

# Multi-Agent Planning Method Using Affordances from Environment

***Sawako Tajima***<sup>1</sup>, *Daiki Takamura*<sup>1</sup>, *Daiki Shimokawa*<sup>1</sup>  
*Reo Kobayashi*<sup>1</sup>, *Reo Abe*<sup>2</sup>, *Satoshi Kurihara*<sup>2</sup>

<sup>1</sup> *Graduate School of Science and Technology. Keio University(Japan)*

<sup>2</sup> *Faculty of Science and Technology. Keio University(Japan)*

Contact email: **sawako\_tajima\_0727@keio.jp**



# Presenter's Bio

- **Sawako Tajima** is a graduate student majoring in administration engineering at Graduate School of Science and Technology, Keio University.
- My research area is Artificial Intelligence (AI), especially autonomous agent.

# Collaborative Project

- Our goal
  - Constructing interactive story-type contents generation system

Realizing the characters  
in the story



3D Scene

This character  
interaction seems  
to be useful...



Creator

→ We need to achieve AI characters that can act autonomously

# Overview

- Research Objective
  - Reducing the explosion in the number of agents in multi-agent planning
- Proposal
  - A method for efficiently obtaining action sequences by using affordances in multi-agent planning
- Result
  - Efficient planning was possible when affordances were used

