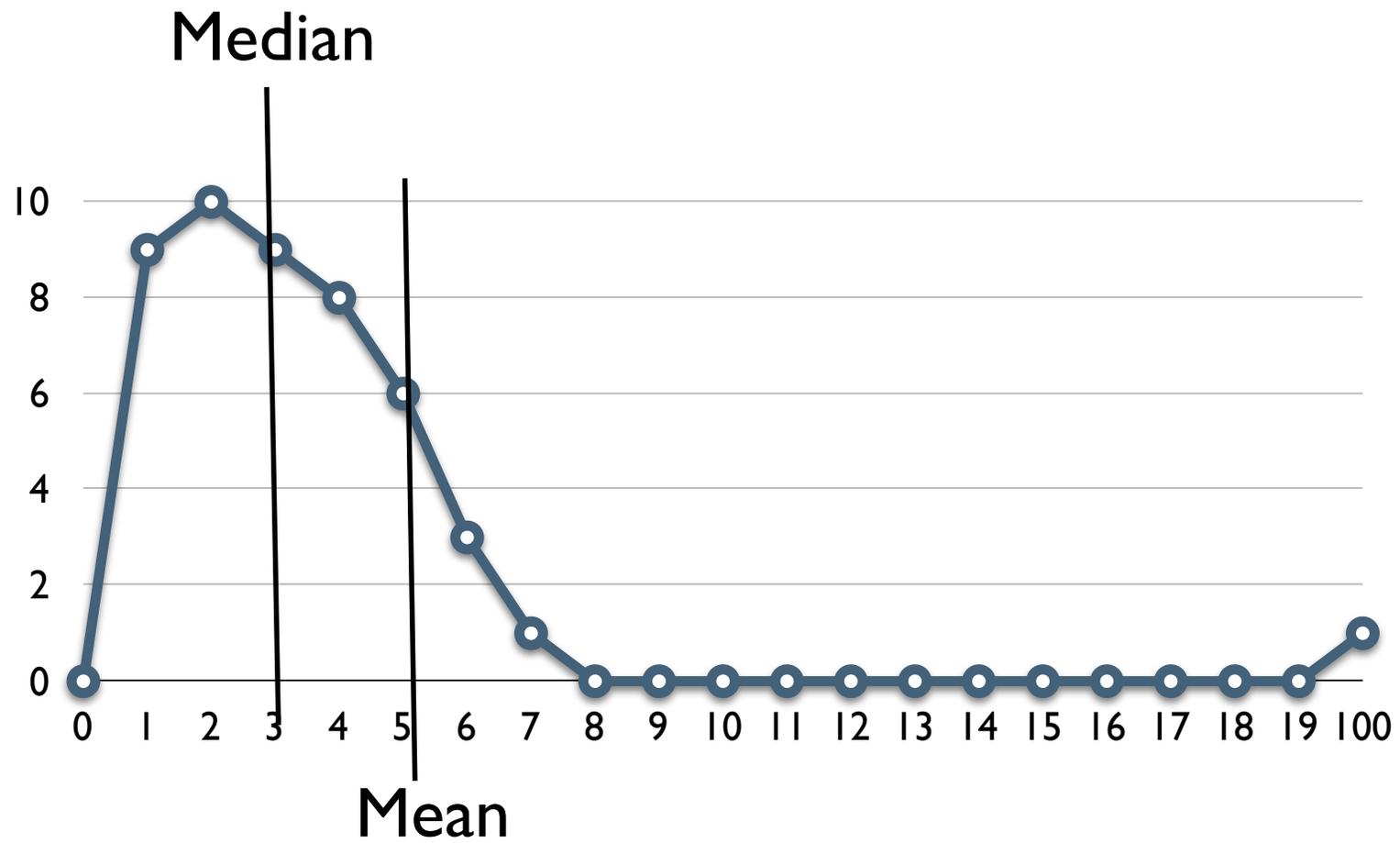
# Why Those Numbers Aren't Data

# Basic Statistical Skepticism

- What do you mean “average”?
- Compared to what?
- What do you mean by “correlated”?







# Average

- Means are only interesting for symmetrical single-peaked curves.
- Your data probably does not make one of them.
- You probably want median, quartiles, or percentiles.
- If you do want a mean, you want a standard deviation.

# What Can You Do?

- Forget the average, look at a picture of the numbers.
- Ask what kind of average it is.
- Ask what the standard deviation is.

# Compared to...

- Is 99.9% accuracy good?
- If your false positive rate on network packets is .1%, you get a false alarm every...
- And your false negative rate?

# Better and Worse

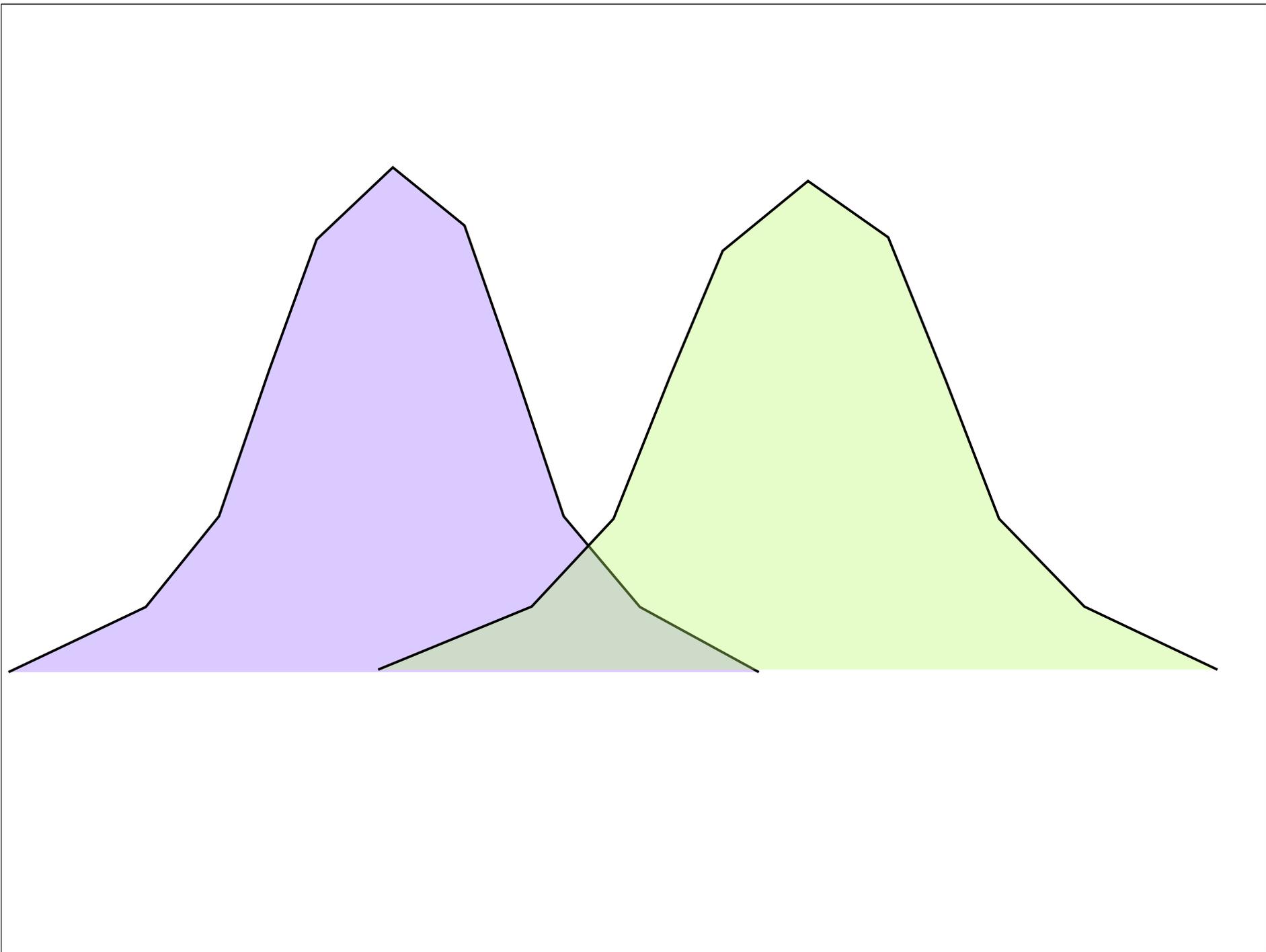
- Is a 200% increase in error rate bad?
- If your initial error rate was 1 in 4, your new error is 3 in 4.
- If your initial error rate was 1 in a million, your new error rate is 3 in a million.

