

In-video Tutors: Revisited and Beyond

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Background

- Demand for online education, specifically in the domain of AI, has exponentially increased over the past decade with instructional videos being the primary method for delivery¹.
- During the onset of COVID-19, this directly impacted two sets of learners:
 - 19.7 million students enrolling in online post-baccalaureate and post-secondary programs².
 - Approx. 100 million adult workers will require upskilling and reskilling^{3,4}.

Primary Benefit & Challenge of Online Learning Compared to In-Person Instruction

- ✓ **Scalability:** empower students with increased access to education across domains and disciplines and enable instructors to focus on teaching by leveraging AI and automation to avoid routine tasks.
- ✗ **Unattainability:** online learning via MOOCs have lower completion ratios and lower student satisfaction than equivalent in-person instruction largely due to passive learning.

Research Questions^{1,5,6}

- RQ1: How may AI help make online education videos foster active learning and learning by doing?
- RQ2: How do we design an effective, repeatable, and scalable online course in the domain of AI?

A Successful Experiment for Teaching Cognitive Systems Online^{1,5,6}

- The CS 7637: Knowledge-Based Artificial Intelligence (KBAI) course taught as part of the Online Master's in Computer Science degree (OMSCS) at Georgia Tech successfully addressed RQ1 and RQ2.
- Key characteristics of the KBAI course includes:
 - Content based on design-based research and principles of cognitive and learning sciences like *learning by doing*, *learning by example*, *learning by reflection* etc.
 - "Using AI to teach AI" approach such as 100 intelligent tutors embedded in online videos ("in-video tutors") to provide adaptive feedback.

Deep Dive into In-video Tutor Concepts^{1,5,6}

- **Context:** Guide students' understanding of one narrowly defined concept or skill using interactive exercises with in-video tutors
 - E.g.: Completing a semantic network representation problem (see Figure 1 & 2)
- **Development tools:** Using the Udacity infrastructure and custom Python code to evaluate multiple user input types.
- **Methodology:** At each step, the tutor contextualizes feedback in terms of the concept demonstrated
 - **Step 1:** Assess the readability of learner's input.
 - **Step 2:** Checks if input matches the rules of the problem.
 - **Step 3:** Assesses if the final state matches the goal state.
- **Evaluation:** Across two modalities and time periods
 - Class assessment outcomes (Fall 2014): Comparing OMSCS students' (n=200) and in-person students' (m=75) performance, OMSCS students **on average outperformed** in-person students across 14 written assessments and exams.
 - End-of-course student surveys (Fall 2014 to Spring 2017)
 - > 80% students (x=1,242) agreed that the interactive exercises were engaging, and the feedback received enhanced course content understanding.

