

ASEE 2022 ANNUAL CONFERENCE

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# Adaptive Virtual Assistant for Virtual Reality-based Remote Learning

Hannah Ava Sloan  
(Computer Science, University of Calgary)

Dr. Richard Zhao  
(Computer Science, University of Calgary)

Dr. Faisal Aqlan  
(Industrial Engineering, University of Louisville)

Dr. Hui Yang  
(Complex Systems Monitoring, Modeling and Control  
Laboratory, Penn State University)

Dr. Rui Zhu  
(Industrial and Systems Engineering, University of Oklahoma)

# Introduction

- This paper discusses the development of an adaptive virtual assistant in an immersive virtual reality (VR) serious game
- VR opens new opportunities for teaching and learning manufacturing, and enables remote learning from any physical location
- We propose an adaptive virtual assistant in the game environment to support the student learning process
- Using reinforcement learning, the virtual assistant can learn and adapt to the student's preference in the types of assistance to provide

# Serious Games

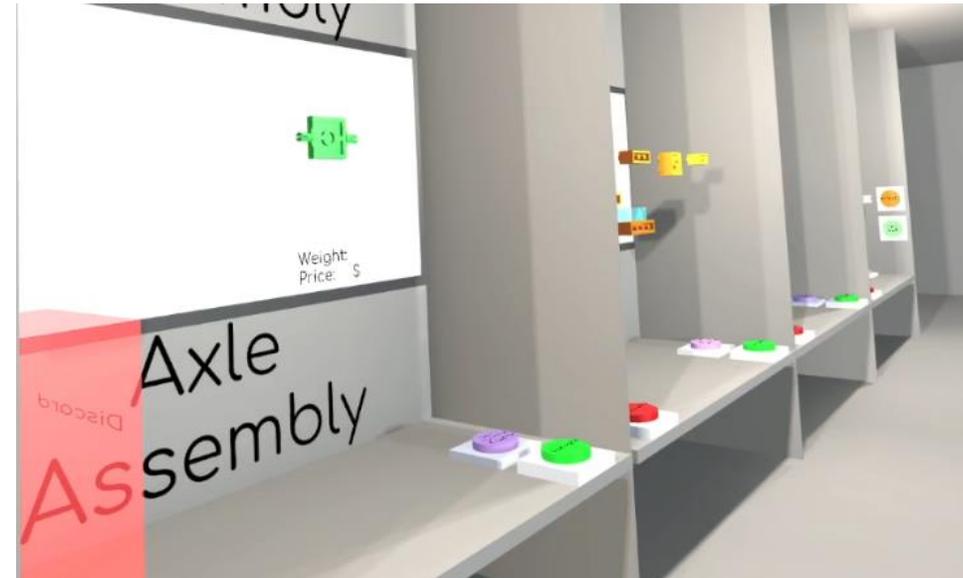
- Using games as a form of training and education has been a growing trend
- Video games have been used to teach
  - computer science
  - computational thinking
  - programming logic
  - biological engineering
  - sustainability
  - Biology
  - problem solving, etc.

# Virtual Reality (VR)

- VR provides a more immersive experience. Getting popular due to lightweight and affordable VR headsets
- Serious games for the purpose of education, training and military applications have started deploying VR technology
- Researchers examined different input schemes, such as head-tracking, or the effects of locomotion methods on the player's experience

