

# PSYC 20100: Psychological Statistics

Autumn 2016

Professor Dan Yurovsky

## Why I love statistics

Undergrad in Computer Science at Carnegie Mellon

- interested in AI and Machine Learning
- Became interested in Human learning

PhD in Cognitive Psychology at Indiana University

- Studied how infants learn language
- "Statistical learning"

Interests

- How we communicate and learn from each-other
- How we are so smart while being so dumb
- Excited about using "big data" to understand how we develop and learn



Communication and Learning Lab

## Why you should love statistics too

1. Statistics are a way to cope with the absurd
2. Statistics are the connection between theory and the natural world
3. Statistics are the glue between theory and the social world

## Statistics and the absurd

"Man stands face to face with the irrational.  
He feels within him his longing for happiness and for reason.  
The absurd is born of this confrontation between the human  
need and the unreasonable silence of the world."



Albert Camus, The Myth of Sisyphus

To understand statistics is to embrace the absurd: *There is no certainty, only degrees of doubt*

## Statistics connect scientific theories to the world

The artifacts of science are models

*All models are wrong, but some are useful*



George Box

Because there is no certainty, no model can be *True*.  
Statistics is a set of tools for helping us to figure which ones are more useful.

## Statistics are an expression of liberty

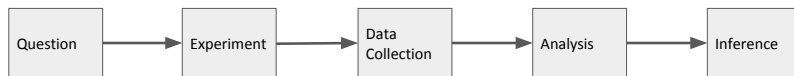
The fundamental premise of inferential statistics: You could be wrong!

The practice of statistics is *doubt* of authority

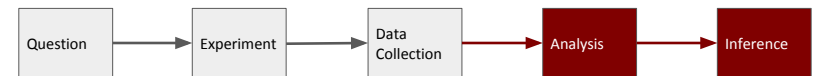
**Ubi dubium ibi libertas**

Thanks to John Kruschke

## Goals for PSYC 20100 and 20200



## Goals for PSYC 20100



## A statistical story



### A multi-scale approach to ambiguity reduction in word learning

A key question in language acquisition is how children and adults map words to their referents despite the ambiguity in naming events....



#### Air Itinerary Details Flights

San Francisco, CA (SFO), US  
Thu, 12 Sep 2013, 10:30 AM  
Airbus 320

Denver, CO (DEN), US  
Thu, 12 Sep 2013, 01:05 PM



Denver 7 – The Denver Channel

## Building a statistical model of flooding

Editor's note

### Boulder's 100-year flood: How to help, and how to talk about it

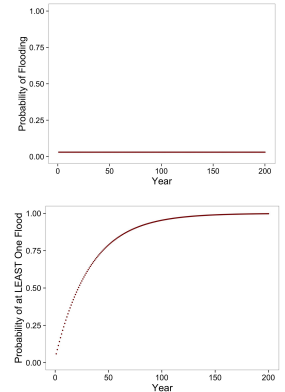
By Jenn Fields

[fields@coloradodaily.com](mailto:fields@coloradodaily.com)

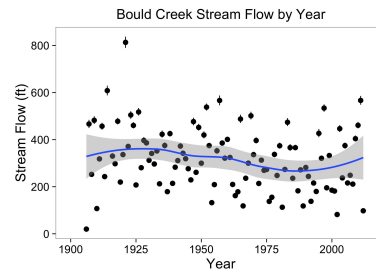
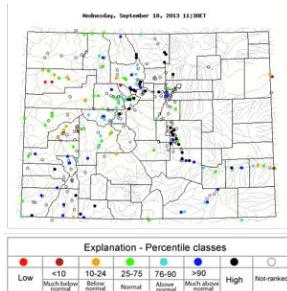


Is the chance of flooding every year an **independent** event?

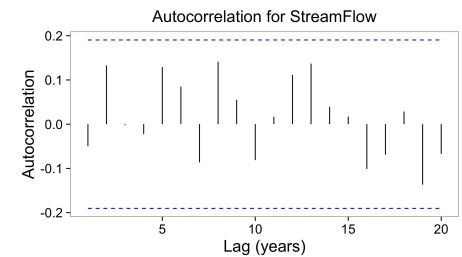
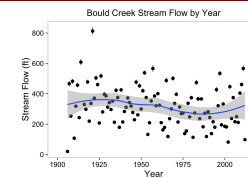
Every year you flip a coin, if it's heads you get a flood. Only the coin is weighted, and tails happens 97/100 flips.



