



university of
 groningen

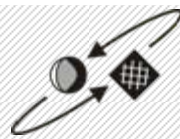
Taking the Teacher's Perspective for User Modeling in Complex Domains

Christian P. Janssen & Hedderik van Rijn

cjanssen@ai.rug.nl

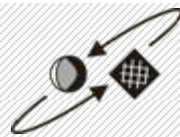
www.ai.rug.nl/~cjanssen

www.ai.rug.nl/cogmod



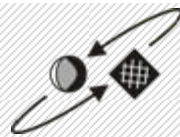


- › Games might be a good training environment
(Aldrich, 2005; Gee, 2003, Michael and Chen, 2005)
- › Serious games: games that focus on more than fun, by training specific content or skill.
- › Mentioned advantages of serious games: (Gee, 2003)
 - It's a natural medium to use for young people
 - Fun might encourage longer learning
 - Learning by doing instead of learning theory outside of context



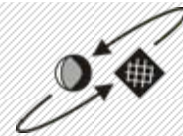
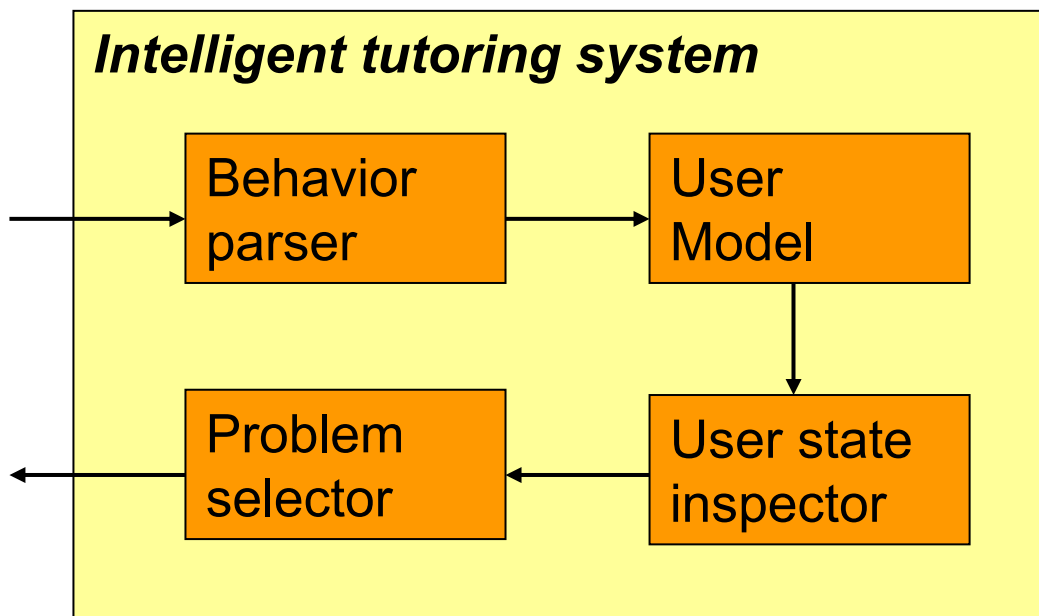


- › Question: how can the training be focused most on the challenges of the player at hand?
 - Develop a user model with estimated capacities of student
 - Adapt training using this user model
- › This has similarities with the domain of intelligent tutoring systems
- › In this presentation a new method for user modeling in complex domains will be presented



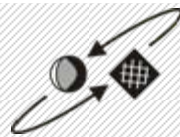


15
27+
?