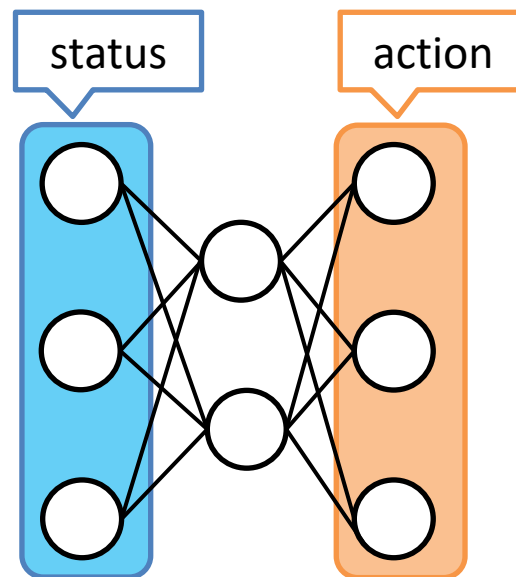
# Planning

- The process of thinking actions to achieve goals from the environment

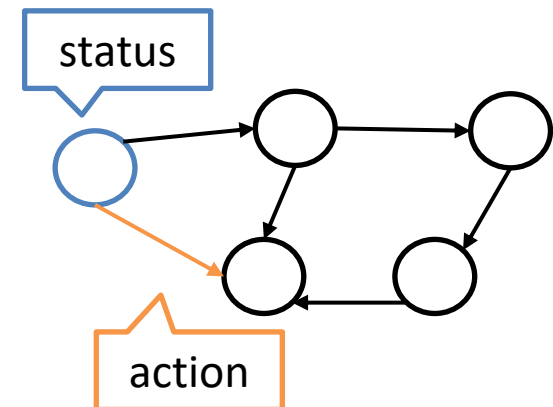
Table

$s_t$	$a_t$
$s_1$	$a_1$
$s_2$	$a_2$
$s_3$	$a_3$
$s_4$	$a_4$

NN

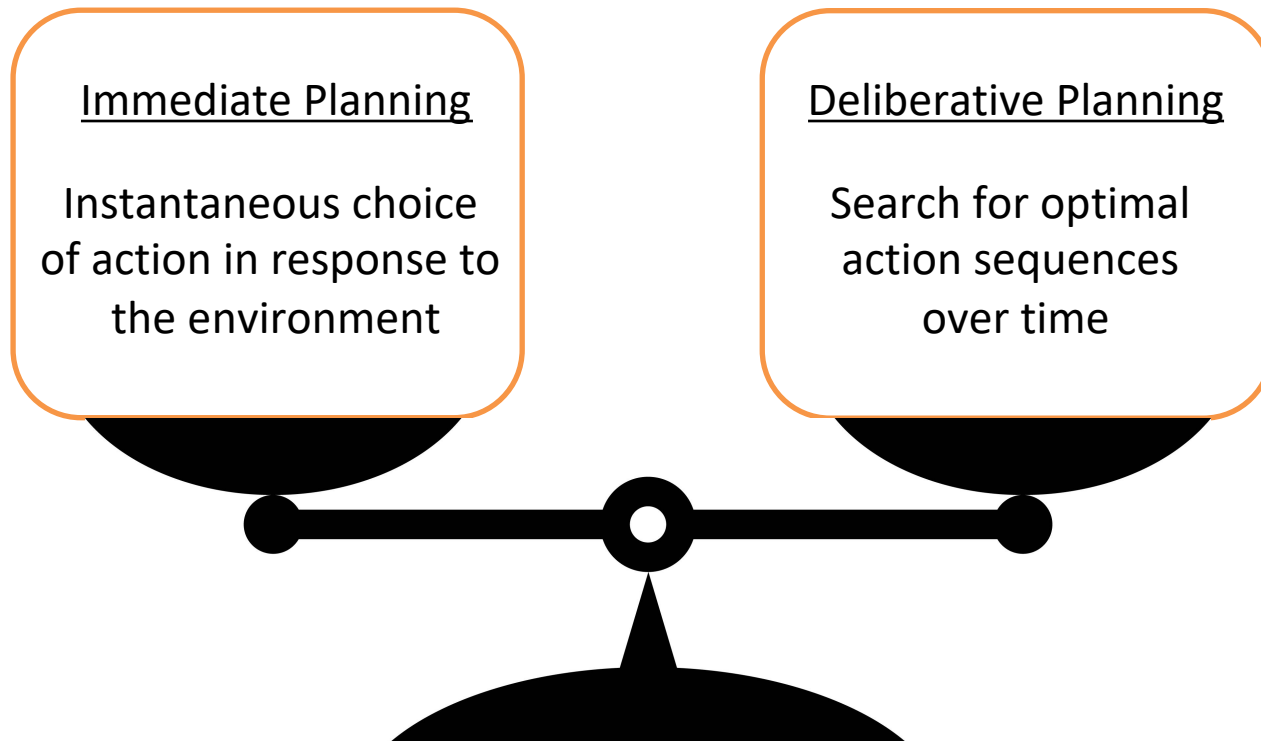


Graph



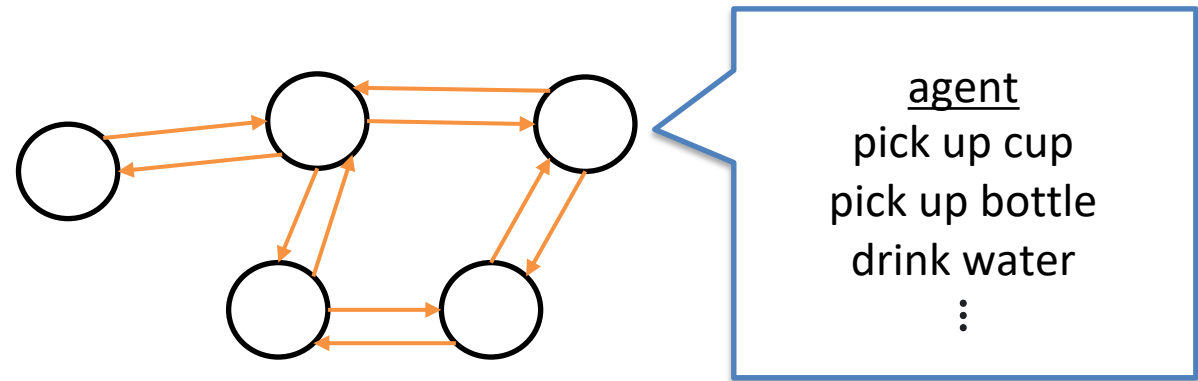
# Planning

- Responding to dynamically changing environments
  - Need for **both immediacy and deliberateness**



# Multi-Agent Planning (MAP)

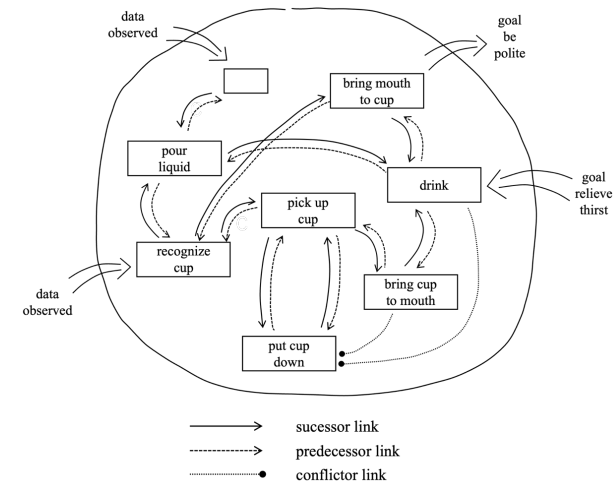
- Planning by interaction (→ in the figure) of agents (equivalent to operants in STRIPS)
- Able to combine immediacy and deliberateness



