



Cognitive Robotics, Embodied Cognition and Human-Robot Interaction

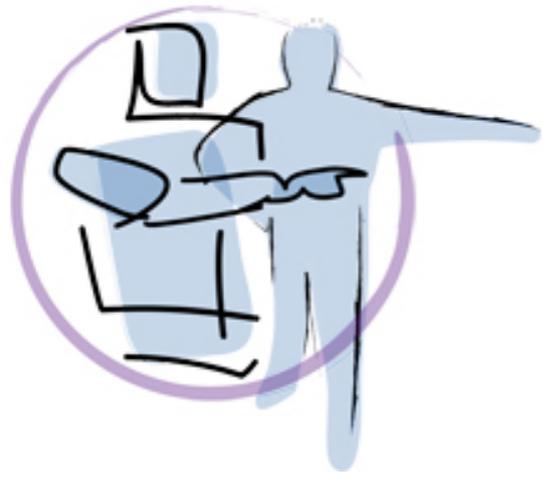
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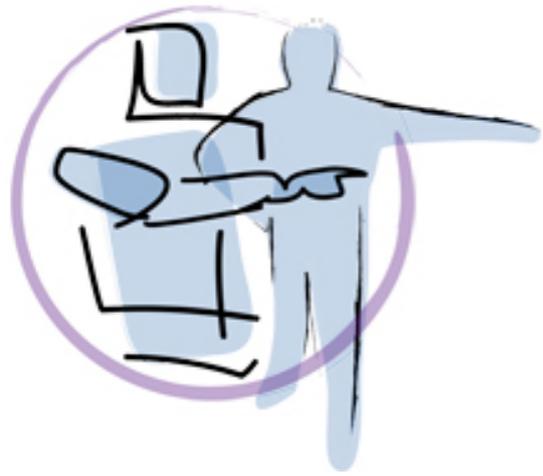
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| 14. ABSTRACT The last decade has seen a strong push for more and better autonomous systems. In many cases, autonomous systems and people need to interact to accomplish joint goals. We have been working on developing autonomous systems that interact with people in a variety of ways. Our approach has been from a strong cognitive science viewpoint: we build models of how people think, perceive, and behave, and then put those models on our robots. I will describe our research paradigm and show several demonstrations of our working robots (through video). | | | | | |
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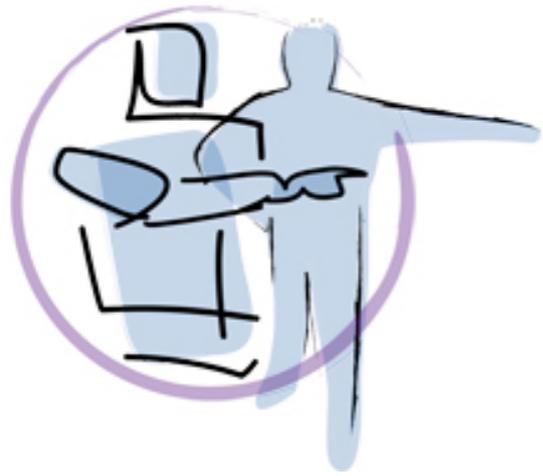
Quick Background

- As a cognitive scientist, I am interested in building computational cognitive models of people
- Our models are process descriptions of how people think, reason, perceive, etc. They express behavior through a series of human representations and strategies
- Building models of people facilitates human-robot interaction: People have expectations on how to interact with other people



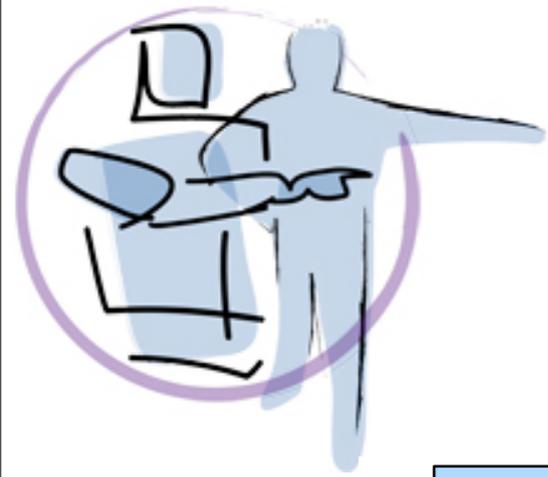
Online vs. Offline Embodied Cognition

- A recent drumbeat in cognitive science is that cognition is for action (embodied cognition)
- We are building embodied models for cognitive robotics and human-robot interaction
- Online Cognition: Our perceptual and motoric processing are primarily for here and now interactions (Brooks)
- Offline Cognition: Even when decoupled from the environment, thinking is grounded in body-based mechanisms (sensing and motor control)