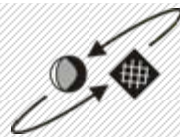


- › Normal approaches are often only applicable in hierarchical & well-studied domains
- › For example: model tracing (Anderson & Gluck, 2001)
 - Tries to develop a model of the (in)correct knowledge of a student
 - These systems map a user's actions to specific mental processes





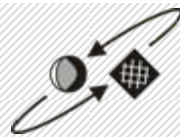
- › Games are often rich environments with a lot of interaction possibilities
 - It is difficult to map each specific action to a specific mental process
- › Problem: incorrect mapping might lead to an *incorrect user model*, and therefore to incorrect training adaptation
- › Possible solution: restrict the set of interactions
 - Serious gaming: no! Multiple routes (Gee, 2003)
 - Intelligent tutoring: no! Be *flexible, adapt* to the individual (Ohlsson, 1986)





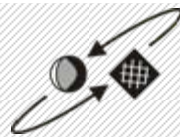
Teacher Modeling

- › Possible solution for domains that lack theories of hierarchy: Teacher Modeling
- › Main idea: as it is difficult to model the thoughts of a *student/player*, try to model aspects of the thoughts of the *teacher*



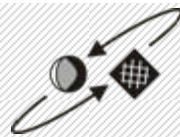


- › Assess the capacities of the student
 - Assessments play a critical role in normal (classroom) learning settings (Shepard, 2000)
- › Act like a teacher who:
 - Does not always know *what* a student is thinking
 - Does know *when* a student is making a mistake
 - Can make an *assessment over time* that states what skills are performed correct and incorrect
 - Needs *observations to prevent forgetting* the assessment





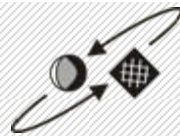
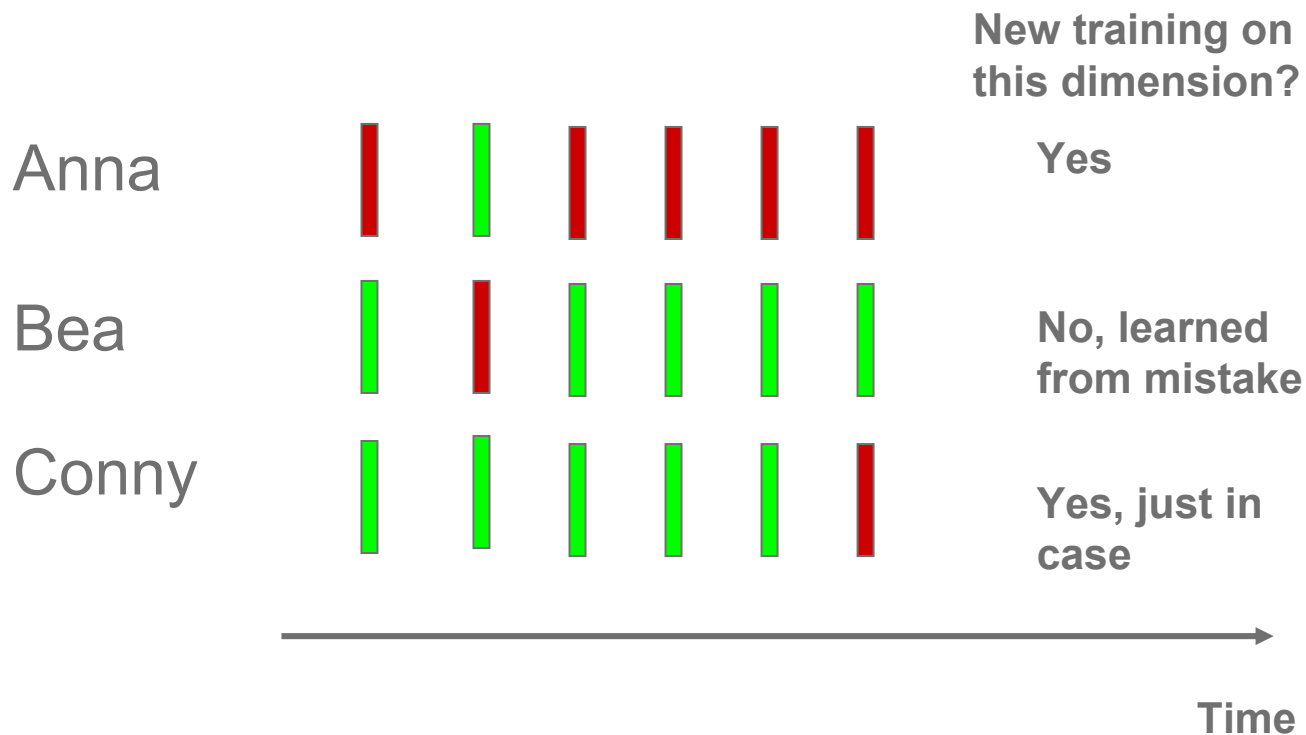
- › Teacher modeling requires:
 - Set of training objectives: training dimensions
 - Will often be very broad, as fine-grained (model tracing) approaches are difficult to apply
 - Set of training exercises
 - With multiple outcomes
 - And an indication of the student's behavior for each outcome and each dimension





