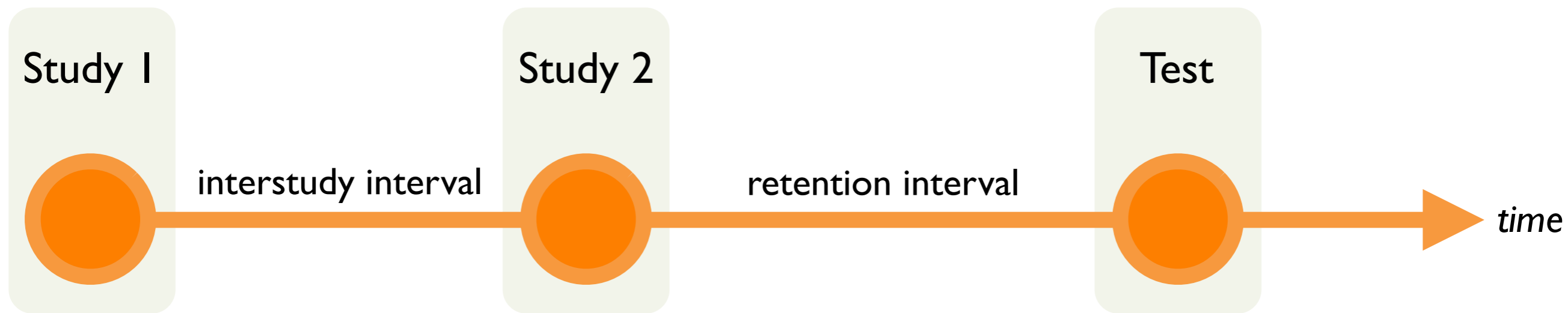


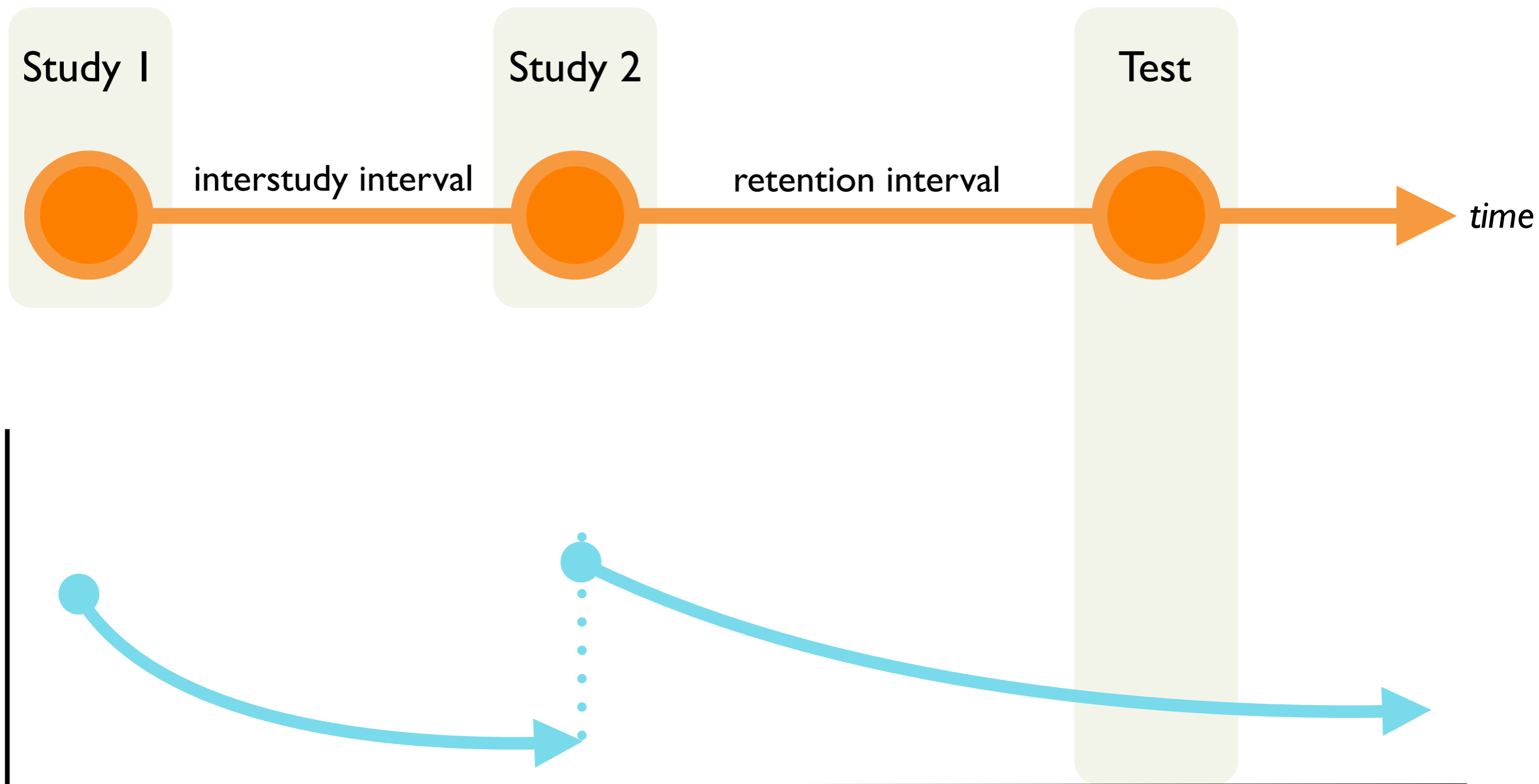
Inferring History-Dependent Memory Strength via Collaborative Filtering

Robert Lindsey, Jeff Shroyer, Michael Mozer
University of Colorado
December 8, 2012

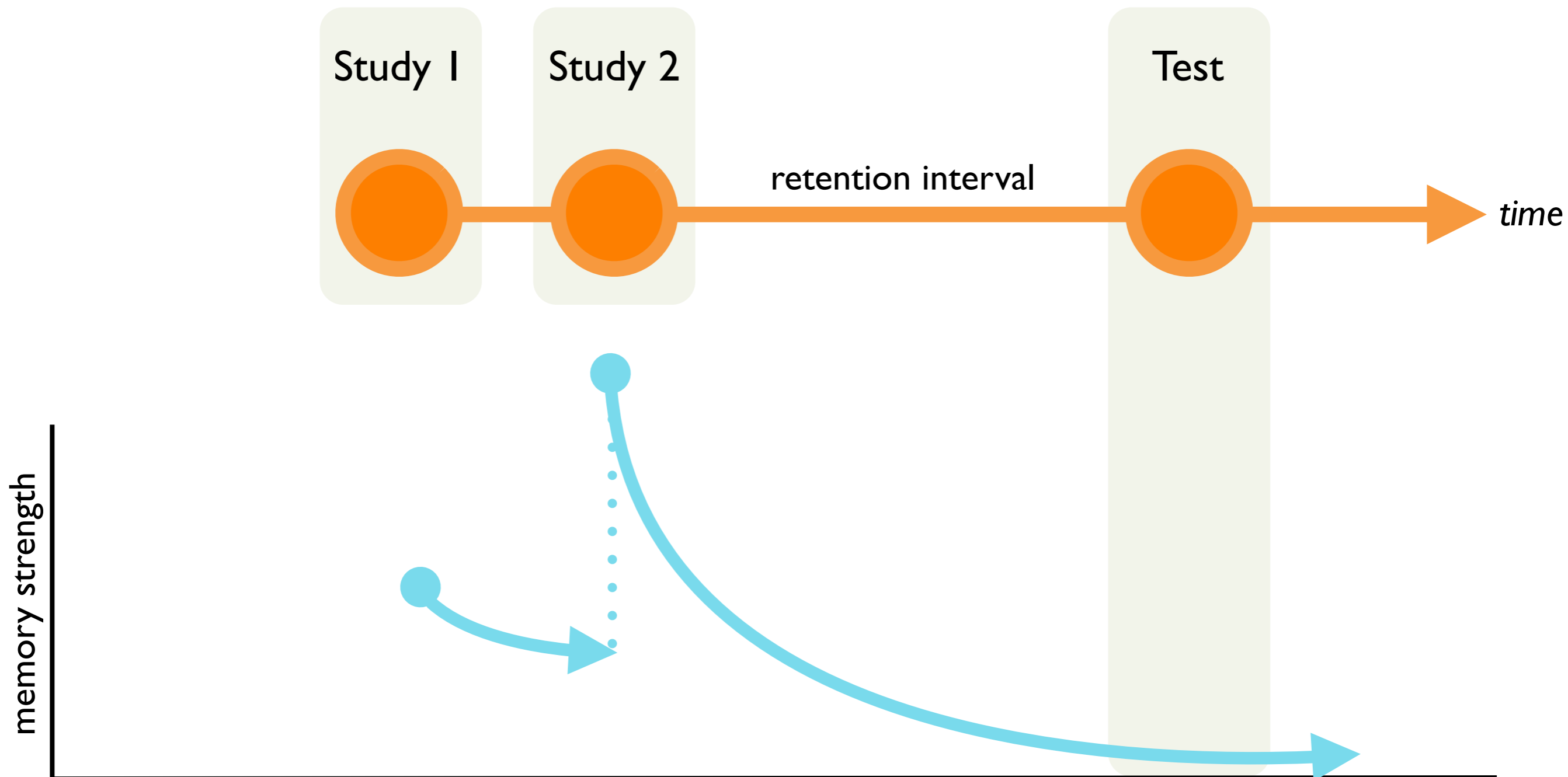
Retrieval probability is strongly influenced by the temporal spacing of study opportunities



