

**Bob the Brown
(and Slightly Gray)*
A Tale of Magic and Change**

*aka Bob delMas, University of Minnesota





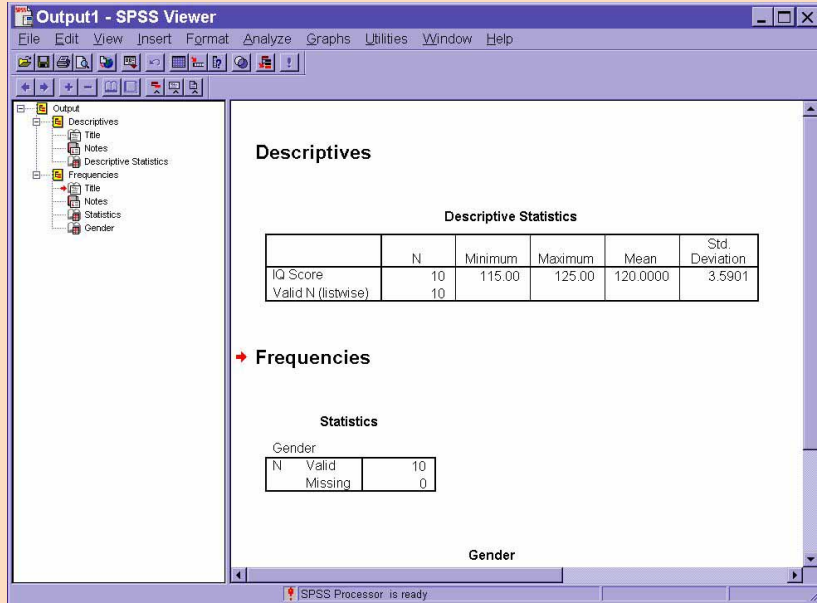
$$E(Y) = \frac{\sum Y}{N}$$

$$\text{Var}(Y) = \frac{\sum (Y - \bar{Y})^2}{N}$$

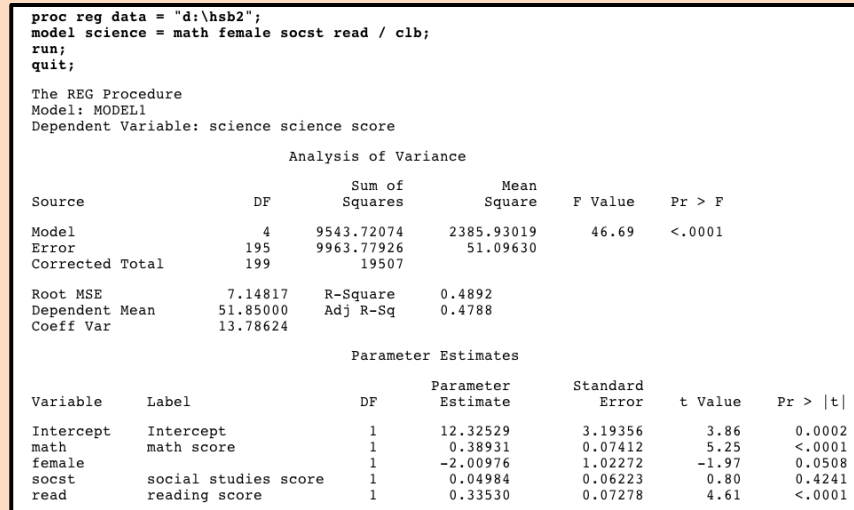
$$H_0: \mu_1 = \mu_2$$

$$P\left(\left|\bar{Y}_j - \bar{Y}_k\right| \geq \left|\bar{Y}_1 - \bar{Y}_2\right| \mid H_0: \mu_1 = \mu_2\right)$$

