

Lecture 1 - Introduction

Sta 102 / BME 102

May 16, 2016

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Course Details

Course goals & objectives

1. Recognize the importance of data collection, identify limitations in data collection methods, and determine how they affect the scope of inference.
2. Use statistical software to summarize data numerically and visually, and to perform data analysis.
3. Have a conceptual understanding of the unified nature of statistical inference.
4. Apply estimation and testing methods to analyze single variables or the relationship between two variables in order to understand natural phenomena and make data-based decisions.
5. Model numerical response variables using a single explanatory variable or multiple explanatory variables in order to investigate relationships between variables.
6. Interpret results correctly, effectively, and in context without relying on statistical jargon.
7. Critique data-based claims and evaluate data-based decisions.
8. Complete an independent research project employing what you learn in this class.

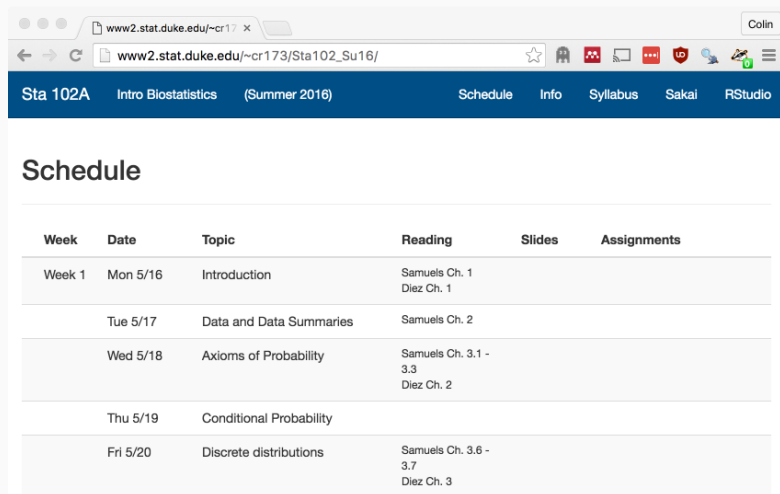
Major topics

- *Introduction to data*: Observational studies and non-causal inference, principles of experimental design and causal inference, exploratory data analysis: description, summary and visualization.
- *Probability and distributions*: The basics of probability and chance processes, Bayesian perspective in statistical inference, the normal distribution.
- *Framework for inference*: Central Limit Theorem and sampling distributions
- *Statistical inference*: Univariate and bivariate analyses for numeric and categorical data, decision errors, power.
- *Simple linear regression*: Bivariate correlation and causality, introduction to modeling.
- *Multiple regression*: Multiple regression, logistic regression.

Course materials

- Statistics for the Life Sciences - Samuels, Witmer, Schaffner
Pearson, 4th Edition, 2012 (ISBN: 9780321652805)
- OpenIntro Statistics - Diez, Barr, Çetinkaya-Rundel
CreateSpace, 3rd Edition, 2015 (ISBN: 194345003X)
- Calculator (\sqrt{x} , $\log(x)$, e^x)

Announcements, slides, assignments, etc. will be posted on course website:



www2.stat.duke.edu/~cr173/Sta102_Su16/

Sta 102A Intro Biostatistics (Summer 2016) Schedule Info Syllabus Sakai RStudio

Schedule

Week	Date	Topic	Reading	Slides	Assignments
Week 1	Mon 5/16	Introduction	Samuels Ch. 1 Diez Ch. 1		
	Tue 5/17	Data and Data Summaries	Samuels Ch. 2		
	Wed 5/18	Axioms of Probability	Samuels Ch. 3.1 - 3.3 Diez Ch. 2		
	Thu 5/19	Conditional Probability			
	Fri 5/20	Discrete distributions	Samuels Ch. 3.6 - 3.7 Diez Ch. 3		

Homework

Goal of the homework is for you develop a more in-depth understanding of the material and help you prepare for exams and the project.

- Questions from the textbooks and outside sources. (Full questions will be downloadable as a PDF from course website)
- Due at the beginning of class on the due date.
- 8 homeworks planned - lowest score will be dropped.
- Show all your work to receive credit.

Goal of the labs is for you to have hands on experience with data analysis using statistical software, provide you with tools for the projects.

