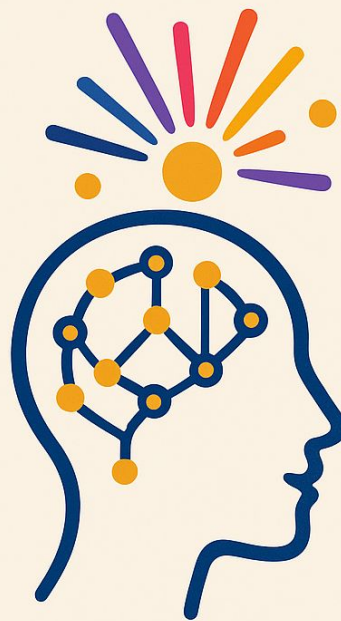


Electronic Conference on Teaching Statistics (eCOTS) 2026

Closing Session

**Nick Horton (Amherst
College)**



Sparking Joy and Discovery

In a World of AI
eCOTS 2026

Congratulations to the R Core Team!

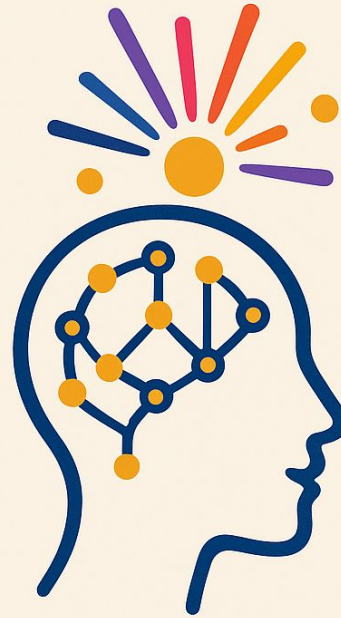
<https://rousseeuwprize.org/2026>

THE ROUSSEEUW PRIZE FOR STATISTICS 2026
The R Project



So Long, and Thanks for All the Fish!

[https://en.wikipedia.org/wiki/So_Long,
_and_Thanks_for_All_the_Fish](https://en.wikipedia.org/wiki/So_Long,_and_Thanks_for_All_the_Fish)



**Sparking
Joy and
Discovery**
In a World of AI
eCOTS 2026

Sponsors



WILEY



PennState
Eberly College of Science

Sponsors

Ross Metusalem: Multivariate
Analysis with JMP



Patti Frazer Lock: Teaching Statistics
using the GAISE College Report

WILEY

Julia Silge: Teaching with Positron



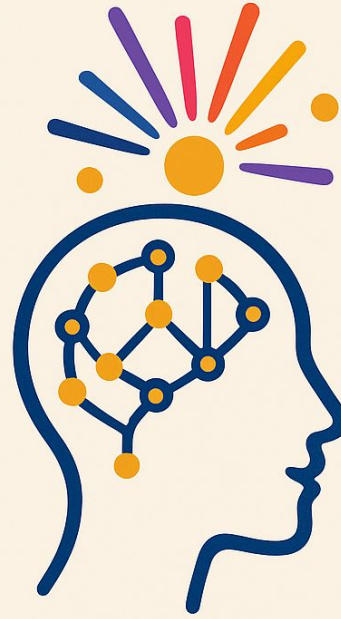
CAUSE + Andrew + Terra + Matt + Lydia!



PennState
Eberly College of Science

CAUSE Research

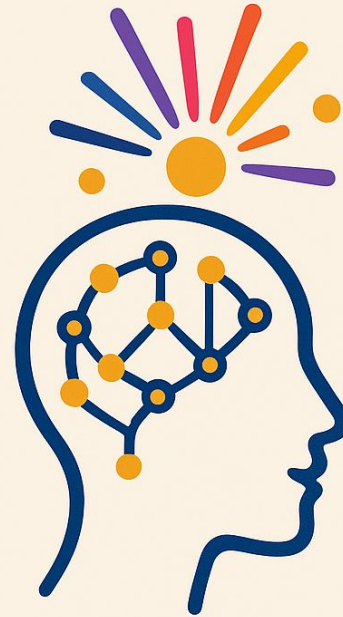
Laura Le (University
of Minnesota) and
Jenny Green
(Michigan State
University)



**Sparking
Joy and
Discovery**
In a World of AI
eCOTS 2026

USPROC/USCLAP/USRESP

- **Ciaran Evans** (Wake Forest University)
- **Ranjini Grove** (University of Washington)
- **Shaoyang Ning** (Swarthmore College)
- **Annie Tang** (Colby College)
- **Victoria Woodard** (Notre Dame University)



**Sparking
Joy and
Discovery**
In a World of AI
eCOTS 2026

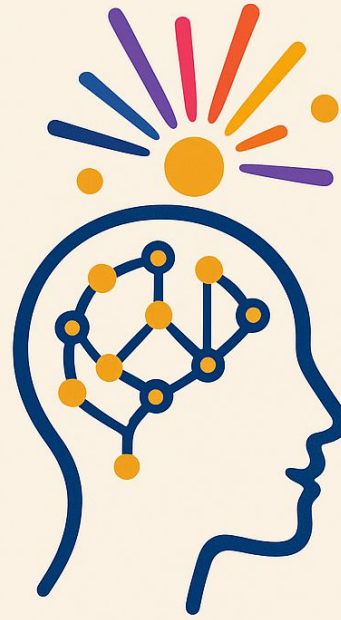
Speed Mentoring

Tyler George

(Cornell College) and

Catalina Medina

(CSU Channel Islands)



**Sparking
Joy and
Discovery**

**In a World of AI
eCOTS 2026**

Speed Mentoring

Kudos to our mentors!

Tori Ellison

Camille Fairbourn

Jennifer Green

Jennifer Kaplan

Tyler Maley

Ariadni Papana

Uma Ravat

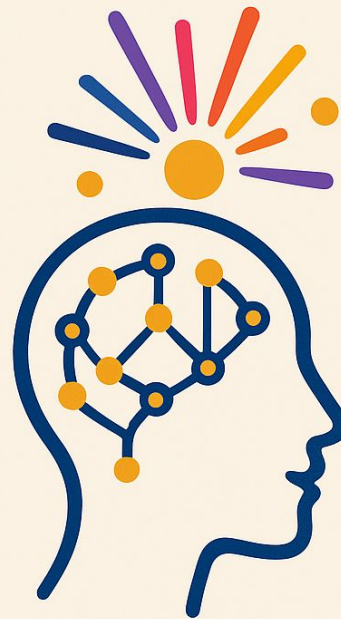
Joel Sanqui

Derek Sollberger

Travis Weiland

Bethany White

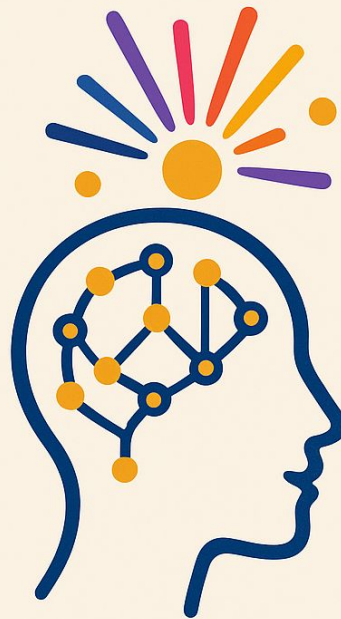
Behind the Scenes Award



**Sparking
Joy and
Discovery**
In a World of AI
eCOTS 2026

Behind the Scenes Award

Andrew Ferguson
(Penn State University)



**Sparking
Joy and
Discovery**
In a World of AI
eCOTS 2026

2026 Program Committee

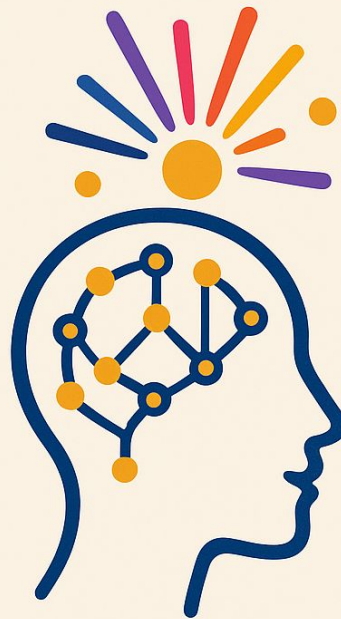
- Heather Cook (poster co-chair), University of Southern Indiana
- Andrew Ferguson, Penn State University
- Tyler George, Cornell College
- Juan Gomez, Carmel High School, CA
- Jacqueline Herman (poster co-chair), Northern Kentucky University
- Nicholas Horton (chair), Amherst College
- Vimal Rao, University of Illinois
- Kelly Spoon, San Diego Mesa College
- Sara Stoudt, Bucknell University
- Maria Tackett, Duke University
- Allison Theobald, Cal Poly, San Luis Obispo
- Jennifer Ward, Clark College
- Jonathan Wells, Grinnell College
- Beverly L. Wood, Embry-Riddle Aeronautical University

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- Maria Tackett, Duke University
- Allison Theobald, Cal Poly, San Luis Obispo
- Jennifer Ward, Clark College
- Jonathan Wells, Grinnell College
- Beverly L. Wood, Embry-Riddle Aeronautical University

Plus the leadership of CAUSE
Director Matt Beckman!

What have we
learned?
Where do we go
from here?



