



Reaping the Rewards of Teaching ACT-R: Class Projects Spring 2002

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Overview

- ✍ Structure of the class
- ✍ Thoughts on 5.0 vs. 4.0
- ✍ Maurier's project: Learning in PRP
- ✍ Fick's project: Contingent capture



Class Structure

- ✍ Starts with general readings on computation and cognition (Pylyshyn, Simon)
- ✍ A couple weeks of connectionism
 - ✍ Readings
 - ✍ Building simple models
- ✍ Transition to symbolic systems
 - ✍ Fodor & Pylyshyn critique
 - ✍ Newell on symbolic systems
- ✍ ACT-R the rest of the way
 - ✍ Readings
 - ✍ CMU tutorial units
 - ✍ Projects



Project Requirement

- ✍ Select a published data set
- ✍ Model it!
 - ✍ Did not have to be with ACT-R
 - ✍ Most used ACT anyway
- ✍ Grades were based more on showing what they learned than on r-squared of fits
- ✍ I wrote all the supporting Lisp code
 - ✍ 9 students, only 8 of whom used ACT



Pros and Cons of 5.0 (vs. 4.0)

✍ Pros

- ✍ More uniform syntax
- ✍ Buffer basis simplifies explaining all the things that happen on a production cycle
 - ✍ Especially dealing with retrieval failures
- ✍ Parallels between declarative retrieval and visual attention
- ✍ Simpler PG-C formulation

✍ Cons

- ✍ Lack of a book like *Atomic Components*
- ✍ Lack of a manual for 5.0
- ✍ Debugging 5.0 seems a little harder

✍ Environment was a wash



David Maurier's Project: Learning in the PRP Paradigm

- ✍ Anyone **not** know the PRP paradigm?
- ✍ Data from Van Selst, Ruthruff, & Johnston (1999)
- ✍ Setup
 - ✍ Task 1: 4-choice tone discrimination with vocal response
 - ✍ Task 2: 8-choice visual character discrimination with manual response
 - ✍ SOAs of 17, 67, 150, 250, 450, and 850 ms
 - ✍ A bunch of difficulty manipulations
- ✍ Focus on first phase of the experiment
- ✍ 18 sessions, roughly 1 hour each



Maurier's Model

- ✍ Attempt to model endpoints
 - ✍ Session 1 performance
 - ✍ Session 18 performance
 - ✍ No initial attempt to have ACT-R do the learning
- ✍ Each task requires three productions:
 - ✍ Register
 - ✍ Relies on buffer-stuffing
 - ✍ Shifts attention to new location in buffer
 - ✍ Retrieve chunk which maps stimulus to appropriate response
 - ✍ Respond
 - ✍ No complex unlocking