The number of agents is vast when the environment is large  
→ Use affordances

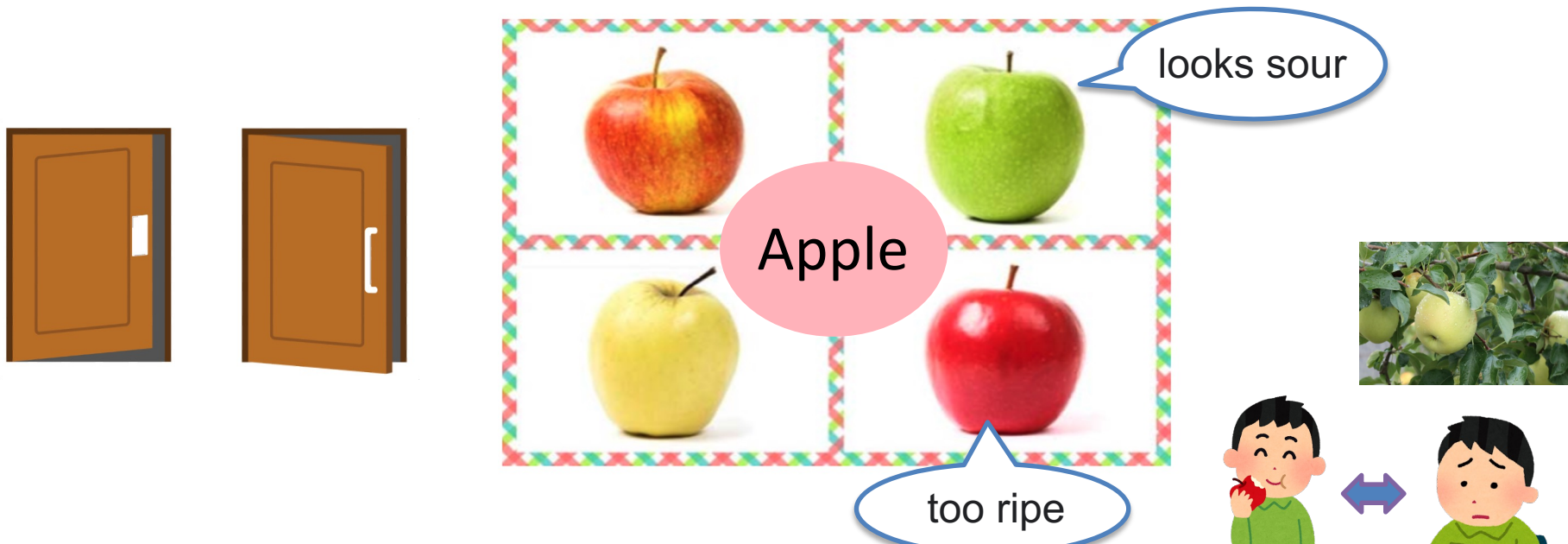
# Agent Network Architecture(ANA)

- Advantage
  - Coexistence of immediacy and deliberateness is possible
- Disadvantage
  - The number of agents is vast when the environment is large



# Affordance

- Animals perceive affordances transmitted by objects



→ Affordances allow animals to **reduce wasteful search**

# Affordance Network Construction Procedure

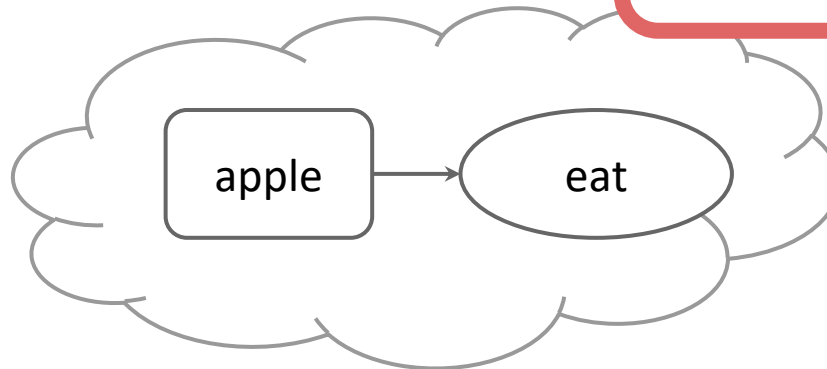


1. Sentence generation with GPT-3

Q : Story about an apple.      A : He ate an apple.



2. Parsing with CoreNLP  
3. Networking dependencies



# Large Language Model (LLM)

The performance of a language model is highly dependent on the size of the model

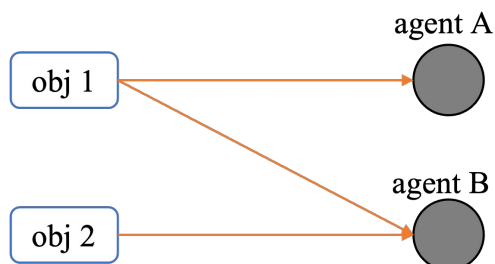


LLM, with many parameters and a large amount of training data, can be used for various tasks and is called Foundation Models

# Proposed Method

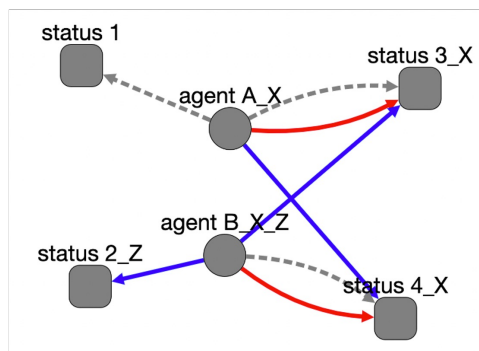
- Extending the use of affordances to address ANA issues

(a)Affordance Network



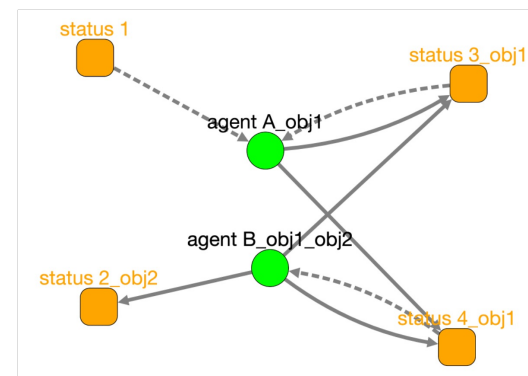
A knowledge graph expressing the relationship between a noun (object) and the verb recalled from that noun