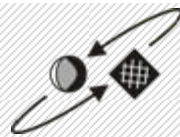
Teacher Modeling

› What would a teacher think of the following students?



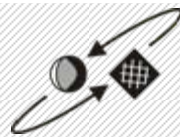


- › Important to keep track of:
 - Positive and negative occurrences
 - Frequency of occurrence
 - Recency of occurrence
- › Formal theories of declarative memory can take this into account (Anderson & Schooler, 1991)
 - Information containing unit: Chunk
- › Chunks have an activation level
 - Represents usefulness of chunk in the past (Anderson *et al.*, 2004; Anderson and Lebiere, 1998)
 - In case of our model: “how representative is this behavior for the current student?”





- › Keep track of a student's skills using three chunks for each dimension:
 - Amount of *positive completed* exercises
 - Amount of *negative completed* exercises
 - Overall *amount of training*

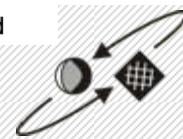
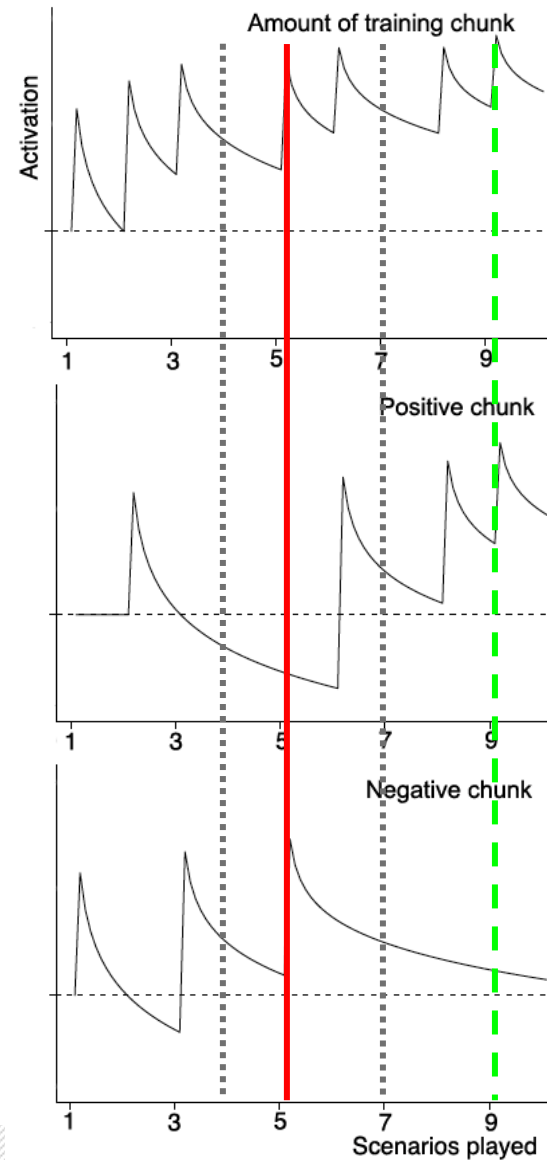




