A/B tests and Online Controlled Experiments: Introduction, Insights, Scaling, and Humbling Statistics

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Most software changes are believed to be positive to the user experience, but are often flat or negative!

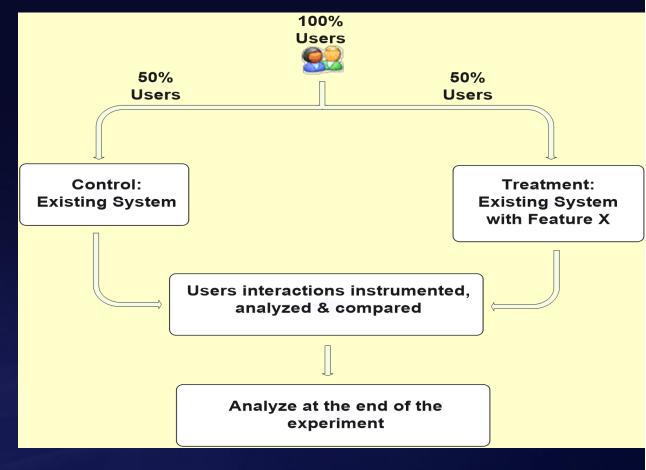
Once you objectively evaluate changes, you're often humbled

Agenda

- Controlled experiments and observational studies
- Examples: you're the decision maker
- Running experiments at scale and best practices
- The cultural challenge
- Two key messages to remember
 - It is hard to assess the value of ideas.
 Get the data by experimenting because data trumps intuition
 - Make sure the org agrees what you are optimizing

Controlled Experiments in One Slide

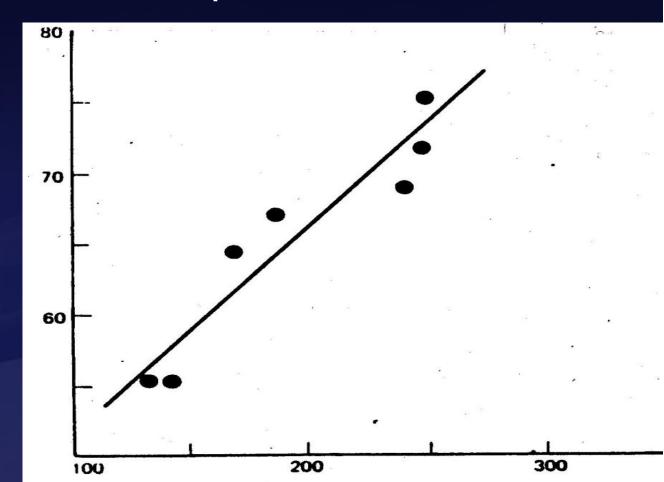
- Concept is trivial
 - Randomly split traffic between two (or more) versions
 - A (Control)
 - B (Treatment)
 - Collect metrics of interest
 - Analyze



- Must run statistical tests to confirm differences are not due to chance
- Best scientific way to prove causality, i.e., the changes in metrics are caused by changes introduced in the treatment(s)

Typical Discovery

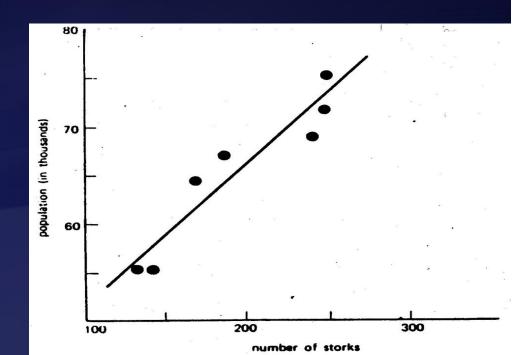
- With data mining, we find patterns, but most are correlational, providing hypotheses for possible causes
- Here is one a real example of two highly correlated variables



Correlations are not Necessarily Causal

- Real Data for the city of Oldenburg, Germany
 - X-axis: stork population
 - Y-axis: human population
- What your mother told you about babies and storks when you were three is not correct, despite the strong correlational "evidence"
- Killing the storks won't solve population growth problems





Personalized Correlated Recommendations

Actual personalized recommendations from Amazon.
 (I was director of data mining and personalization at Amazon back in

2003, so I can ridicule my work.)