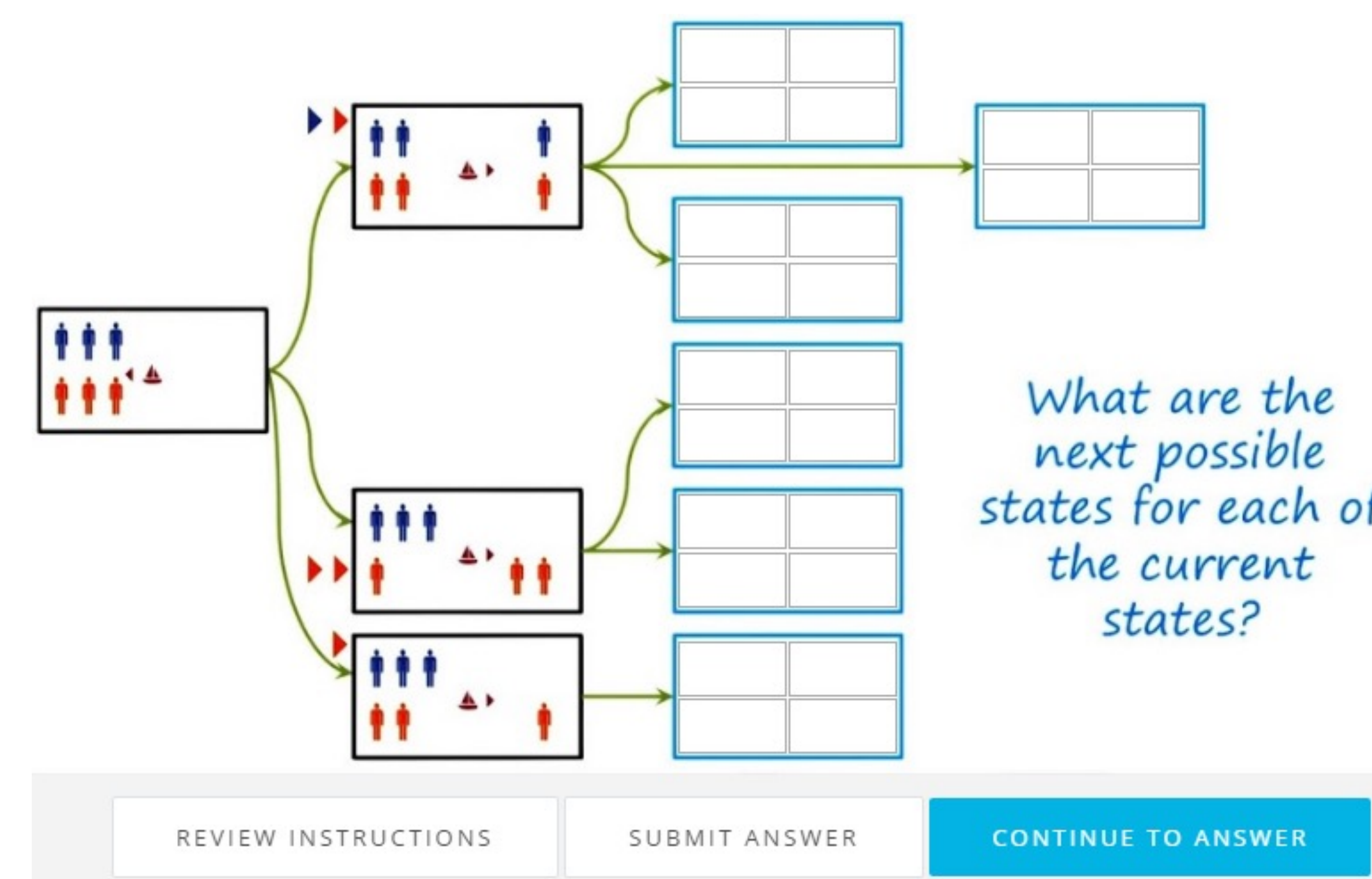


Figure 1: Example exercise from the KBAI course in Fall 2014. Here, students are asked to fill in 24 boxes to represent the possible next states of a problem in means-ends analysis in accordance with rules provided^{1,5}

