

An ACT-R model of collaborative skill acquisition for Coop Space Fortress

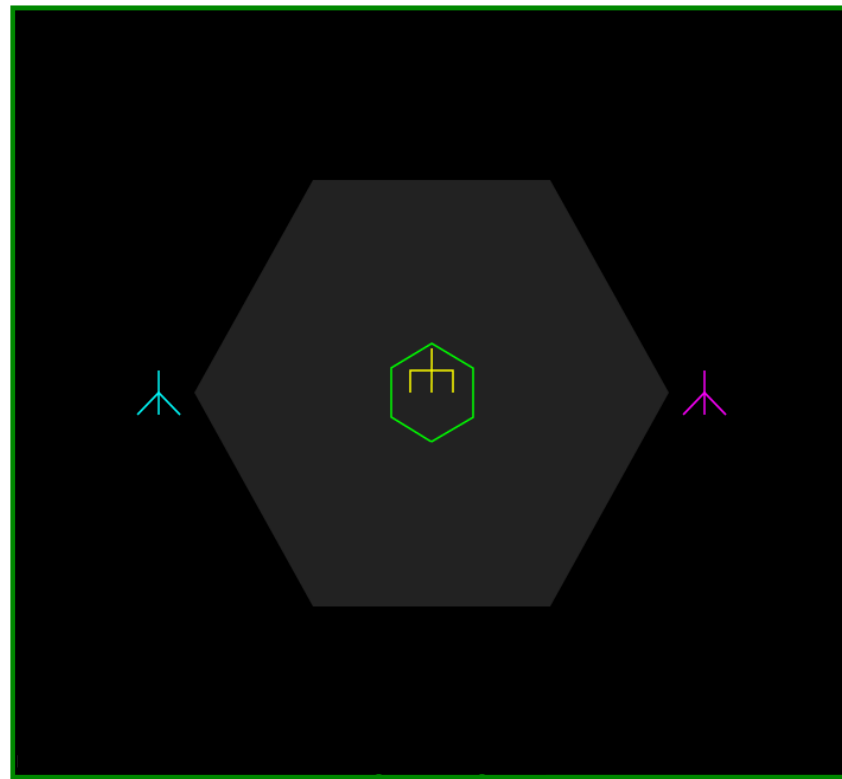
Cvetomir Dimov
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Coop Space Fortress

Game description

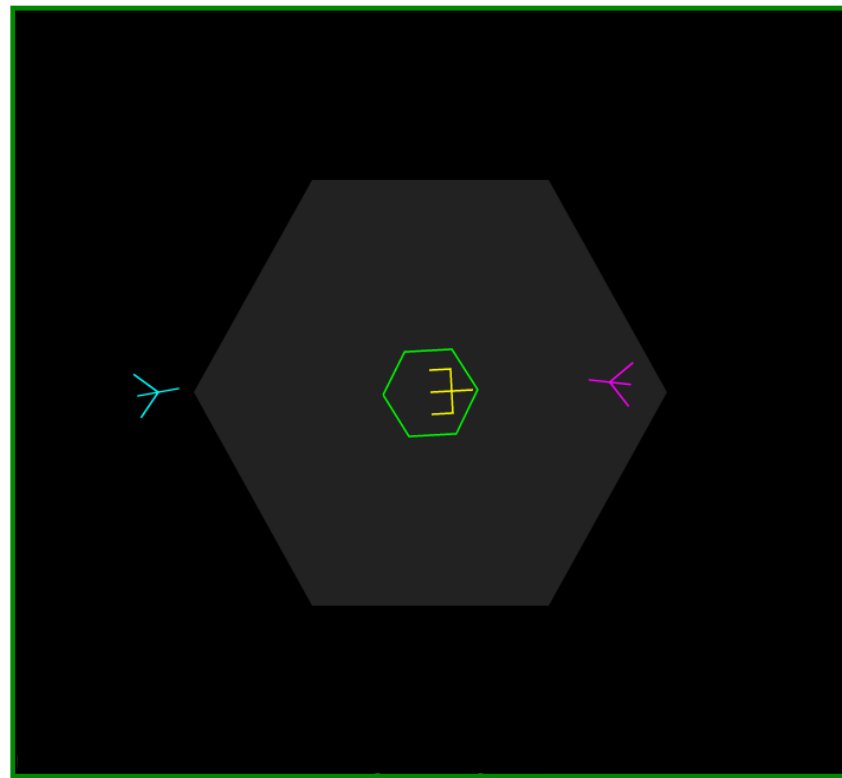
- ♦ **Goal:** kill fortress located at center of screen
- ♦ **Roles:** one player baits the fortress, the other shoots it down; bait should fly slowly
- ♦ **Incentivization:** point gains from killing fortress and losses from dying or missing shots



Coop Space Fortress

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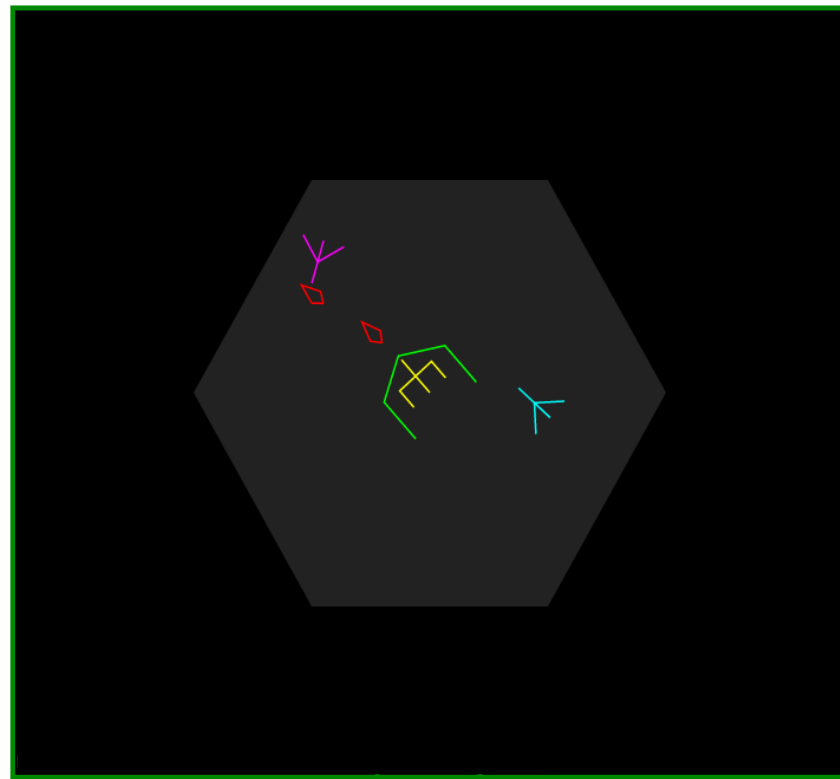
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Coop Space Fortress

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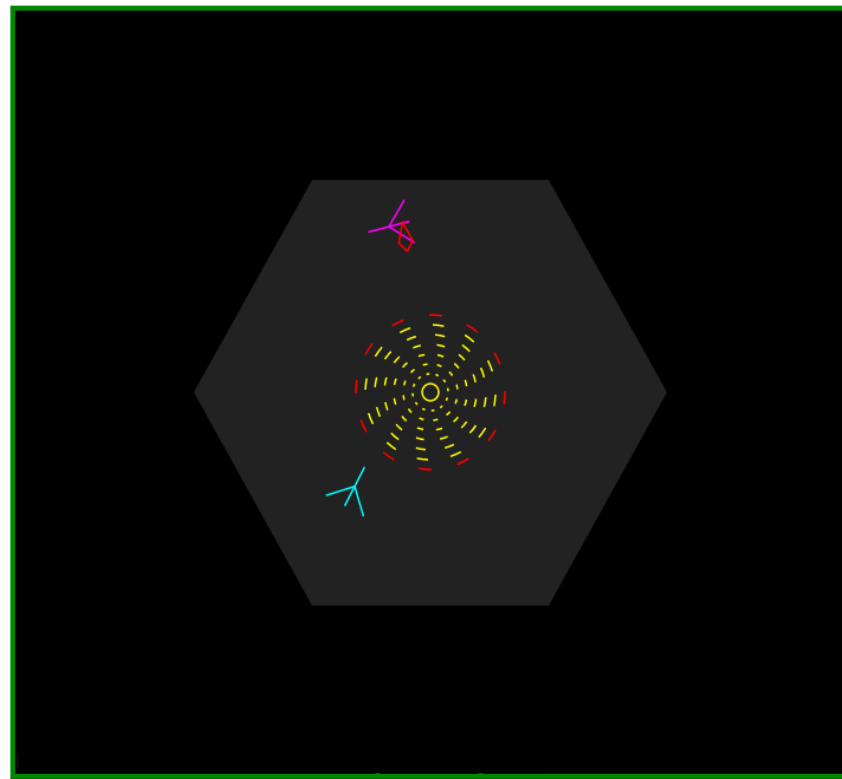
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Studies with Coop Space Fortress

Study 1

Goal: pilot

Duration: ~ 1 h (20 games x 3 minutes)