**Sparking
Joy and
Discovery**
In a World of AI
eCOTS 2026

**What would Dennis Pearl
Say?**

**What can we
learn from the
world of chess?**

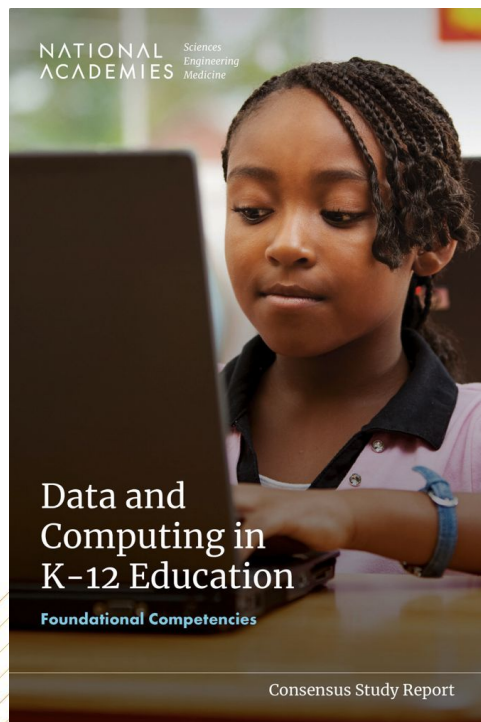
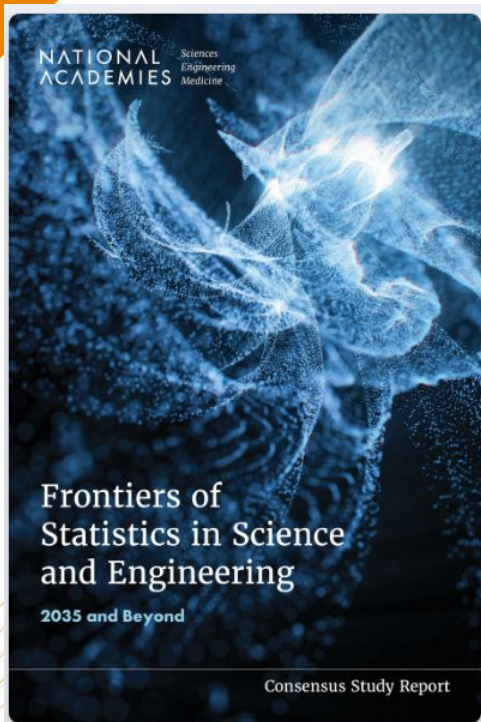


**What would Dennis Pearl
Say?**

**Congrats again to
Dennis for the 2026
George Cobb Lifetime
Achievement Award!**



Liz Stuart (Johns Hopkins) and Lance Waller (Emory University) keynote



Both reports freely downloadable (with free NASEM account)

Modernizing Mathematics Education for Grades 9-14 consensus report

Information Gathering Open Session to be held June 23
Noon-4pm ET (Virtual)

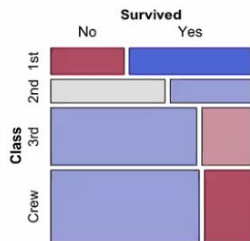
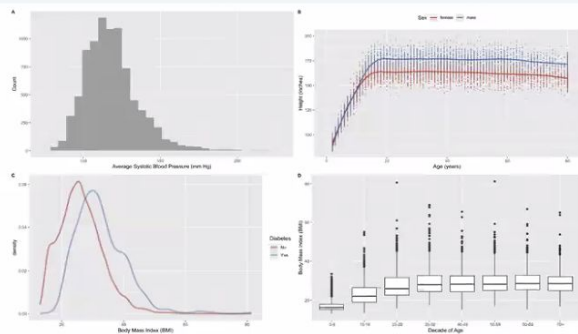
<https://www.nationalacademies.org/projects/DEPS-MSEB-25-01>



COMPETENCY 2

Producing & Working with Data

Measure, organize, explore, visualize



COMPONENTS

- Produce data: measurement & data provenance
- Organize data: case/attribute structure & data moves
- Exploration & data visualization

KEY INSIGHTS

- Data is actively “produced,” not simply collected — every measurement involves decisions about what and how
- Data provenance matters: who collected it, when, where, why, and who was left out?
- Organizing data through wrangling, filtering, and aggregation is essential before meaningful analysis

Takeaways from this talk

Barbara Means keynote

- ❖ Students are using gen AI extensively and in many different ways.
- ❖ While gen AI is influencing student course behaviors, the basic principles of how people learn remain applicable and have implications for effective instruction.
- ❖ Digital Promise research shows that use of learning sciences-based teaching practices in statistics course matters for affective engagement, conceptual learning, and grade outcomes.
- ❖ As gen AI use continues to grow and proliferate, uses that complement or enable active learning and metacognition are likely to improve learning while those that support efficiency at getting right answers may undermine meaningful learning.
- ❖ Course assessment and grading practices need to change in light of the availability of AI.
- ❖ We should be assessing reasoning and inquiry processes rather than number of correct answers.
- ❖ In the near future, gen AI tools may support more efficient assessment of student reasoning on an ongoing basis.

The 10 Recommendations in the 2025 GAISE College Report embody many of these strategies

<i>Strategy</i>	<i>Definition</i>
Active Learning	Practices that ask the learner to construct and/or apply knowledge
Collaborative Active Learning	Practices that call on students to construct/apply knowledge collaboratively with peers
Building on Prior Knowledge	Cues that activate students' background knowledge and experiences with relevance to course content
Formative Assessment	Opportunities to practice skills and receive timely targeted feedback
Supporting Metacognition	Practices that help students take control of the learning process and learn how to be a better learner
Transparency	Providing students with a clear description of course content, learning objectives, and criteria to be used in assessing their work
Fostering Belonging & Inclusion	Practices that enable all students to feel that they and their unique background are welcomed in the course
Data-informed instruction	Use of near real-time student learning data to inform teaching

See <https://amstat.quarto.pub/college-gaise/>

13

What has changed?

In the age of AI

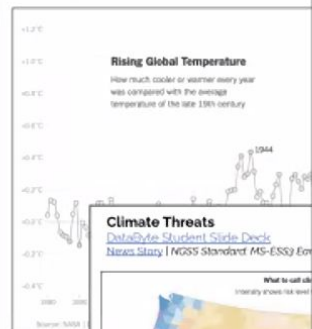
Rohan Alexander keynote

1. The importance of teaching skepticism.
 - Evals rule everything around me.
2. The importance of teaching editing and understanding code
 - “Learning Opportunities: A Claude Code and Codex Skill for Deliberate Skill Development” by Cat Hicks: <https://github.com/DrCatHicks/learning-opportunities>.
3. The importance of teaching students to push back.
 - AI helps most where the user knows least (Mythos model card, p. 27)!

Rising Global Temperature

[DataByte Student Slide Deck](#)

[News Story](#) | NGSS Standard MS - ESS3 Earth and Human Activity



Caption The 1880-1889

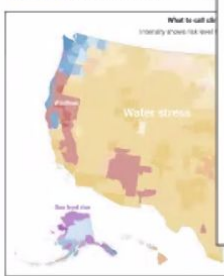
Making Sense

1. What is
2. What d

Climate Threats

[DataByte Student Slide Deck](#)

[News Story](#) | NGSS Standard MS-ESS3 Earth and Human Activity



Caption The map shows different climate threats across regions. States. Each color represents a different threat. Labeled on the map.

Making Sense

1. What is the first thing you think about or wonder when you see
2. What do you notice in the graph?

Potential teacher prompts: Are there any locations that are not susceptible?

Personal Connections

3. Do you have a personal connection with the data or patterns in why not? Which people or groups do you think would feel more connection with the data or patterns here?
4. Does this graph help you understand how climate affects you? Information would you want or need to better understand how it affects you?
5. Who do you think made this graph? Why did they collect this information?

Exploring Carbon Dioxide and Costs for Vehicles

[DataByte Student Slide Deck](#)

[News Story](#) | NGSS Standard MS - ESS3 Earth and Human Activity



Caption This graph

vehicle per mile, it.

Making Sense

1. What is this fr
2. What do you

Potential teacher prompts: Why do you think the data shows off?

Personal Connections

3. Do you have a
4. Why not? Why

Potential teacher prompts: Why do you think the connection is there?

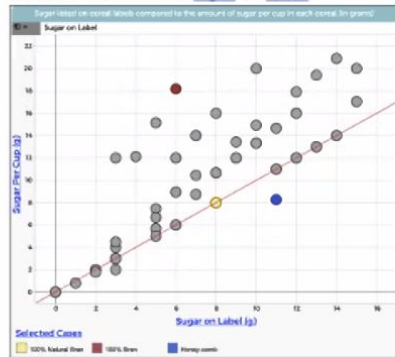
Examining Cereal Labels

[DataByte Student Slide Deck](#)

NGSS Standards: LS2: From Molecules to Organisms: Structures and Processes

Bonus: Engage with data for this DataByte as an [interactive CODAP file](#)!

Learn more about how to use CODAP with [this guide](#) or this [PDF video](#)



Caption The graph shows the amount of sugar listed on cereal labels compared to the amount of sugar contained in a cup of the cereals (in grams). If a cereal is on the line, then the sugars on the label match the sugars per cup. Three cereals are below; and on the line are colored.

Making Sense

1. What is the first thing you think about or wonder when you see these graphs?

Potential teacher prompts: What do you think the red line represents? Why do you think most of the cereals are above the line?

Michelle Wilkerson keynote

Making sense of the data.

Building personal connections to the data.

Reflecting on the context and history of the data.

Envisioning future uses of the data.

AI + Data Science Concepts/Interpretations

AI tools like ChatGPT make **A LOT of incorrect claims/interpretations about data science concepts.**

Tori Ellison (UIUC)

Common Conceptual Data Science AI Slop

Type	Professional Penalty	Course Penalty
<u>“ChatGPT linguistic style”</u> (ANY content written in this style) <i>Ex: Too many emdashes.</i> What counts as “ChatGPT linguistic style”? (Just like in your future career, it is incumbent upon <i>you</i> to know and keep up to date with what people think “looks” like “ChatGPT linguistic style” if you plan to use it.) 	- Even if characteristic “ChatGPT linguistic style” is used for non-technical content, reader loses some trust that you actually understood and/or looked over the technical work. - Potentially not the appropriate tone/audience for the analysis. - Overhyped tone can be misleading.	-1.5 points additionally deducted from assignment.
<u>“Provable” Unchecked AI Content Generation</u> <i>Ex: Leaving the AI prompt/response/instructions in a copy-paste “Sure, here is a k-means elbow plot below”.</i> <i>Ex: A hallucinated written response that is completely unrelated/misaligned with what other code/elements of your assignment says.</i> 	Reader loses pretty much all trust that you actually understood and/or looked over the technical work. The work could be full of errors, for all they know. Reading and using the insights from this work becomes very risky for the reader, if they do not wish to have a flawed understanding of the topic.	0 on the assignment (“drop the 2 lowest” doesn’t count). - The assignment counts as a “AI Slop Warning” (see tier system) Repeated “Provable” Unchecked AI Content Generation of this nature is subject to an academic integrity violation.