Buy a 30" monitor because you bought a DisplayPort cable

Buy Atonement movie DVD because you bought a Maglite flashlight

Buy Organic Virgin Olive Oil because you bought Toilet Paper



Dell UltraSharp U3011 30" Monitor

Average Customer Review: ******

(125)

by Dell (September 17, 2010)

List Price: \$1,299.99 Price: \$1.099.99

In Stock



Advantage of Controlled Experiments

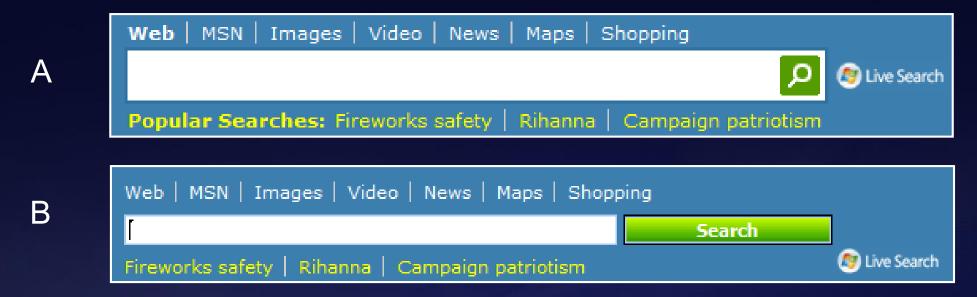
- Controlled experiments test for causal relationships, not simply correlations
- When the variants run concurrently, only two things could explain a change in metrics:
 - 1. The "feature(s)" (A vs. B)
 - 2. Random chance But of course: 3. Mistake somewhere. RB
 - Everything else happening affects both the variants
 - For #2, we conduct statistical tests for significance ("Student's t-test")
- The gold standard in science and the only way to prove efficacy of drugs in FDA drug tests
- Controlled experiments are not the panacea for everything. Issues discussed in the journal <u>survey paper</u>

Examples

- Three experiments that ran at Microsoft
- Each helps share interesting lessons
- All had enough users for statistical validity
- Game: see how many you get right
 - Everyone please stand up
 - Three choices are:
 - A wins (the difference is statistically significant)
 - A and B are approximately the same (no stat sig diff, < 2% delta)</p>
 - B wins

MSN Home Page Search Box

OEC: Clickthrough rate for Search box and popular searches



Differences: A has taller search box (overall size is the same), has magnifying glass icon, "popular searches"

B has big search button

- Raise your left hand if you think A Wins
- Raise your right hand if you think B Wins
- Don't raise your hand if they are the about the same

Search Box

- <deleted>
- Insights
 - Stop debating, it's easier to get the data
 - Most people are overly confident that their idea will work. How confident were you? Reality: most ideas fail to deliver (statistics in later slides)
 - To get insights try OFAT: One Factor At a Time. Don't tweak too many things at once. (But be careful not to fall into Incrementalism)

MSN US Home Page: Search Box

- A later test showed that changing the magnifying glass to an actionable word (search, go, explore) was highly beneficial.
- This:



is better than



In line with Steve Krug's great book: Don't Make Me Think

Bing Ads with Site Links

- Should Bing add "site links" to ads, which allow advertisers to offer several destinations on ads?
- OEC: Revenue, ads constraint to same vertical pixels on avg

Esurance® Auto Insurance - You Could Save 28% with Esurance.
www.esurance.com/California
Get Your Free Online Quote Today!