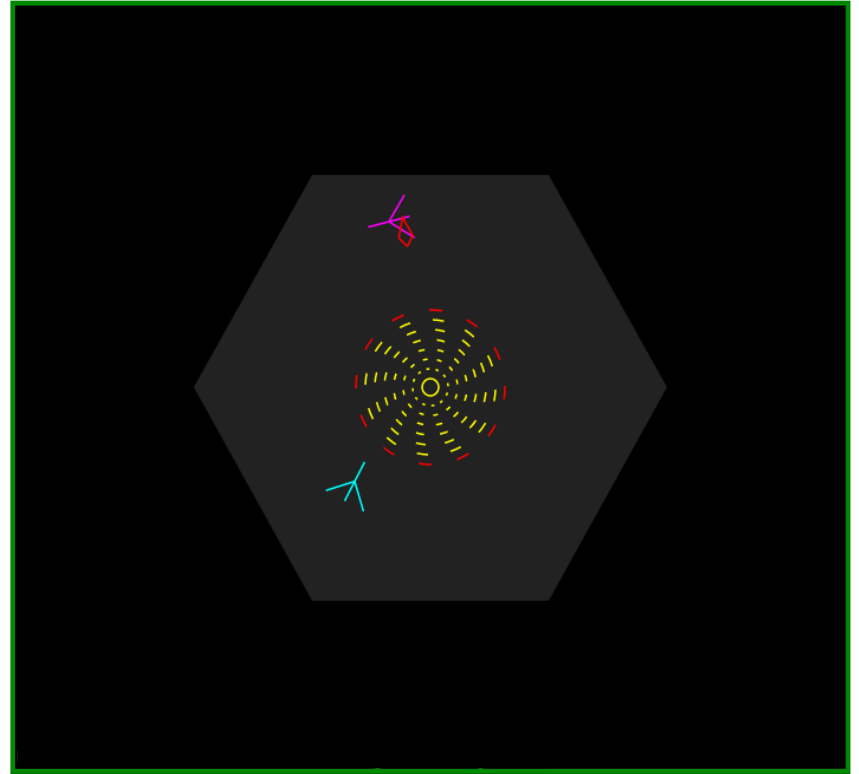
Sample: 14 teams

Study 2

Goal: transfer from Space Track

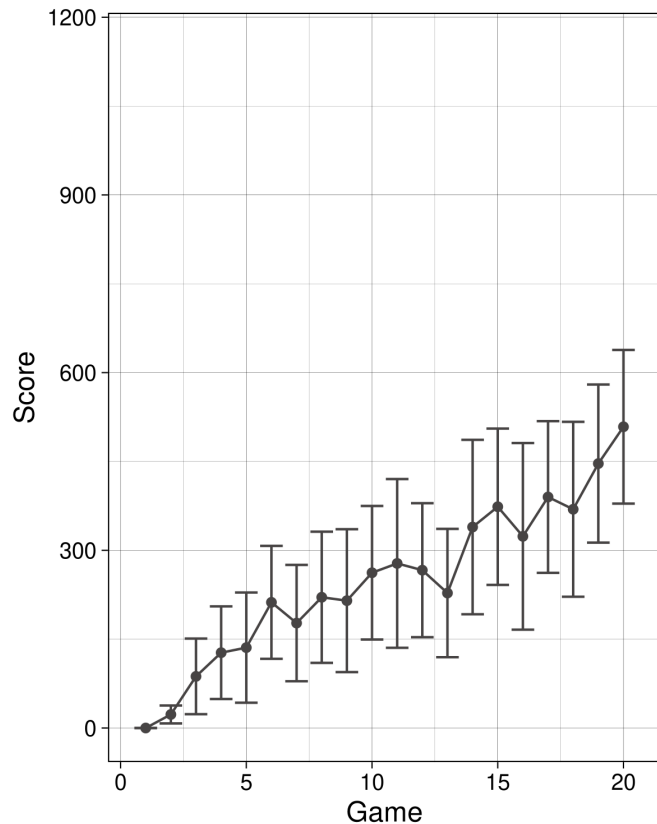
Duration: ~ 2 h (2 sessions)

