



国防科技大学
NATIONAL UNIVERSITY
OF DEFENSE TECHNOLOGY



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MRBTP: Efficient Multi-Robot Behavior Tree Planning and Collaboration

Session: Robotics (2/4)

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Introduction

❑ Multi-robot systems (MRS)

- robots with diverse capabilities, improved performance and fault tolerance
- requires an efficient and robust control architecture

❑ Behavior Trees (BTs)

- a popular control architecture for robot behaviors
- modularity, interpretability, reactivity, and robustness

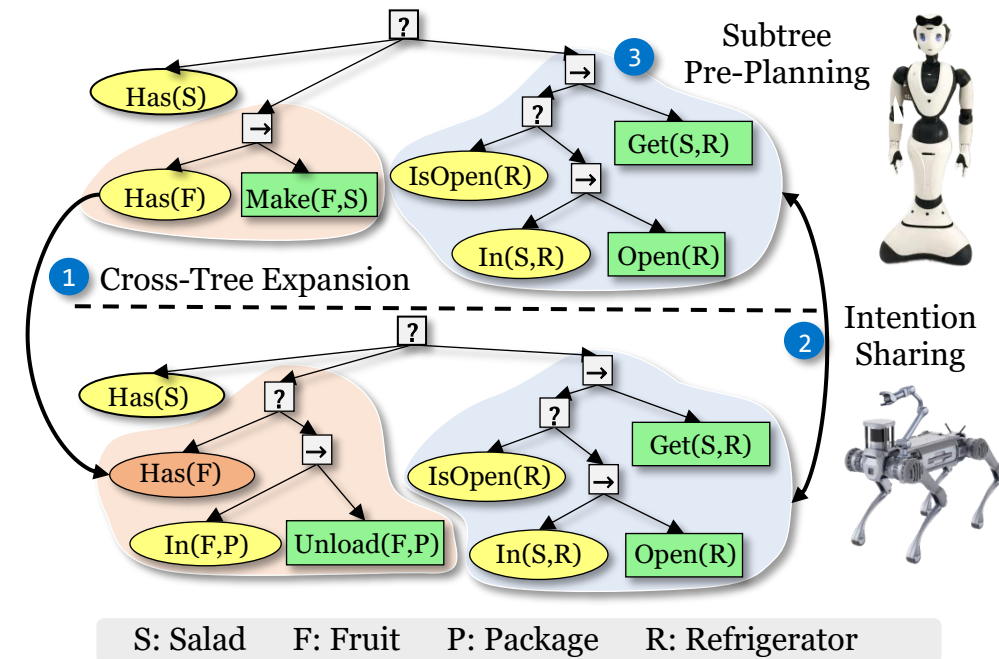
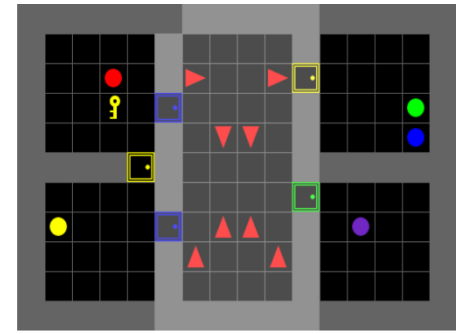
❑ BT in MRS

■ BT Planning for a Single Robot

- **Sound and Complete** (Proven by BT Expansion^[1])

■ Our Contribution in BT Planning for Multi-Robots

- **MRBTP**: theoretical guarantees of both **soundness and completeness**
- **Cross-Tree Expansion** and **Intention Sharing**: Maintain robustness while enhancing execution performance
- **Optional Plugin** (LLM-based subtrees pre-planning): Improves planning and execution efficiency



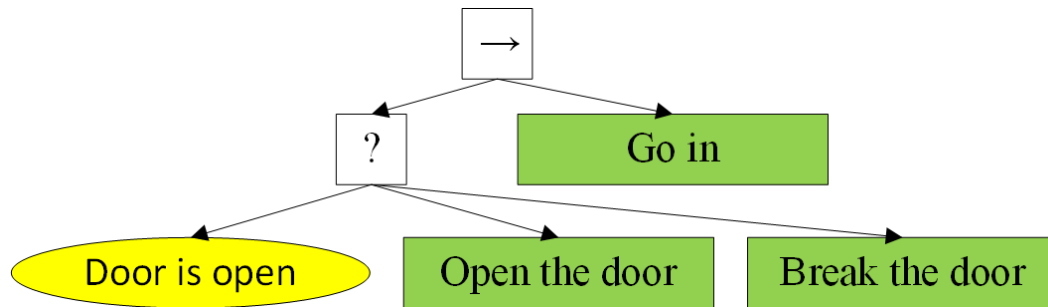
Background

❑ Classic Behavior Tree Representation

- BT: a directed rooted tree
 - Root at the top (usually omitted in graphics)
 - Internal: control-flow nodes
 - Leaf: condition/action nodes
 - Returns to the parent: $R = \{\text{Success, Running, Failure}\}^*$

❑ Behavior Trees: a running example

- Robot entering a room



Sequence nodes

S: All children Succeed
F: One child Fails
R: Others

Fallback nodes

S: One child succeeds
F: All children fail
R: Others

Other control-flow nodes: Parallel nodes and decorator nodes.

The BT can react and adapt to different run-time situations:

Door is open	→	Directly go in
Door not open	→	Open the door
Opening fails	→	Break the door
Someone helps open it	→	Skip breaking and go in

*Condition nodes only return success or failure.

For more details, we refer the readers to the book *Behavior Trees in Robotics and AI: An Introduction*, Michele Colledanchise, Petter Ögren, CRC Press.