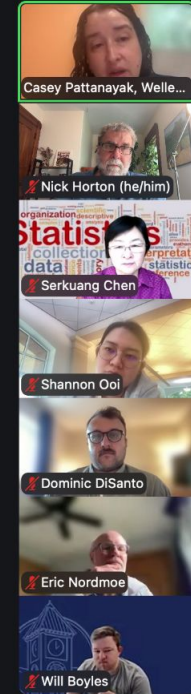
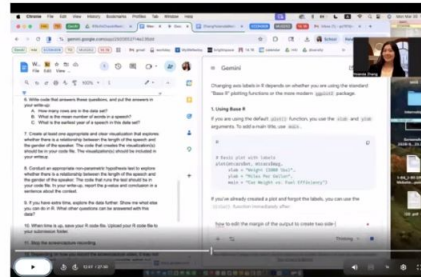
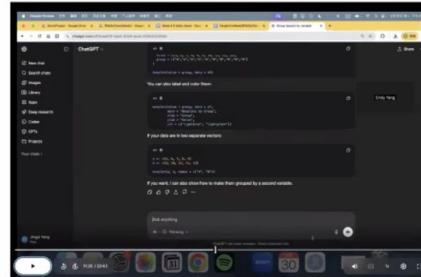
Revisions to course policies and structures

AI allowed for coding on homework

Screencapture videos of timed coding tasks, with AI allowed

Individual feedback on process

Analysis of video collection and student self-reflection in progress



Week 2: Using AI to Assist with Time Management

This is the actual lesson as students experience it — embedded in a course support structure, taking approximately **10 minutes of class time**. Here's how it unfolds:

- 1** Reflect on Successes & Challenges
Students briefly inventory what's working and what isn't from the previous week
- 2** Describe Their Weekly Schedule
Students articulate their full reality: classes, work, family, commute, and personal obligations
- 3** Engage the AI Tool
Students input a structured prompt describing their schedule and asking for a realistic study plan
- 4** Evaluate AI Recommendations
Students critically assess what the AI suggests — what's realistic, what isn't, and why
- 5** Select Strategies
Students choose which suggestions they'll actually implement based on their own judgment
- 6** Reflect on Implementation
Students write a brief reflection on what they tried and what they'd adjust going forward

Kadron Johnson (The Dana Center and The University of Texas at Austin) and Edith Aguirre (El Paso Community College)

Allison Davidson (Muhlenberg College)

What brings you Joy?



Photo Credit: Upsplash Joe Caione

Lightning talks: Community members were asked to share something that you are excited about that connects to the program

- Beth Chance (Cal Poly San Luis Obispo)
- Jessie Oehrlein (Fitchburg State University)
- Lisa Chan (The School District of Springfield Township, Pennsylvania)
- Michal Dvir (University of Haifa)
- Zachary del Rosario (Olin College)

Lightning Talk: Beth Chance (Cal Poly San Luis Obispo)

Choose Joy



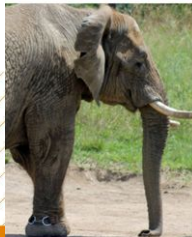
Example 1: *ISCAM*



INVESTIGATING STATISTICAL CONCEPTS, APPLICATIONS, AND METHODS

Beth L. Chance
Allan J. Rossman

Fourth Edition



**Investigating Statistical Concepts,
Applications, and Methods (4th edition, JMP/R,
pdf)**

\$ 5.00

**Investigating Statistical Concepts,
Applications, and Methods (4th edition, JMP/R,
print)**

\$ 20.00

Runestone

ISCAM: Investigating Statistical Concepts, Applications, and Methods

Beth L. Chance, Allan J. Rossman

Contents

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Front Matter

To the Student

Using Technology with This Book

1 Preliminaries

2 ISCAM Ch. 1 Intro

3 Analyzing a Process Probability

4 Normal Approximations for Sample Proportions

5 Sampling from a Finite Population

6 Chapter 1 Wrap-Up

7 ISCAM Ch. 2 Intro

8 Descriptive Statistics for Quantitative Data

9 Inference for a Population Mean

10 Inference for Other Statistics

11 Chapter 2 Wrap-Up

ISCAM: Investigating Statistical Concepts, Applications, and Methods

Beth L. Chance
Cal Poly - San Luis Obispo

Allan J. Rossman
Cal Poly - San Luis Obispo

Build: November 20, 2025 - 3:45 PM

To the Student

Using Technology with This Book

Carry out the simulation. Use the One Proportion Inference applet to carry out 1,000 repetitions of the simulation. Check the **Summary Statistics** box and report the mean and standard deviation of your simulated distribution of "could have been" outcomes.

Describe process:

Probability of heads: ☐ Summary Statistics
Number of tosses:
Number of repetitions:
☒ Show animation
☐ Draw Samples
Total Repetitions = 0

Choose statistic:

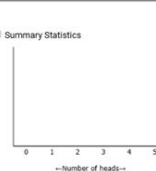
☒ Number of heads
☐ Proportion of heads

Count samples

As extreme as ☐ Show previous results
☐ Show sliders

Options:

☐ Two-sided
☐ Exact Binomial
☐ Normal Approximation



7. Summarize your results.

Mean: Standard deviation:

► Solution.

8. Evaluate Surprise Level.

Based on your simulation results, which assumes Yukon is guessing, should we be very surprised to see 6 correct guesses? Explain/support your reasoning.

► Hint.

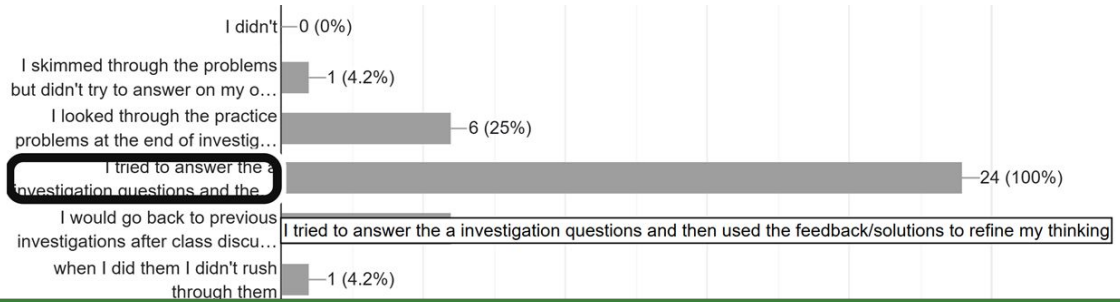
Write your answer here

► Solution.

Changed how I taught the course

Wednesday/Thursday - Factors that impact p-value (Day 4 [ppt](#) →)

- Inclass: Recap 1.4 and 1.5
- Activity 4 ([solutions](#) →)
- **Submit** [I Wonder question](#) for Day 4
- To Do before next class
 - Investigation 1.6 in Runestone



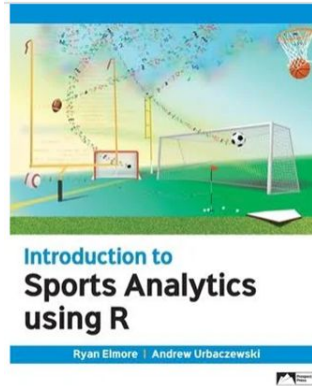
“Favorite: Built in applets + context for each investigations.
Least: lack of "demonstrations" prior to questions, forces you to "guess" and learn from the answer key.”

Example 2: Sports Analytics Course



SCORE Network

<https://scorenetwork.org>



Amanda Glazer,
UT Austin

ASA Statistics in Sports Section

What I needed/got

- Brainstorming
- New examples, contexts
- Programming best practices
- Pedagogical advice
- Better assessment questions
 - Example question: how are regularized adjusted plus minus (RAPM) and shrinkage estimation related, different?

Understanding the big ideas

Bias vs. Variability in Mean Squared Error

Why ridge-regularized RAPM can outperform unregularized APM

1 The MSE Decomposition

$$\text{MSE} = \text{Bias}^2 + \text{Variance}$$

$$\text{MSE}(\hat{\theta}) = \mathbb{E}[(\hat{\theta} - \theta)^2]$$

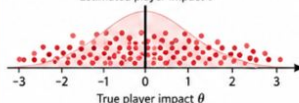
A good estimator balances being close to the truth on average (low bias) and being stable from sample to sample (low variance).

2 Two Estimators in Sports Analytics

Unregularized APM / OLS

Repeated samples (different lineups, games)

Estimated player impact $\hat{\theta}$



Low bias
High variability

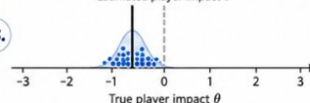
Can overreact to lineup noise, collinearity, and small-minute players.

VS.

Regularized RAPM / Ridge

Repeated samples (different lineups, games)

Estimated player impact $\hat{\theta}$



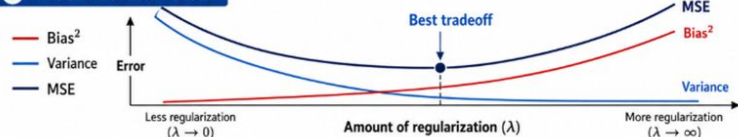
More bias
Lower variability

Shrinkage reduces instability from collinearity and sparse minutes.



In RAPM, a little bias can buy a big reduction in variance, lowering overall MSE.

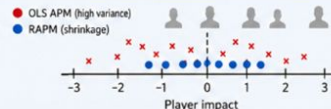
3 The Bias-Variance Tradeoff



Cross-validation chooses λ to minimize prediction error.