Sample: 39 teams

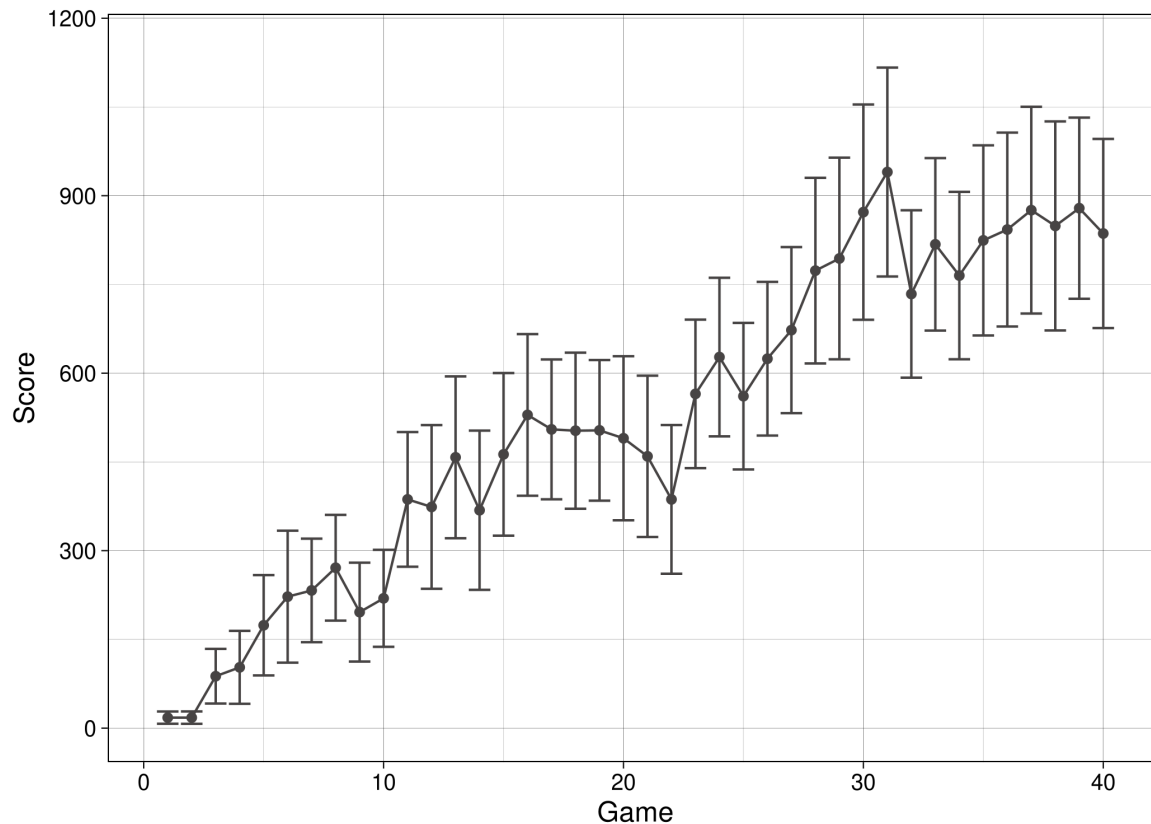


Study Results: Performance

Study 1: Score

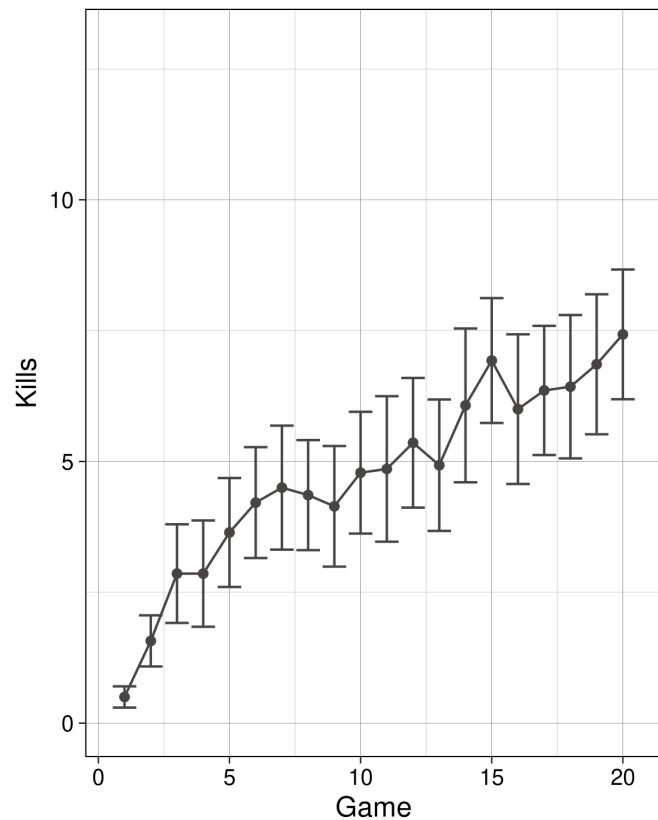


Study 2 (control condition): Score

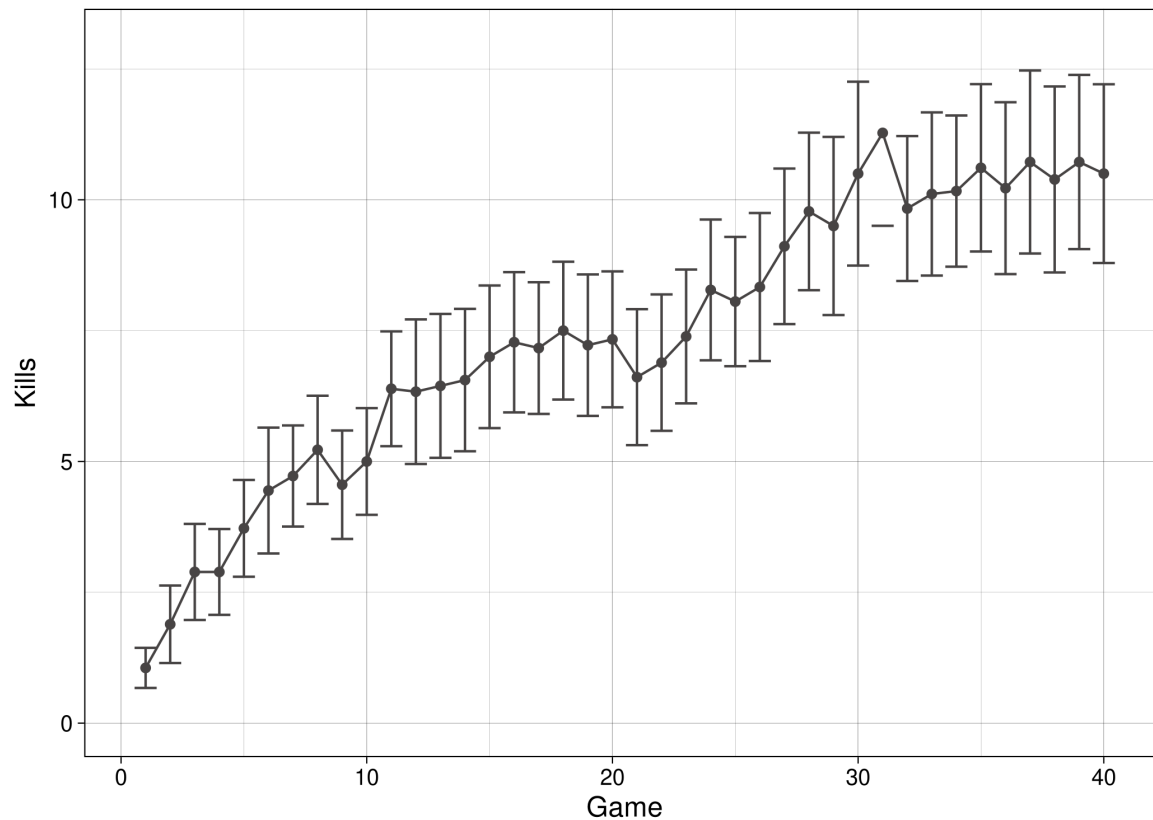


Study Results: Performance

Study 1: Kills

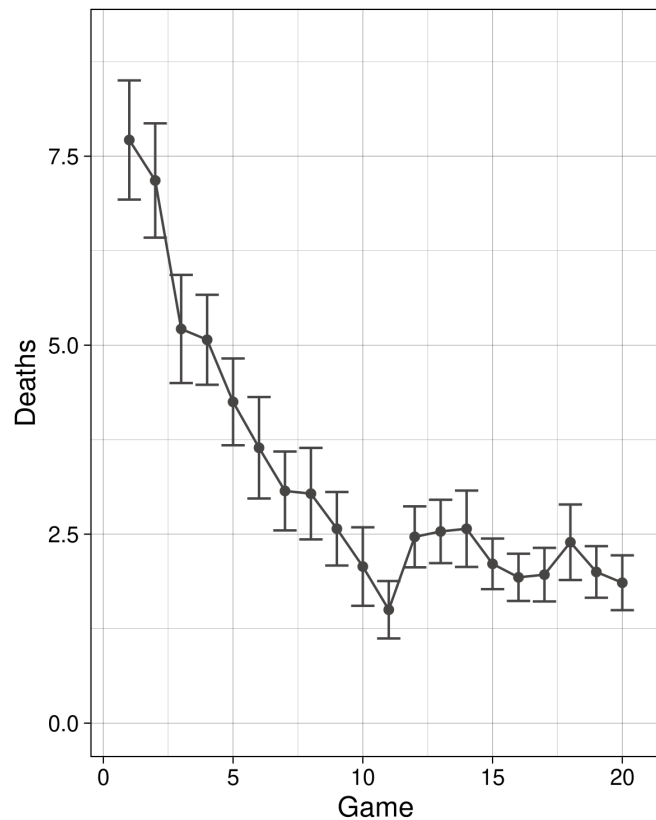


Study 2 (control condition): Kills

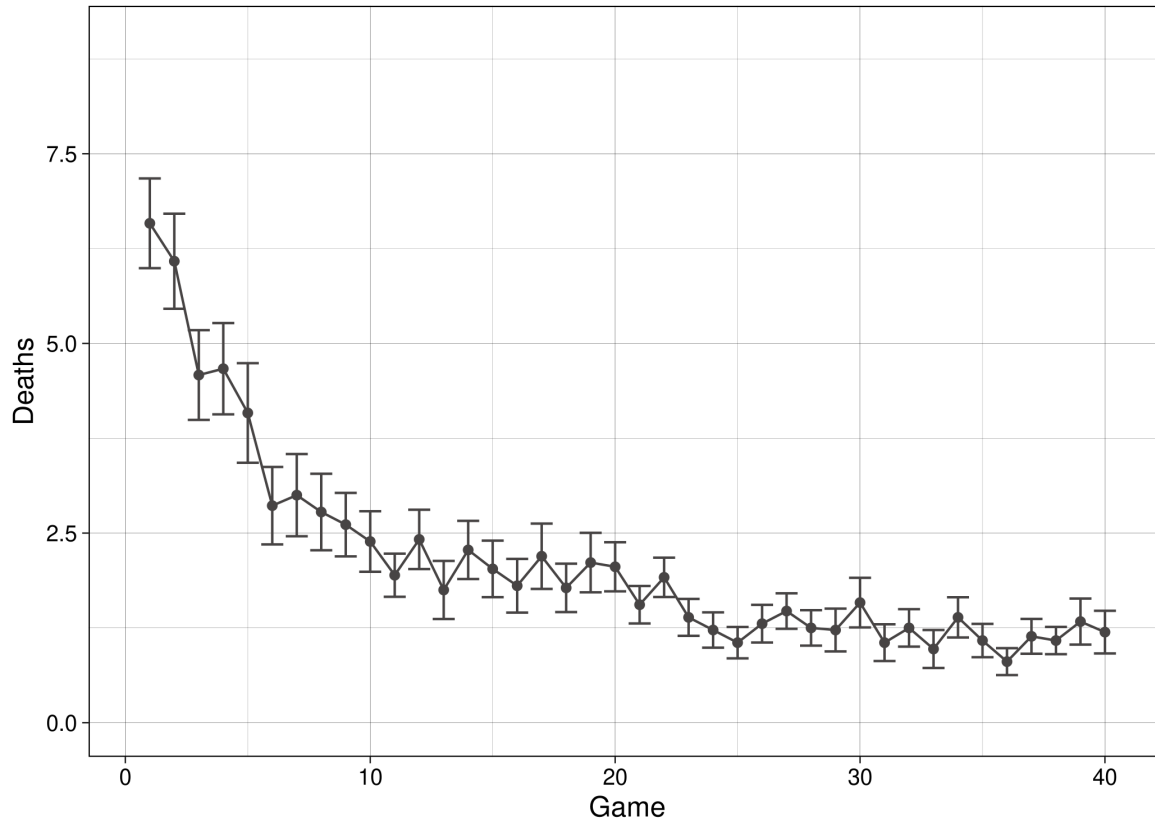


Study Results: Performance

Study 1: Deaths

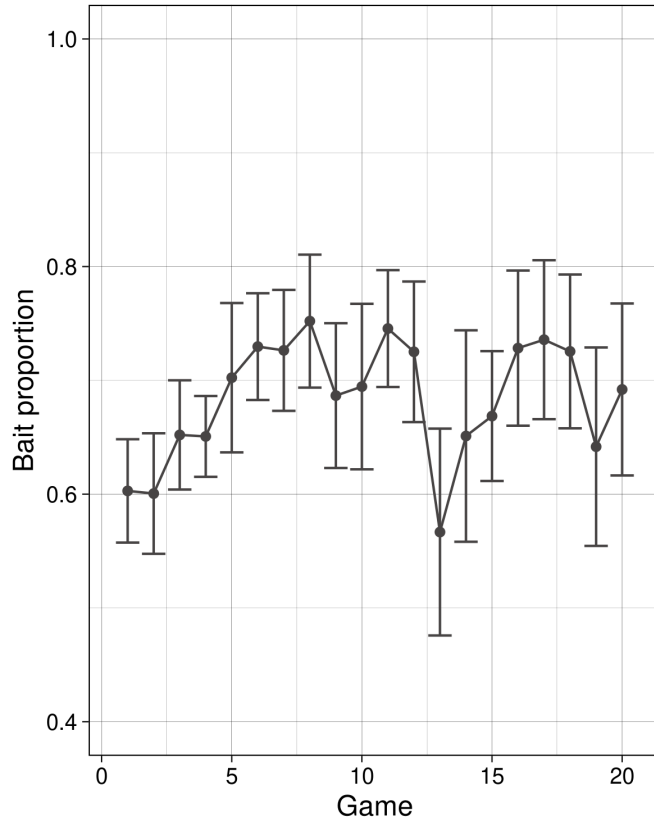