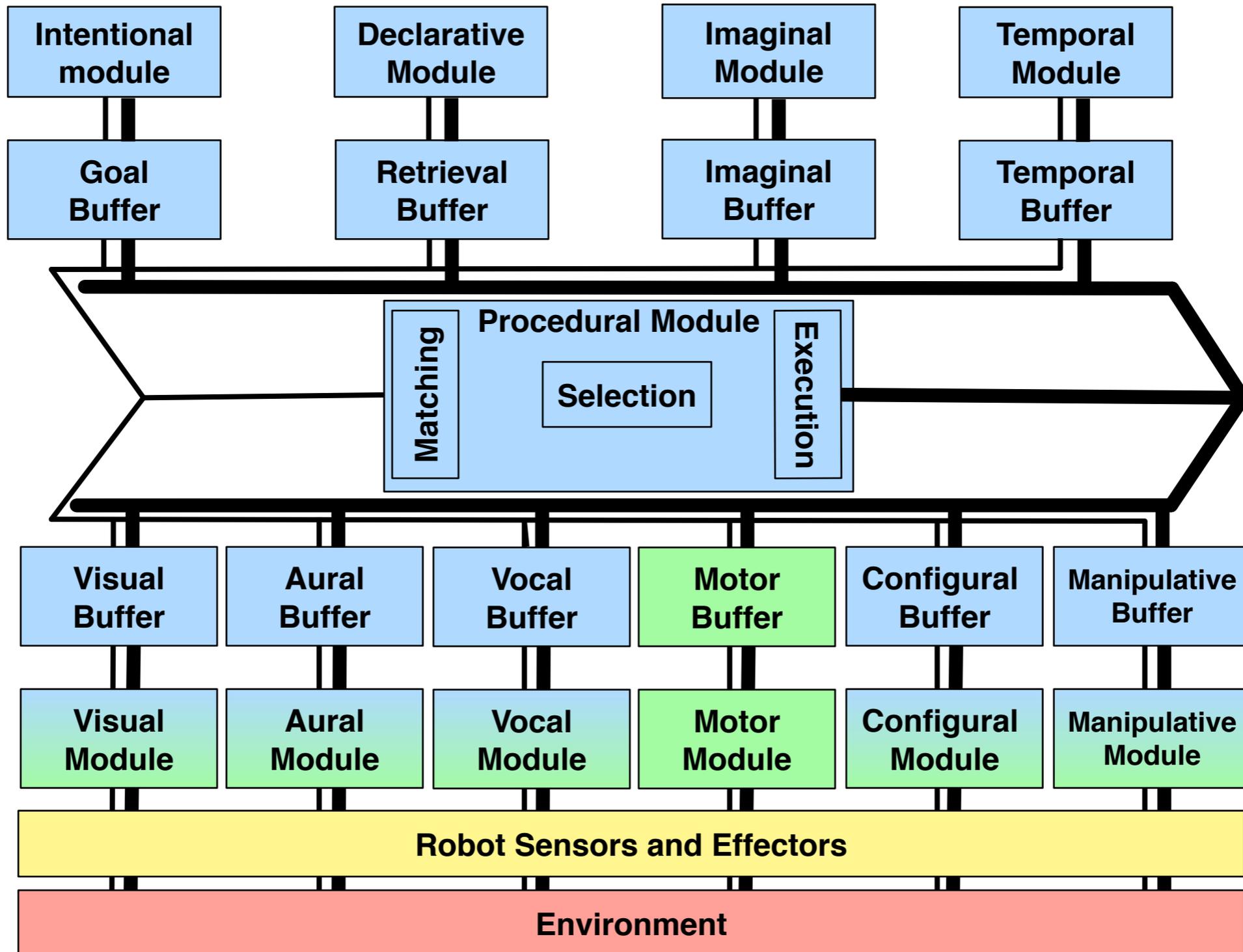


Cognitive Models

- We build models for online and offline cognition using a computational cognitive architecture (ACT-R; Anderson et al., 2004; Anderson, 2007)
- A cognitive architecture is a specification of the structure of the brain at a level of abstraction that explains how it achieves the function of the mind (Anderson, 2007)
- Hybrid symbolic and sub-symbolic system
- Connects with psychological data (experiments) and neuroscience data (fMRI)
- ACT-R models are precise process descriptions of how people perform different tasks



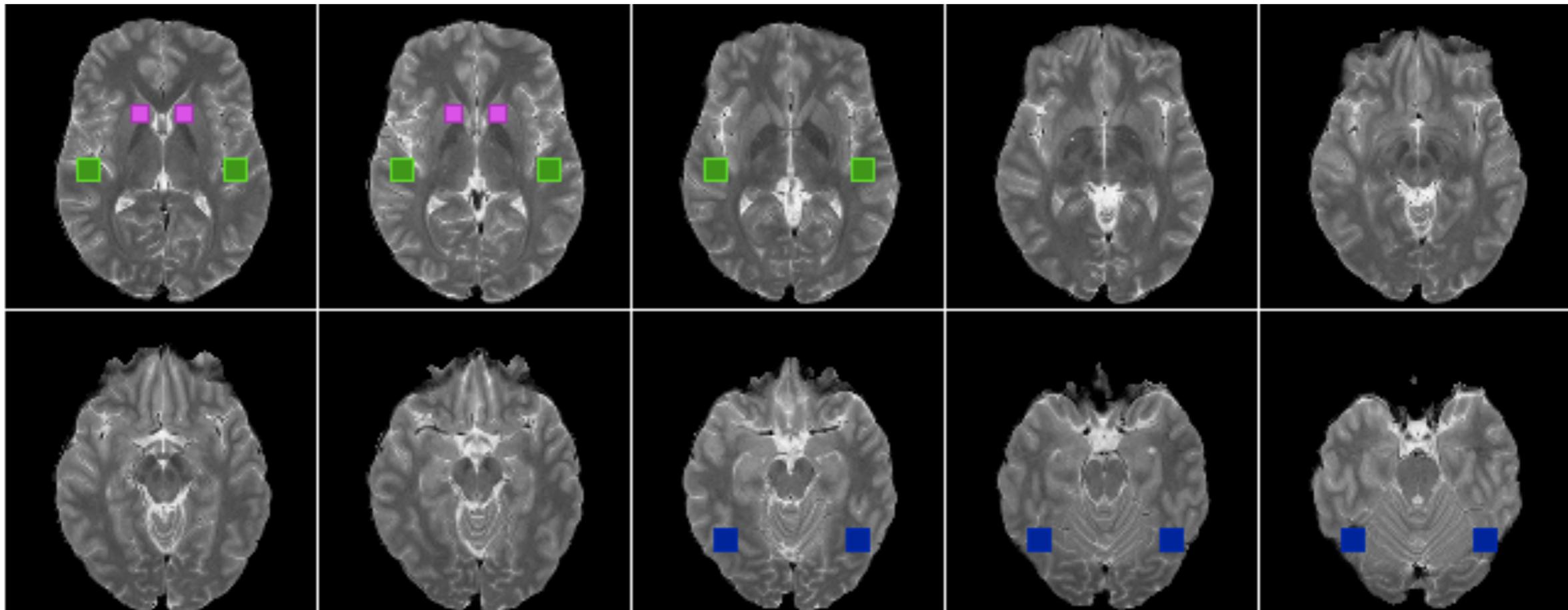
ACT-R/E (Embodied)





ACT-R can make predictions about brain regions (fMRI)

procedural
aural
visual





Embodied Cognitive Modeling

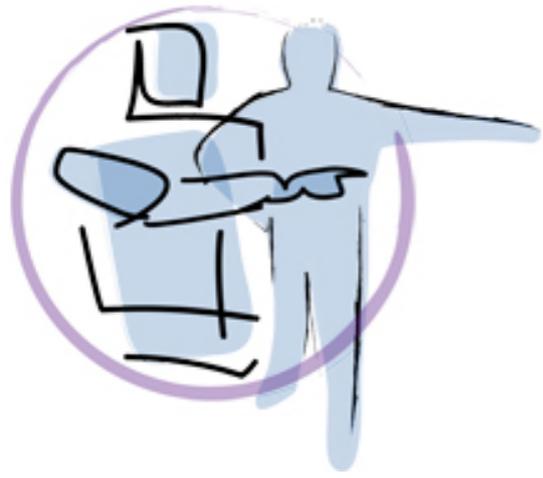
- We use an MDS robot (Trafton et al., 2010) [Mobile/Dextrous/Social] named Octavia
 - 2 color cameras (eyes)
 - ToF camera (swiss ranger in forehead)
 - Segway base
 - 52 DOFs





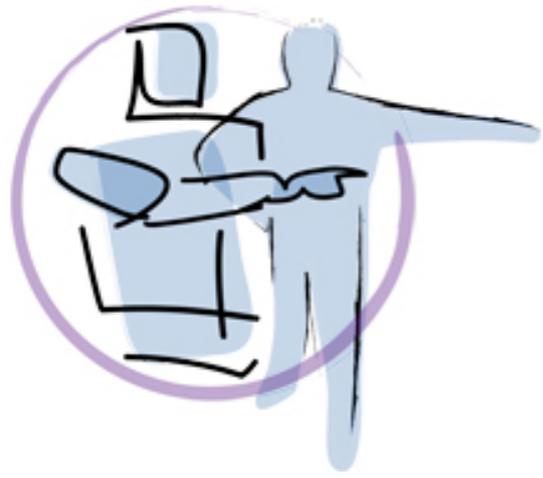
ACT-R/E

- We've hooked up ACT-R/E to our robotic sensors and effectors
- Visual input enters into ACT-R's visicon and is then processed the same way ACT-R processes any other visual stimuli
- Spatial information is also extracted through our sensors
- ACT-R/E motor commands move the robot
 - Eyes follow attention, then head, then body
- Auditory information enters into the audicon
- We have lots of good sensor / perception systems which I will not talk about here...