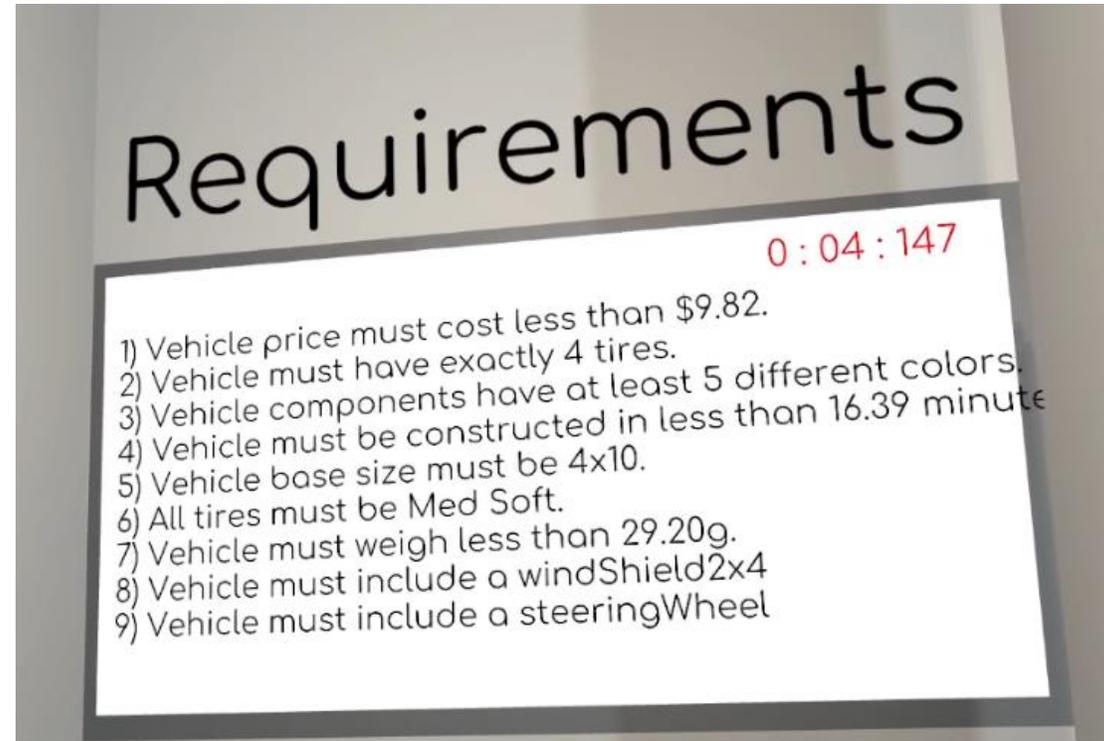
# Research Methodology

- The proposed research develops a gamified VR learning environment together with a virtual assistant to enhance student understanding of manufacturing concepts
- The task is toy car assembly on an assembly line



