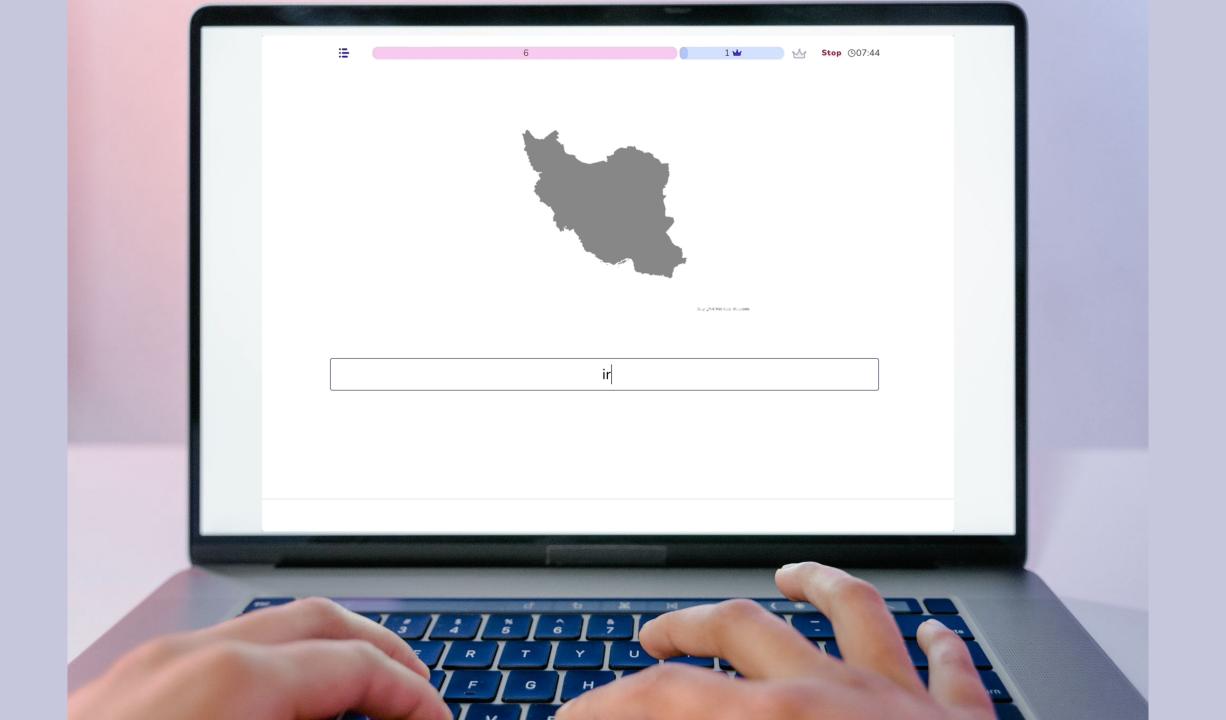
# Predicting Longer-Term Retention from Learners' Retrieval Practice Performance and its application in education