(b)Variable Agent Behavior Network



ANA's network of agents and states with noun parts as variables

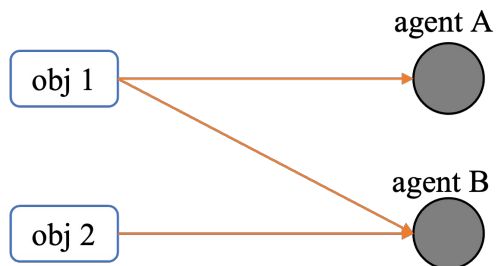
(c)Agent Behavior Network



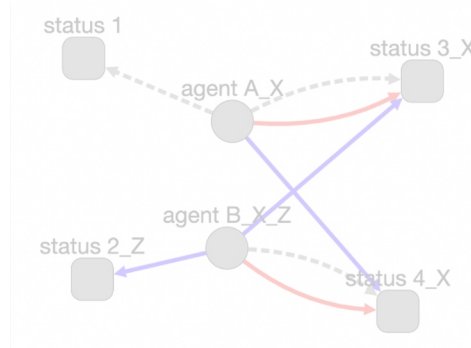
A network with a concrete object assigned to it

# Proposed Method

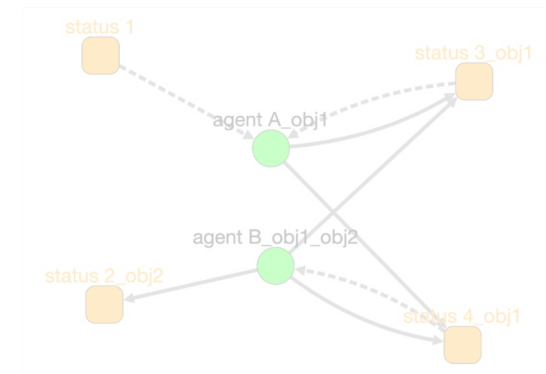
(a)Affordance Network



(b)Variable Agent Behavior Network



(c)Agent Behavior Network



A knowledge graph expressing the relationship between a noun (object) and the verb recalled from that noun

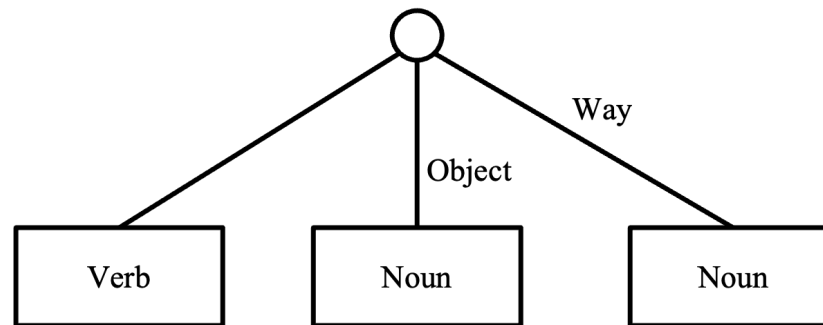
# Proposed Method

- Affordance Network

- Extracting affordances from the relationship between verbs and nouns in sentences output by LLM

- Verb-Object
- Verb-Way

Common sense and tacit knowledge are contained within.



Empty Node  
○: Action Node

# Affordance Network Construction Procedure

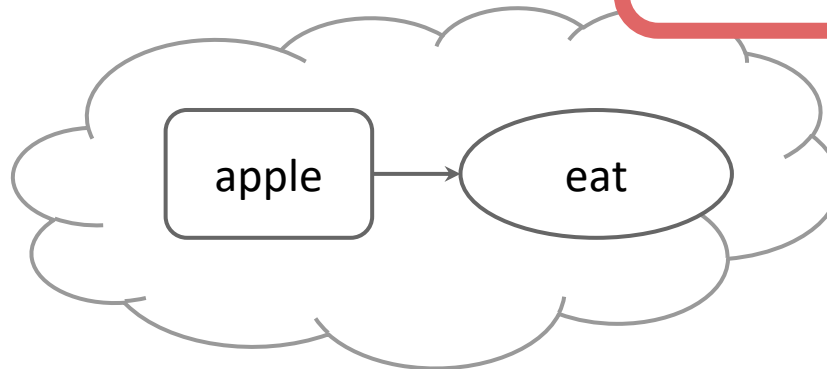
  
**GPT-3**

1. Sentence generation with GPT-3

Q : Story about an apple.      A : He ate an apple.

