



Opportunities in Statistics Education Research

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Who am I?

- ❖ Undergraduate math major
- ❖ PhD Operations Research
 - Minor in Education
 - *Behavior Characterization and Estimation for General Hierarchical Multivariate Linear Regression Models*
- ❖ Mathematics department, University of the Pacific
- ❖ Statistics department, Cal Poly – SLO (1999)

Overview

- ❖ What can we learn from the history of statistics education research?
- ❖ What kind of research is being done?
 - Classroom based research
 - Qualitative research
- ❖ Examples
- ❖ Resources/Advice

*What can we learn from the
“history” of statistical research?*



Beginnings of Stat Ed Research

- ❖ Psychology research, Educational psychology, Cognitive science
- ❖ Mathematics education
 - International group for the psychology of mathematics education
 - ICME: International congress of mathematics education
 - *JRME: Journal of Research in Mathematics Education*
- ❖ Statistics education research
 - IASE: International association for statistical education
 - *SERJ: Statistics Education Research Journal*
 - ICOTS: International conference on teaching statistics (USCOTS)

Beginnings of Stat Ed Research

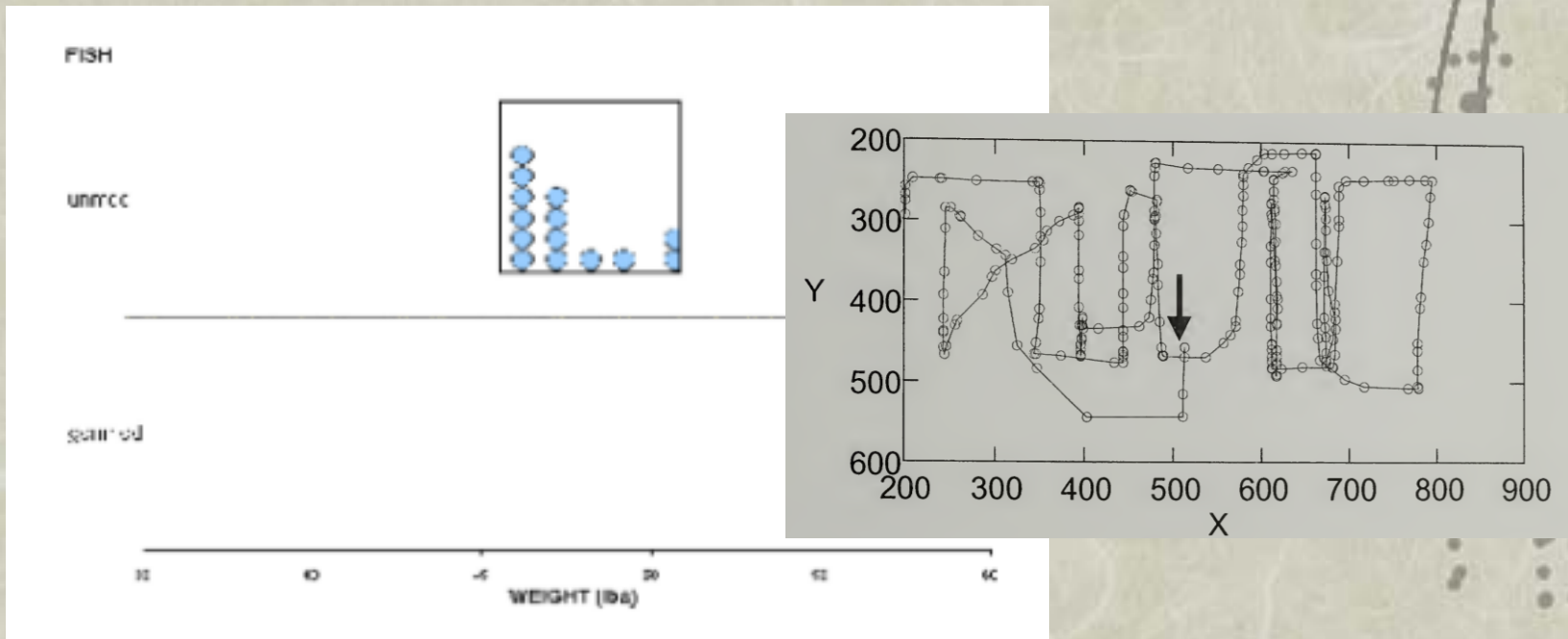
- ❖ Initially limited focus
 - Identification of misconceptions
 - Comparison of instructional modes
 - Prediction of achievement
- ❖ Rarely focused on statistical reasoning
- ❖ Still “straddling two worlds” a bit