Background

□ BT Expansion^[1]: A sound and complete algorithm for single-robot BT planning

■ STRIPS-style planning states and actions

- An action a is a three tuple $\langle \mathbf{pre(a)}, \mathbf{add(a)}, \mathbf{del(a)} \rangle$, consisting of the precondition, add effects and delete effects of the action.

$$add(a) \cap del(a) = \emptyset$$

$$add(a) \cap pre(a) = \emptyset$$

Properties for well-defined actions*

$$s_{t+k} = f_a(s_t) = s_t \cup add(a) \setminus del(a)$$

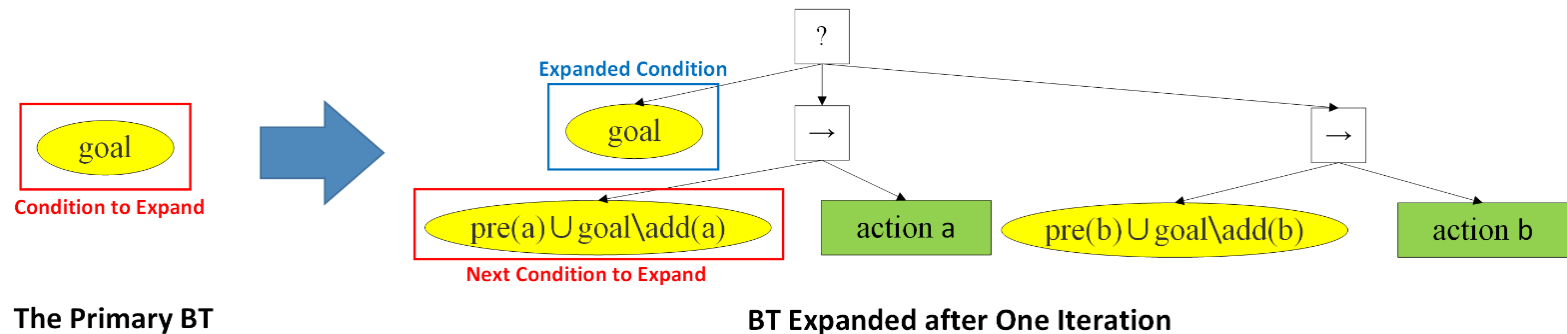
State transition of a BT executing a^*

■ Input & Output

- **Input:** Goal, Initial State, the finite set of actions

- **Output:** Goal-Oriented BT

■ The one-step expansion



Background

□ BT Expansion^[1]: A sound and complete algorithm for single-robot BT planning

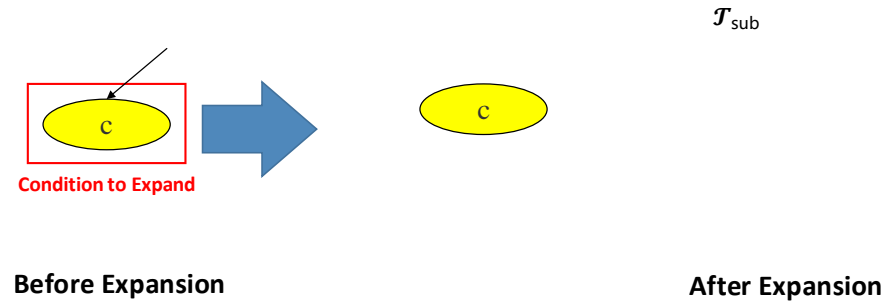
■ Input & Output

➤ Input: *Goal, Initial State, the finite set of actions*

➤ Output: *Goal-Oriented BT*

■ The one-step expansion

1. Preserve condition c



Background

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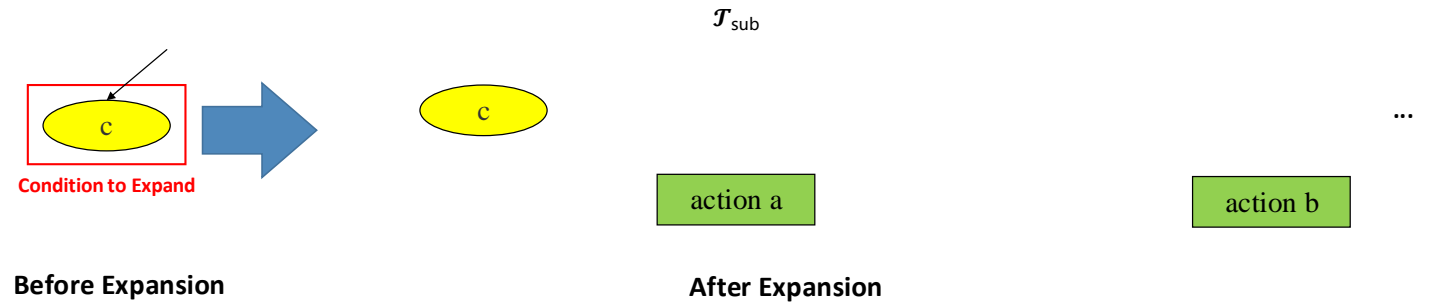
■ Input & Output

➤ Input: *Goal, Initial State, the finite set of actions*

➤ Output: *Goal-Oriented BT*

■ The one-step expansion

1. Preserve condition c
2. Select actions



Background

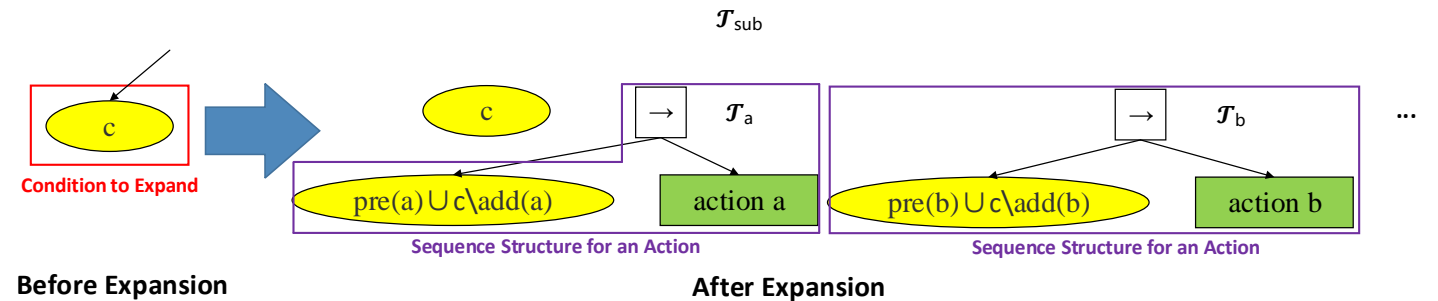
□ BT Expansion^[1]: A sound and complete algorithm for single-robot BT planning

■ Input & Output

- Input: *Goal, Initial State, the finite set of actions*
- Output: *Goal-Oriented BT*