Study 2 (control condition): Deaths

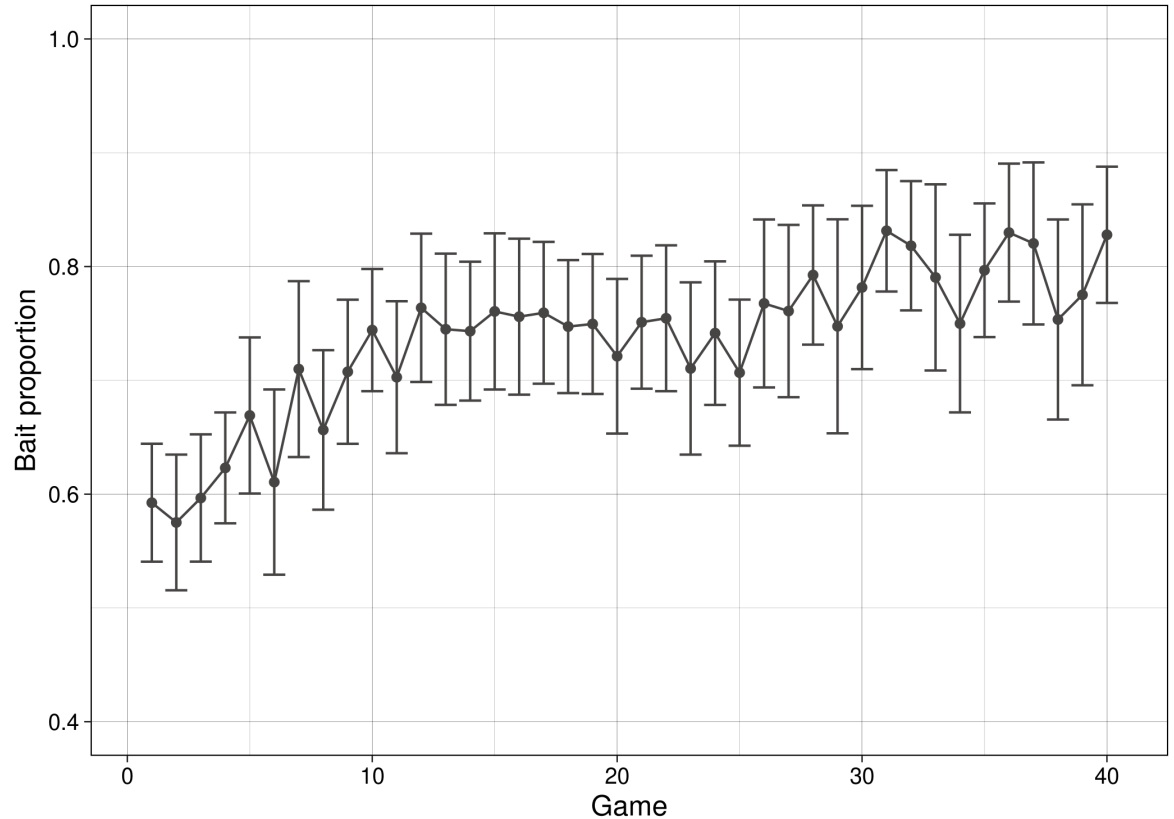


Study Results: Role Consistency

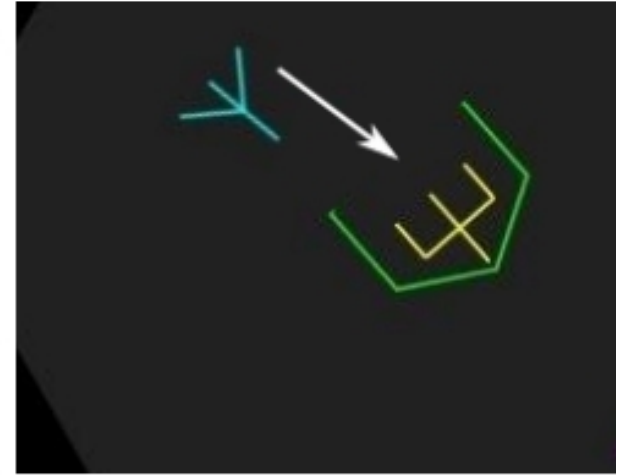
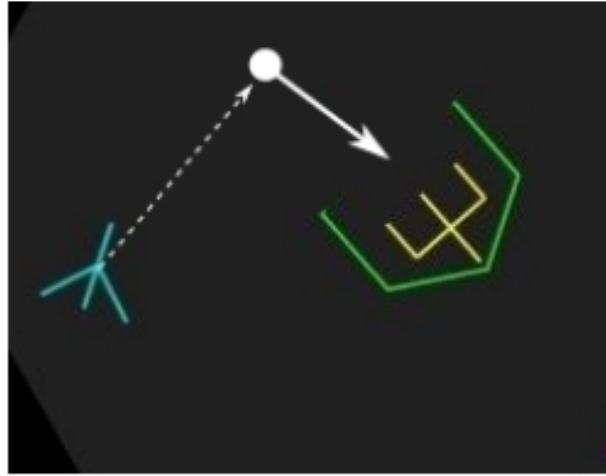
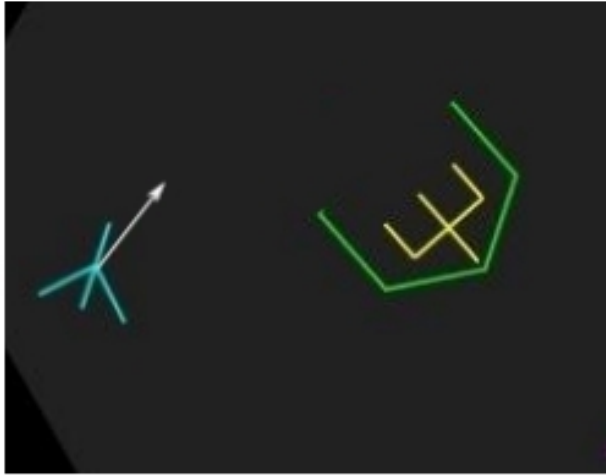
Study 1: Bait proportion



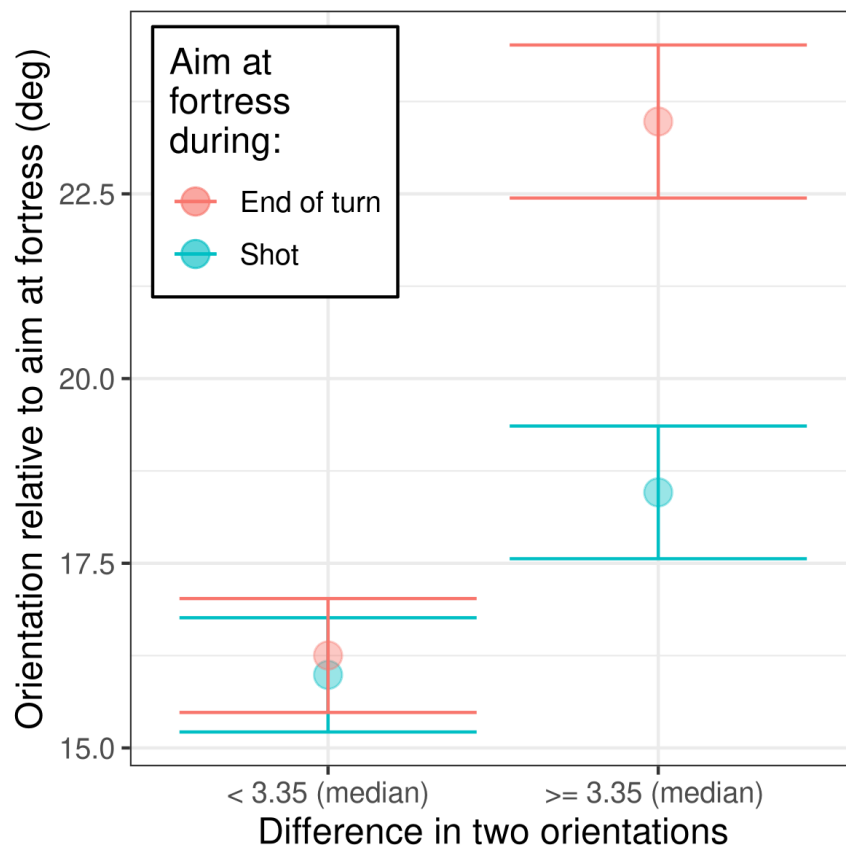
Study 2 (control condition): Bait proportion



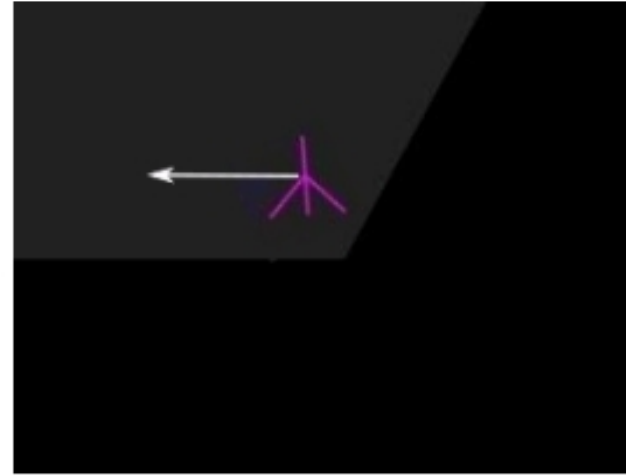
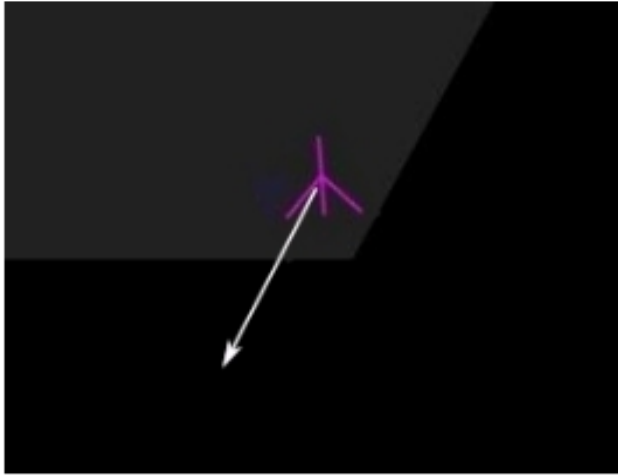
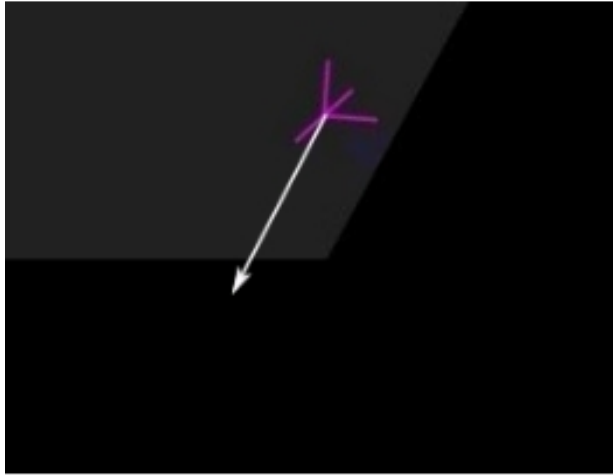
Study Results: Anticipatory Actions