# Research Methodology

- The task is toy car assembly on an assembly line



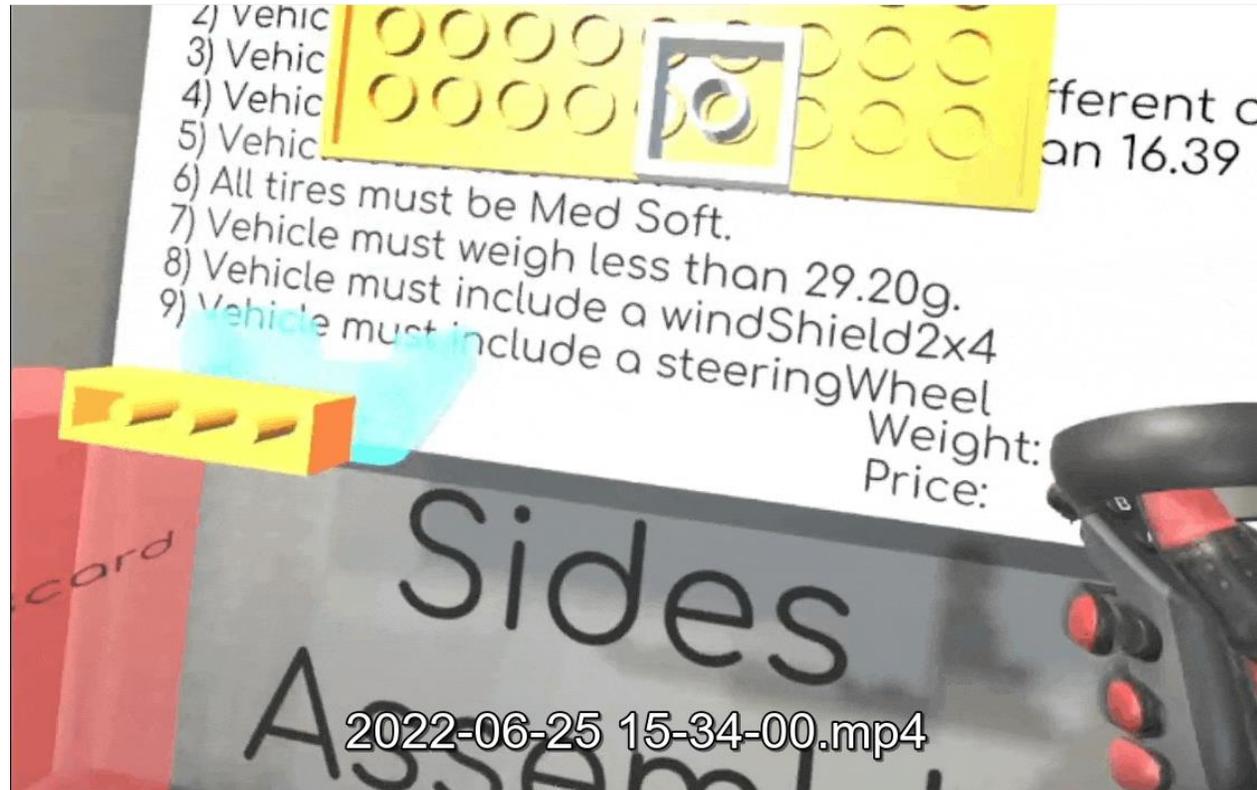
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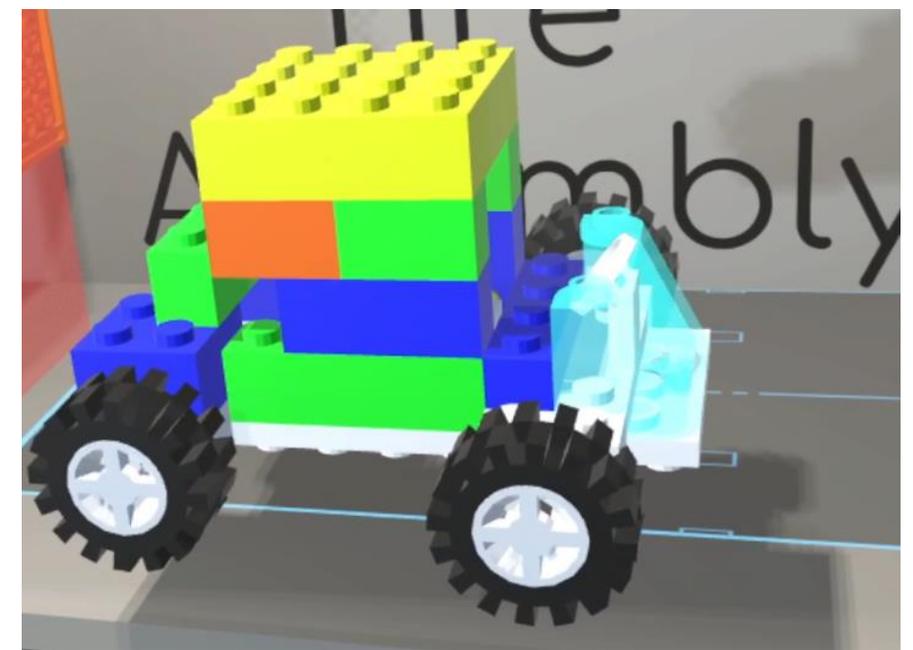