Retrieval probability is strongly influenced by the temporal spacing of study opportunities



Retrieval probability is strongly influenced by the temporal spacing of study opportunities





memory strength

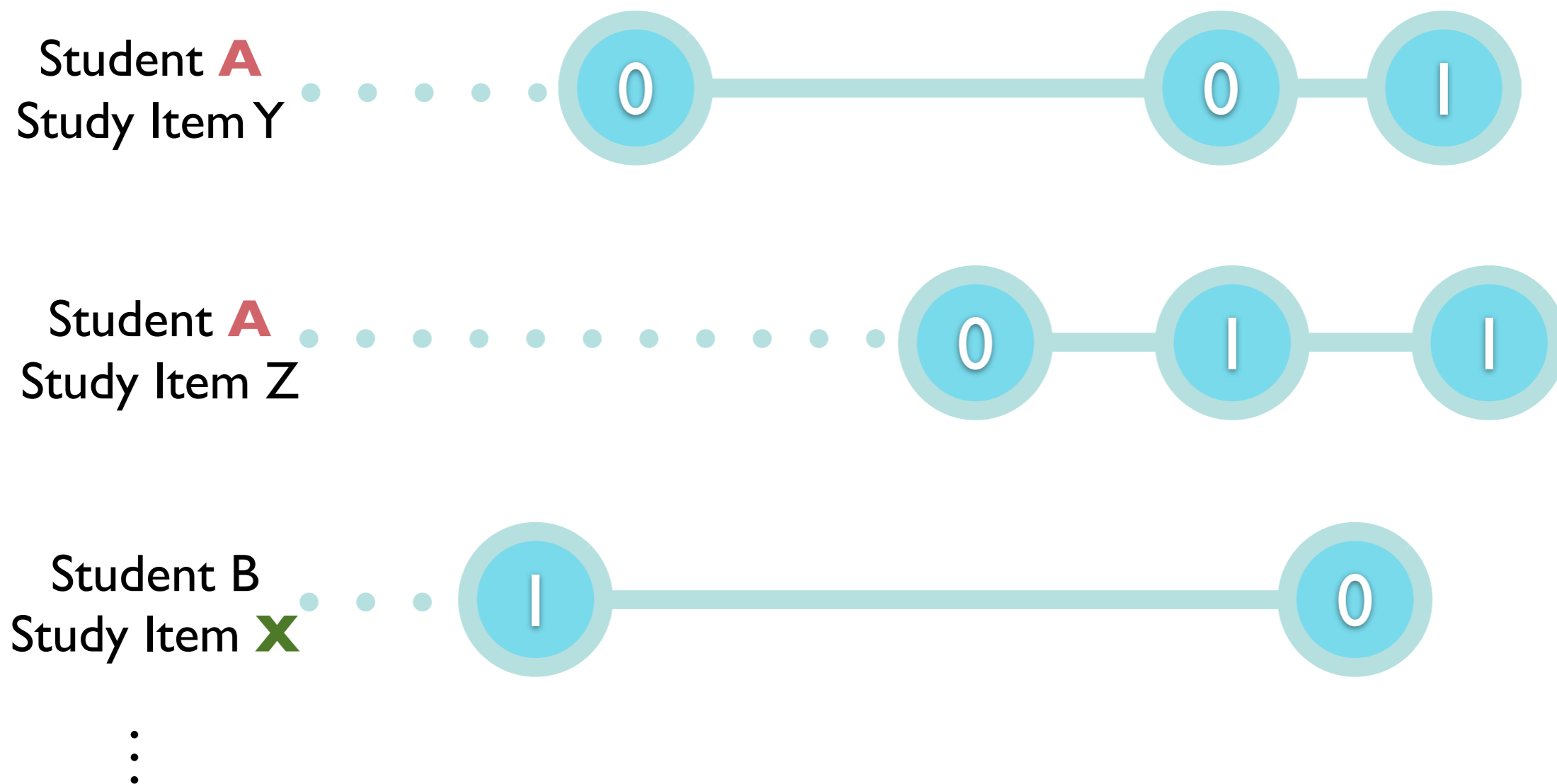
Item response theory is focused on measuring a **fixed** memory state at an instant in time