■ The one-step expansion

1. Preserve condition c
2. Select actions
3. Construct sequence structures



Background

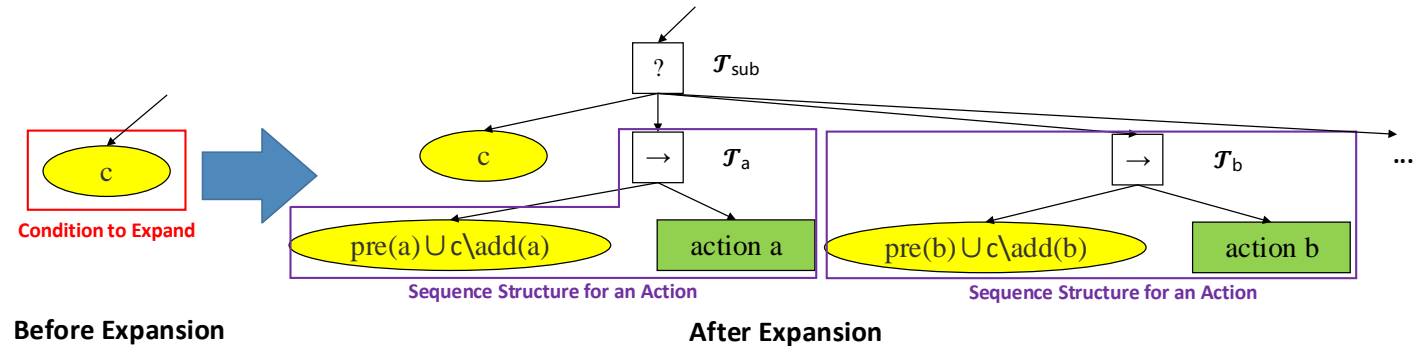
□ BT Expansion^[1]: A sound and complete algorithm for single-robot BT planning

■ Input & Output

- Input: *Goal, Initial State, the finite set of actions*
- Output: *Goal-Oriented BT*

■ The one-step expansion

1. Preserve condition c
2. Select actions
3. Construct sequence structures
4. Link to the top fallback node



Background

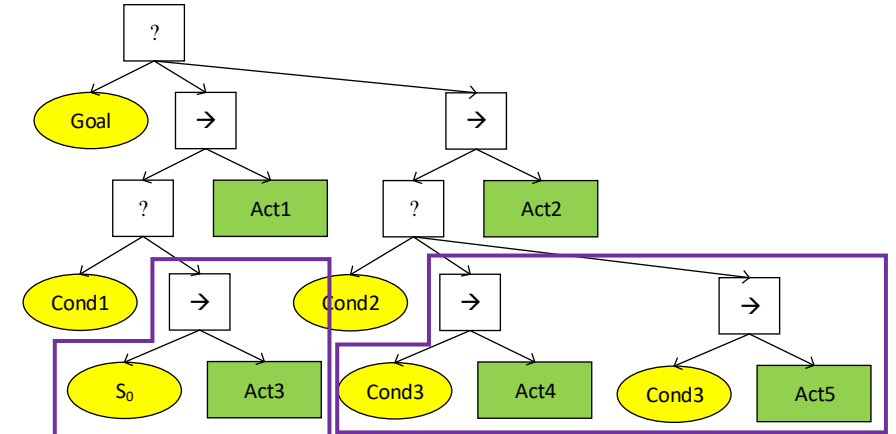
□ BT Expansion^[1]: A sound and complete algorithm for single-robot BT planning

■ Input & Output

- Input: *Goal, Initial State, the finite set of actions*
- Output: *Goal-Oriented BT*

■ The one-step expansion -> BT Expansion

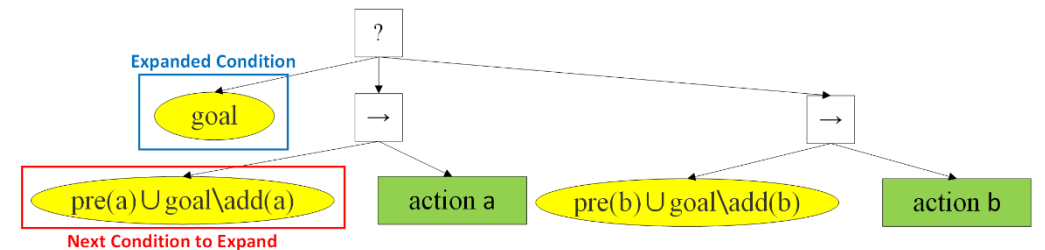
1. Preserve condition c
2. Select actions
3. Construct sequence structures
4. Link to the top fallback node
5. Repeat 2-4 until the BT can run in *Initial State*, or all conditions expanded.