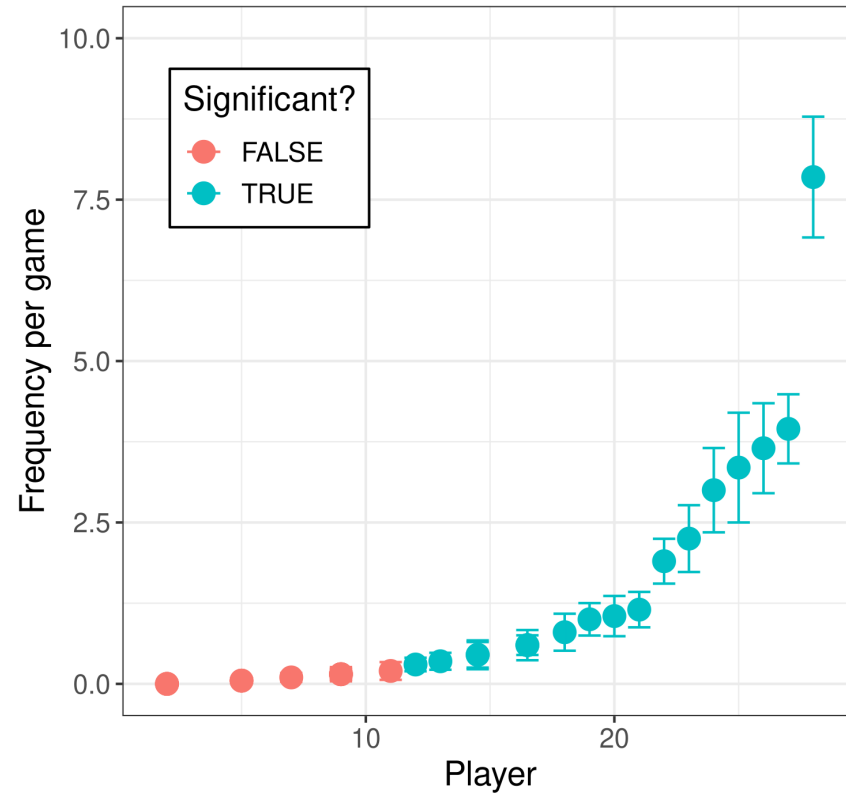
Study Results: Anticipatory Actions



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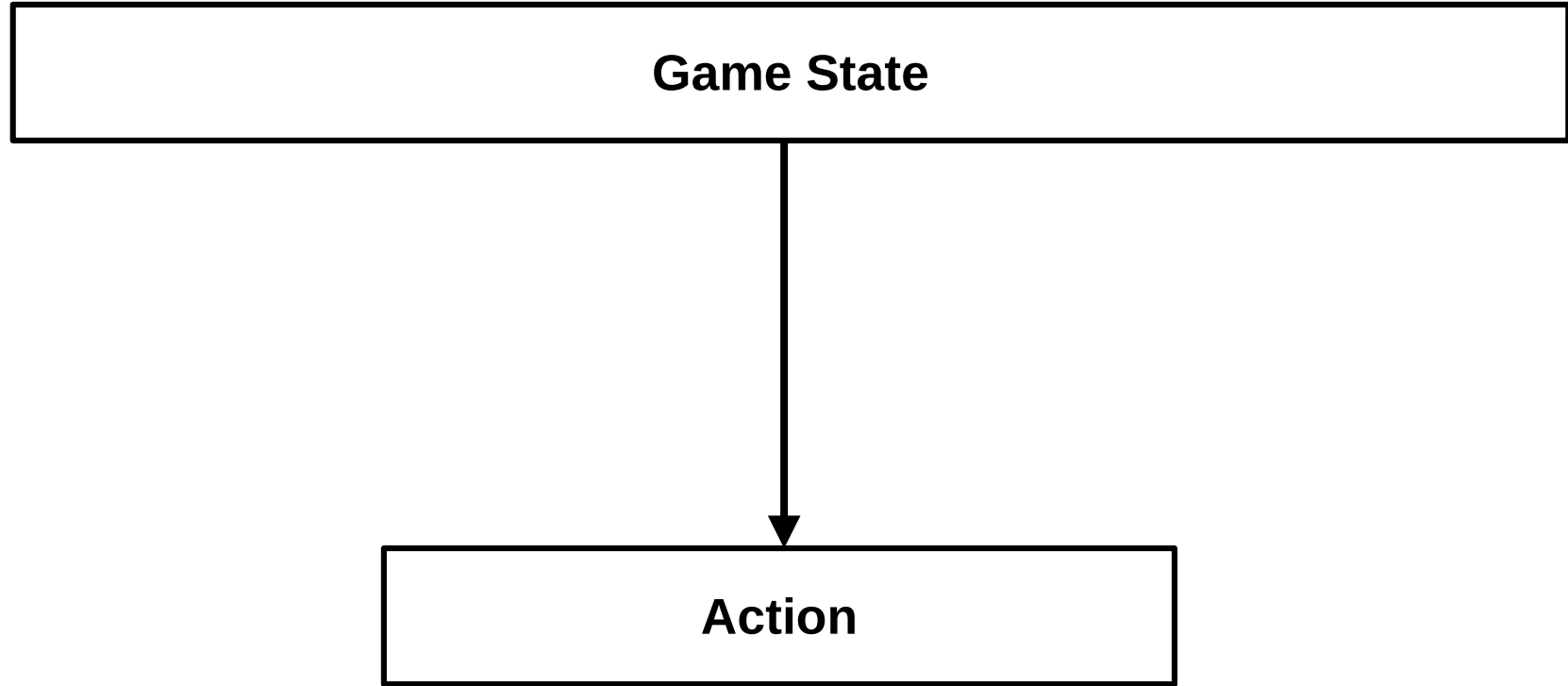
Other anticipatory actions:

- Preparing to stop when outside
- Avoiding crashing into outer border or fortress
- Bait prepares to slow down before having entered

