

The Size of Message Set Needed for the Optimal Communication Policy

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Background

- Multi-agent coordination with communication.

Main objective :

To find the optimal *action policy* δ^A and *communication policy* δ^M

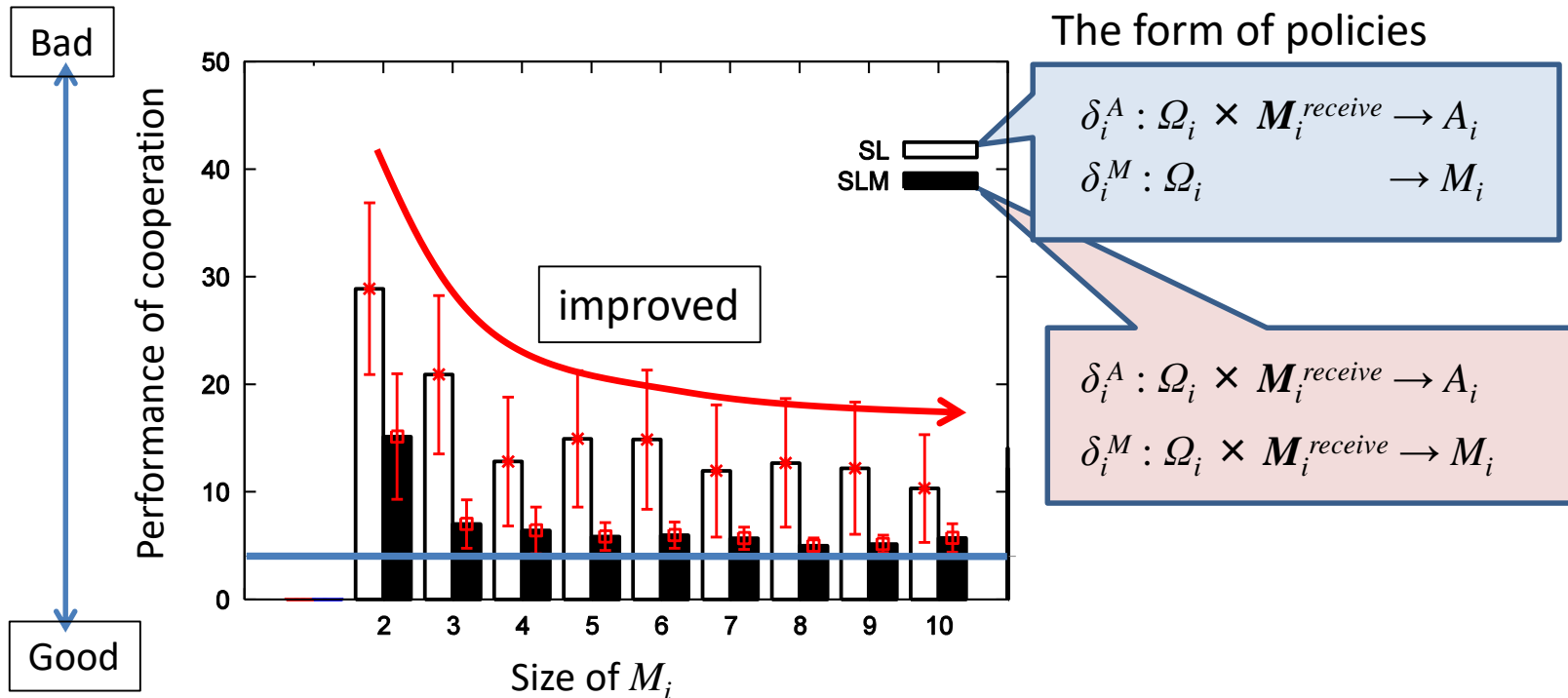
- We are interested in an approach based on autonomous learning.
- Definition of policies for agent i in our proposed methods

	Signal Learning (SL) [Kasai+ 08]	Signal Learning with Messages (SLM) [Kasai+ AAMAS09]
<i>Action Policy</i>	$\delta_i^A : \Omega_i \times M_i^{receive} \rightarrow A_i$ A set of observations (points to Ω_i) and A set of actions (points to A_i)	
<i>Communication Policy</i>	$\delta_i^M : \Omega_i \rightarrow M_i$ A set of received messages (points to $M_i^{receive}$)	$\delta_i^M : \Omega_i \times M_i^{receive} \rightarrow M_i$ A set of messages to send other agents (points to M_i)

(SL and SLM are based on Multi-Agent Reinforcement Learning framework)

Motivation

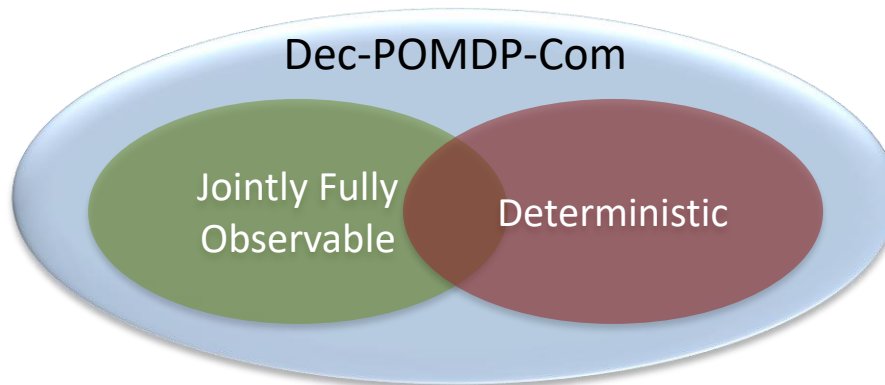
- Actual learning results of SL and SLM [Kasai+ AAMAS09]
- The performance of cooperation when the size of M_i is increases.



