

AIT-Rescue

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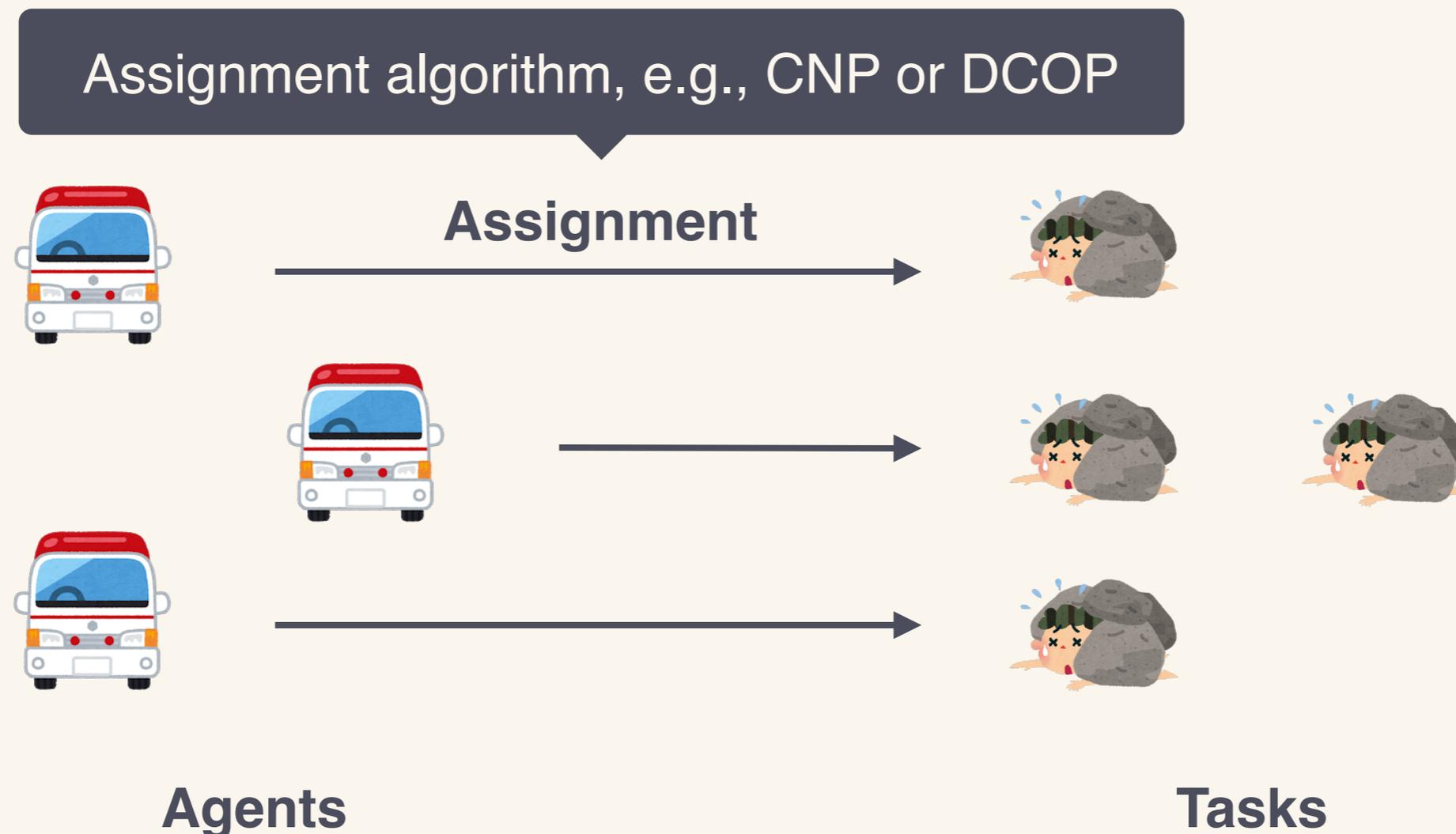
Aichi Institute of Technology / Aichi University

Agenda

- ① The DCOP framework is useful for solving task assignment problems.
- ② We propose an environment that makes it possible to use DCOP on RoboCupRescue Simulation (RRS).
- ③ We demonstrate simulations using our proposed environment & show their results.

Task Assignment Problem

- ▶ A typical problem in RRS.
- ▶ It is finding an approach to assign n agents to m tasks.



Task Assignment Problem

- ▶ A typical problem
- ▶ It is finding

**DCOP attracted attentions at AAI-18.
→ DCOP is standard on Multi-Agent**

Assignment algorithm, e.g., CNP or DCOP



Assignment



Our proposal: an environment that allows DCOP to be used for task assignment problems on RRS.