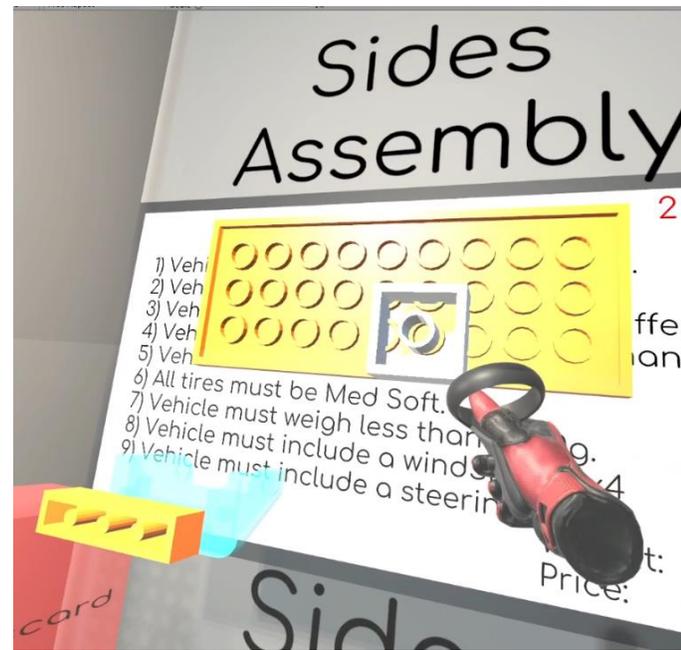
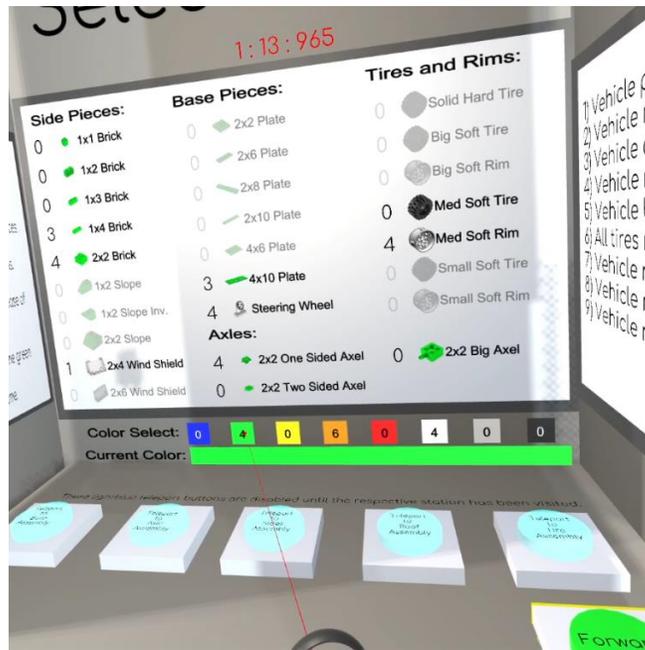
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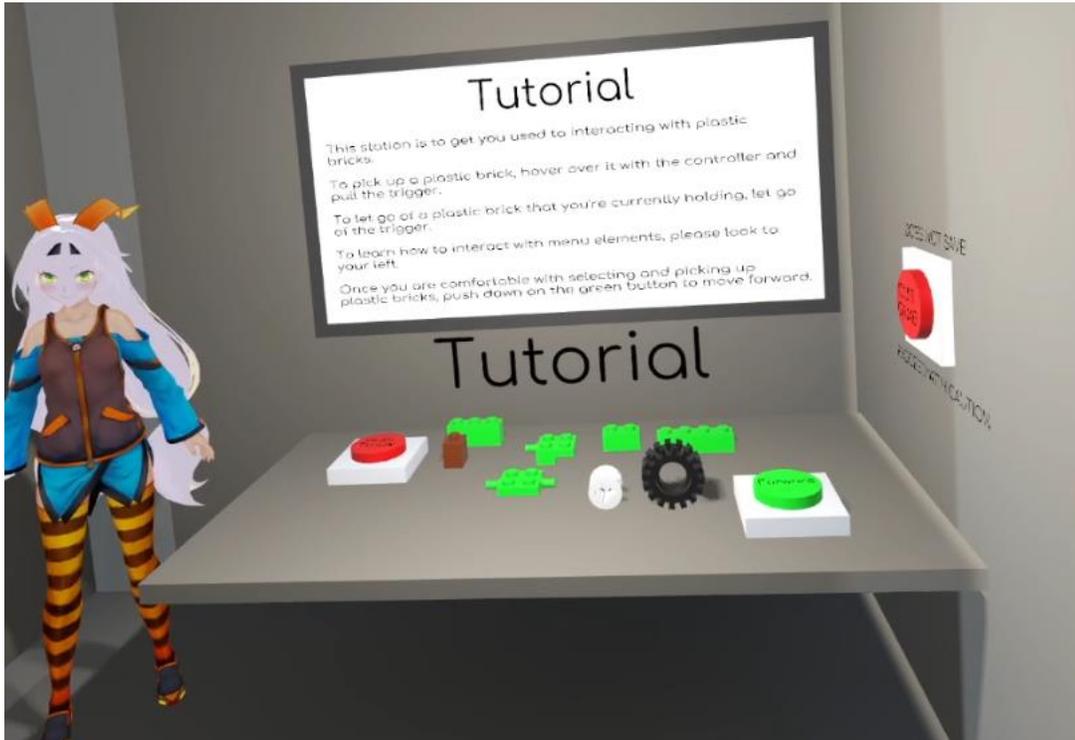
# Research Methodology