Effective teaching requires predicting the **dynamic** properties of memory

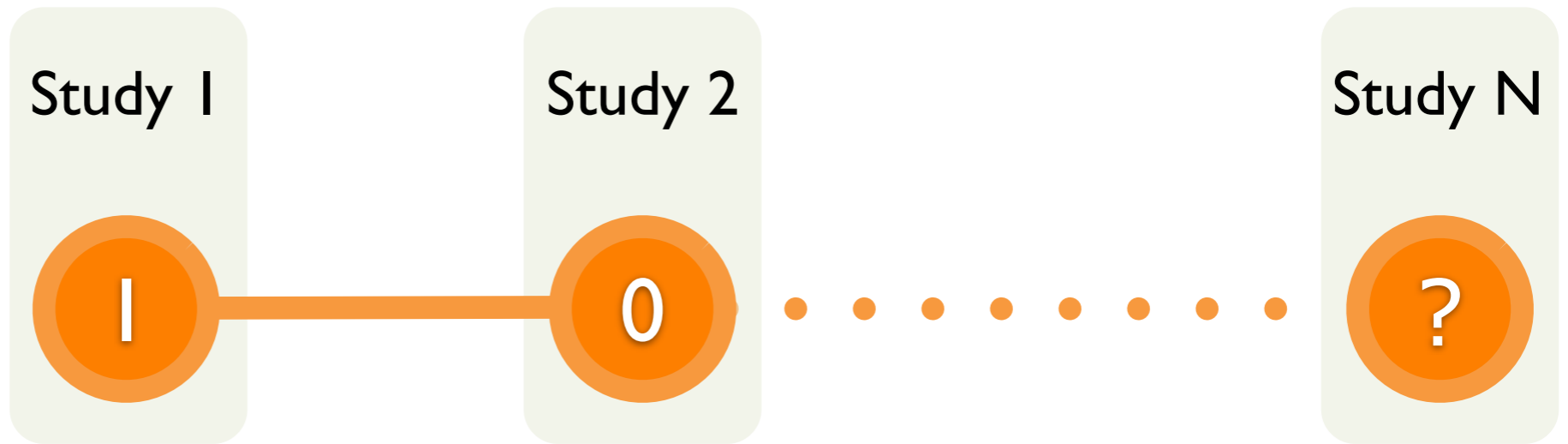
Inference Problem



Training Data



Student **A**
Study Item **X**

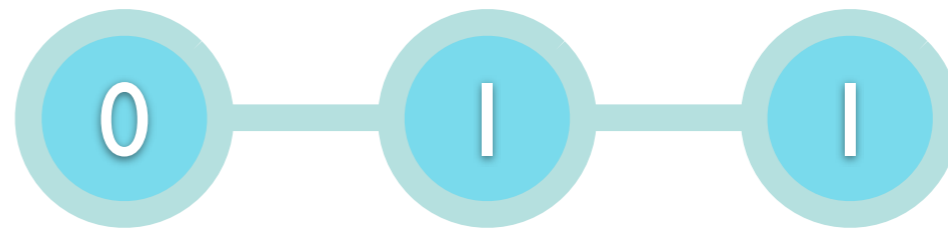


Student **A**
Study Item **Y**



Same student,
other items

Student **A**
Study Item **Z**



Student **B**
Study Item **X**



Same item,
other students


⋮

Naive Approach: Ignore Study History

logistic
function

Likelihood

$$\pi_n = \sigma(\alpha_s - \delta_i)$$


 ability difficulty

individuals
linked
together via
shared prior

Hierarchical Priors

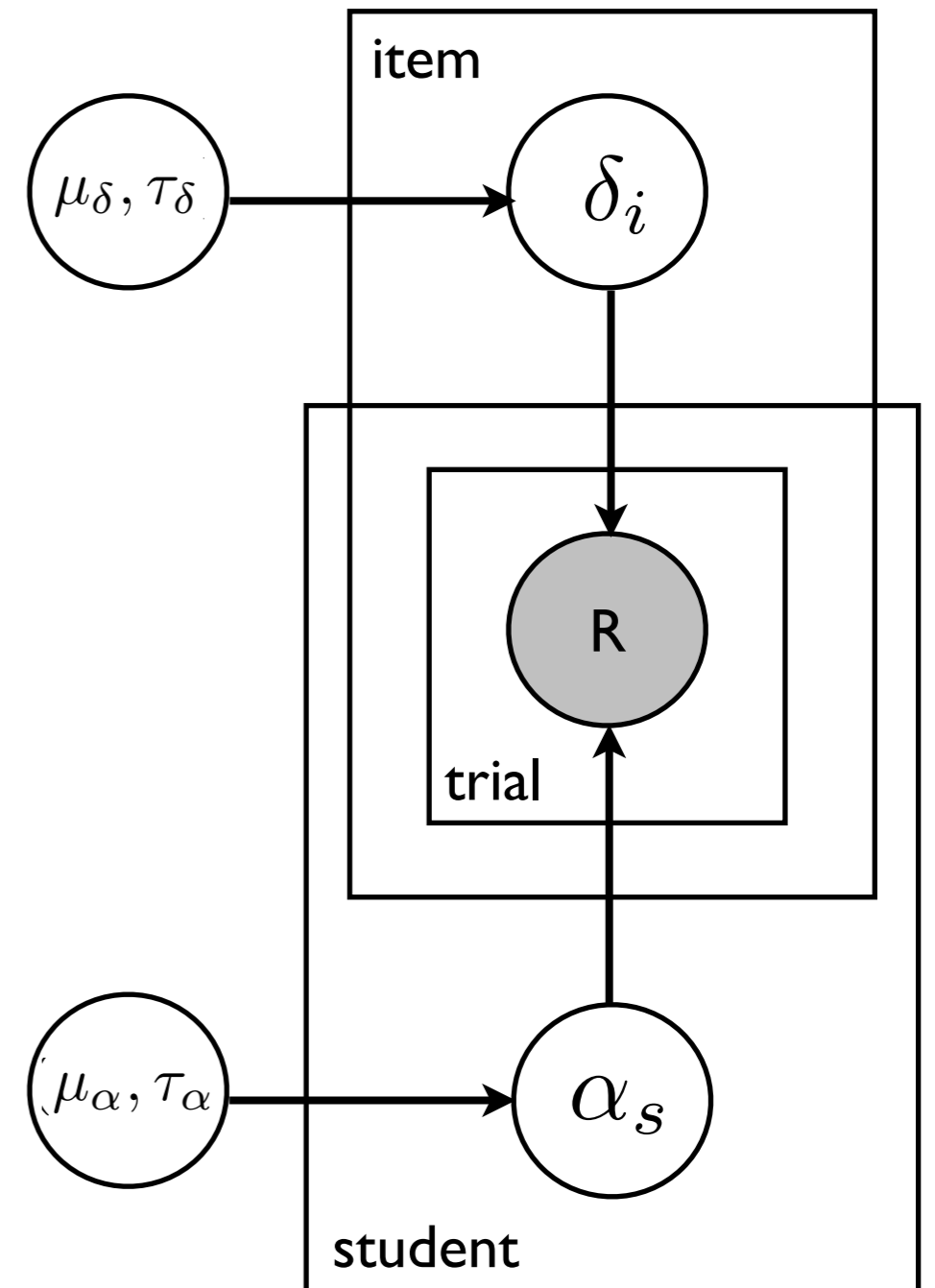
$$\alpha_s \sim \text{Normal}(\mu_\alpha, \tau_\alpha)$$

$$(\mu_\alpha, \tau_\alpha) \sim \text{NG}(\mu_0^{(\alpha)}, \kappa_0^{(\alpha)}, a_0^{(\alpha)}, b_0^{(\alpha)})$$

$$\delta_i \sim \text{Normal}(\mu_\delta, \tau_\delta)$$

$$(\mu_\delta, \tau_\delta) \sim \text{NG}(\mu_0^{(\delta)}, \kappa_0^{(\delta)}, a_0^{(\delta)}, b_0^{(\delta)})$$