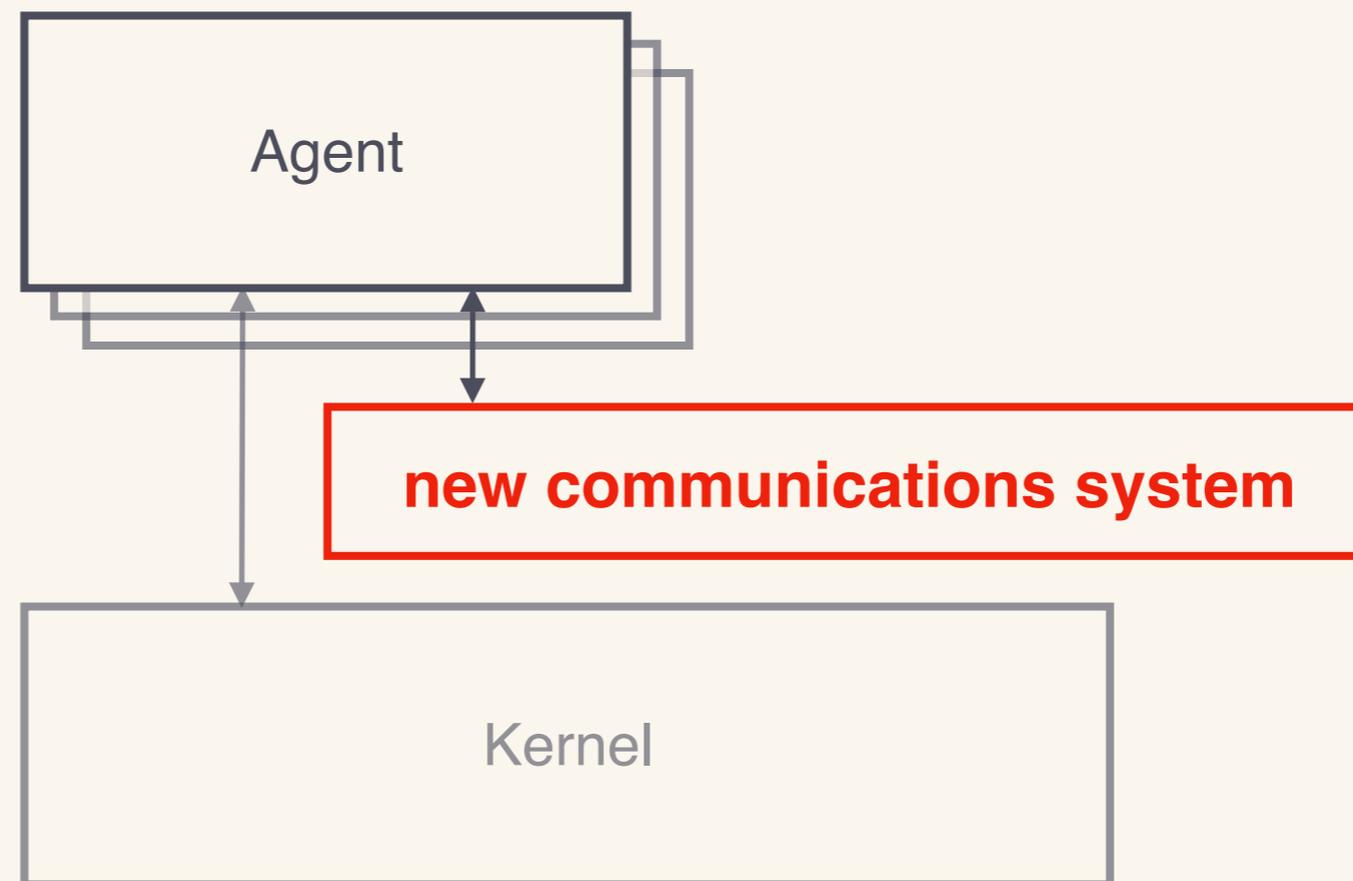


Agents

Tasks

Task Assignment Problem

- ▶
- ▶ **If we use a new communications system on the RRS system to allow the use of the DCOP...**



Importance of DCOP for RRS

- ▶ Task assignment problem features on RRS can be modeled as an extended DCOP.

RRS

- Multi-Objective
- Dynamic environment
- Partial observation
- Incomplete comm.
- Discrete time

DCOP Extensions

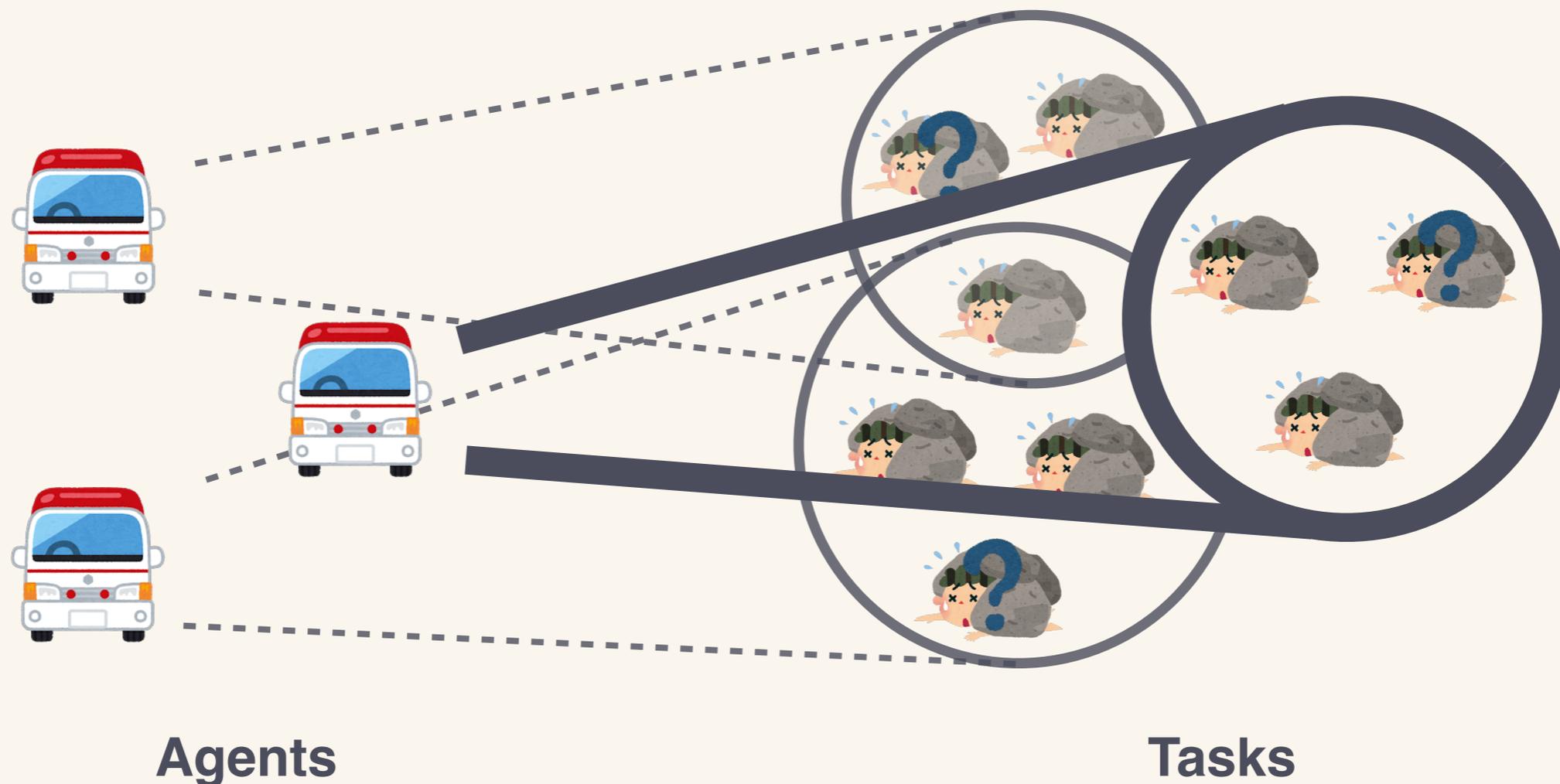
- Multi-Objective
- Dynamic
- Probabilistic

Open Problems

We can research new DCOP algorithm by combining these extensions on RRS.

Task Assignment Problem (Decentralized)

- ▶ When each agent performs task assignment individually, each agent selects a task from tasks in each scope.
- ▶ These kinds of problems can be modeled precisely as DCOP.



Definition of DCOP

Distributed Constraint Optimization Problem

- ▶ The problem of determining a combination of variable values that maximize utility.

Formalization

F. Fioretto, E. Pontelli, and W. Yeoh: "Distributed Constraint Optimization Problems and Applications: A Survey"

$$DCOP = \langle \mathbf{A}, \mathbf{X}, \mathbf{D}, \mathbf{F}, \alpha \rangle$$

$\mathbf{A} = \{ a_1, \dots, a_n \}$: a set of agents.

$\mathbf{X} = \{ x_1, \dots, x_m \}$: a set of variables.

$\mathbf{D} = \{ D_1, \dots, D_m \}$: a set of ranges for each variable.

$\mathbf{F} = \{ f_1, \dots, f_k \}$: a set of utility functions (constraints).

$\alpha : \mathbf{X} \rightarrow \mathbf{A}$: a mapping function.

