

Using Assessment System Data to Generate Individualized Learning Materials for Students

What We Learned from Developing the iDAP

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Overview

- 1 Background
 - AP-CAT
 - iDAP
- 2 Lessons learned
 - Teacher Access & Teacher Engagement
 - Plan for Heterogeneity
 - Toolchain
 - Quality Control
 - Process Data
- 3 Future
 - Learning Modules

Background

- ❖ Full Name: “Cognitive Diagnostic Computerized Adaptive Testing (CD-CAT) for AP Statistics”
- ❖ A five-year CAREER project funded by the National Science Foundation
- ❖ Two major components
 - ❖ Research
 - ❖ Educational outreach

Development



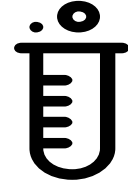
Item Bank
Web Platform

Refinement



Mobile
Diagnostic Score Report
Pilot Data

RCT



Feedback Types
Learning Outcomes



Testing



Bugs
UX/UI
Item Keys

Observation



Engagement Survey
Item Calibration
Adaptive Testing

Item Bank

- ❖ 850 items
 - ❖ 4 sections
 - ❖ 16 main topics
 - ❖ 158 attributes

All Questions

#	Question	Creator	Type	Status	Attributes	Time Created	Operation
1	Temperature Boxplot Median Qid: 1	Alison_Cheng	Multiple Choice	Updated	Statistics - 1.C.1.	Jul 28, 2014 - 10:48 p.m.	Review Edit Delete
2	Temperature Boxplot Percentile Qid: 2	Alison_Cheng	Multiple Choice	Approved	Statistics - 1.B.17.	Jul 29, 2014 - 10:11 a.m.	Review Edit Delete
3	Temperature Summary Outlier 1 Qid: 3	Alison_Cheng	Blank Field	Updated	Statistics - 1.B.5.	Jul 29, 2014 - 10:14 a.m.	Review Edit Delete
4	Temperature Summary Outlier 2 Qid: 4	Alison_Cheng	Blank Field	Updated	Statistics - 1.B.5.	Jul 29, 2014 - 10:16 a.m.	Review Edit Delete
5	Empirical Rule Area Qid: 5	Alison_Cheng	Blank Field	Updated	Statistics - 3.C.1.	Jul 29, 2014 - 10:18 a.m.	Review Edit Delete
6	Donut Probability 1 Qid: 6	Alison_Cheng	Blank Field	Updated	Statistics - 1.B.10.	Jul 29, 2014 - 10:21 a.m.	Review Edit Delete
7	Donut Probability 2 Qid: 7	Alison_Cheng	Blank Field	Updated	Statistics - 3.A.4.	Jul 29, 2014 - 10:38 a.m.	Review Edit Delete



Toolbox



Help



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Assignment Group:

- ☒ AlexMockClass
☐ AlexMockClass2

Is the order of questions random? ☒

Due Date: 11/06/2019

Due Time: 12:00 AM

Optional tag for special survey load: Default is empty

- ▼ Exploring Data
 - ▼ Comparing distributions of univariate data (dotplots, back-to-back stemplots, parallel boxplots)
 - ☒ Comparing and interpreting centers
 - ☐ Comparing and interpreting spreads
 - ☒ Comparing and interpreting shapes
 - ☒ Comparing and interpreting clusters/gaps
 - ☐ Comparing and interpreting outliers
 - ▶ Constructing and interpreting graphical displays of distributions of univariate data
 - ▼ Exploring bivariate data
 - ☒ Identifying shape, direction, strength of a Scatterplot
 - ☐ Knowing properties of a Correlation coefficient
 - ☒ Interpreting coefficient of determination: R square
 - ☐ Writing Least Squares Regression Line (LSRL) from computer output
 - ☒ Obtaining LSRL from dataset
 - ☐ Writing LSRL from summary statistics (using $b1=r*(sy/sx)$ formula) (not likely)
 - ☐ Interpreting slope
 - ☐ Interpreting intercept
 - ☐ Finding a predicted y-hat
 - ☐ Finding a residual value
 - ☐ Interpreting a residual
 - ☐ Using residual plot to determine if transformation is needed, i.e., is linear model appropriate
 - ☐ Understanding lower reliability in extrapolation
 - ☐ Identifying influential points and their impact on model (leverage)
 - ☐ Identifying correct re-expression (log, power, exponential)
 - ☐ Correctly calculating a predicted value under a transformation (log/power/exponential)
 - ▶ Exploring categorical data
 - ▶ Summarizing distributions of univariate data
 - ▶ Sampling and Experimentation
 - ▶ Randomness, probability and simulation
 - ▶ Statistical Inference