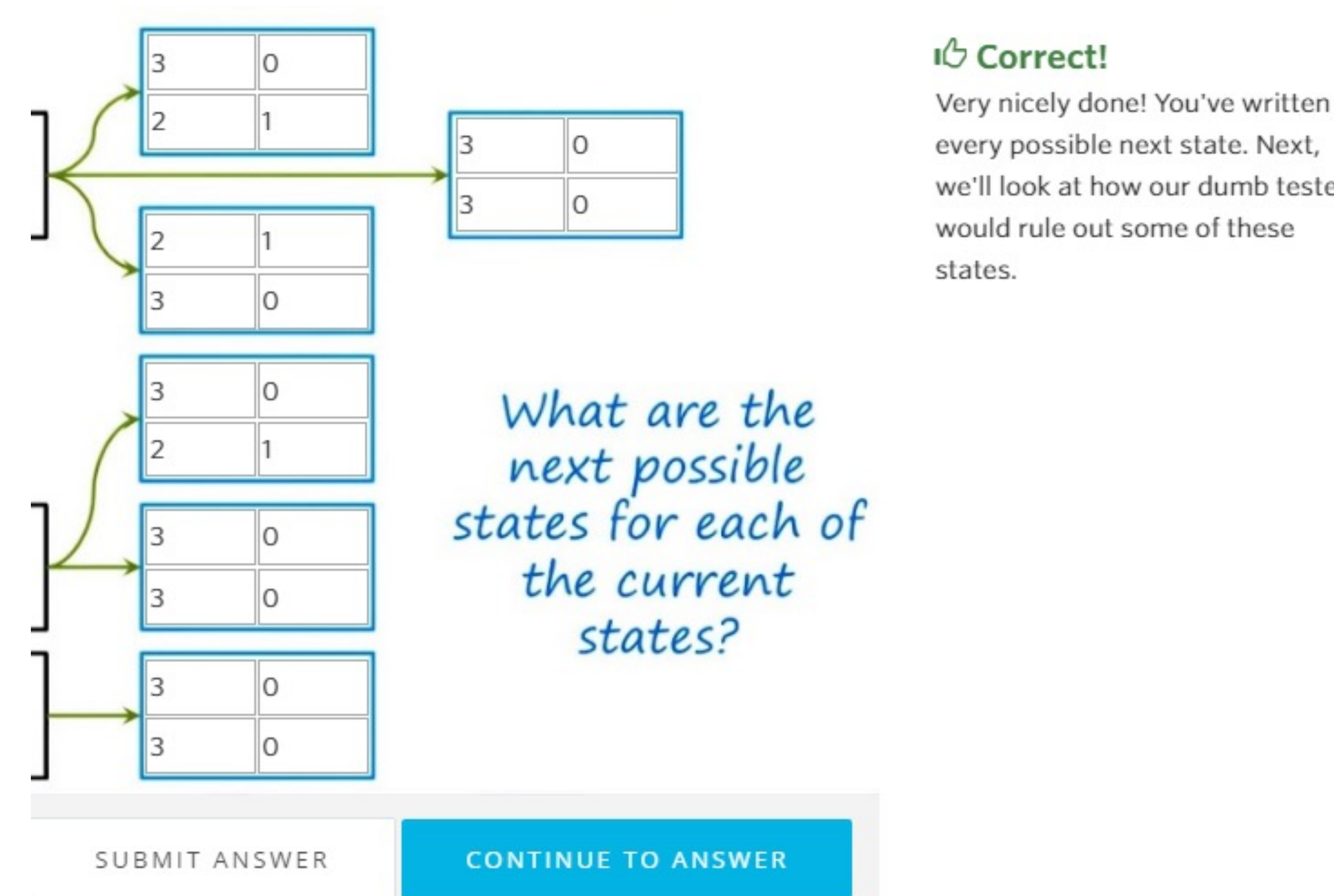
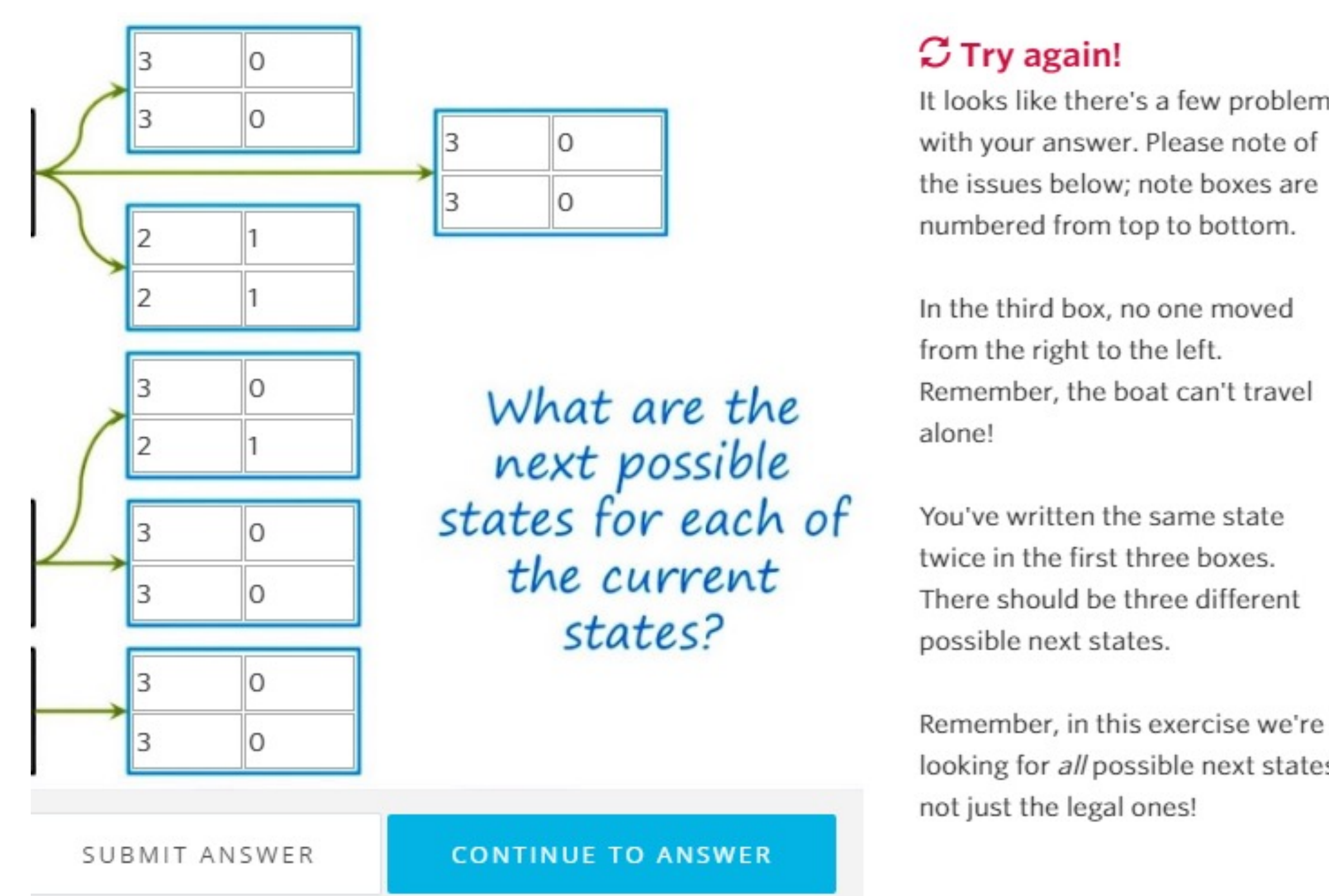