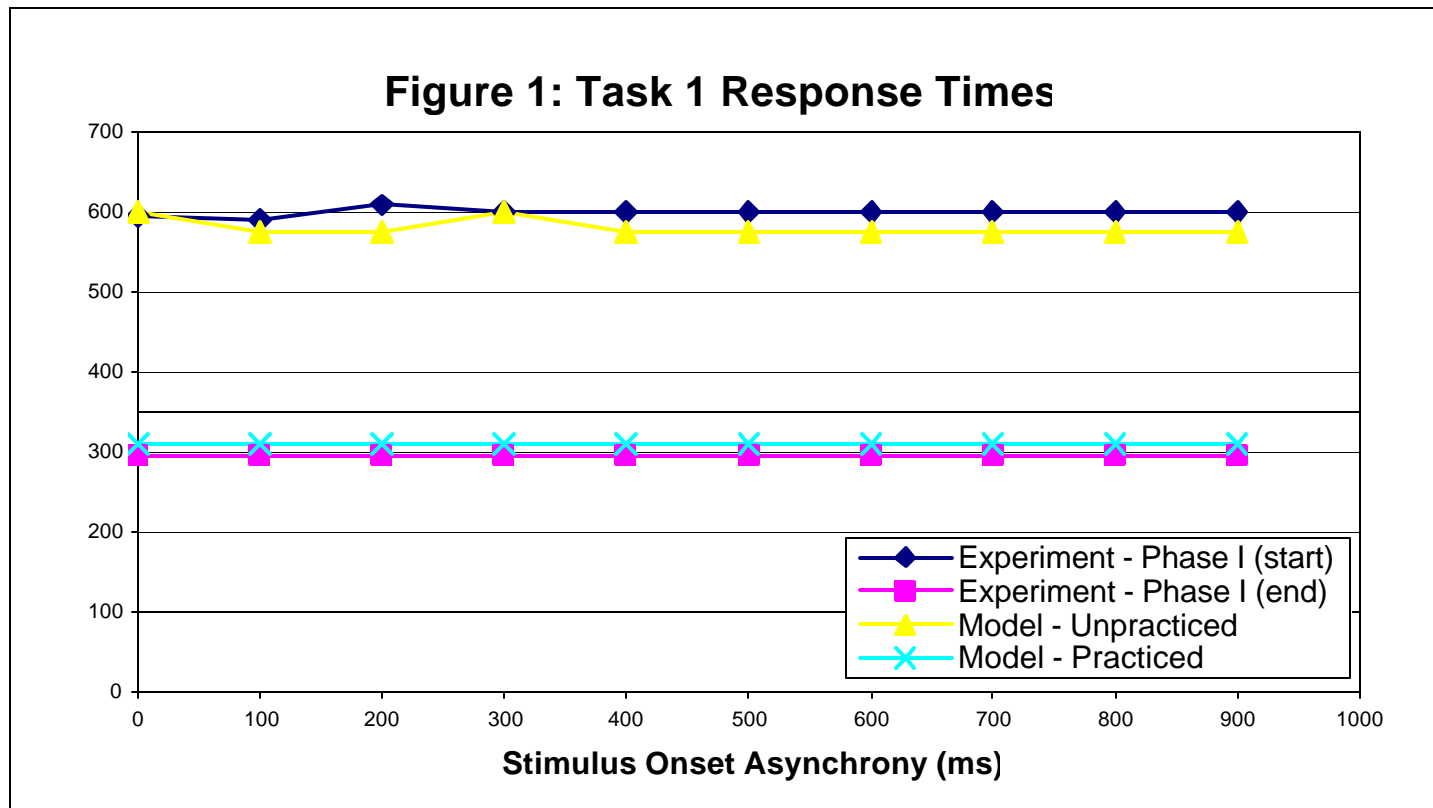


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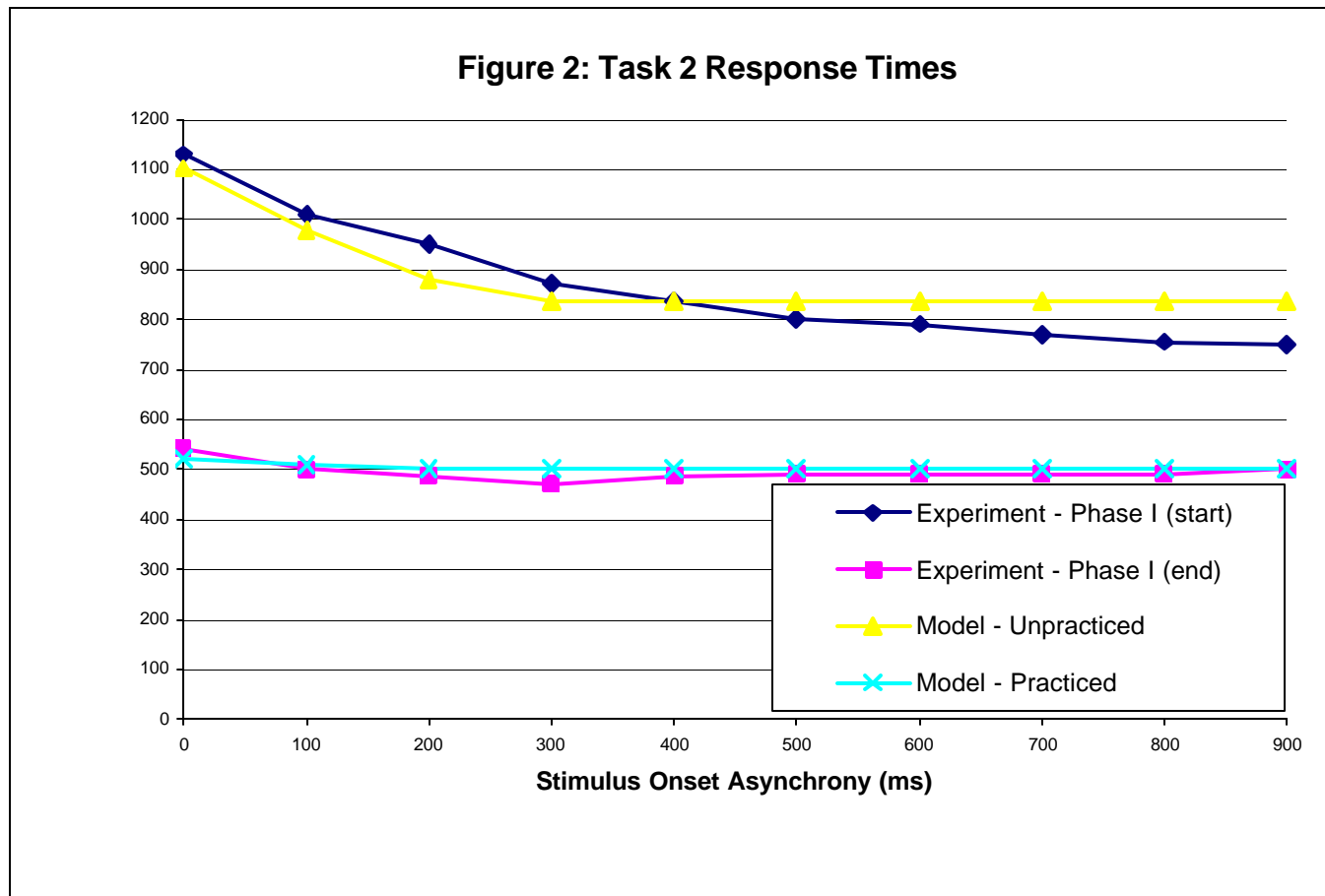
- ✍ Session 1 model
 - ✍ Tone recode 100 ms
 - ✍ SR mapping chunk activation
 - ✍ For Task 1: 1.6
 - ✍ For Task 2: 1.0
 - ✍ Default cognitive cycle time of 50 ms
- ✍ Session 18 model
 - ✍ Tone recode time 70 ms
 - ✍ SR mapping chunk activation to 16 and 15
 - ✍ Cognitive cycle time to 5.5 ms (!)