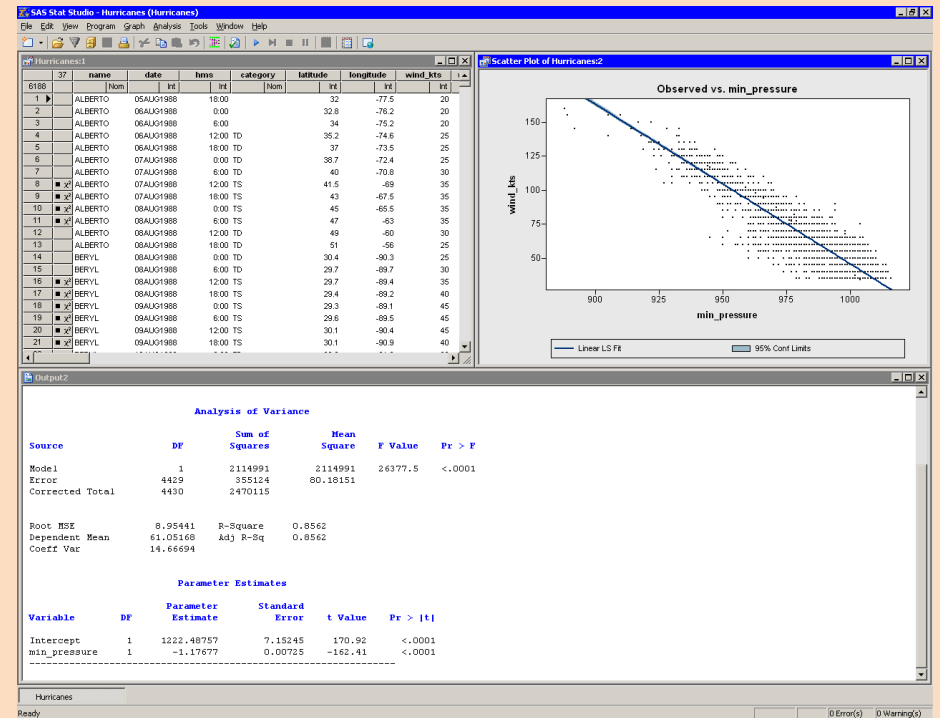
SPSS



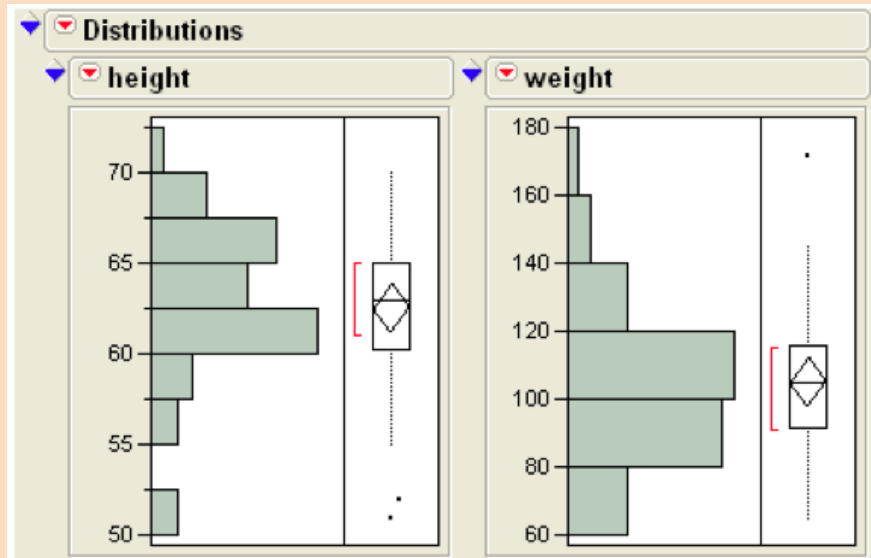
SAS



Minitab



JMP

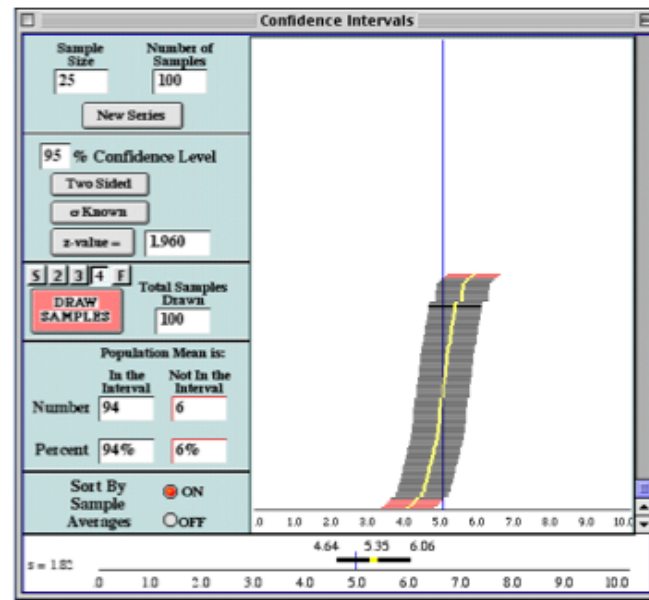
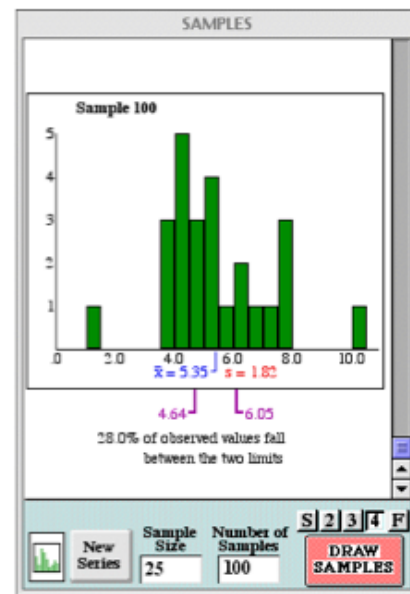
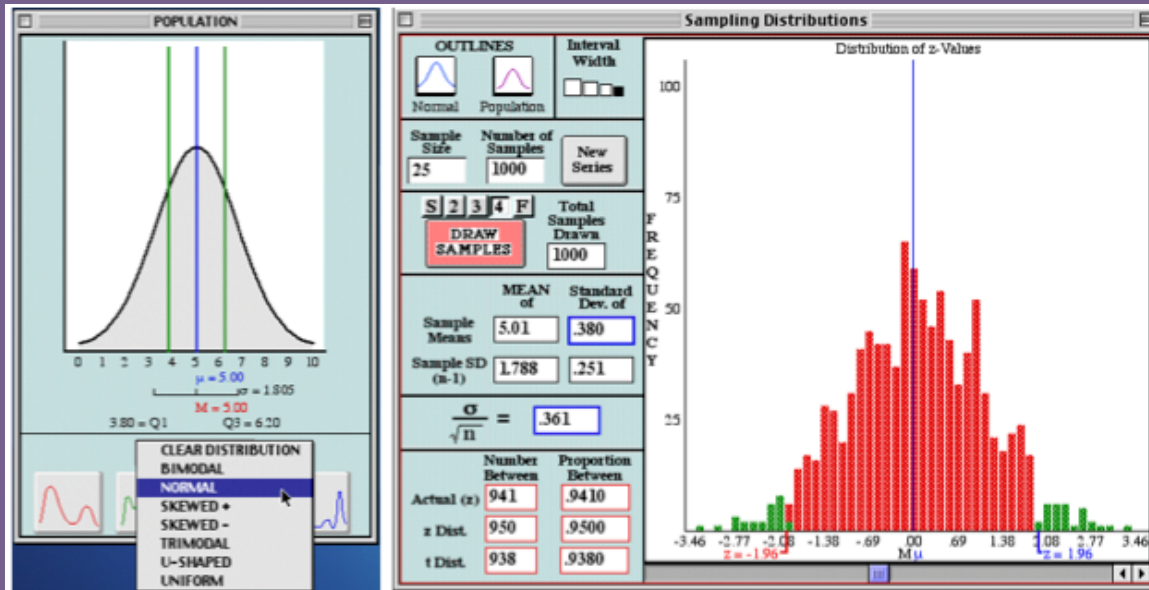


Let's see ... the output says $p = 0.384$, and that's the probability that H_0 is true, so less than a 50% chance that H_0 is true, meaning it's not possible, so I reject H_0 - there is a difference!





Sampling SIM



Sampling SIM Instructions

Sampling Distributions

OUTLINES
 Normal Population

Interval Width

Sample Size **Number of Samples** **New Series**

DRAW SAMPLES **Total Samples Drawn**

Sample Means **MEAN of** **Standard Dev. of**

Sample Stand. Dev.

$\frac{\sigma}{\sqrt{n}} =$

	Number Below	Proportion Below
Actual	<input type="text"/>	<input type="text"/>
Normal Dist.	<input type="text"/>	<input type="text"/>

Distribution of Sample Means

Specify the sample size of each sample drawn (e.g. n = 1)

Specify how many samples you want to draw

Indicates how many samples have been drawn (e.g. 10)

The standard deviation of the sample means

The average of the sample means

The average of the sample standard deviations.