- We have an interest about how much size of M_i for constructing the optimal policy ?

Scheme of talk

- We show *minimum required sizes* $|M_i|$ for achieving the optimal policy for
 - Signal Learning on *Jointly Fully Observable Dec-POMDP-Com*
 - Signal Learning with Messages on *Deterministic Dec-POMDP-Com*



Outline

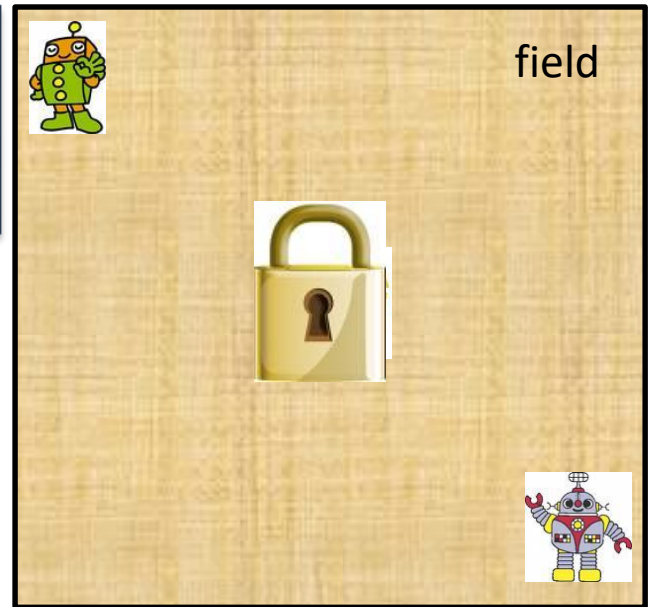
- Background
- Scheme of talk
- Review : Dec-POMDP-Com [Goldman+ 04]
- Constrained model
 - Jointly Fully Observable Dec-POMDP-Com [Goldman+ 04]
 - Deterministic Dec-POMDP-Com (we define)
- Theoretical analysis
- Conclusion

(Decentralized Partially Observable Markov Decision Process with Communication)

- A decentralized multi-agent system, where agents can communicate with each other and only observe the restricted information.

- Two agents get a treasure cooperatively.
- The treasure is locked.
- Both agents must reach the treasure at the same time to open the lock.

Example of model



(Decentralized Partially Observable Markov Decision Process with Communication)

- A decentralized multi-agent system, where agents can communicate with each other and only observe the restricted information.

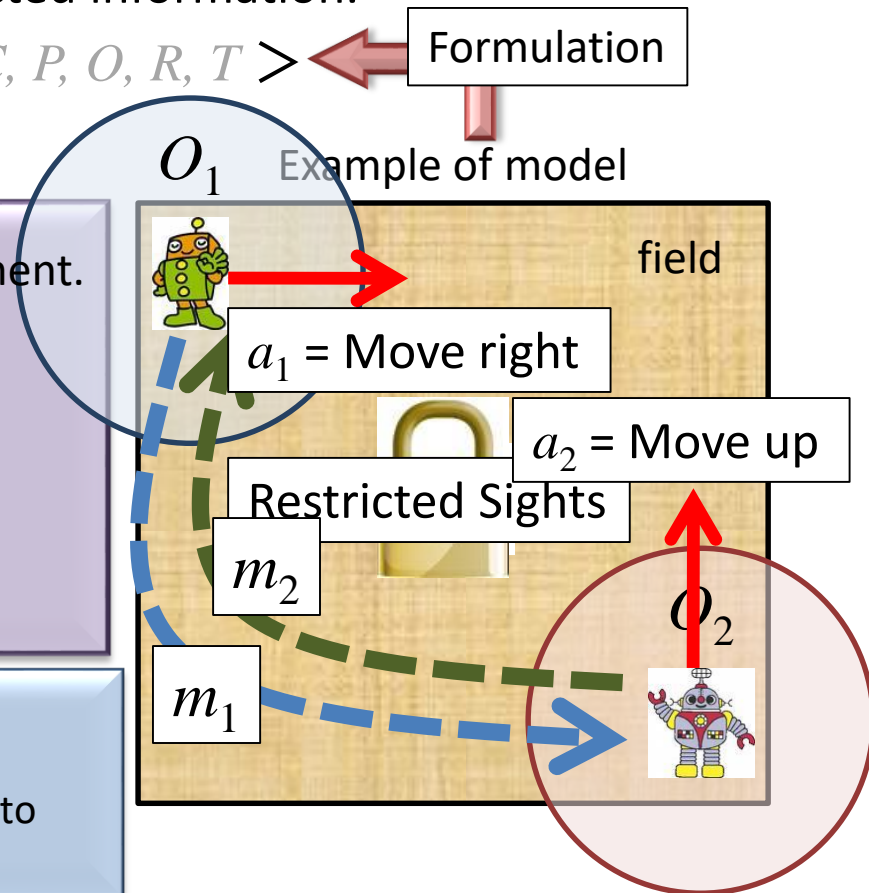
- Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$ ← Formulation

1step for agent i on Dec-POMDP-Com

1. Receive an observation O_i from the environment.
2. Send a message m_i to the other agents.
3. Perform an action a_i in the environment.