BT Expansion



The Primary BT



BT Expanded after One Iteration

Proposed Algorithm: MRBTP

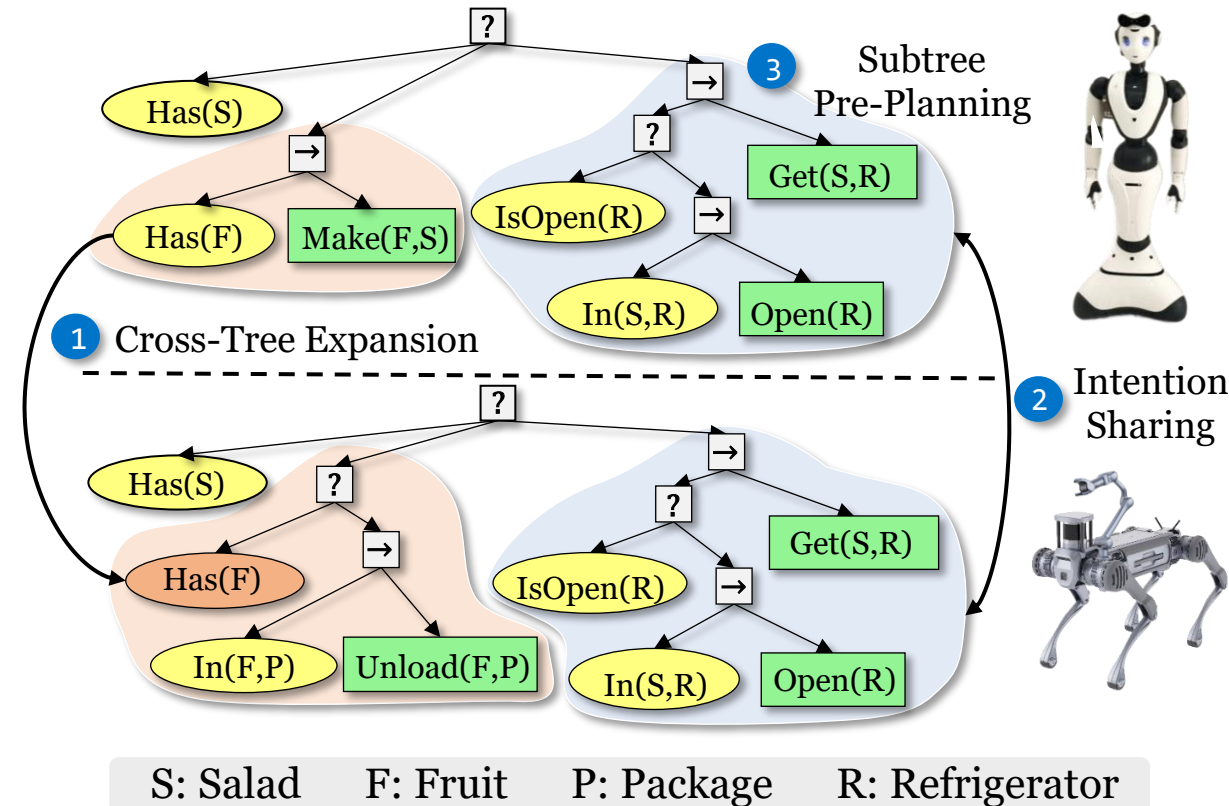
□ BT Planning for Multi-Robots

■ Input & Output

- Input: *Team Goal, Initial State, **Finite Action Set for each Robot***
- Output: **Goal-Oriented BT for each Robot**

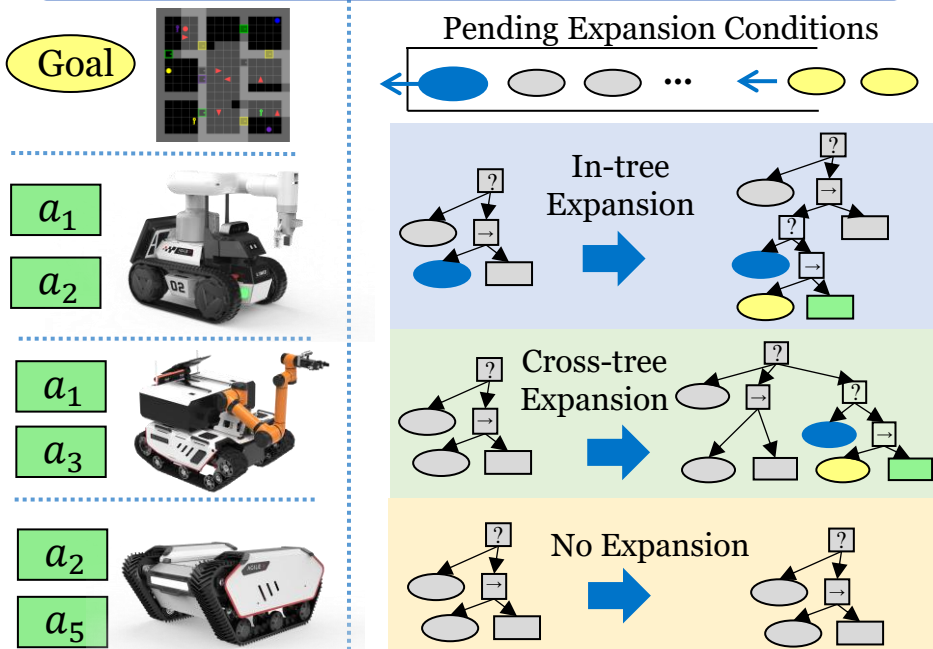
□ Challenges

- How to coordinating **heterogeneous actions** across multiple BTs to achieve team goals?
- How to enhancing fault tolerance for **homogeneous actions** without redundant execution?