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$$\alpha : \mathbf{X} \rightarrow \mathbf{A}$$

Selects a value
per variable

Maximizes
a utility



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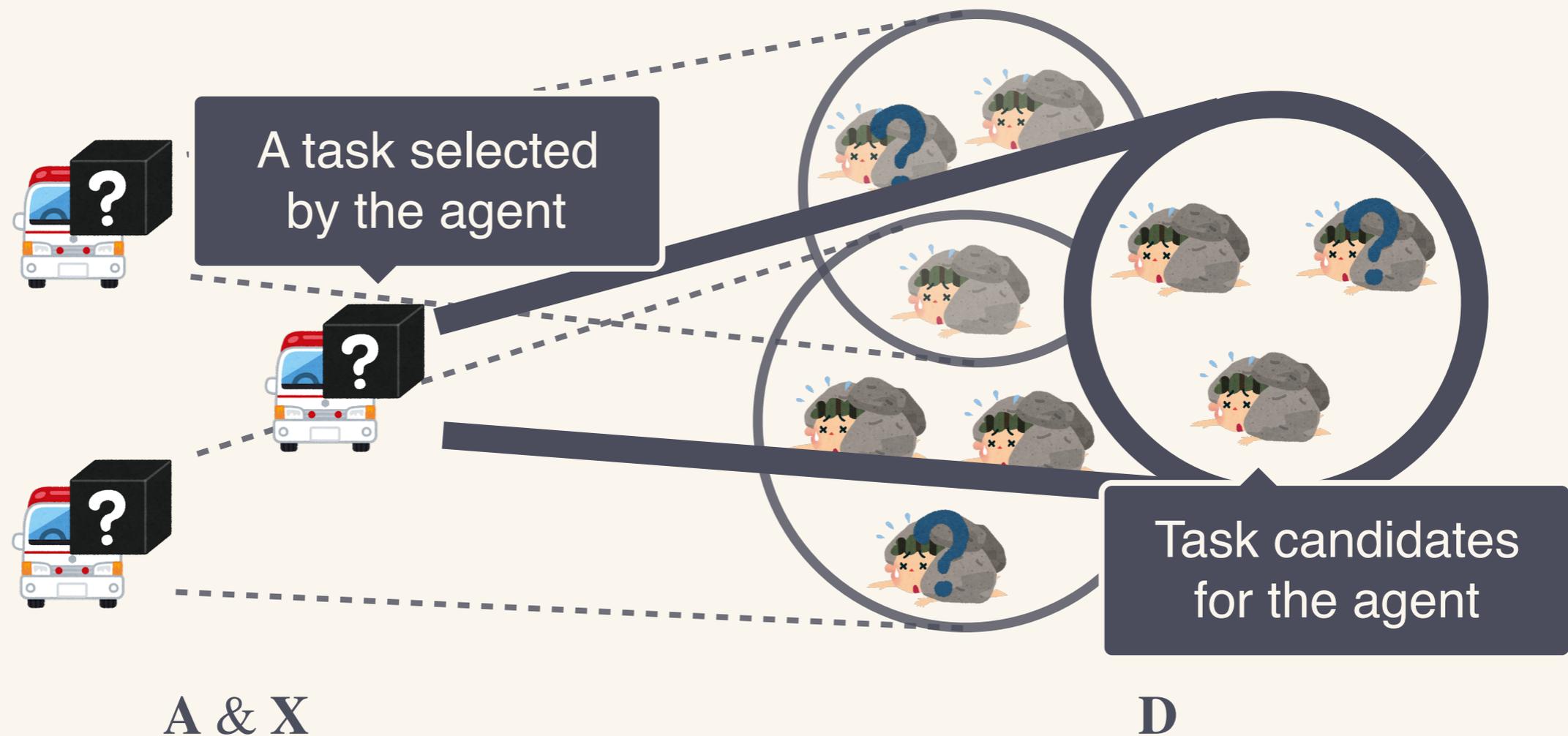
$$\alpha : \mathbf{X} \rightarrow \mathbf{A}$$

Allowing an agent to control 1/more variables;
can add attributes & abilities to variables
such as location & communication.

※ In most cases, each agent controls 1 variable.

Task Assignment Problem on DCOP

- ▶ A task assignment problem can be modeled as a DCOP.

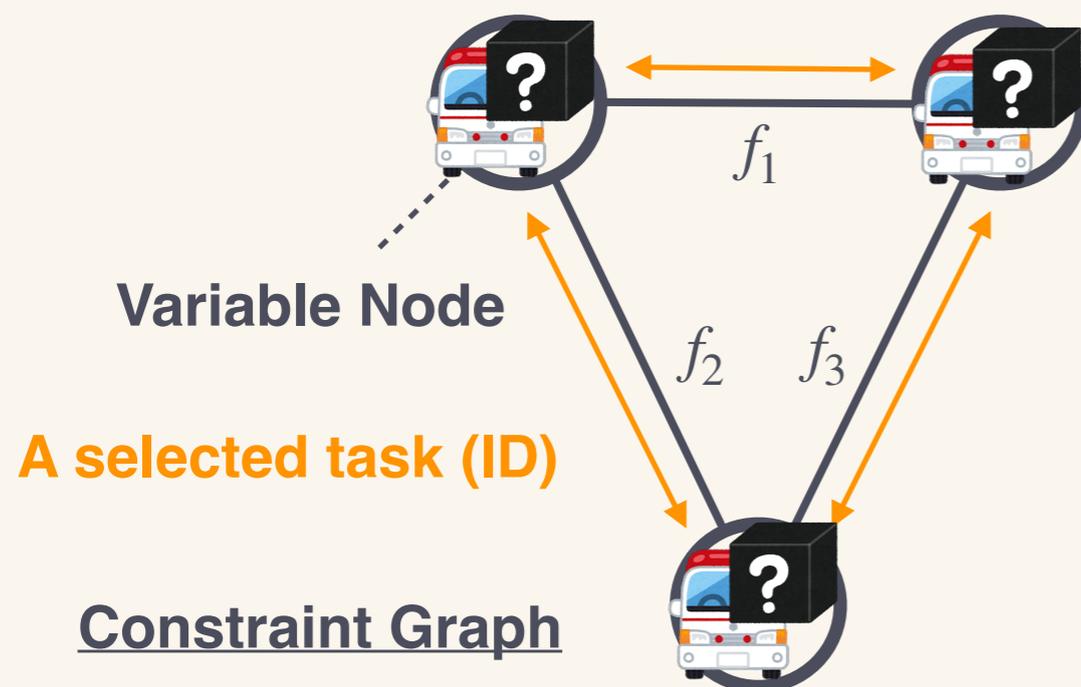


Two Major Real-Time Algorithms for DCOP

- ▶ DCOP algorithms work on a graph that represents a problem.
- ▶ Each node selects a value by communicating with the other nodes.

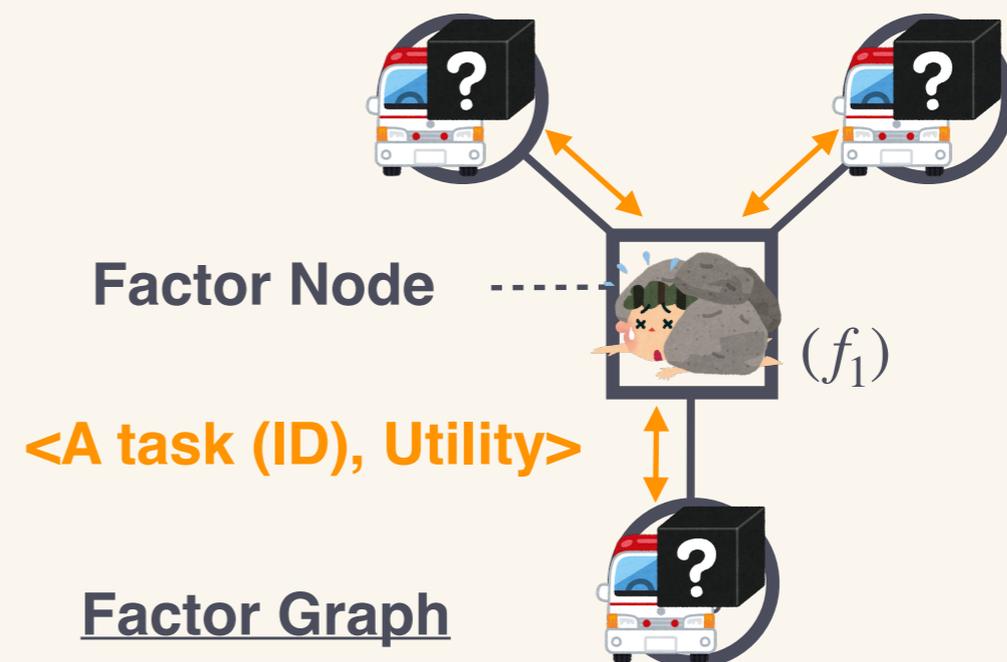
DSA

Uses values of other nodes
& random numbers.