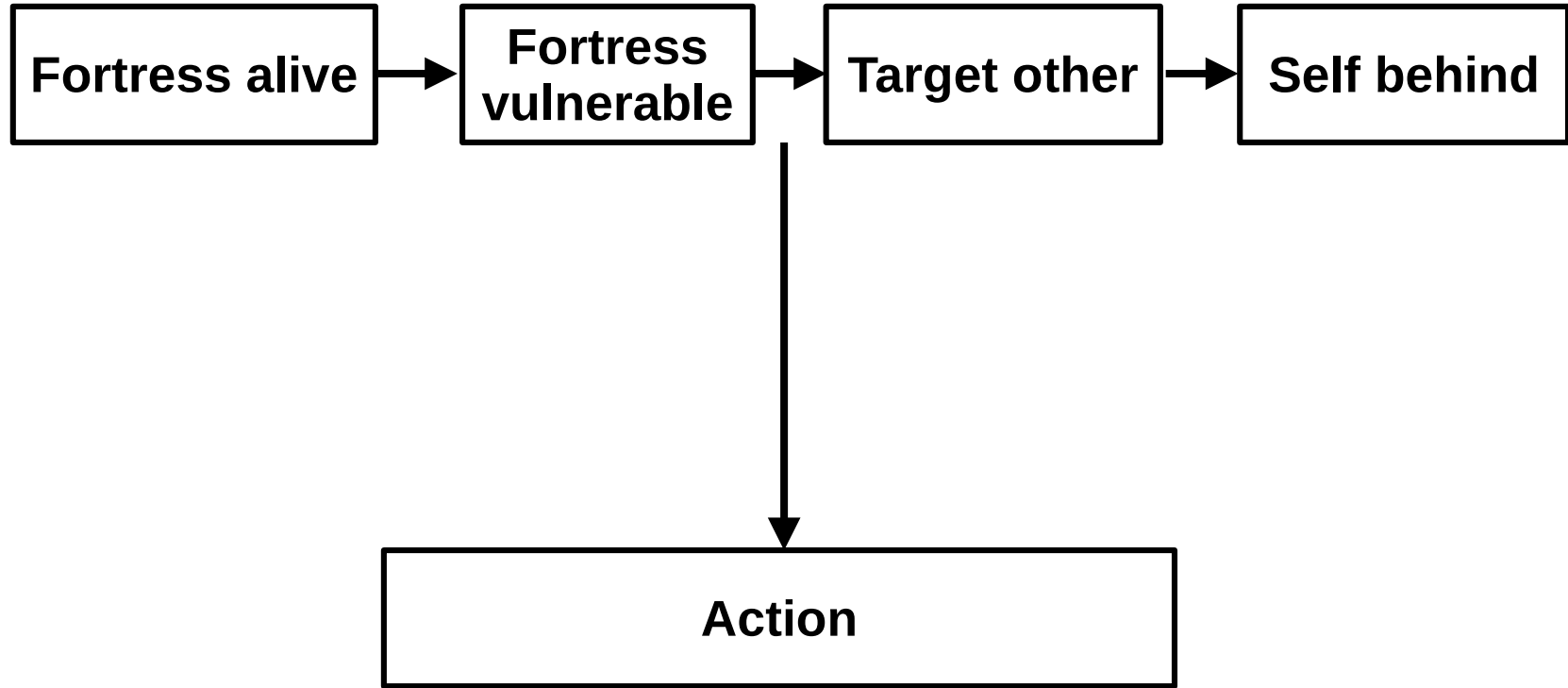
ACT-R Model: Skill Acquisition Process

1. Cognitive operators
2. Transition from declarative phase to procedural phase
3. Control tuning

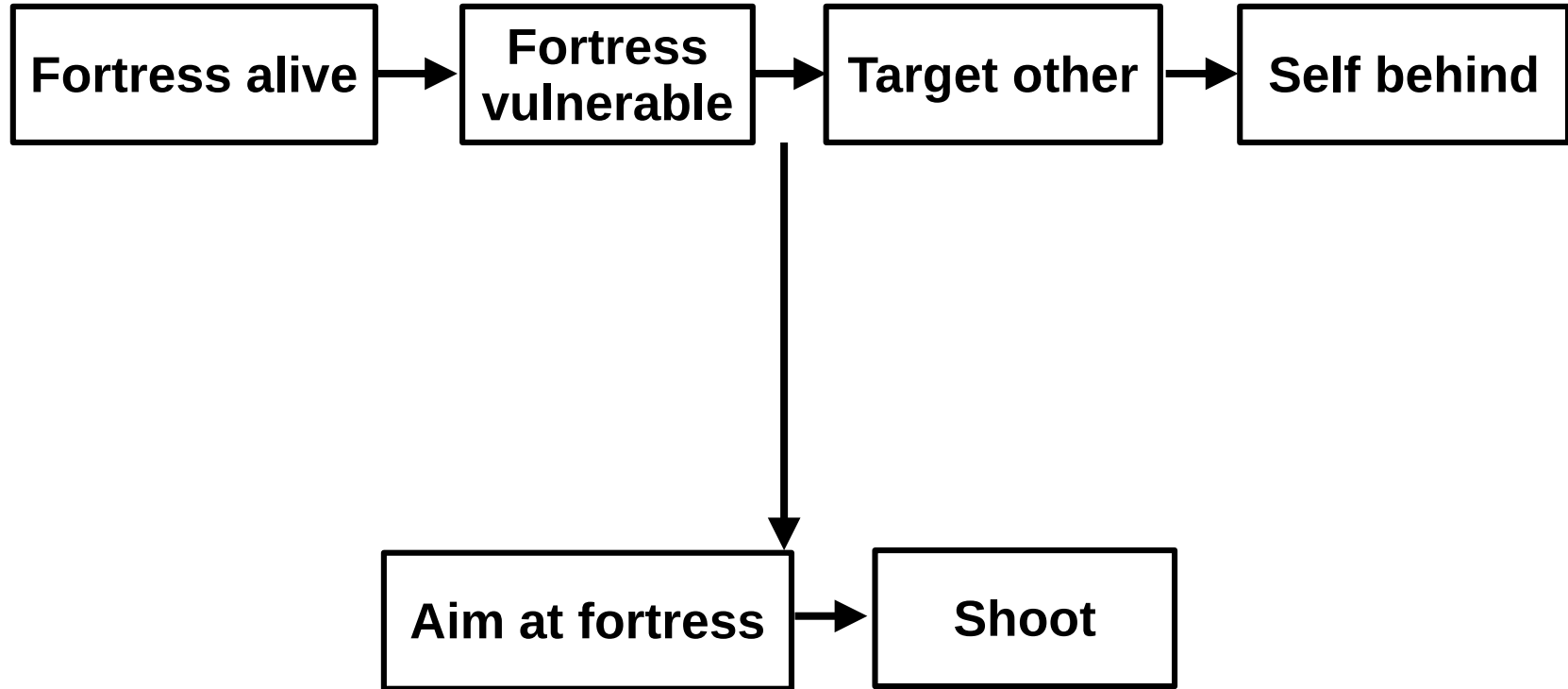
ACT-R Model: Operators



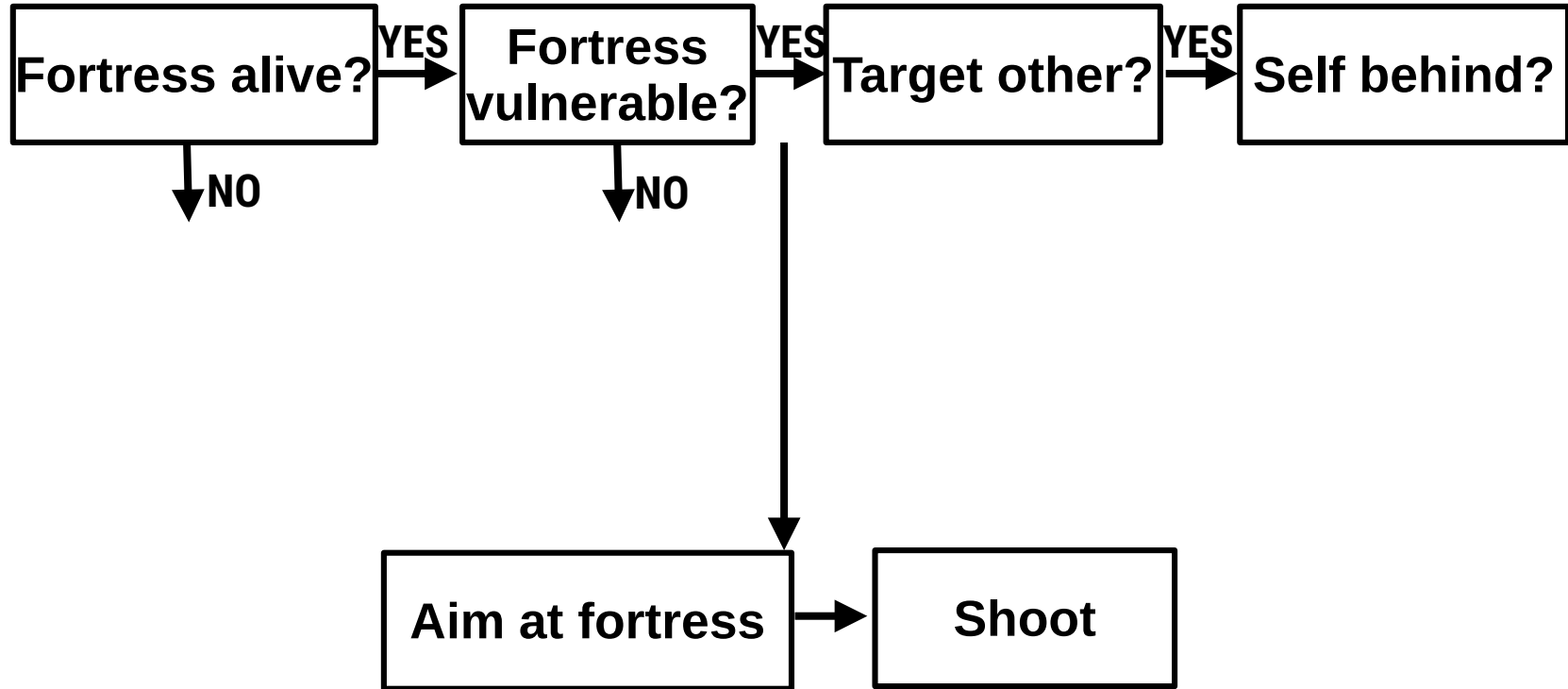
ACT-R Model: Operators



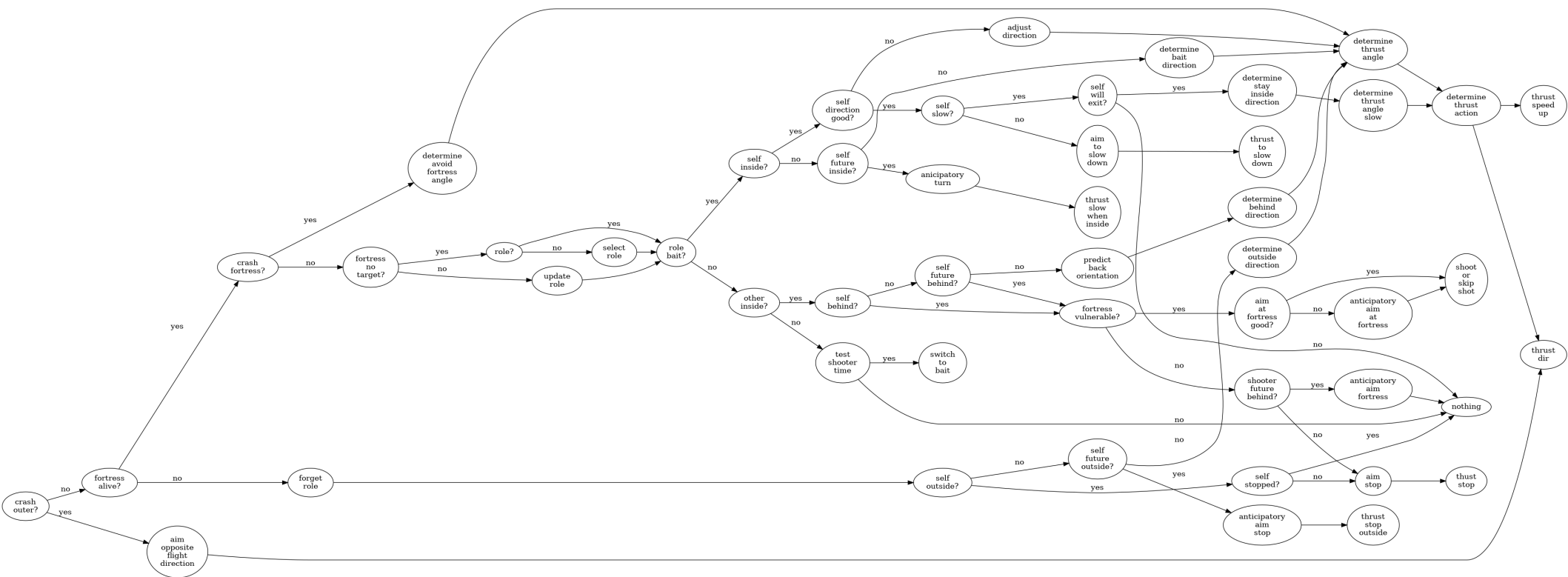
ACT-R Model: Operators



ACT-R Model: Operators



ACT-R Model: Full Operator Tree



ACT-R Model: Production Compilation

P1: Trigger operator retrieval



R1: Retrieve operator



P2: Execute operation

ACT-R Model: Production Compilation

P1: Trigger operator retrieval



R1: Retrieve operator



P2: Execute operation



P2b: Directly execute
operation

ACT-R Model: Control Tuning

- Control tuning tunes parameters to the specific task from feedback about reward rate:
 1. Assume a quadratic relationship between parameter values and reward rate
 2. Assume an exponentially decaying noise around reward rate
 3. Pick a parameter value and sample its reward for a certain period (positive and negative feedback associated with events)
- Example: Learning **bait speed** (positive feedback from fortress missiles and negative feedback from dying)

ACT-R Model: Control Tuning

- Stochastic gradient descent: Adjust control parameter as a function of **error magnitude** and **error direction** after an action
- Example: learning when to **release key press** when turning as a function of feedback about **how far off** the resultant orientation was from the target orientation and whether it **under- or overshoot** the target

ACT-R Model: Anticipatory Actions

- All anticipatory actions in model rely on **motion extrapolation** based on the ship's current position and speed:
 1. Avoid crashing into obstacles
 2. Determine aim at fortress at time of shot
 3. If bait's future location is inside the hexagon, prepare to slow down
 4. If future location is outside of hexagon, prepare to stop