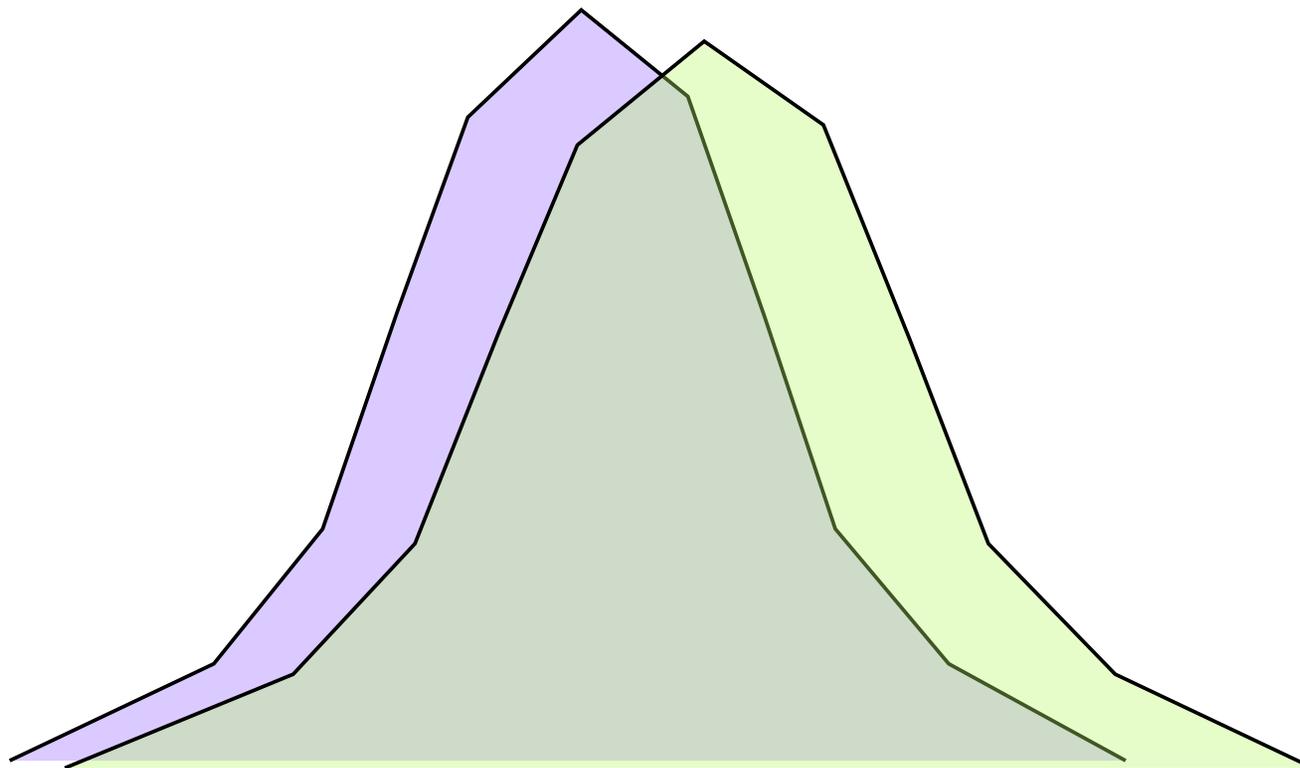
# Error Rates Again

- Suppose both routers have the same error rate
- but one of them eats every millionth packet (random error)
- and the other eats every packet of a rare type (systematic error)

# Correlations

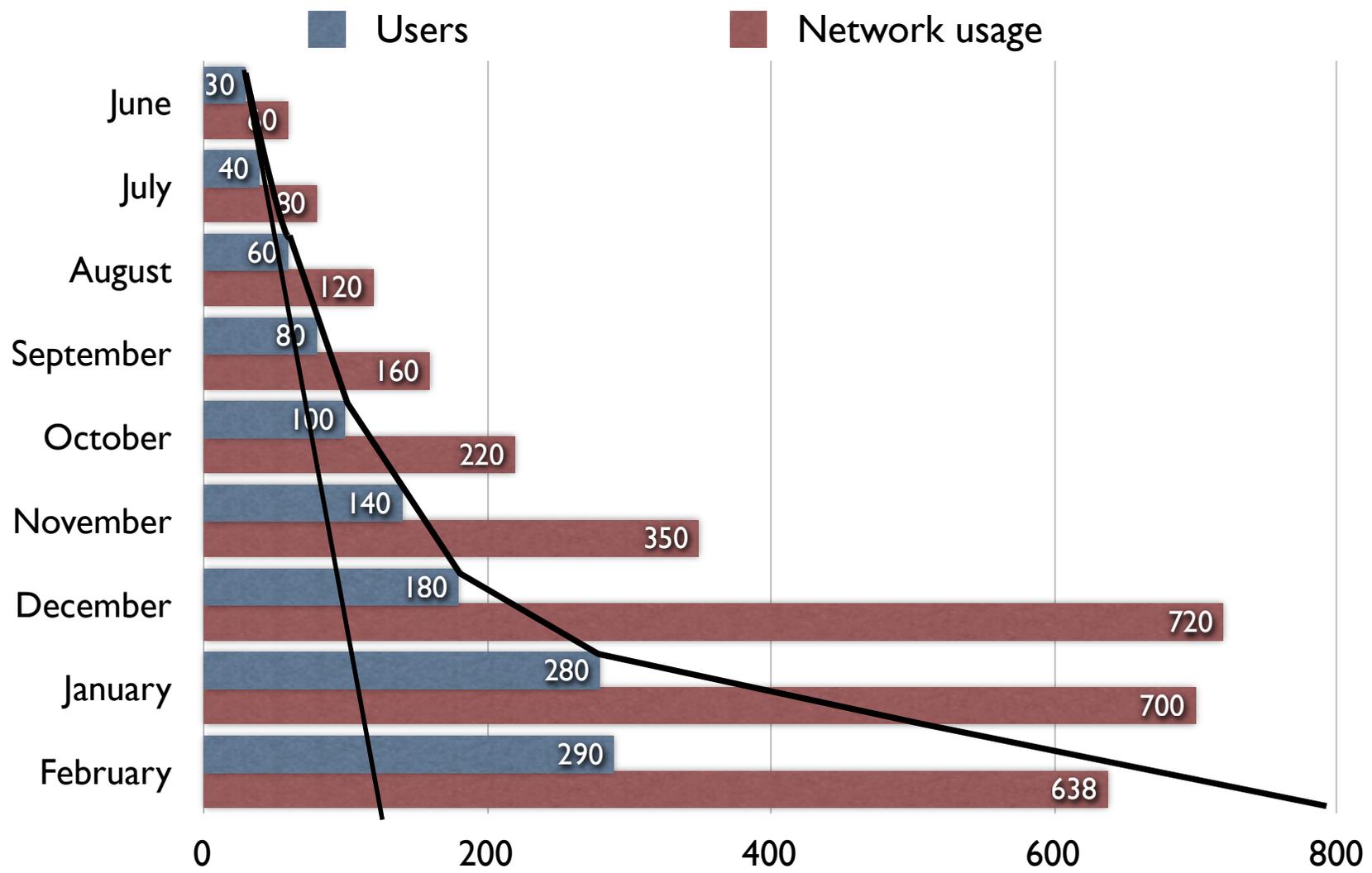
- “Sites that use Brand A routers are more likely to report slowness.”
- Correlation does not imply causation.
- Some correlations are weak.
- If you look at enough correlations, some of them will be “strong”.