Sports analytics example: For players with limited minutes, OLS APM may give extreme estimates. RAPM shrinks those estimates toward average, improving stability and often reducing MSE.



New times, new skills

- Modelling effective use, iteration, refinement
- Asking better questions
- Statistical thinking
- Critical thinking
- Project management

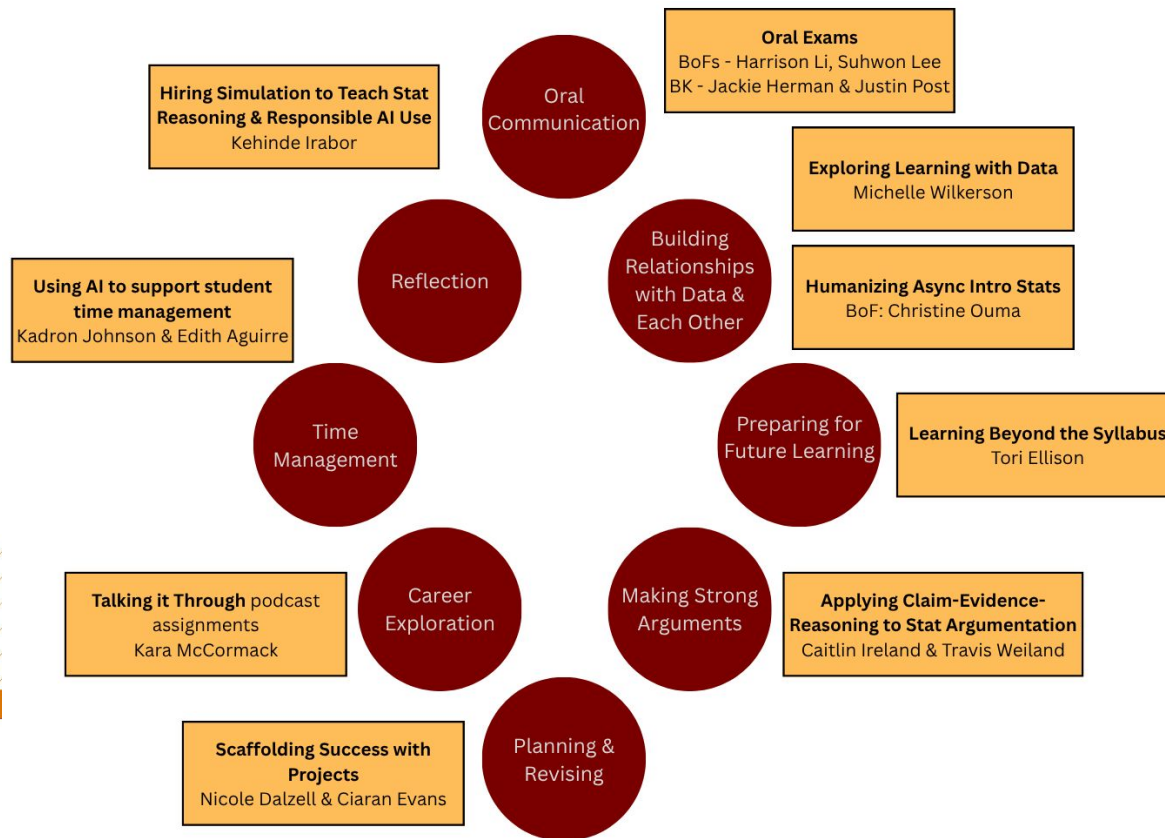
LET'S WORK TOGETHER!

Lightning Talk: Jessie Oehrlein (Fitchburg State University)

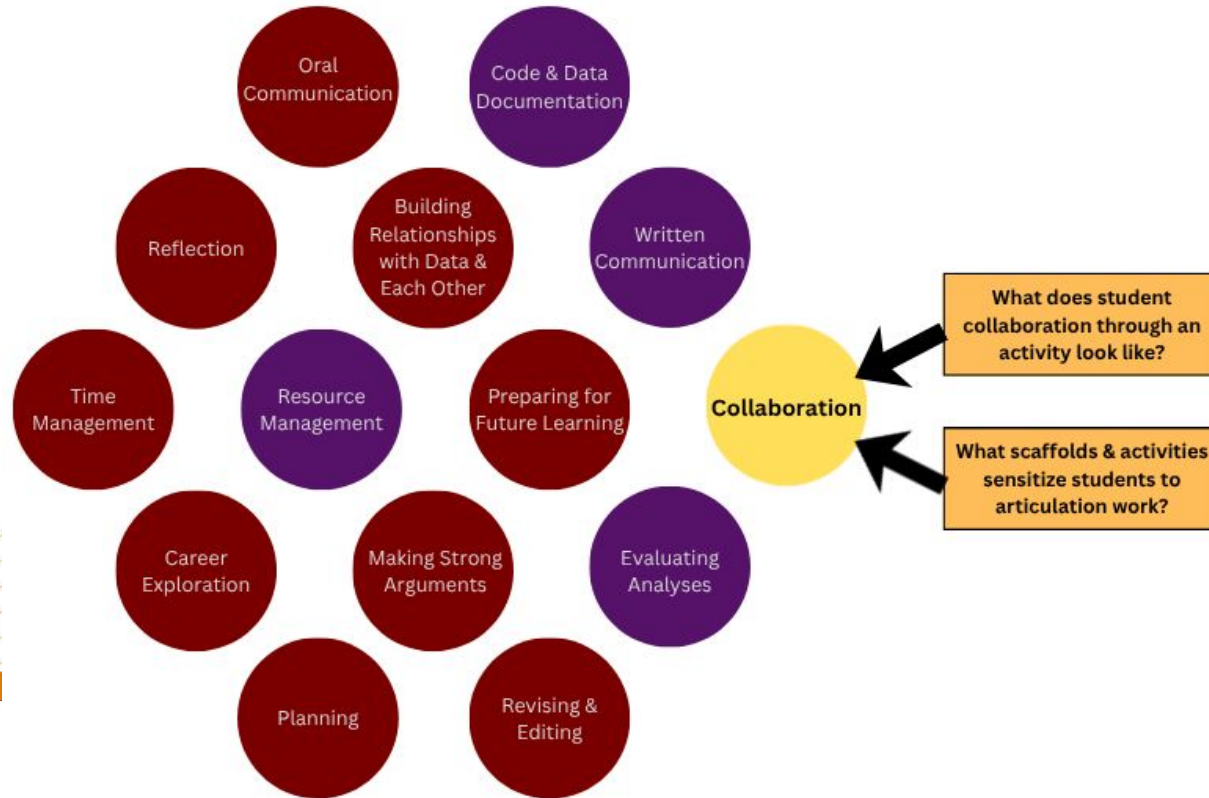


Discovering the Process Along
the Way

All the Process Along the Way



So many process skills to study & support!



Lisa Chan (The School District of Springfield Township, Pennsylvania)



Statistics Education

- Many statistics education resources including STEW
 - Lesson plans, activities and worksheets for various grade levels
- Data analysis vs. probability in statistics courses

Lightning Talk: Michal Dvir (University of Haifa)



Transformative bridges: Connecting
Student Joy with the Dynamic Data
Science Culture

Two different foci

Rohan Alexander
Telling Stories with Data
in the AI Age



Disciplinary
culture

Elizabeth Stuart &
Lance Waller
The Future of Statistics
and Data Science
Education

Two different foci



Students'
culture

Michelle Wilkerson
Exploring Learning with
Data Across the
Curriculum and Beyond

Barbara Means
Building Human
Intelligence in a World of
AI

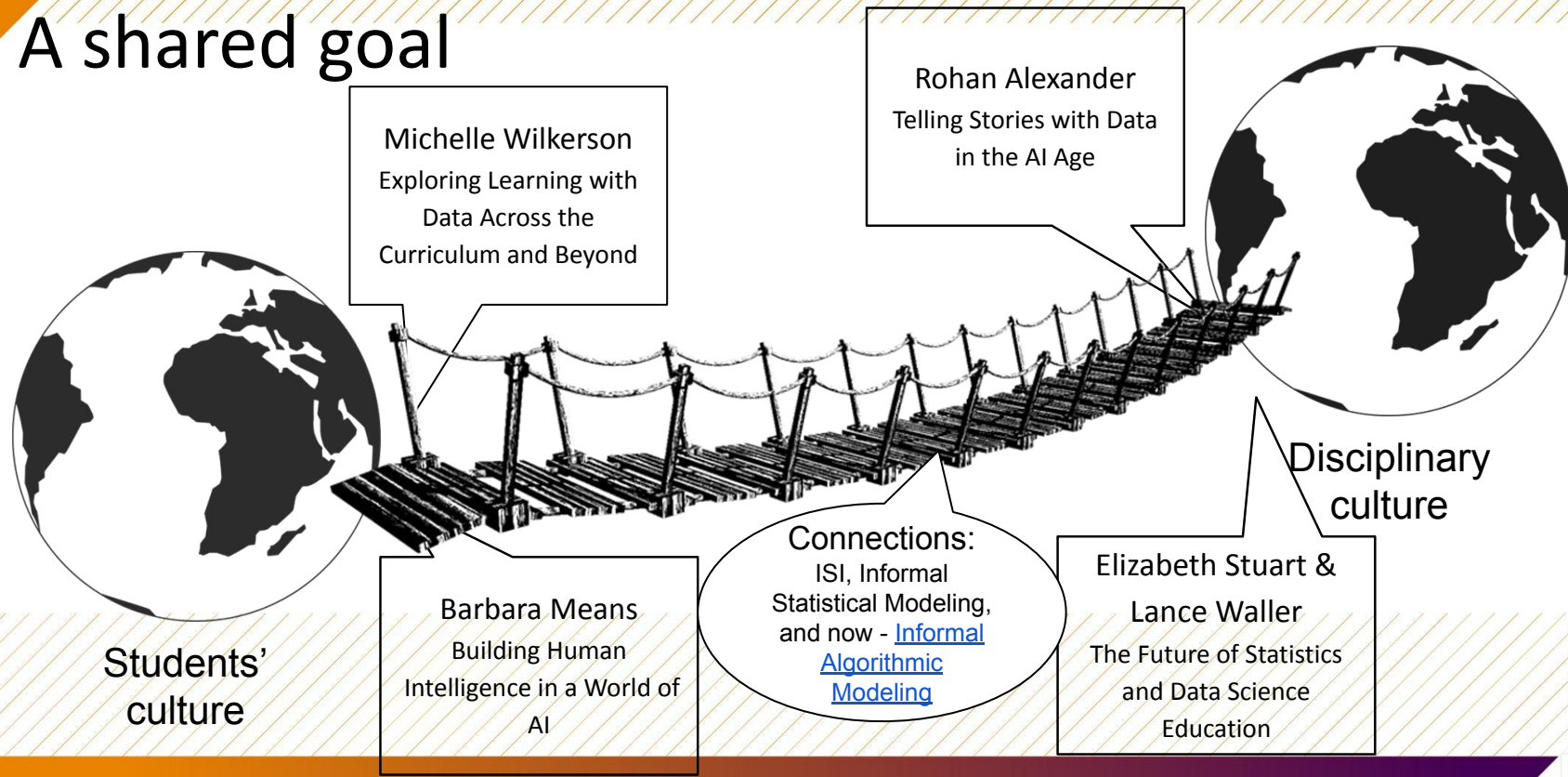
Rohan Alexander
Telling Stories with Data
in the AI Age