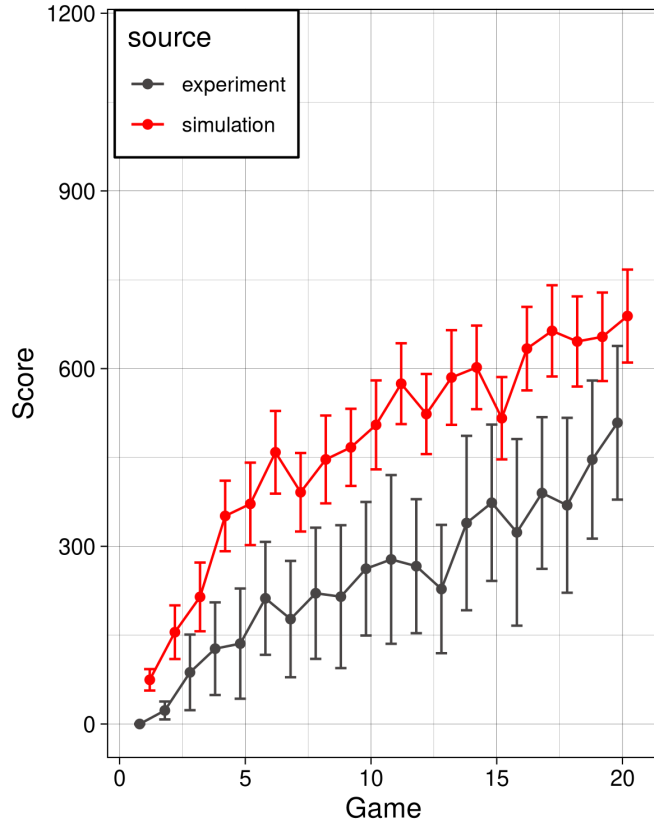
ACT-R Model: Role Adoption

Role adoption is modeled through **reinforcement learning**. Two productions that select bait/shooter compete with each other. The production utility is updated:

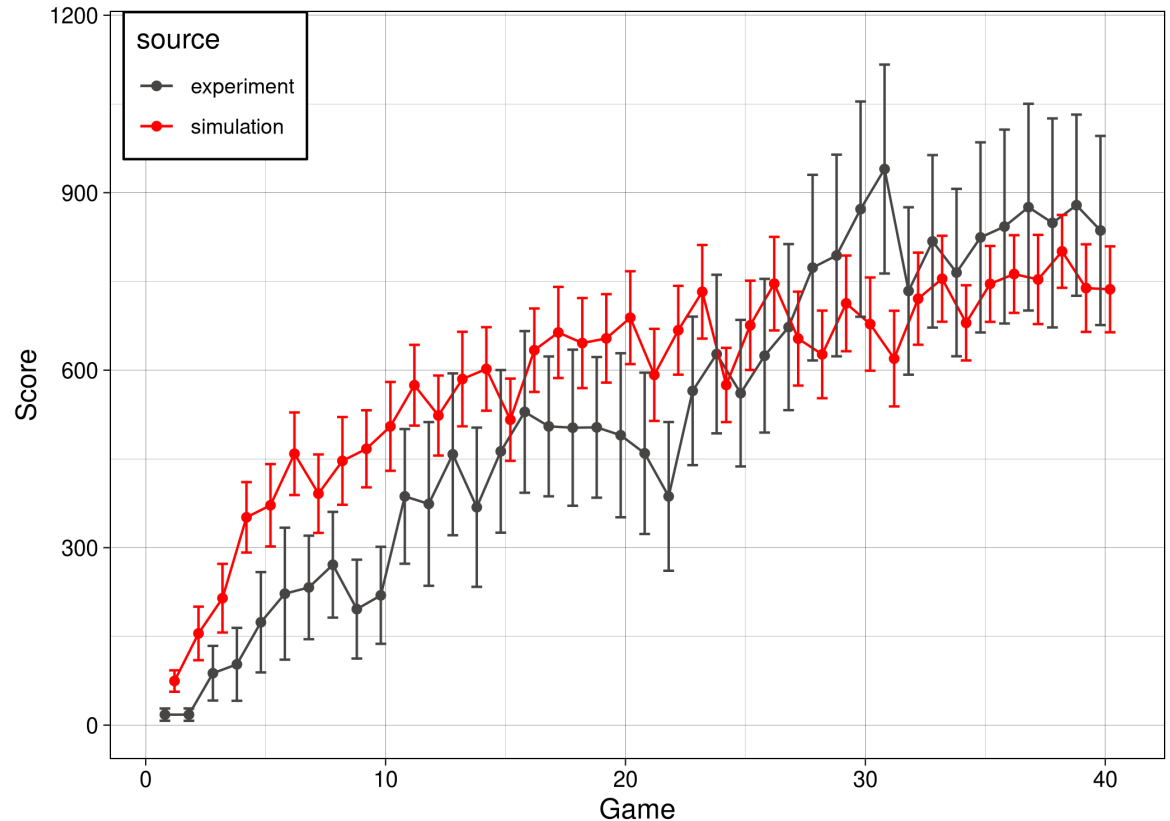
- Each fortress kill brings positive reward
- A role switch brings negative reward