Esurance® Auto Insurance - You Could Save 28% with Esurance.
www.esurance.com/California
Get Your Free Online Quote Today!
Get a Quote · Find Discounts · An Allstate Company · Compare Rates

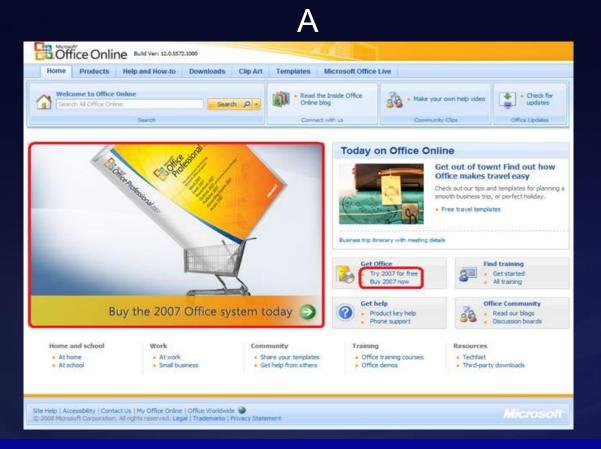
A

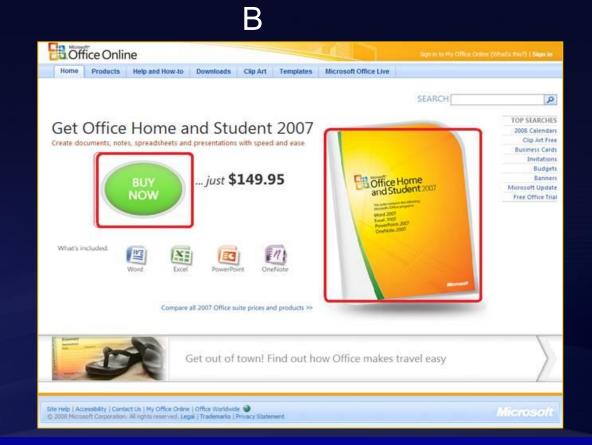
В

- Pro: richer ads, users better informed where they land
- Cons: Constraint means on average 4 "A" ads vs. 3 "B" ads
 Variant B is 5msc slower (compute + higher page weight)
- Raise your Left hand if you think A Wins
- Raise your Right hand if you think B Wins
- Don't raise your hand if you think they're about the same

Office Online

OEC: Clicks on revenue generating links (red below)





- Raise your left hand if you think A Wins
- Raise your right hand if you think B Wins
- Don't raise your hand if they are the about the same

Twyman's Law

Any figure that looks interesting or different is usually wrong

- If something is "amazing," find the flaw!
- Examples
 - If you have a mandatory birth date field and people think it's unnecessary, you'll find lots of 11/11/11 or 01/01/01
 - If you have an optional drop down, do not default to the first alphabetical entry, or you'll have lots of: jobs = Astronaut
 - For most web sites, traffic will be lower 2AM-3AM March 9, 2014, relative to the same hour a week prior. Why?
- Previous Office Example
- More at http://bitly.com/twymanLaw

Hard to Assess the Value of Ideas: Data Trumps Intuition

- Features are built because teams believe they are useful. But most experiments show that features fail to move the metrics they were designed to improve
- We joke that our job is to tell clients that their new baby is ugly
- In Uncontrolled, Jim Manzi writes Google ran ...randomized experiments... with [only] about 10 percent of these leading to business changes.
- In an Experimentation and Testing Primer by Avinash Kaushik, authors of Web Analytics: An Hour a Day, he wrote

80% of the time you/we are wrong about what a customer wants

Hard to Assess the Value of Ideas: Data Trumps Intuition

- Based on experiments at Microsoft (paper)
 - 1/3 of ideas were positive ideas and statistically significant
 - 1/3 of ideas were flat: no statistically significant difference
 - 1/3 of ideas were negative and statistically significant
- Our intuition is poor: 60-90% of ideas do not improve the metric(s) they were designed to improve (domain dependent). Humbling!