Measurement

- ❖ Standardized exams, Final exams, Student ratings
 - Students who earn good grades on final exams often demonstrate poor statistical reasoning skills (Hawkins, Joliffe, & Glickman, 1992)
 - Most existing high-impact standardized exams are poorly aligned with national standards for instruction and assessment (Lesh & Lovitts, 2000)
 - Traditional exams too often emphasize final answer over the process” Garfield (1993)
- ❖ If goals change, so must assessment strategies

New kinds of “data”

- ❖ Cliff Konold, “The virtues of building on sand”, plenary presentation, ICOTS 8, 2010



Research Techniques

- ❖ Randomized comparative experiments
 - *Using statistics effectively in mathematics education research* (Scheaffer et al., 2007)
- ❖ Cautions
 - long vs. short-term
 - confounding variables
 - realism
 - time delays
 - ethical issues
 - external perspective

Qualitative Research

- ❖ *SERJ* special issue: Qualitative approaches in statistics education research (Nov. 2010)
- ❖ Standards
 - Validity
 - Generalizability
 - Reliability
 - Objectivity
- ❖ Consistent, Replicable, Well-documented, Fair and equitable

Classroom-Based Research

- ❖ “Teachers researching their own practice of teaching.”
 - Feldman & Minstrell in Kelly & Lesh (2000)
- ❖ “It is most simply defined as ongoing and cumulative intellectual inquiry by classroom teachers into the nature of teaching and learning in their own classrooms.”
 - Cross and Steadman (1986)

Classroom-Based Research

- ❖ Narrows gap between theory and practice
 - direct link to classroom environment
- ❖ Further insight into classroom, students
 - combined with nonparticipant viewpoint
- ❖ Dynamic
- ❖ Open to alternative student interpretations
- ❖ Focus on process

To Experiment or Not to Experiment?

- ❖ Variety of tools should be employed
 - “Different techniques generate different types of information, and it is often the case that a single technique will not provide the breadth of information necessary to answer unequivocally the research questions under investigation.” – Mestre (2000)
 - qualitative and quantitative data
 - observe before manipulate

Human Subjects

- ❖ Talk to your institution's Institutional Review Board (IRB)
 - Exemption

Best Practices

- ❖ *Handbook of Research Design in Mathematics and Science Education*
 - Kelly and Lesh, Eds. (2000)
- ❖ *International Handbook of Research in Statistics Education*
 - Ben-Zvi, Makar, and Garfield, Eds. (2018)
- ❖ Research agenda/Instruments and Methods

Examples

- ❖ 1. Tools for Teaching, Assignment Statistical Inference Project
 - 5-year collaboration with Joan Garfield and Bob delMas
 - Diverse background/perspective
 - Diverse institutions
 - Developed new instructional technology
 - Developed new assessment tools (visual), student interviews

Examples

- ❖ 2. Simulation-Based Inference Project
 - 10-year collaboration with Tintle, Cobb, Rossman, Roy, Swanson, VanderStoep
 - Developed new instructional materials
 - Developed new instructional technology (applets)
 - Cross-institutional assessment (pre/post)
 - Workshops for teachers