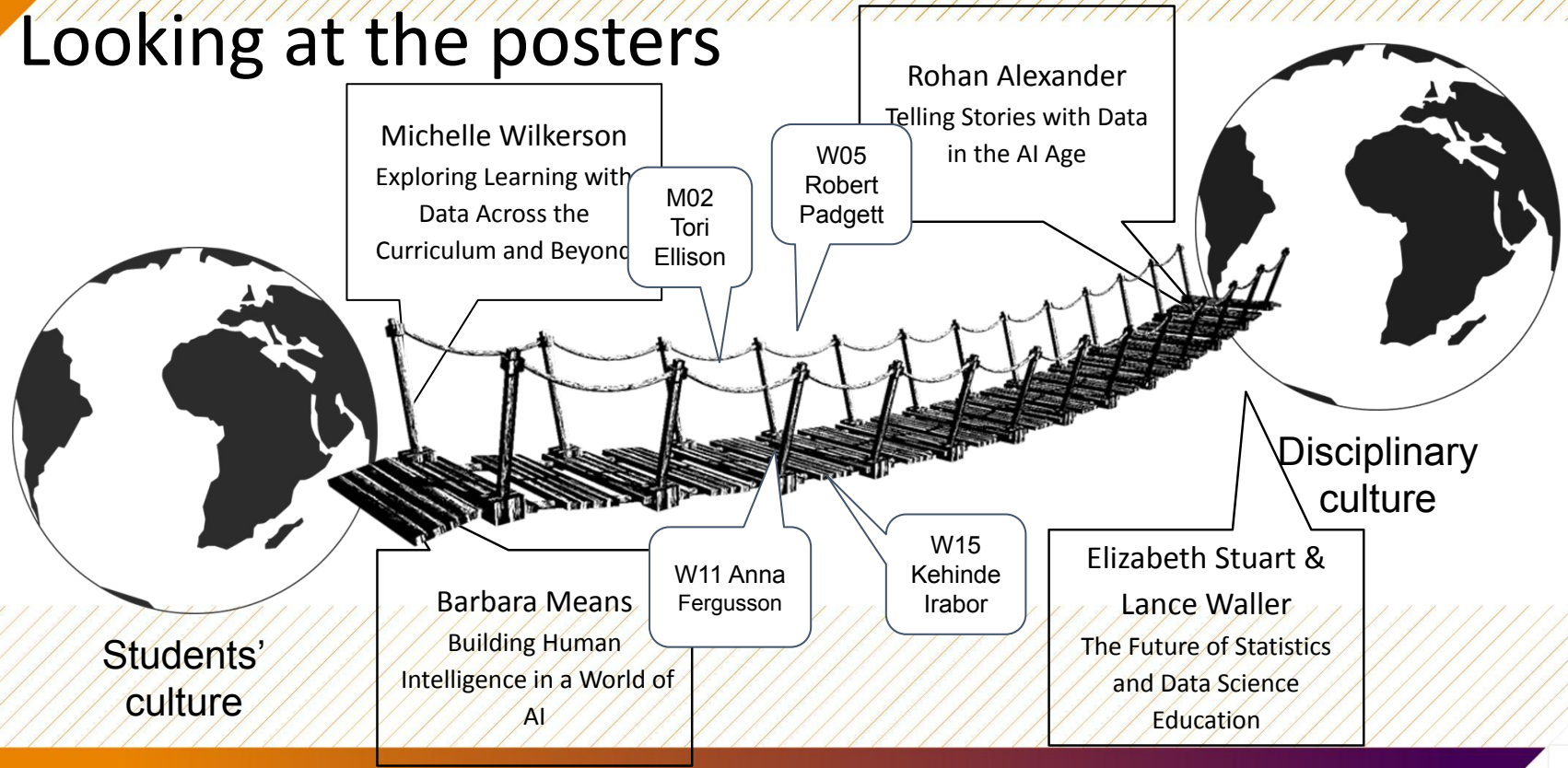
Disciplinary
culture

Elizabeth Stuart &
Lance Waller
The Future of Statistics
and Data Science
Education

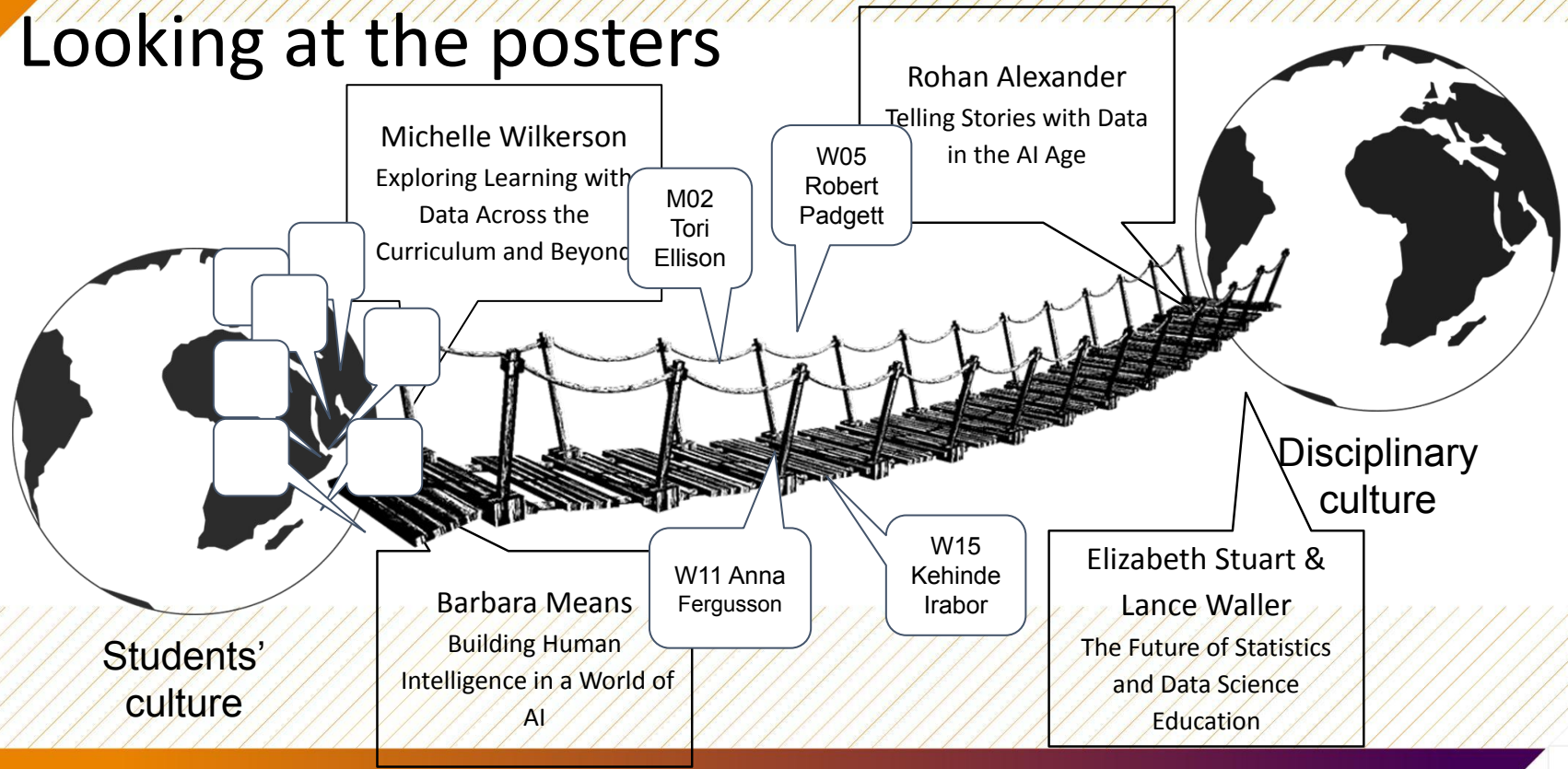
A shared goal



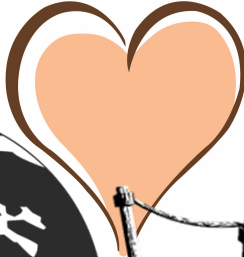
Looking at the posters



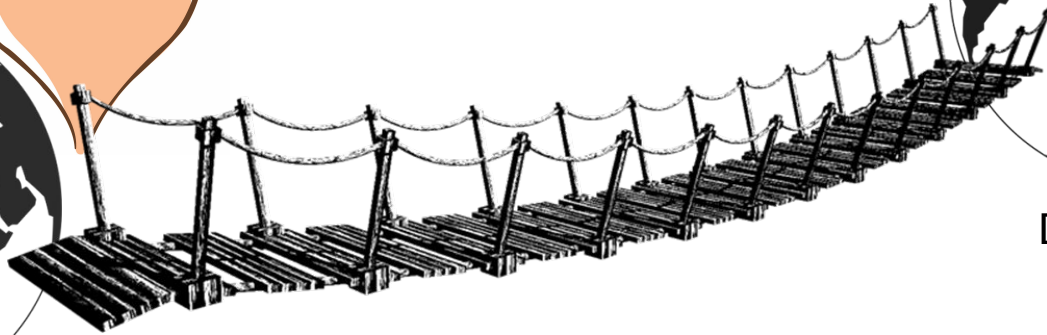
Looking at the posters



I wonder...



