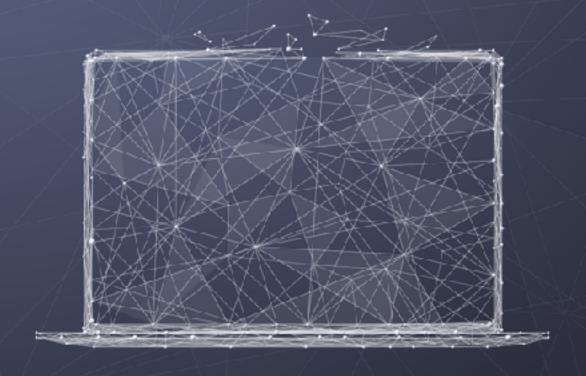
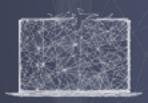
# Data Science Foundations of Decision Making

**AB** testing



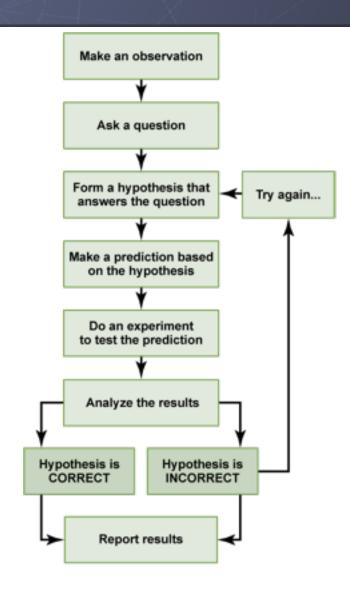
PURDUE UNIVERSITY

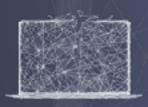
College of Science



# Conjecture and test

- The best data scientists are skeptics
- "If a statistic/figure looks interesting or unusual it is probably wrong." Twyman's Law
- If you don't formulate a conjecture about your data/process/model and then test your idea, you will make mistakes
- Example: Bing experimentation platform for online A/B testing

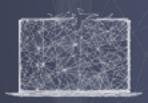




# Testing conjectures in data science

- Making a claim about discovered pattern or estimated model?
  - What population are you generalizing to?
- Making a claim about a model/algorithm within a single domain?
  - Do you want to predict model/algorithm accuracy or choose between methods?
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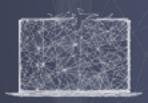
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#### Artificial intelligence faces reproducibility crisis

#### Matthew Hutson

See all authors and affiliations.

Science 16 Feb 2018; Vol. 359, Issue 6377, pp. 725-726 DOI: 10.1126/science.359.6377.725



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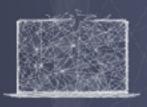
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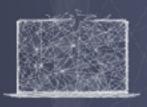
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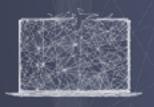
- Microsoft team proposed to change the way ad titles were displayed on Bing (2012)
  - One of hundreds of ideas proposed, other features were ranked as more valuable
  - Implementation was delayed for >6 months
  - Engineer decided it was trivial to implement in a few days, so started a controlled experiment (A/B test) to evaluate





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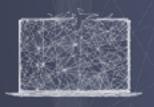




#### Result

- When running A/B test, a system alert fired that Bing was making too much money... the idea increased Bing's revenue by 12% without hurting other metrics
- Hundreds of engineers work on Bing Ads and increase revenue by 1.5% in a good month. Thus, simple change to titles was worth the equivalent of over 100 person years of work
- Takeaway: We are terrible at assessing the value of ideas. The best revenuegenerating idea in Bing's history was badly rated and delayed for months!

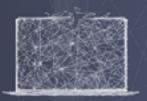




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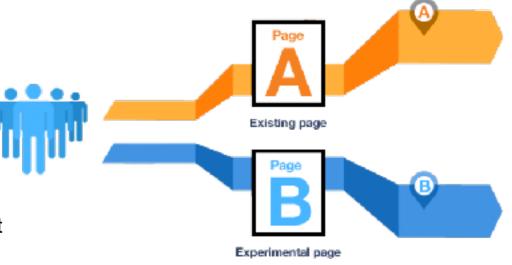
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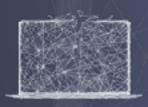




# A/B Testing

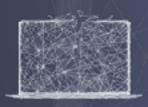
- Randomly split traffic between two (or more) versions:
  - A (Control) vs. B (Treatment)
  - Collect metrics of interest, and analyze
- A/B test is the simplest controlled experiment,
  - A/B/n refers to multiple treatments
- Must run statistical tests to confirm differences are not due to chance
- Best scientific way to prove causality, i.e., the changes in metrics are caused by changes introduced in the treatment(s)





- Features are built because teams believe they are useful. But most experiments show that features fail to improve metrics they were designed for
- Experiments at Microsoft shows that only 1/3 of ideas improve performance and 1/3 actually decrease performance
- Bing success rate is lower. The low success rate has been documented many times across multiple companies



