

# Feedback Promotes Learning and Knowledge of the Distribution of Values Hinders Exploration in an Optimal Stopping Task

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### Summary

In many naturalistic situations such as deciding on an apartment to rent or selecting a life partner, people explore options before making a selection. We investigate the following:

- How do feedback and knowledge of the distribution of option values affect learning in sequential search?
- How do people **deviate from optimal** based on these factors, and can this be modeled with a **cognitive model** of decisions from experience?

### Introduction

- Previous work indicates that people are suboptimal at stopping exploration and often stop earlier than optimal [1, 5].
- People may learn to stop at the optimal point with experience [3].
- We investigate factors that may influence learning in stopping decisions and extend our previous modeling work [2] to this task.

### Methods

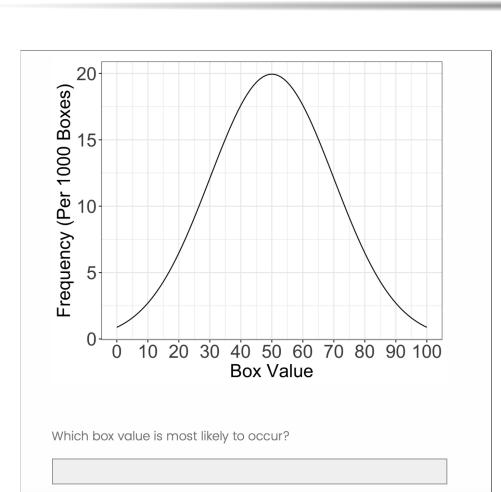
### **Human Participants**

- Participants: 256 from Amazon Mechanical Turk
- Design: Between-subjects
- Conditions: 2 (Knowledge of the Distribution: Known or Unknown) x 3 (Feedback: No Feedback, Outcome, or Detailed)

### Model Agents

• Simulate 256 IBL model and optimal agents

# Optimal Stopping Task



which box value is most likely to occur?

(a) Known Distribution Condition

# Round: 1 of 50 Total points: 0 Box 1 of 10 Pass Select (b) Choice

### Feedback Conditions

• Outcome Feedback:

Your choice was [Correct/Wrong]. You chose Box [Number] with a value of [Value].

• Detailed Feedback:

Your choice was [Correct/Wrong]. You chose Box [Number] with a value of [Value]. The maximum value box was Box [Number] with a value of [Value].

• No Feedback:

You chose **Box** [Number] with a value of [Value].

### Cognitive Model: Instance-Based Learning Model

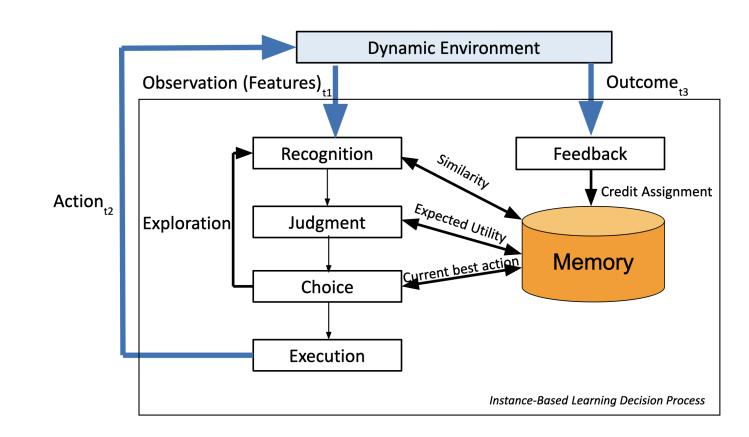
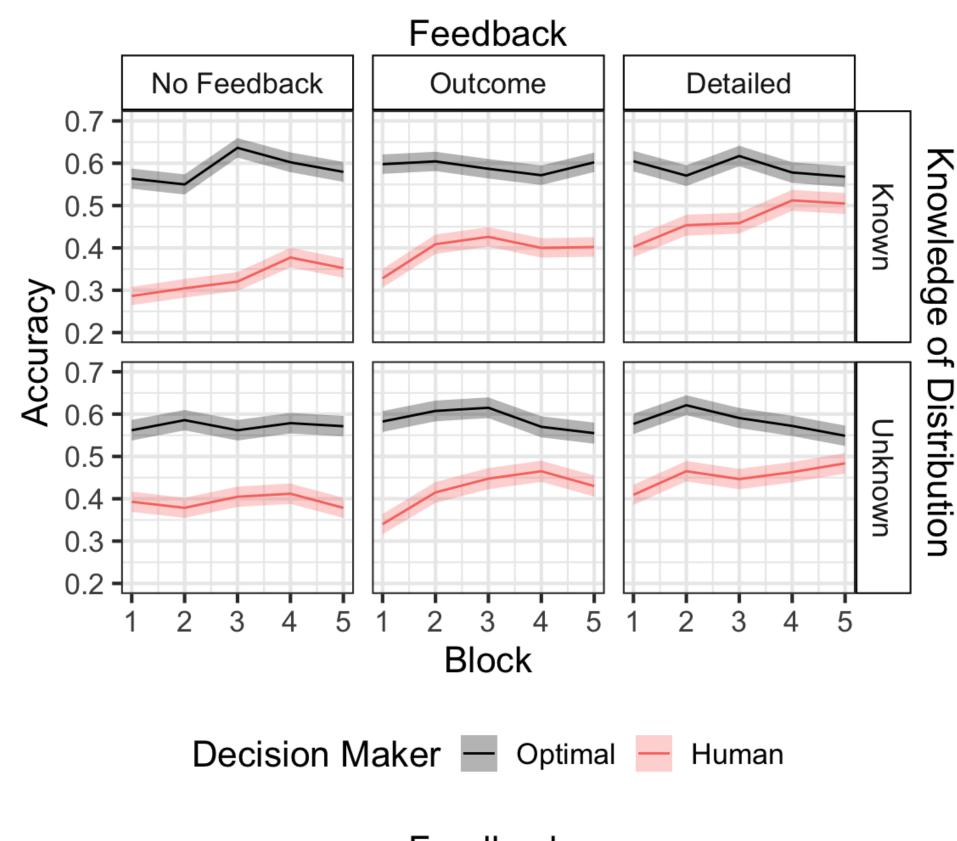