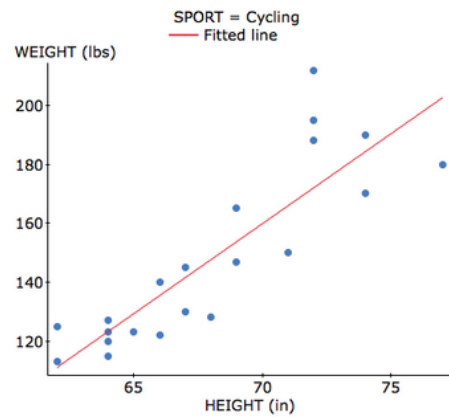


Question 1/1

Show All

Save and resume later

The heights and weights of the 2016 USA Olympic cycling team are graphed below. We can predict an Olympian's weight from their height using the least squares regression equation below:

$$\text{WEIGHT} = -267 + 6.1 \text{ HEIGHT (in)}$$


Interpret the y-intercept in context.

- A. ☐ This model predicts that a cyclist who is zero inches tall will weigh around -267 pounds.
- B. ☐ This model predicts that a cyclist who is zero inches tall will weigh around 110 pounds.
- C. ☐ This model predicts that a cyclist who weighs zero pounds will be -267 inches tall.
- D. ☐ This model predicts that a cyclist who weighs zero pounds will be around 110 inches tall.
- E. ☐ This model predicts that a cyclist who weighs zero pounds will be around 61 inches tall.

Next

Scaffolding

- ❖ Step-by-step Solutions
 - ❖ Visible after assignment completion



Aerobics Class IQR Interpretation

The middle 50% of ages of aerobics participants has a range of 9.

The middle 25% of ages of the aerobics participants is between 23.5 and 28.5. 3 s

In a random sample of 24 people in an aerobics class, their ages are given as follows:

21 38 32 29 27 27 46 18 24 23 30 31
28 20 35 34 31 29 19 48 28 25 22 33

Explain what the interquartile range means for these data.

- Fifty percent of the time, the age of an aerobics participant is between 23.5 and 32.5 years old.
- The interquartile range is the average range of the ages of participants in this aerobics class.
- The middle 25% of ages of the aerobics participants is between 23.5 and 28.5.
- The interquartile range is the participants' average distance away from the mean age.
- ✓
The middle 50% of ages of aerobics participants has a range of 9.

See Scaffolding Solutions



Outlier Sensitivity

Standard Deviation

✓
Standard Deviation

3 s



House Prices Y-Intercept

39.4

✓
39.4

8 s

Aerobics Class IQR Interpretation

The middle 50% of ages of aerobics participants has a range of 9.

The middle 25% of ages of the aerobics participants is between 23.5 and 28.5.

3 s

Solution (1/4)

1. Find the first quartile and the third quartile:

- Split the data into two halves.
- Since there are 24 numbers in the dataset and 24 is an even number, each half will have 12 numbers.
- Q1 is the median of the lower half and Q3 is the median of the upper half. Each half has an even number of data points, so the median of the lower half is

$$Q_1 = \frac{23 + 24}{2} = 23.5, \text{ and the median of the upper half is } Q_3 = \frac{32 + 33}{2} = 32.5.$$

- ❖ Granular feedback
 - ❖ Areas of opportunity



Home



My tests



Help



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Test results: 2017 - Mock Assignment 1 - Form A

Back to my tests

Global results

Attributes

Detailed results

Below are displayed the different levels of attribute performance for this assignment

- represents attributes that you are proficient in.
- represents attributes where your skills are still developing.
- represents attributes that you need to work on more.