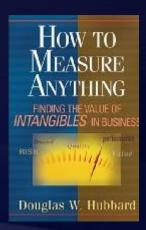
Key Lessons

- Avoid the temptation to try and build optimal features through extensive planning without early testing of ideas
- Experiment often
 - To have a great idea, have a lot of them -- Thomas Edison
 - If you have to kiss a lot of frogs to find a prince, find more frogs and kiss them faster and faster
 - -- Mike Moran, Do it Wrong Quickly
- Try radical ideas. You may be surprised
 - Doubly true if it's cheap to implement (e.g., shopping cart recommendations)
 - If you're not prepared to be wrong, you'll never come up with anything original – Sir Ken Robinson, TED 2006 (#1 TED talk)



The OEC

- If you remember one thing from this talk, remember this point
- OEC = Overall Evaluation Criterion
 - Agree early on what you are optimizing
 - Getting agreement on the OEC in the org is a huge step forward
 - Suggestion: optimize for customer lifetime value, not immediate short-term revenue
 - Criterion could be weighted sum of factors, such as
 - Time on site (per time period, say week or month)
 - Visit frequency
 - Report many other metrics for diagnostics, i.e., to understand the why the OEC changed and raise new hypotheses



OEC for Search

- KDD 2012 paper (*)
- Search engines (Bing, Google) are evaluated on query share (distinct queries) and revenue as long-term goals
- Puzzle
 - A ranking bug in an experiment resulted in very poor search results
 - Distinct queries went up over 10%, and revenue went up over 30%
 - What metrics should be in the OEC for a search engine?
- Degraded (algorithmic) search results cause users to search more to complete their task, and ads appear more relevant

Puzzle Explained

Analyzing queries per month, we have

$$\frac{Queries}{Month} = \frac{Queries}{Session} \times \frac{Sessions}{User} \times \frac{Users}{Month}$$

where a session begins with a query and ends with 30-minutes of inactivity. (Ideally, we would look at tasks, not sessions).

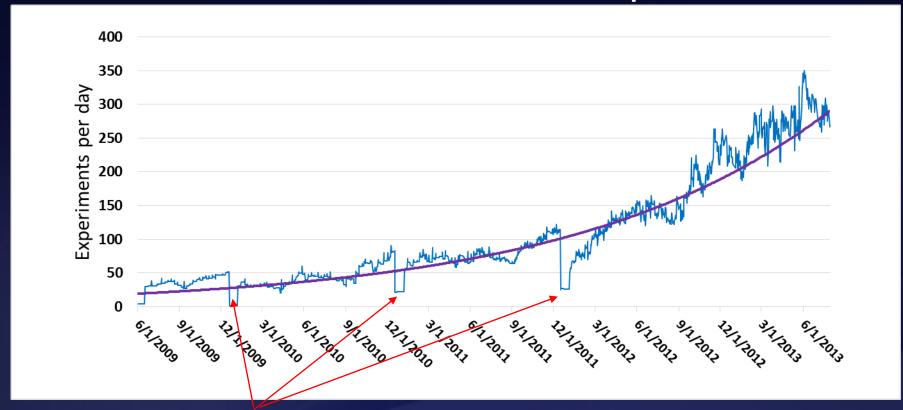
- Key observation: we want users to find answers and complete tasks quickly, so queries/session should be smaller
- In a controlled experiment, the variants get (approximately) the same number of users by design, so the last term is about equal
- The OEC should therefore include the middle term: sessions/user

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Scaling Experiments at Bing

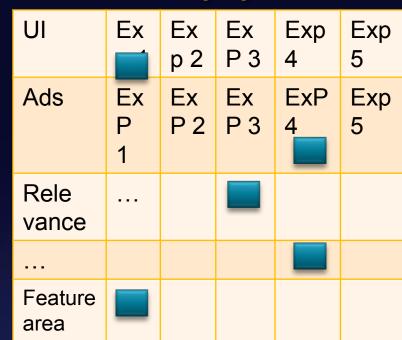
- KDD 2013 paper to appear: http://bit.ly/ExPScale
- We now run over 250 concurrent experiments at Bing