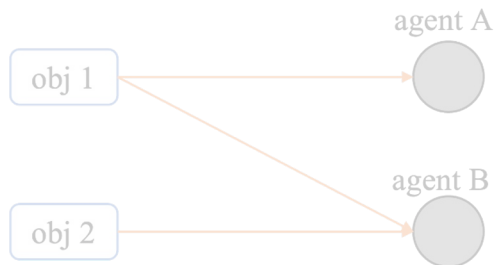


2. Parsing with CoreNLP  
3. Networking dependencies

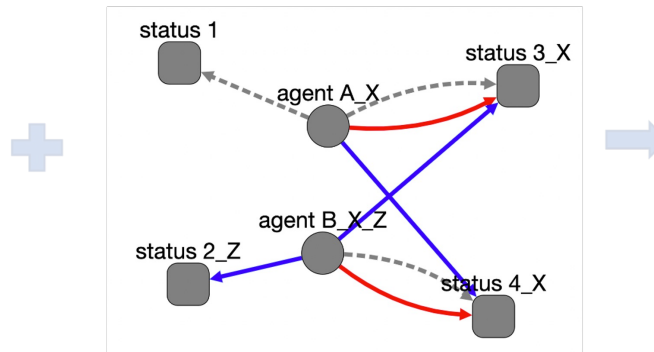


# Proposed Method

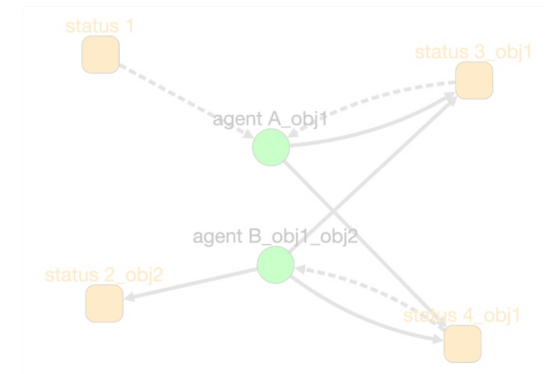
(a)Affordance Network



(b)Variable Agent Behavior Network



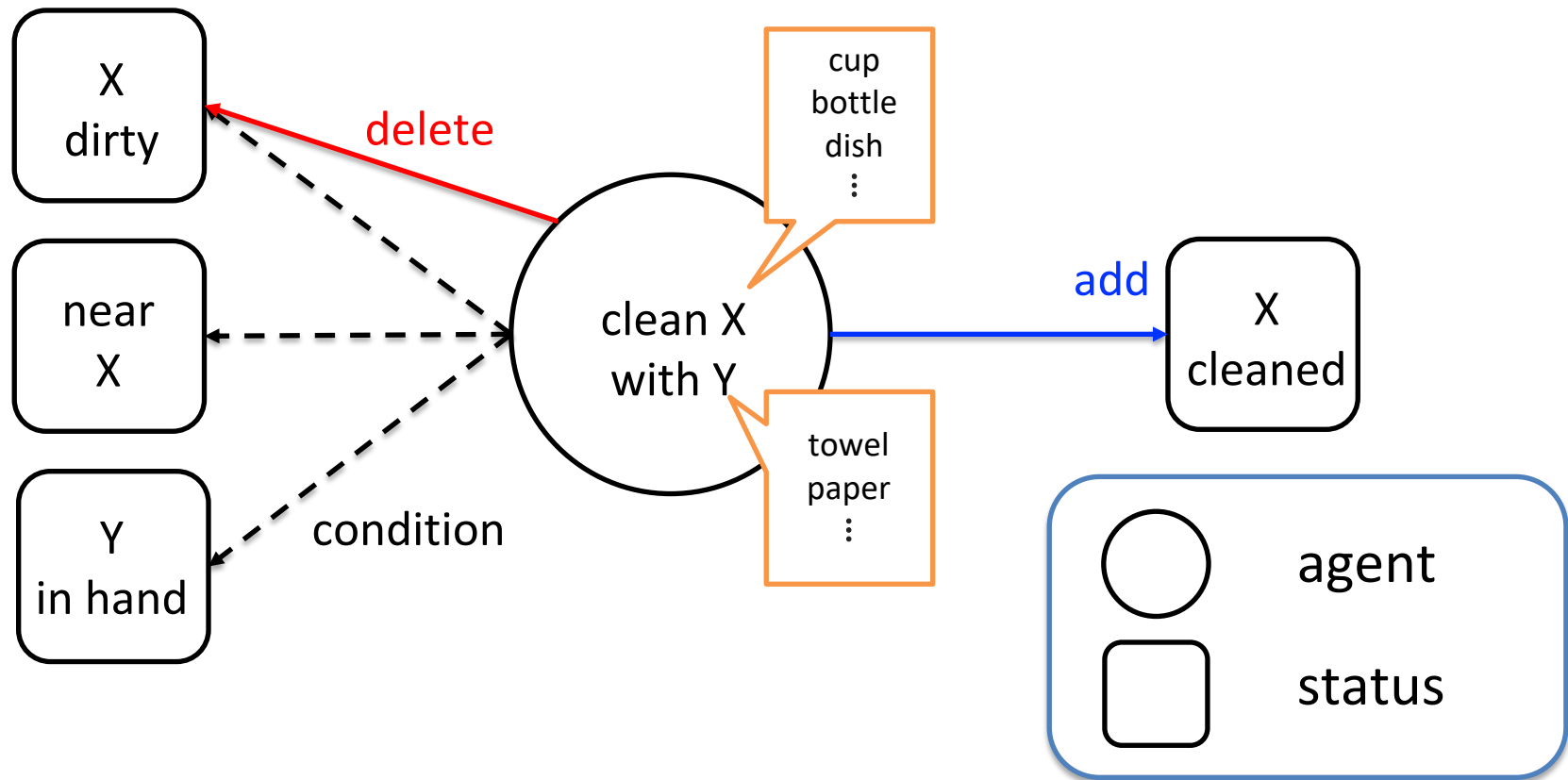
(c)Agent Behavior Network



ANA's network of agents  
and states with noun parts  
as variables