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# Gamified VR

- Tutorial and score board



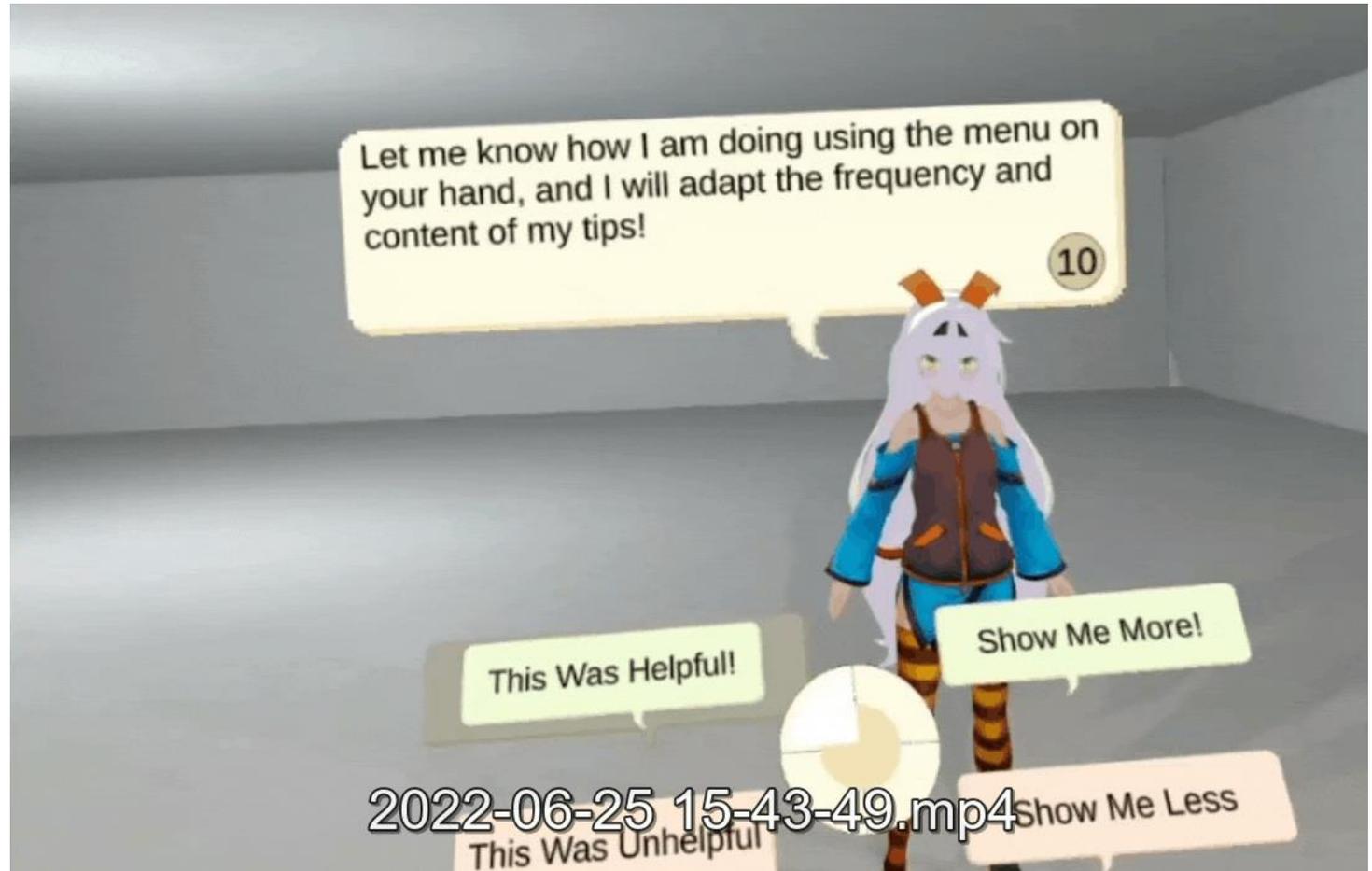
# Gamified VR

- Leader board
- Level selection



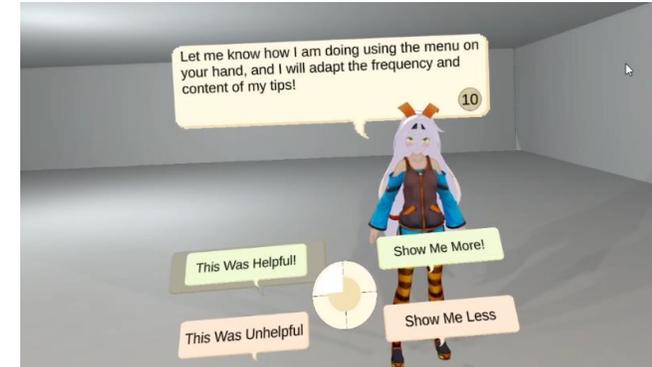
# Adaptive Virtual Assistant

- The purpose of the virtual assistant is to help the student as the student completes the problem-solving task



# Adaptive Virtual Assistant

- The purpose of the virtual assistant is to help the student as the student completes the problem-solving task
- Support from the virtual assistant can be categorized into two categories, hints and reminders.
  - Hints can be generated when students make mistakes in the tasks, such as when an obvious wrong plastic component was chosen for a task
  - Reminders can be generated when a student is at risk of failing to achieve a requirement, such as when the toy car being assembled is getting close to the maximum weight allowed