We used to lockdown for Dec holidays. No more

Running Controlled Experiments at Scale (1)

- In a visit, you're in about 15 experiments
 - There is no single Bing.
 There are 30B variants (5^15)
 - 90% of users are in experiments.10% are kept as holdout
- Sensitivity: we need to detect small effects
 - 0.1% change in the revenue/user metric > \$1M/year
 - Not uncommon to see unintended revenue impact of +/-1% (>\$10M)
 - Sessions/UU, a key component of our OEC, is hard to move, so we're looking for small effects
 - Important experiments run on 10-20% of users



Running Controlled Experiments at Scale (2)

Challenges

- QA. You can't QA all combinations, of course.
 What are the equivalence classes?
 For UI change, no need to QA combinations of relevance exps
- Alarming on anomalies is critical: notify experiment owners that there's a big delta on metric M (100 metrics) for browser B
- Interactions (optimistic experimentation): everyone experiments.
 Run statistical tests for pairwise interactions, and notify owners.
- Carryover effects: reuse of "bucket of users" from one experiment to the next is problematic

Important Lesson: Performance

- Bing server time is under one second at the 95th percentile
- Is it worth improving?
- We ran slowdown experiments to see the impact: we introduce an artificial server delay
- Performance matters a LOT. Here's the summary:

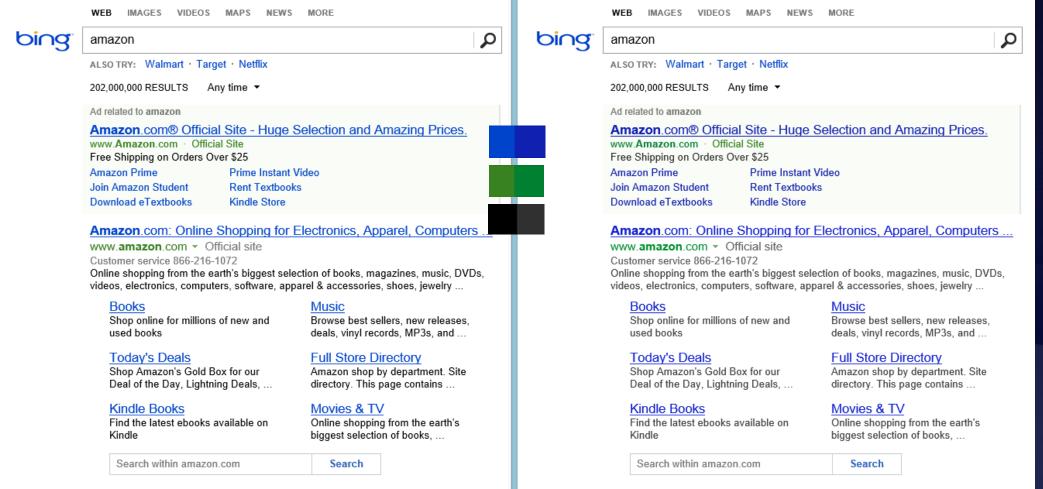
An engineer that improves server performance by 10msec (that's 1/30 of the speed that our eyes blink) more than pays for his fully-loaded annual costs