Maarten van der Velde & Hedderik van Rijn







#### Multi-session retrieval practice

10 min

1 min

1.00 -

0.75 -

0.50 -

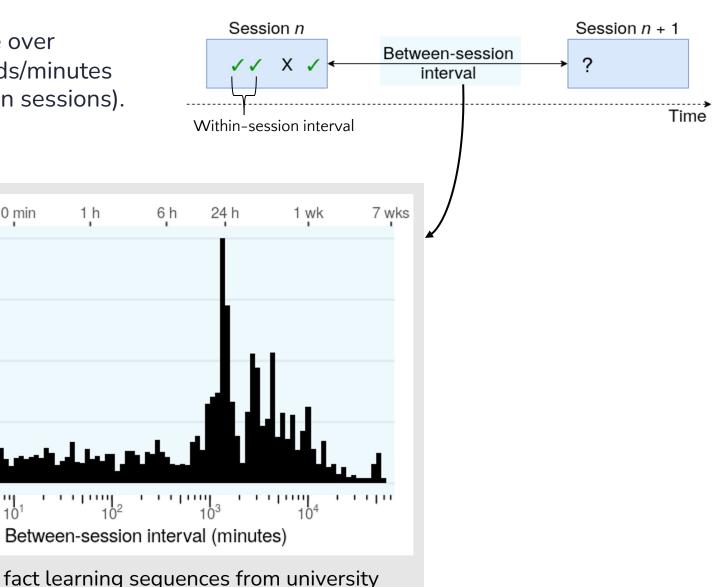
0.25 -

0.00 -

Density

1 h

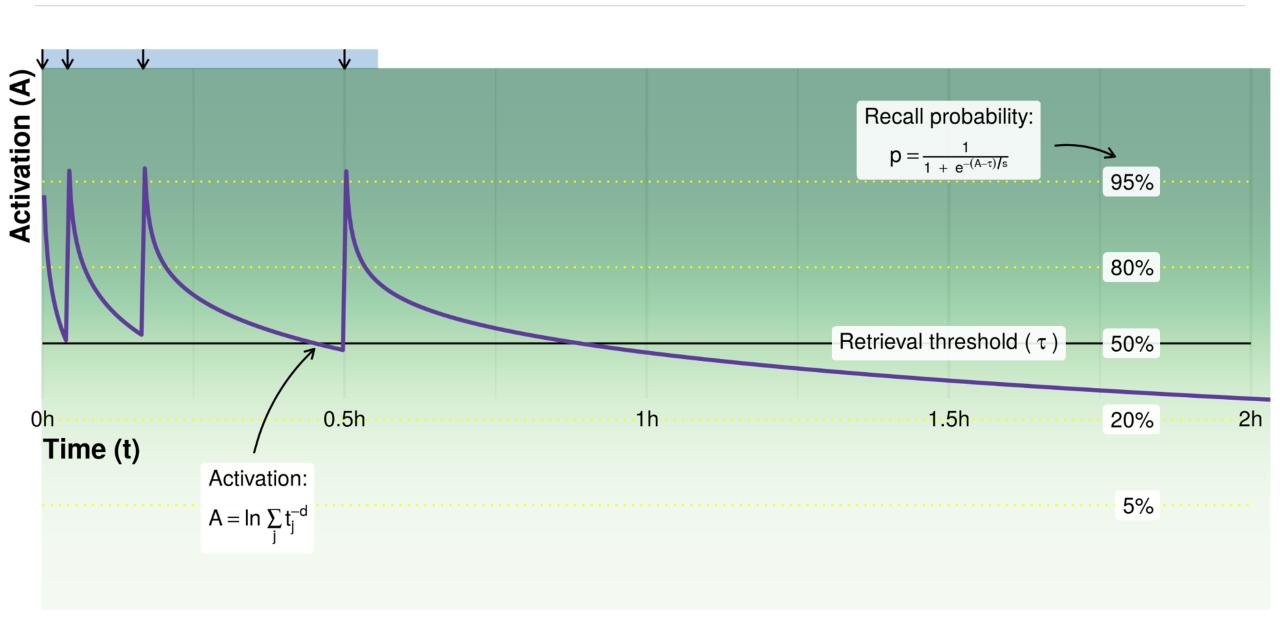
Learners typically do spaced retrieval practice over multiple sessions. Intervals range from seconds/minutes (within session) to hours/days/weeks (between sessions).