Gaze-Following

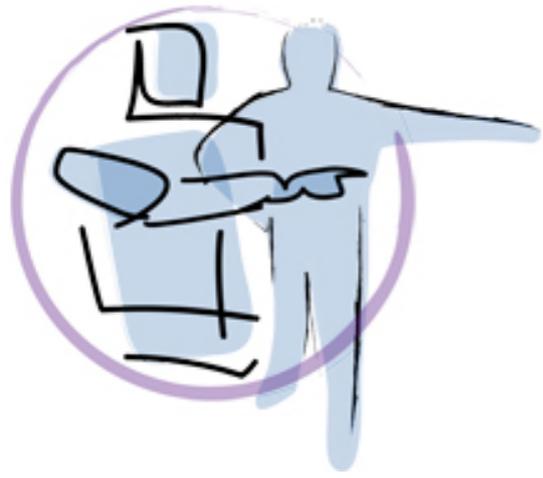
- As a concrete example of online embodied cognition, I will show a classic gaze-following experiment with young children from Corkum & Moore (1998)
 - Debate in the literature about maturation (via stages) vs. learning
- Gaze following is important for joint attention, early social development, theory of mind, later interaction



Corkum & Moore, 1998

Participants

- 3 ages:
 - 6-7 months old
 - 8-9 months old
 - 10-11 months old



Corkum & Moore, 1998 Procedure

- Child came in and sat on their parent's lap
- Parent was blindfolded so they could not give the child cues
- Experimenter got child's attention so the child would look at the experimenter



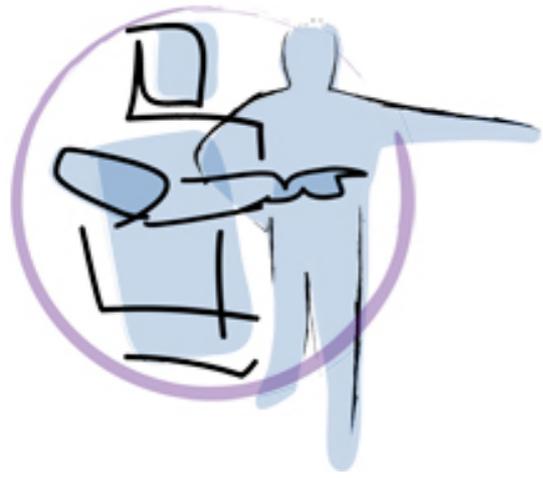
Child

Toy 1 ★

★ Toy 2

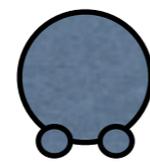


Experimenter



Corkum & Moore, 1998 Procedure

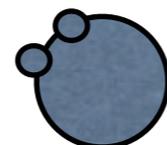
- The experimenter looked at one of two toys
- The experimenter did not point or talk to the child while looking at the toy



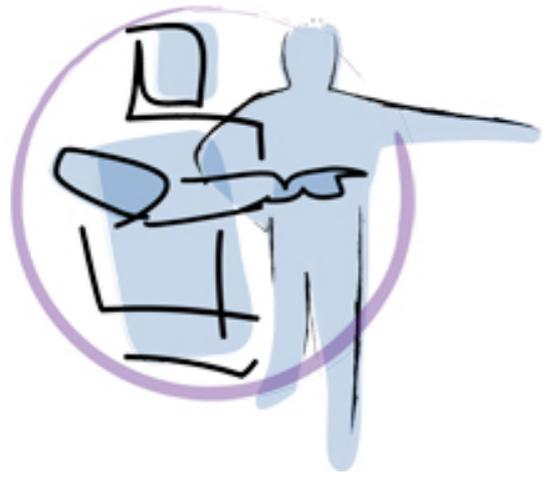
Child

Toy 1 ★

★ Toy 2



Experimenter



Corkum & Moore, 1998 Procedure

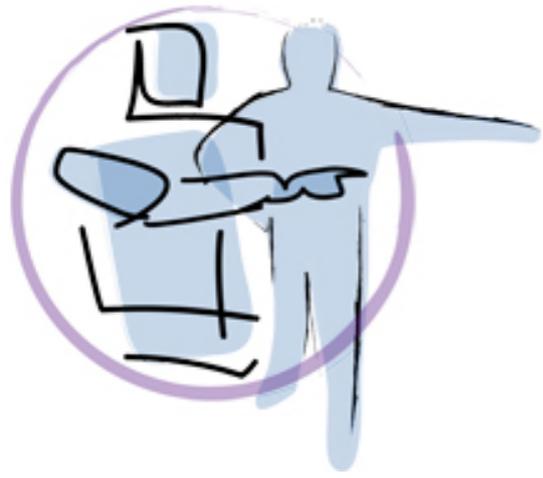
- If the child looked at the same toy the experimenter did, it was coded as a “joint gaze”



Toy 1 ★

★ Toy 2





Corkum & Moore, 1998 Procedure

- If the child looked at the toy the experimenter was not looking at, it was coded as a “non-target gaze”
- Navel gazing and non-looks were not coded



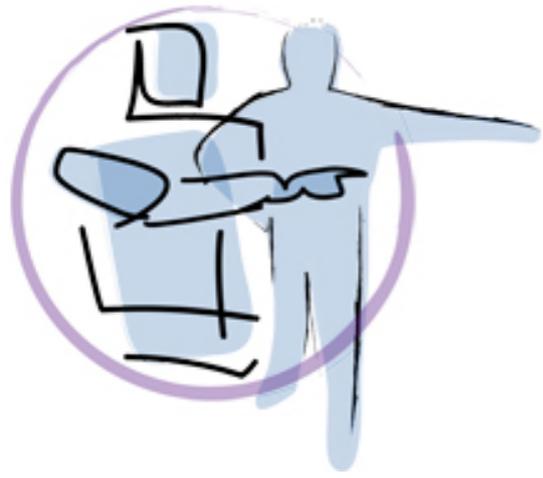
Child

Toy 1 ★

★ Toy 2



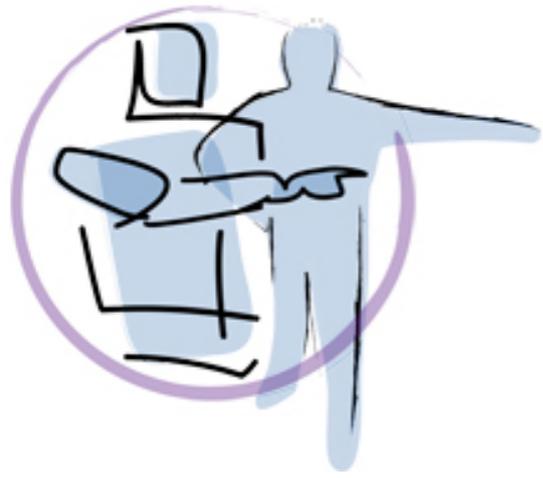
Experimenter



Corkum & Moore, 1998 Procedure

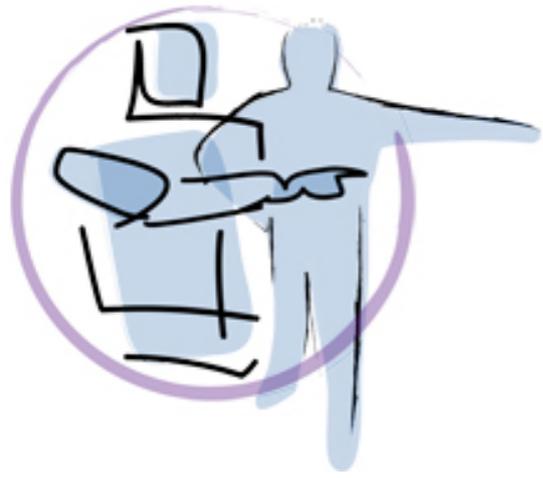
- There were three phases to the experiment
- For all trials, experimenter looked at a toy

| Phase | # Trials | Description |
|--------------|----------------------|---|
| Baseline | 4 (2 on each side) | Toy did not light up |
| Shaping | 4 (2 on each side) | Regardless of what infant did, toy lit up and rotated when experimenter gazed at it |
| Testing | 20 (10 on each side) | Toy lit up and rotated only if both experimenter and infant gazed at it |