Students'
culture

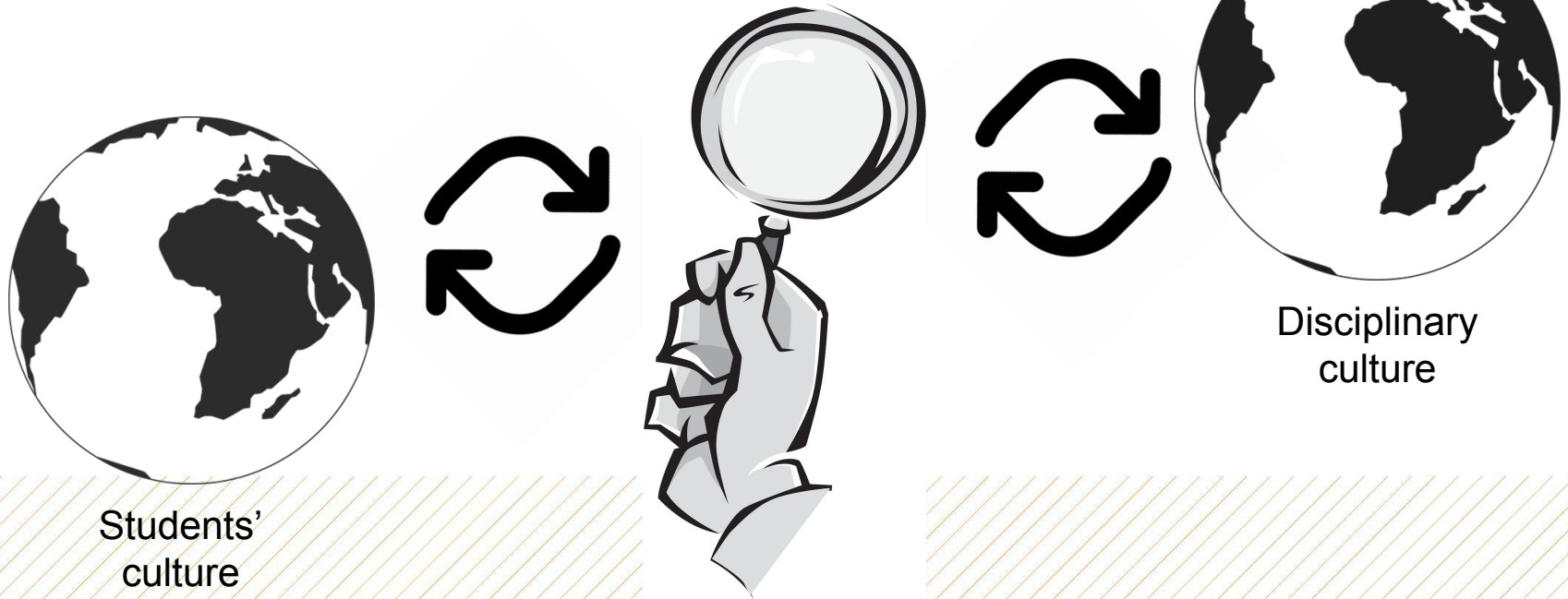


?



Disciplinary
culture

Transformative bridges: Connecting Student Joy with the Dynamic Data Science Culture

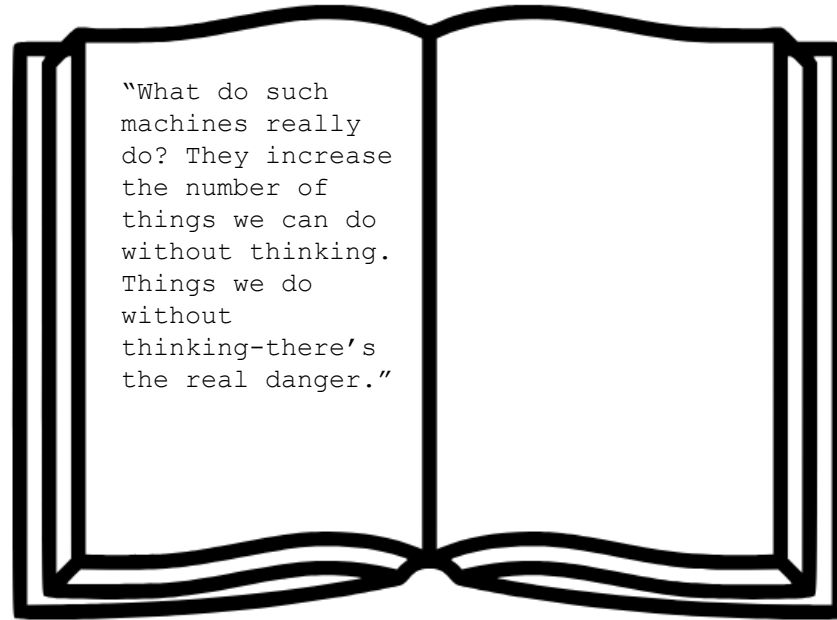


Lightning Talk: Zachary del Rosario (Olin College)



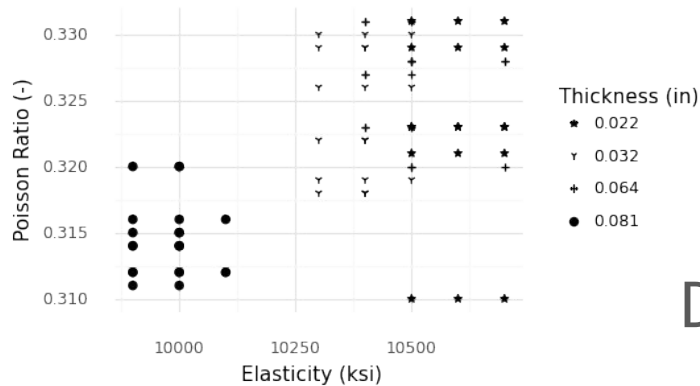
“What do such machines really do? They increase the number of things we can do without thinking. Things we do without thinking—there’s the real danger.”

— Frank Herbert



"What do such
machines really
do? They increase
the number of
things we can do
without thinking.
Things we do
without
thinking-there's
the real danger."

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DATA

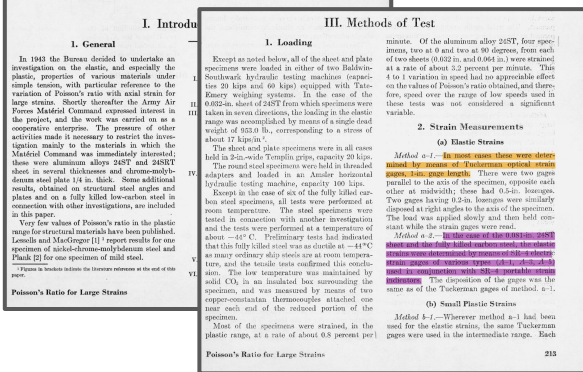
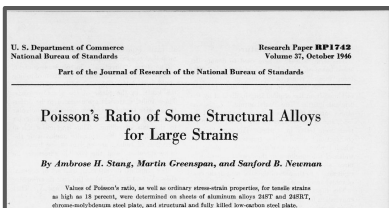
+

Human
Understanding

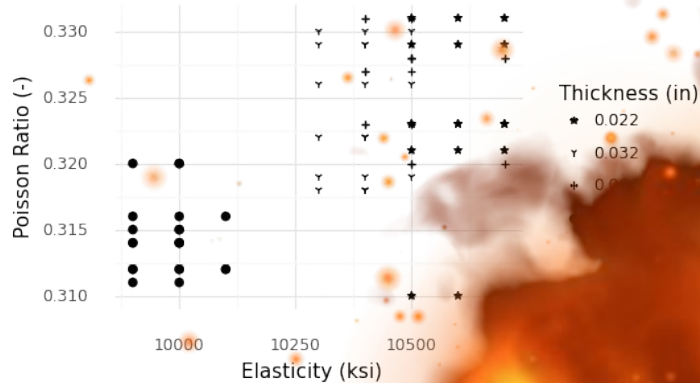


MEANING

CONTEXT



<https://tinyurl.com/stang03>



U. S. Department of Commerce
National Bureau of Standards

Research Paper **RP1742**
Volume 37, October 1966

Part of the Journal of Research of the National Bureau of Standards

Poisson's Ratio of Some Structural Alloys for Large Strains

By *Ambrose H. Stang, Martin Greenspan, and Sanford B. Neuman*

Values of Poisson's ratio, as well as ordinary stress-strain properties, for tensile strains as high as 18 percent, were determined on sheets of aluminum alloys 2024 and 248T, ducum-molybdenum steel plate, and structural and fully killed low-carbon steel plate.

I. Introduction

1. General

In 1943 the Bureau decided to undertake an investigation on the elastic, and especially the plastic, properties of various materials under simple tension, with particular reference to the variation of Poisson's ratio with axial strain for large strains. Shortly thereafter the Army Air Force Materiel Command, expressed interest in the project, and the work was carried on as a cooperative enterprise. The pressure of other Materiel Command was immediately interested; these were aluminum alloys 248T and 248N, steel in several thicknesses and ducum-molybdenum steel plate 1/4 in. thick. Some additional results, obtained on structural steel angles and plates and on a fully killed low-carbon steel in connection with other investigations, are included in this paper.