- 8 labs planned - lowest score will be dropped.
- Write ups due the following lab session - majority of each lab can be completed in class, turned in via Sakai.
- Submit both Rmd and HTML files.

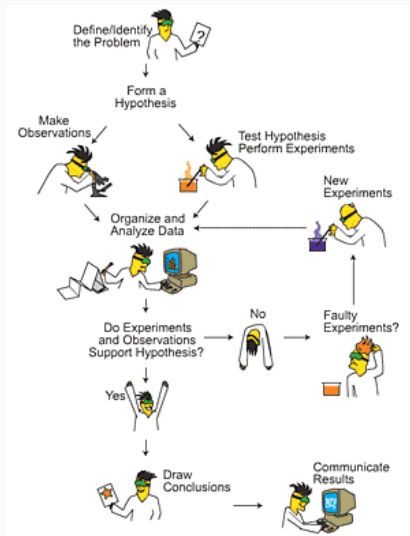
Research Projects

The goal of the project is to give you independent applied research experience using real data

- Open ended research project.
- You find a data set, choose a research question, select relevant data, analyze it, write up your results.
- Multiple stages: proposal, EDA, analysis.

Why (Bio)Statistics

Statistics and the Scientific Method



ANNALS OF SCIENCE

DECEMBER 13, 2010 ISSUE

THE TRUTH WEARS OFF

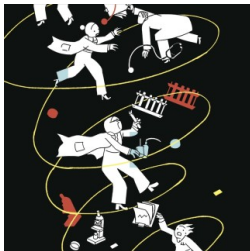
Is there something wrong with the scientific method?

BY JONAH LEHRER

Many results that are rigorously proved and accepted start shrinking in later studies.

ILLUSTRATION BY LAURENT CILLUFFO

On September 18, 2007, a few dozen neuroscientists, psychiatrists, and drug-company executives gathered in a hotel conference room in Brussels to hear some startling news. It had to do with a class of drugs known as atypical or second-generation antipsychotics, which came on the market in the early nineties. The drugs, sold under



Essay

Why Most Published Research Findings Are False

John P. A. Ioannidis

Summary

There is increasing concern that most current published research findings are false. The probability that a research claim is true may depend on study power and bias, the number of other studies on the same question, and, importantly, the ratio of true to no relationships among the relationships probed in each scientific field. In this framework, a research finding is less likely to be true when the studies conducted in a field are smaller; when effect sizes are smaller; when there is a greater number and lesser preselection of tested relationships; when there is greater flexibility in designs, definitions, outcomes, and analytical modes; when there is greater financial and other

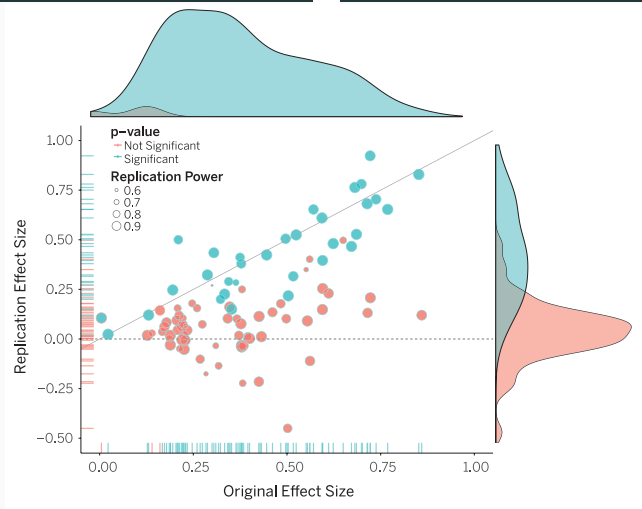
factors that influence this problem and some corollaries thereof.