Figure 2: Examples of two pieces of feedback a learner may receive from the in-video tutor based on provided input. The top image represents feedback from the tutor given an incorrect learner response. The bottom image represents feedback from the tutor given a correct learner response^{1,5}

Looking Beyond

- Rethink the ad-hoc process used to develop in-video tutors for the 2014 KBAI course
- Systematically formalize the methodology, architecture and interface ("tech framework") for in-video tutors that is generalizable and scalable for online courses that have two main characteristics:
 - The primary audience are adult learners.
 - The course content focuses on skill-based learning.

Tech Framework Conceptual Approach

- **Methodology** (part of content creation, no tutors involved):
 - Step 1: For a given online class where lectures and exercises are in recorded format, list out all the concepts and skills that an educator would like a student to learn
 - Step 2: For each concept ($1..n$), an educator would list out common student misconceptions ($1..m$)
 - Step 3: For each concept and associated set of misconceptions, an educator would create an exercise to better understand the relationship between incorrect answers (provided by a subset of students) and those misconceptions - this will be handled by the tutor (part of the architecture).
- **Architecture:** build a tutor for each exercise that can (1) recognize incorrect student responses, (2) correlate an incorrect answer with a misconception, and (3) explain the reasoning
 - For a given exercise, if there are i possible solutions where $i-1$ solutions are incorrect, and there are m possible misconceptions, an in-video tutor would be able to understand student responses and "match" $i-1$ incorrect solutions with m misconceptions - potentially as a many-to-many relationship
 - For a given (incorrect answer, misconception) pair, the tutor would be able to give an explanation.
 - To what extent will the above "architecture" be accomplished by a low-level architecture is yet to be determined, however, the primary objective of this "architecture" is to not hinder an educator with these (low-level) technical architecture details.
- **Interface:** Develop an intuitive, high-level interface allowing any educator to build in-video tutors with a course content based on the above-mentioned methodology.

References

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