

Assessing Online Computer Coaches for Problem Solving: Educational Impact

Presented by: Evan Frodermann¹

Qing (Xu) Ryan^{1,3}, Ken Heller¹, Leon Hsu¹, Jia-Ling Lin¹,
Judy Hill¹, Bijaya Aryal², Koblar Alan Jackson⁴

1. University of Minnesota–Twin Cities

2. University of Minnesota–Rochester

3. University of Colorado–Boulder

4. Central Michigan University

AAPT Summer 2014 Meeting
Minneapolis, MN

Supported by NSF DUE #0715615 and DUE-1226197.
and by the University of Minnesota



UNIVERSITY OF MINNESOTA
Driven to DiscoverSM

Research Question

- Do the computer coaches improve students' problem-solving?

Version 1 Summary

- 3 sections of a university calculus-based introductory mechanics course
 - Fall 2011: One section of 221 students
 - Students could complete their homework using WebAssign or the coaches.
 - Spring 2013: Two sections of 148/103 students.
 - Although coaches were available to help with some problems, students were required to complete their homework using WebAssign.



Final exam scores Sp. 2013

Final Exam	Male	Female	Total
Low (0-20%) Coach Use (N=72)	70%±3%	71% ± 6%	70% ± 3%
Medium (40-60%) Coach Use (N=38)	65% ± 4%	68% ± 4%	66% ± 3%
Heavy (80-100%) Coach Use (N=49)	73% ± 2%	68% ± 4%	71% ± 2%

- Heavy users \approx Light users in final exam
 - However, the confidence and FCI score of the students started lower as shown in previous talk.
 - Need to examine this closer.