Modeling the Framework for False Positive Findings

Several methodologists have pointed out [9–11] that the high rate of nonreplication (lack of confirmation) of research discoveries is a consequence of the convenient, yet ill-founded strategy of claiming conclusive research findings solely on the basis of a single study assessed by formal statistical significance, typically for a p -value less than 0.05. Research is not most appropriately represented and summarized by p -values, but, unfortunately, there is a widespread notion that medical research articles

is characteristic of the field and can vary a lot depending on whether the field targets highly likely relationships or searches for only one or a few true relationships among thousands and millions of hypotheses that may be postulated. Let us also consider, for computational simplicity, circumscribed fields where either there is only one true relationship (among many that can be hypothesized) or the power is similar to find any of the several existing true relationships. The pre-study probability of a relationship being true is $R/(R + 1)$. The probability of a study finding a true relationship reflects the power $1 - \beta$ (one minus the Type II error rate). The probability of claiming a relationship when none

Reproducibility Project: Psychology



From Science - <http://science.sciencemag.org/content/349/6251/aac4716>

ASA Statement of p-values

- P-values can indicate how incompatible the data are with a specified statistical model.
- P-values do not measure the probability that the studied hypothesis is true, or the probability that the data were produced by random chance alone.
- Scientific conclusions and business or policy decisions should not be based only on whether a p-value passes a specific threshold.
- Proper inference requires full reporting and transparency.
- A p-value, or statistical significance, does not measure the size of an effect or the importance of a result.
- By itself, a p-value does not provide a good measure of evidence regarding a model or hypothesis.

ON

THE ORIGIN OF SPECIES

BY MEANS OF NATURAL SELECTION,

OR THE

PRESERVATION OF FAVOURED RACES IN THE STRUGGLE
FOR LIFE.

By CHARLES DARWIN, M.A.,

FELLOW OF THE ROYAL, GEOLOGICAL, LINNEAN, ETC., SOCIETIES;
AUTHOR OF 'JOURNAL OF RESEARCHES DURING H. M. S. BEAGLE'S VOYAGE
ROUND THE WORLD.'

INTRODUCTION Page 1

CHAPTER I.

VARIATION UNDER DOMESTICATION.

Causes of Variability — Effects of Habit — Correlation of Growth —
Inheritance — Character of Domestic Varieties — Difficulty of
distinguishing between Varieties and Species — Origin of Domestic
Varieties from one or more Species — Domestic Pigeons, their
Differences and Origin — Principle of Selection anciently followed,
its Effects — Methodical and Unconscious Selection — Unknown
Origin of our Domestic Productions — Circumstances favourable
to Man's power of Selection 7-43

CHAPTER II.

VARIATION UNDER NATURE.

Variability — Individual differences — Doubtful species — Wide
ranging, much diffused, and common species vary most — Spe-
cies of the larger genera in any country vary more than the species
of the smaller genera — Many of the species of the larger genera
resemble varieties in being very closely, but unequally, related
to each other, and in having restricted ranges 44-59

CHAPTER III.

STRUGGLE FOR EXISTENCE.

Bears on natural selection—The term used in a wide sense—Geometrical powers of increase—Rapid increase of naturalised animals and plants—Nature of the checks to increase—Competition universal—Effects of climate—Protection from the number of individuals—Complex relations of all animals and plants throughout nature—Struggle for life most severe between individuals and varieties of the same species; often severe between species of the same genus—The relation of organism to organism the most important of all relations .. Page 60-79

CHAPTER IV.

NATURAL SELECTION.

Natural Selection—its power compared with man's selection—its power on characters of trifling importance—its power at all ages and on both sexes—Sexual Selection—On the generality of intercrosses between individuals of the same species—Circumstances favourable and unfavourable to Natural Selection, namely, intercrossing, isolation, number of individuals—Slow action—Extinction caused by Natural Selection—Divergence of Character, related to the diversity of inhabitants of any small area, and to naturalisation—Action of Natural Selection, through Divergence of Character and Extinction, on the descendants from a common parent—Explains the Grouping of all organic beings 80-130

CHAPTER V.

LAWS OF VARIATION.