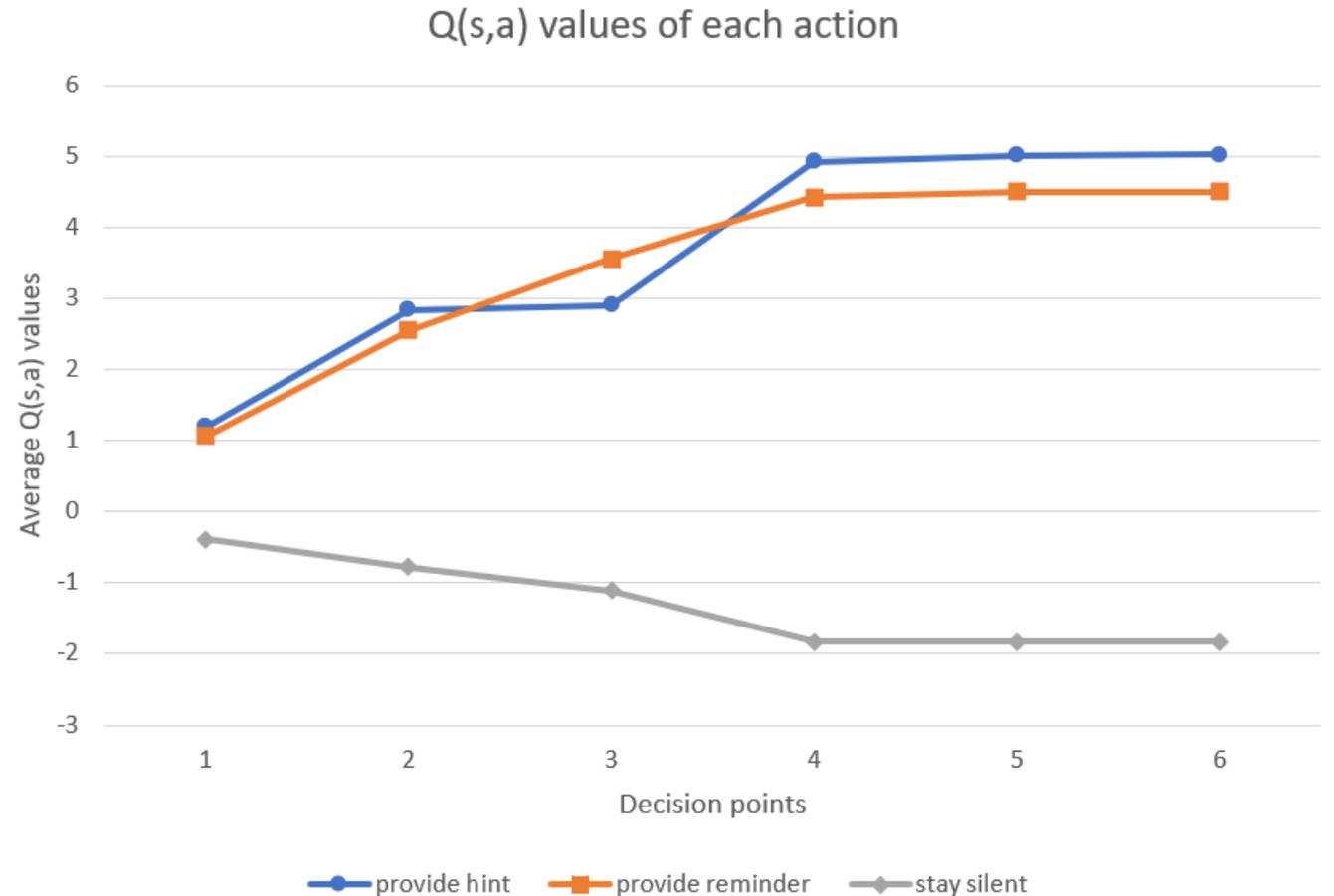
# Adaptive Virtual Assistant

- Adaptation through Reinforcement Learning
- Reinforcement learning is an area of machine learning where an AI agent is not told what to do and has to discover the appropriate actions to maximize a notion of a numerical reward.
- In RL
  - a set of world states  $s \in S$
  - a set of actions  $a \in A$
  - an unknown reward function  $R(s,a) \rightarrow r$  that outputs a reward  $r$  for each state  $s$  and action  $a$
  - an unknown state transition function  $T(s,a) \rightarrow s'$  that takes a state  $s$  into the next state  $s'$ .
  - a  $Q(s,a)$  value represents how good an action  $a$  is, in the state  $s$

Student action	Reward given	Rationale
Student responds to the hint/reminder by selecting "Show me more!"	+5	Explicit responses result in the largest positive reward
Student responds to the hint/reminder by selecting "Show me less"	-5	Explicit responses result in the largest negative reward
Student looks at the hint/reminder for more than 2 seconds and turns away without a response	+3	No explicit response, but student reads what the assistant provides
Student looks at the hint/reminder for less than 2 seconds and turns away without a response	-3	No explicit response, and student did not read what the assistant provides
Student ignores the provided help	-2	Student does not want to interact with assistant
Student does not look for help when the assistant remains silent	+3	Student does not want to interact with assistant and assistant correctly remains non-intrusive
Student looks for help when the assistant remains silent	-3	Student wants to interact with assistant