Fit to Task 1 RTs



Fit to Task 2 RTs



Discussion

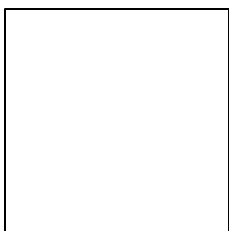
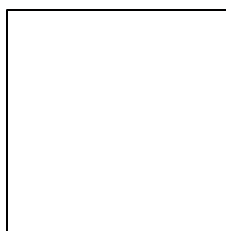
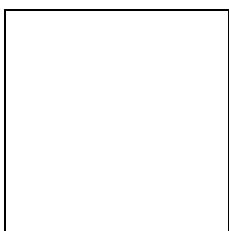
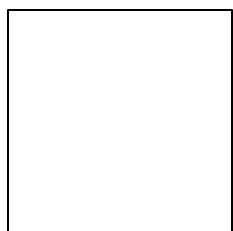
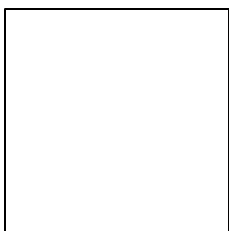
- ✍ Could ACT-R learn it?
 - ✍ Chunk activations would certainly go up a lot with that much practice
 - ✍ However, might not be necessary with production rule learning on
 - ✍ Production learning might also solve the problem of reducing the cycle time
 - ✍ Cannot right now learn to reduce tone recoding time
 - ✍ Not clear if it would learn at the right rate
- ✍ Hopefully we'll have more to report at ICCM