Effects of external conditions—Use and disuse, combined with natural selection; organs of flight and of vision—Acclimatisation—Correlation of growth—Compensation and economy of growth—False correlations—Multiple, rudimentary, and lowly organised structures variable—Parts developed in an unusual manner are highly variable: specific characters more variable than generic: secondary sexual characters variable—Species of the same genus vary in an analogous manner—Reversions to long-lost characters—Summary 131-170

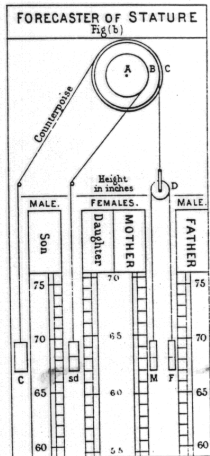
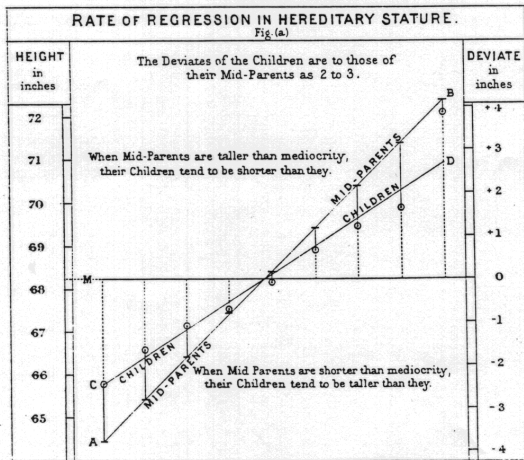
TABLE I.

NUMBER OF ADULT CHILDREN OF VARIOUS STATURES BORN OF 205 MID-PARENTS OF VARIOUS STATURES.
(All Female heights have been multiplied by 1·08).

Heights of the Mid-parents in inches.	Heights of the Adult Children.													Total Number of		Medians.	
	Below	62·2	63·2	64·2	65·2	66·2	67·2	68·2	69·2	70·2	71·2	72·2	73·2	Above	Adult Children.		Mid-parents.
Above	1	3	..	4	5	..
72·5	1	2	1	2	7	2	4	19	6	72·2
71·5	1	3	4	3	5	10	4	9	2	2	43	11	69·9
70·5 ..	1	..	1	..	1	1	3	12	18	14	7	4	3	3	68	22	69·5
69·5	1	16	4	17	27	20	33	25	20	11	4	5	183	41	68·9
68·5 ..	1	..	7	11	16	25	31	34	48	21	18	4	3	..	219	49	68·2
67·5	3	5	14	15	36	38	25	38	19	11	4	211	33	67·6
66·5	3	3	5	2	17	17	14	13	4	78	20	67·2
65·5 ..	1	..	9	5	7	11	11	7	7	5	2	1	66	12	66·7
64·5 ..	1	1	4	4	1	5	5	..	2	23	5	65·8
Below ..	1	..	2	4	1	2	2	1	1	14	1	..
Totals ..	5	7	32	59	48	117	138	120	167	99	64	41	17	14	929	205	..
Medians	66·3	67·8	67·9	67·7	67·9	68·3	68·5	69·0	69·0	70·0

NOTE.—In calculating the Medians, the entries have been taken as referring to the middle of the squares in which they stand. The reason why the headings run 62·2, 63·2, &c., instead of 62·5, 63·5, &c., is that the observations are unequally distributed between 62 and 63, 63 and 64, &c., there being a strong bias in favour of integral inches. After careful consideration, I concluded that the headings, as adopted, best satisfied the conditions. This inequality was not apparent in the case of the Mid-parents.

Plate IX.





“I occasionally meet geneticists who ask me whether it is true that the great geneticist R.A. Fisher was also an important statistician.”
- L.J. Savage (Annals of Statistics, 1976)

Source: http://www.swlearning.com/quant/kohler/stat/biographical_sketches/Fisher_3.jpeg

Biology:

- Heterozygote advantage
- Population genetics (Modern evolutionary synthesis)
- Fisherian runaway selection
- ...

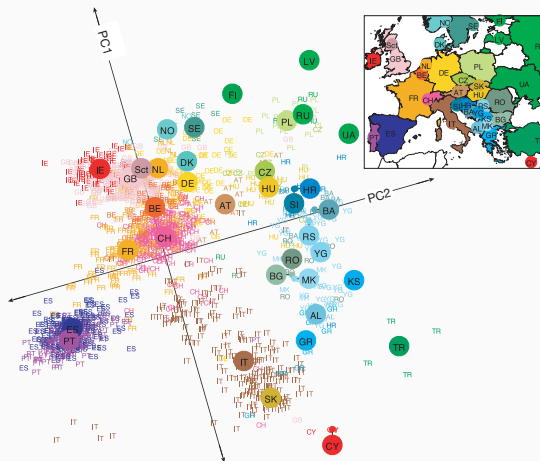
Statistics:

- Analysis of Variance
- Null hypothesis
- Maximum Likelihood
- F distribution
- Fisher's Exact test
- Fisher Information
- Randomization testing
- ...