Rasch Model



Additive Factor Model: Binning Approach

logistic
function

Likelihood

$$\pi_n = \sigma(\alpha_s - \delta_i + h_{si})$$

↑
summarization of
study history

$$h_{si} = \sum_k w_k f_{sik}$$

↑
observed covariate

individuals
linked
together via
shared prior

Hierarchical Priors

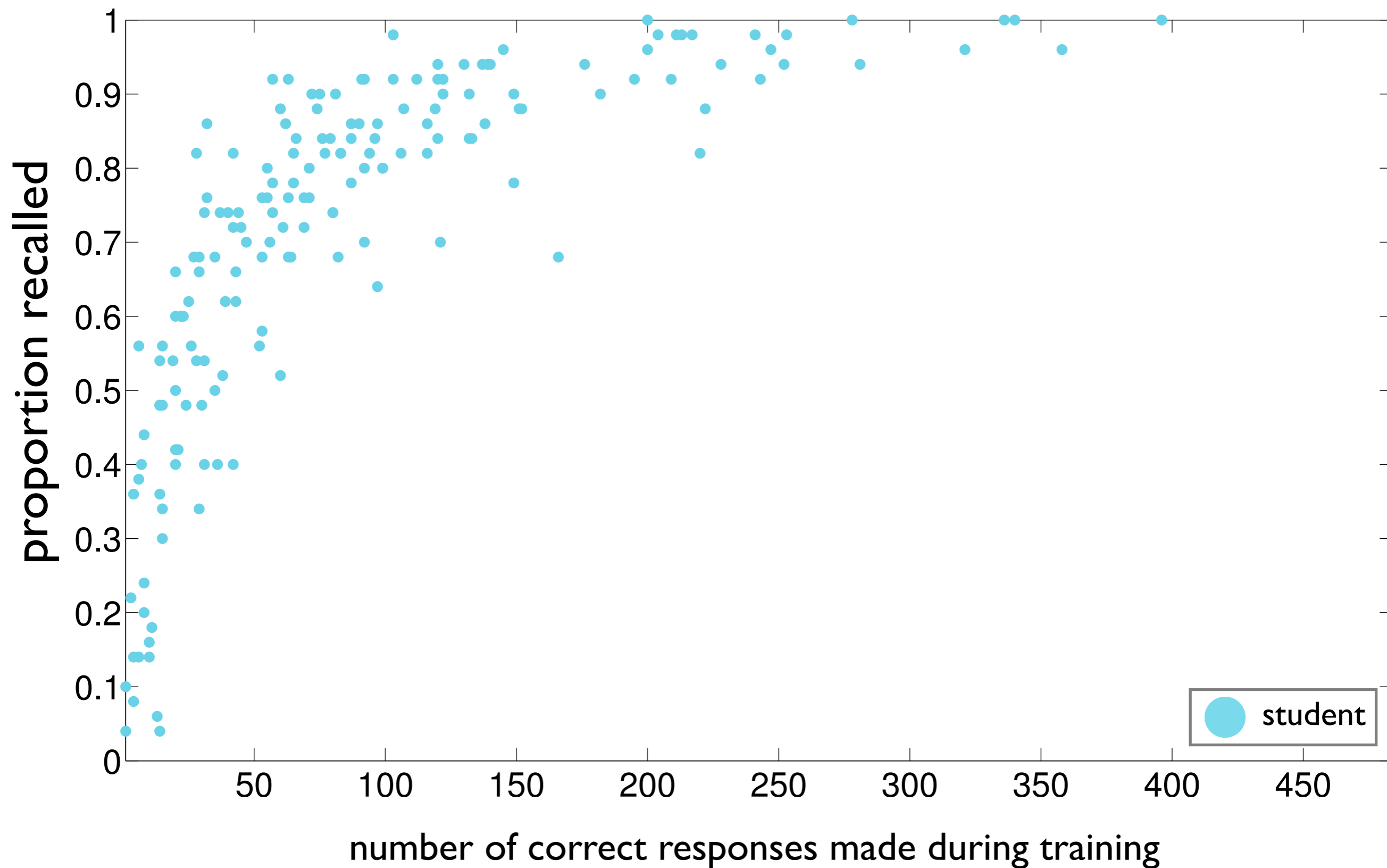
$$\begin{aligned} \alpha_s &\sim \text{Normal}(\mu_\alpha, \tau_\alpha) \\ (\mu_\alpha, \tau_\alpha) &\sim \text{NG}(\mu_0^{(\alpha)}, \kappa_0^{(\alpha)}, a_0^{(\alpha)}, b_0^{(\alpha)}) \\ \delta_i &\sim \text{Normal}(\mu_\delta, \tau_\delta) \\ (\mu_\delta, \tau_\delta) &\sim \text{NG}(\mu_0^{(\delta)}, \kappa_0^{(\delta)}, a_0^{(\delta)}, b_0^{(\delta)}) \end{aligned}$$

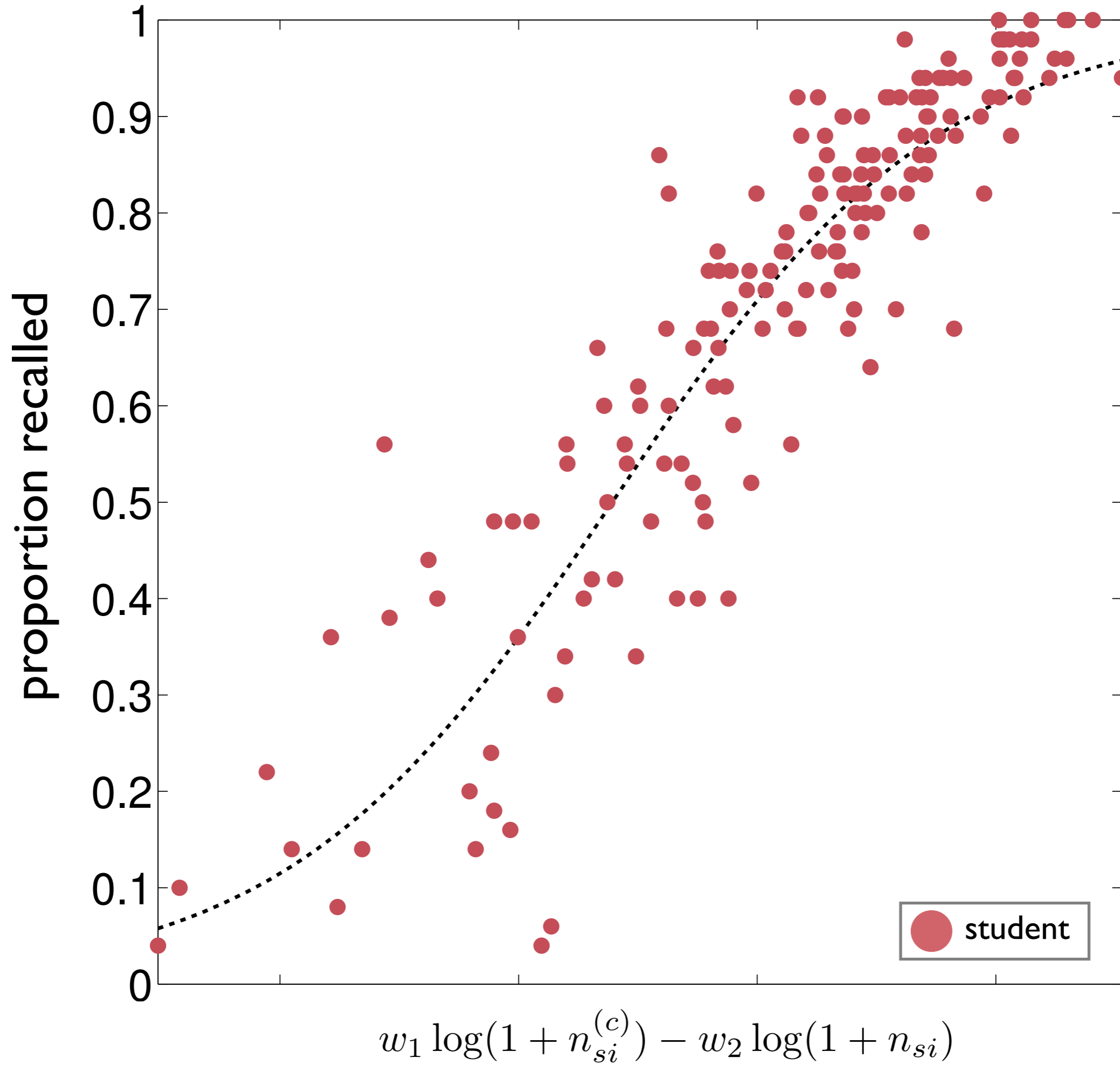
$$f_{sik} = ?$$

$$f_{sik} = ?$$

Simple Idea:

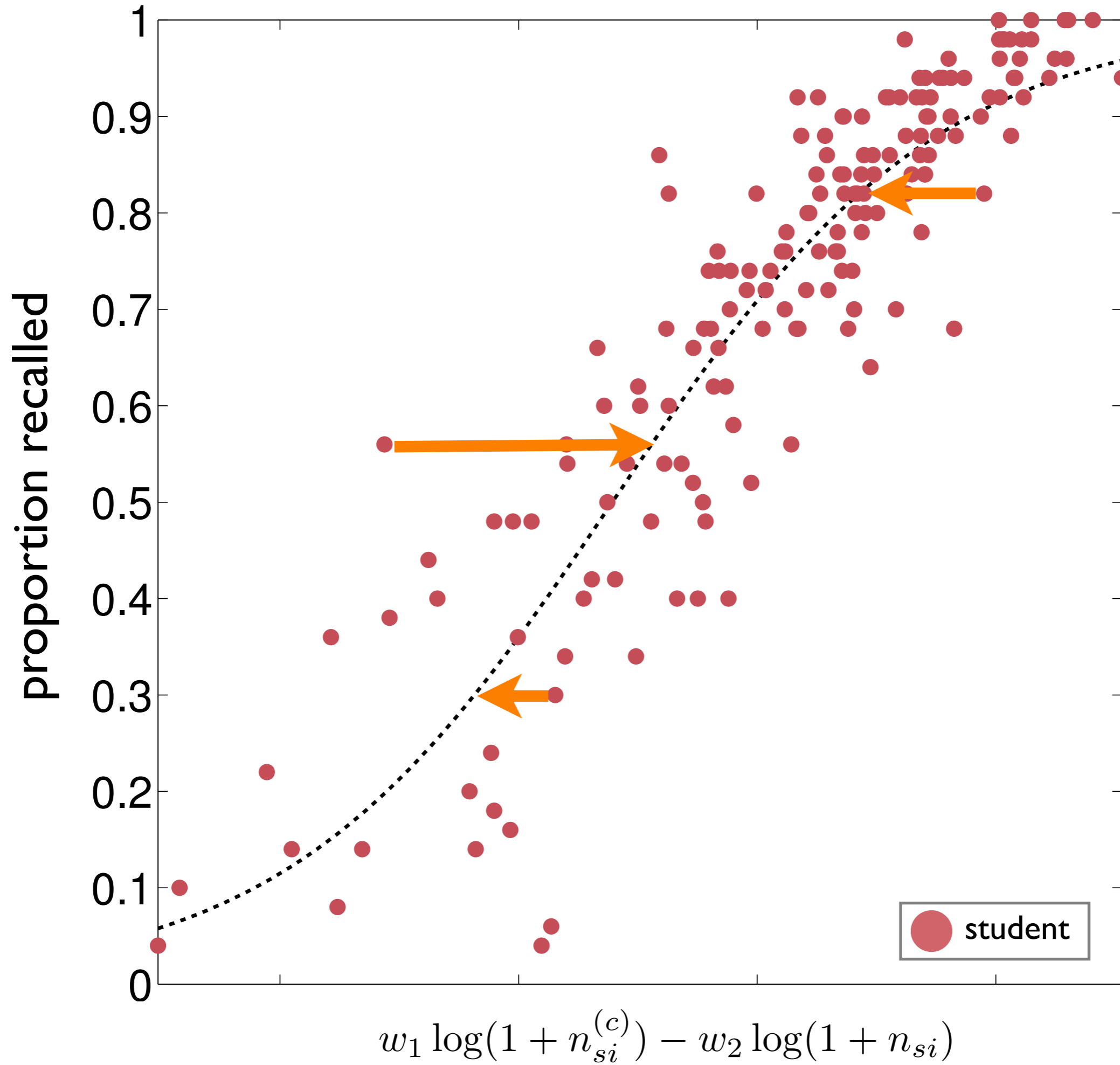
count the number of correct responses made and the total number of trials





$$w_1 \log(1 + n_{si}^{(c)}) - w_2 \log(1 + n_{si})$$

● student

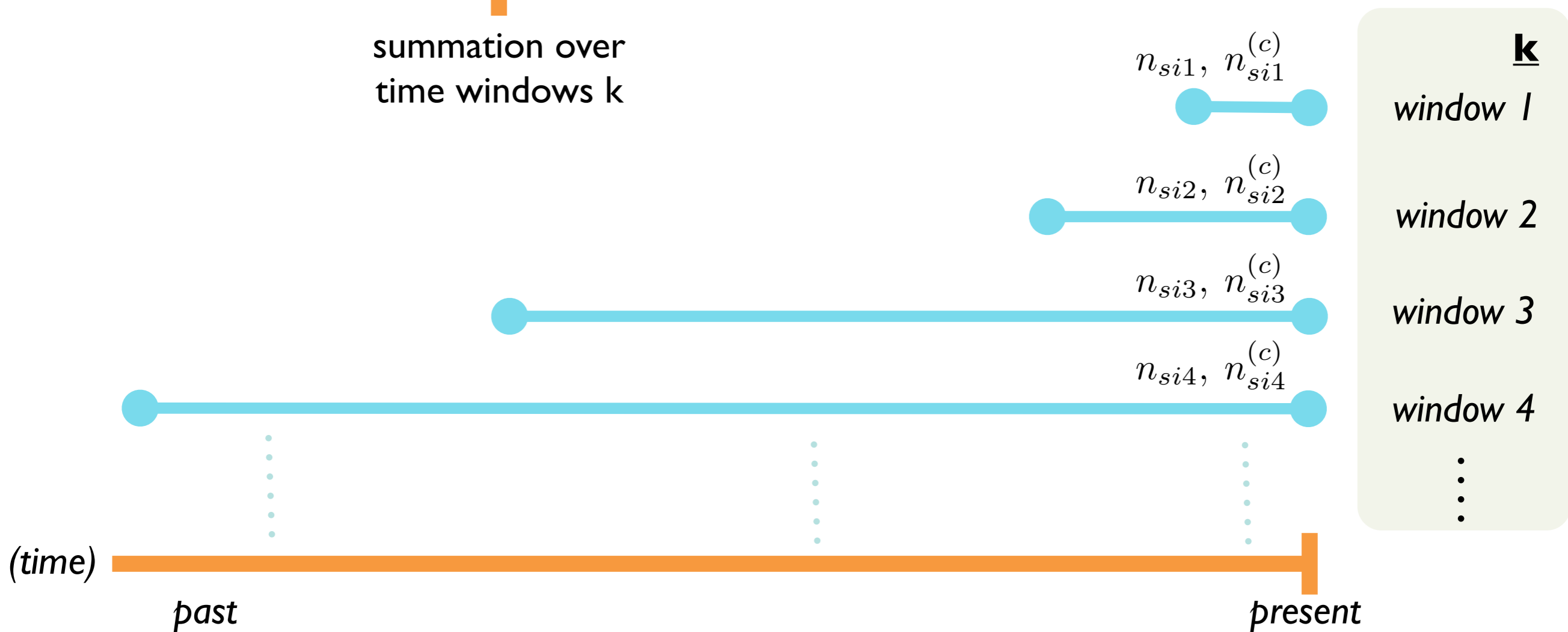


$$\pi_n = \sigma(\alpha_s - \delta_i + h_{si})$$

counts in a particular time window

$$h_{si} = \sum_k \left[w_{k1}^{(c)} \log(1 + n_{sik}^{(c)}) - w_{k2} \log(1 + n_{sik}) \right]$$