Corkum & Moore, 1998 Scoring

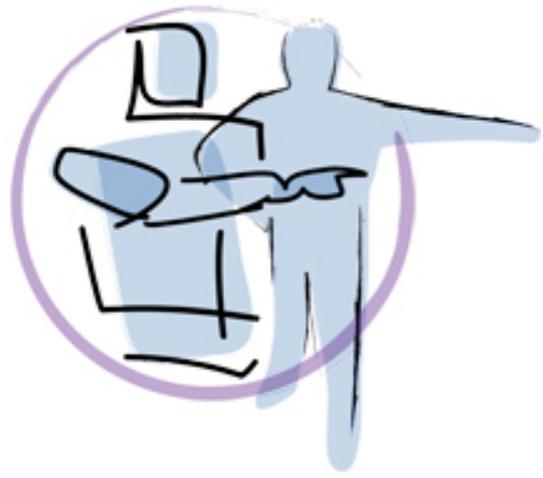
- Accuracy was scored as $\frac{\# \text{ Joint Gazes}}{\text{Total \# Gazes}}$
- Perfect gaze-following = 100%
- Random gaze-following = 50%
- Perfect anti-gaze-following = 0% (should not happen)



Corkum & Moore, 1998

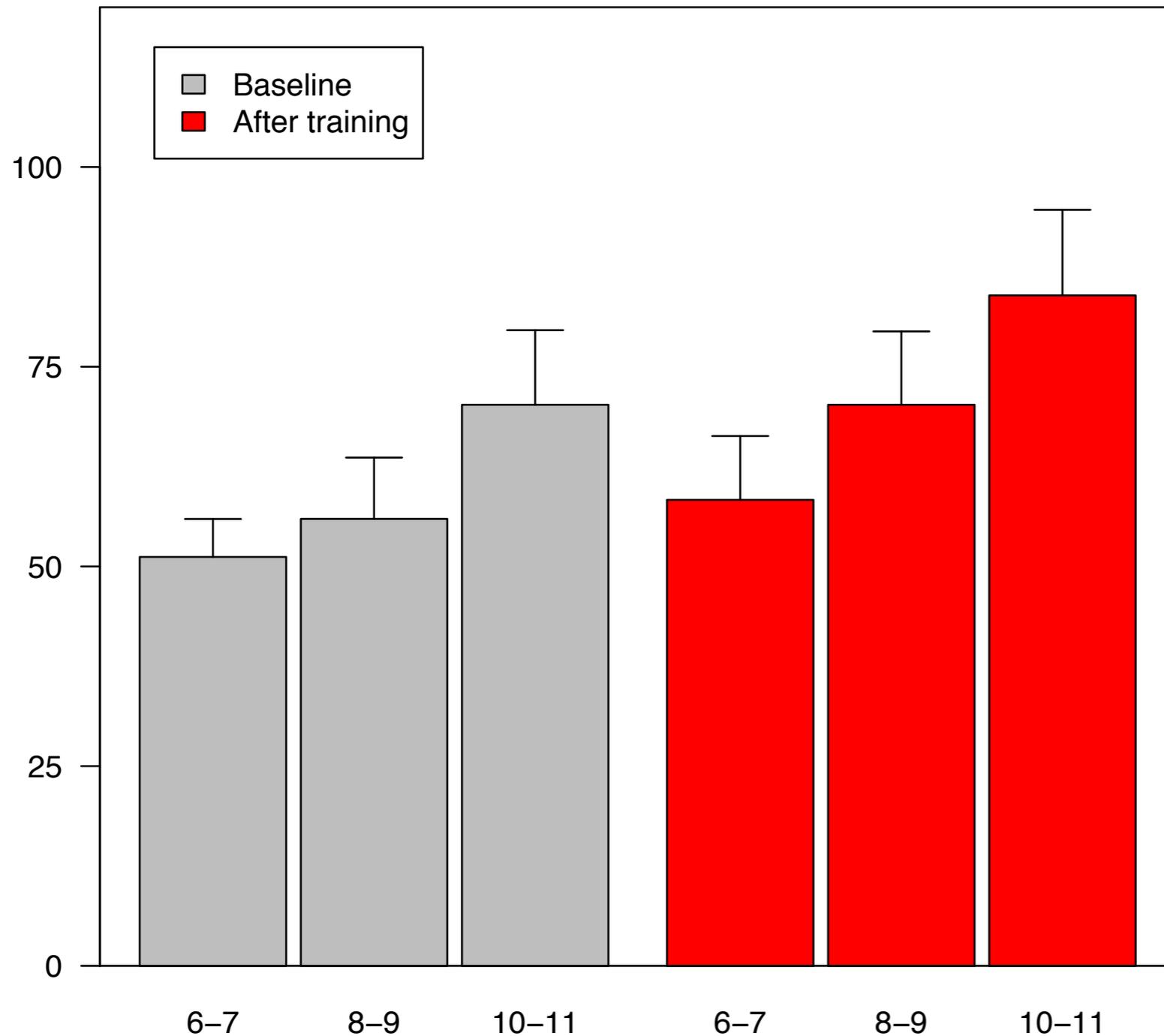
Research Questions

- At what age does gaze-following naturally occur?
- Can children who did not show gaze-following at baseline learn to follow gaze?
 - This “learning to follow gaze” hypothesis was quite novel when it came out

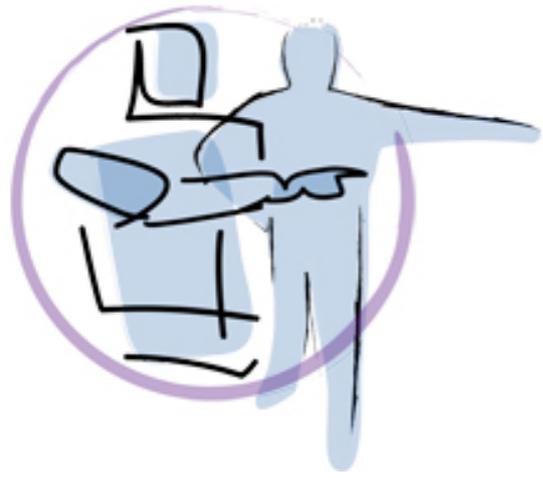


Corkum & Moore, 1998 Results

Gaze-Following
From Corkum and Moore (1998)

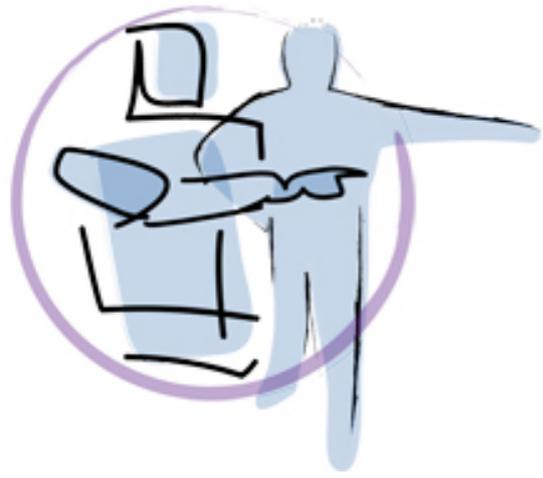


- At Baseline:
 - 6-7 m.o.: Chance
 - 8-9 m.o.: Chance
 - 10-11 m.o.: **Joint!**
- After Training:
 - 6-7 m.o.: Chance
 - 8-9 m.o.: **Joint!**
 - 10-11 m.o.: **Joint!**