Runaway Selection

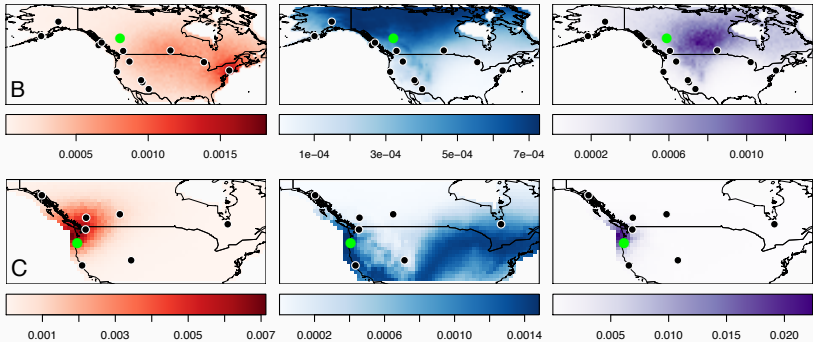


Source: [Irish Elk](#) - [Fiddler Crab](#) - [Peafowl](#)

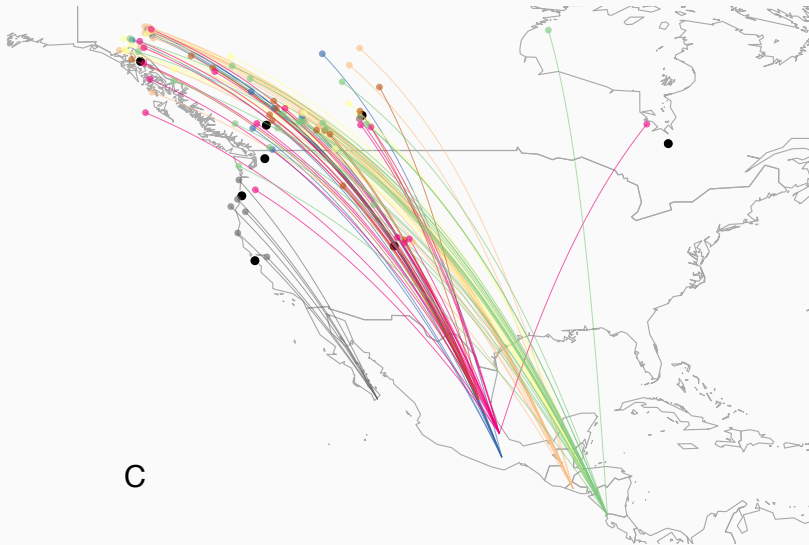


Analysis of 197,146 SNPs in 1,387 Europeans with known family origins

Spatial Mapping



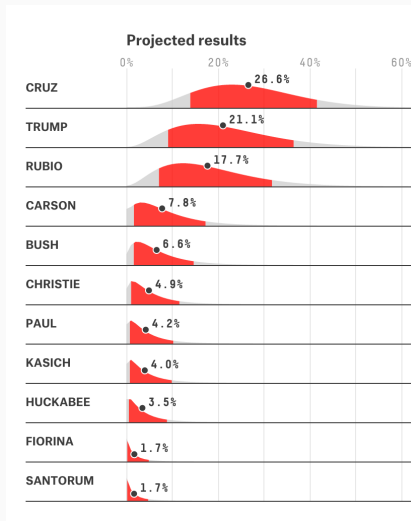
Migratory Connectivity



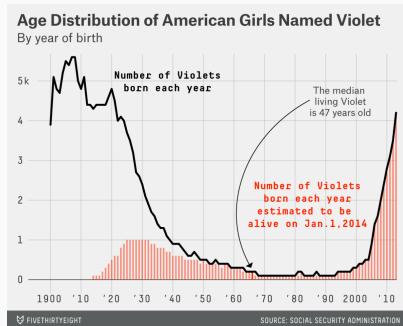
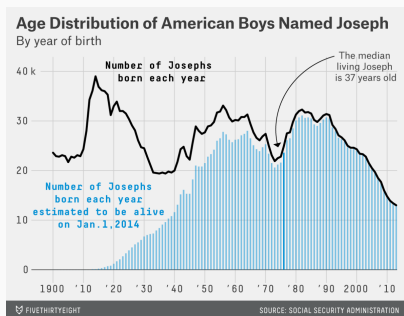
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Other Applications