Example

- › Base-level learning equation:

$$A_i = \ln \left(\sum_{j=1}^n t_j^{-d} \right)$$





1. Transform activation score into probabilistic score
(Anderson *et al.*, 2004; Anderson and Lebiere, 1998)

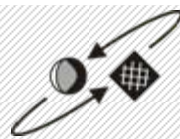
$$P_i = \frac{1}{1 + e^{-(A_i - \tau)/s}}$$

“How certain am I that the student performs the behavior in this manner?”

2. Overall measure of dimension performance:

$$Total_j = P_{positive(j)} + (1 - P_{negative(j)}) + P_{training(j)}$$

3. Train dimension with lowest total score

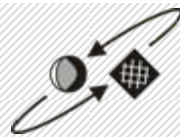




Does it work?

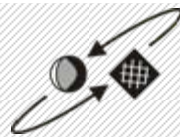
- › *“Does the model correctly adapt to the characteristics of individual students?”*
 - By focusing training mostly on the most challenging dimensions

- › Tested in simulations





- › Train 4 dimensions during two hundred training exercises
 - Each exercise “trains” one dimension and has two possible outcomes: positive or negative
- › Time between trainings was constant
- › Always keep training varied:
 - Dimension with highest score on amount of training chunk was excluded from selection

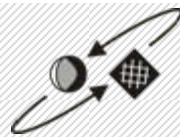




- › 4 types of students were simulated

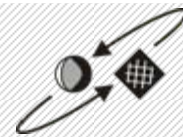
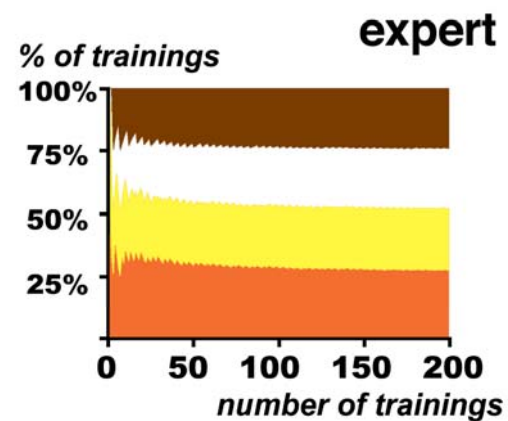
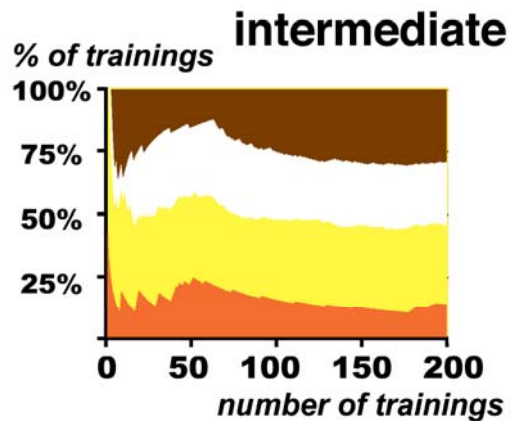
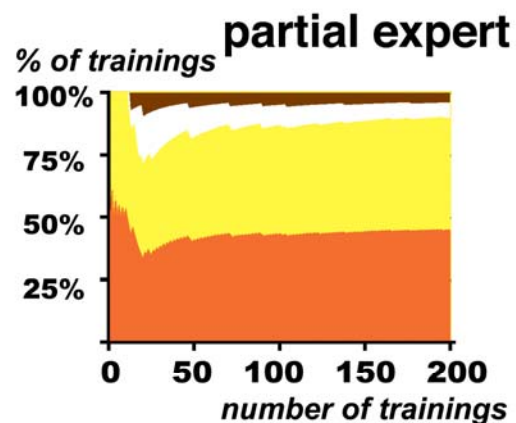
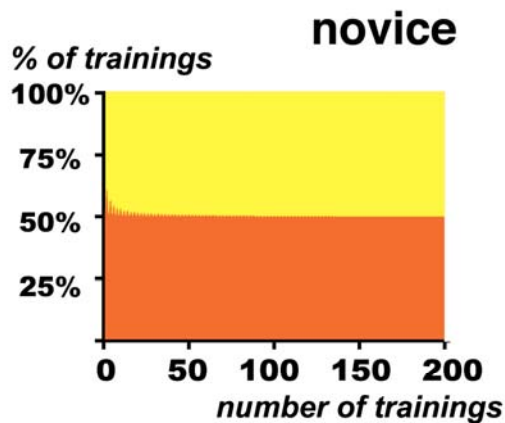
	% of questions correct , per dimension			
	A	B	C	D
Novice	0%	0%	0%	0%
Intermediate	50%	50%	50%	50%
Partial expert	25%	25%	75%	100%
Expert	75%	100%	100%	100%