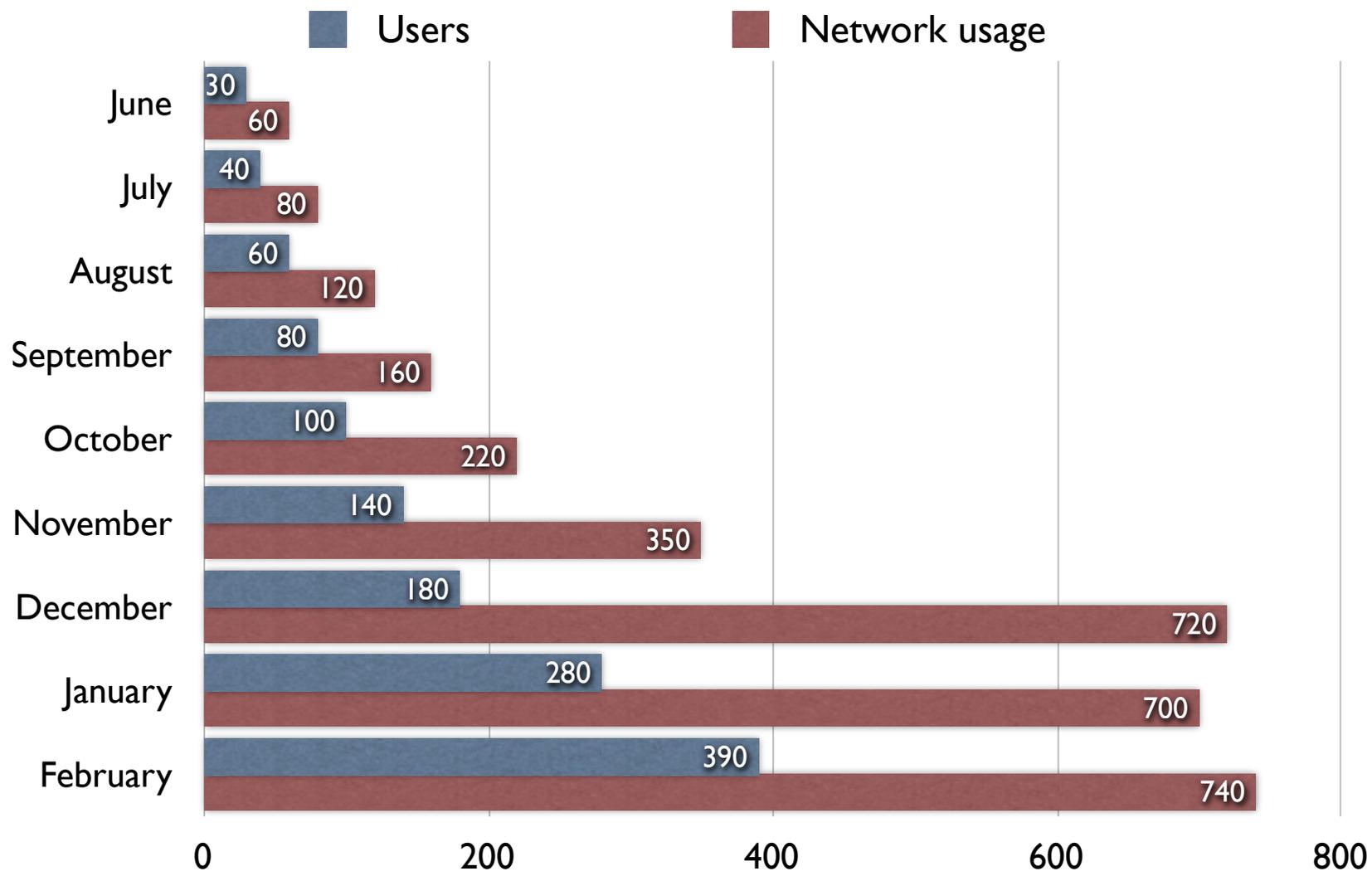


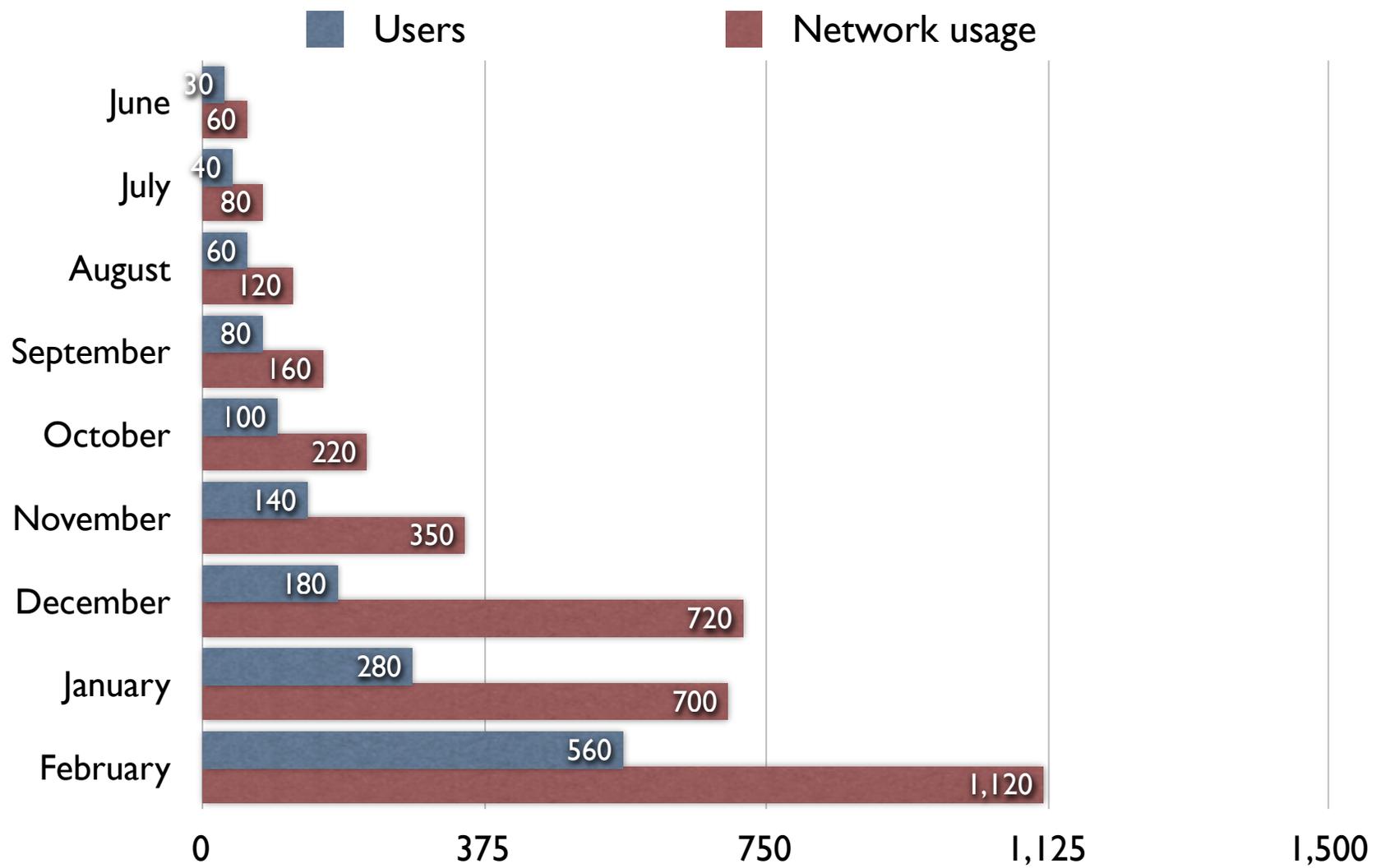


# What Is It About?

- “47 users complained about network slowness yesterday”
  - is real data
  - about users
- “Network usage is increasing rapidly”







# What Is It About?

- Most data is about lots of things
- The users are complaining it's slow because
  - it's slower
  - they changed applications
  - they're unhappy

# What conclusions?

- From the data I've shown:
  - Either your network will be overprovisioned most of the year, or December is going to be nasty.

# What Conclusions?

- Data is a lot easier to find than truth.
- Be very cautious in the conclusions you draw from data.
- Correlation does not imply causation.

# Gathering Data

# Basic Tools

- A programming language, preferably one that's good with text.
- Some programs for looking at the guts of things.
- Some programs for making data into pictures.

# Looking at Guts

- trace, dtrace, truss
- wireshark, tcpdump
- Windows sysInternals

# Making Data into Pictures

- Your favorite spreadsheet
- GraphViz
- gnuplot

# Basic Knowledge

- Regular expressions
- SQL
- XML
- Basic statistics

# Finding Data

- Mine existing sources
- Collect data
- Simulate and/or extrapolate
- Find somebody else with data
- Make stuff up

# Mine Existing Data

- How many files have we got? Count them.
- What are people's names like? Look them up.
- Those log files must be good for something

# Collect Data

- Add logging
- Save snapshots of changing data
- Use tracing or network sniffing
- Run tests

# Simulate and/or Extrapolate

- Set up a test situation
- Find a similar situation
- And then go back to mining or collecting data

# Find Somebody Else With Data

- Published sources
- Friends and colleagues
- Get the rawest available data
- Know as much about it as possible

# Make Stuff Up

- If all else fails, try guessing
- Get a lot of guesses
- Base guesses on knowns as much as possible
- Play around to see how changing guesses changes outcomes

# Backups

- How much data will a given backup scheme backup?
- Mining: pull data from existing backup system.
- Collection: record statistics by day
- Simulation: make up a model of how people behave, see how much data

# Educating Users on Security

- Mining: What do people currently look for or read?
- Collection: What do they do with changed content?
- Research: What do we know about naive users and security?

# Collecting Data About People

- Human Subjects Boards and ethics
- Random sampling is good
- If you can't be right,
  - be qualitative instead of quantitative
  - be wrong lots of different ways
  - at least understand why you're wrong

# What Next?

- Maybe fascinating things will just jump out at you.
- Maybe you just need to ask “why”?
- Maybe you’re going to use that data.