summation over
time windows k



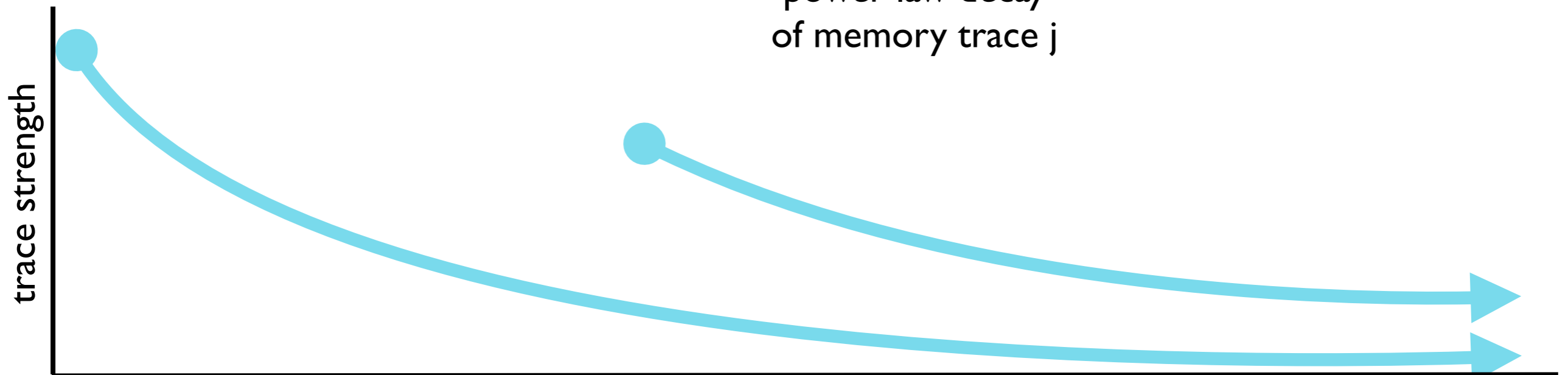
AFM Factor Model: Trace Decay Approach

$$\pi_n = \sigma(\alpha_s - \delta_i + h_{si})$$

summation over
previous
presentations j

$$h_{si} = c \log\left(1 + \sum_j m_j t_{sij}^{-d}\right)$$

power law decay
of memory trace j



The screenshot shows a web interface for a vocabulary tutor. At the top, the word "to receive" is displayed in a large, bold font. Below it is a text input field. Underneath the input field are three buttons: "Accent", "I don't know", and "Submit". Below the buttons is a table with three columns: "Time Spent", "Cards Completed", and "Correct Answers". The values in the table are "0:04", "0", and "0" respectively. At the bottom of the table, there are links for "Back" and "Logout".

Time Spent	Cards Completed	Correct Answers
0:04	0	0

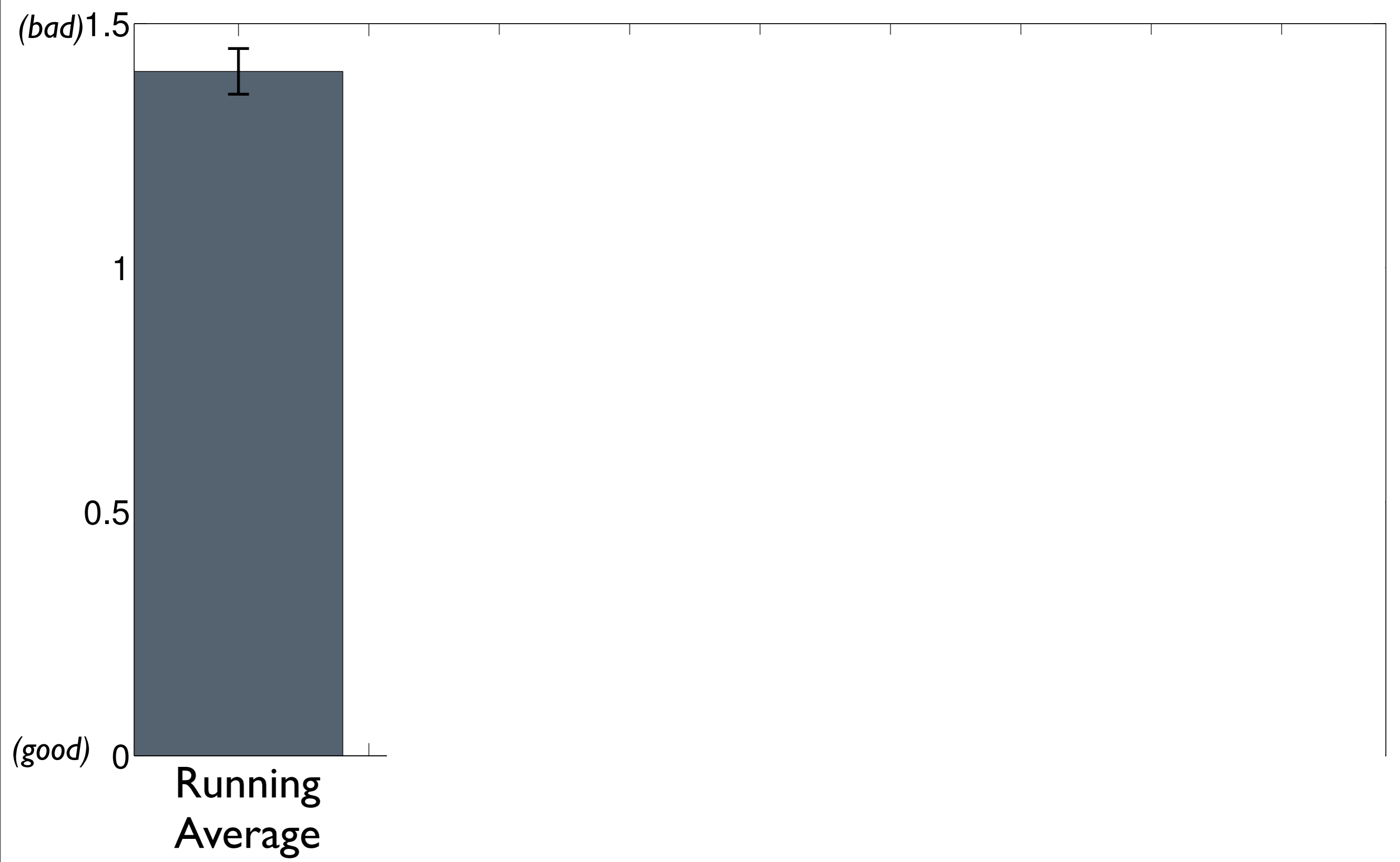
Dataset

- Online vocabulary tutor
- 8th grade Spanish classes
 - 180 students
 - 409 study items
- 14 week study
- 600,000+ study trials

