Carnegie Mellon University

HCI - Learning Discontinuity (Second Progress Report)

Jie Luo, Naifei Pan, Yiwen Zhang

Agenda

Introduction/ Clients Information

- Methods & Preliminary results
- Next Steps & Roadblocks

Introduction

Advisor:

Zach Branson, Assistant Teaching Professor, CMU

Members:

Name: Yiwen Zhang

Background: MSP

Name: Naifei(Julia) Pan

Background: MSP

Name: Jie Luo

Background: MSP

Client Info

Vincent Aleven

- Professor and Director of Undergraduate Programs in Human-Computer Interaction Institute, CMU
- Co-founder of Carnegie Learning & MathTutor



Introduction - Background

- This project is tasked with developing a way to detect learning discontinuities within tutor log data to measure effects of out-of-tutor events in Intelligent Tutoring System (ITS)s.
- Research Questions:
 - Do these interventions put students on a different learning trajectory, with respect to the specific skills?
 - How can we measure effect?
 - Do we see struggles before tutor interventions?
- Purposes:
 - Improve Learning with tutor system
 - Improve scientific understanding of learning with ITS and teachers

- 2 Datasets: Students transaction dataset (104,550 transactions). Student-Step dataset (195 students).
- Transaction dataset: transaction time, **tutor response**, problem name, relevant KC, student actions, ...
- Student-step dataset: **opportunity**, problem name, relevant KC, Derived from Transaction dataset.
- **KC**: A Knowledge Component is "an acquired unit of cognitive function or structure that can be inferred from performance on a set of related tasks". We have 7 KCs in our dataset. (Combine variable terms, Compute quotient for constant, etc)
- Each problem can test more than 1 KC.
- **Opportunity**: An opportunity is the first chance on a step for a student to demonstrate whether he or she has learned the associated KC. Opportunity number is therefore a count that increases by one each time the student encounters a step with the listed knowledge component.

- Incorrect attempt: once a student makes a mistake or asks for a hint in one attempt, we would call it an incorrect attempt
- Error rate: the proportion of incorrect attempts among total attempts



- Tutor intervention: only available in Transaction dataset (observations listed by transaction time), manually combined with Student-step dataset
- Tutor intervention time: the earliest exposure to the tutor for a specific student. Here we assume that a tutor would influence all KCs at the same time.





• Predicted error rate: estimated by AFM model











Methods

AFM model: logistic regression for predicting the success of the next step

Additive Factors Model (AFM)



Fitting 2 AFM models (ideally with different slope and intercept) KC by KC Start from KC where sample can be easily separated



Result- preliminary



Next Steps

1. Intervention

a. Happens to the struggling students at their struggling time?

- i. Do students who never receive tutor help have lower error rate?
- ii. Do we see long step duration before intervention?



Next Steps

- 1. Intervention
 - **a**. Happens at the struggle time?
 - b. Accelerate learning rate?
 - i. Fit two AFM models
 - 1. For pre-tutor data: fit a AFM
 - 2. For post-tutor data: fit another AFM (If intervention happens at opportunity M, then opportunity M+1 will be treat as opportunity 1)
 - 3. Compare the two AFM model (jump?)

Next Steps

2. Improve AFM model (adding intervention to the model)

Fit PFA (Performance Factors Analysis) model

AFM:

$$m(i, j \in KCs, n) = \alpha_i + \sum_{j \in KCs} (\beta_j + \gamma_j n_{i,j})$$
PFA:

$$m(i, j \in KCs, s, f) = \sum_{j \in KCs} (\beta_j + \gamma_j s_{i,j} + \rho_j f_{i,j})$$



Contacts

Project Members:

Jie Luo <<u>jieluo@andrew.cmu.edu</u>>

Naifei Pan <<u>npan@andrew.cmu.edu</u>>

Yiwen Zhang <<u>viwenzh3@andrew.cmu.edu</u>>

Faculty Advisor:

Zach Branson <<u>zach@stat.cmu.edu</u>>