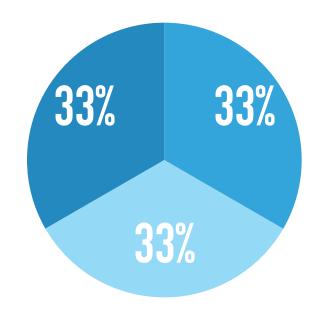


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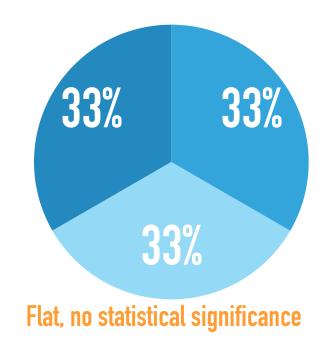
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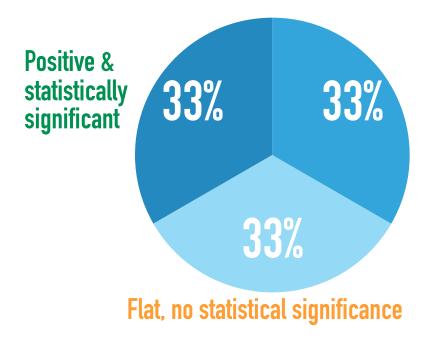
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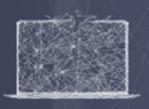




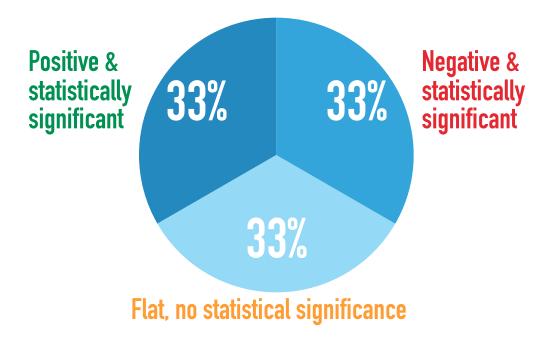
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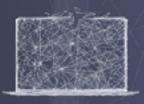




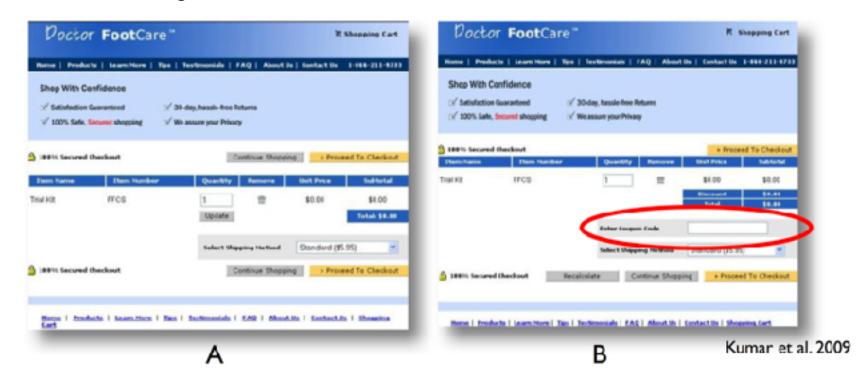
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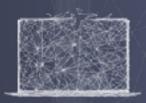




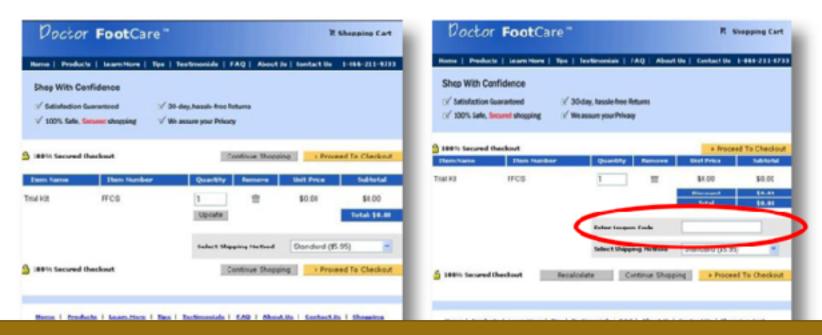


Can you guess which page has a higher conversion rate and whether the difference is significant?

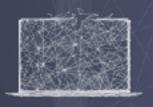




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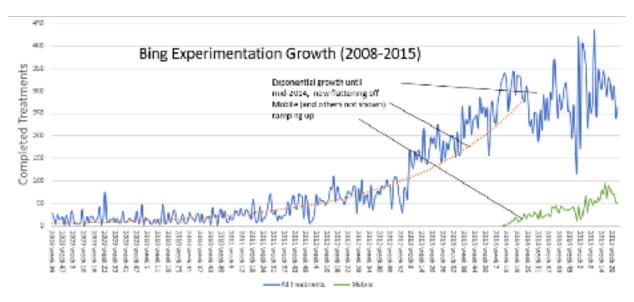


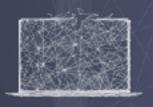
Using version B the site lost 90% of their revenue. Why? "There maybe discount coupons out there that I do not have. The price may be too high..." (Kumar et al. 2009)