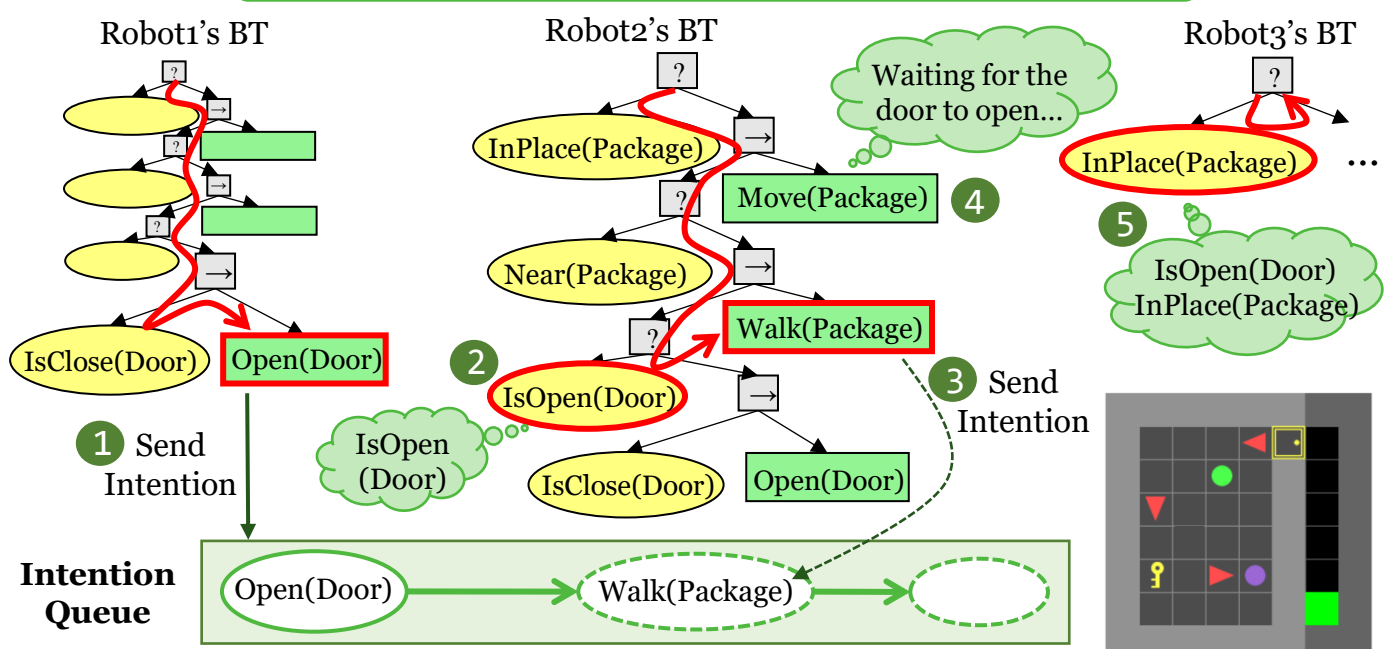


Proposed Algorithm: MRBTP

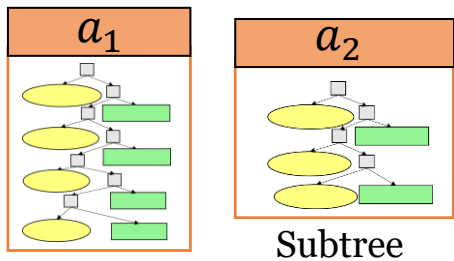
Multi-Robot Behavior Tree Planning



Intention Sharing for Multi-BT Execution



Optional Plugin: Subtree Pre-Planning



Subtree Pre-Planning

Robot1 }
Robot2 ...

Walk (self, pie)
Walk (self, stove)

RightGrab (self, pie)
Open (self, stove)

Walk
(self, table)

RightPut
(self, pie, table)

RightPut
(self, pie, table)

sk-Related

Task-Related Action Sequences



LLM



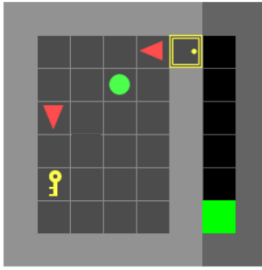
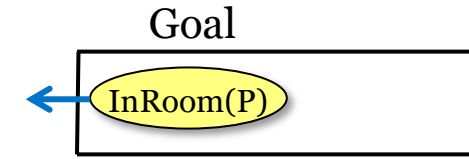
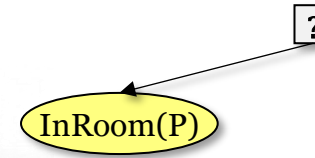
Proposed Algorithm: MRBTP

① Multi-Robot Behavior Tree Planning

1. Preserve condition g

Move(P)

Open(Door)



Algorithm 2: MABTP

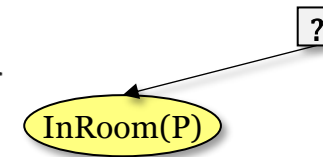
Input: problem $\langle \mathcal{S}, \mathcal{L}, \{\mathcal{A}_i\}_{i=1}^n, \mathcal{M}, s_0, c \rangle$

Output: solution $\Phi = \{\mathcal{T}_i\}_{i=1}^n$

```

1:  $\mathcal{C}_U \leftarrow \{g\}$  ▷ conditions to be explored
2:  $\mathcal{C}_E \leftarrow \emptyset$  ▷ expanded conditions
3: for  $i = 1$  to  $n$  do
4:    $\mathcal{T}_i \leftarrow \text{Fallback}(g)$  ▷ init the BTs
5: while  $\mathcal{C}_U \neq \emptyset$  do
6:    $c \leftarrow \text{Pop}(\mathcal{C}_U)$  ▷ explore  $c$ 
7:   if  $\text{HasSubSet}(c, \mathcal{C}_E)$  then continue ▷ prune
8:   for  $i = 1$  to  $n$  do
9:      $\mathcal{T}_i, \mathcal{C}_{new} \leftarrow \text{ExpandOneRobot}(\mathcal{T}_i, \mathcal{A}_i, c)$ 
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11:      return  $\Phi = \{\mathcal{T}_i\}_{i=1}^n$  ▷ return a solution
12:     $\mathcal{C}_E \leftarrow \mathcal{C}_E \cup \mathcal{C}_{new}$ 
13:     $\mathcal{C}_U \leftarrow \mathcal{C}_U \cup \mathcal{C}_{new}$  ▷ add new conditions
14: return Unsolvale
  
```

Open(Door)



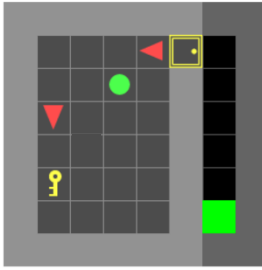
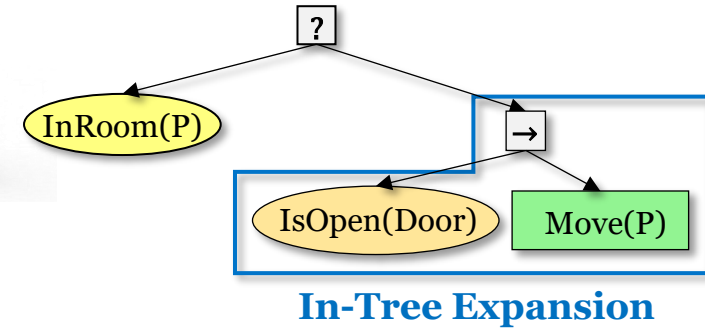
Proposed Algorithm: MRBTP

① Multi-Robot Behavior Tree Planning

1. Preserve condition g
2. One-step expansion of each robot's BT

Move(P)

Open(Door)