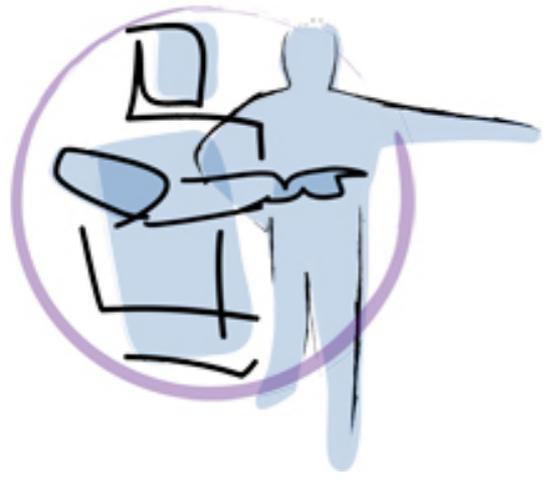
Corkum & Moore

- Corkum & Moore showed that gaze-following can be increased during a lab setting but:
 - They just showed that children could learn how to follow gaze, not the mechanism of learning (e.g., what type of learning)
 - Were 10-11 m old children at the right “stage?” Or was it continuous learning?
 - They did not show that the experimental trial learning was what happened during ‘normal’ development
 - There are no process descriptions of this finding
- So we built an ACT-R/E model (details light here)



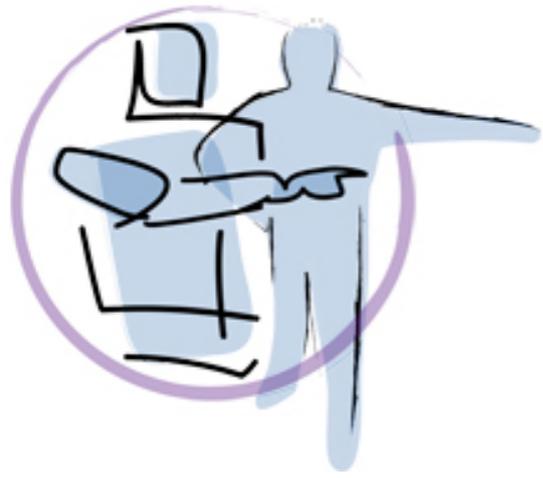
ACT-R/E Model of Gaze-Following

- When the model is young, it looks around the world for interesting objects
- When it habituates to an object, it looks for another interesting object
- When it looks at the same object a person does, it gets a small reward, which is propagated back in time
 - In “real-life” the object of gaze is relevant to both parent and child (Deak et al., 2008) [reward is tied to relevance]
- As model ages, it gradually learns to select the right spatial information and follow gaze



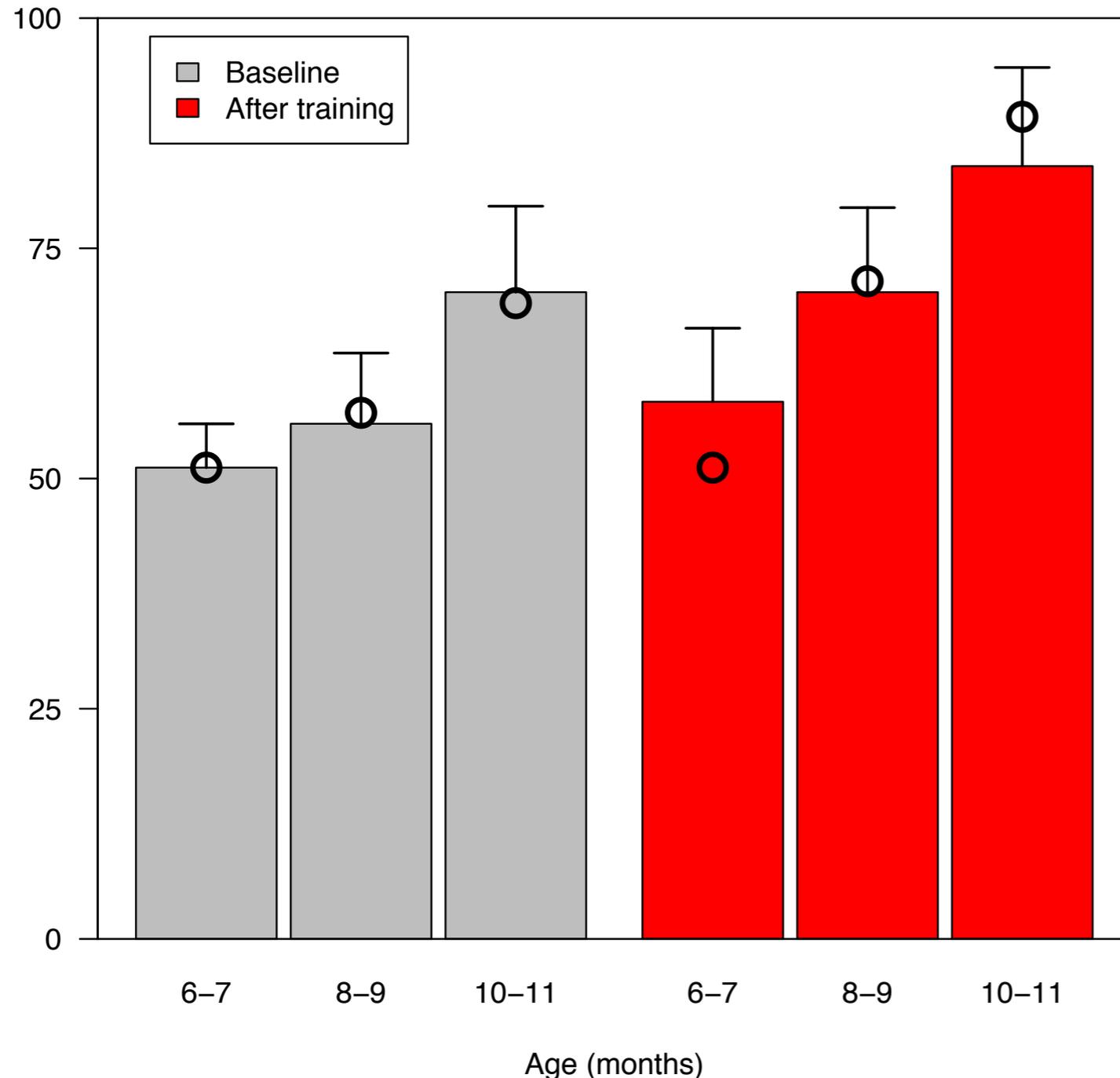
ACT-R/E Model of Gaze-Following

- Age is modeled by providing the model with opportunities to follow gaze
- The older the model, the more gaze-following experience it has had
- After aging each model, we ran it through the exact same experimental procedure that Corkum & Moore (1998) used
- We collected accuracy at each age group for the experiment and compared it to the empirical data



ACT-R/E Model of Gaze-Following

Gaze-Following
From Corkum and Moore (1998)



- Dots are model fits
- $R^2 = .95$
- $\text{RMSE} = .3$
- All model points are within 95% CI
- Process support for learning hypothesis, not stage maturation

Trafton, Harrison, Fransen, &
Bugajska, 2009



Embodied Cognitive Modeling

- We also ran our model up to 10-11 m on our MDS robot, Octavia
- Several reasons for emphasizing embodied cognition
 - No cheating allowed: very explicit theory of spatial cognition, learning, development
 - Must deal with the complexity of the environment (dynamic, 3D, etc): the complexity of the real world tests the model, theory, framework