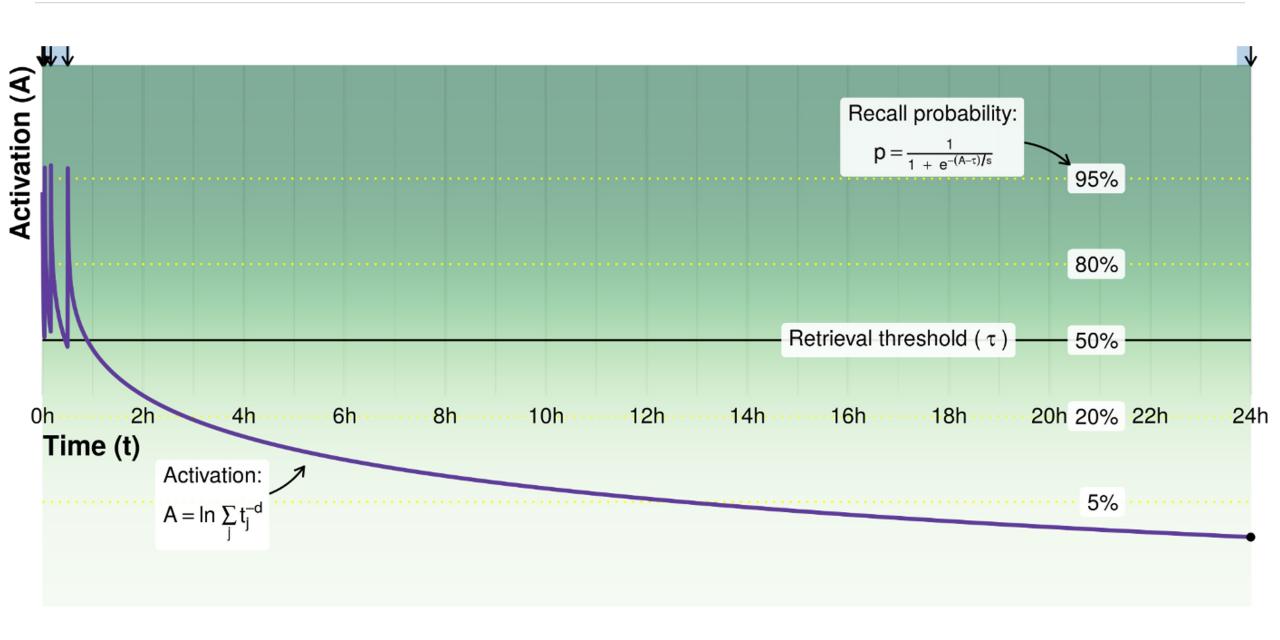


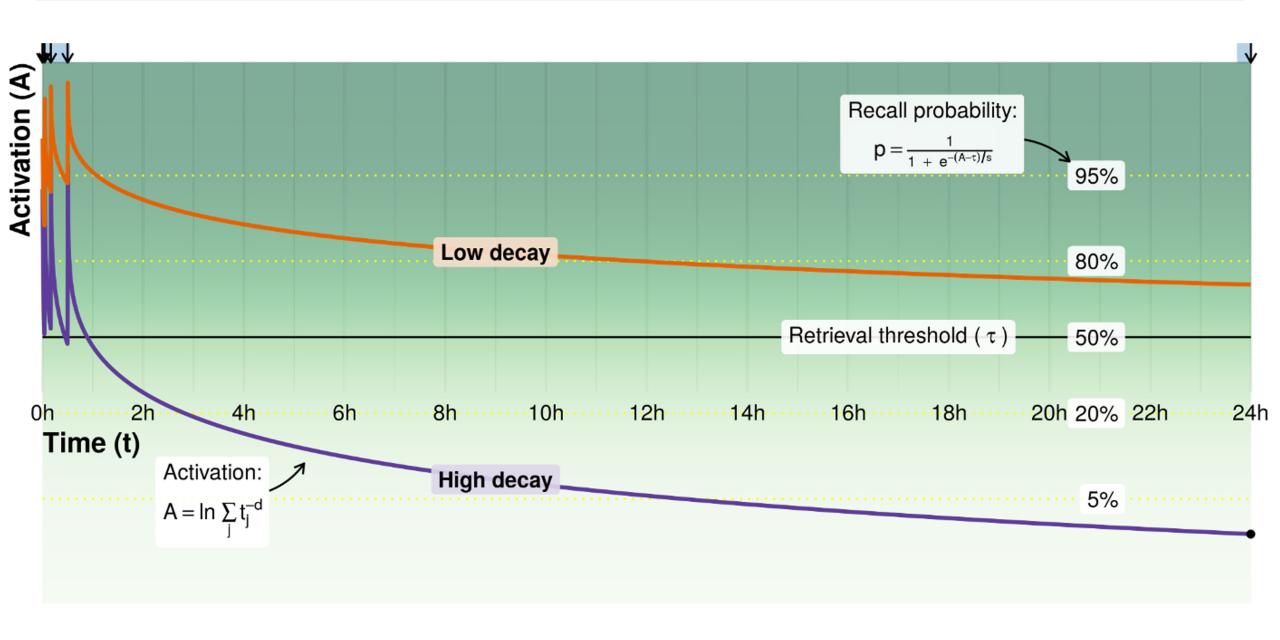
Data set: 25,843 fact learning sequences from university students in a Cognitive Psychology course.

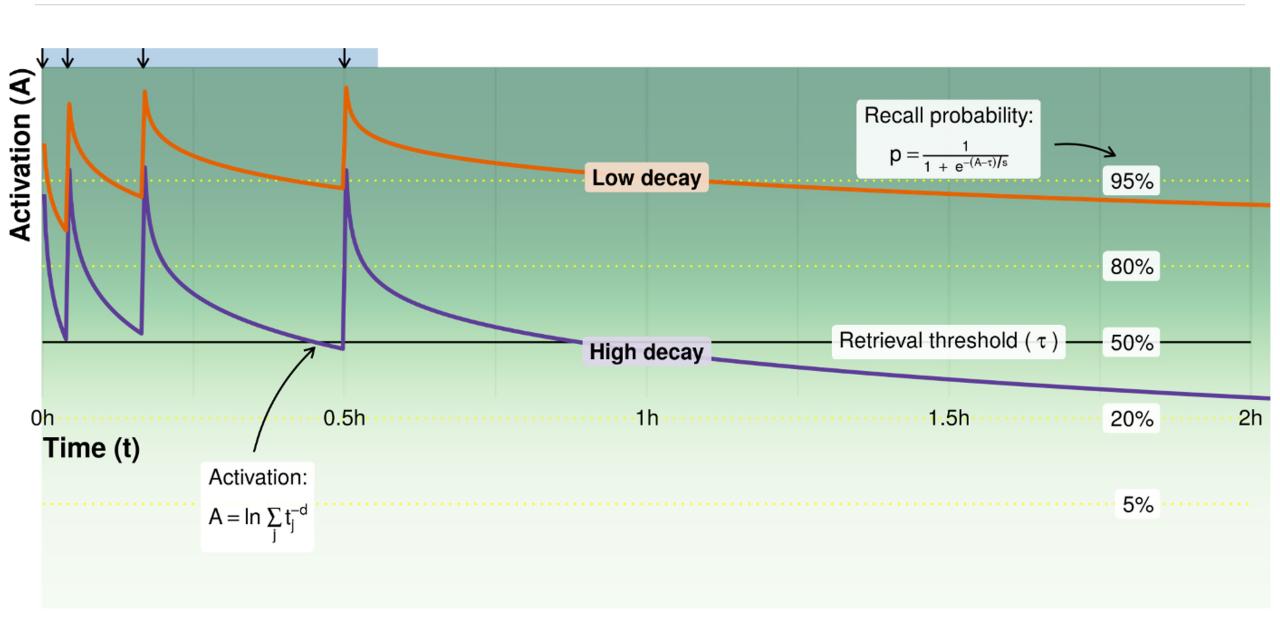
#### Challenges

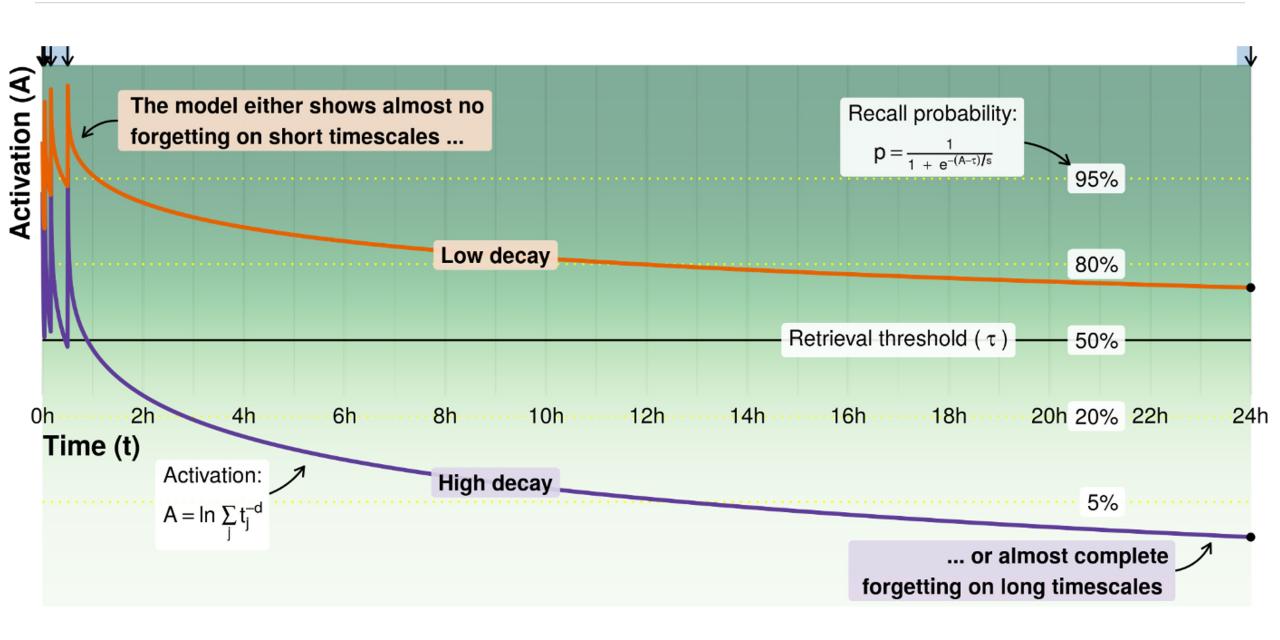
- Can we predict longer-term retention from retrieval practice performance?
- In terms of ACT-R: Can a single model describe cognition on multiple different timescales (seconds/minutes hours/days/weeks)?
- Are these predictions sufficiently robust to be used in practical applications?