# Cuckoo's Egg

- Cliff Stoll tracks a quarter



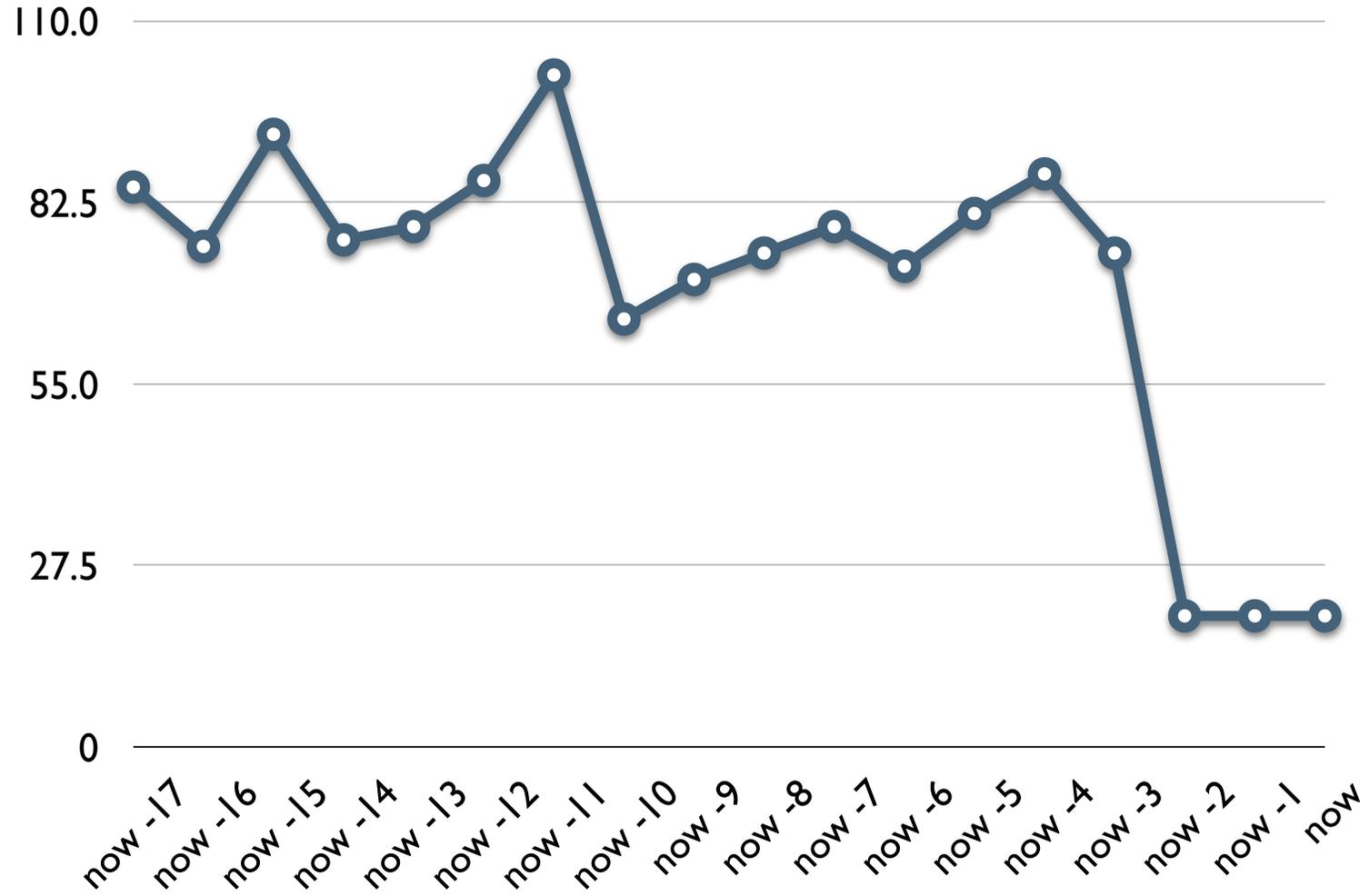
# Sanity Checking

- Another reason you might be asking “why”?
- Some data collection is wrong
- Some data collection reveals other problems

# Analyzing Data

- Let the data lead you
- Know what questions you want to ask
- Humans are good at very specific sorts of pattern recognition

○ Mystery Measurement



# Humans are Good At

- Noticing abrupt change
- Finding correlation
- Seeing faces

# Humans are Bad At

- Evaluating probability
- Finding non-correlation
- Perceiving slow change
- Perceiving correlation with time delay

# Displaying Data

- Decide what you want to say
- Display that with only minimal other facts

# Not Lying With Graphs

- Up is good, down is bad.
- Humans perceive area, but not well.
- Whenever possible, start at 0.