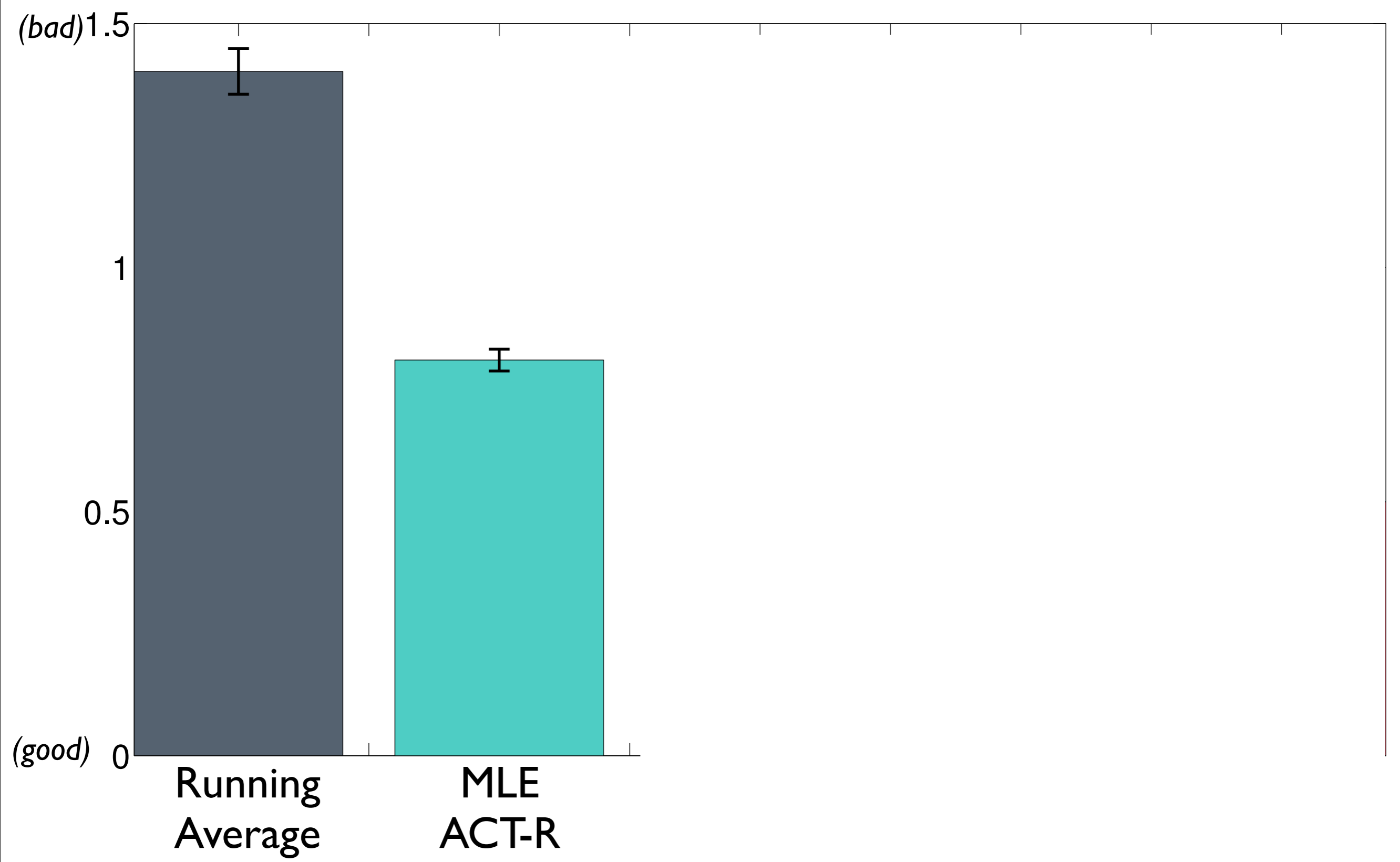
Simulations

Predict recall on day X given data from days 1 through $X-1$

Average Surprisal

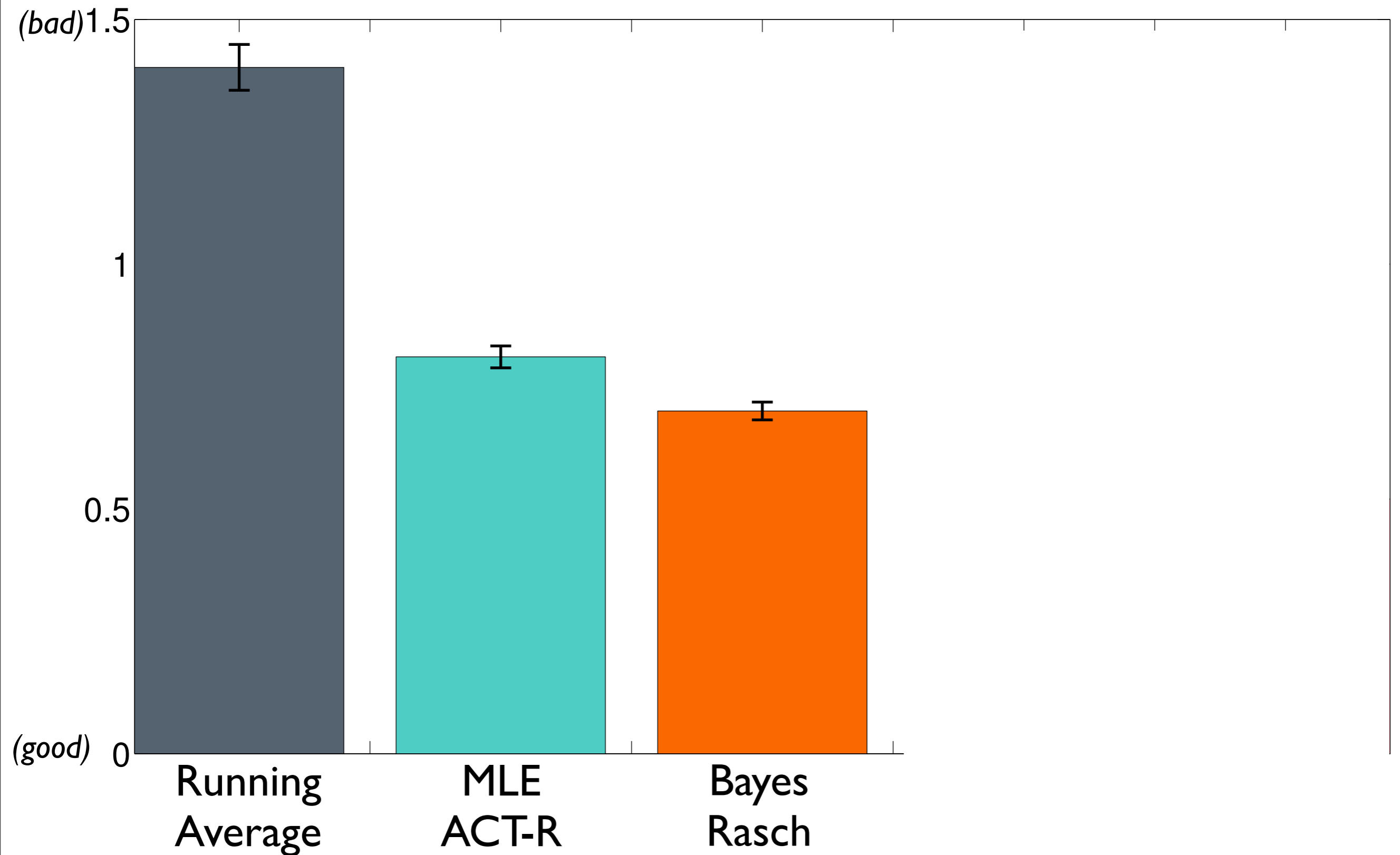


Average Surprisal



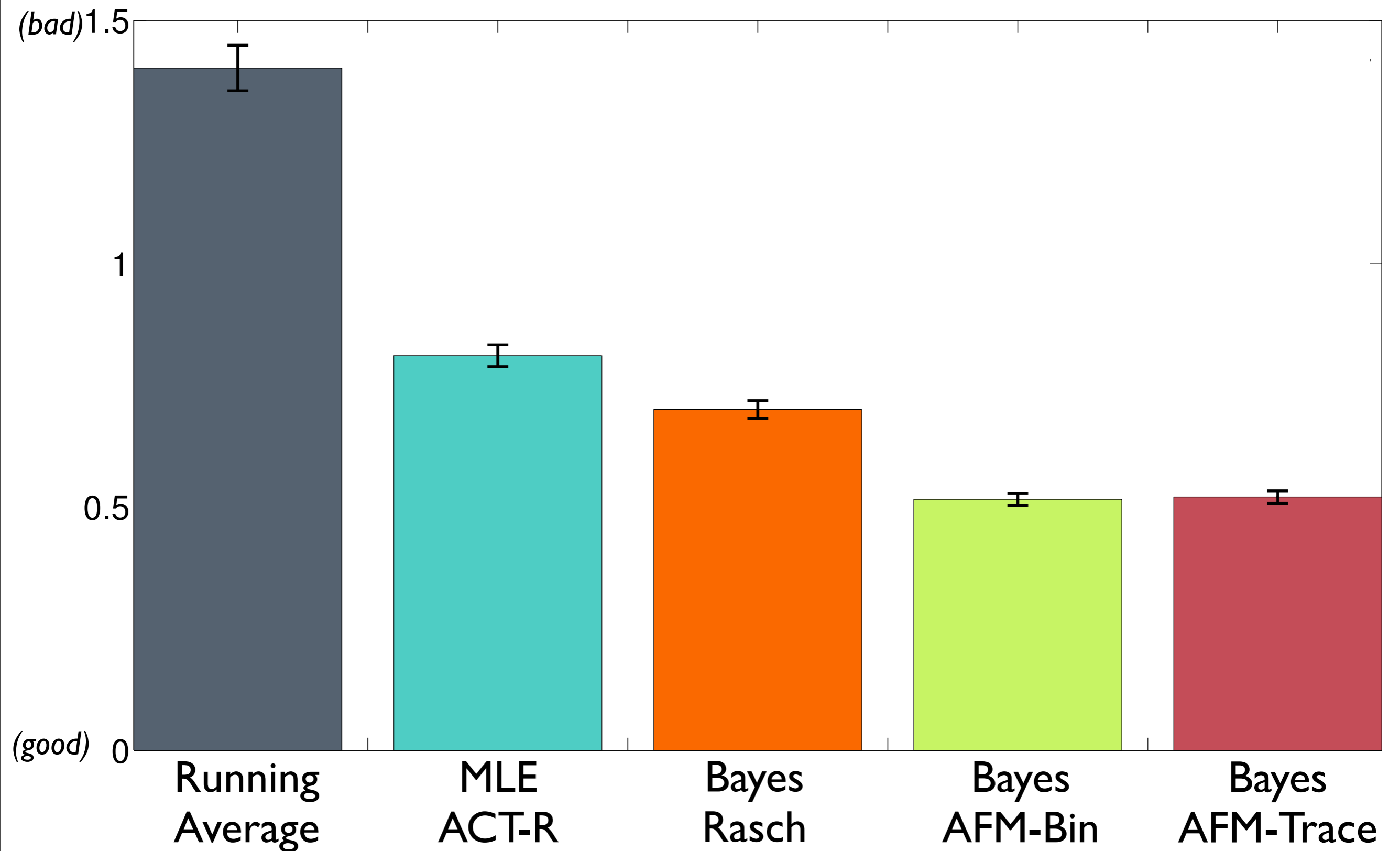
(Pavlik & Anderson, 2008)