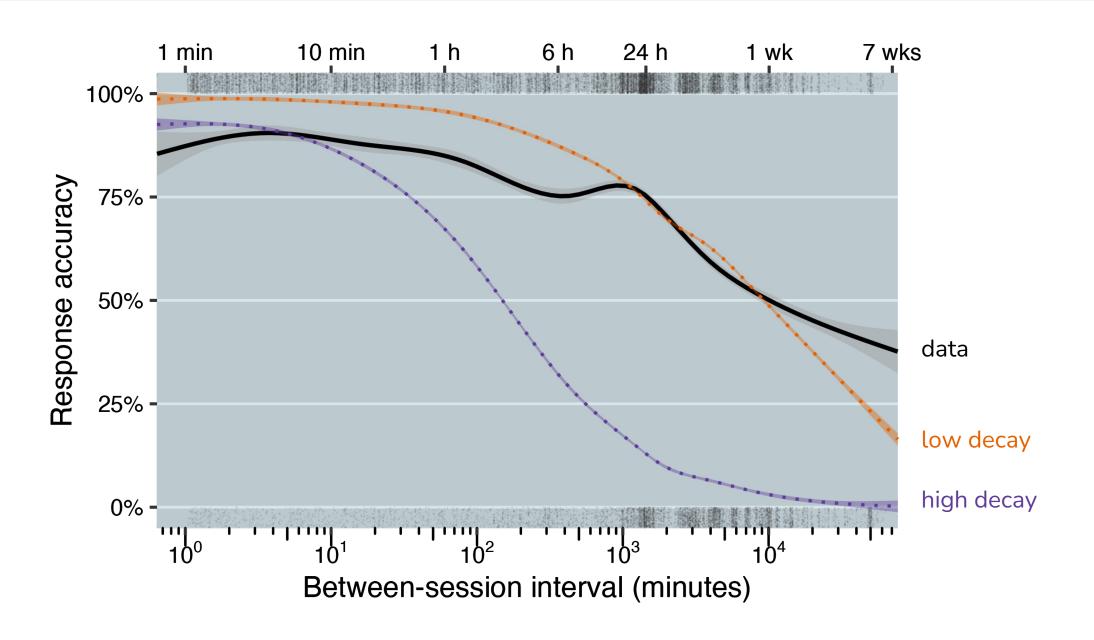








## MemoryLab How does ACT-R do on our multi-session retrieval practice data?



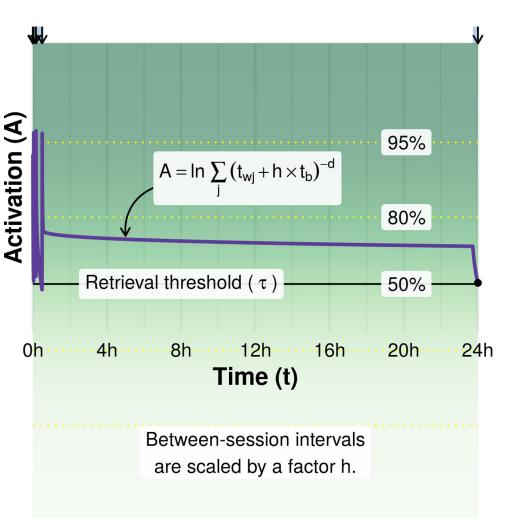
#### Solution: "psychological time"?

Shrink between-session intervals by a scaling factor h.

- There's less decay between sessions than expected based on elapsed clock time Elliott & Anderson (1995); Anderson, Fincham, & Douglass (1999); Pavlik & Anderson (2003)
- Is forgetting slower because of fewer intervening events? See also: context drift in SAM (Mensink & Raaijmakers, 1988), MCM (Mozer et al., 2009)
- "Slowed-clock" model: scale between-session intervals by a factor h (between 0 and 1).
  Different studies find different values: 0.00046, 0.0172, 0.025, 0.031