**76.13% Bounce Rate**



**00:01:22 Avg. Time on Site**



**62.13% % New Visits**

■ Region I

100

75

50

25

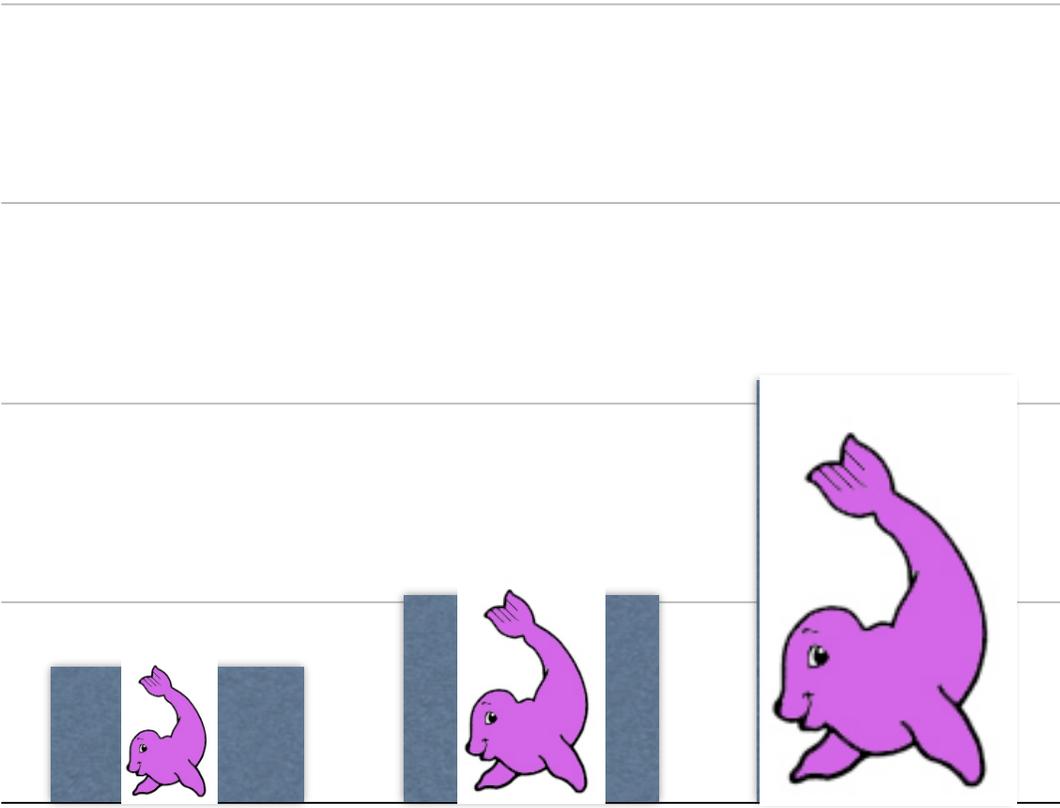
0

2007

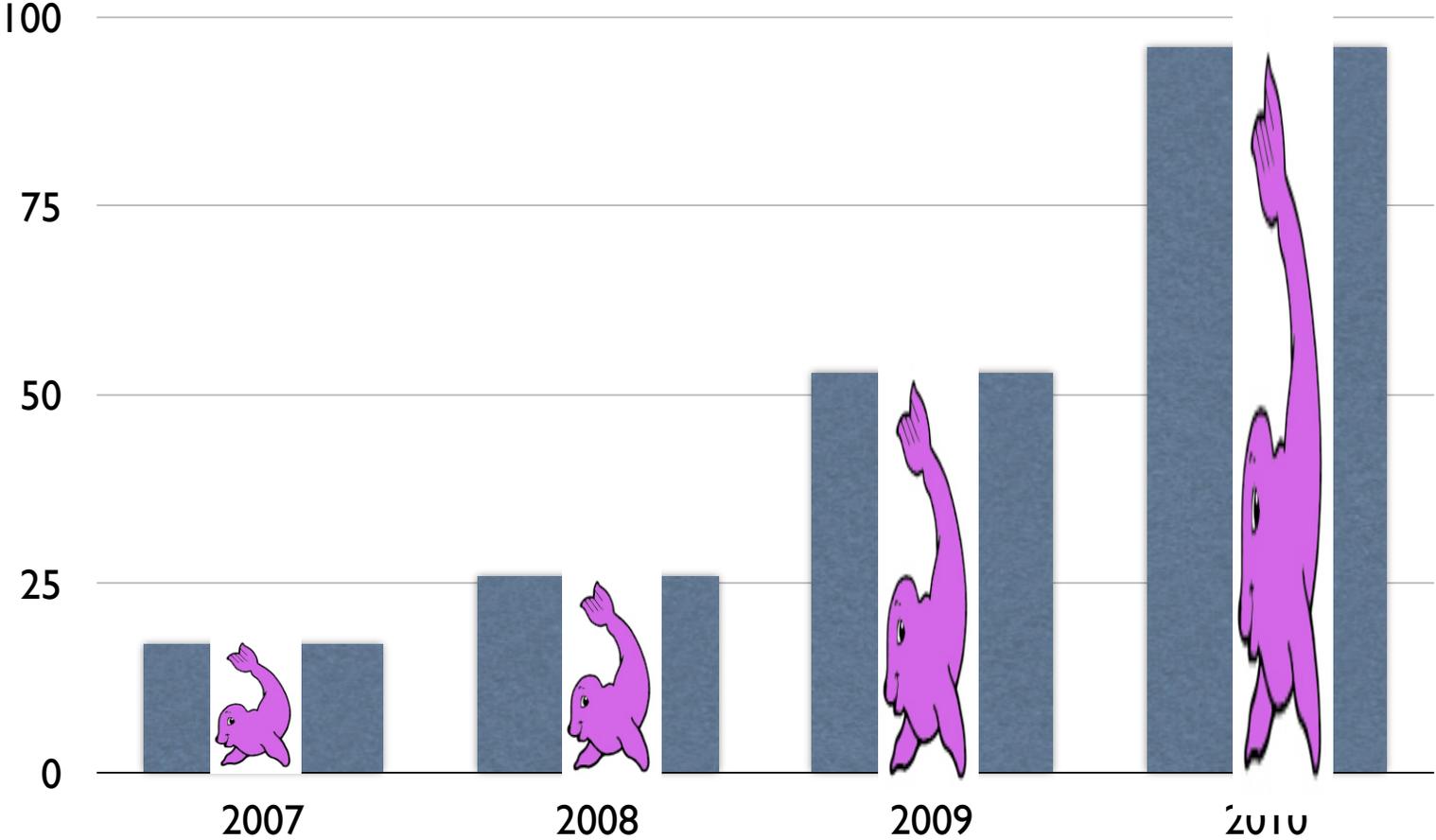
2008

2009

2010



■ Region I

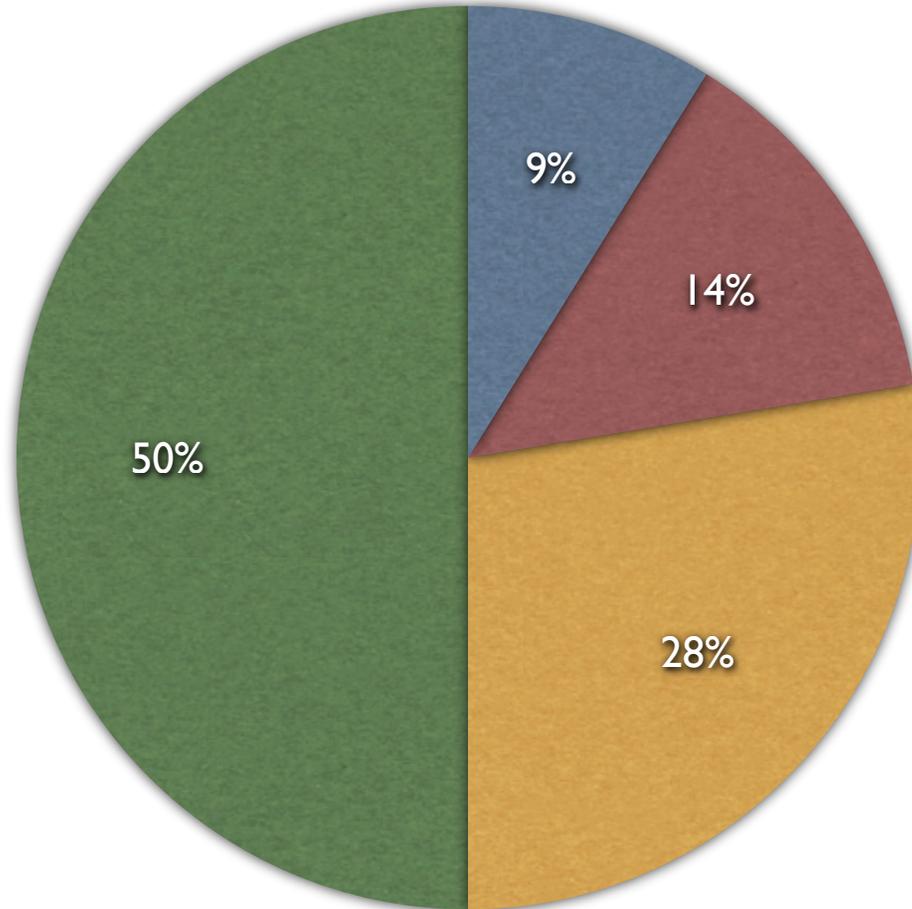


● 2007

● 2008

● 2009

● 2010