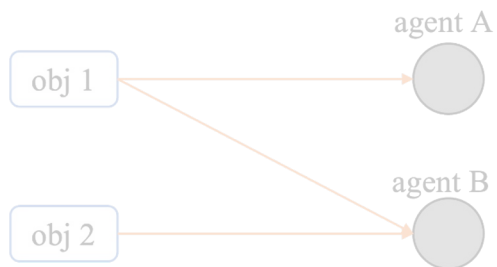
# Variable Agent Behavior Network

- Variable ANA agents and statuses

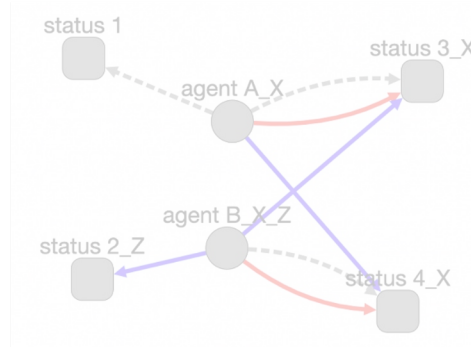


# Proposed Method

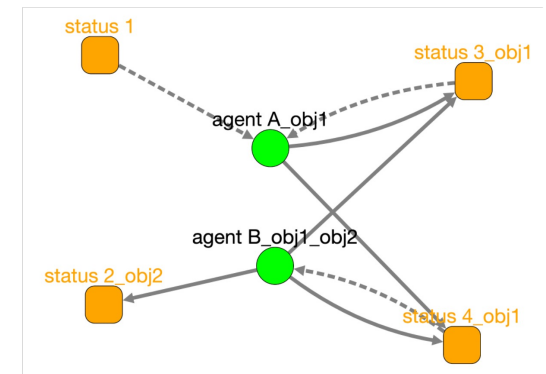
(a)Affordance Network



(b)Variable Agent Behavior Network



(c)Agent Behavior Network

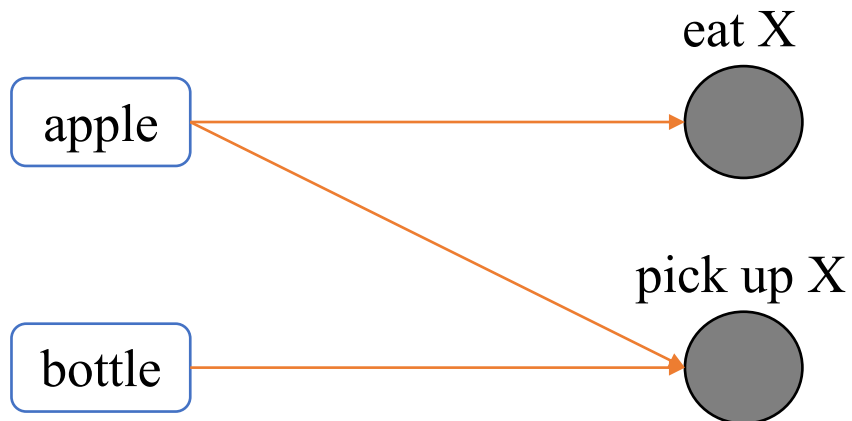


A network with a concrete object assigned to it

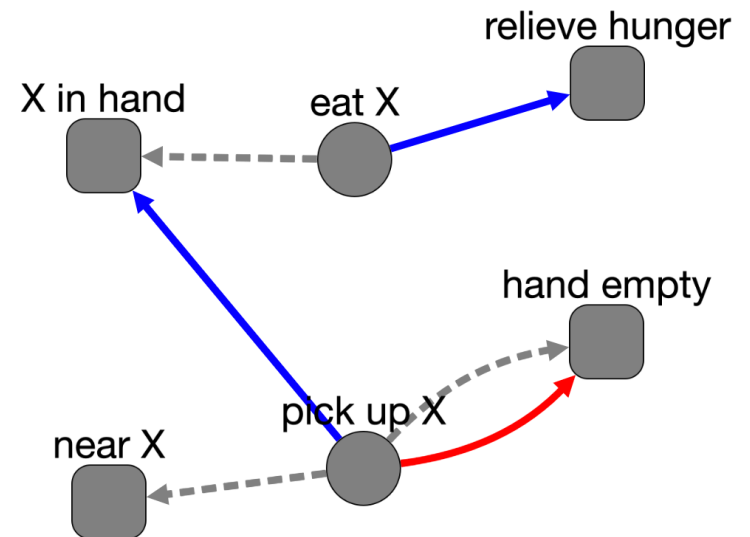