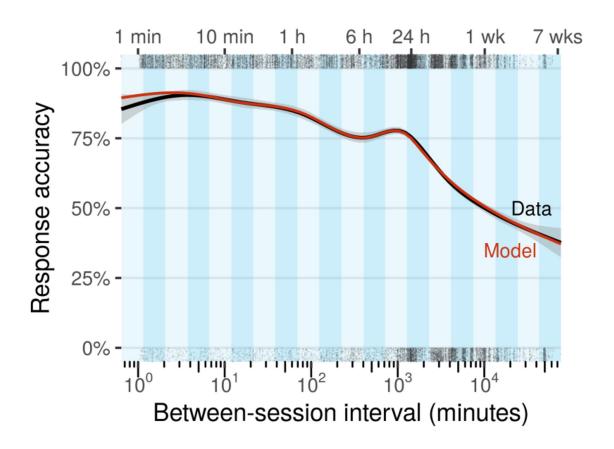
Pavlik, Bolster, Wu, Koedinger, & Macwhinney (2008); Pavlik & Anderson (2008); Pavlik & Anderson (2005); Pavlik & Anderson (2003)

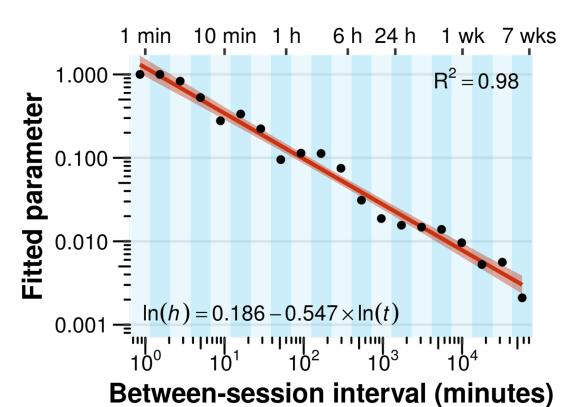
What is the right value of h? Does it depend on the interval?



### Finding a time-variant h(t) in three steps

- Bin learning sequences based on between-session interval (here: 20 bins)
- Find best-fitting h for each bin
- Fit function h(t)

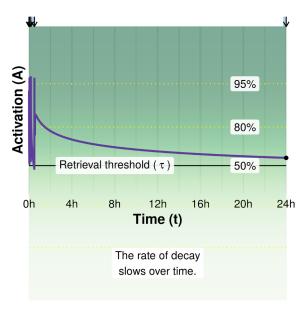




#### But there are (at least) two other solutions!

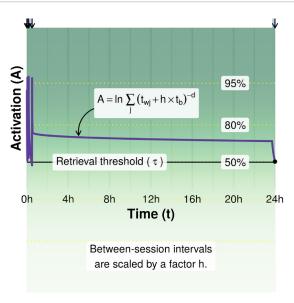
#### Scaling between-session intervals by a scaling factor h(t)

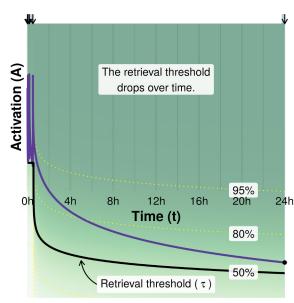
Less interference  $\rightarrow$  shorter "psychological time". Bridge to context-drift-based accounts of forgetting.



#### Time-variant decay d(t)

Lower decay over time → ever stronger "persistence" consolidation. See Ribot's gradient: older memories are more resistant to disruption.