Comparing distributions of univariate data (dotplots, back-to-back stemplots, parallel boxplots) **INTERMEDIATE**

- Comparing and interpreting outliers : 100.0%
- Comparing and interpreting centers : 50.0%
- Comparing and interpreting spreads: 100.0%
- Comparing and interpreting shapes: 50.0%

Exploring bivariate data **INTERMEDIATE**

- Knowing properties of a Correlation coefficient: 50.0%
- Interpreting coefficient of determination: R square : 100.0%
- Writing Least Squares Regression Line (LSRL) from computer output: 100.0%
- Obtaining LSRL from dataset: 100.0%
- Writing LSRL from summary statistics (using $b1=r*(sy/sx)$ formula) (not likely): 0.0%
- Interpreting a residual: 0.0%

Exploring categorical data **NOVICE**

- Knowing the Difference between categorical and quantitative variables: 50.0%
- Calculating Joint frequencies for two-way tables: 0.0%

- ✦ Intelligent Diagnostic Assessment Platform (i-DAP) for High School Statistics Education
 - ✦ GOAL: Develop a holistic, personalized, learning system integrated into the classroom

- Leverage attribute system developed in the AP-CAT
 - Map to 36 common core state standards on “statistics and probability”



Lessons learned

Lesson 1

Teacher Access & Teacher Engagement

Teacher Access & Teacher Engagement

- ❖ Teacher approval
 - delivery engine
 - reporting
 - feature requests
 - rollout of assignments
 - Example: In-Class Assignment Length

Lesson 2

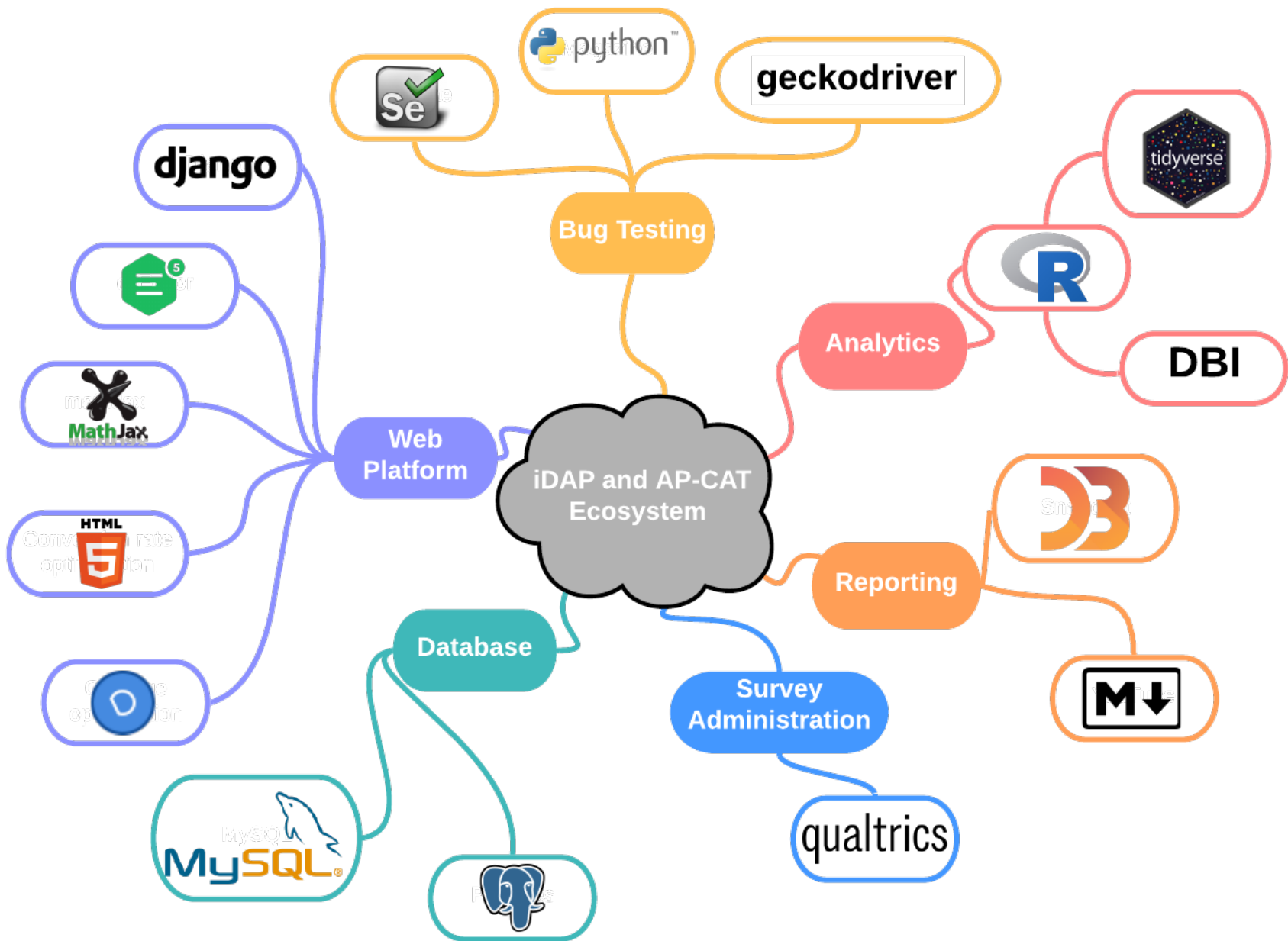
- Plan for Heterogeneity
 - in learning environment
 - in students