Every millisecond counts

Lesson: Small Changes can have High ROI

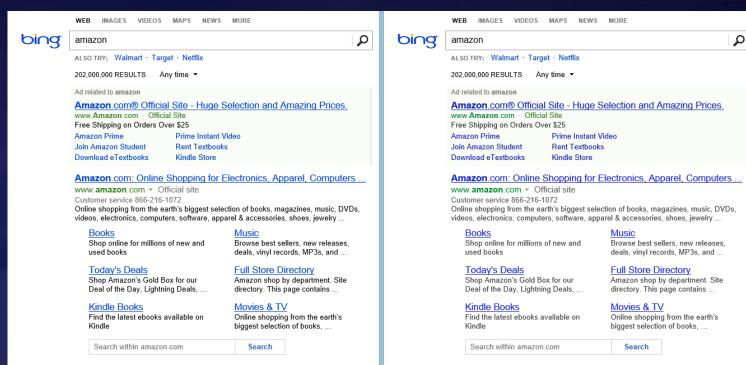
We made small changes to font colors in August 2013

Can you see? Can you figure out which is better?



Lesson: Small Changes (2)

- The change was from the left version to the right version
- Users were more successful in their tasks (SSR)
- Users completed tasks faster (time-to-success)
- We made more money (over \$10M annually)
- Companies set standard company color/fonts without appreciating the impact it can have



Best Practice: A/A Test

- Run A/A tests simple, but highly effective
 - Run an experiment where the Treatment and Control variants are coded identically and validate the following:
 - 1. Are users split according to the planned percentages?
 - 2. Is the data collected matching the system of record?
 - 3. Are the results showing non-significant results 95% of the time?

This is a powerful technique for finding problems

- Generating some numbers is easy
- Getting correct numbers you trust is much harder!

Remove Bots for Analysis

- Bots are lucrative business, but they skew the statistics
- At Bing, >50% of traffic comes from bots

Actual picture I took



Best Practice: Ramp-up

- Ramp-up
 - Start an experiment at 0.1%
 - Do some simple analyses to make sure no egregious problems can be detected
 - Ramp-up to a larger percentage, and repeat until 50%
- Big differences are easy to detect because the min sample size is quadratic in the effect we want to detect
 - Detecting 10% difference requires a small sample and serious problems can be detected during ramp-up
 - Detecting 0.1% requires a population 100^2 = 10,000 times bigger
- Abort the experiment if treatment is significantly worse on key metrics



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The Cultural Challenge

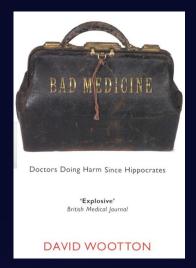
It is difficult to get a man to understand something when his salary depends upon his not understanding it.

-- Upton Sinclair

- Why people/orgs avoid controlled experiments
 - Some believe it threatens their job as decision makers
 - At Microsoft, program managers select the next set of features to develop. Proposing several alternatives and admitting you don't know which is best is hard
 - Editors and designers get paid to select a great design
 - Failures of ideas may hurt image and professional standing. It's easier to declare <u>success</u> when the feature launches
 - We've heard: "we know what to do. It's in our DNA," and "why don't we just do the right thing?"

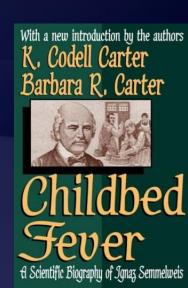
Cultural Stage 1: Hubris

- The org goes through stages in its cultural evolution
- Stage 1: we know what to do and we're sure of it
 - True story from 1849
 - John Snow claimed that Cholera was caused by polluted water
 - A landlord dismissed his tenants' complaints that their water stank
 - Even when Cholera was frequent among the tenants
 - One day he drank a glass of his tenants' water to show there was nothing wrong with it
- He died three days later
- That's hubris. Even if we're sure of our ideas, evaluate them
- Controlled experiments are a powerful tool to evaluate ideas



Cultural Stage 2: Insight through Measurement and Control

- Semmelweis worked at Vienna's General Hospital, an important teaching/research hospital, in the 1830s-40s
- In 19th-century Europe, childbed fever killed more than a million women
- Measurement: the mortality rate for women giving birth was
 - 15% in his ward, staffed by doctors and students
 - 2% in the ward at the hospital, attended by midwives