Max-Sum

Uses value & its utility pairs
that were calculated on other nodes.

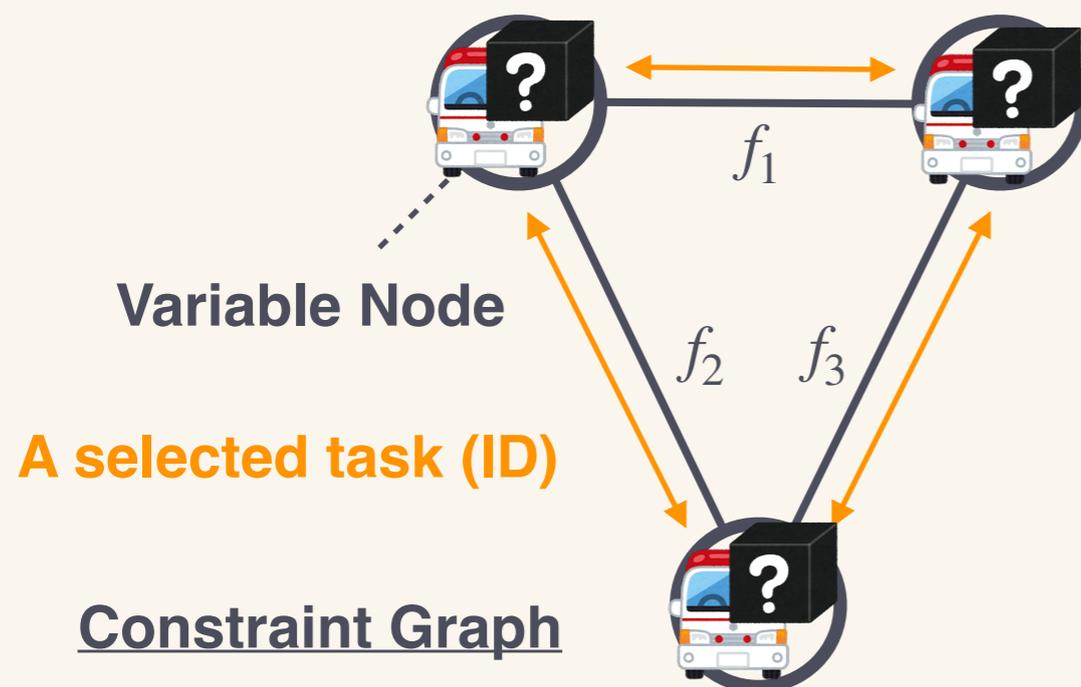


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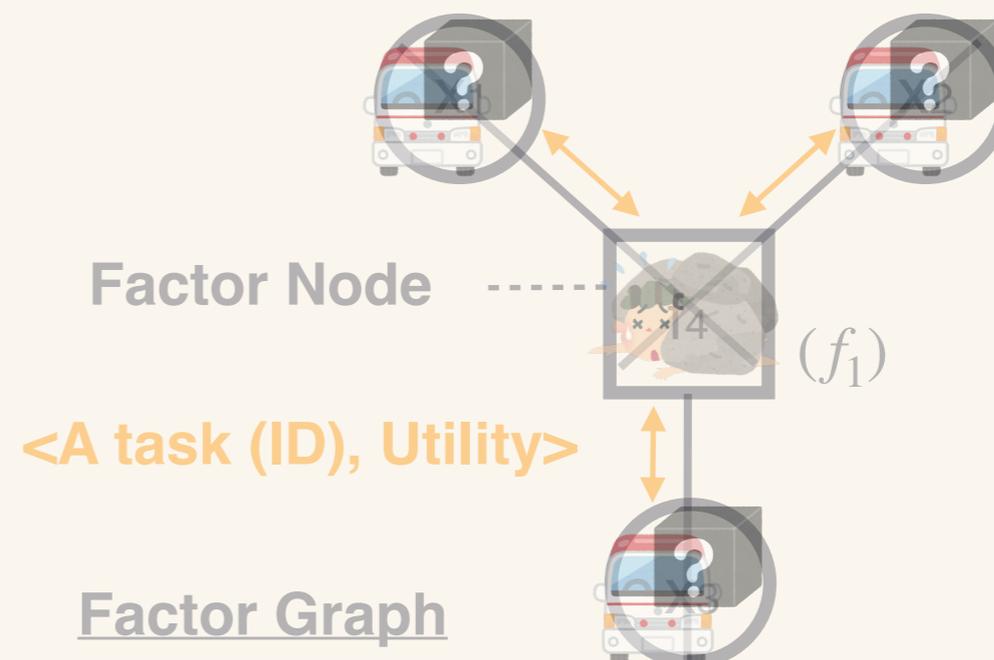
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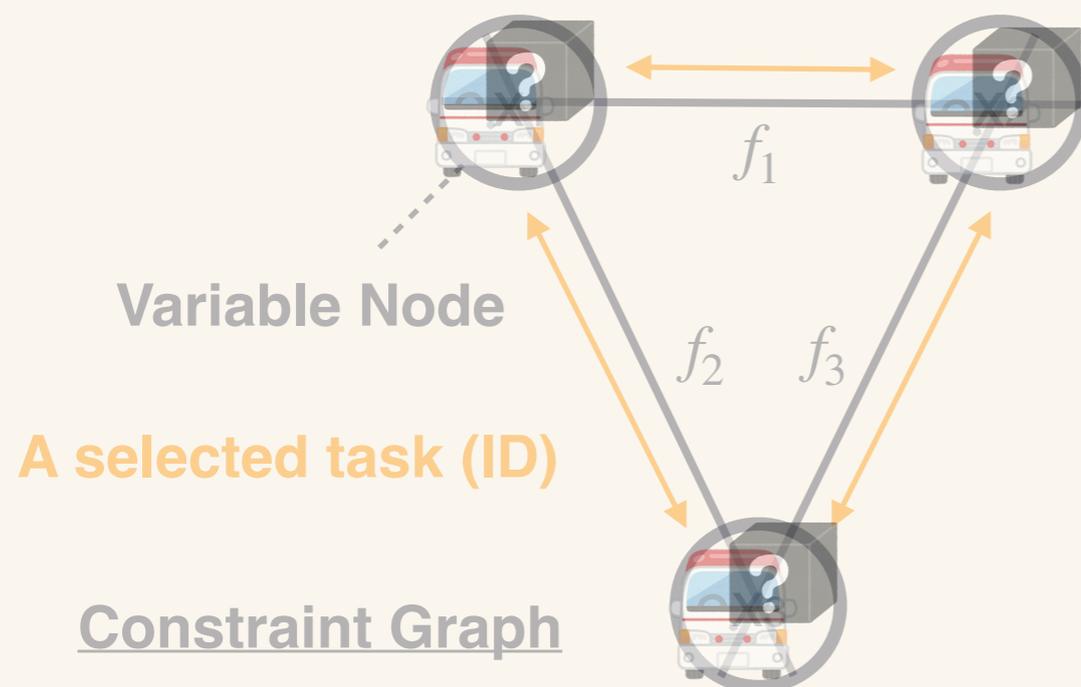