NSF Grants

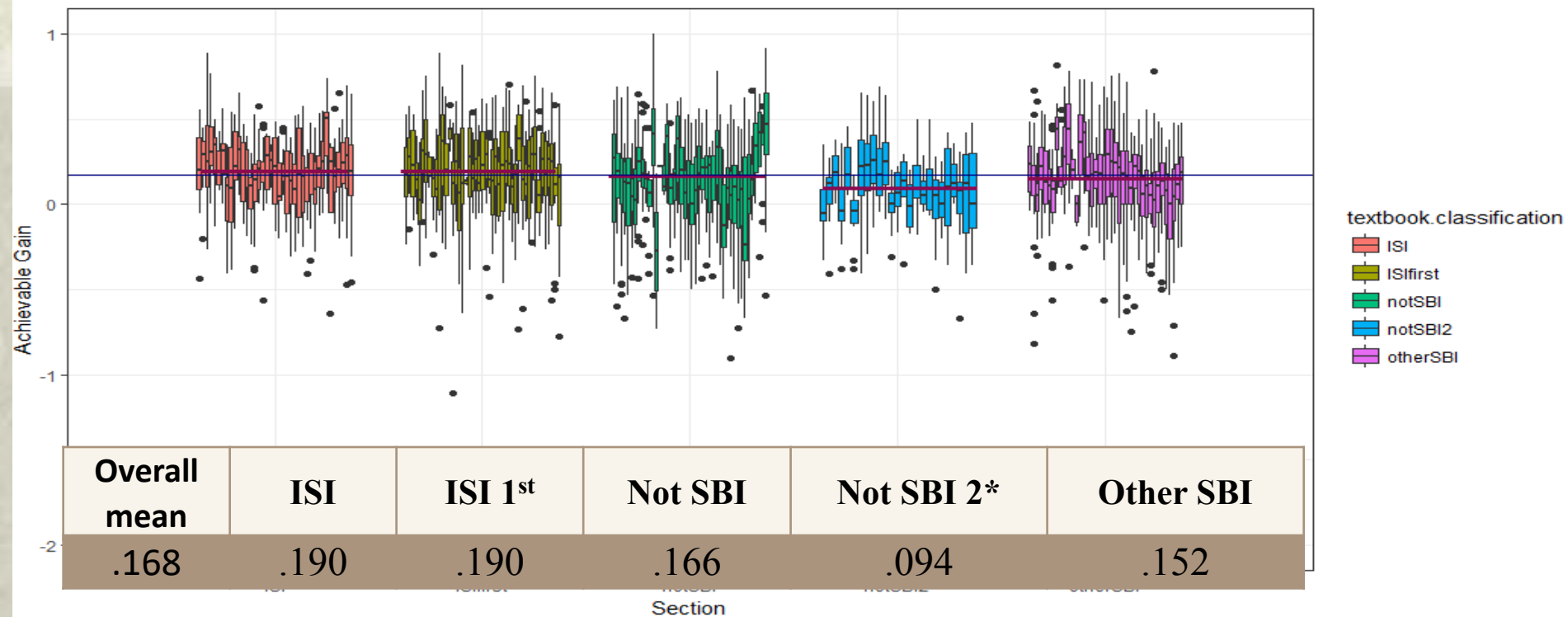
- ❖ CCLI #0633349: *Concepts of Statistical Inference: A Randomization-Based Curriculum,*
- ❖ NSF TUES Type I Project, Award #1140629: *Developing an Innovative Randomization-based Introductory Statistics Curriculum*
- ❖ NSF/TUES/DUE- Phase II, Award #1323210: *Broadening the impact and evaluating the effectiveness of randomization-based curricula for introductory statistics*
- ❖ NSF/IUSE/HER #1612201: *Developing and Assessing a Conceptual Approach to an Algebra-based Second Course in Statistics*
- ❖ RCN-UBE #1730668: *Statistical Thinking in Undergraduate Biology (STUB) Network: A network for coordinating the teaching and assessment of statistical thinking in introductory biology*