- › Simulations had a consistent character
 - no “learning”





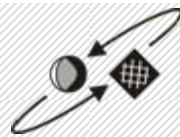
Simulation 1





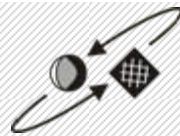
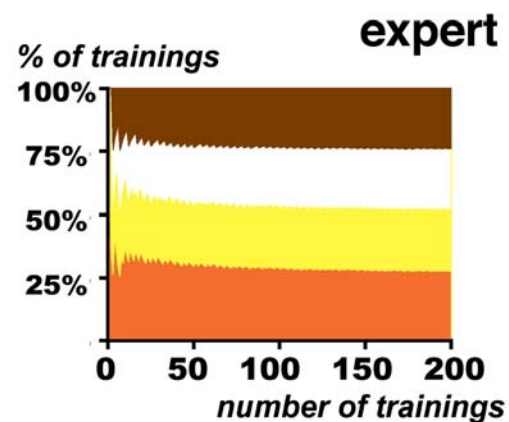
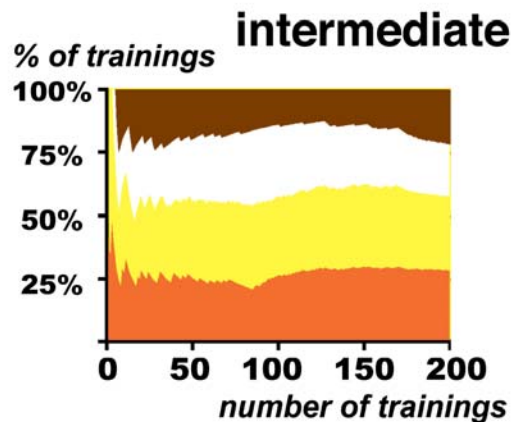
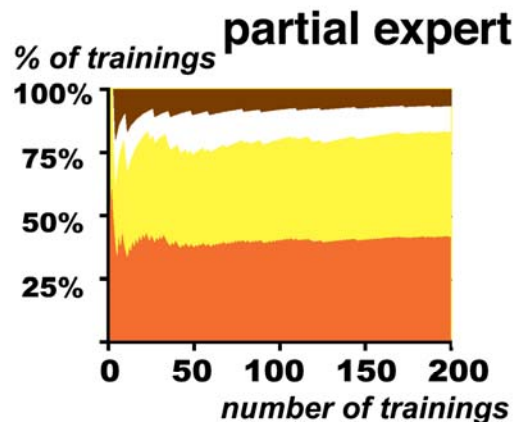
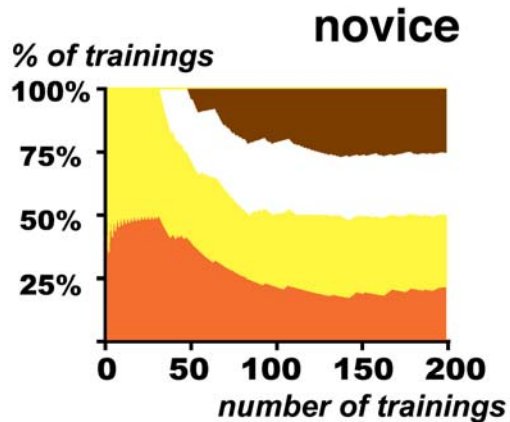
*“What if students learn during the training,
and change their behavior?”*