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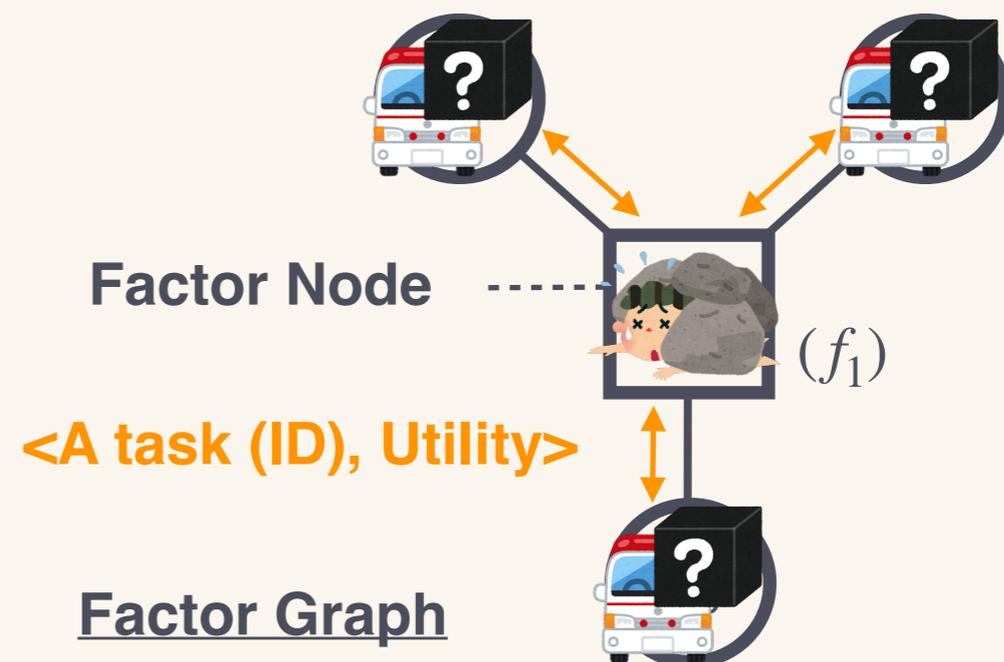
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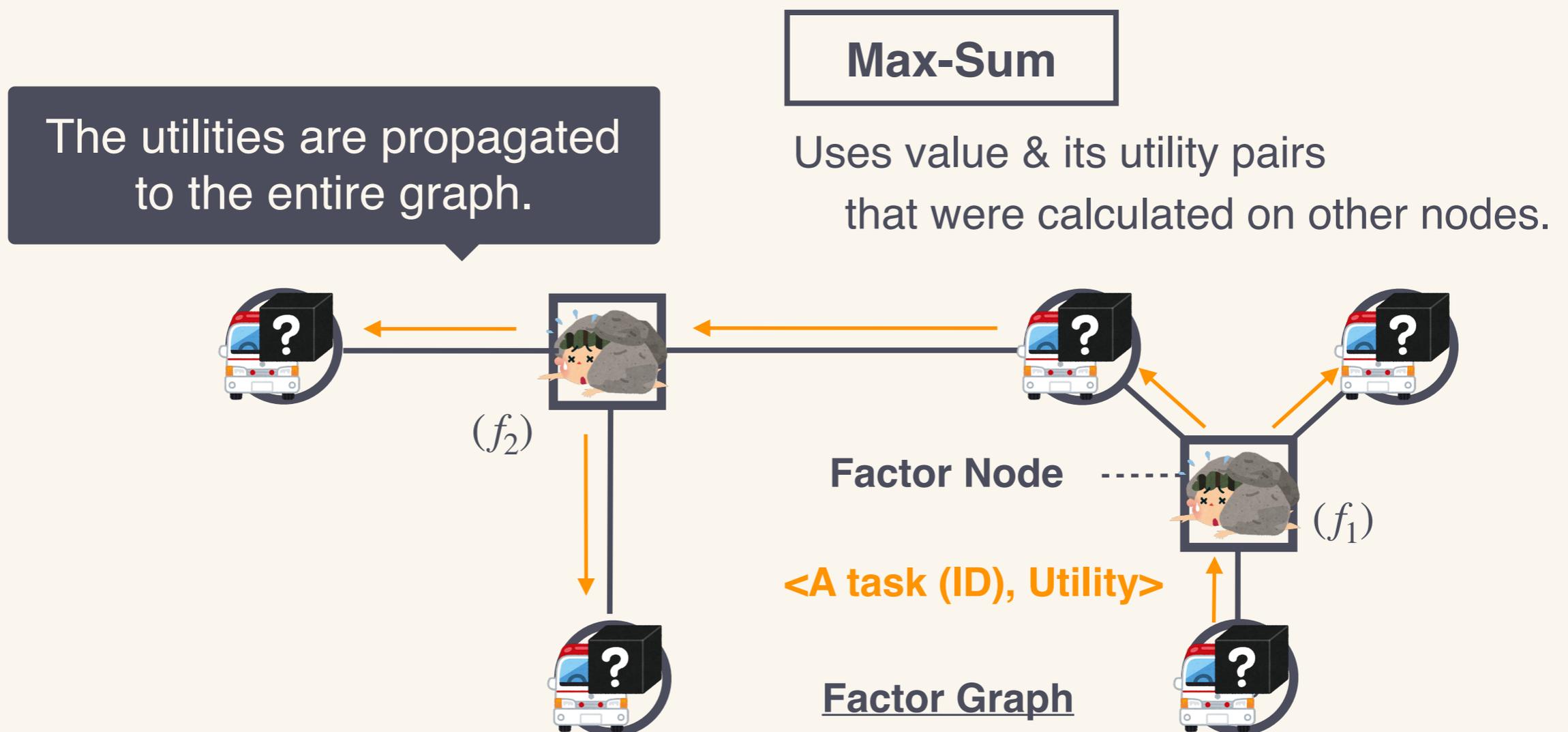
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The utilities are propagated to the entire graph.

Max-Sum

Uses value & its utility pairs that were calculated on other nodes.

DCOP algorithms need to communicate repeatedly.
→ This is difficult for the RRS system.