Measurements

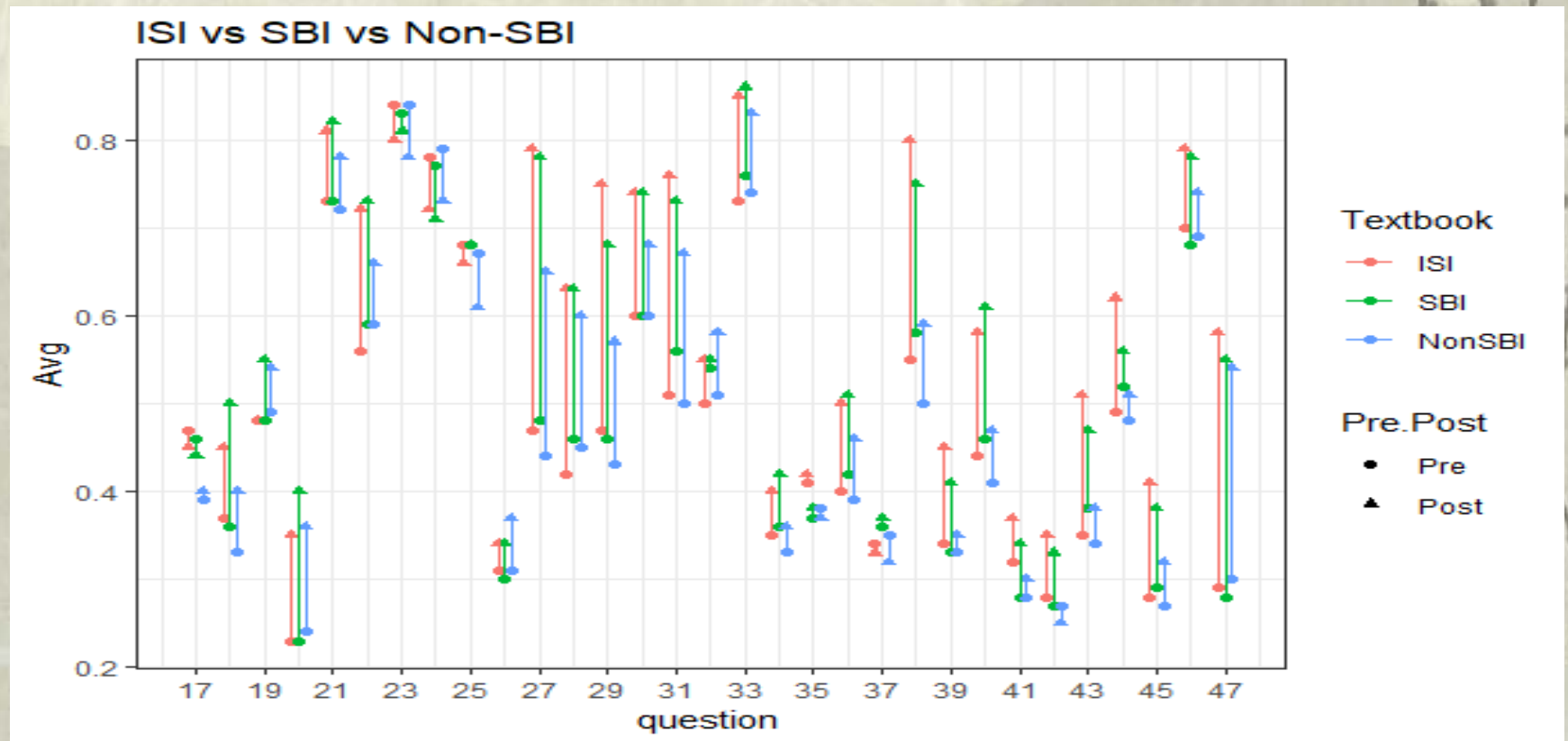
- ❖ Multi-institutional assessment pre/post
 - Concept inventory adapted from CAOS
 - Comprehensive Assessment of Outcomes in first Statistics course (see also GOALS)
 - ~35 questions (24 “sets”)
 - National comparison data
 - SATS = Students Attitudes Toward Statistics (Schau, 2003; see *SERJ* special issue – Nov 2012)
 - 6 Subscales (e.g., Affect, Difficulty, Effort, Cognitive competence, Value, Interest)
 - Demographic data

Multilevel Models!

Boxplots of Achievable Gains by Section and Textbook (2016-2017)



Data Visualization



Where to begin?



Key Principles – How students learn statistics

- ❖ Garfield (1995), Garfield & Ben-Zvi (2007)
 - Students learn by constructing knowledge
 - Students learn by active involvement in learning activities
 - *Active learning increases student performance in science, engineering, and mathematics, Freeman et al (PNAS, 2014)*
 - Students learn to do well only what they practice doing
 - Do not underestimate student difficulty

Key Principles – How students learn statistics

- ❖ Garfield (1995), Garfield & Ben-Zvi (2007)
 - Students need to become aware of and confront their errors in reasoning
 - Use technology to visualize and explore
 - Consistent and helpful feedback on their performance
 - Students learn to value what they know will be assessed

Some Current Efforts

- ❖ Service learning (e.g., Doehler; Nordmoe; Hydorn; Phelps), Experiential learning (e.g., Morris)
- ❖ Context-driven statistics (e.g., Dierker, ProCivicStats, Strengthening Data Literacy across the Curriculum)
- ❖ Beyond the first course (e.g., Kuiper; Tintle et al.; Chihara & Hesterberg; Nolan)
- ❖ Connections to research (e.g., Son & Stigler)
- ❖ Assessment, Adaptive testing (e.g., Beckman; Sabbag; Broaddus; Cheng)
- ❖ Interdisciplinary collaboration (e.g., STUB)

Some Current Questions

- ❖ Expert vs. Novice
- ❖ Student experience vs. instructor demonstration
- ❖ Large classes
- ❖ Analyzing student interaction with technology
- ❖ Preparation of future teachers
- ❖ Retention
- ❖ Student attitudes
- ❖ Statistics vs. Data Science

Advice – Designing a Lesson

- ❖ What are the learning goals?
 - What are common student difficulties
- ❖ How will I assess whether students have met those goals?
- ❖ How does it connect to content before/after this lesson?
- ❖ What is an engaging context?
- ❖ How/when do I actively engage the students
 - Directly confront student difficulties
- ❖ Will technology be helpful?