Very few values of Poisson's ratio in the plastic range for structural materials have been published. Lowth and MacGregor [1] report results for one specimen of nickel-chrome-molybdenum steel and Paul [2] for one specimen of mild steel.

¹ Figures in brackets indicate the literature reference at the end of this paper.

III. Methods of Test

1. Loading

Except as noted below, all of the sheet and plate specimens were loaded in either of two Baldwin-Buschway hydraulic testing machines (capacities 20 kips and 60 kips) equipped with Teflon-Energy weighing systems. In the case of the 0.022-in. sheet of 248T from which specimens were taken in seven directions, the loading in the elastic range was accomplished by means of a single dead weight of 552.0 lb., corresponding to a stress of about 17 kips/in.².

The sheet and plate specimens were in all cases held in 2-in.-wide Teflon grips, capacity 20 kips. The round steel specimens were held in threaded adapters and loaded in an Ancker horizontal cylindrical testing machine, capacity 100 kips.

Except in the case of six of the fully killed carbon steel specimens, all tests were performed at room temperature. The steel specimens were tested in connection with similar investigations and the tests were performed at a temperature of about -42° C. Preliminary tests had indicated that this fully killed steel was as ductile at -44° C as at room temperature. The low temperature was maintained by solid CO₂ in an insulated box surrounding the specimens, and was measured by means of two resistance-temperature thermocouples attached one near each end of the reduced portion of the specimen.

Most of the specimens were strained, in the plastic range, at a rate of about 0.5 percent per minute. Of the aluminum alloy specimens, two were strained at a rate of about 0.2 percent per minute, two at 0 and two at 10 percent per minute. The two sheets of 0.022 in. and 0.062 in. thickness were strained at a rate of about 3.2 percent per minute. The 4 to 1 variation in speed had no appreciable effect on the values of Poisson's ratio obtained, and therefore, speed over the range of low speeds used in these tests was not considered a significant variable.

2. Strain Measurements

a) Elastic Strains

Method a-1.—In most cases these were done using by means of Tuckerman optical strain gages, thin, non-ferrous. There were two gages parallel to the axis of the specimen, opposite each other at midlength. These had 0.5-in. long legs. Two gages having 0.2-in. long legs were similarly disposed at right angles to the axis of the specimen. The load was applied slowly and then held constant while the strain gages were read.

Method a-2.—In the case of the 0.022-in. sheet and the fully killed carbon steel, the elastic strains were determined by means of the Tuckerman gages of various types, which were held in the specimen by means of small metal clips.

Method a-3.— The disposition of the gages was the same as for the Tuckerman gages of method a-1.

b) Small Plastic Strains

Method b-1.—Wherever method a-1 had been used for the elastic strains, the same Tuckerman gages were used in the intermediate range. Each



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MEANING??

Self Actualization

OR

JOB PREPARATION

OR

Civic Engagement



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Cognitive



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OR

Social



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OR

Liberatory



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This "or" perspective limits from
being bold

Here's my perspective:

Hi! I'm Zach del Rosario
Associate Professor of
Engineering and Applied Statistics

...

... but what are you *really*?

☒ Asian / Pacific Islander

☒ White

☐ Engineer

☐ Statistician

... why only one?

Cognitive



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AND

Social



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AND

Liberatory

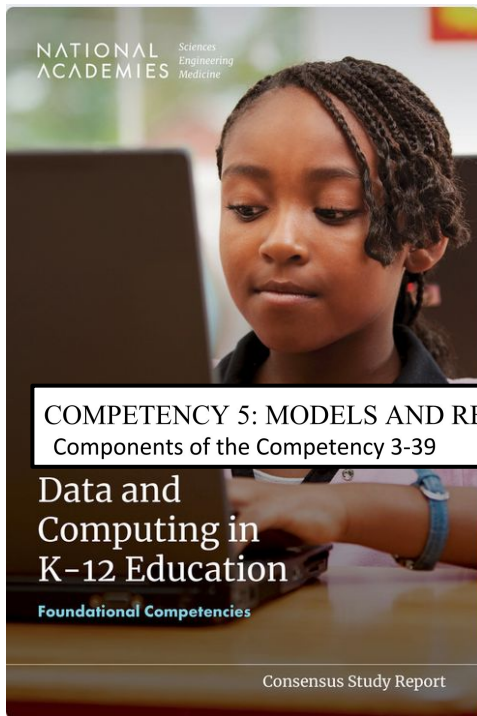


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... how can we put these together?

... can I tell you about my dream?

It starts with a *movement*...



COMPETENCY 5: MODELS AND REPRESENTATIONS 3-35

Components of the Competency 3-39

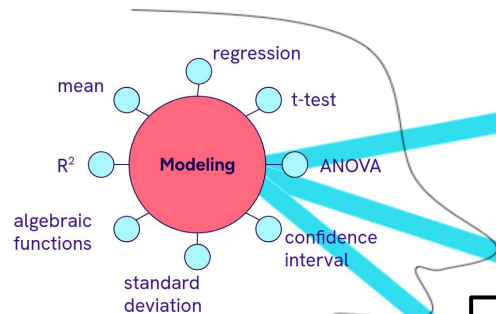
Data and Computing in K-12 Education

Foundational Competencies

Consensus Study Report

Modeling is a movement

Principles: 7. Modeling authentic applications



Models as practicing connections



The Summit will drive national reform in undergraduate



pathways, and a roadmap for nationwide implementation.



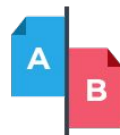
May 18 - 20, 2026



HARVARD UNIVERSITY

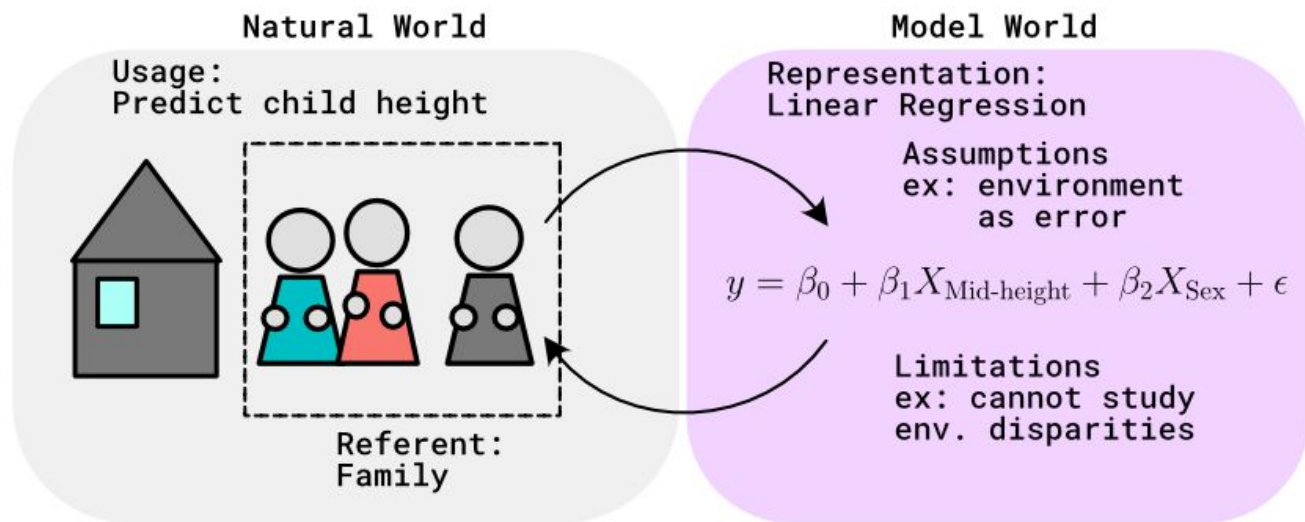
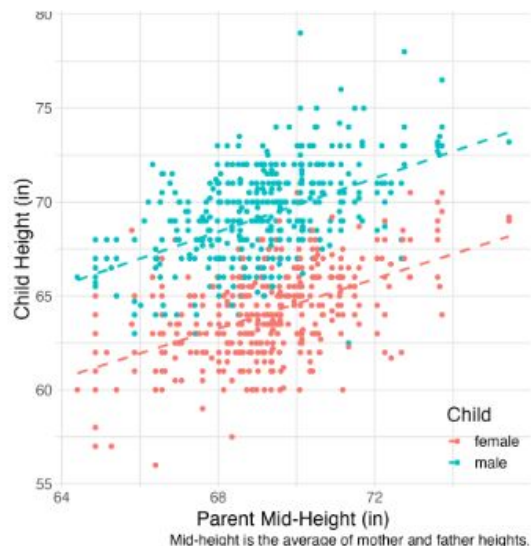


Visit Our Website
www.tpsemath.org



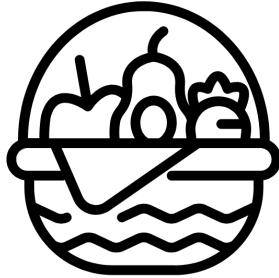
CourseKata

My goal with modeling:
combine **cognitive** and **humanist** outcomes



So *multidisciplinary*,
right?