Theoretical value for the standard error of the sample means

Values of Sample Means (\bar{x})

Teacher's Corner

Applying Cognitive Theory to Statistics Instruction

Marsha C. LOVETT and Joel B. GREENHOUSE

This article presents five principles of learning, derived from cognitive theory and supported by empirical results in cognitive psychology. To bridge the gap between theory and practice, each of these principles is transformed into a practical guideline and exemplified in a real teaching context. It is argued that this approach of putting cognitive theory into practice can offer several benefits to statistics education: a means for explaining and understanding why efforts work; a set of guidelines that can help instructors make well-informed design decisions when implementing instructional innovations.

KEY WORDS: Instructional technique; statistical education.

... time period, research in cognitive psychology has raised the question of how learning can be defined in similar terms—skills that enables new and the question of how to be respective. In cognitive theory, skill

Capturing and modeling the process of conceptual change

Stella Vosniadou

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University of Illinois at Urbana-Champaign, U.S.A.

[http://dx.doi.org/10.1016/0959-4752\(94\)90018-3](http://dx.doi.org/10.1016/0959-4752(94)90018-3)

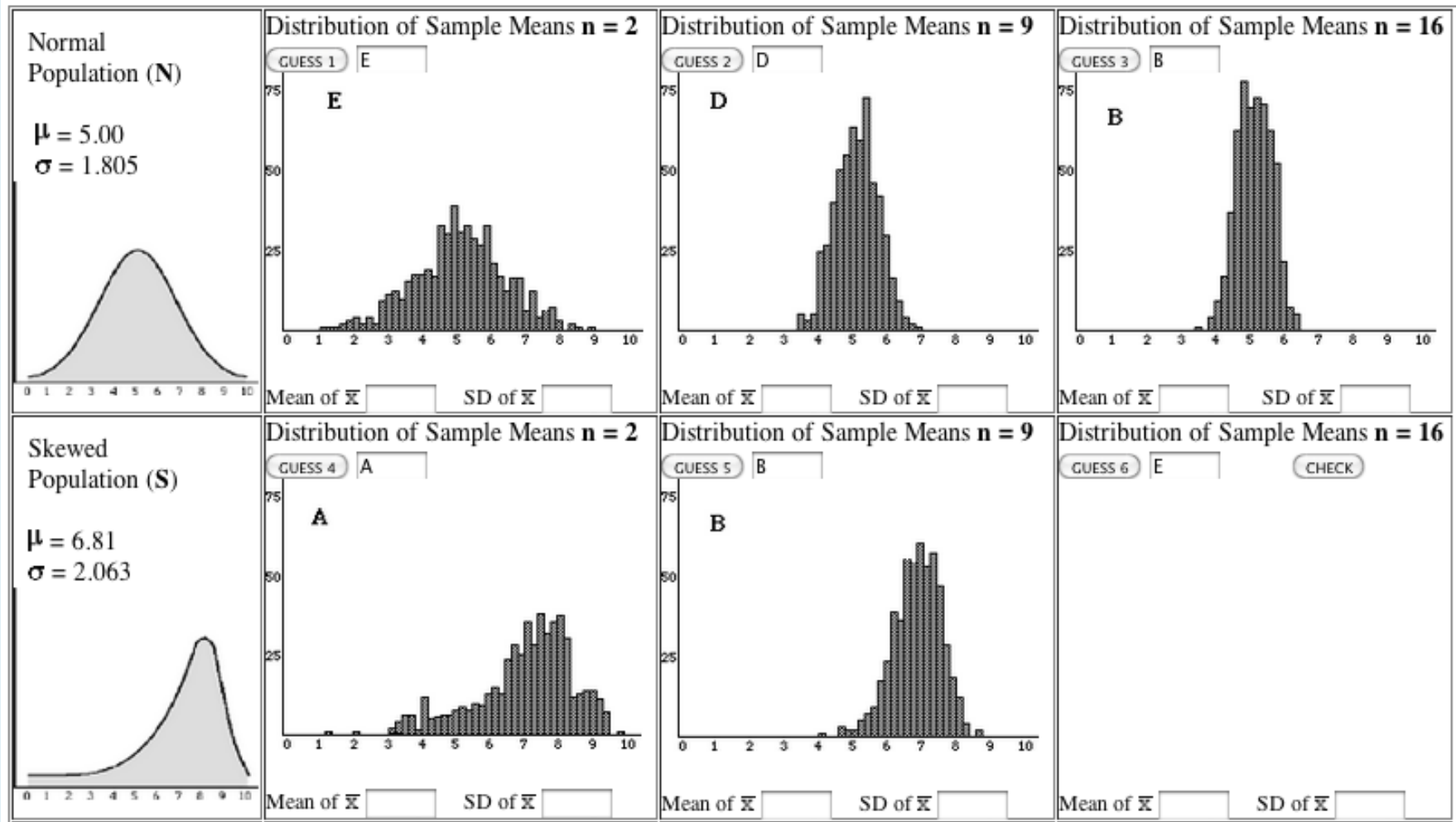
Permissions & Reprints

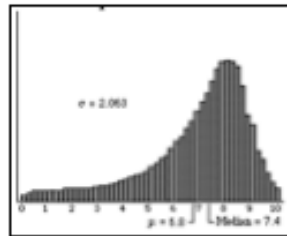
[How to Cite or Link Using DOI](#)

Abstract

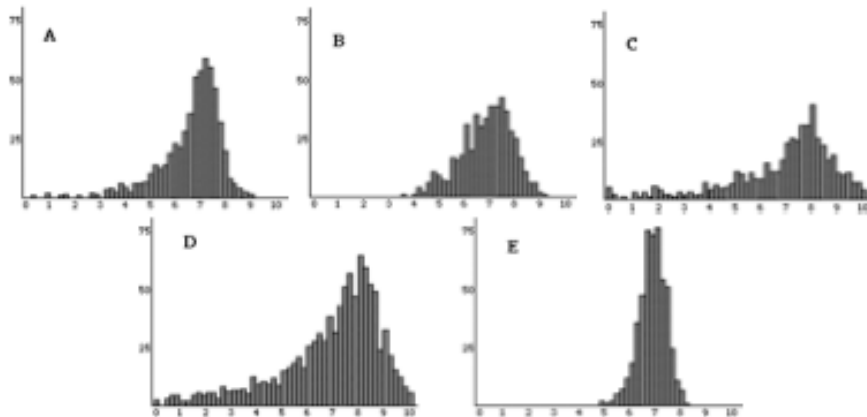
A theoretical framework is outlined in this article that attempts to explain the nature of conceptual change that takes place in the learning of physical science. It is argued that a naive framework theory of physics is established early on in infancy and forms the basis of individuals' ontology and epistemology. The presuppositions of this framework theory act as constraints on the way individuals interpret their observations and the information they receive from the culture to construct specific theories about the physical world. The specific theories formed through this process are continuously enriched and modified. Some kinds of conceptual change require the simple addition of new information and presuppositions are revised. It is proposed that conceptual change is particularly difficult to achieve and very likely to give rise to misconceptions when it requires the revision of fundamental presuppositions of the framework. Misconceptions are interpreted as individuals' attempts to assimilate new information into existing conceptual structures that contain information contradictory to the new information.

Sampling SIM Scrapbook





Each of the five graphs below labeled A to E represent possible distributions of sample means from random samples drawn from Talia's population.



A distribution of sample means from samples of size $n = 16$ should look more like the population. I predict the distribution of sample means will look like D.





"I argue that despite broad acceptance and rapid growth in enrollments, the consensus curriculum is still an unwitting prisoner of history...Randomization-based inference makes a direct connection between data production and the logic of inference that deserves to be at the core of every introductory course."