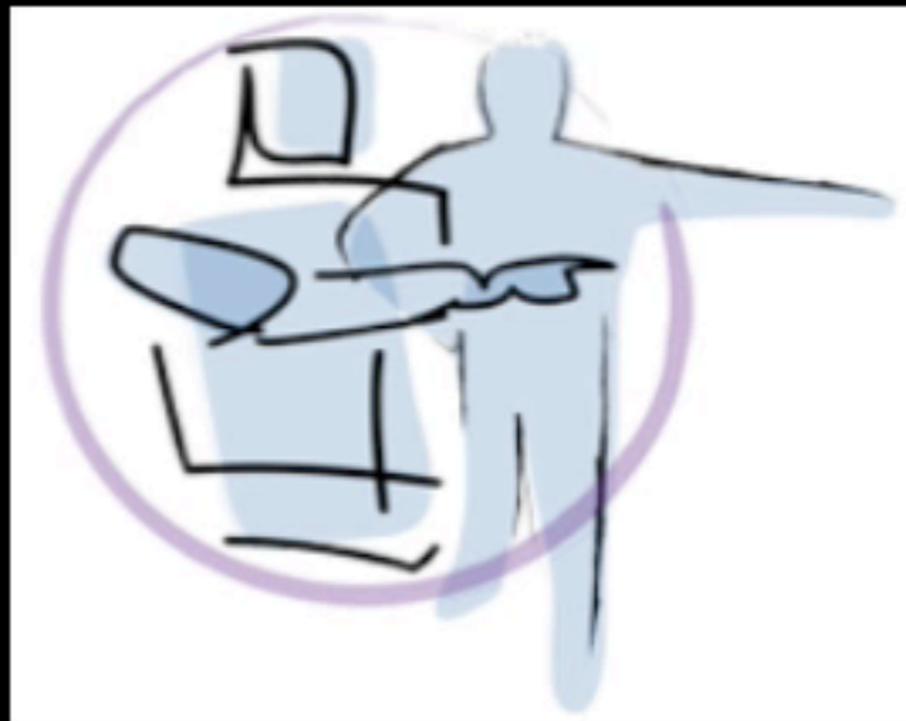
Model of 10-11 month old

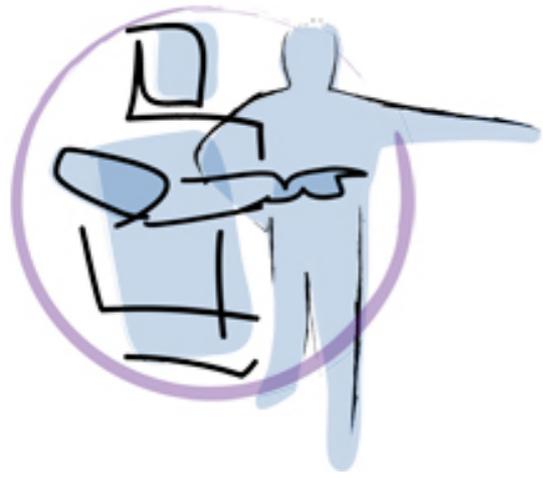




We can also use head position to infer what someone sees

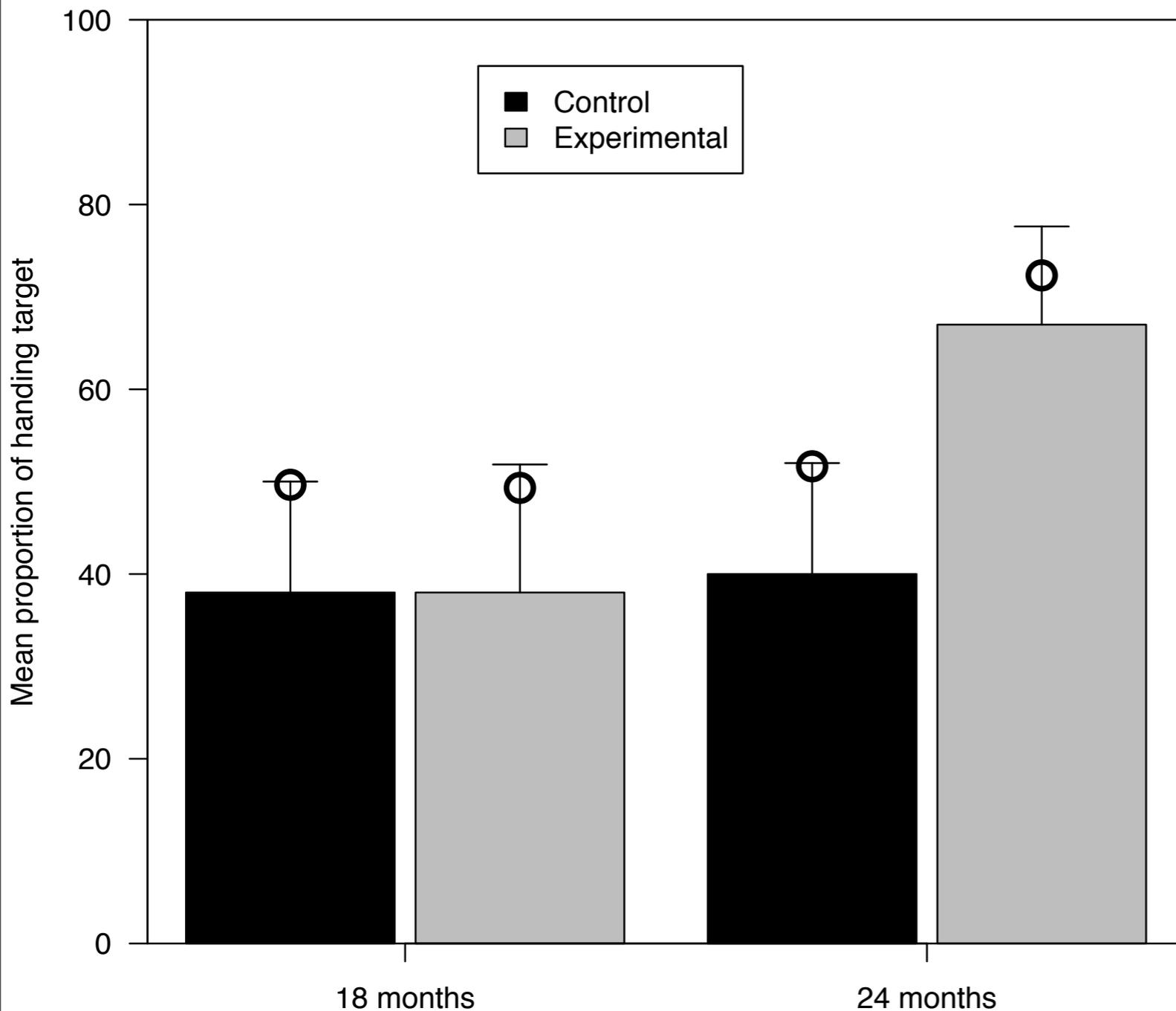
Level I Visual Perspective Taking (Based on Moll & Tomasello, 2006)