Repeat until both agent arrive at the treasure.

- Two agents get a treasure cooperatively.
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(Decentralized Partially Observable Markov Decision Process with Communication)

- A decentralized multi-agent system, where agents can communicate with each other and only observe the restricted information.

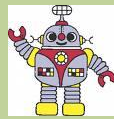
● Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$ Formulation

- A set of agents' indices

- e.g., $I = \{1, 2\}$



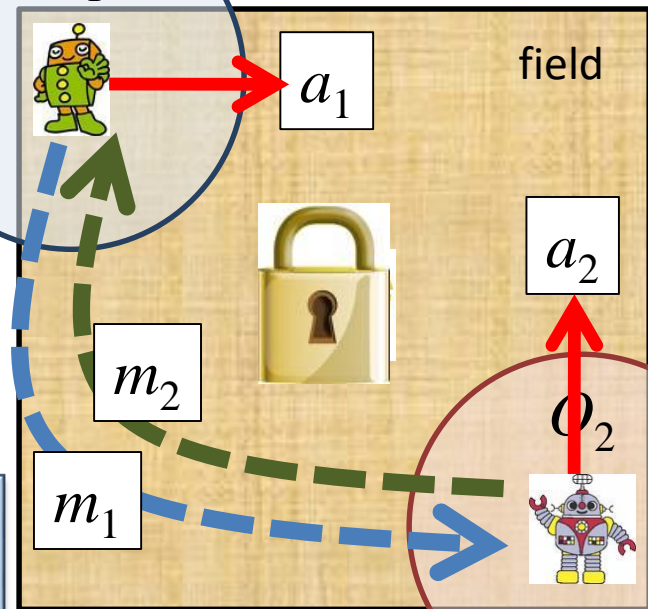
= 1



= 2

- Two agents get a treasure cooperatively.
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O_1 Example of model



(Decentralized Partially Observable Markov Decision Process with Communication)

- A decentralized multi-agent system, where agents can communicate with each other and only observe the restricted information.

● Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$

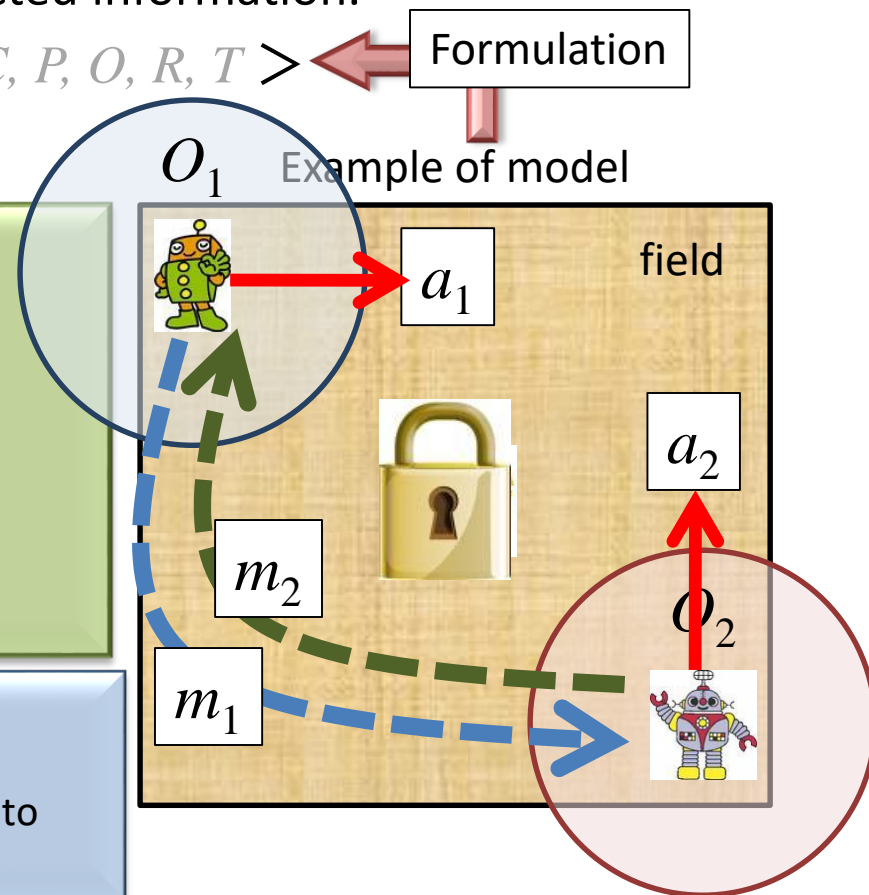
Formulation

- A set of global states

- e.g., $s = (\text{position of agent 1, position of agent 2, position of treasure})$

, $s \in S$

- Two agents get a treasure cooperatively.
- The treasure is locked.
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(Decentralized Partially Observable Markov Decision Process with Communication)

- A decentralized multi-agent system, where agents can communicate with each other and only observe the restricted information.

● Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$

Formulation

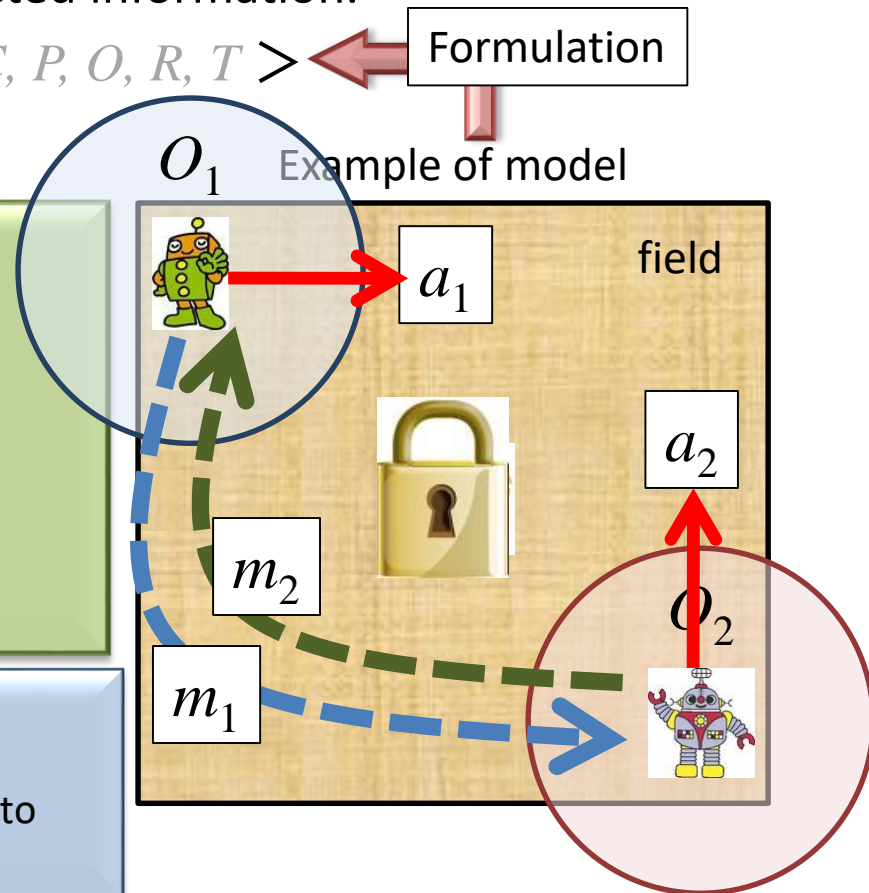
- Ω : a set of joint observations

- $\Omega = \Omega_1 \times \Omega_2$,
where Ω_i is a set of observations for agent i

- A : a set of joint actions

- $A = A_1 \times A_2$

- Two agents get a treasure cooperatively.
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(Decentralized Partially Observable Markov Decision Process with Communication)

- A decentralized multi-agent system, where agents can communicate with each other and only observe the restricted information.

● Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$

Formulation

- M : a set of joint messages

- $M = M_1 \times M_2$

- $C : M \rightarrow \mathcal{R}$ is a cost function

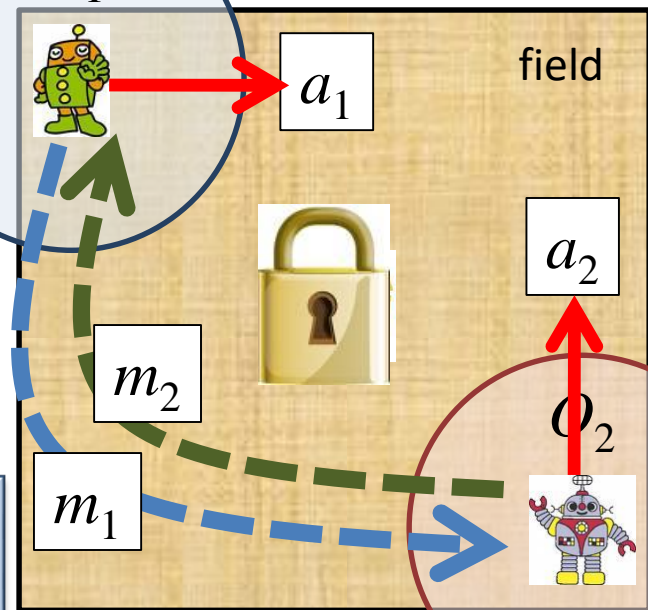
- $C(m)$ represent the total cost of transmitting the messages sent by all agents.

- Two agents get a treasure cooperatively.

- The treasure is locked.

- Both agents must reach the treasure at the same time to open the lock.

O_1 Example of model



(Decentralized Partially Observable Markov Decision Process with Communication)

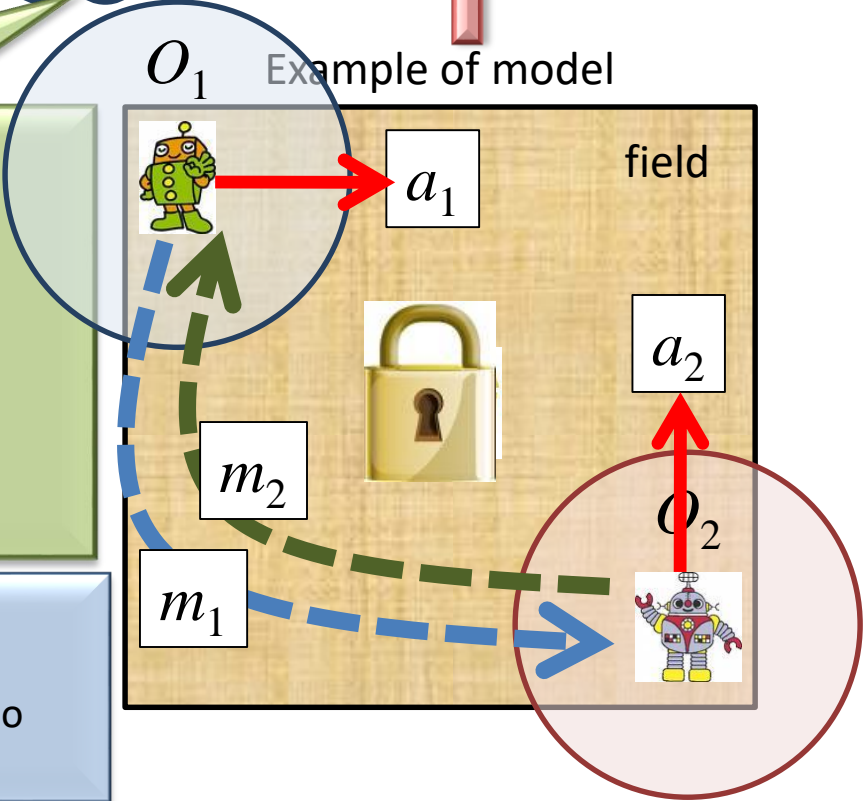
- A decentralized multi-agent system, where agents can communicate with each other and only observe the restricted information.

● Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$

Formulation

- P : a transition probability function
- O : an observation probability function

- Two agents get a treasure cooperatively.
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(Decentralized Partially Observable Markov Decision Process with Communication)

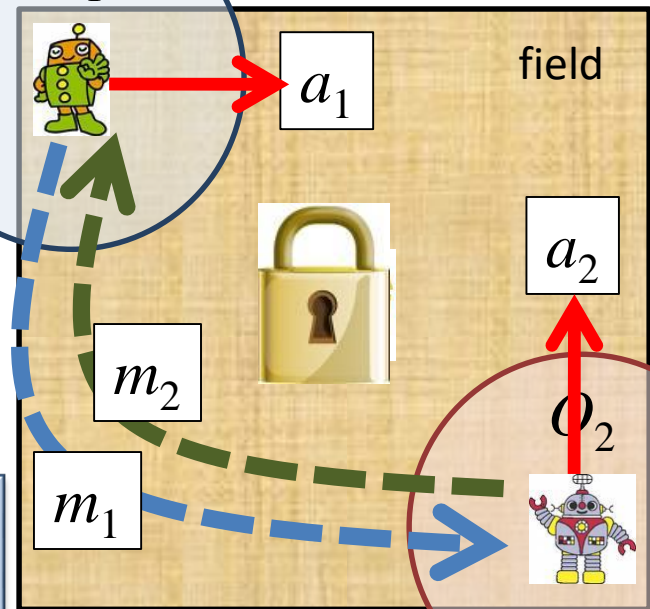
- A decentralized multi-agent system, where agents can communicate with each other and only observe the restricted information.

● Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$

Formulation

Example of model

- R : a reward function
- e.g., the treasure obtained by agents
- T : a time horizon



- Two agents get a treasure cooperatively.
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Outline

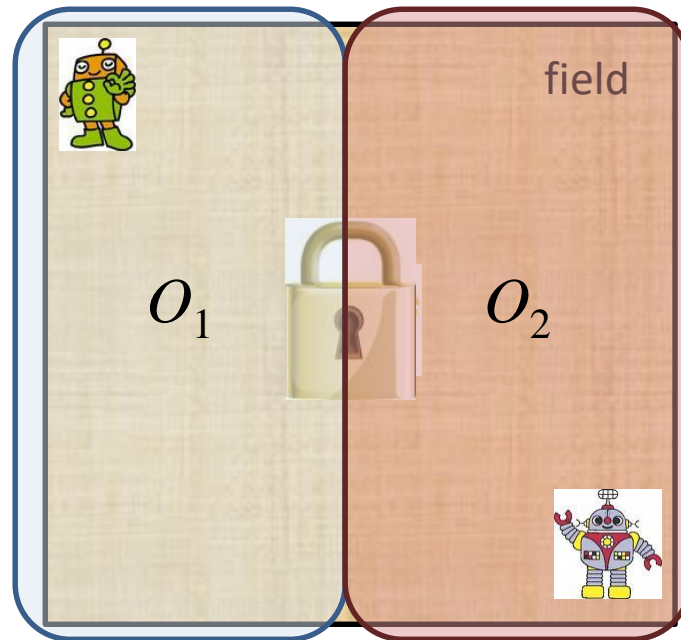
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Jointly Fully Observable Dec-POMDP-Com

[Goldman+ 04]

- The Dec-POMDP-Com such that the combination of the agents' observations leads to the global state.

Jointly fully Observable Dec-POMDP-Com



$O_1 + O_2 = \text{global state}$
(That is Jointly fully observable)

Deterministic Dec-POMDP-Com

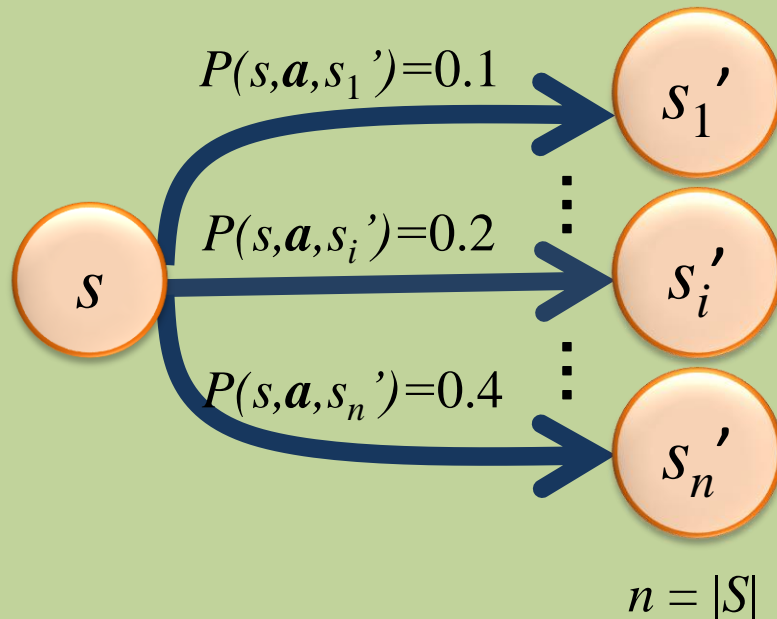
- The model where P and O on the definition are constrained.
- Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$