Algorithm 2: MABTP

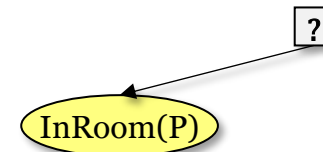
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14: return Unsolvble
  
```

Open(Door)



Proposed Algorithm: MRBTP

① Multi-Robot Behavior Tree Planning

1. Preserve condition g
2. One-step expansion of each robot's BT
 - In-Tree Expansion
 - Cross-Tree Expansion

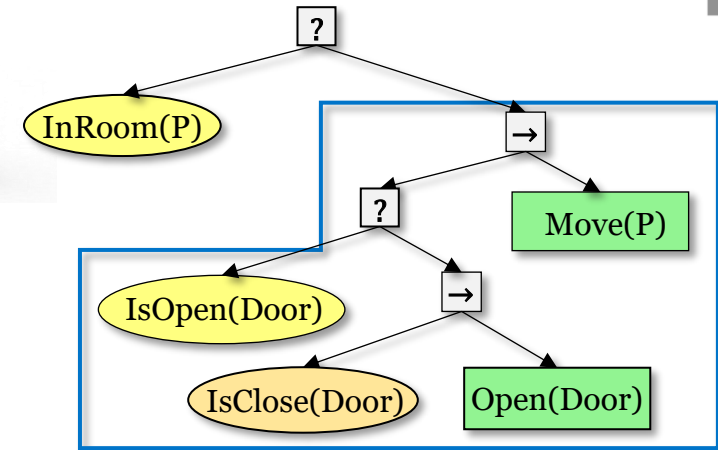
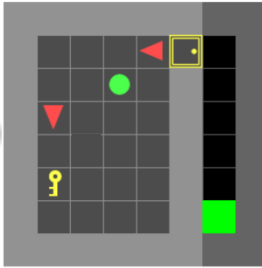
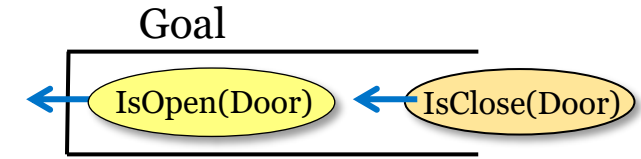
Algorithm 1: One-step cross-tree expansion

```

1: function ExpandOneRobot ( $\mathcal{T}, \mathcal{A}, c$ )
2:    $\mathcal{T}_{new} \leftarrow c$  ▷ newly expanded subtree
3:    $\mathcal{C}_{new} \leftarrow \emptyset$  ▷ newly expanded conditions
4:   for each action  $a \in \mathcal{A}$  do
5:     if  $c \cap (pre(a) \cup add(a) \setminus del(a)) \neq \emptyset$  and  $c \setminus del(a) = c$  then
6:        $c_a \leftarrow pre(a) \cup c \setminus add(a)$ 
7:        $\mathcal{T}_a \leftarrow Sequence(c_a, a)$ 
8:        $\mathcal{T}_{new} \leftarrow Fallback(\mathcal{T}_{new}, \mathcal{T}_a)$ 
9:        $\mathcal{C}_{new} \leftarrow \mathcal{C}_{new} \cup \{c_a\}$ 
10:  if  $\mathcal{C}_{new} \neq \emptyset$  then
11:    if ConditionInTree( $c, \mathcal{T}$ ) then
12:      Replace  $c$  with  $\mathcal{T}_{new}$  in  $\mathcal{T}$  ▷ in-tree expand
13:    else if  $\mathcal{T}_{new} \neq c$  then
14:       $\mathcal{T} \leftarrow Fallback(\mathcal{T}, \mathcal{T}_{new})$  ▷ cross-tree expand
15:  return  $\mathcal{T}, \mathcal{C}_{new}$ 
    
```

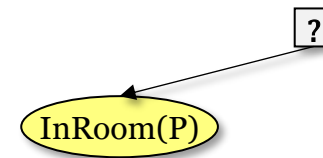
Move(P)

Open(Door)



In-Tree Expansion

Open(Door)



Proposed Algorithm: MRBTP

① Multi-Robot Behavior Tree Planning

1. Preserve condition g
2. One-step expansion of each robot's BT
 - In-Tree Expansion
 - Cross-Tree Expansion

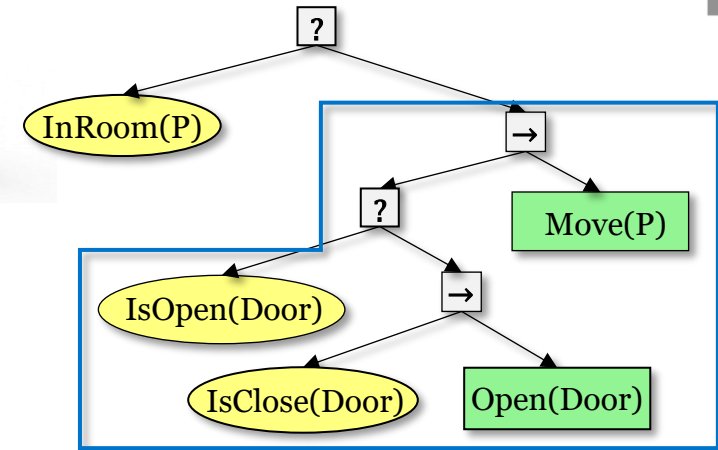
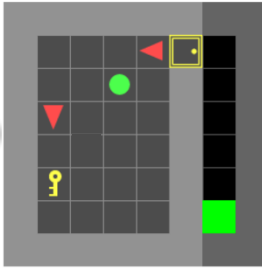
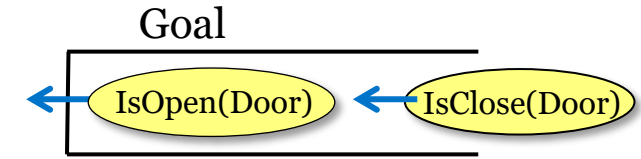
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10:  if  $\mathcal{C}_{new} \neq \emptyset$  then
11:    if ConditionInTree( $c, \mathcal{T}$ ) then
12:      Replace  $c$  with  $\mathcal{T}_{new}$  in  $\mathcal{T}$  ▷ in-tree expand
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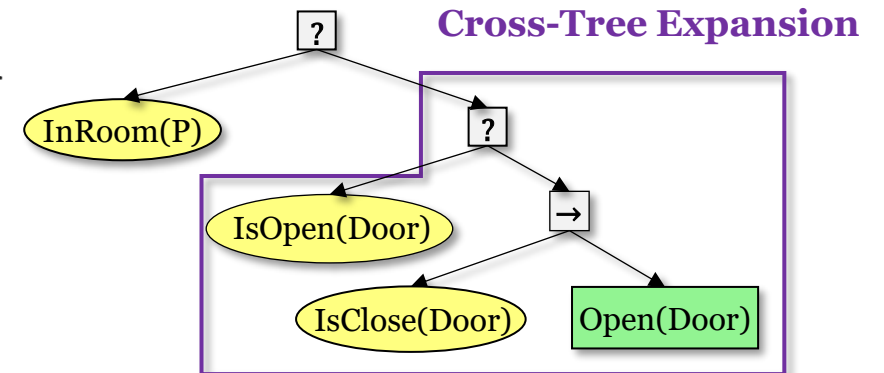
Move(P)

Open(Door)



In-Tree Expansion

Open(Door)



Proposed Algorithm: MRBTP

① Multi-Robot Behavior Tree Planning

1. Preserve condition g
2. One-step expansion of each robot's BT
 - In-Tree Expansion
 - Cross-Tree Expansion

□ Proof of Soundness and Completeness

Definition 1 (Multi-BT System). A n -robot BT system is a four-tuple $\langle \Phi, f_\Phi, r_\Phi, \Delta t_\Phi \rangle$, where $\Phi = \{\mathcal{T}_i\}_{i=1}^n$ is the set of BTs, $f_\Phi : \mathcal{S} \mapsto \mathcal{S}$ is the team state transition function, Δt_Φ is the team time step, $r_\Phi : \mathcal{S} \mapsto \{S, R, F\}$ is the team region partition.

Definition 2 (Finite Time Successful). Φ is finite time successful (FTS) from region of attraction (ROA) R to condition c , if $\forall s_0 \in R$ there is a finite time τ such that for any $t < \tau$, $r_\Phi(s_t) = R$, and for any $t \geq \tau$, $r_\Phi(s_t) = S$, $c \in s_t$.

Algorithm 2: MABTP

Input: problem $\langle \mathcal{S}, \mathcal{L}, \{\mathcal{A}_i\}_{i=1}^n, \mathcal{M}, s_0, c \rangle$

Output: solution $\Phi = \{\mathcal{T}_i\}_{i=1}^n$