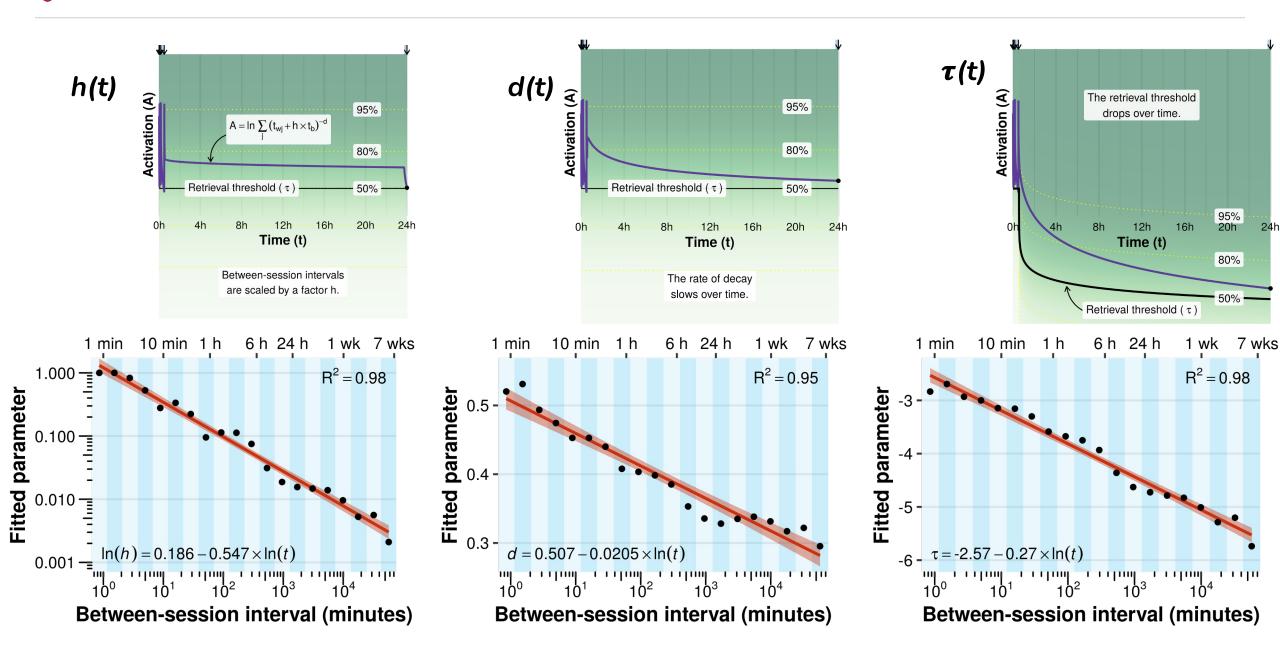


#### Time-variant *retrieval threshold* $\tau(t)$

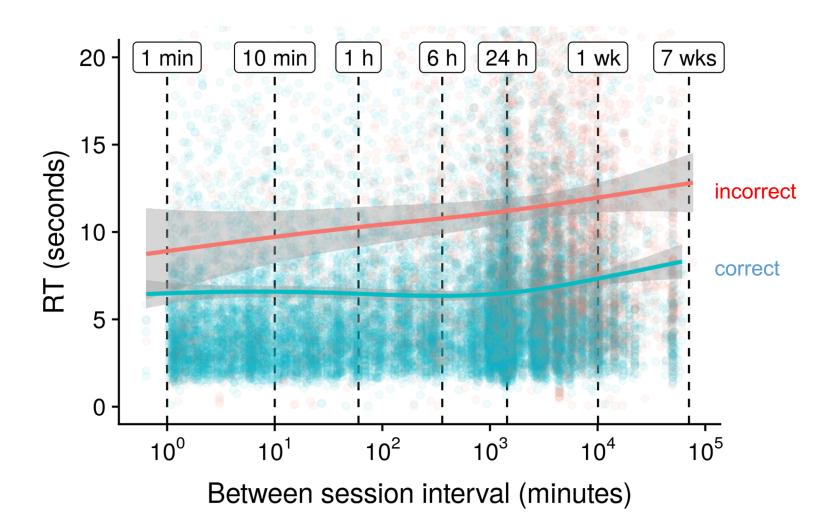
Lower threshold over time → items retrievable at lower activation. Suggests an increase in potential invested retrieval effort.