Advice – Designing a Research Question

- ❖ What is my audience?
- ❖ What are the learning goals?
 - What are common student difficulties
- ❖ What do I plan to do differently?
 - What are my preconceptions?
- ❖ How does it connect to prior research?
- ❖ How will I assess whether students have met those goals/whether it works?

Advice – Designing a Research Study (Grant)

- ❖ Familiarize yourself with the research, assessment tools
 - NSF Award Search
- ❖ Connect with others (e.g., causeweb.org)
 - Across institutions
 - Across disciplines
 - New and “Old” folks
- ❖ Talk with program officer
- ❖ Be open to alternative research methodologies
 - Synergy with your “real” research

e.g., NSF grant funding

- ❖ The IUSE program (formerly TUES) at the National Science Foundation supports curricular innovation, experimentation, and implementation

Track	Level
Engaged student learning	Level 1/Level 2/Level 3
Institutional and community transformation	Capacity building/Level 1/Level 2

Summary

- ❖ Look to history
- ❖ Importance of collaboration
- ❖ Student involvement
- ❖ Not all randomized experiments
 - Qualitative research, Think-aloud protocols, Learning trajectories, Classroom-based research
- ❖ New measurement tools
- ❖ Not only about students
 - Teacher preparation
 - Role of technology in teaching
 - Integration with data science, other disciplines

Any Questions?

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1. Tools for Teaching, Assessing Statistical Inference Project (Garfield, delMas, Chance)

- ❖ Can interaction with simulation program improve student reasoning about sampling distributions?
 - how to best integrate technology into instruction
 - why particular techniques more effective
 - how student understanding evolves

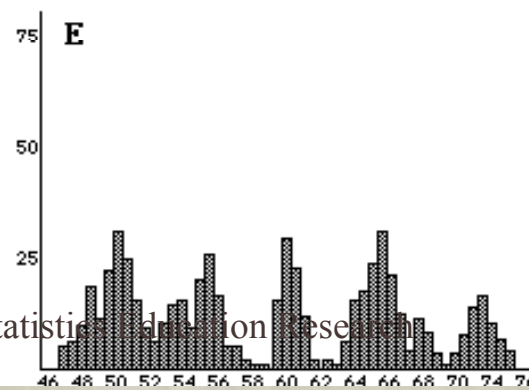
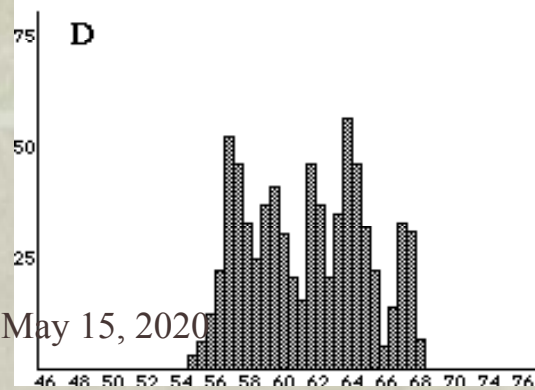
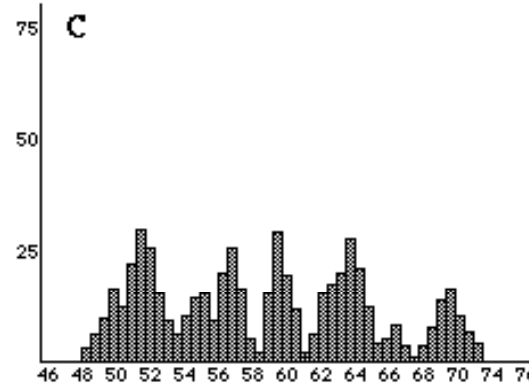
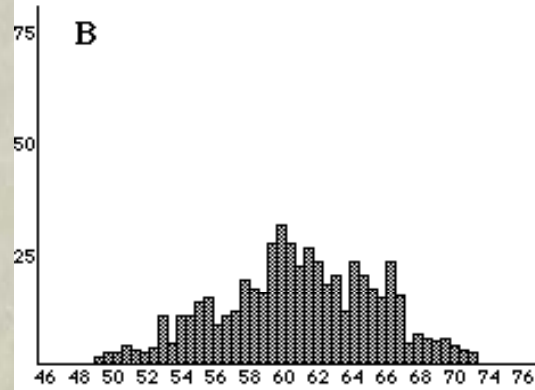
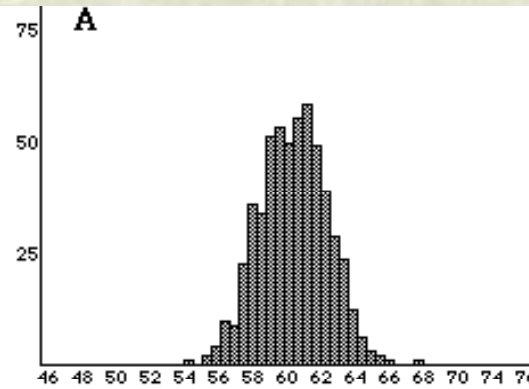
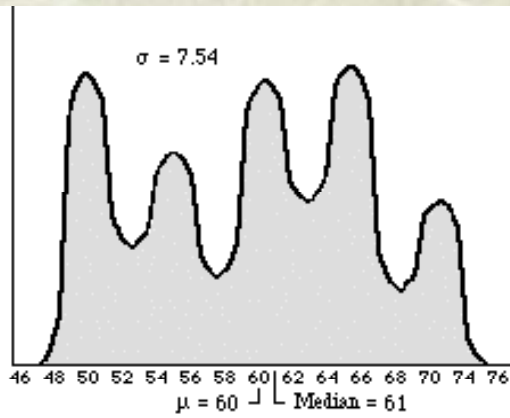
This project was supported, in part, by the National Science Foundation, DUE-9752523