<A task (ID), Utility>

Factor Graph



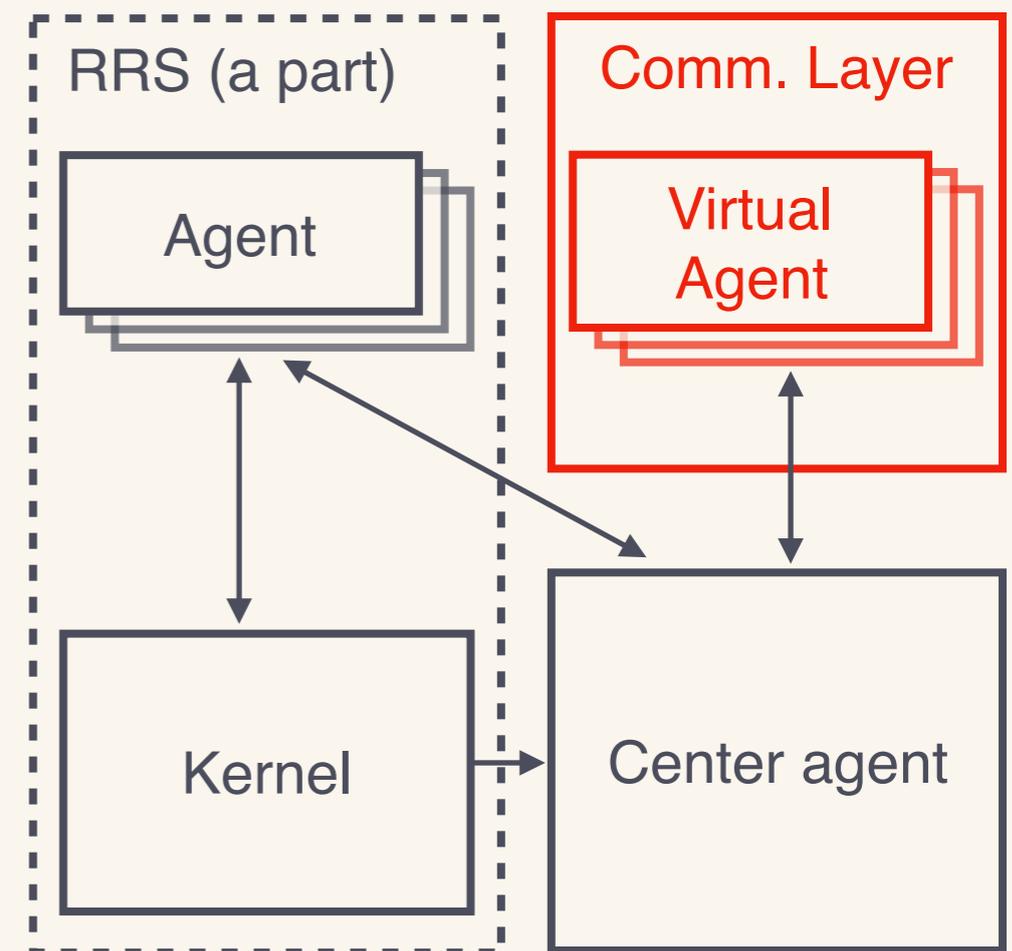
Related Research: RMASBench

- ▶ A benchmark system for DCOP algorithm using the RRS system.
- ▶ Its purpose was unclear to our community.
- ▶ It is incompatible with ADF.

- ▶ It introduced an independent unconstrained communication system on RRS.

DSA needs 60 communication round trips to solve a problem.
RRS system needs 2 steps for each round trip communication.

- It is difficult to utilize a DCOP algorithm in 200 - 300 steps.



Purpose

- ▶ We propose an environment in which DCOP algorithms can be used for RRS task assignment problems.
 - Our aim for this environment was to create a mechanism that follows the current RRS system as closely as possible.

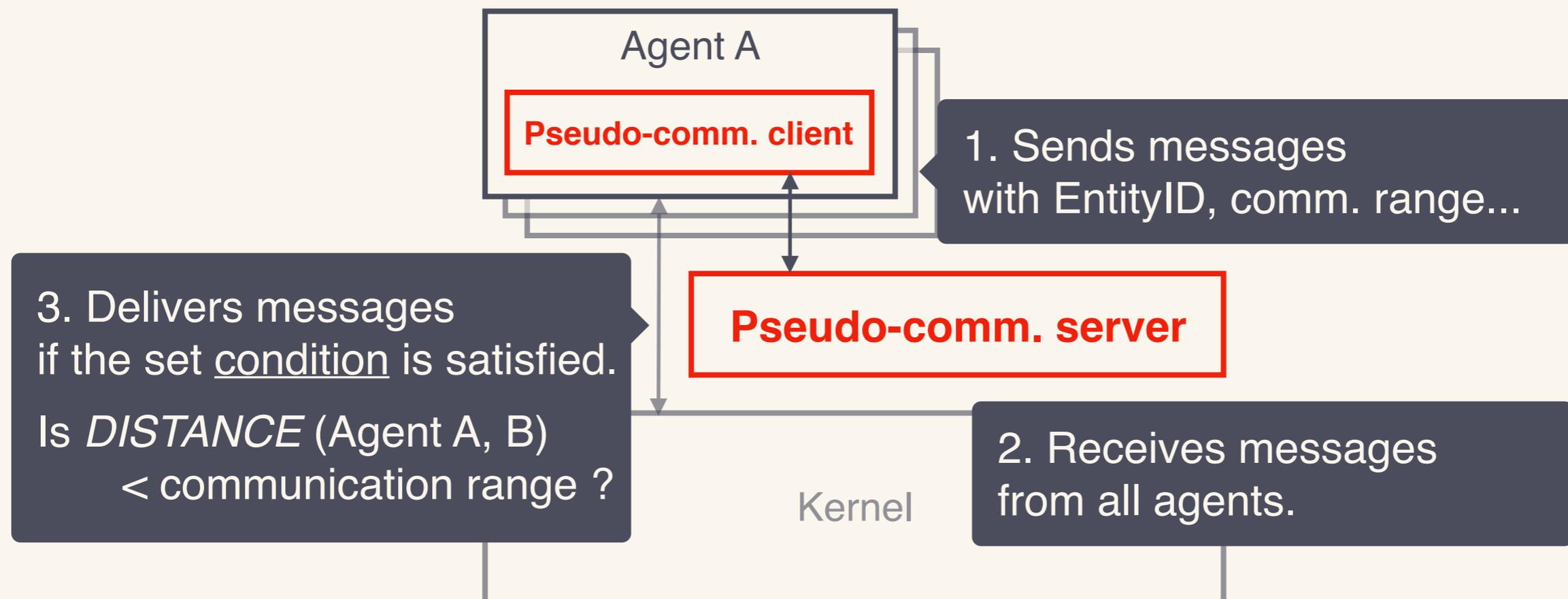


We can now implement the task assignment module on ADF.
→ We can also use other ADF modules.

However, we really need
a new communication system.