```

1:  $\mathcal{C}_U \leftarrow \{g\}$   $\triangleright$  conditions to be explored
2:  $\mathcal{C}_E \leftarrow \emptyset$   $\triangleright$  expanded conditions
3: for  $i = 1$  to  $n$  do
4:    $\mathcal{T}_i \leftarrow \text{Fallback}(g)$   $\triangleright$  init the BTs
5: while  $\mathcal{C}_U \neq \emptyset$  do
6:    $c \leftarrow \text{Pop}(\mathcal{C}_U)$   $\triangleright$  explore  $c$ 
7:   if  $\text{HasSubSet}(c, \mathcal{C}_E)$  then continue  $\triangleright$  prune
8:   for  $i = 1$  to  $n$  do
9:      $\mathcal{T}_i, \mathcal{C}_{new} \leftarrow \text{ExpandOneRobot}(\mathcal{T}_i, \mathcal{A}_i, c)$ 
10:    if  $\text{HasSubSet}(s_0, \mathcal{C}_{new})$  then
11:      return  $\Phi = \{\mathcal{T}_i\}_{i=1}^n$   $\triangleright$  return a solution
12:     $\mathcal{C}_E \leftarrow \mathcal{C}_E \cup \mathcal{C}_{new}$ 
13:     $\mathcal{C}_U \leftarrow \mathcal{C}_U \cup \mathcal{C}_{new}$   $\triangleright$  add new conditions
14: return Unsolvable
    
```

Algorithm 1: One-step cross-tree expansion

```

1: function  $\text{ExpandOneRobot}(\mathcal{T}, \mathcal{A}, c)$ 
2:    $\mathcal{T}_{new} \leftarrow c$   $\triangleright$  newly expanded subtree
3:    $\mathcal{C}_{new} \leftarrow \emptyset$   $\triangleright$  newly expanded conditions
4:   for each action  $a \in \mathcal{A}$  do
5:     if  $c \cap (\text{pre}(a) \cup \text{add}(a) \setminus \text{del}(a)) \neq \emptyset$  and  $c \setminus \text{del}(a) = c$  then
6:        $c_a \leftarrow \text{pre}(a) \cup c \setminus \text{add}(a)$ 
7:        $\mathcal{T}_a \leftarrow \text{Sequence}(c_a, a)$ 
8:        $\mathcal{T}_{new} \leftarrow \text{Fallback}(\mathcal{T}_{new}, \mathcal{T}_a)$ 
9:        $\mathcal{C}_{new} \leftarrow \mathcal{C}_{new} \cup \{c_a\}$ 
10:  if  $\mathcal{C}_{new} \neq \emptyset$  then
11:    if  $\text{ConditionInTree}(c, \mathcal{T})$  then
12:      Replace  $c$  with  $\mathcal{T}_{new}$  in  $\mathcal{T}$   $\triangleright$  in-tree expand
13:    else if  $\mathcal{T}_{new} \neq c$  then
14:       $\mathcal{T} \leftarrow \text{Fallback}(\mathcal{T}, \mathcal{T}_{new})$   $\triangleright$  cross-tree expand
15:  return  $\mathcal{T}, \mathcal{C}_{new}$ 
    