Cultural Stage 2: Insight through Measurement and Control

- He tries to control all differences
 - Birthing positions, ventilation, diet, even the way laundry was done
- He was away for 4 months and death rate fell significantly when he was away. Could it be related to him?
- Insight:
 - Doctors were performing autopsies each morning on cadavers
 - Conjecture: particles (called germs today) were being transmitted to healthy patients on the hands of the physicians
- He experiments with cleansing agents
 - Chlorine and lime was effective: death rate fell from 18% to 1%

Cultural Stage 3: Semmelweis Reflex

- Success? No! Disbelief. Where/what are these particles?
 - Semmelweis was dropped from his post at the hospital
 - He goes to Hungary and reduced mortality rate in obstetrics to 0.85%
 - His student published a paper about the success. The editor wrote We believe that this chlorine-washing theory has long outlived its usefulness... It is time we are no longer to be deceived by this theory
- In 1865, he suffered a nervous breakdown and was beaten at a mental hospital, where he died
- Semmelweis Reflex is a reflex-like rejection of new knowledge because it contradicts entrenched norms, beliefs or paradigms
- Only in 1800s? No! A 2005 study: inadequate hand washing is one of the prime contributors to the 2 million health-care-associated infections and 90,000 related deaths annually in the United States

Cultural Stage 4: Fundamental Understanding

- In 1879, Louis Pasteur showed the presence of Streptococcus in the blood of women with child fever
- 2008, 143 years after he died, there is a 50 Euro coin commemorating Semmelweis



True Story – Scurvy and Vitamin C

- Without fundamental understanding, you make mistakes
- Scurvy is a disease that results from vitamin C deficiency
- It killed over 100,000 people in the 16th-18th centuries, mostly sailors
- First known controlled experiment in 1747
 - Dr. James Lind noticed lack of scurvy in Mediterranean ships
 - Gave some sailors limes (treatment), others ate regular diet (control)
 - Experiment was so successful, British sailors are still called limeys
- But Lind didn't understand the reason
 - At the Royal Naval Hospital in England, he treated Scurvy patients with concentrated lemon juice called "rob."
 - He concentrated the lemon juice by heating it, thus destroying the vitamin C
 - He lost faith in the remedy and became increasingly reliant on bloodletting
- In 1793, a formal trial was done and lemon juice became part of the daily rations throughout the navy; Scurvy was quickly eliminated

Summary: Evolve the Culture



- In many areas we're in the 1800s in terms of our understanding, so controlled experiments can help
 - First in doing the right thing, even if we don't understand the fundamentals
 - Then developing the underlying fundamental theories

Summary

The less data, the stronger the opinions

1. Empower the HiPPO with data-driven decisions

- HiPPO = Highest Paid-Person in Org, or Highest Paid-Person's Opinion
- Hippos kill more humans than any other (non-human) mammal (really)
- OEC: make sure the org agrees what you are optimizing (long term lifetime value)

2. It is hard to assess the value of ideas

- Listen to your customers Get the data
- Prepare to be humbled: data trumps intuition

3. Compute the statistics carefully

Getting a number is easy. Getting a number you should trust is harder

4. Experiment often to accelerate innovation

Triple your experiment rate and you triple your success (and failure) rate.
Fail fast & often in order to succeed

Resources and Q&A

- This talk: http://bit.ly/expQCon
- http://exp-platform.com has papers, talks including
 - Controlled Experiments on the Web: Survey and Practical Guide (Data Mining and Knowledge Discovery journal)
 - Online experiments at Microsoft
 (Third Workshop on Data Mining Case Studies and Practice Prize)
 - Trustworthy Online Controlled Experiments: Five Puzzling Outcomes Explained (KDD 2012)
 - Online Controlled Experiments at Large Scale (KDD 2013)
- Nice Etsy talk: http://www.slideshare.net/danmckinley/design-for-continuous-experimentation