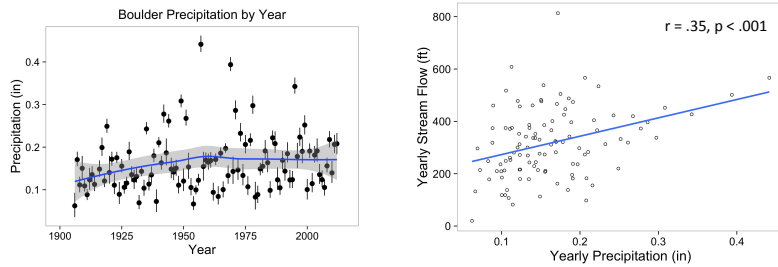
## Let's get some data to answer the question



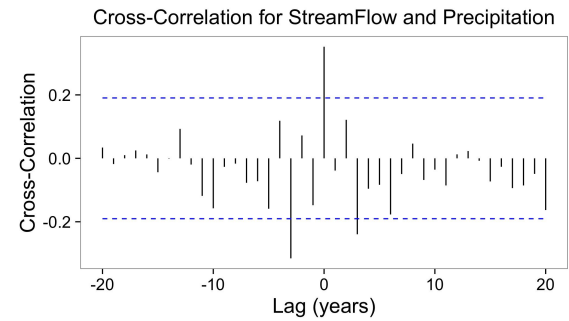
## Autocorrelation: A way of testing for independence



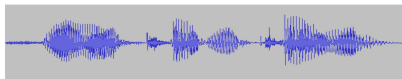
## Trying to predict streamflow



## Yearly precipitation *predicts* streamflow?



## How do you know what words are?



He re ki tt y ki tt y

Word boundaries are not marked by silences! But we can hear them anyway

Thanks to Mike Frank

## How do you know what words are?



bigoku vs. dobigo

Thanks to Julie Sedivy

## Segmenting words by detecting dependence

o look what a pretty baby  
what a pretty shirt  
oh look at the happy baby  
it's pretty late already  
there's a baby can you see it

If you just heard **ty**, you can't predict whether you will next hear **ba**  
*They are independent*

If you just heard **ba**, you are very likely to next hear **by**

Thanks to Mike Frank

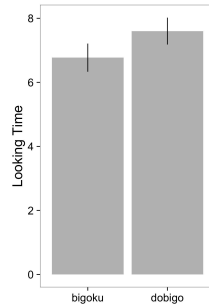
## Segmenting words by detecting dependence

buladobigokudatibabuladotadupabigoku

Test: **bigoku** (word) vs. **dobigo** (partword)

Thanks to Mike Frank

## Segmenting words by detecting dependence



Saffran, Aslin, & Newport (1996)

## By the end of the quarter, you should be able to:

1. Understand how the way that data is collected affects what you can learn from it
2. Use statistical software to summarize this data numerically and visually
3. Build statistical models of the data and understand which models are better and why
4. Make predictions about what kind of data you would expect to see in the future
5. Ask questions about the data, and make statistical inferences to answer them
6. Present these results in a transparent way to others
7. Understand the claims that others make from data and be able to critique them.

## Course information

### Teaching Team

Professor	Dr. Dan Yurovsky	yurovsky@uchicago.edu
Section 1 TA	Heather Manglesdorf	hharden@uchicago.edu
Section 2 TA	Omid Kardan	okardan@uchicago.edu

We want to help!

Come to our office hours, send us email, ask us questions!

### Online Resources

Course Website:

<https://dyurovsky.github.io/psyc20100/>

- Find syllabus, slides, etc.

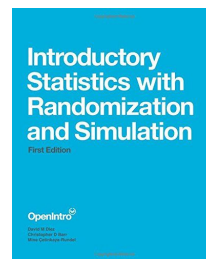
Google Classroom:

<https://classroom.google.com/>

- Submit assignments, post questions, etc.

## Two parallel roads to the goal

### Theory: Lectures and Textbook



### Application: Labs and Project



## Assessment and Grading

CAOS Pre and Post tests	5%
Quizzes	10%
Problem sets	30%
Labs	30%
Project	25%

**Theory**

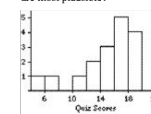
**Application**

## Comprehensive Assessment of Outcomes in a first Statistics Course (CAOS) Test



<https://apps3.cehd.umn.edu/artist/caos.html>

e.g. For this graphical display of Quiz Scores, which estimates of the mean and median are most plausible?



- median = 13.0 and mean = 12.0
- median = 14.0 and mean = 15.0
- median = 16.0 and mean = 14.3
- median = 16.5 and mean = 16.2

You will take a CAOS Pre and Post Test. *These will be graded for completion, not correctness.*

## Assessing your understanding of theory

### Quizzes

There will be a **quiz** every monday at the start of lecture. Quizzes are designed to give both you and your instructors rapid feedback about you understanding of the theory.

Your lowest grade will be dropped.

### Problem Sets

There will be a **problem set** assigned for each of the 4 units. These are designed to give you practice reasoning about the theory of statistics more deeply. You are encouraged to work together, **but must submit your own work**.

## Assessing your understanding of application

### Labs

Every friday, you will have a **lab** assignment. These are designed to give you practice applying the theoretical ideas you are learning to thinking about real data.

These will likely be challenging, especially if they are your first exposure to programming. But we are here to help, and so is a sizeable chunk of the internet!

These skills are useful, transferrable, and empowering. Seriously, you want to learn this!

### Project

The capstone assessment for the class is a **final project**. You will be given a dataset, and your goal will be to show something interesting about it. Think of this a larger, less structured lab assignment.

If you can do this, you (and we) will know that you really learned something!

### HOW MATH WORKS:



## The Curse of Knowledge

- These ideas are challenging
- If you don't understand them right away, don't worry!
- They took centuries to develop