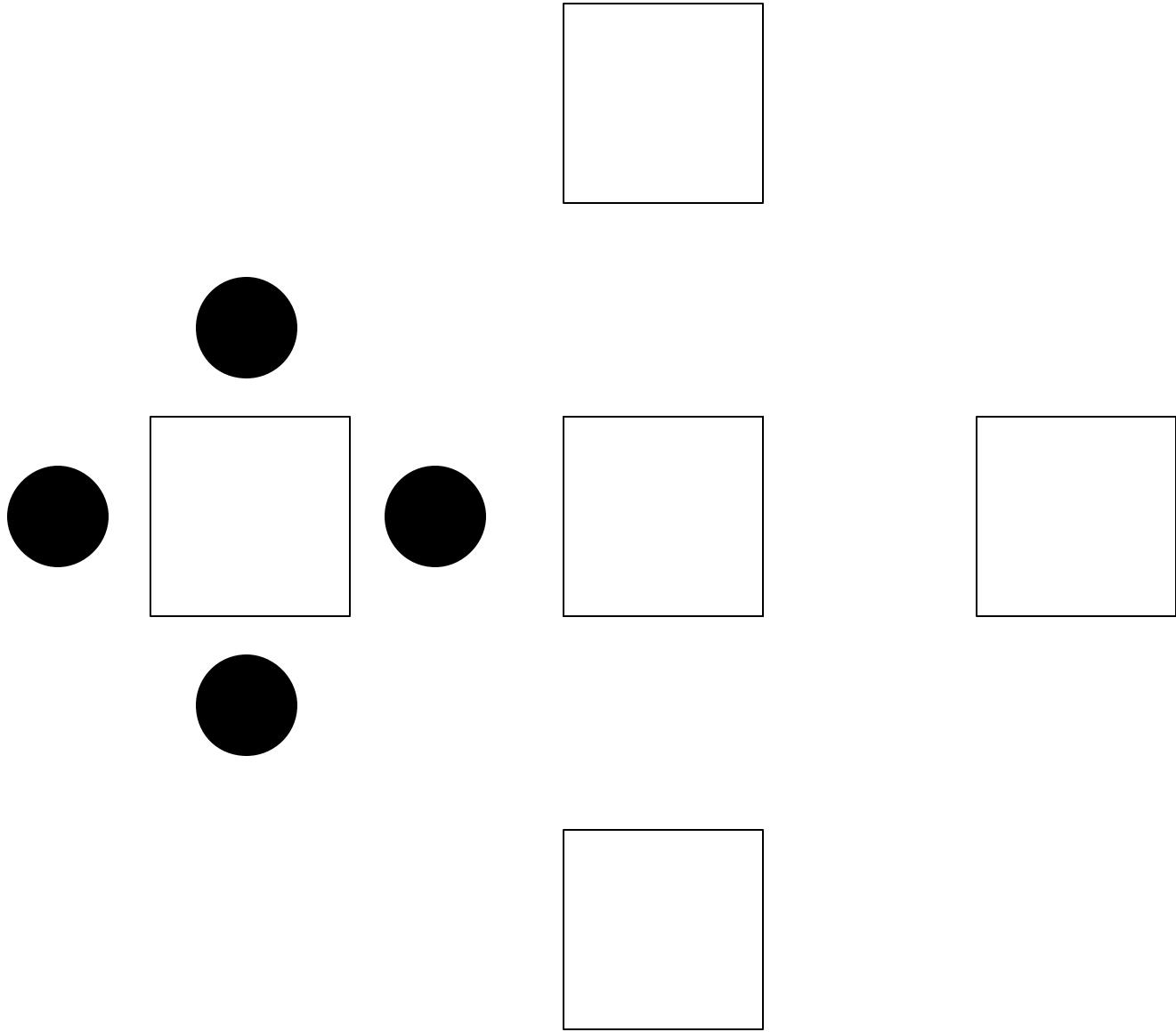


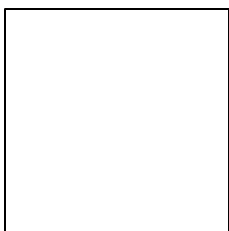
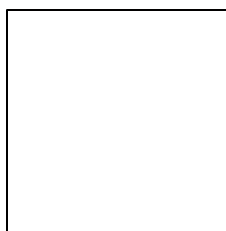
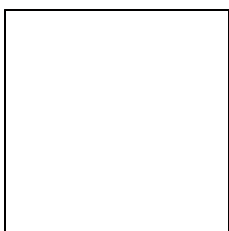
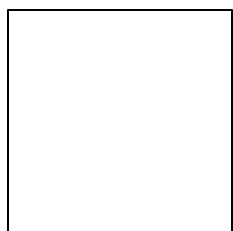
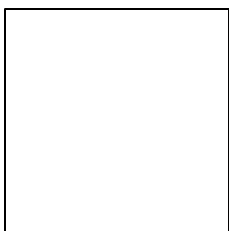
Chris Fick's Project: Contingent Orienting

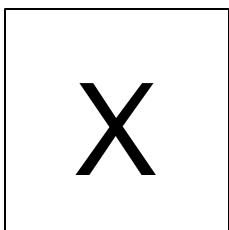
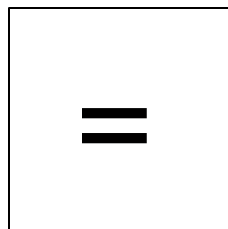
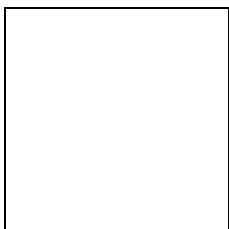
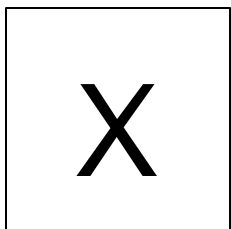
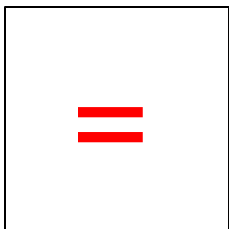
- ✍ Main research question: Do onsets of new visual stimuli “capture” attention?
- ✍ Widely believed to be the case until famous paper by Folk, Remington, & Johnston (1992)
- ✍ Three cue types
 - ✍ Color singleton
 - ✍ Onset singleton
 - ✍ None
- ✍ Two target conditions
 - ✍ Onset singleton
 - ✍ Color singleton
- ✍ Cues could be valid or invalid

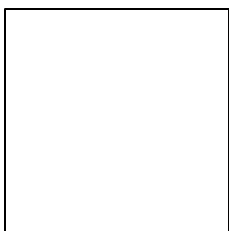
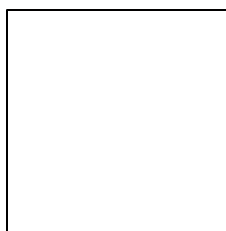
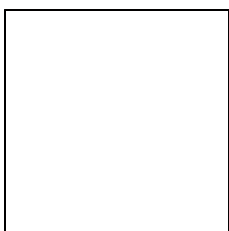
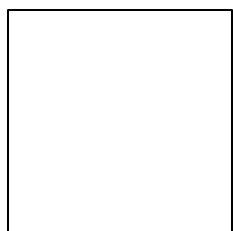
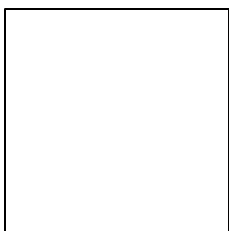