ACT-R Model: Performance

Study 1: Score

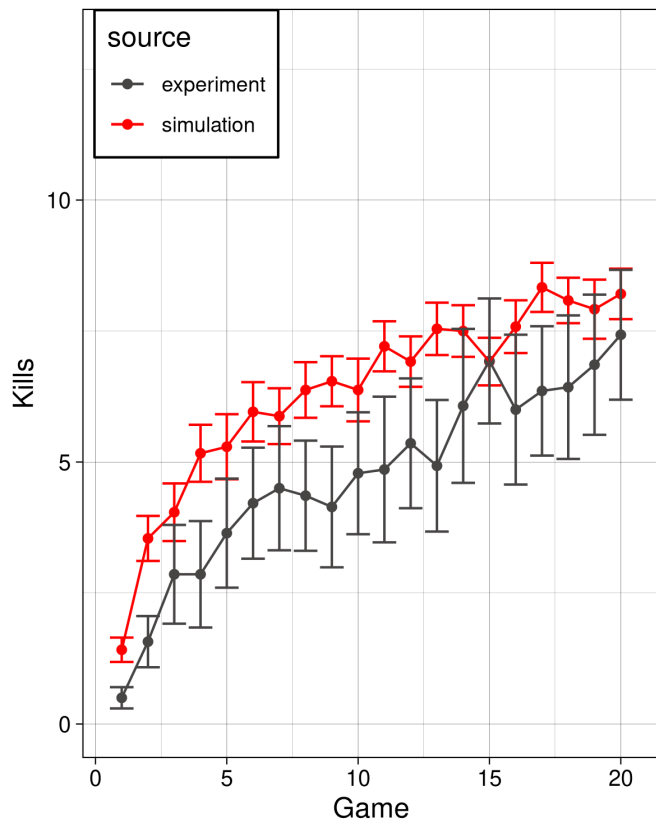


Study 2 (control condition): Score

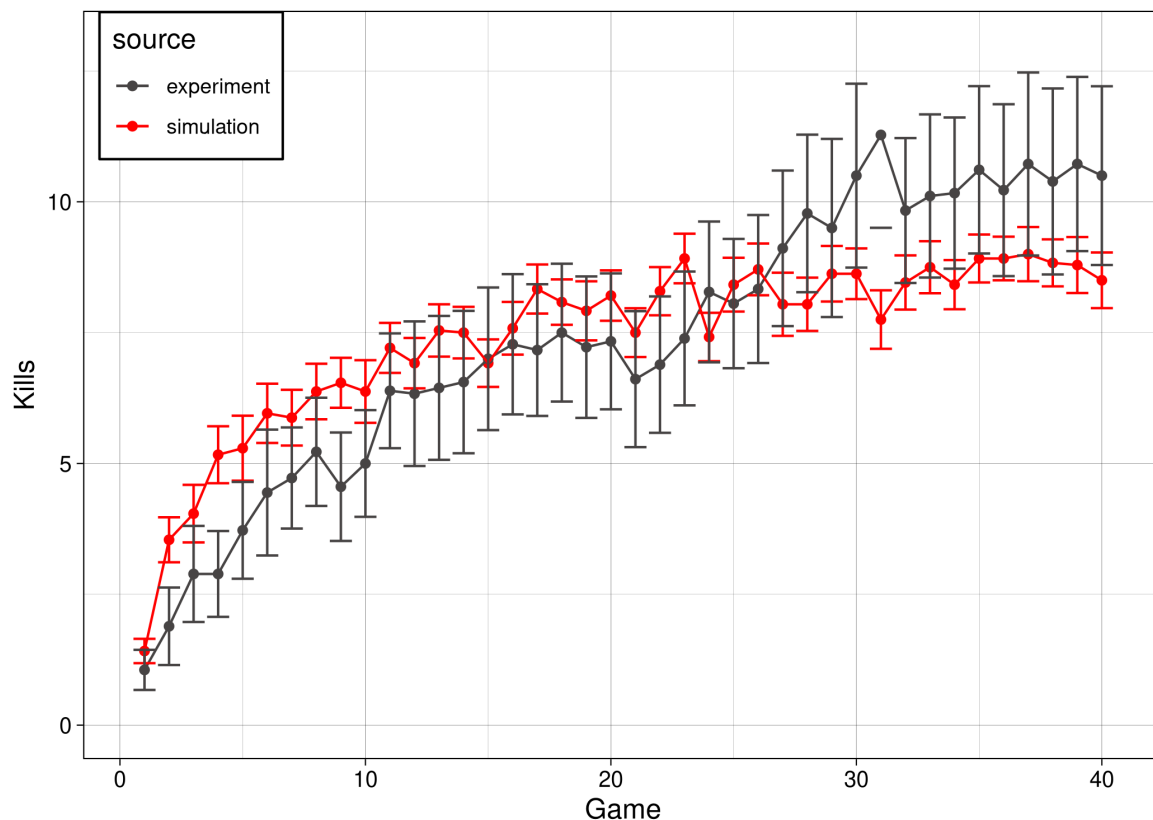


ACT-R Model: Performance

Study 1: Kills

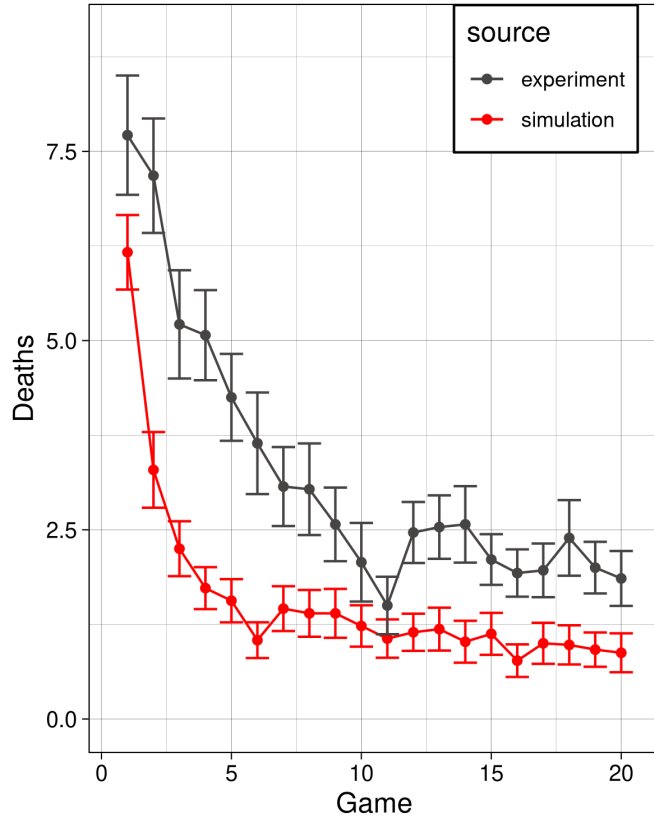


Study 2 (control condition): Kills

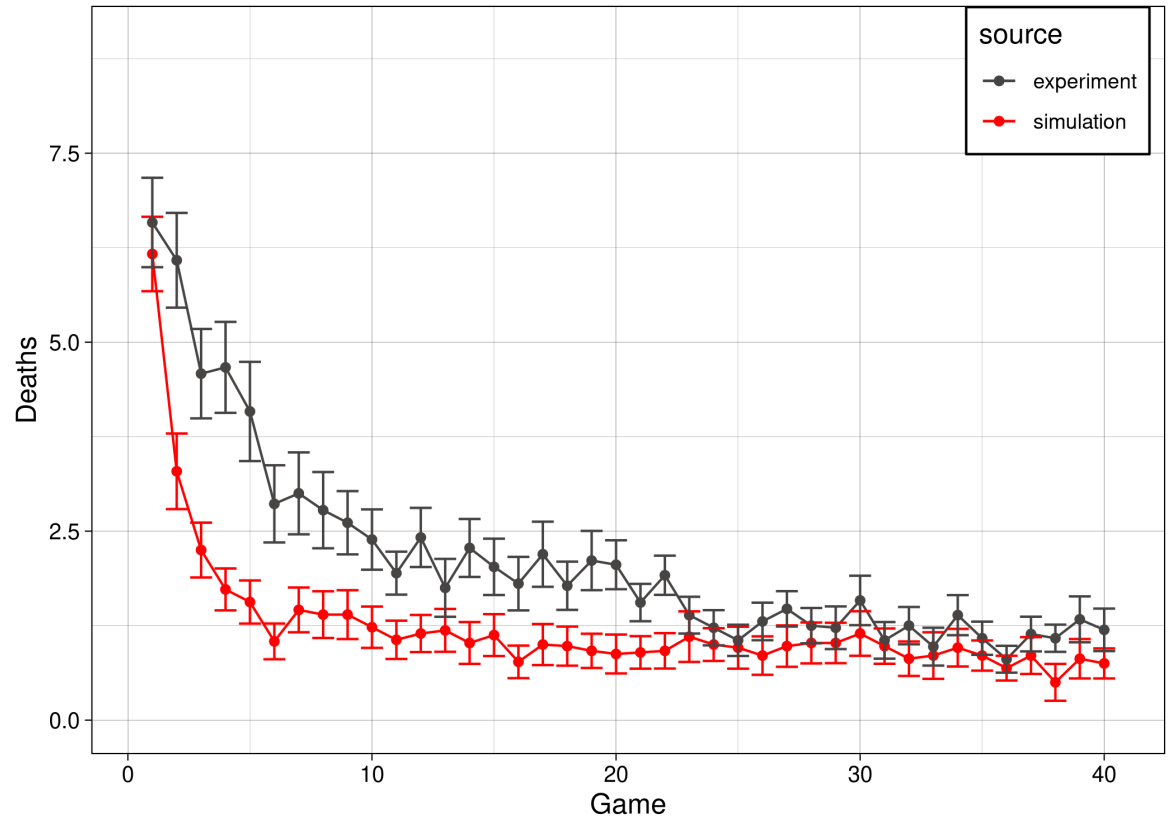


ACT-R Model: Performance

Study 1: Deaths

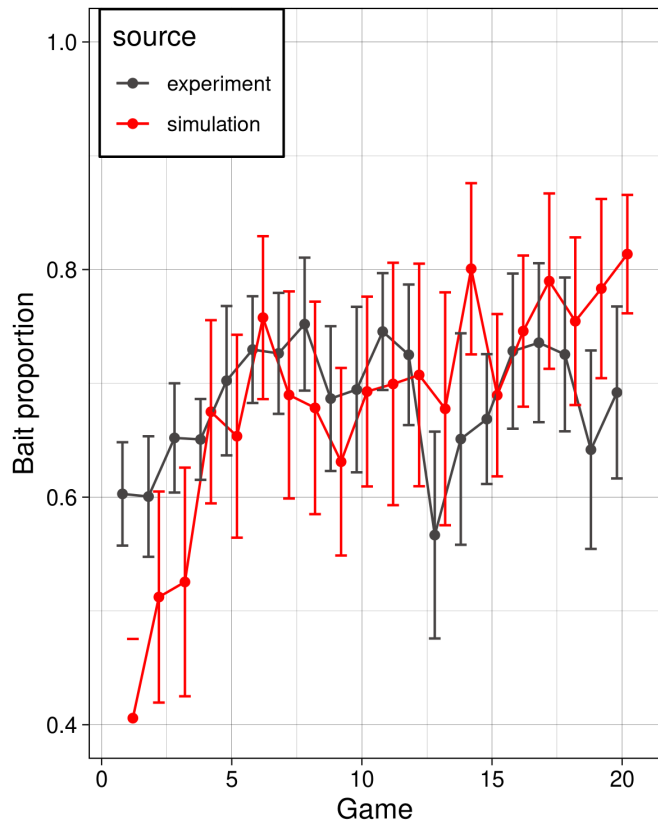


Study 2 (control condition): Deaths

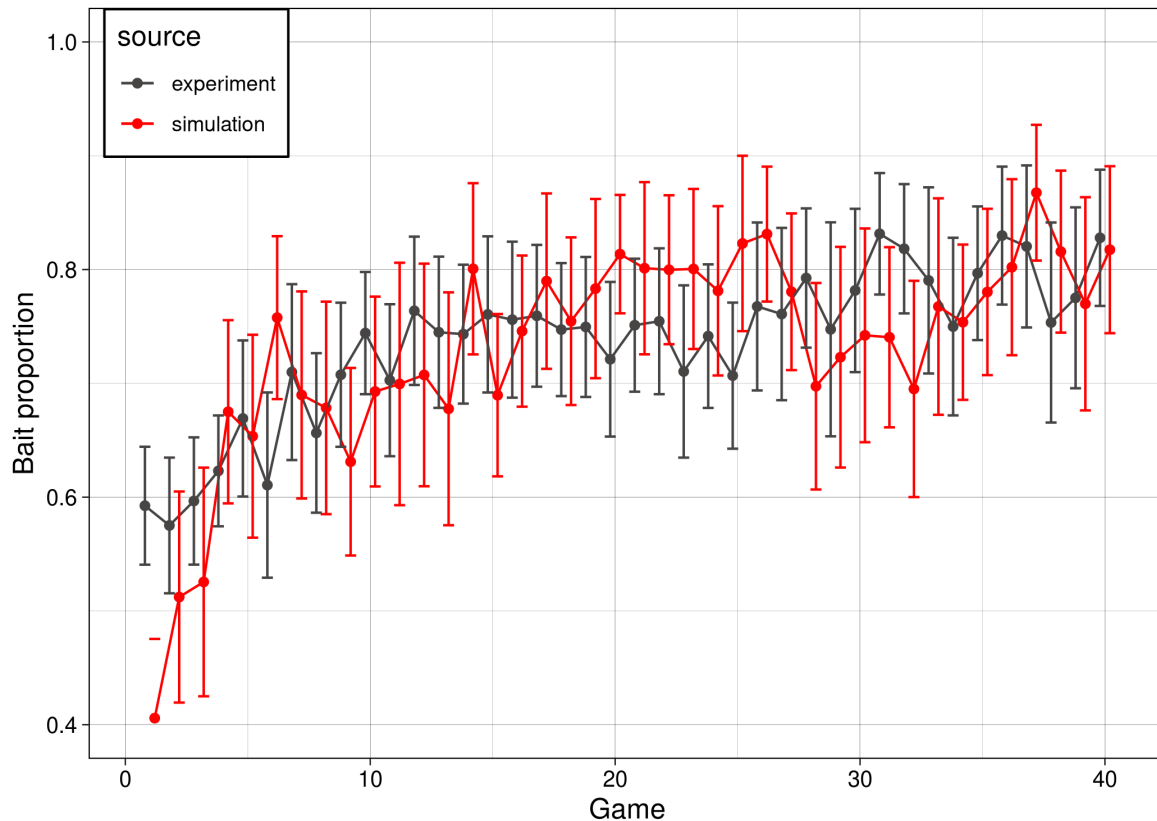


ACT-R Model: Role Consistency

Study 1: Bait proportion



Study 2 (control condition): Bait proportion





Thank you