Deterministic Dec-POMDP-Com

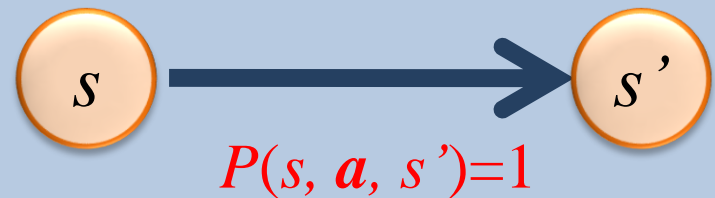
- The model where P and O on the definition are constrained.
- Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$

Restriction 1 : Deterministic transitions

- P is a transition probability function



- For any state $s \in S$ and any joint action $a \in A$, there exists a state $s' \in S$ such that $P(s, a, s') = 1$.



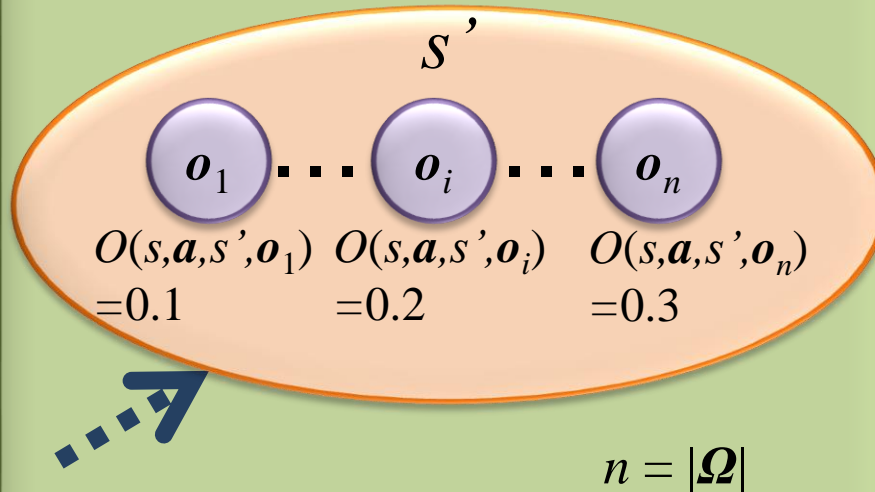
The next global state is decided uniquely.

Deterministic Dec-POMDP-Com

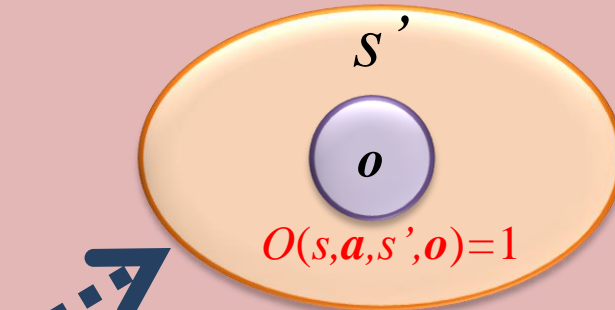
- The model where P and O on the definition are constrained.
- Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$

Restriction 2 : Deterministic observable

- O is a observation probability function



- For any state $s, s' \in S$ and any joint action $a \in A$, there exists a joint observation $o \in \Omega$ such that $O(s, a, s', o) = 1$.



The current observation is decided uniquely.

Deterministic Dec-POMDP-Com

- The model where P and O on the definition are constrained.
- Dec-POMDP-Com := $\langle I, S, \Omega, A, M, C, P, O, R, T \rangle$

Restriction 1 : Deterministic transitions

The next global state is decided uniquely.

Restriction 2 : Deterministic observable

The current observation is decided uniquely.

- When Dec-POMDP-Com has Restriction 1 and 2, it is called **Deterministic Dec-POMDP-Com**

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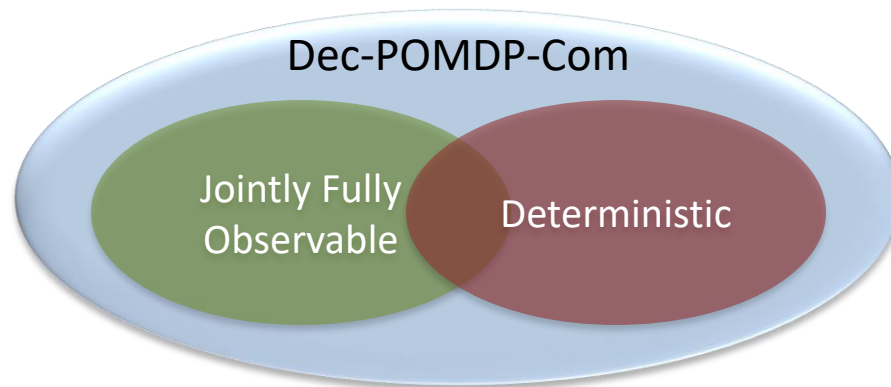
Main results