# Proposed Method

- Use of Affordances

Affordance Network

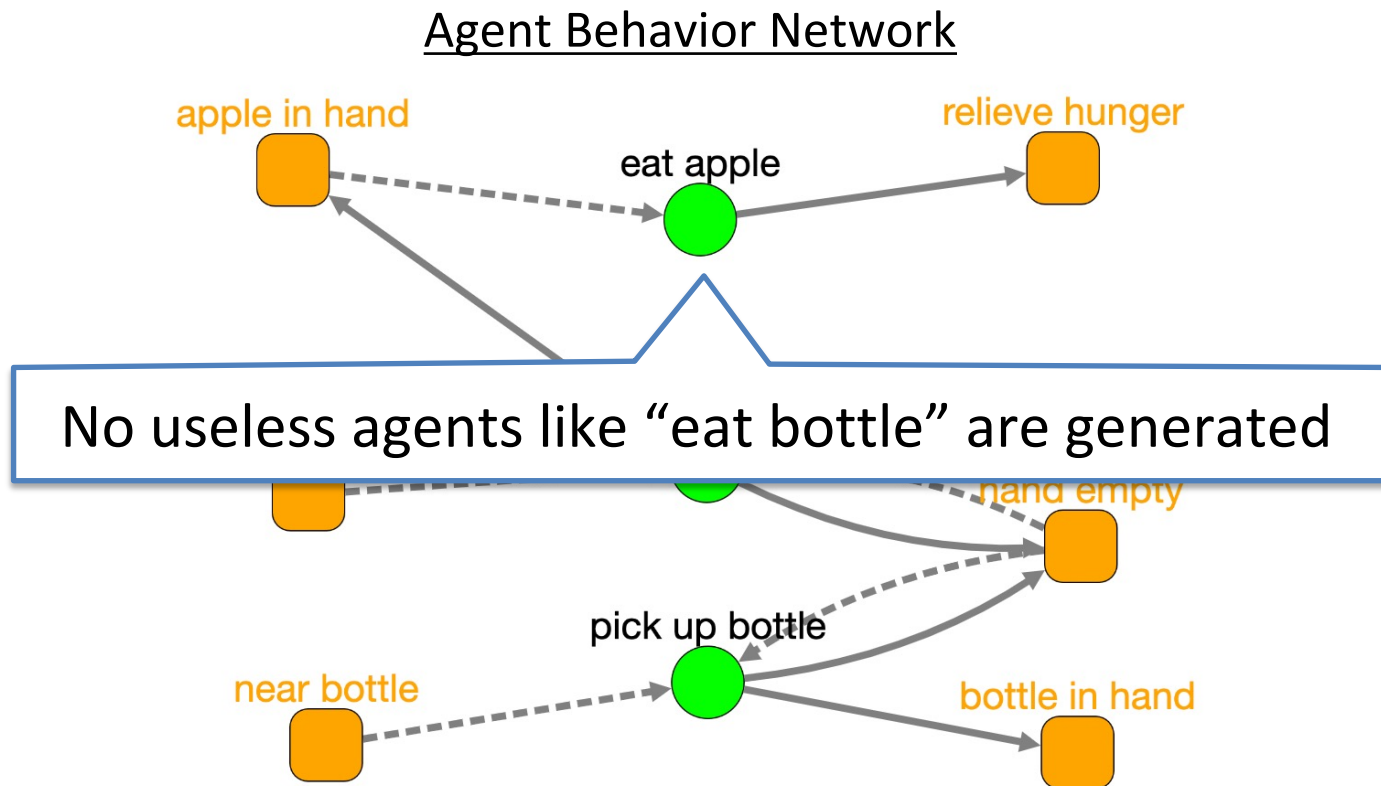


Variable Agent Behavior Network



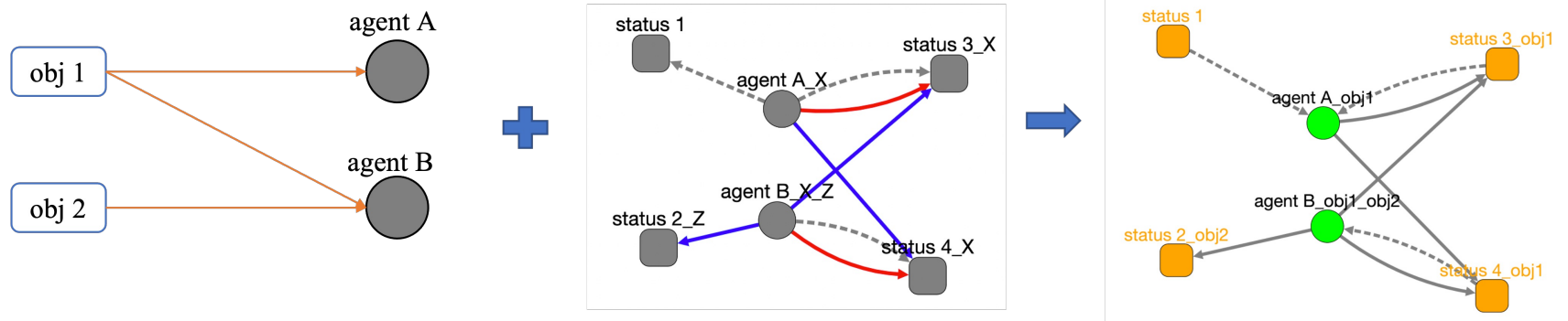
# Proposed Method

- Use of Affordances



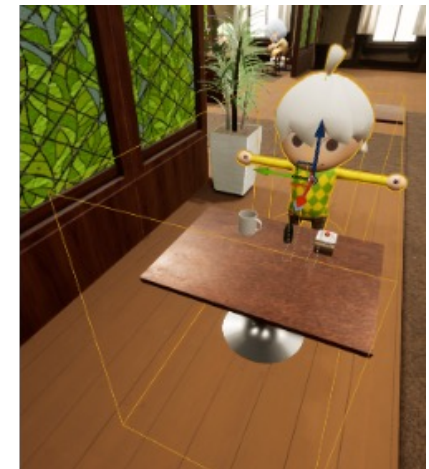
# Experimental Objective

- To verify whether affordance can be applied to seek a sequence of actions to achieve goals efficiently