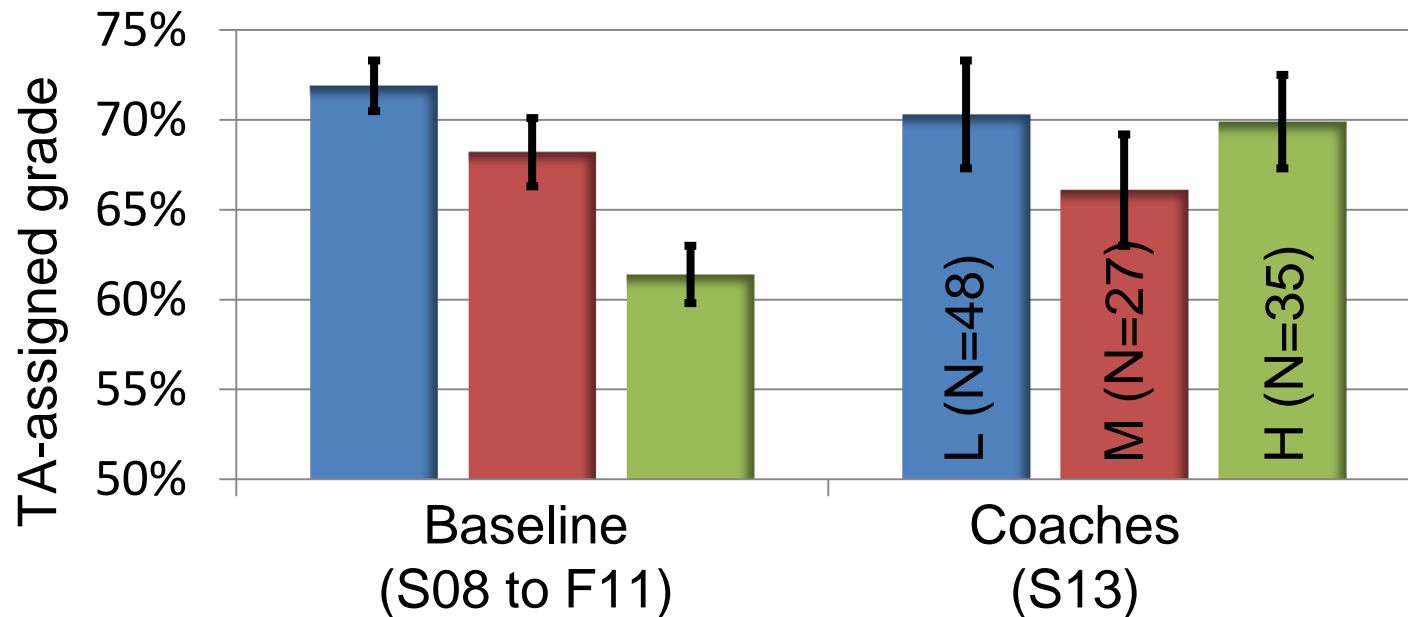


Matching Historical Data

- 4 to 1 match of historical students to Sp2013 students.
 - Matched on pre-class FCI, gender, expected grade, and expected study time.
 - ~85% student perfect match. 3145 students from Spring 2008 to Fall 2011.
- Normalize the exam scores.
 - Exam differences absorbed by equating the exam averages.
 - Retained the exam distributions.
- Avoided unintended selection bias or specific class bias into the results by using varied selection rules.



Final Exam Problem-Solving Grade



- Low-use characterized students in historical match predict there should be a significant difference ($11\% \pm 2\%$) as compared to the high-user characterized students.
- Coached group shows no significant difference ($0\% \pm 4\%$) between heavy and low user groups in coached class from Sp2013.



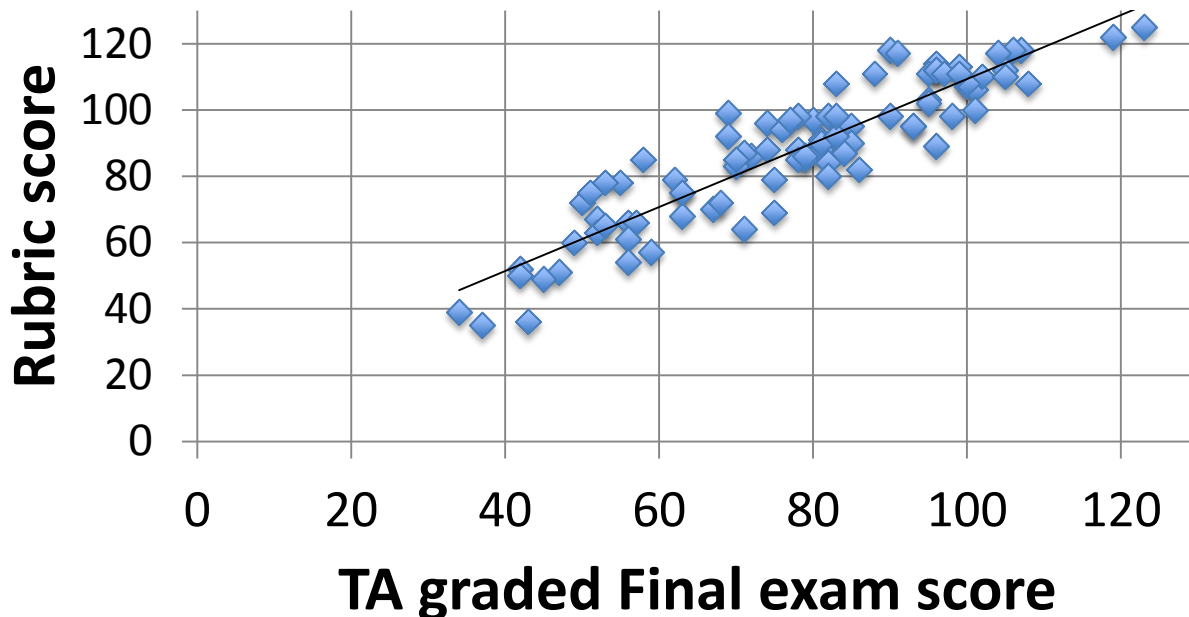
How to assess the impact of the coaches?