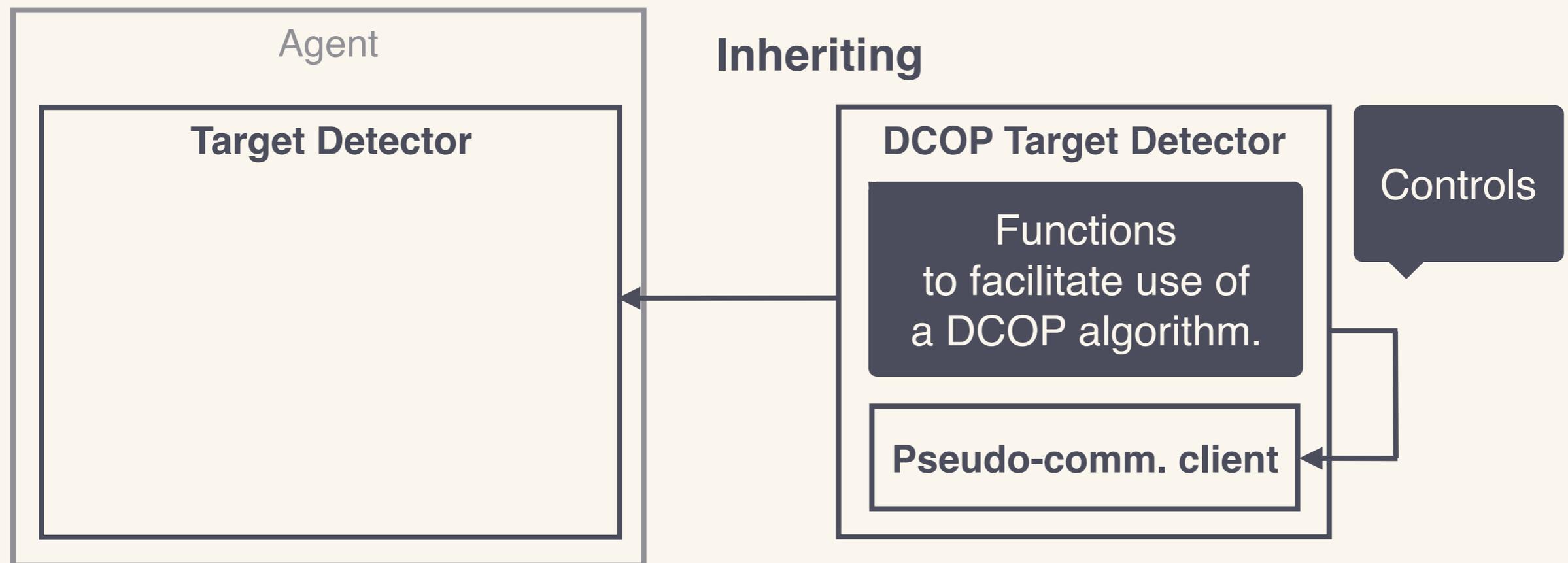
New Communication System

- ▶ DCOP algorithms need a communication system that is not restricted by the RRS discrete time.
- ▶ We introduce a new communication system using a server & clients.



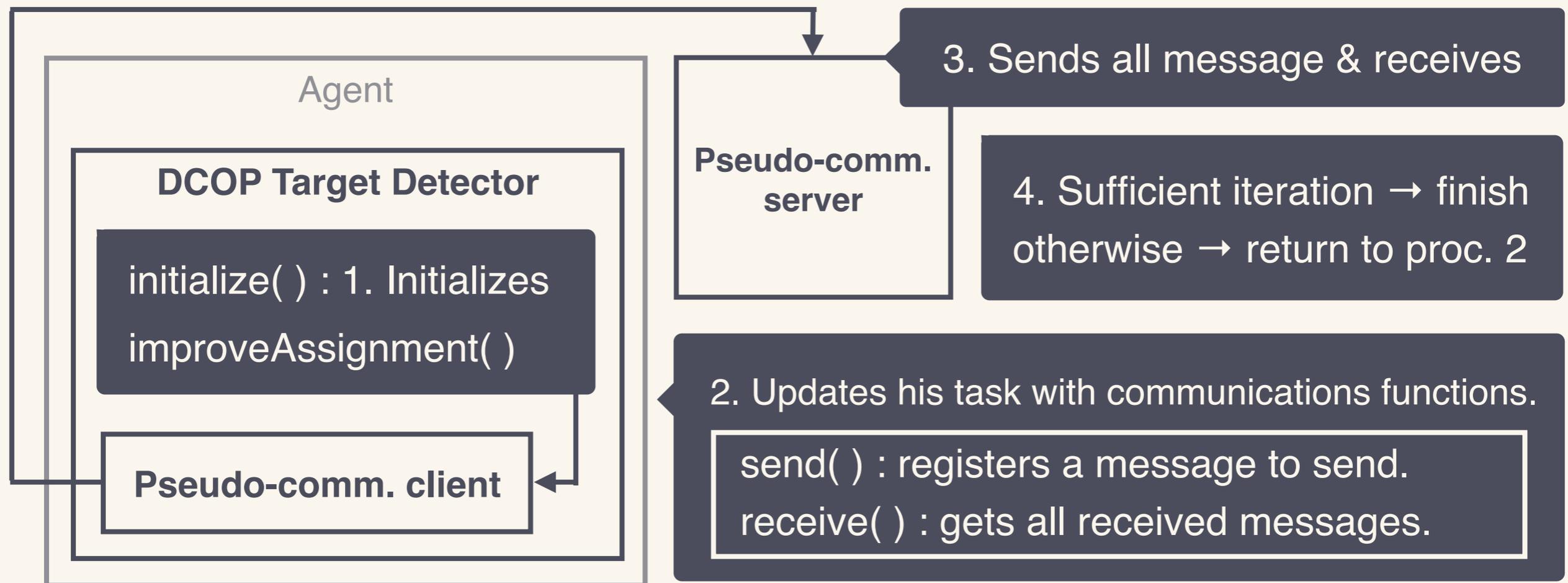
ADF Extension

- ▶ We design a task assignment module using the DCOP algorithm & the new communication system.



ADF Extension

- ▶ We design a task assignment module using the DCOP algorithm & the new communication system.



Procedures for running Max-Sum

- ① Implement a task assignment module using Max-Sum on our environment.
- ② We also implement Closest & Greedy to make comparisons.
- ③ Make them run simulations on some conditions (30 times / condition).

Agents : Only Ambulance Teams

Scenarios : VC3 / Eindhoven3 of RoboCup 2018

Agents are diverged / converged

Communication Range : the size of map x 1/4 or 2/4 or 3/4 or 4/4

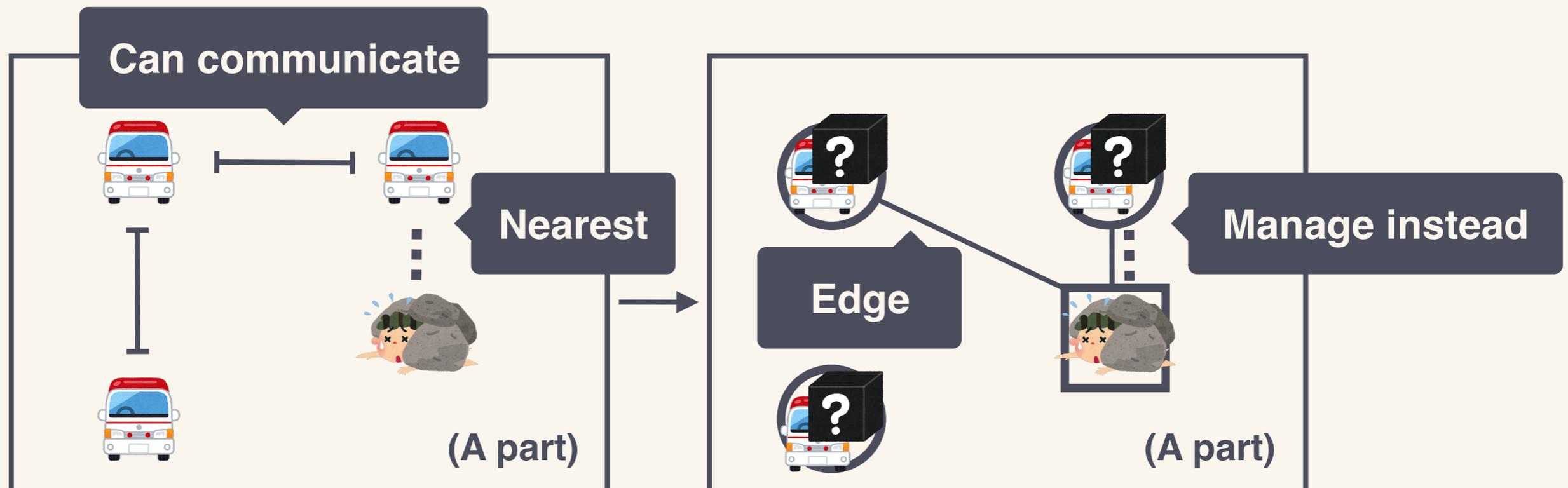
- ④ Compare simulation scores.