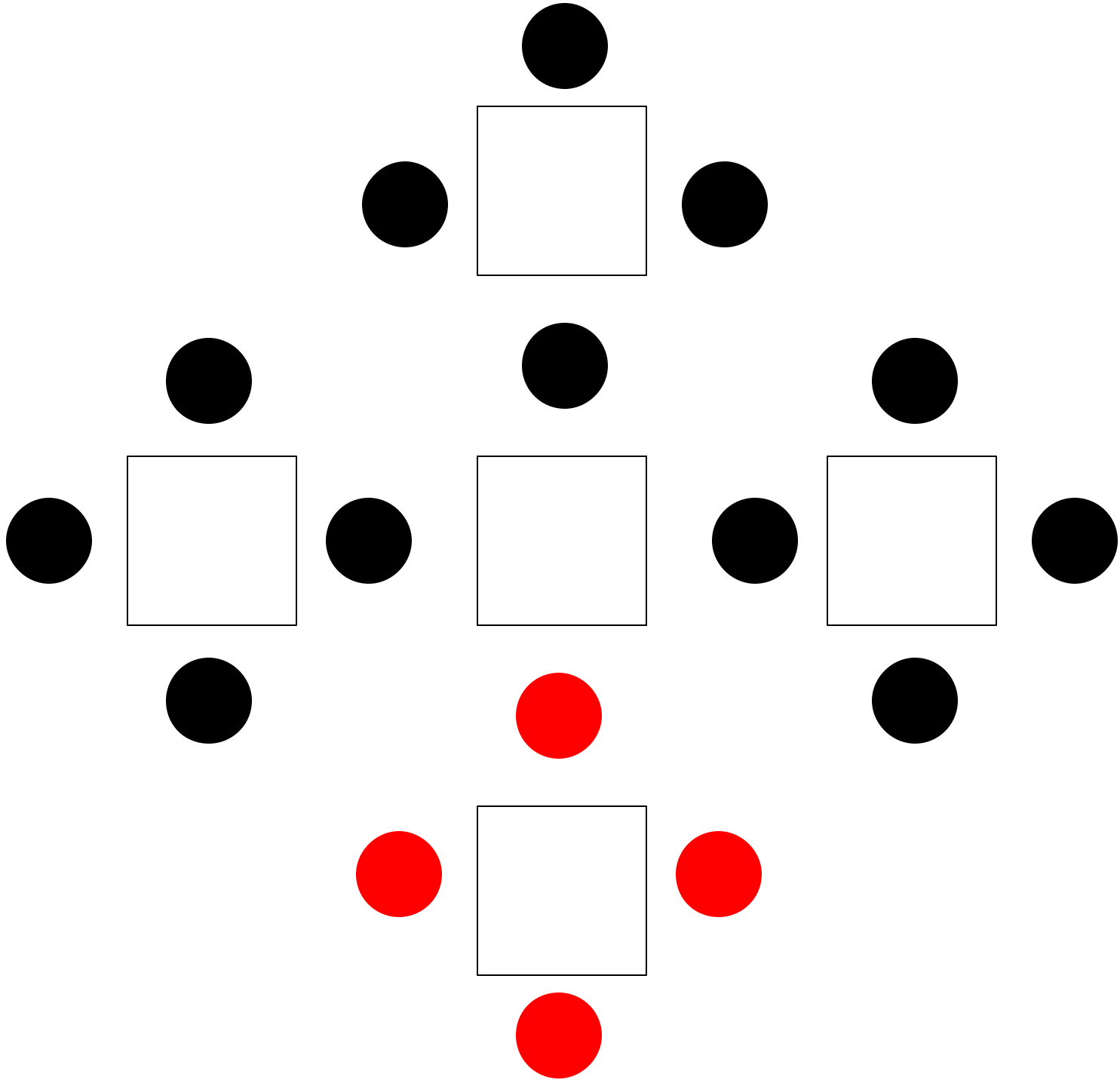


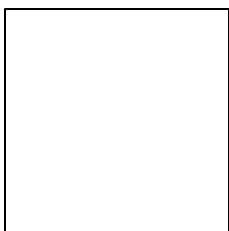
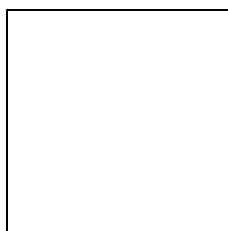
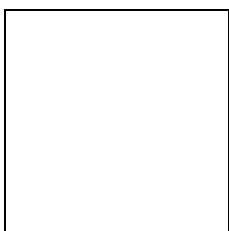
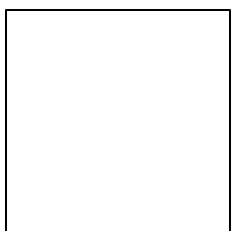
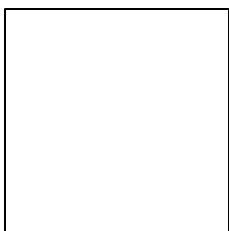