#### MemoryLab All three parameters change predictably with interval



#### Conflicting evidence in RT

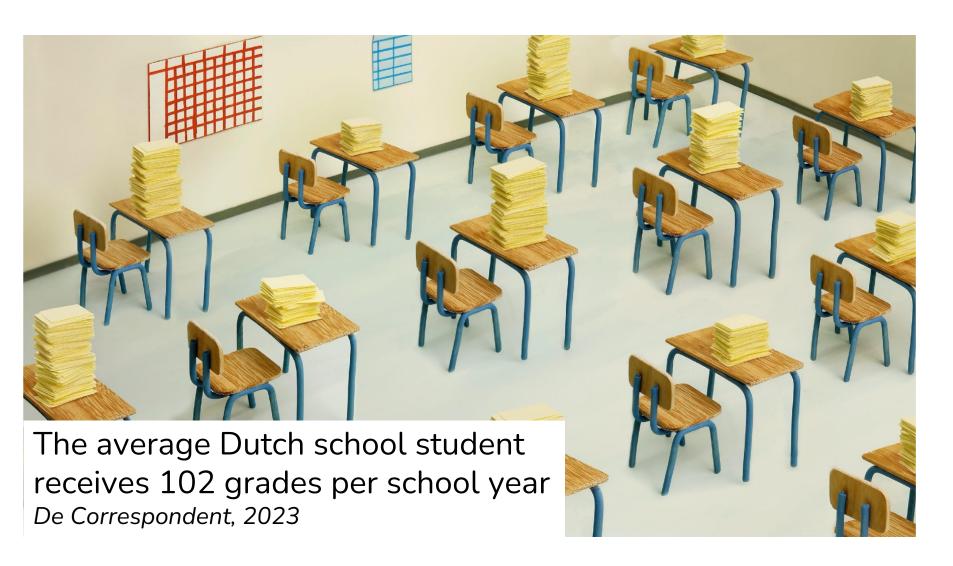


$$RT_f = F * e^{-\tau} + t_{er}$$

$$RT_c = F \cdot e^{-A_c} + t_{er}$$



## Can predictions from practice replace traditional knowledge tests?











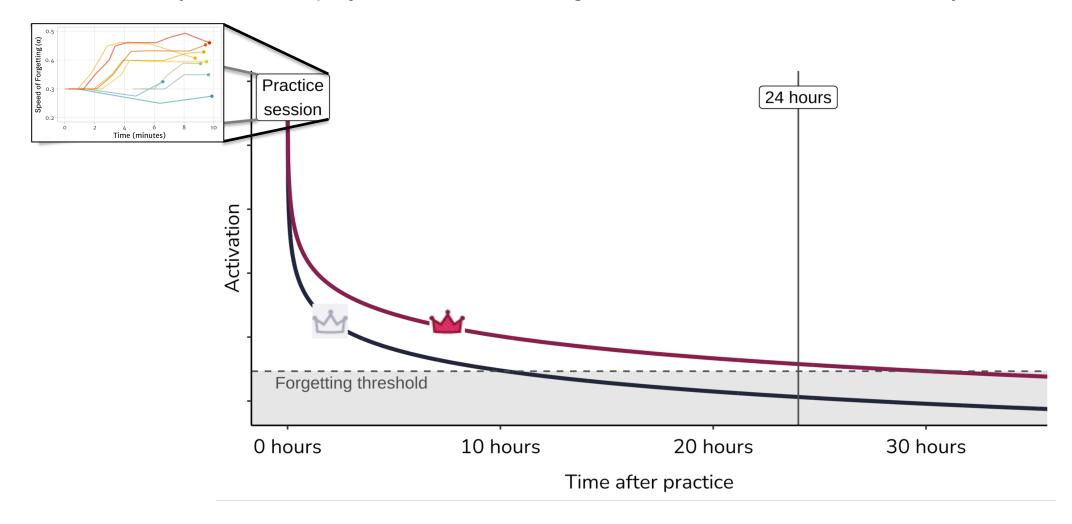


## Can predictions from practice replace traditional knowledge tests?



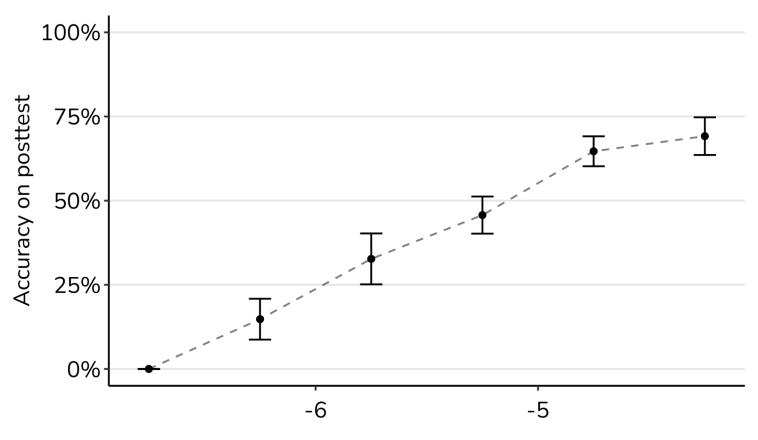
## Model-based assessment of mastery

Define mastery in terms of projected retention, using the same threshold-based memory model:





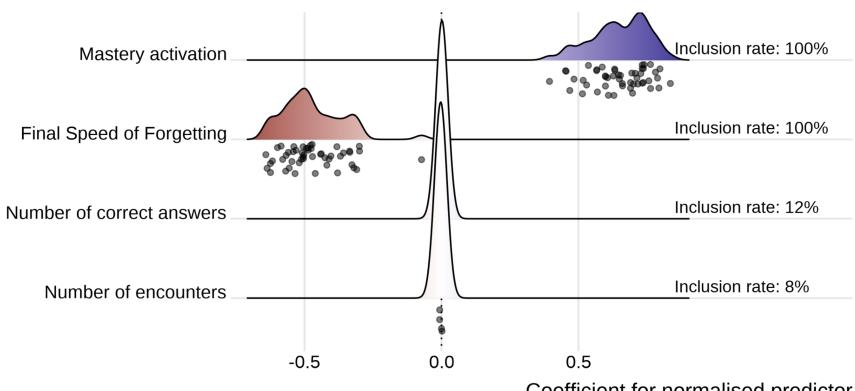
## Memory activation predicts test performance



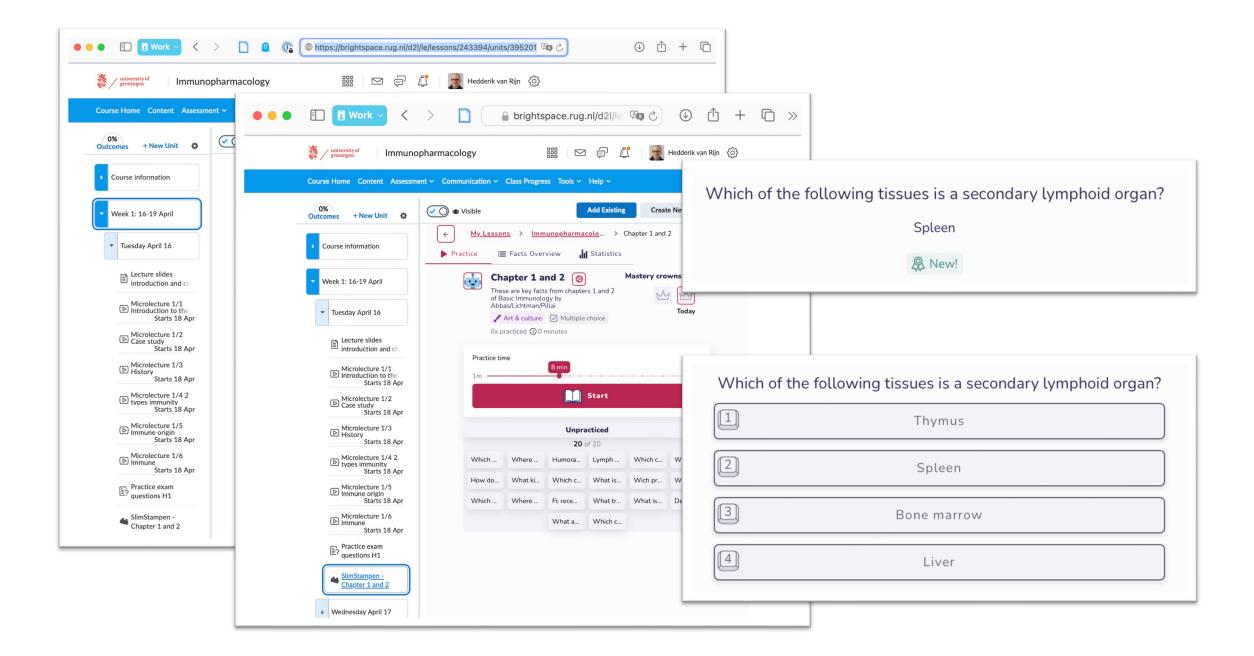
Mastery activation at end of study session



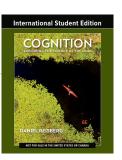
## Model-based predictions outperform "raw" measures



Coefficient for normalised predictor LASSO model fitted through K-fold cross-validation