- Other tools to assess student problem-solving?
- TA grading vs. rubric assessment
 - Rubric from Jennifer Docktor (2009)
 - J Docktor et al, PERC proceedings (2009) AIP
 - Built upon previous work: P. Heller et al, Vol 60 #7, AJP (1992)
 - 5 categories assessed on scale 0-5
 - Useful Description, Physics Approach, Specific Application, Mathematical Procedures, Logical Progression



TA grading vs. Rubric

RUBRIC FINAL EXAM score: Coach



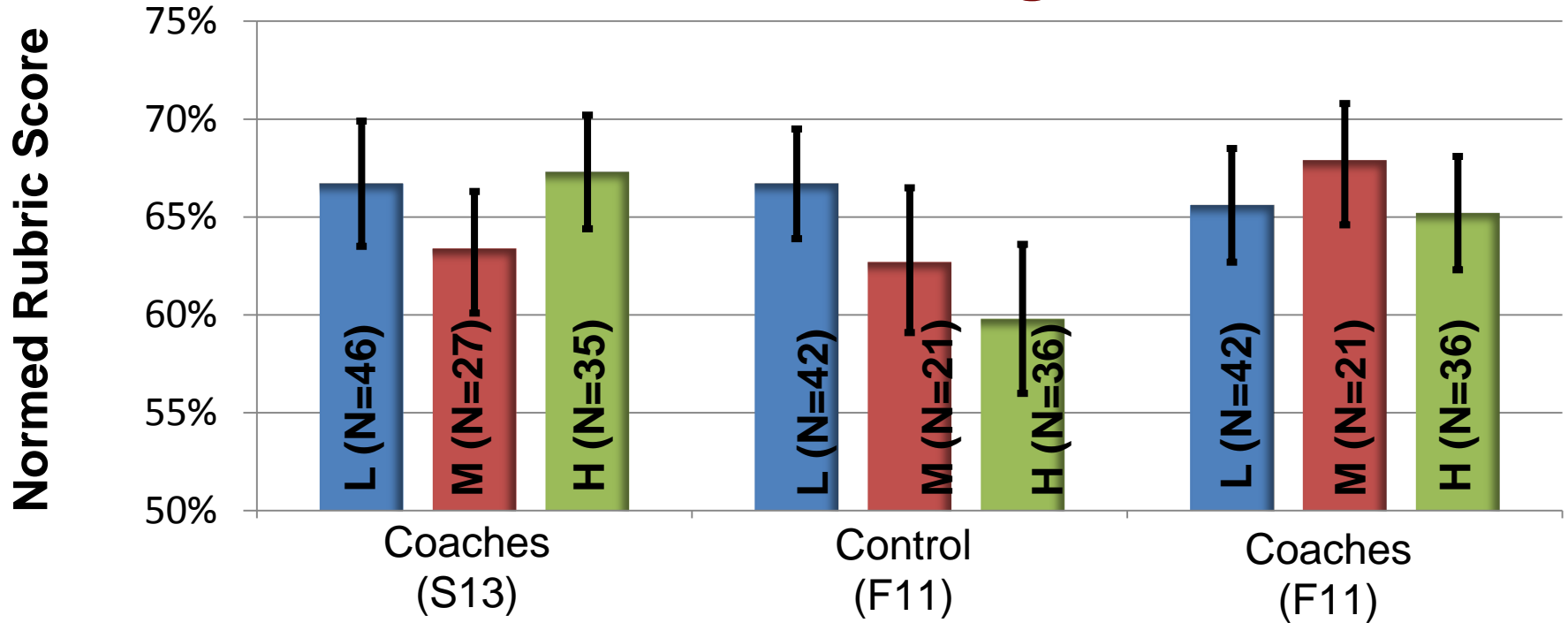
Fall 2011,
Combined 5 Final
exam problems:
C=0.91 coach;
C=0.88 control;

Individual problems:
C=0.79 ~ 0.85

- The TA grades are highly correlated with independently assessed rubric scores.
 - Note: Our TAs have gone through TA training in the Minnesota model but not the rubric assessment.



Final Exam Problem-Solving Rubric Score



- Test differences controlled through a match of low users from Spring 2013 to the Fall 2011 control class low user characterization.
- No perceived difference in low user characterization students.
- The difference in heavy use characterization still present in Fall 2011.



Summary

- Coaches improved problem solving scores
 - Near full letter grade difference
 - Less prepared, less self confident student characteristics
 - More female than the rest of the class (about 1/3 of the class)
 - **70% \pm 3% Coached heavy users; 61% \pm 2% Non-coached matched**
- For TAs with our TA support, grading is a good indication of problem solving skills.
 - High correlation between TA grading and Rubric assessment
- For more information see:
 - Talks: GC09/10, FD08/09 (Wednesday)
 - Posters: PST2C13/14/15 (Tuesday)