Implementing Max-Sum (Ambulance Team)

- ▶ We describe 3 important points to use Max-Sum.
- ① We had to make the module form a graph based on its environment & run the algorithm on each step to handle a dynamic environment.
- ② We need to design how to form the factor graph.
- ③ We also need to design the utility functions.

Forming a Factor Graph

- ▶ An agent (variable) is regarded as a variable node, a task related to a utility function is regarded as a factor node.
- ▶ Each agent manages own node → 1 graph is created by all agents.
- ▶ The nearest agent to each task also manages its factor node.
- ▶ An edge created when 2 agents communicate with each other.



Defining Utility Functions

- ▶ $\mathbf{F} = \{ f_1, \dots, f_k \}$: a set of utility functions.
- ▶ The purpose of Ambulance Teams is to rescue as many Civilians as possible.
- ▶ We designed the utility functions to **minimize** the movement costs & assign a sufficient number of agents to each task.

$$\begin{aligned}
 \text{Objective Function } \mathbf{F}_g(\mathbf{X}) &= \sum_{f \in \mathbf{F}} f(\mathbf{X}) \\
 &= \sum_{x \in \mathbf{X}} \underline{\text{COST of } x} + \sum_{d \in \text{UD}_i} \underline{\text{PENALTY of } d}
 \end{aligned}$$

Movement cost

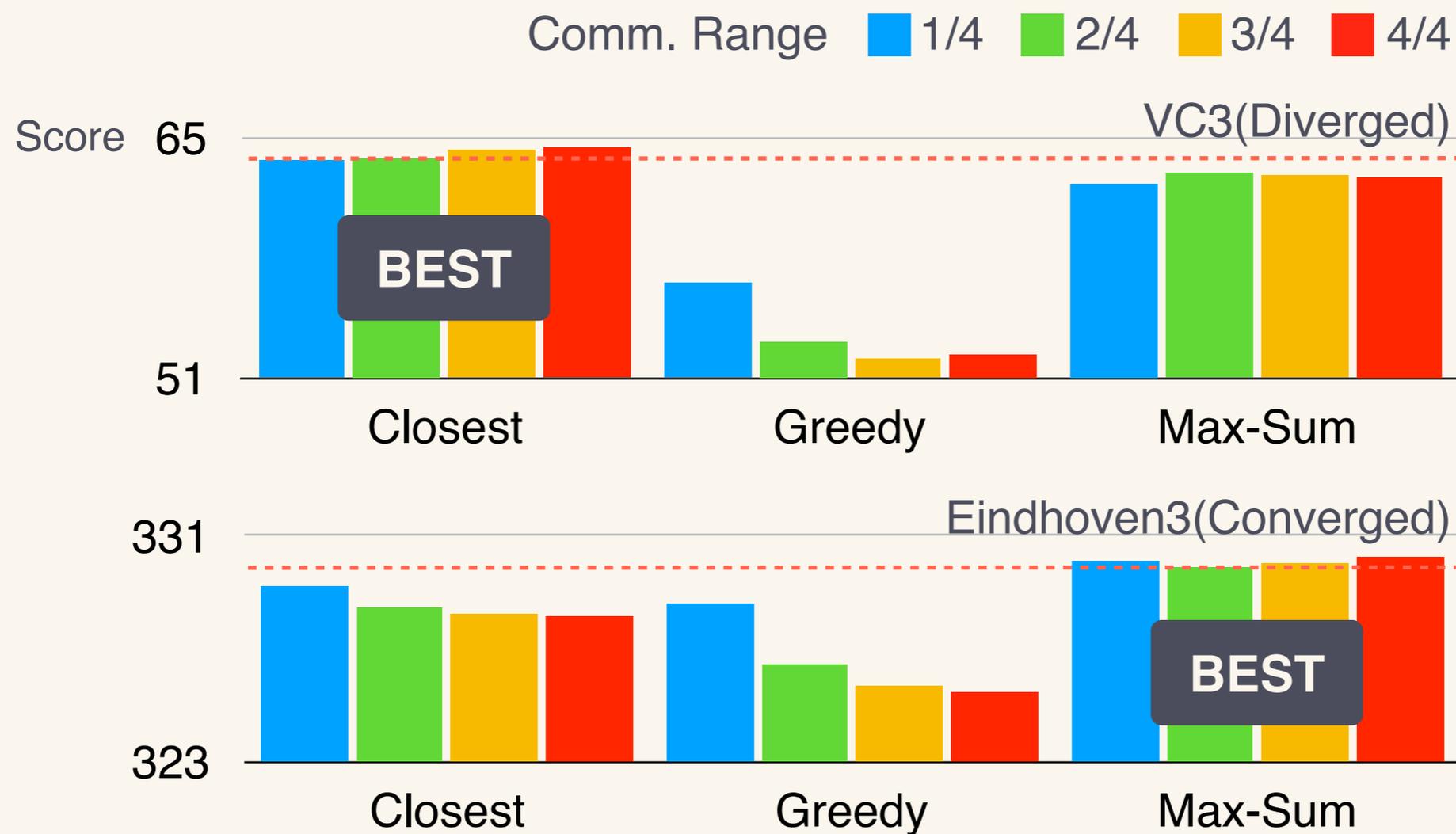
→ used in each variable node

Penalty arising from
an insufficient no. of agents

→ used in each factor node

Results

- ▶ Max-Sum can result stable & high scores.
- ▶ It even works well in scenario where agent have converged.



Conclusion

- ① DCOP framework is useful for solving task assignment problems.
- ② We proposed an environment for utilizing DCOP on RRS.
 - It introduced a new communication system.
 - It provides an ADF task assignment module to use a DCOP algorithm & the communication system.
- ③ We conducted simulations in our environment & showed their results.
 - Max-Sum could produce high, stable scores on RRS.

Contribution

- ▶ We provided an environment that allows DCOP algorithm utilization on RRS.
- ▶ We will research DCOP algorithms based on the following themes:



Implement new algorithms & improve utility functions

Fast Max-Sum, Bounded Max-Sum, etc...

Combine Multi-Objective & Dynamic DCOP

Test in environments containing some noise

Future Work

- ▶ We will attempt to allow the communication component of RRS kernel control the pseudo-communication server & work to make the server easier to configure & use.