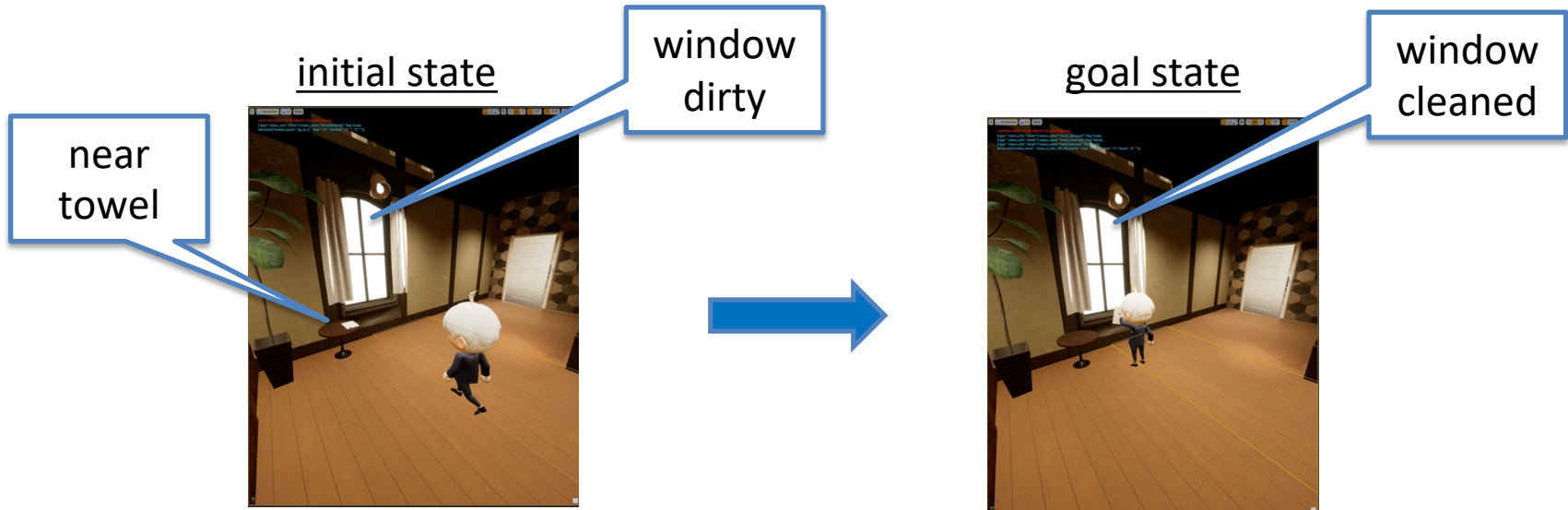
# Experimental Environment

- Unreal Engine
  - Cafe 3D simulation
- Character
  - Actuators (execution of actions)
  - Sensors (acquisition of perceptual information)



# Experimental Scenario

- Nouns used
  - towel, window
- initial state and goal state



# Experimental Scenario

- Action sequences required to achieve the goal



go to window



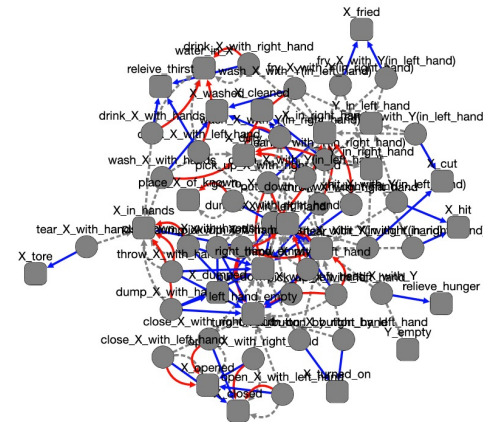
pick up towel



clean window with towel

# Data used in experiment

- Variable Agent
  - 35 kinds
- Affordance Network

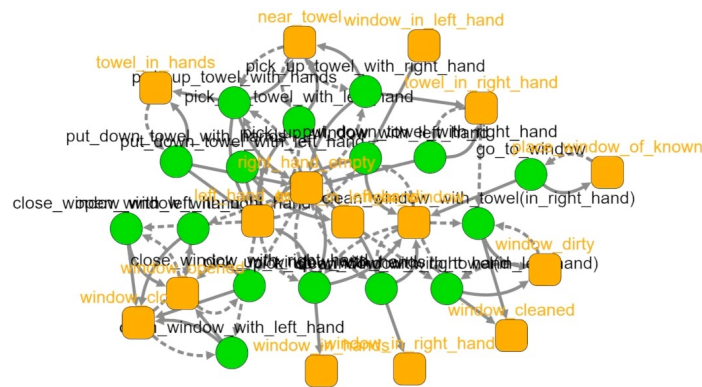


Data	個数
Verb Node	1,476
Noun Node	3,924
Verb-Object link	13,218
Verb-Way link	1,521



# Results

- Video of UE in action (With Affordances)



→ Properly executed actions as directed by the planner

# Conclusion and Future Work

- Conclusion:
  - The use of affordances in MAP was effective
- Future Work:
  - Use of affordances that take into account the attributes and positional relationships of objects  
-> Enables more situationally appropriate agent generation