Sample Heterogeneity

- ❖ Not surprising
 - ❖ learning outcomes depend on school
- ❖ Surprising
 - ❖ Student self-predictions of their final exam scores depend on school (Ober et al., In Prep)
 - ❖ Patterns of incorrect responses

Lesson 3

Choose your toolchain wisely



```
1 con <- DBI::dbConnect(RMySQL::MySQL(), ## Connect to DB
2   host = "domain.name.edu",
3   port = 3306,
4   user = "db_user",
5   password = "db_pw",
6   dbname = "db_name")
7
8 tbl(con, "activity") %>%
9   filter(role == "student") %>% ## Only student data
10  select(-firstname, -lastname, ## de-identify
11    -role, -username,
12    -id, -meta)
13
```

Lesson 4

Have Quality Control and Contingency Plans

Lesson 4

- ❖ Not all data are worth being collected
- ❖ Not all data that have been collected are worth being analyzed

```
# Source: lazy query [?? x 4]
# Database: mysql 5.7.27-0ubuntu0.18.04.1
# [user@domain.name:/dname]
  url          tabname          t          info
  <chr>        <chr>          <chr>      <chr>
1 /results/461 results(student) 2019-09-25 11:30:28 <NA>
2 /results/461 results(student) 2019-09-25 11:30:30 <NA>
3 /results/461 student_info_btn 2019-09-25 11:30:34 "btn":"info","qid":
4 /results/461 student_info_btn 2019-09-25 11:30:36 "btn":"info","qid":
5 /results/461 student_info_btn 2019-09-25 11:31:30 "btn":"info","qid":
6 /results/461 student_info_btn 2019-09-25 11:31:38 "btn":"info","qid":
7 /results/461 Attributes(student) 2019-09-25 11:33:20 <NA>
8 /results/461 results(student) 2019-09-25 11:33:29 <NA>
9 /results/787 results(student) 2019-09-25 11:43:31 <NA>
10 /results/787 results(student) 2019-09-25 11:46:55 <NA>
# ... with many, many, many more rows
```

Lesson 4

- ❖ Quality Control
 - ❏ Dummy items
 - ❏ Response time data (Qualtrics hack)
 - ❏ Open text responses to surveys
- ❖ Contingency Plan
 - ❏ End users are used to having "delete" mean "put in trash"
 - ❏ Choosing not to take the exam