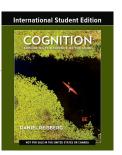




## **Cognitive Psychology**

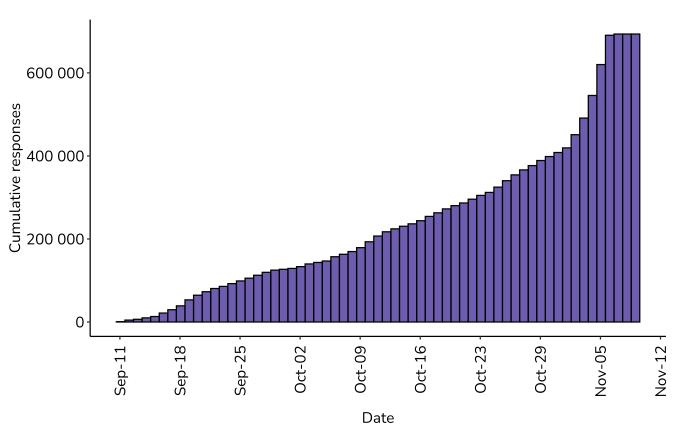
- Students have to know all glossary items by heart (of some of these, the definition will be given on the exam, with the answer being the term)
  - Account for 30% of total grade
- Previous years: students could use MemoryLab to study (=> 6.2/10 grade)
- 2023-2024: *if* two Mastery Credits per chapter were obtained, students were garanteed a *7.5/10 grade* on the fact-part of the exam.
  - The exam could also be taken without MemoryLab studying.





## **Students Start Earlier!**

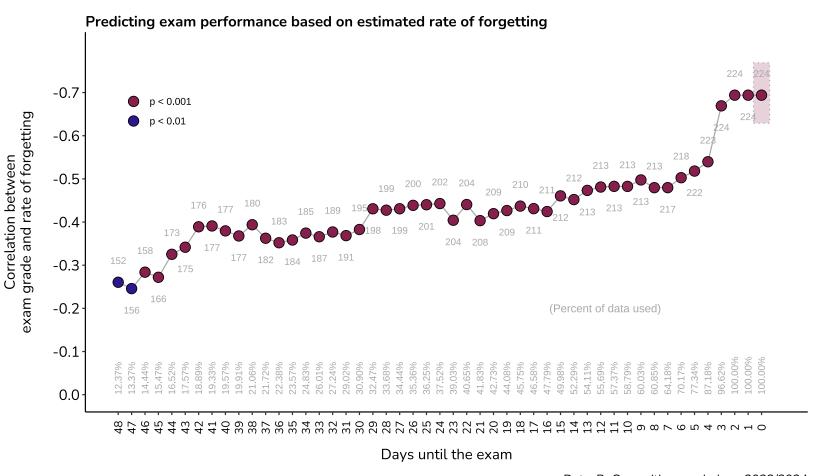
#### **Cumulative usage**



Data: RuG Cognitive Psychology 2023/2024

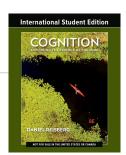


## **Prediction accuracy increases**

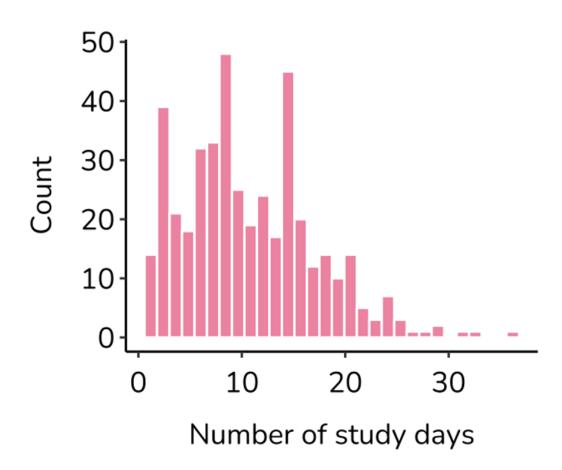


Data: RuG cognitive psychology 2023/2024

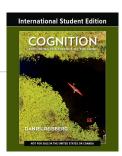




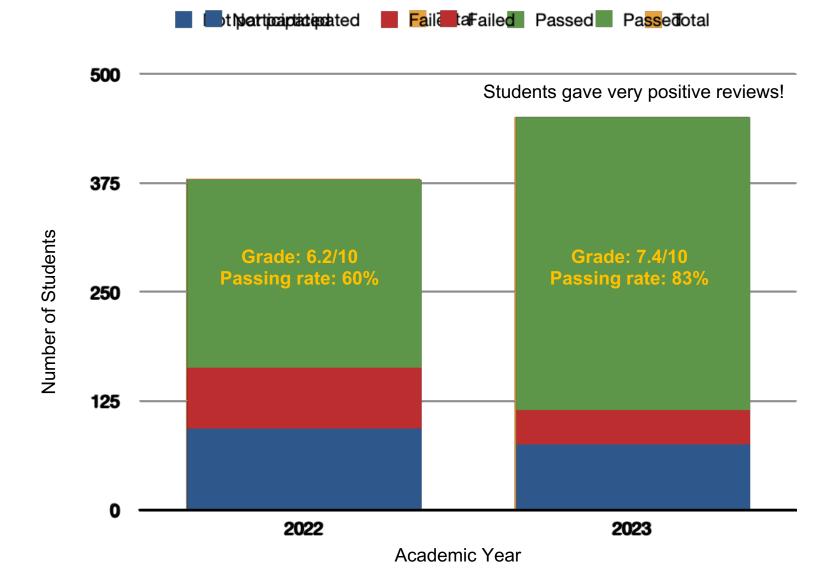
## **Students Distribute Learning**







## **Proof of the Pudding**



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