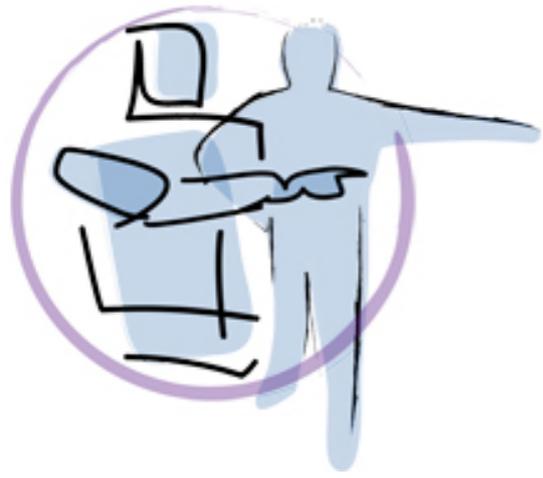


ACT-R/E Model of Level 1 Visual PT

Accuracy of Level 1 Perspective-Taking
From Moll and Tomasello (2006)

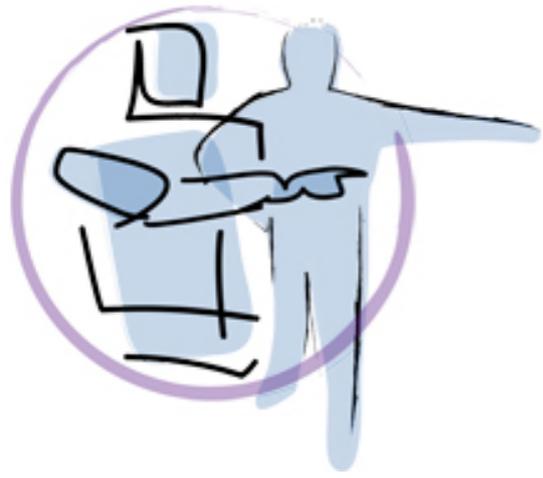


- Dots are model fits
- $R^2 = .99$
- $\text{RMSE} = 10.4$
- All model points are within 95% CI
- Trafton, Harrison, Fransen, & Bugajska, 2009



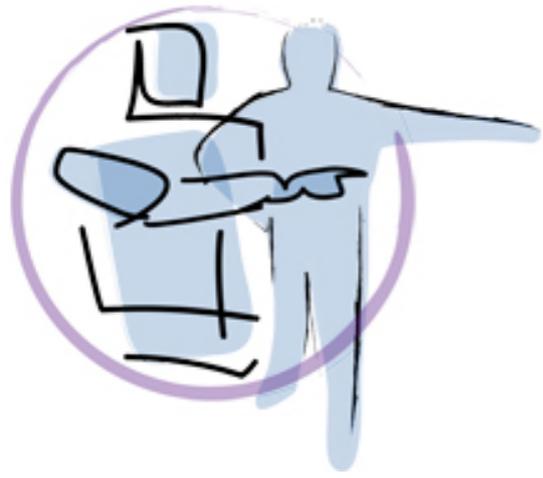
Beyond Gaze-Following

- Gaze following is a strong example of online embodied cognition: very much “here and now”
- What about another classic cognitive phenomena that uses offline embodied cognition: Theory of Mind?



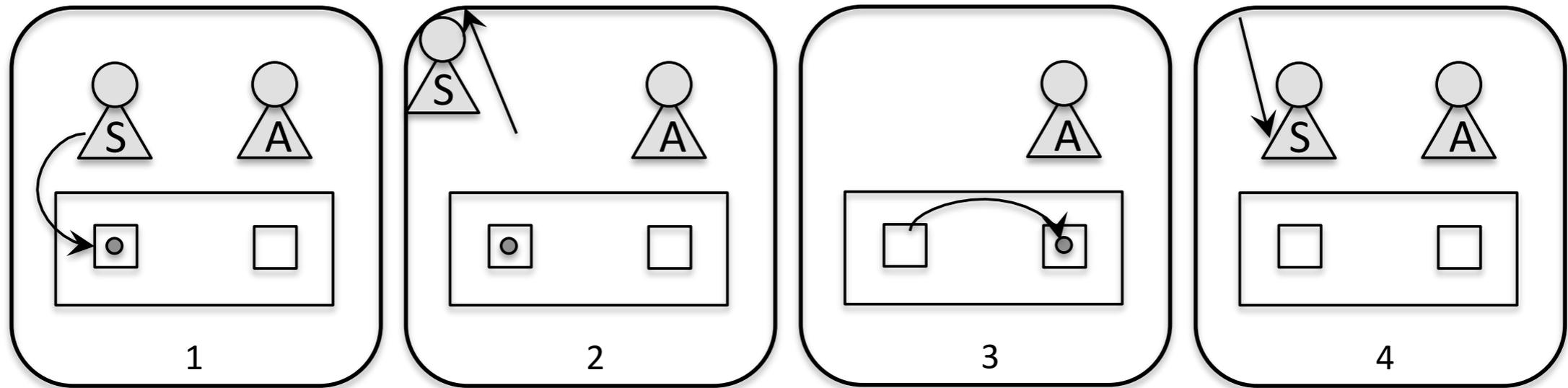
Theory of Mind

- Recognition that others can have beliefs, desires and intentions different from your own
 - Typically develops around age 4
- Ability to infer what the beliefs, desires and intentions of others
- Critical ability for interacting with others
- The Sally-Anne task is one of the typical tasks used to study Theory of Mind



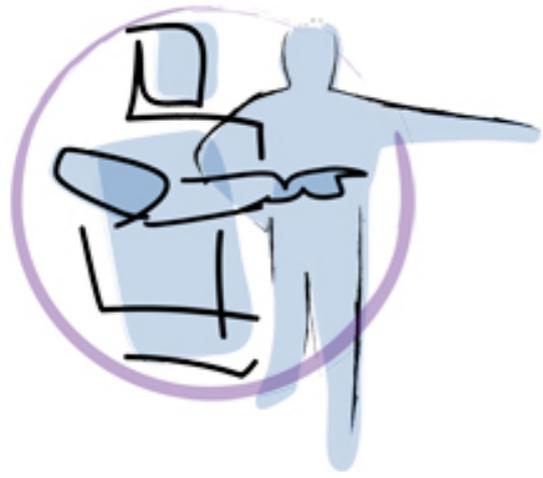
Sally-Anne Task

(Baron-Cohen et al., 1985)



A child watches while:

1. Sally puts a marble in her box
2. Sally leaves the room
3. Anne moves the marble to Anne's box
4. Sally returns to the room
5. The child is asked where Sally believes the marble is