- Corollary 1 :

*Minimum required sizes $|M_i|$ for Signal Learning on **Jointly Fully Observable Dec-POMDP-Com***

- Theorem 2 :

*Minimum required sizes $|M_i|$ for Signal Learning with Messages on **Deterministic Dec-POMDP-Com***



Theorem 1 [Goldman 04]

Theorem 1

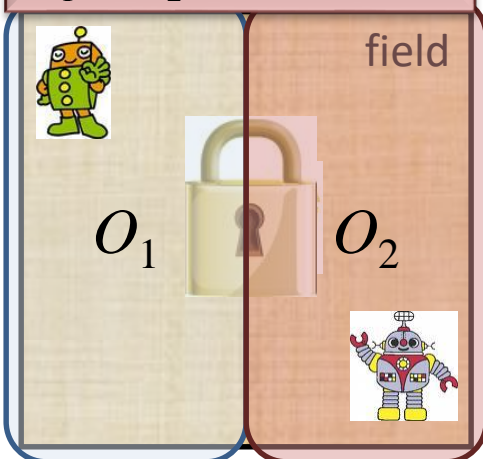
For any *jointly fully observable Dec-POMDP-Com*, the following equation holds.

$$\forall \mathbf{M}, \max_{\delta \in D} V_{\delta, \mathbf{M}}^T(s_0) \leq \max_{\delta \in D} V_{\delta, \mathbf{M}'}^T(s_0)$$

the value of the optimal joint policy with respect to any joint message set \mathbf{M}

the value of the optimal joint policy with respect to the joint message set $\mathbf{M}' := \Omega$.

$O_1 + O_2 =$ global state



This theorem means that the optimal communication policy of each agent is **to send its own observation** in jointly fully observable Dec-POMDP-Com (i.e. for agent i , $m_i := o_i$).

Each agent can always know the current global state by own observation and received message.

(e.g., for agent 1, $o_1 + m_2 = o_1 + o_2 =$ global state)

Therefore, each agent always perform the optimal actions.

Corollary 1

Corollary 1

For *any jointly fully observable Dec-POMDP-Com*, if the size $|M_i|$ of the message set of each agent i satisfies the condition,

$$|M_i| \geq |\Omega_i|$$

then the following equation holds:

$$\max_{\delta \in D^{SL}} V_{\delta}^T(s_0) = \max_{\delta' \in D} V_{\delta'}^T(s_0)$$

the value of the optimal joint policy on SL

the value of the optimal joint policy with history

From theorem 1 by Goldman,
the optimal communication policy of each agent is
to send its own observation in jointly fully observable Dec-POMDP-Com.

Therefore, If each agent has $|M_i|$ that is larger than $|\Omega_i|$,
it is possible to constructing the optimal policy such that each agent can
send its own observation.

Theorem 2

Theorem 2

For *any deterministic Dec-POMDP-Com*, if the size $|M_i|$ of the message set of each agent i satisfies the condition,

$$|M_i| \geq \max_{j \in I} \max_{o \in \Omega_j} |S_j^{obs}(o)|$$

then the following equation holds:

$$\max_{\delta \in D^{SLM}} V_{\delta}^T(s_0) = \max_{\delta' \in D} V_{\delta'}^T(s_0)$$

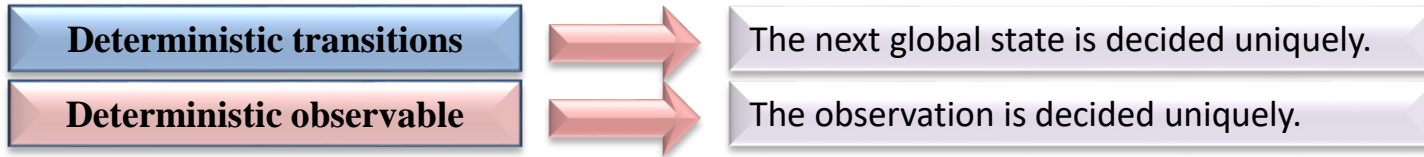
the value of the optimal joint policy on SLM

the value of the optimal joint policy with history

First, I explain a function S_j^{obs} .

S_j^{obs}

- Deterministic Dec-POMDP-Com has the following properties.