The most famous statistician in the world ...



538 - How to Tell Someone's Age When All You Know Is Her Name



<http://fivethirtyeight.com/features/>

[how-to-tell-someones-age-when-all-you-know-is-her-name/](http://fivethirtyeight.com/features/how-to-tell-someones-age-when-all-you-know-is-her-name/)

Why you probably shouldn't be playing ...



<http://graphics.latimes.com/powerball-simulator/>

Data collection and study design

Using a sample to make inferences about the population

- Ultimate goal: make inferences about populations

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Suppose we want to know how many offspring female lemurs have, on average. It's not feasible to obtain offspring data from on all female lemurs, so we use data from the Duke Lemur Center. We use the sample mean from these data as an estimate for the unknown population mean. Can you see any limitations to using data from the Duke Lemur Center to make inferences about all lemurs?

Sampling is natural

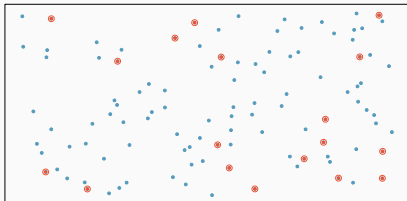


- When you taste a spoonful of soup and decide the spoonful you tasted isn't salty enough, that's *exploratory analysis*
- If you generalize and conclude that your entire soup needs salt, that's an *inference*
- For your inference to be valid, the spoonful you tasted (the sample) needs to be *representative* of the entire pot (the population)

Sampling methods

Simple random:

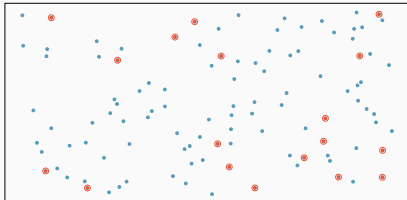
Drawing names from a hat



Sampling methods

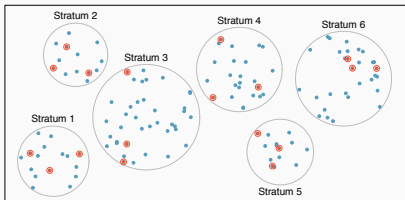
Simple random:

Drawing names from a hat



Stratified: homogenous strata

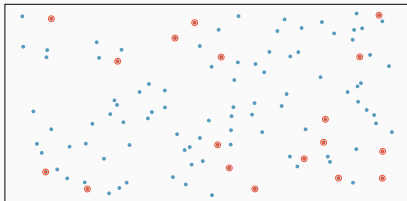
Stratify to control for SES



Sampling methods

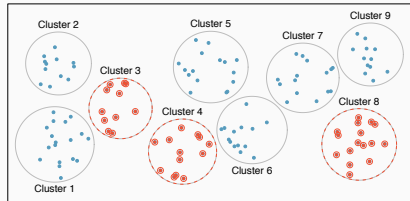
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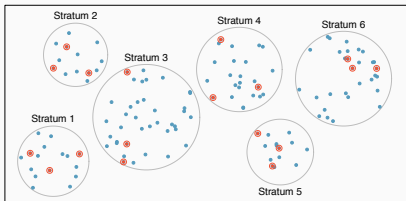
Cluster: heterogenous clusters

Sample all chosen clusters



Stratified: homogenous strata

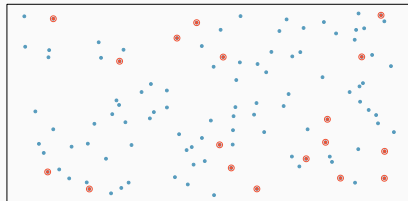
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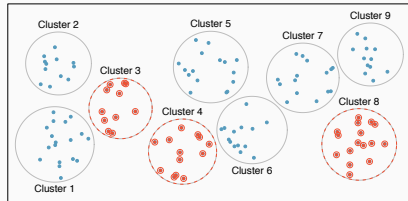
Simple random:

Drawing names from a hat



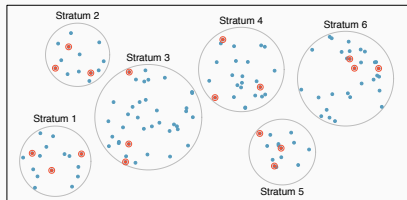
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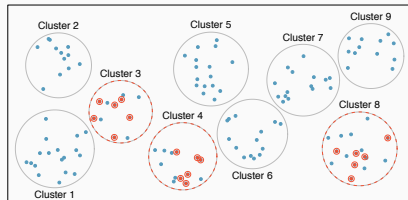
Stratified: homogenous strata

Stratify to control for SES



Multistage:

Random sample in chosen clusters



Clicker question

A city council has requested a household survey be conducted in a suburban area of their city. The area is broken into many distinct and unique neighborhoods, some including large homes, some with only apartments, and others a diverse mixture of housing structures. Which approach would likely be the *least* effective?

- (a) Simple random sampling
- (b) Stratified sampling, where each stratum is a neighborhood
- (c) Cluster sampling, where each cluster is a neighborhood

Biases in study design

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- *Convenience sample*: Individuals who are easily accessible are more likely to be included in the sample

Clicker question

A school district is considering whether it will no longer allow high school students to park at school after two recent accidents where students were severely injured. As a first step, they survey parents by mail, asking them whether or not the parents would object to this policy change. Of 6,000 surveys that go out, 1,200 are returned. Of these 1,200 surveys that were completed, 960 agreed with the policy change and 240 disagreed. Which of the following statements are true?

- I. Some of the mailings may have never reached the parents.
 - II. Overall, the school district has strong support from parents to move forward with the policy approval.
 - III. It is possible that majority of the parents of high school students disagree with the policy change.
 - IV. The survey results are unlikely to be biased because all parents were mailed a survey.
- (a) Only I (b) I and II (c) I and III (d) III and IV (e) Only IV

What type of study is this? What is the scope of inference (causality / generalizability)?

Facebook Tinkers With Users' Emotions in News Feed Experiment, Stirring Outcry

By VINDU GOEL JUNE 29, 2014

The New York Times

In [an academic paper](#) published in conjunction with two university researchers, the company reported that, for one week in January 2012, it had altered the number of positive and negative posts in the news feeds of 689,003 randomly selected users to see what effect the changes had on the tone of the posts the recipients then wrote.

The researchers found that moods were contagious. The people who saw more positive posts responded by writing more positive posts. Similarly, seeing more negative content prompted the viewers to be more negative in their own posts.

A study that surveyed a random sample of otherwise healthy adults found that people are more likely to get muscle cramps when they're stressed. The study also noted that people drink more coffee and sleep less when they're stressed. What type of study is this?

What is the conclusion of the study?

Can this study be used to conclude a causal relationship between increased stress and muscle cramps?

Four principles of experimental design

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Why is this important? Can you think of other variables to block for?

Random sampling helps generalizability, random assignment helps causality

<i>ideal experiment</i>	Random assignment	No random assignment	<i>most observational studies</i>
Random sampling	Causal conclusion, generalized to the whole population.	No causal conclusion, correlation statement generalized to the whole population.	Generalizability
No random sampling	Causal conclusion, only for the sample.	No causal conclusion, correlation statement only for the sample.	No generalizability
<i>most experiments</i>	Causation	Correlation	<i>bad observational studies</i>

Summary

Summary of main ideas

1. Use a sample to make inferences about the population
2. Ideally use a simple random sample, stratify to control for a variable, and cluster to make sampling easier
3. Sampling schemes can suffer from a variety of biases
4. Experiments use random assignment to treatment groups, observational studies do not
5. Four principles of experimental design: randomize, control, block, replicate
6. Random sampling helps generalizability, random assignment helps causality