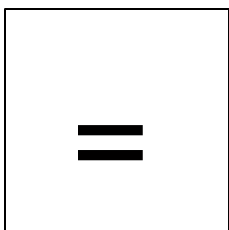
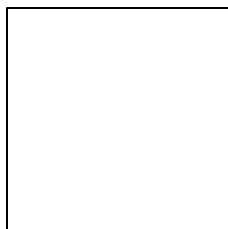
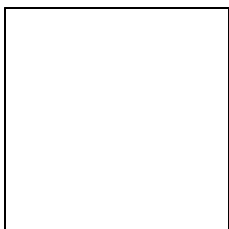
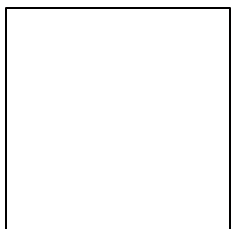
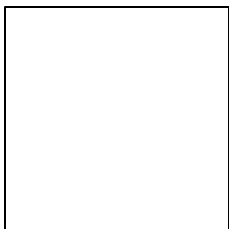






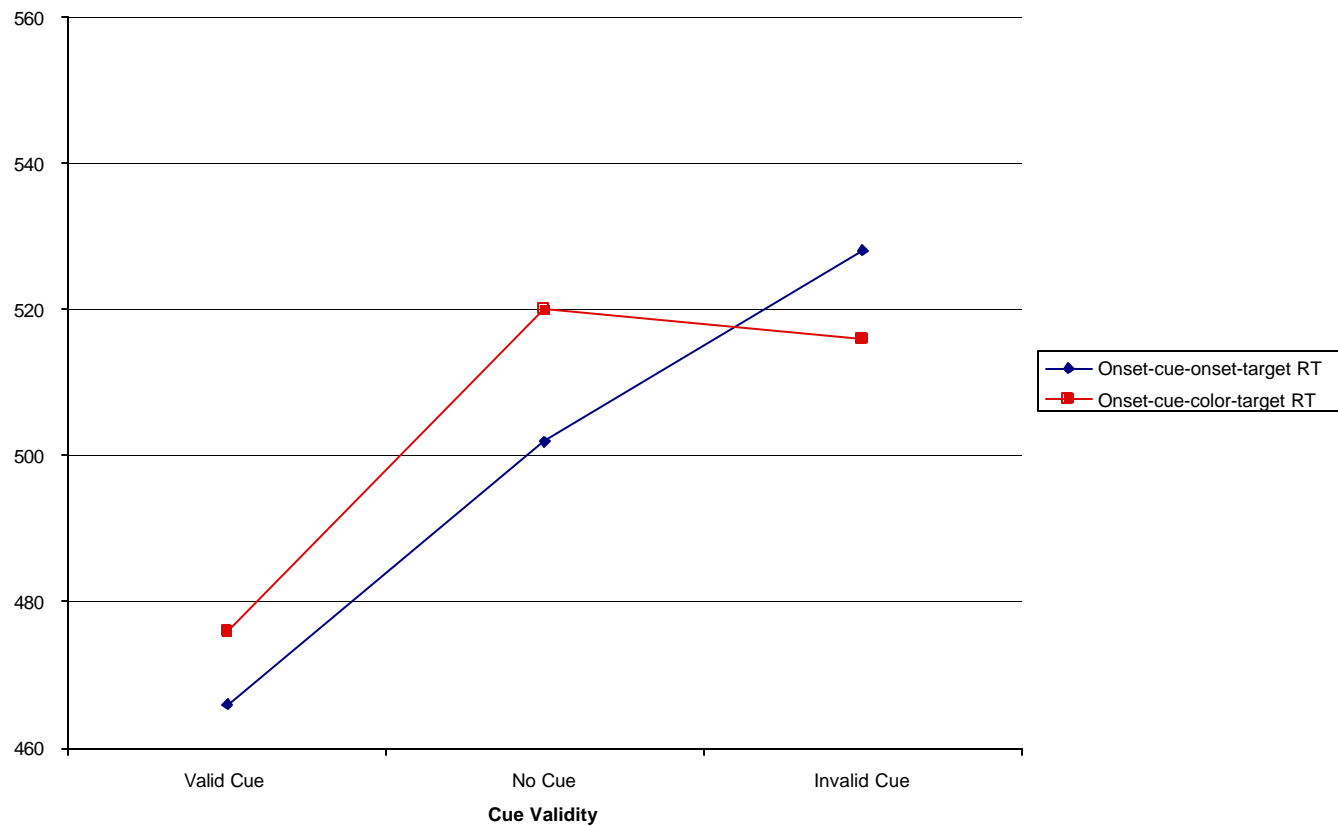




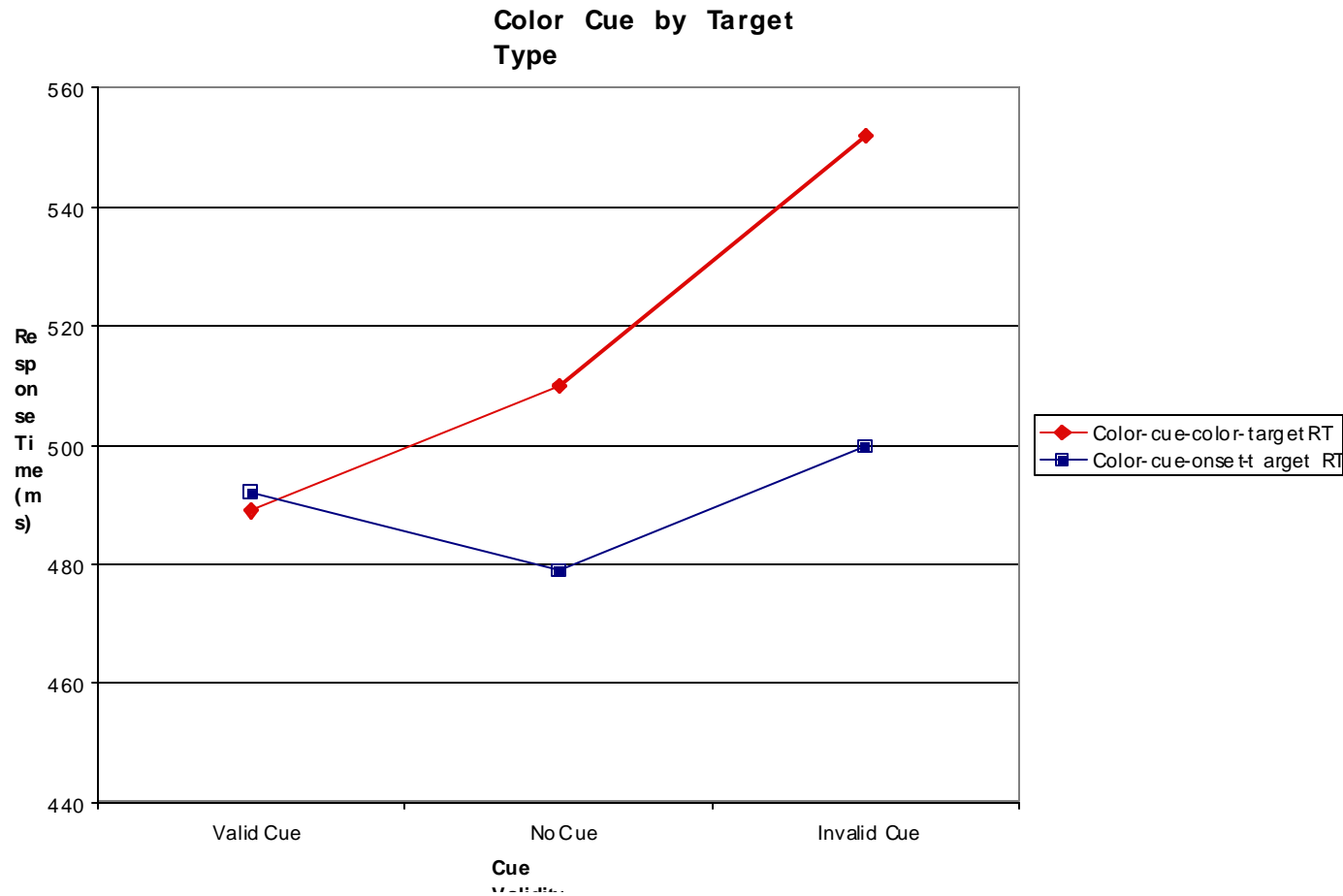


Results: Onset Cues

Onset Cue by Target Type



Results: Color Cues



The Model

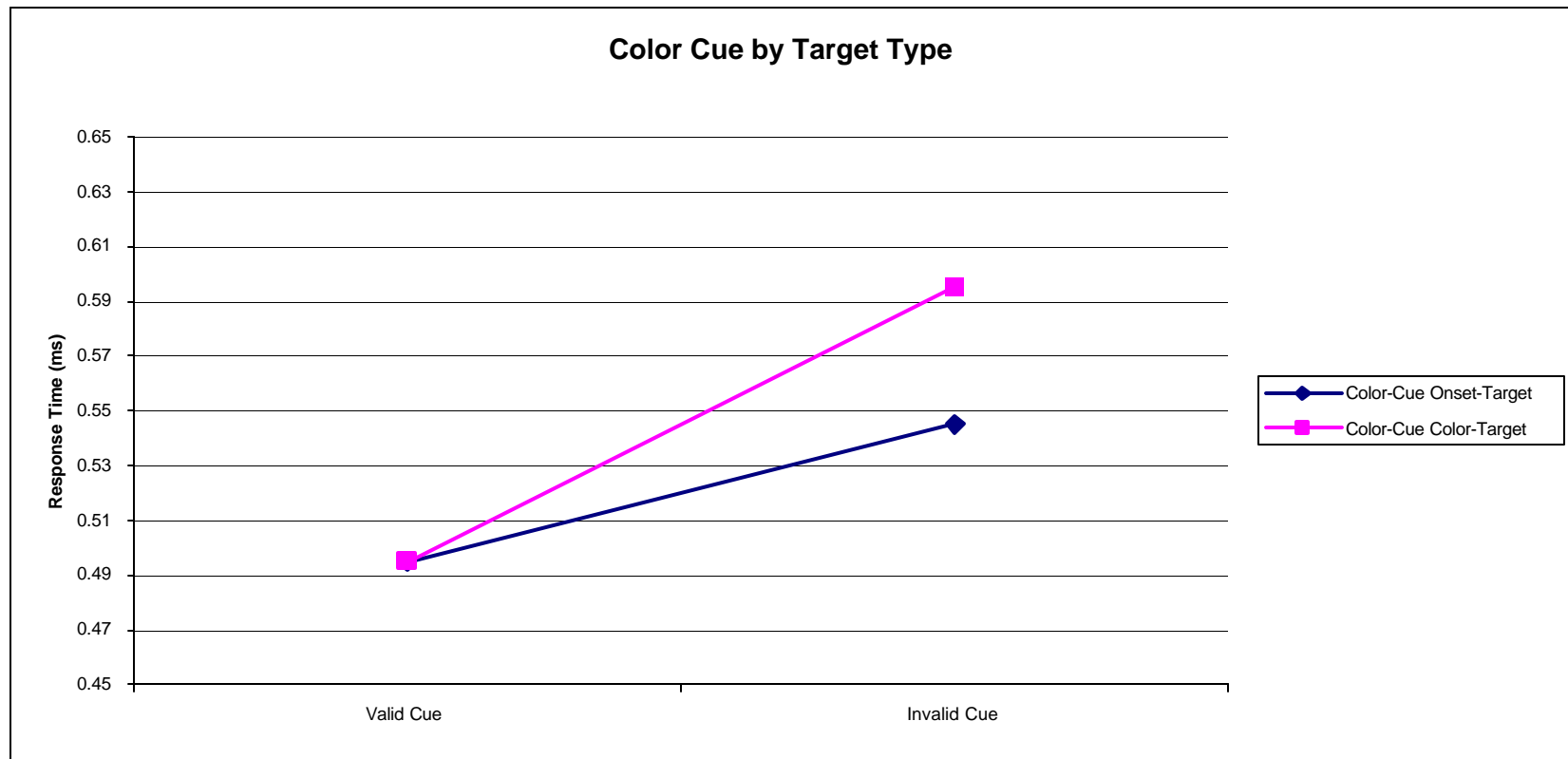
- ✍ Didn't model "no cue" conditions
- ✍ Capitalizes on strange aspect of the original experiment
 - ✍ Trial types were blocked
 - ✍ Including blocking by validity
 - ✍ Model knows if cue is valid
- ✍ Handling cues