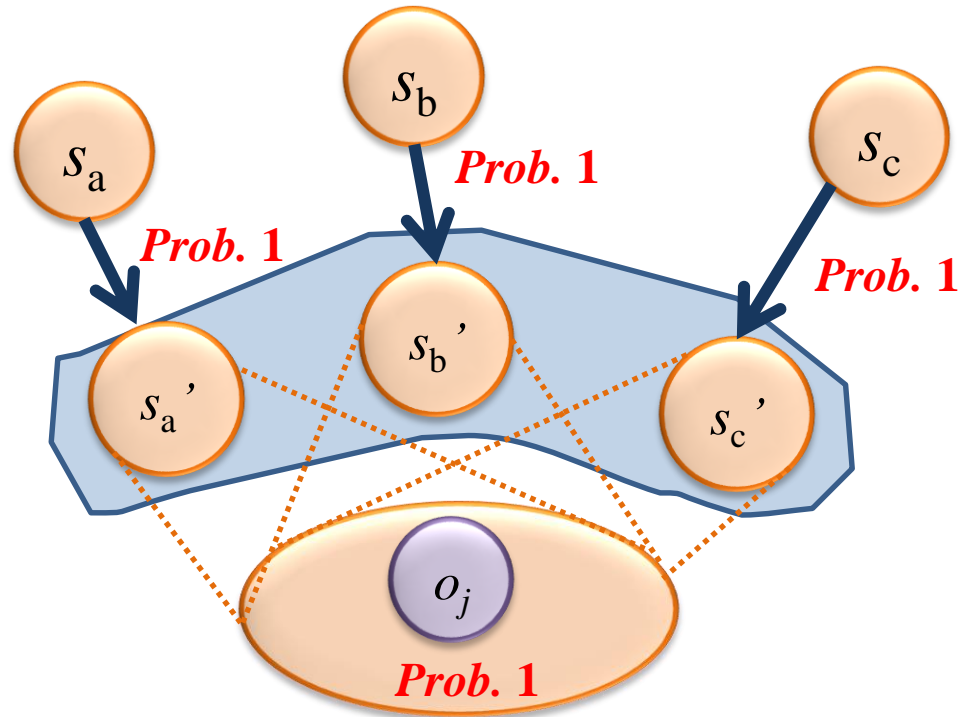


There exists some transitions such that agent j observes the same observation.

From the properties on deterministic Dec-POMDP-Com, we can compute the following function.

$$S_j^{obs}(o_j) = \{s_a', s_b', s_c'\}$$

S_j^{obs} returns the set of all states where agent j observes o_j .



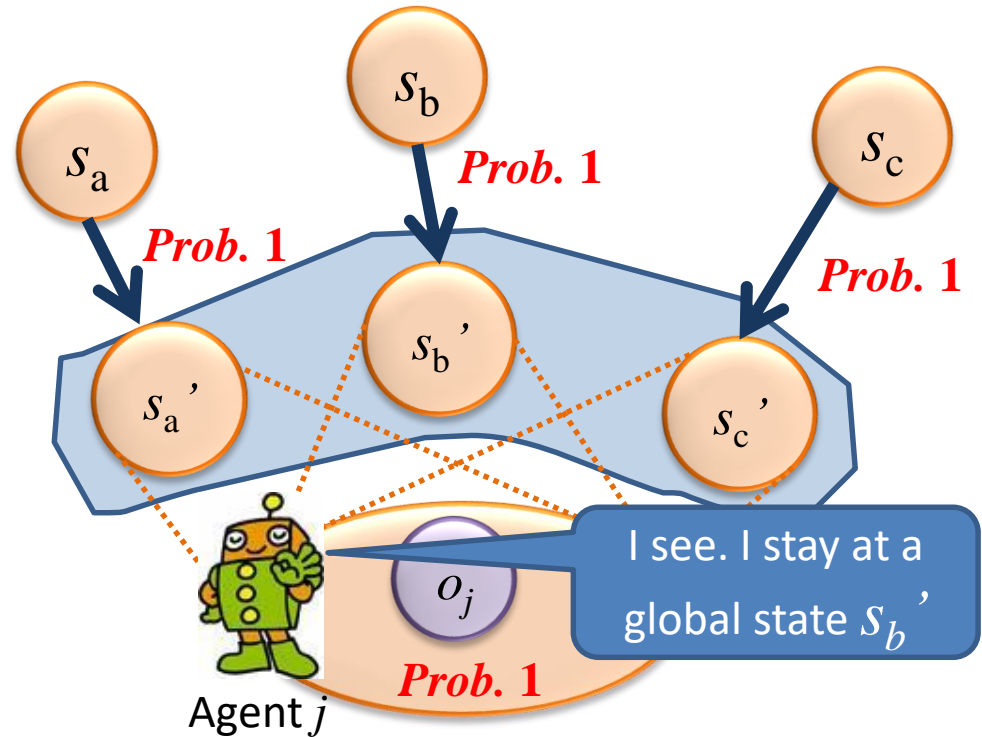
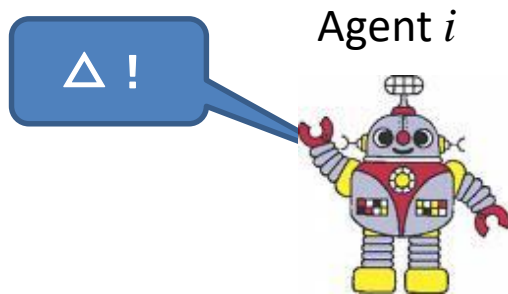
Proof sketch of Theorem 2

The condition of theorem 2 is $|M_i| \geq \max_{j \in I} \max_{o \in \Omega_j} |S_j^{obs}(o)|$



The condition shows that agent j can know the global state based on the message received from agent i by setting the set of message which have the maximum size of $S_j^{obs}(o_j)$.

$$S_j^{obs}(o_j) = \{s_a', s_b', s_c'\}$$

$$M_i = \{\square, \triangle, \circ\}$$



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Conclusion

- We defined deterministic Dec-POMDP-Com for theoretical analysis

Restriction 1 : Deterministic transitions

The next global state s' is decided uniquely.

Restriction 2 : Deterministic observable

The observation o is decided uniquely.

- We showed *Minimum required sizes* $|M_i|$ for
 - Signal Learning on *Jointly Fully Observable Dec-POMDP-Com* at corollary 1
 - Signal Learning with Messages on *Deterministic Dec-POMDP-Com* at Theorem 2