Average Surprisal



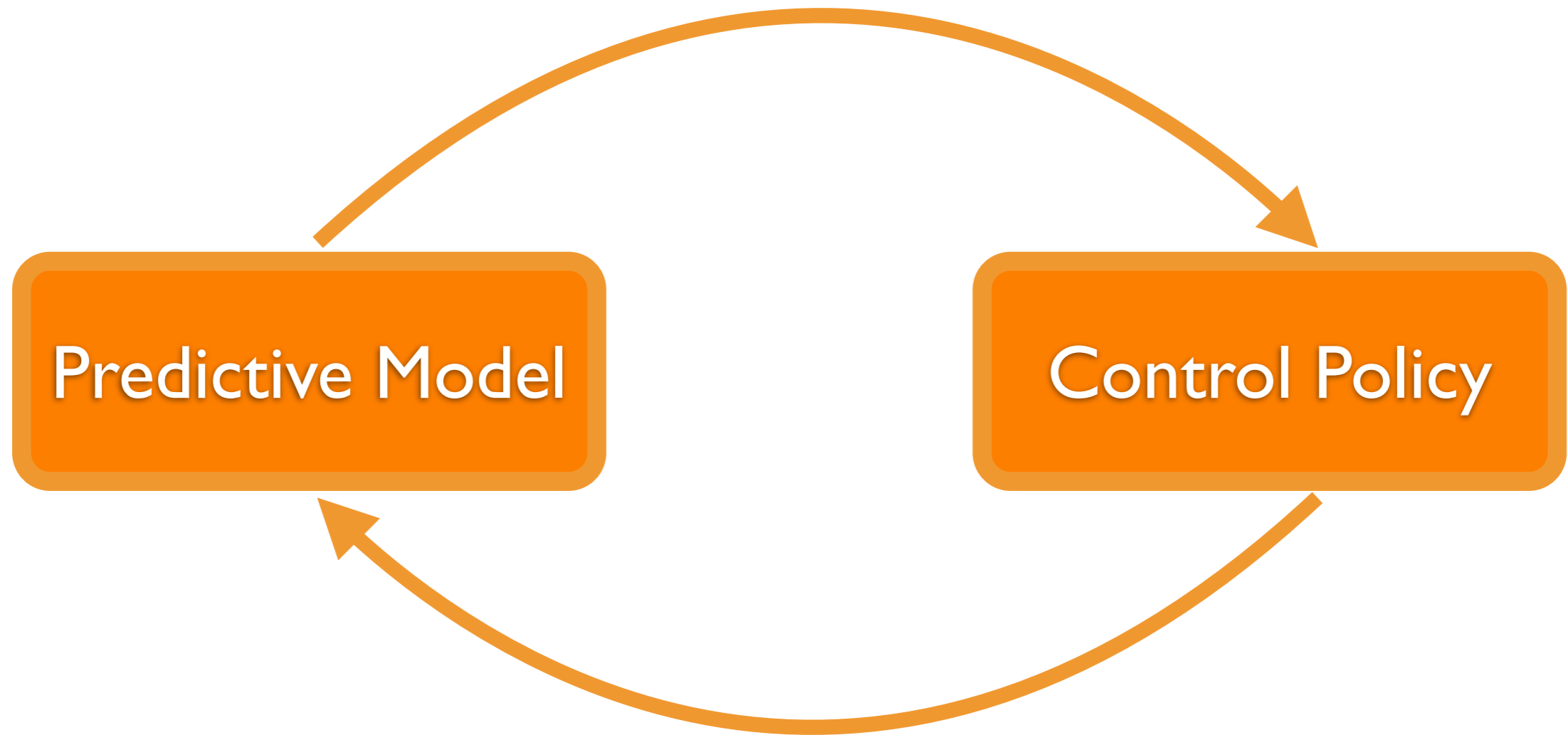
(Pavlik & Anderson, 2008)

Average Surprisal



(Pavlik & Anderson, 2008)

individualized predictions



intelligently chosen material reviewed by student

To what extent can this scheme improve long term retention?

individualized predictions

Predictive Model

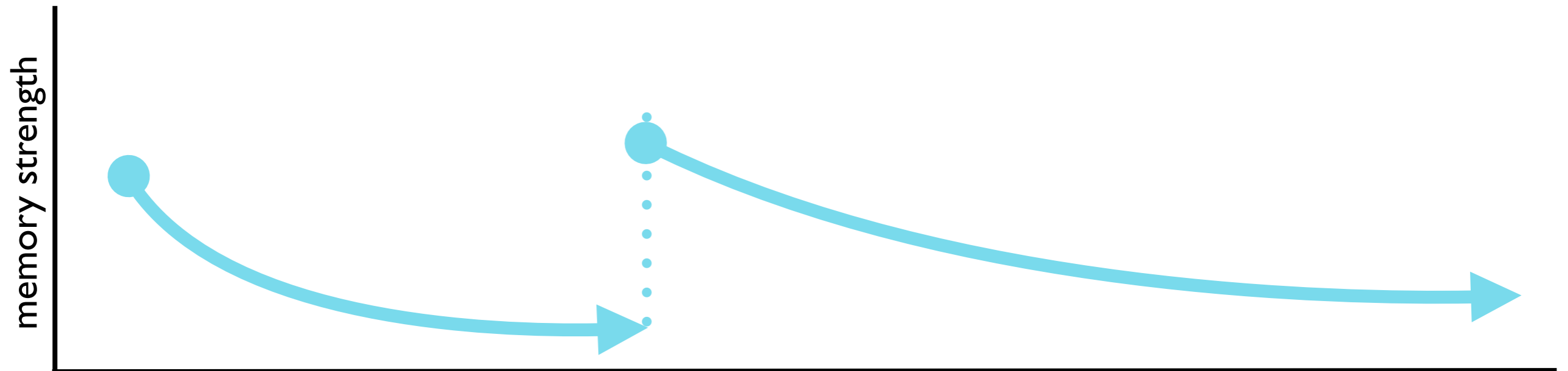
Control Policy

intelligently chosen material reviewed by student

Experiment Timeline



Regardless of what's being learned, forgetting occurs.
Accounting for it is important.



Collaborative filtering enables us to make strong inferences despite weak behavioral data

We can use the models to intelligently choose individual items to present to individual students