#### Experimentation at scale (Ronny Kohavi, Microsoft Research)

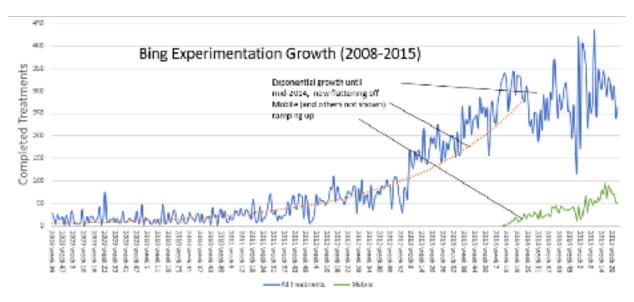
- ~300 experiment treatments are completed at Bing every week
- Each variant is exposed to between 100K and 10M users
- 90% of eligible users are in experiments (10% are a global holdout changed once a year)
- There is no single Bing. Since a user is exposed to 15 concurrent experiments, they get one of 5<sup>15</sup> = 30 billion variants

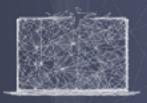




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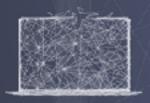




## Advantage of controlled experiments

- Controlled experiments test for causal relationships, not simply correlations
  - The gold standard in science
  - The only way to prove efficacy of drugs in FDA drug tests
- When the variants run concurrently, only two things can explain differences:
  - The "feature(s)" (A vs. B)
  - Random chance
- All other effects are the same in both the conditions
- To control for random chance, statistical tests are used to test for significance

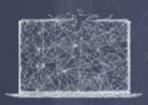




#### First controlled experiment for medical purposes

- Scurvy is a disease that results from vitamin C deficiency
- Killed over 100,000 people in the 16th-18th centuries, mostly sailors
  - E.g., Lord Anson's circumnavigation voyage from 1740 to 1744 started with 1,800 sailors and only about 200 returned; most died from scurvy
- 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

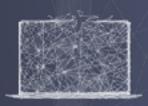




# Lind's experimental details

- Lind's hypothesis was that scurvy was due to putrefaction of the body which could be helped by acids
  - The experiment was done on 12 sailors split into 6 pairs
  - Each pair got a different treatment: cider, elixir vitriol, vinegar, sea-water, nutmeg+barley water, oranges+lemon
  - The sailors given two oranges and one lemon per day and recovered
- Lind didn't understand the reason and tried treating Scurvy with concentrated lemon juice called "rob." But the lemon juice was concentrated by heating it, which destroyed the vitamin C.

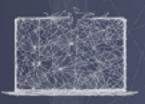




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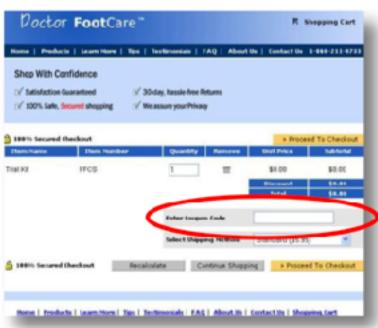
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**Lesson:** Even when you find a significant effect, the reasons are often not understood. Controlled experiments tell you which variant won, not why.



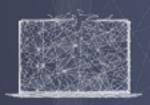
- Sample 5000 customers for each treatment
- Measure number of shopping carts that are "converted" to purchases in conditions A and B





В

Kumar et al. 2009



	Not converted	Converted
Α	4461	539
В	4522	478

from scipy.stats import chi2\_contingency

obs = np.array([[4461,539],[4522,478]])
chi2\_contingency(obs)[:2]

(3.9405799942664563, 0.047134524006671369)





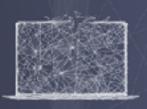
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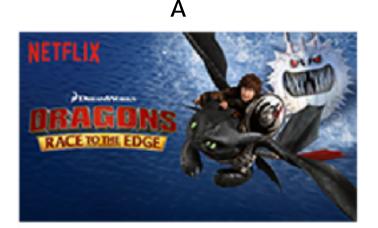
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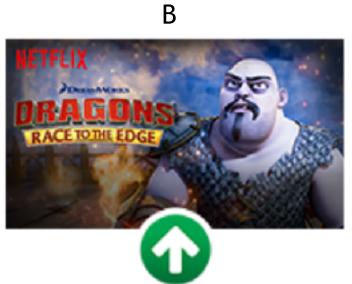
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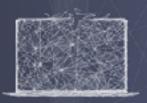
Pvalue <0.05 so conclude effect is significant, but in this case conversions(B) < conversions(A) so impact is negative



- Sample 100,000 users for each treatment
- Measure streaming hours per user in each treatment



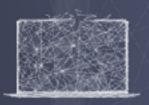




	Mean	Std
Α	6.730	2.5
В	6.762	2.5

```
from scipy.stats import ttest_ind
# generate pseudo data randomly
dA = np.random.normal(loc=6.730, scale=2.5, size=100000)
dB = np.random.normal(loc=6.762, scale=2.5, size=100000)
ttest_ind(dA, dB)
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