Lesson 5

System paradata (Process data) can provide a wealth of information

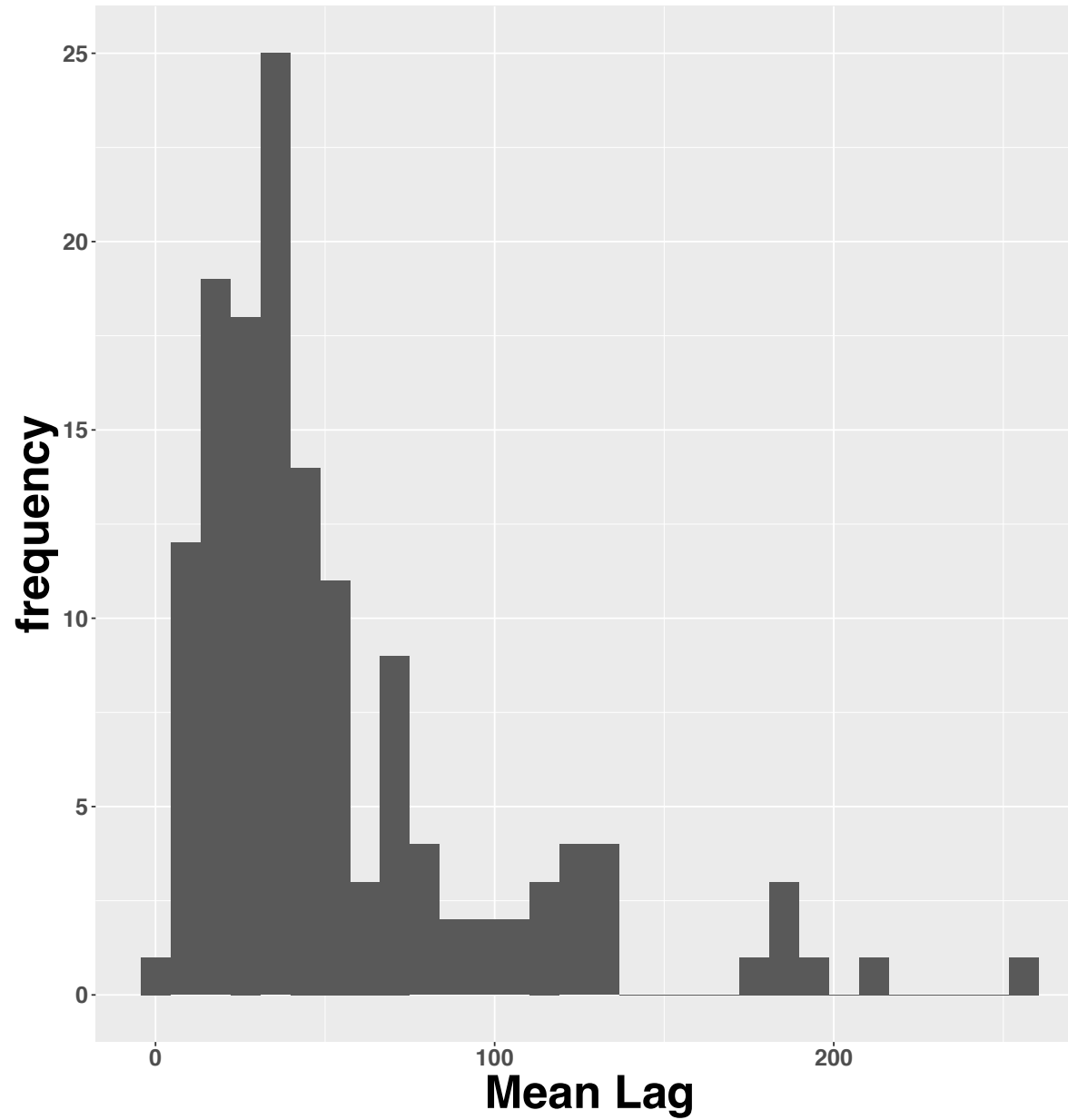
Example: Procrastination

- Active Procrastination (Chu & Choi, 2005)
 - Criticisms (e.g. Delay?) (Krause & Freund, 2014)
- Attempting to validate a survey scale via process data