- › Characters became 1% more likely to complete an exercise correctly, each time they trained the dimension



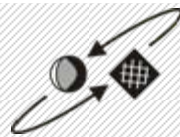


Simulation 2



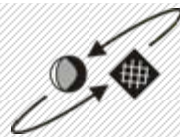


- › Research on serious games might benefit from techniques for intelligent tutoring
 - Hierarchical domain -> traditional methods such as model tracing
 - No (complete) theory of hierarchy -> teacher modeling
- › Mechanism of teacher modeling:
 - Identify set of training dimensions
 - Use 3 chunks per dimension: amount of training, positive encounters and negative encounters
 - Combine chunks in a total dimension score and train dimensions with a low dimension score





- › Teacher modeling
 - Takes frequency and recency of training (observations) into account
 - Uses continuous assessments to get insight in student performance
- › The method has been tested in a simulation
 - It adapts training selection to the individual & adapts to (changing) behavior
- › Each specific domain will require user studies to test the learning gain
- › We are currently developing a serious game in which the user model is tested (Janssen *et al.*, 2007)
- › Method might also be useful for other recommender systems, if they have to categorize broad interests





university of
groningen

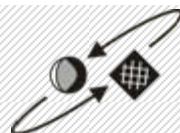
Thank you for your attention Questions?

Christian P. Janssen & Hedderik van Rijn

cjanssen@ai.rug.nl

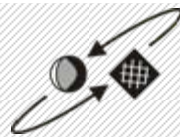
www.ai.rug.nl/~cjanssen

www.ai.rug.nl/cogmod





- › Aldrich, C. (2005). *Learning by doing: A comprehensive guide to simulations, computer games, and pedagogy in e-learning and other educational experiences*. San Francisco: Pfeiffer.
- › Anderson, J. R., Bothell, D., Byrne, M. D., Douglass, S., Lebiere, C., & Qin, Y. (2004). An integrated theory of the mind. *Psychological Review*, 111(4), 1036-1060.
- › Anderson, J. R., & Gluck, K. (2001). What role do cognitive architectures play in intelligent tutoring systems. In S. M. Carver & D. Klahr (Eds.), *Cognition & instruction: Twenty-five years of progress* (pp. 227-262). Mahwah, NJ: Erlbaum.
- › Anderson, J. R., & Lebiere, C. J. (1998). *The atomic components of thought*. New Jersey: Lawrence Erlbaum Associates.
- › Anderson, J. R., & Schooler, L. J. (1991). Reflections of the environment in memory. *Psychological Science*, 2(6), 396-408.





- › Gee, J. P. (2003). *What video games have to teach us about learning and literacy*. New York: Palgrave Macmillan.
- › Janssen, C. P., Van Rijn, H., Van Liempd, G., & Van der Pompe, G. (2007). *User modeling for training recommendation in a depression prevention game*. In: Proceedings of the first NSVKI student conference, Nijmegen, The Netherlands, 29-35.
- › Michael, D. R., & Chen, S. L. (2005). *Serious games: Games that educate, train, and inform*. Boston: Thomson course technology.
- › Ohlsson, S. (1986). Some principles of intelligent tutoring. *Instructional Science*, 14(3), 293-326.
- › Shepard, L. A. (2000). The role of assessment in a learning culture. *Educational Researcher*, 29(7), 4-14.