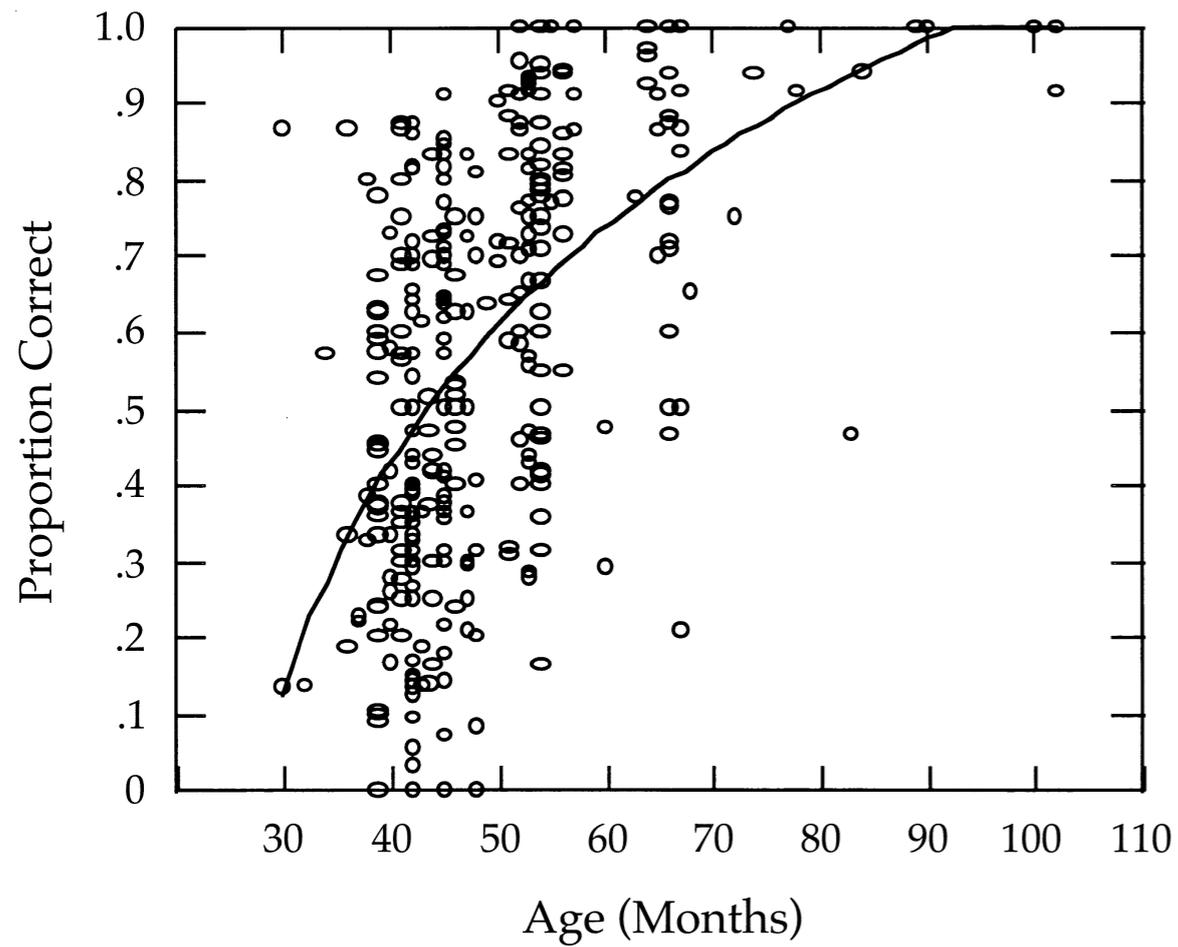
Theory of Mind

- Wellman (2001) performed a meta-analysis on the existing experiments with the Sally-Anne false-belief task (and there were very many)
 - False-beliefs are beliefs that one may have but which are not actually true in the physical world
- They built a statistical model (not a process model)
- We built a process model of the Sally-Anne task and ran it many times at different age levels
 - (Had a combination of maturation + learning)



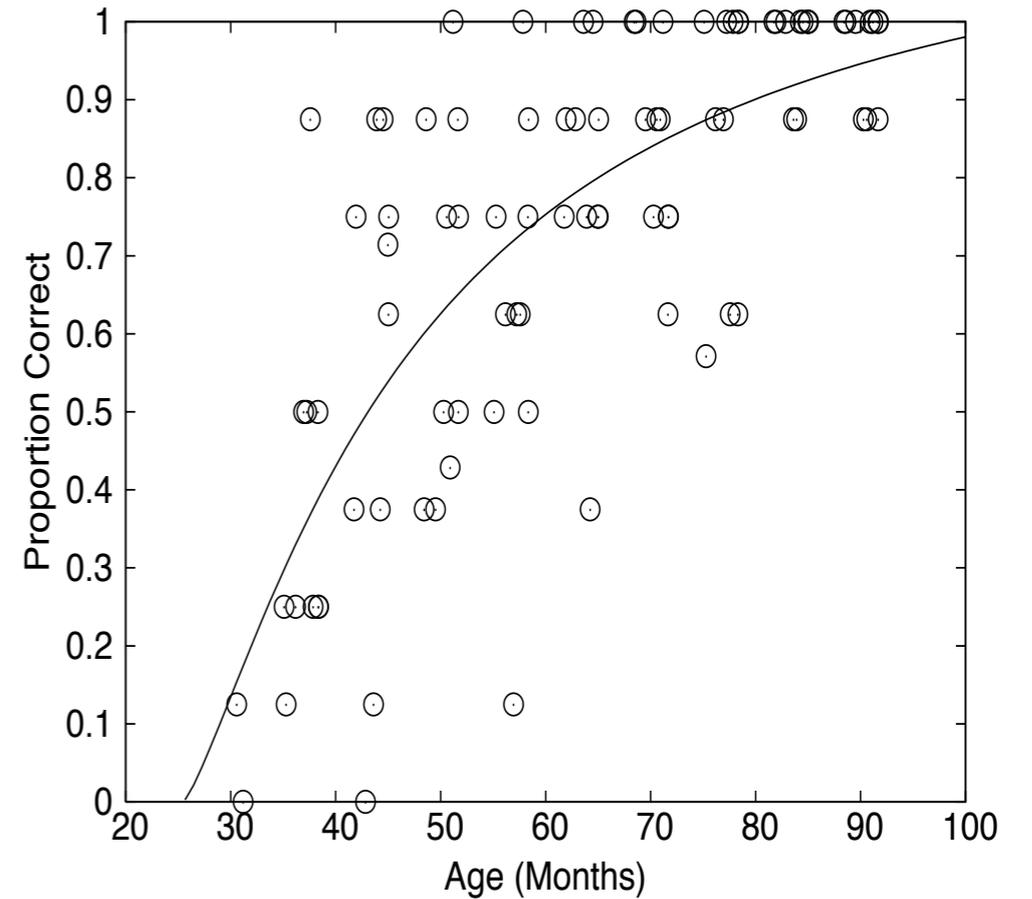
Results

Welman, 2001



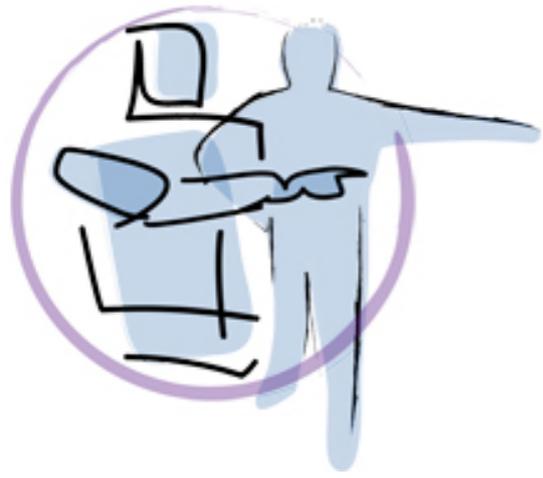
$$R^2 = 0.55$$

Our data



$$R^2 = 0.56$$

Hiatt & Trafton, 2010



Theory of Mind

- Work so far has focused on development
 - It is impossible to program every situation an autonomous agent will encounter, so we have focused on learning because people are the best, most robust learning systems we know of
 - Besides building theoretical models of people (which advances cognitive science by providing process models of how people think, learn, embodied cognition, etc.), we can show the robustness of our models by using/demonstrating them in a different context

Naval Research Laboratory

&

NCARAI

Proudly present



Embodied Cognition

- We build computational cognitive models that takes embodiment seriously
- Our models are process models of how people think
 - Our models match human data at multiple levels
- We take our models and put them on robots so they are functional and have an embodied presence and interaction
- Our approach is radically different from typical robotics architectures (and from other EC researchers)
- Our last video shows interactive behavior with a focus on computational vision and perception...

5



Thanks!