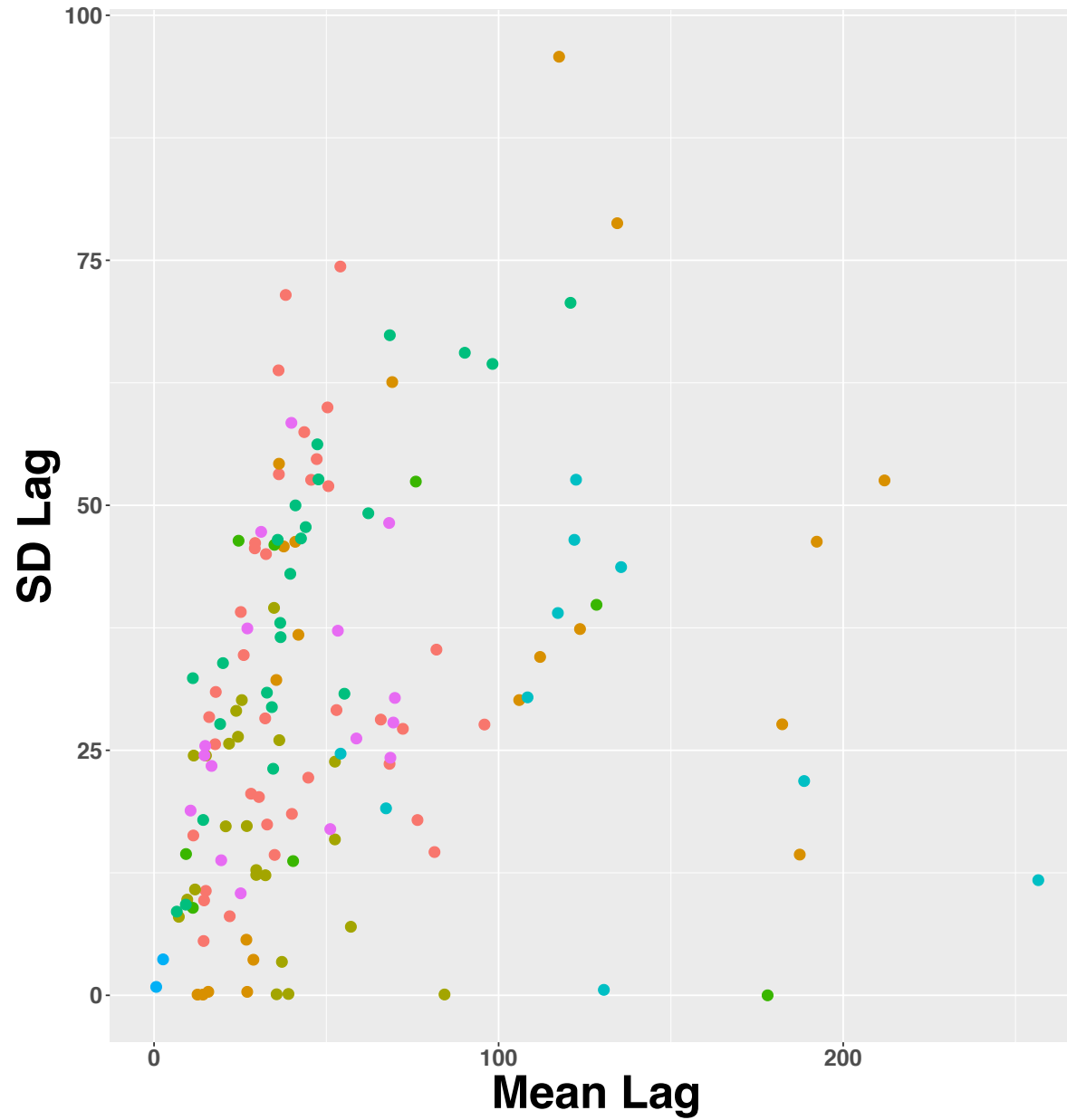
Data Source

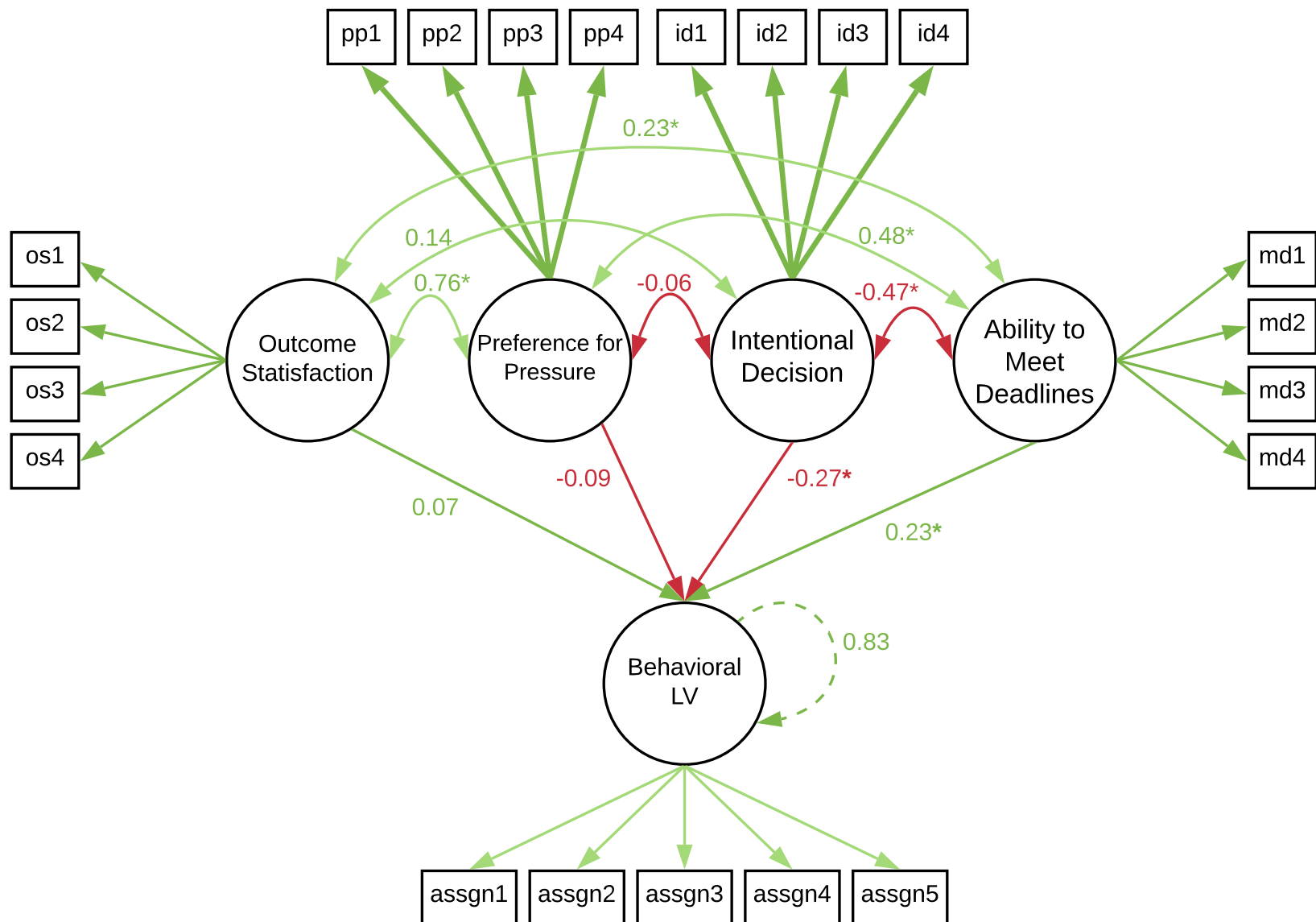
- ❖ Data
 - ❖ $\text{lag} = \text{Assignment Deadline} - \text{Assignment Submission Time}$
 - ❖ across 5 Assignments
 - ❖ Active Procrastination Scale
- ❖ Modeling lag as a behavioral indicator of procrastination

Histogram of Time to Due Date



Mean vs Standard Deviation





Model Fit

Model Fit















Model	cfi	tli	rmsea (5%)	rmsea (95%)	srmr
Active Procrastination	0.964	0.957	0.074	0.082	0.071

Summary

- ❖ Summary
- ❖ Limitations
- ❖ Future Studies
 - ❖ Estimate Latent Procrastination to provide reminders
 - ❖ Other applications

Future

TOOLBOX

-  Single Input
-  Checkbox
-  Radiogroup
-  Comment
-  Rating
-  Boolean
-  HTML
-  Expression (read-only)
-  Matrix (single choice)
-  Matrix (multiple choice)
-  Matrix (dynamic rows)
-  Multiple Text
-  Panel
-  Panel (dynamic panels)