Methods

- ❖ Diverse tertiary environments
 - private liberal arts college, college of education, developmental education college
- ❖ Diverse student groups
- ❖ Researchers with diverse backgrounds
- ❖ Creation of desired learning environment
- ❖ Incorporation of existing theory
- ❖ Generating new models

Measurements

- ❖ Graphics-based test items
- ❖ Open-ended questions, justifications
- ❖ Multiple choice categorizations
- ❖ Pre-test vs. Post-test performance
- ❖ Post-test application problems
- ❖ Using assessment to create dissonance
- ❖ In-depth interviews and videotape analysis



m

$n=1?$

$n=4?$

$n=16?$

$n=25?$

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Statistical Inference Research

Student interviews/Videos

“I’m going to go for C for $n=4$ and then 16 for ... $n=16$ for A. And partially because ... with $n=4$, I’m thinking you are going to have a larger range ... yeah, a larger range for $n=4$ than that you would for $n=16$.

Student interviews/Videos

“Because before I was guessing and I thought that the standard deviation for a larger sample size would be closer to the original than the standard deviation for $n=4$.”

Meeting the Standards

- ❖ Validity: prolonged investigation, immersion, triangulation, member checks
- ❖ Generalizability: extensive description, multi-site design
- ❖ Reliability: multiple perspectives, participant involvement
- ❖ Objectivity: documentation, dissemination
 - “acknowledged and controlled subjectivity”

Additional Benefits

- ❖ Narrows gap between theory and practice
 - direct link to classroom environment
- ❖ Further insight into classroom, students
 - combined with nonparticipant viewpoint
- ❖ Dynamic
- ❖ Open to alternative student interpretations
- ❖ Focus on process

1. Tools for Teaching, Assessing Statistical Inference Project (Garfield, delMas, Chance)

A model of classroom research in action: Developing simulation activities to improve students' statistical reasoning

Article (PDF Available) in [Journal of Statistics Education](#) 7(3) · January 1999 with 396 Reads

2. Simulation-Based Inference Project (Tintle, Cobb, Rossman, Roy, Swanson, VanderStoep...)

- ❖ Does use of simulation-based inference in introductory statistics courses improve students' ability to learn statistics and/or students' attitudes towards statistics
 - Curriculum materials
 - Technology tools
 - Assessment
 - Dissemination
 - NSF grants:



Methods

- ❖ Web-based applets
- ❖ Textbooks
 - Process, Spiraling, Active learning, Assessment
- ❖ SBI Blog (www.causeweb.org/SBI)
- ❖ Workshops for teachers