Figure 2: Instance-Based Learning Theory [4]

- Learning occurs through the accumulation of memory units called *instances*.
- Past instances are *retrieved* based on *similarity* to the current situation, frequency, and recency.
- A blended value (BV) is calculated based on the utility of retrieved instances.
- The agent *chooses* the option with the highest BV.

State	Action	Utility
Value Boxes Remaining	{Select, Pass}	$\{0, 1\}$
Table 1: Instance Structure		

### Results



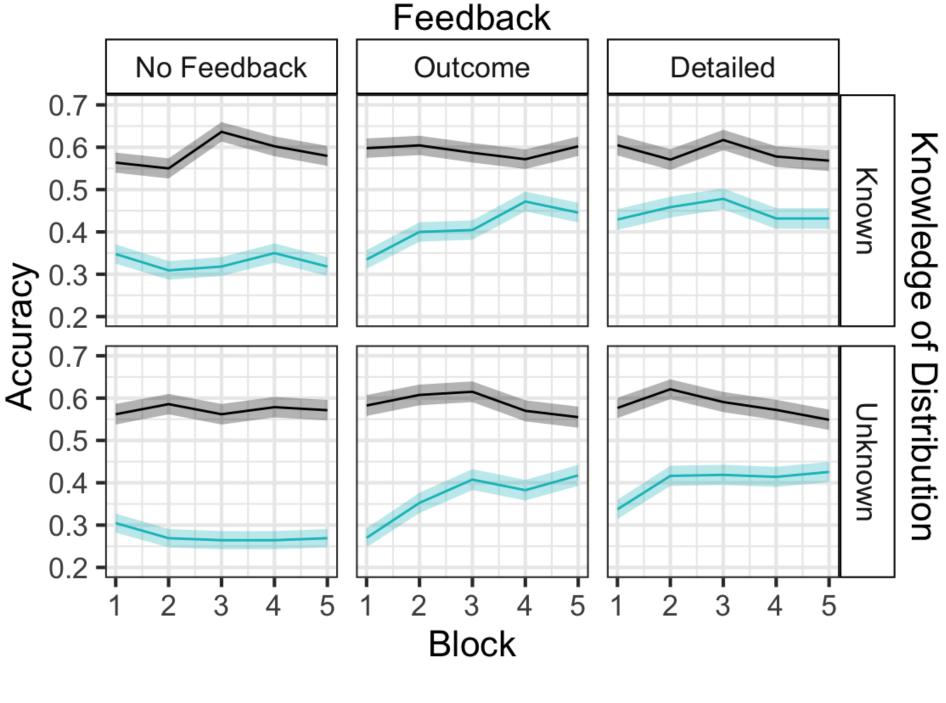


Figure 3: Accuracy for Human Participants (top, red), IBL Agents (bottom, blue), and Optimal Agents (both, black) over Blocks of 10 Problems

# Conclusion

Decision Maker — Optimal — IBL Model

### **Key Findings:**

- Feedback Promotes Learning: Participants in the feedback conditions stopped closer to the optimal position and achieved higher accuracy.
- Knowledge of the Distribution Hinders Exploration: Participants with distribution knowledge explored less by stopping earlier than those without this knowledge.
- Model Predictive Accuracy: The IBL model accurately predicted human stopping behavior across conditions.

### Future Work:

• Investigate additional factors (variability of sequence length, crowd decisions)

### References

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  Management Science, 66(3):1375–1394, March 2020.
- [4] Cleotilde Gonzalez, Javier F. Lerch, and Christian Lebiere. Instance-based learning in dynamic decision making. Cognitive Science, 27, 2003.
- [5] Maime Guan, Ryan Stokes, Joachim Vandekerckhove, and Michael D. Lee. A cognitive modeling analysis of risk in sequential choice tasks.

  \*Judgment and Decision Making, 15(5):823–850, September 2020.

## Additional Information

- Lab Website: cmu.edu/ddmlab
- Personal Website: erinbugbee.com
- For Materials, Paper, Poster: Scan QR code