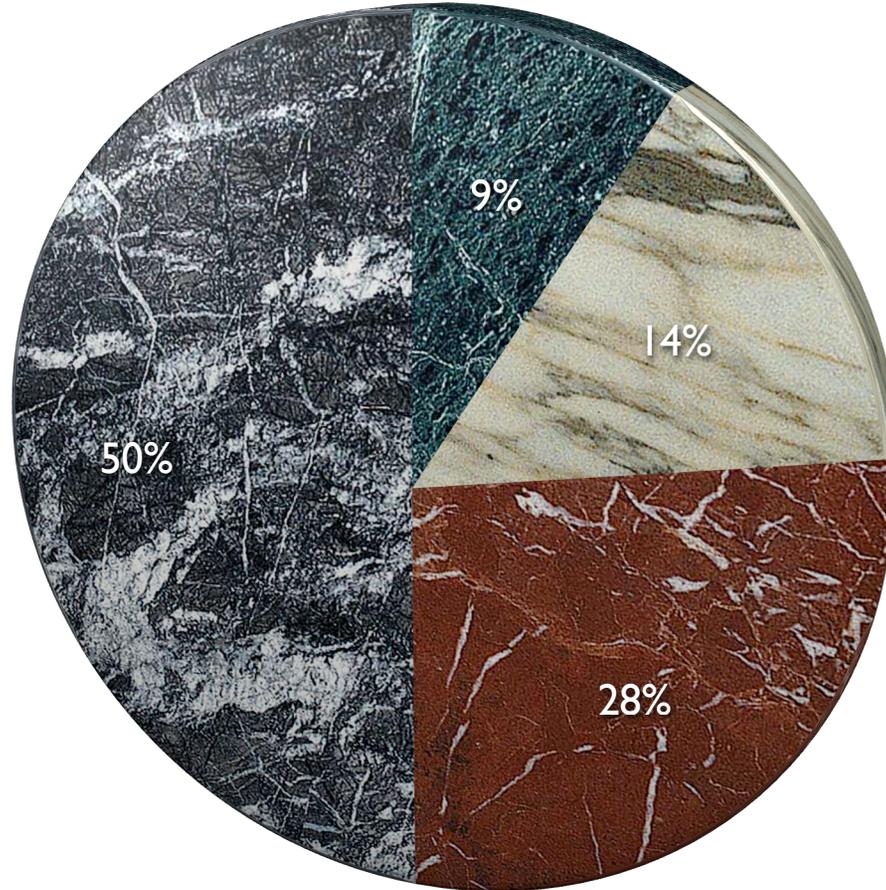


● 2007

● 2008

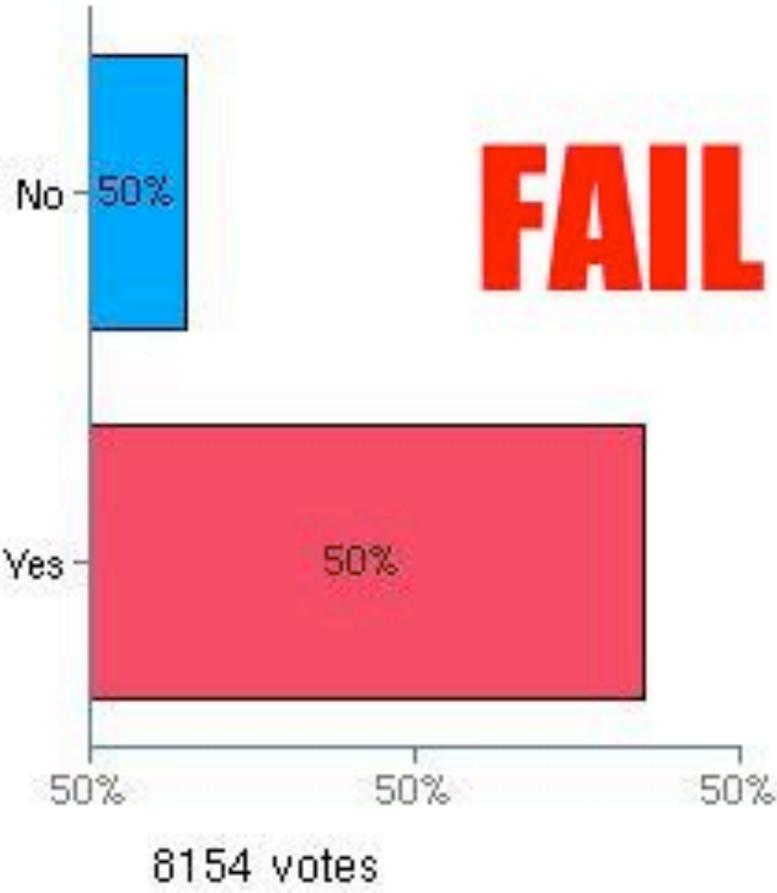
● 2009

● 2010



# Obama's Nobel Prize

Does U.S. President Barack Obama deserve to have won the 2009 Nobel Peace Prize?



This is a non-scientific user poll. Results are not statistically valid and cannot be assumed to reflect the views of Washington Post users as a