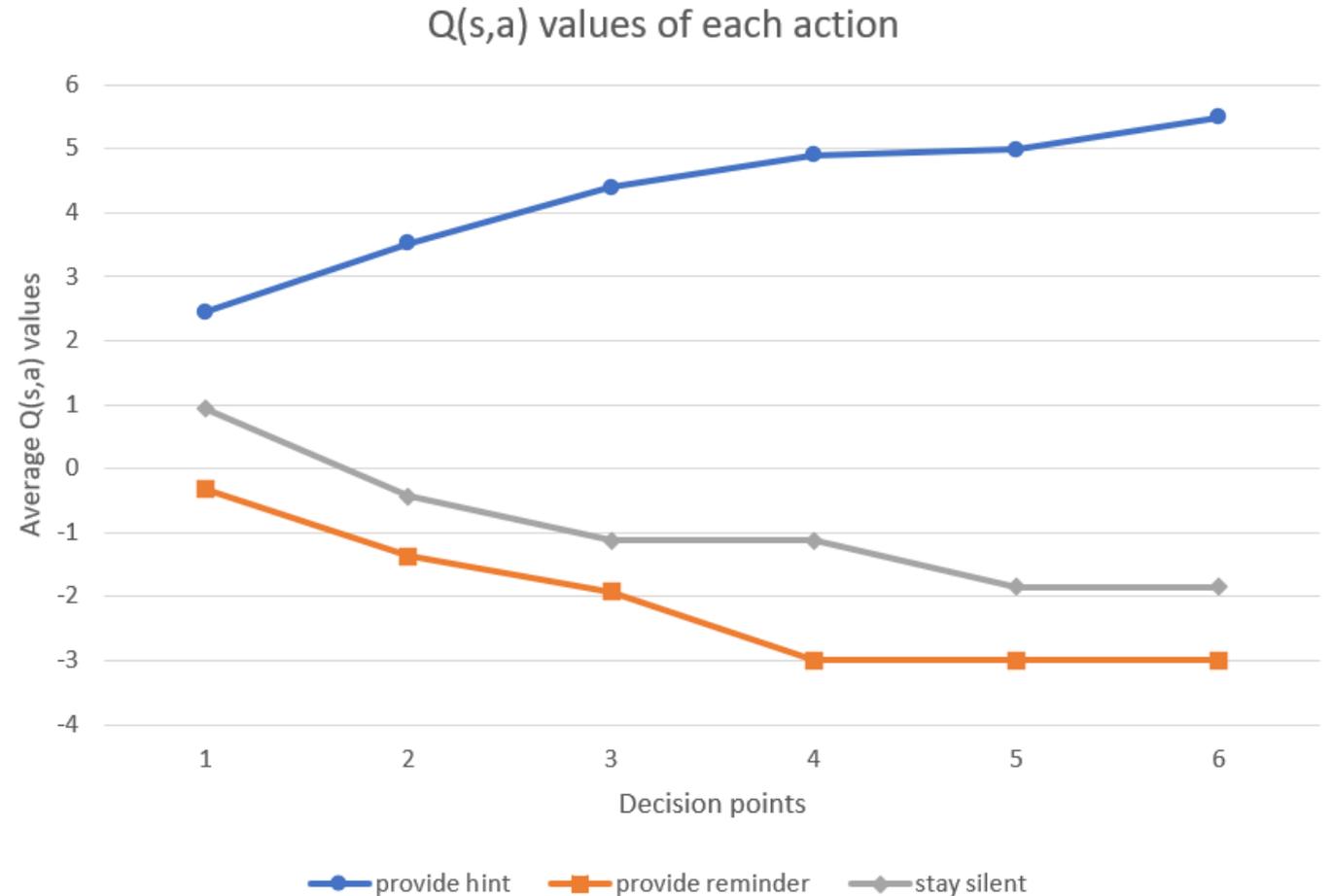
# Experimental Results

- Learned  $Q(s,a)$  values of each action over 6 decision points where the simulated student is receptive to help.



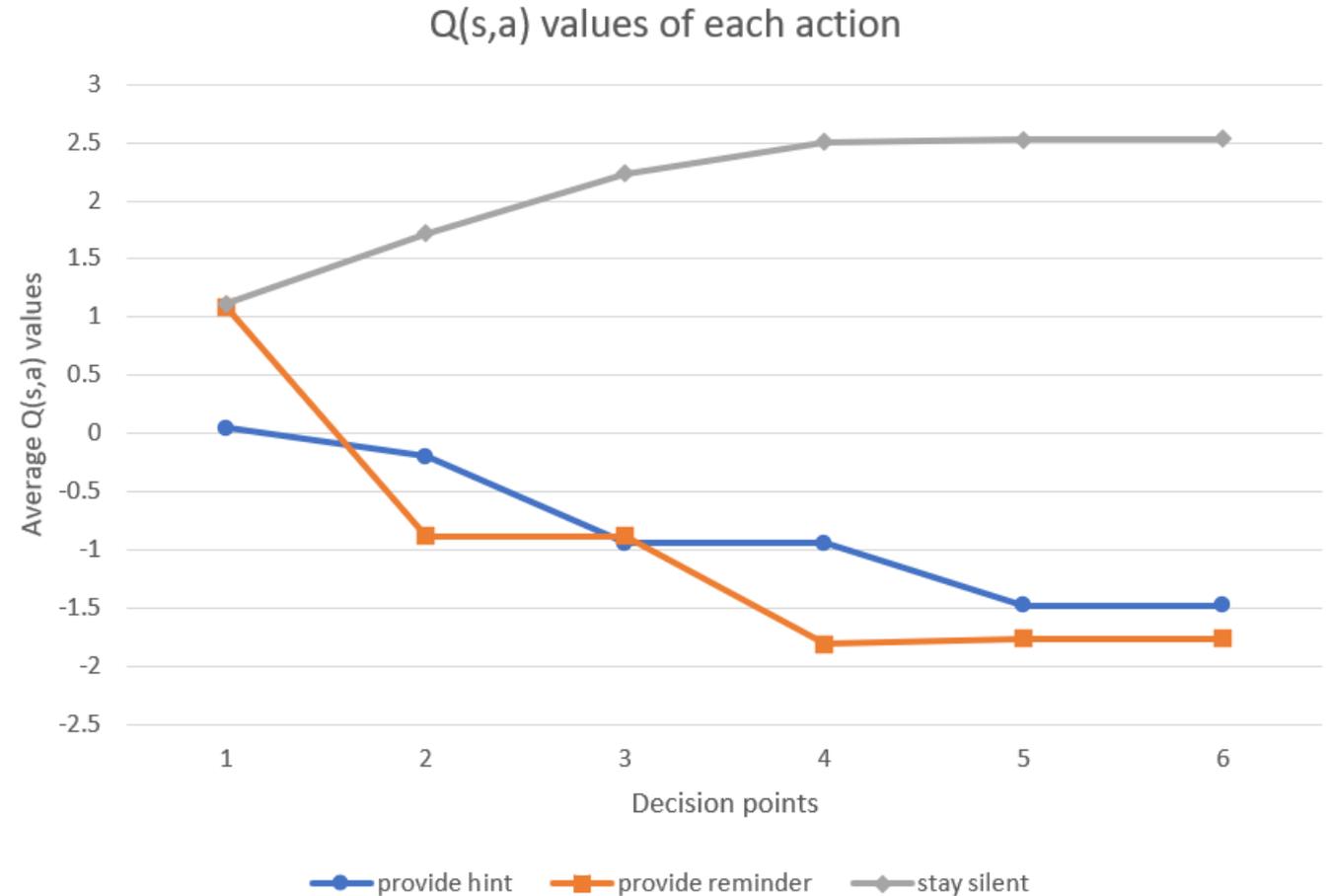
# Experimental Results

- Learned  $Q(s,a)$  values of each action over 6 decision points where the simulated student is receptive to hints, not reminders.