 Undo Redo Learning Module Settings

Save Learning Module

Submit Learning Module

page1


page1 

To calculate the sample mean, first find the sum of all the values:

$$\sum x_i$$

 EditRadiogroup **1 . What is the sum?**

- ☐ 42
- ☐ 37
- ☐ 64

 Other

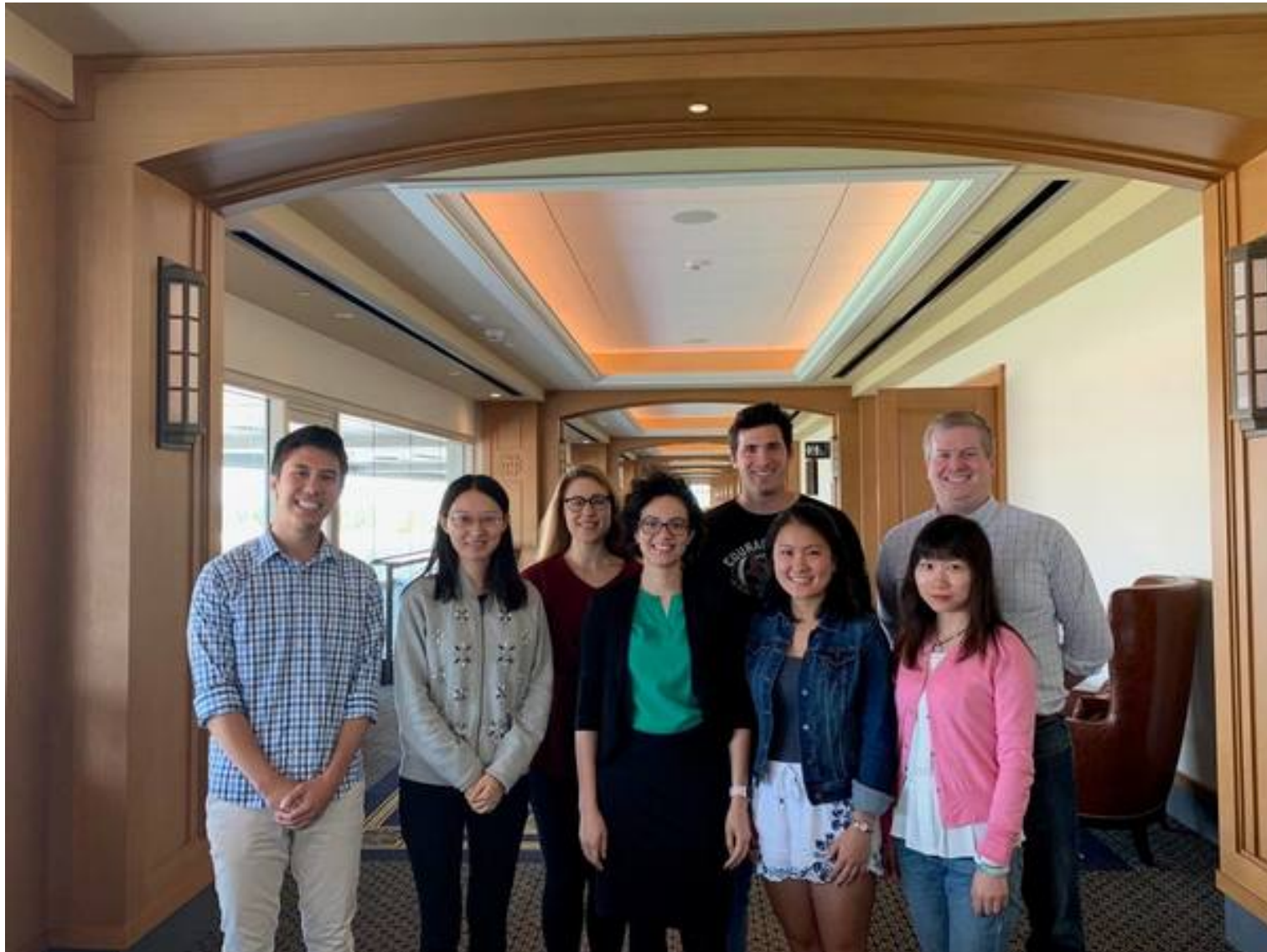
A holistic student view

❖ Multiple Data Sources

- ❖ Assessment Data
- ❖ Big 5
- ❖ Statistics Anxiety
- ❖ Free Response
- ❖ Help Forum

Thank You

Acknowledgments



Thank You

<https://lambslab.nd.edu/>

Thank you

Questions?

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