# A Complex Example

- Help desk performance
- Time to resolve == unhappy customers, unhappy partners
- Customer satisfaction?

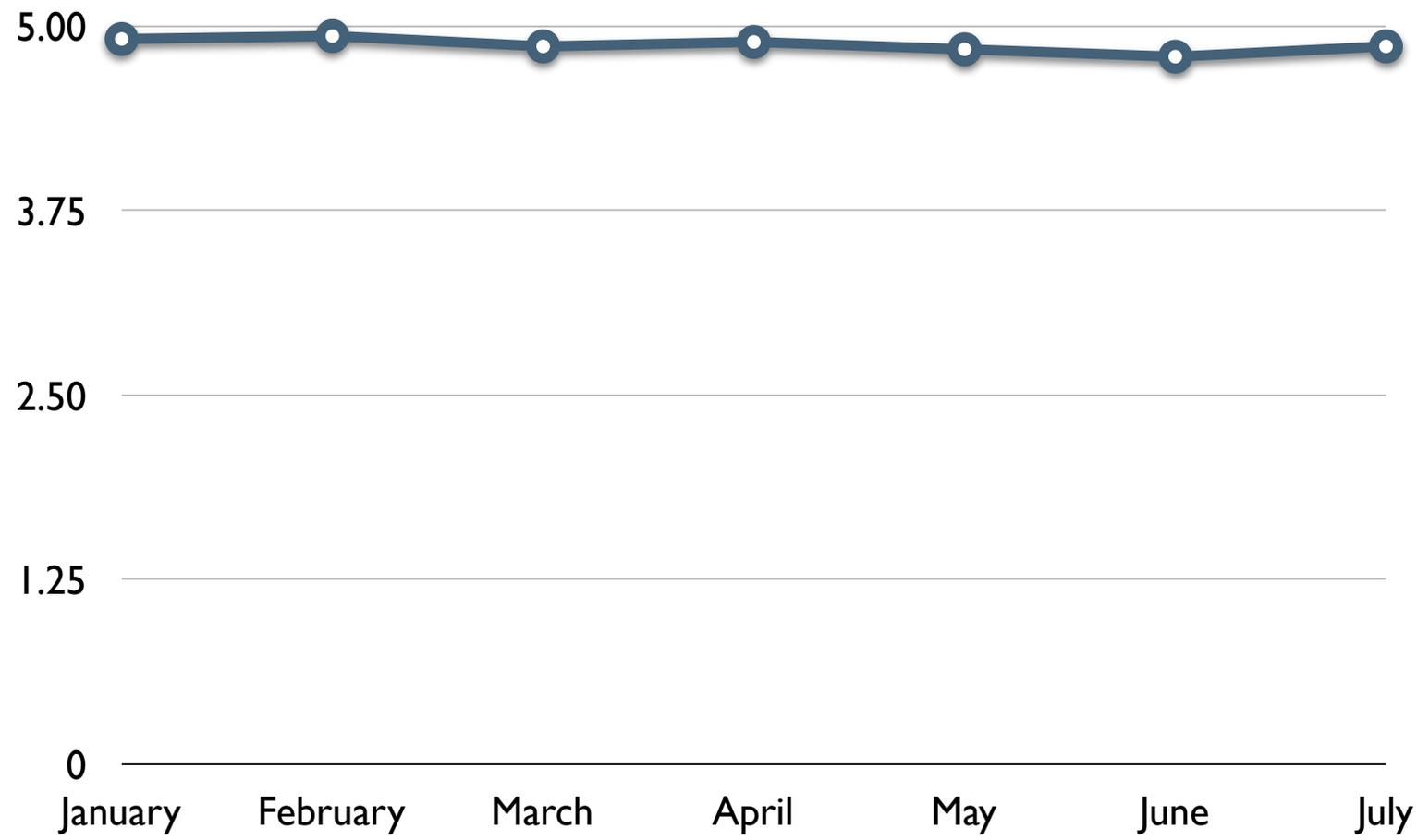
# Customer Satisfaction

- Self-selected sample
- People who are especially unhappy or happy
- People who follow instructions