Multi-disciplinary



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Inter-disciplinary



Created by Arthur Shlain
from Noun Project

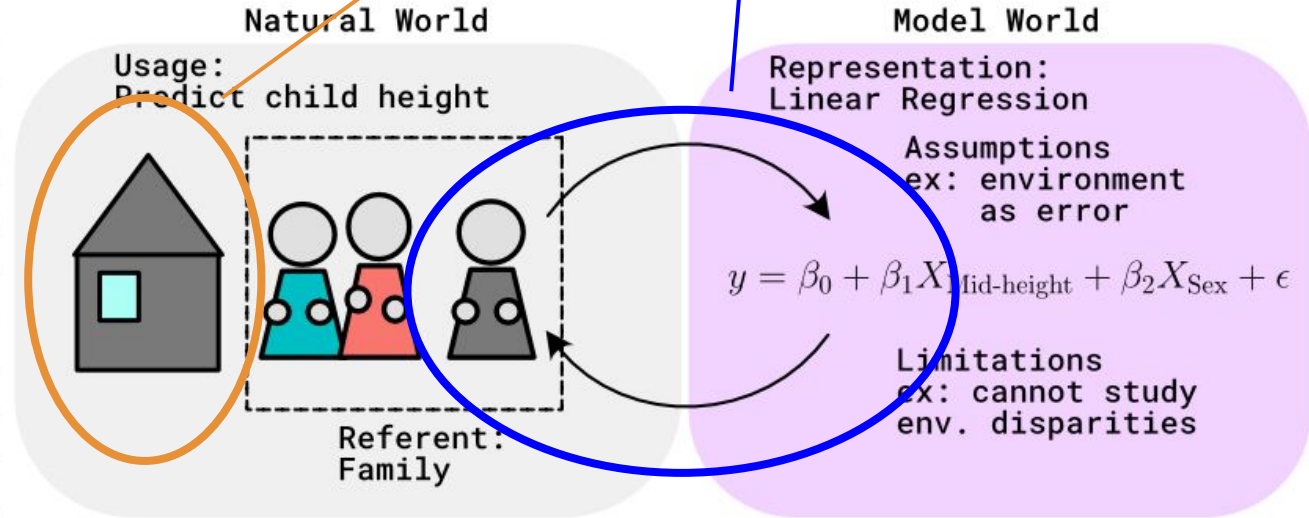
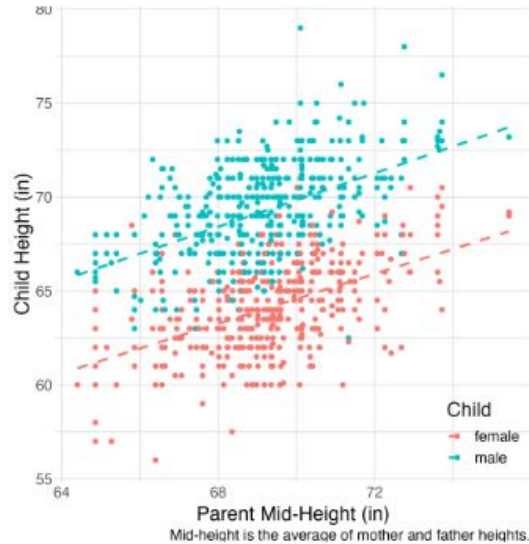
Trans-disciplinary



Created by Febrian Hidayat
from Noun Project

Transdisciplinary teaching (Tear down boundaries)

Make it obvious the **cognitive**
belongs with the **humanist**



The way we teach can reinforce boundaries

Day 1
Math

Day 2
Math

Day 3
Math

Day 4
Ethics

Day 6
Math

Day 7
Math

Day 8
Math



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"I guess this is
a **math** class."

"Wait, what
was that?!"

My invitation:

What ideas would you want to *tie* together?
And *how* could you do that?

Day 1



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"We're going to learn
some cool **technical**
stuff!"

Day 2



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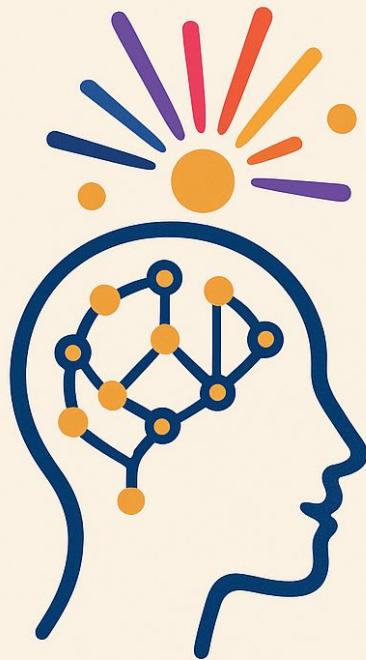
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from Noun Project

"How could you ever
do the **analysis**
without the **context**?"



Why Now in a World of AI?

- Make new meaning(s) no AI could produce
- Resist techno-solutionism
- Show the joy of statistics in *every* backyard



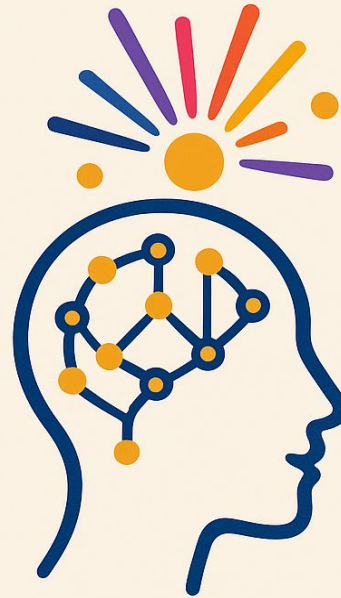
**Sparking
Joy and
Discovery**

In a World of AI
eCOTS 2026

Next steps from here

JSM Birds of a Feather
Sessions (virtual ones
do not require JSM
registration):

Sign Up for Updates: <https://bit.ly/JSM26BOF>

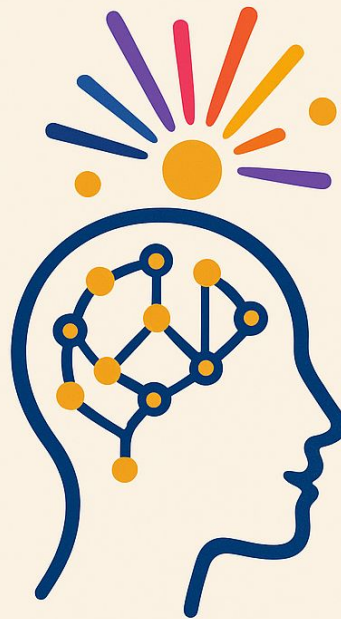


**Sparking
Joy and
Discovery**

**In a World of AI
eCOTS 2026**

Journals + Publication

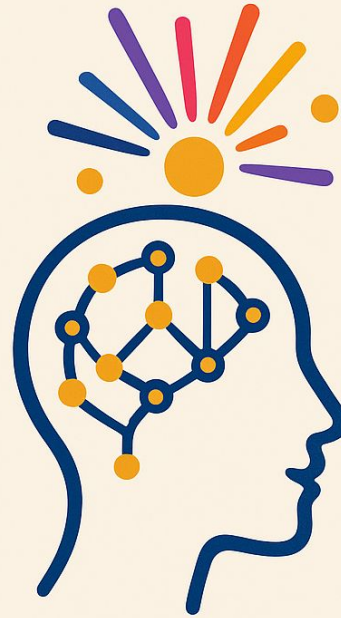
- Journal of Statistics and Data Science Education
- Teaching Statistics
- Harvard Data Science Review
- Statistics Education Research Journal
- Technology Innovations in Statistics Education



**Sparking
Joy and
Discovery**
In a World of AI
eCOTS 2026

Next steps from here

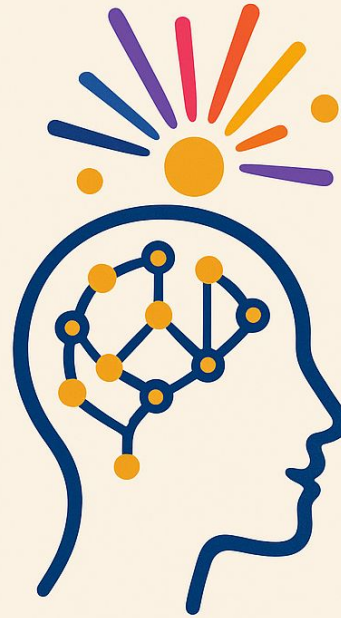
CAUSE Slack
Workspace (posts
expire after some time
but a chance to
interact)



**Sparking
Joy and
Discovery**
In a World of AI
eCOTS 2026

Closing Thoughts

Matt Beckman
(CAUSE Director)



**Sparking
Joy and
Discovery**

In a World of AI
eCOTS 2026

USCOTS 2027: June 22nd - 26th

- **Timing**
 - **Main conference:** Lunch on Thursday June 24th through lunch on Saturday June 26th
 - **Workshops:** Tuesday, June 22nd through morning of Thursday June 24th
 - **Research Satellite:** Evening of Wednesday June 23rd through morning of Thursday June 24th
- **Location:** The Penn Stater, State College, PA
- And, the **theme** will be...



USCOTS 2027 Theme: Constancy and Disruption

In an age of disruption, what **guiding principles** do you aim to hold constant amid the disruption?



On the flipside, how do you embrace the disruption to **transform** your teaching and education research?



Want to get involved in planning USCOTS 2027?

Reach out to Kelly McConville (k.mcconville@bucknell.edu), Allan Rossman (arossman@calpoly.edu), or Matt Beckman (mdb268@psu.edu).

“Ask Good Questions” Symposium on July 14 - 15, 2027 at Cal Poly - San Luis Obispo, CA

- Program:
 - Wed, July 14th: Dinner banquet
 - Thu, July 15th: Talks
- Hosted in honor of Allan Rossman
- Organized by the Statistics Department of Cal Poly - San Luis Obispo, CA
- Contact Beth Chance (bchance@calpoly.edu) for more information
- Hope to see you there!



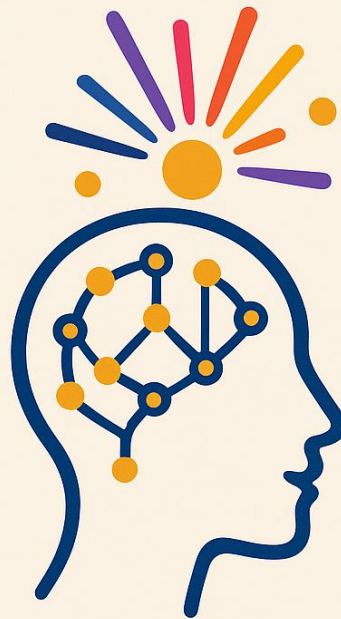
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- Hope to see you there!



Electronic Conference on Teaching Statistics (eCOTS) 2026

Hope to see you next
summer!



**Sparking
Joy and
Discovery**
In a World of AI
eCOTS 2026