Modeling and Simulation with TinkerPlots™



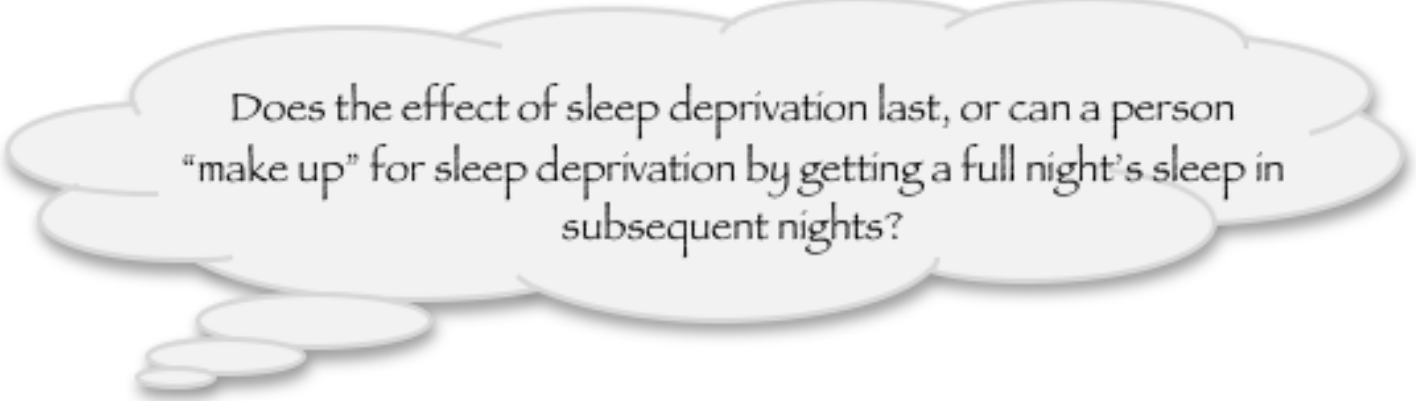
CATALST Course Curriculum

DAY	ACTIVITY
Day 1	<ul style="list-style-type: none"> ▪ Introduction
Day 2	<ul style="list-style-type: none"> ▪ Exploring the Behavior of “Random” [iPod Shuffle MEA]
Day 3	<ul style="list-style-type: none"> ▪ Understanding Human Intuitions about Randomness [Can you beat randomness-part I] ▪ Modeling “Random” Behavior [Checking You Intuitions-dice and coins - Add Intuitions HW]
Day 4	<ul style="list-style-type: none"> ▪ Modeling “Random” Behavior [Checking You Intuitions extend w/count and collect] ▪ Coins, dice, hw2 scenario
Day 5	<ul style="list-style-type: none"> ▪ Modeling Complex Phenomena [One Son until boy/one of each, describe model]
Day 6	<ul style="list-style-type: none"> ▪ Modeling “Blind Guessing” [Matching Dogs w/o replacement, match, counter, describe model, and trial]
Day7	<ul style="list-style-type: none"> ▪ Modeling Variation for a Statistic Based on “Blind Guessing” [Helper / Hinderer, remove p-value, likely/unlikely, conditional on model]
Day 8	<ul style="list-style-type: none"> ▪ Introduction to Randomization Tests [Memorizing Letters]
Day 9	<ul style="list-style-type: none"> ▪ Randomization Tests [Sleep Deprivation]
Day 10	<ul style="list-style-type: none"> ▪ Designing Experiments: Role of Random Assignment Tests [Shoe Strength]

Sleep Deprivation Study

Researchers have established that sleep deprivation has a harmful effect on visual learning. But do these effects linger for several days, or can a person “make up” for sleep deprivation by getting a full night’s sleep in subsequent nights?

A recent study (Stickgold, James, and Hobson, 2000) investigated this question by randomly assigning 21 subjects (volunteers between the ages of 18 and 25) to one of two groups: one group was deprived of sleep on the night following training and pre-testing with a visual discrimination task, and the other group was permitted unrestricted sleep on that first night. Both groups were then allowed as much sleep as they wanted on the following two nights. All subjects were then re-tested on the third day.



Does the effect of sleep deprivation last, or can a person “make up” for sleep deprivation by getting a full night’s sleep in subsequent nights?

Subjects’ performance on the test was recorded as the minimum time (in milliseconds) between stimuli appearing on a computer screen for which they could accurately report what they had seen on the screen. The sorted data and plots presented here are the improvements in those reporting times between the pre-test and post-test (a negative value indicates a decrease in performance):

Statistical Thinking



A Simulation Approach
to Modeling Uncertainty

Catalysts for Change



WARNING
BLATANT AND
UNABASHED
PROMOTION OF A
TEXTBOOK





WARNING

BLATANT AND UNABASHED
PROMOTION OF A BREAKOUT SESSION

Breakout Session #2C (Hope)

Friday

3:00-4:15pm

*Evaluating the Impact of Change in
Curriculum and Teaching*