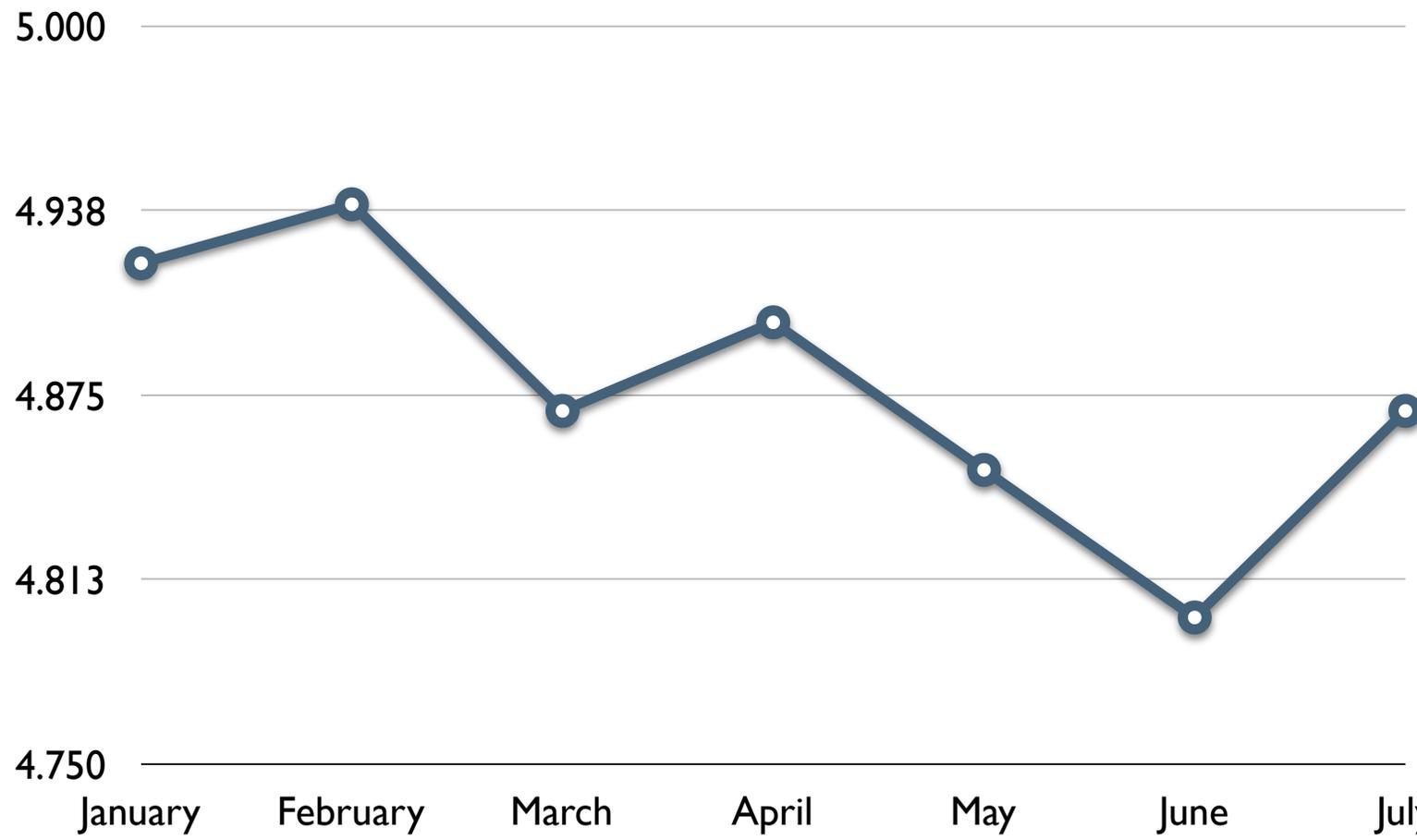
# The Problem

- Help desk operators say users are unhappy
- Help desk management looks at numbers, says there's no problem

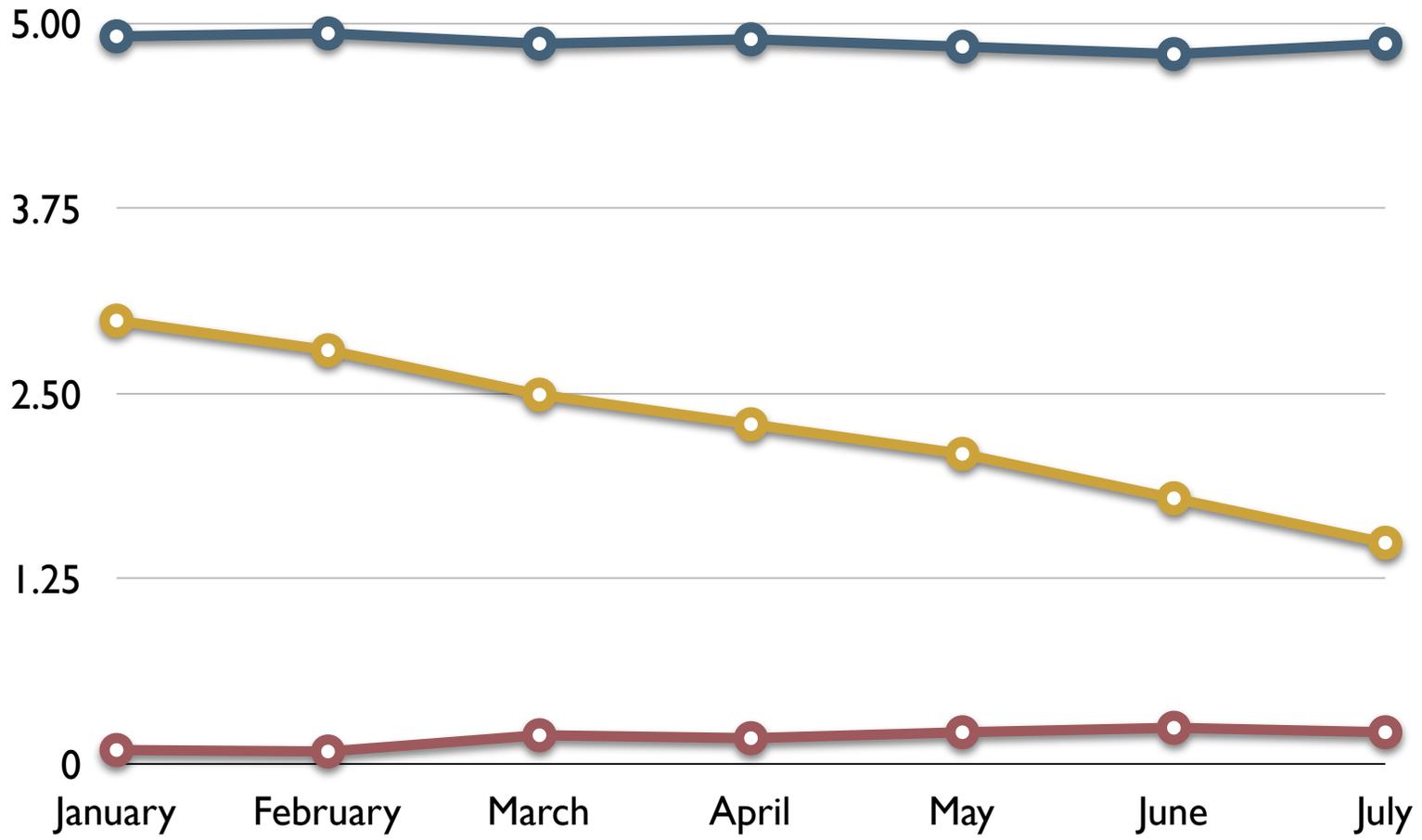
○ Customer Satisfaction



○ Customer Satisfaction



○ Customer Satisfaction      ○ Percent Is      ○ Engineering



# Most Relevant Books

- *Automating System Administration with Perl* by David Blank-Edelman
- *Visualizing Data* by Ben Fry
- *Data Crunching* by Greg Wilson

# Classics

- *How to Lie With Statistics* by Darrell Huff
- *The Visual Display of Quantitative Information*  
by Edward Tufte

# Background

- *Head First Statistics* by Dawn Griffiths
- *Predictably Irrational* by Dan Ariely
- *The Logic of Failure* by Dietrich Dörner

# Blogs about data

- Junk charts: [http://junkcharts.typepad.com/junk\\_charts/](http://junkcharts.typepad.com/junk_charts/)
- Chris Jordan: <http://www.chrisjordan.com>
- Chart Porn: <http://chartporn.org/>

# Blogs that think this way

- Cognitive Daily: <http://scienceblogs.com/cognitivedaily/>
- Language Log: <http://languagelog.idc.upenn.edu>
- Bad Science: <http://www.badscience.net/>

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