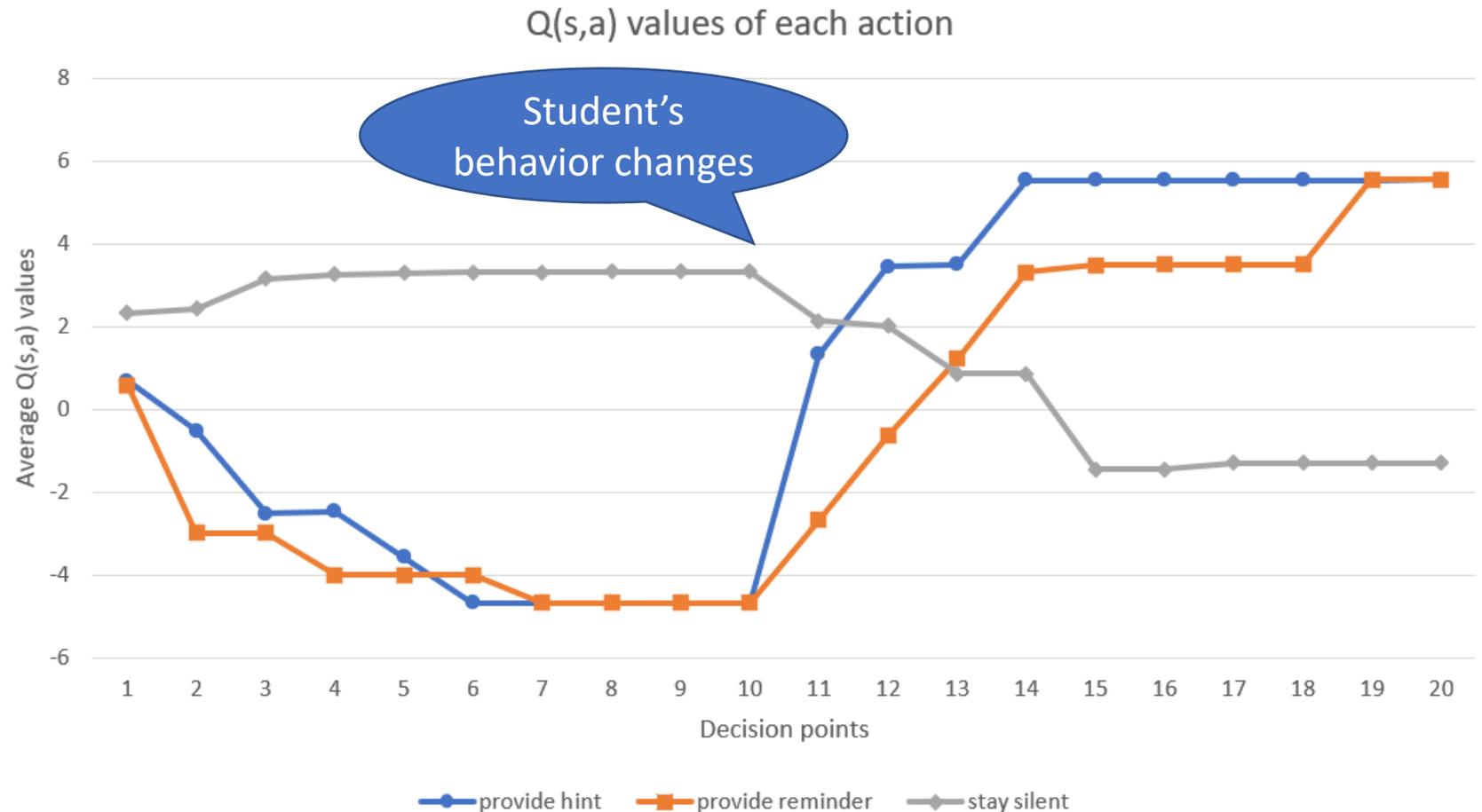
# Experimental Results

- Learned  $Q(s,a)$  values of each action over 6 decision points where the simulated student is not receptive to help



# Experimental Results

- student is not receptive to help during the first 10 decision points, but changes mind to ask for help



# Conclusions

- We integrated gamification with reinforcement learning-based adaptive virtual assistant who can provide help as students complete the problem-solving tasks
- Using reinforcement learning, the virtual assistant can learn and adapt to the student's preference in the types of assistance to provide
- Adaptive virtual assistants can be deployed in many types of learning environments for different subject areas
- Thank you!