 - ✍ Onset cue stuffs buffer
 - ✍ Color cue requires feature search, but that's fast
- ✍ Shift attention contingently depending on validity and cue-target match



Model: Onset Cues



Results: Color Cue



Discussion

- ✍ Model is a little too slow overall
 - ✍ Probably fixable
- ✍ Needs no-cue condition in model
 - ✍ Proposal: have model randomly select location to attend
- ✍ Contingency is not principled, not clear why it would work the way it does
 - ✍ May be a function of the way trials were blocked
 - ✍ Raises question: What happens when trials are not blocked?



Other Projects

- ✍ David Huss will tell you about his shortly
- ✍ Another related serial recall/working memory kind of task
- ✍ Two people modeled mental rotation experiments with somewhat varied levels of success
- ✍ Postcompletion errors
- ✍ Goal management (TOH)



Questions?



Emergency PRP Slide

- ✍ Very simple dual task
- ✍ Two tasks, Task 1 and Task 2
 - ✍ Usually choice RT tasks
- ✍ Stimulus onset for Task 2 stimulus delayed relative to onset of Task 1 stimulus (SOA)
- ✍ Subjects instructed to give priority to Task 1
- ✍ Basic findings
 - ✍ Task 1 RT unaffected by SOA
 - ✍ Task 2 RT a function of SOA; smaller SOA yields higher RT with approximately -1 slope
- ✍ Lots of contention, but used to argue for seriality