```

Proposition 1. Given \mathcal{T} is FTS from R to g , if \mathcal{T} is expanded by Algorithm 1 to \mathcal{T}' given c , c is in \mathcal{T} and $\mathcal{C}_{new} \neq \emptyset$, then \mathcal{T}' is FTS from $R' = R \cup \{s \in \mathcal{S} | c_a \subseteq s, c_a \in \mathcal{C}_{new}\}$ to g .

Proposition 2. If \mathcal{T} is expanded by Algorithm 1 to \mathcal{T}' given c , c is not in \mathcal{T} and $\mathcal{C}_{new} \neq \emptyset$, then \mathcal{T}' is FTS from $\mathcal{S}_{new} = \{s \in \mathcal{S} | c_a \subseteq s, c_a \in \mathcal{C}_{new}\}$ to c .

Proposition 4. Algorithm 2 is sound, i.e. if it returns a result Φ rather than Unsolvale, then Φ is a solution of Problem 1.

Proposition 5. Algorithm 2 is complete, i.e., if Problem 1 is solvable, the algorithm returns a Φ which is a solution.

Proposed Algorithm: MRBTP

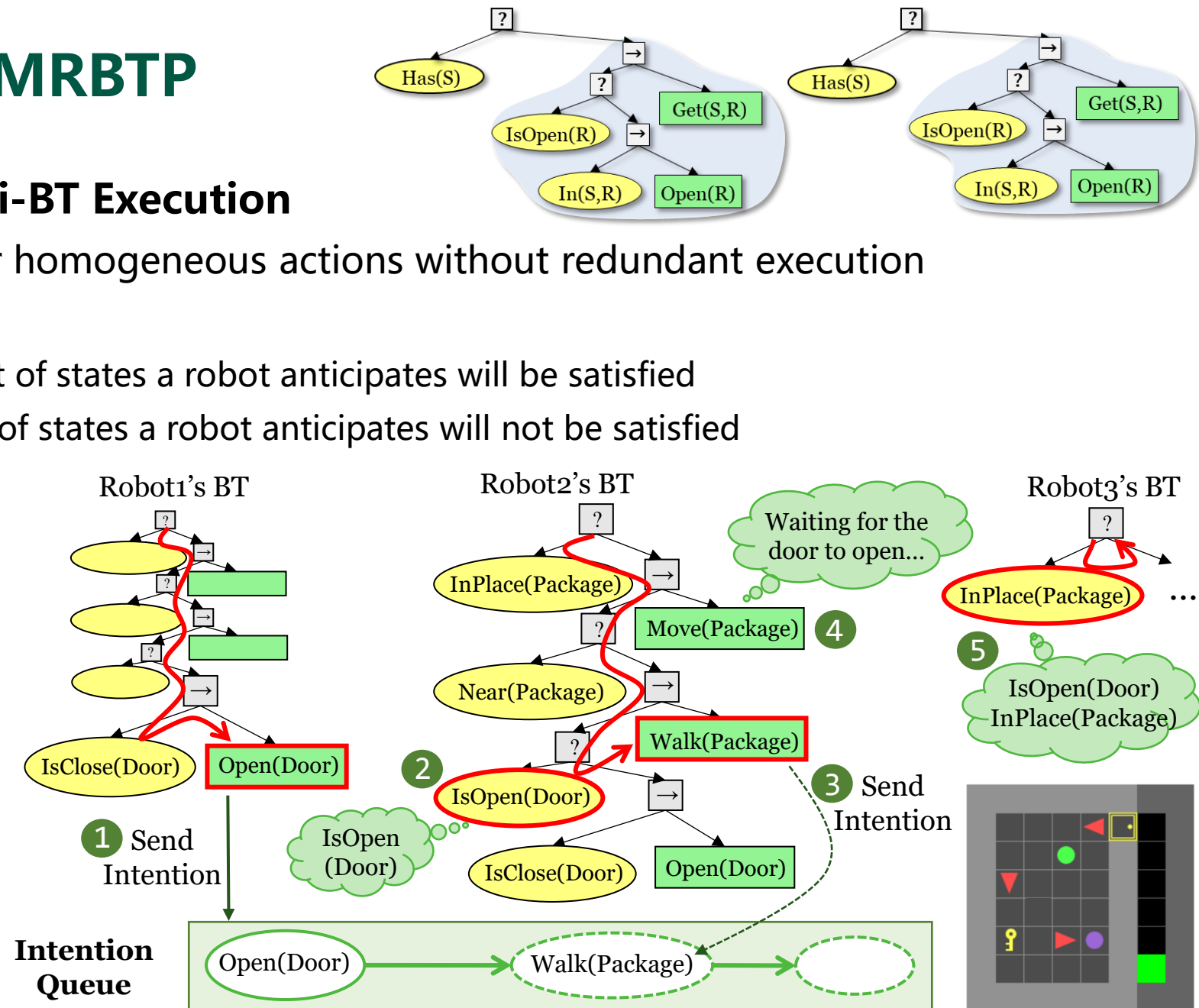
② Intention Sharing for Multi-BT Execution

- Enhancing fault tolerance for homogeneous actions without redundant execution
- Intention Queue
 - Belief Success Space: The set of states a robot anticipates will be satisfied
 - Belief Failure Space: The set of states a robot anticipates will not be satisfied
- Parallelism and Blocking

$$\mathcal{B}_i^S = \bigcup_{k=1}^{j-1} (add(a_k) \setminus del(a_k))$$

$$\mathcal{B}_i^F = \bigcup_{k=1}^{j-1} (del(a_k) \setminus add(a_k))$$

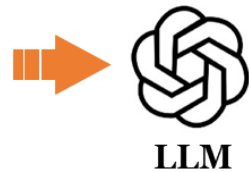
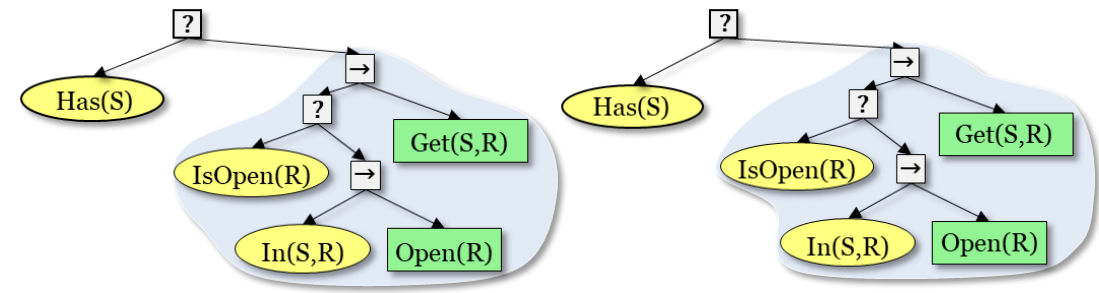
Robot i's Belief Space
(j: Index in the Queue)



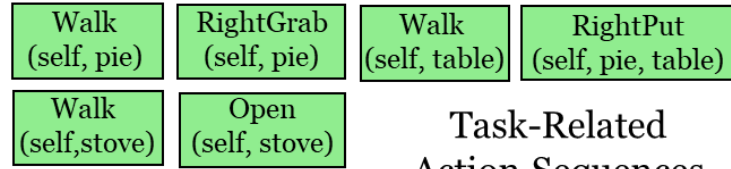
Proposed Algorithm: MRBTP

③ Optional Plugin: Subtree Pre-Planning

- During planning: Repeated tree generation due to overlapping action spaces
- During execution: High communication overhead from short-horizon atomic actions

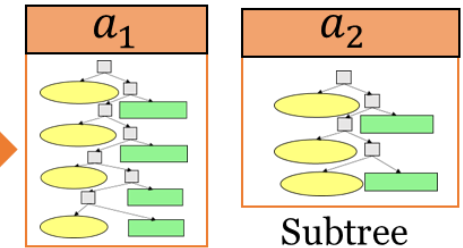


Robot1
Robot2 ...



Task-Related
Action Sequences

Subtree Pre-Planning



- Acquiring task-related action sequences via LLM
 - Preconditions, Add Effects, and Delete Effects of the Action Sequence $A = (a_1, a_1, \dots, a_1)$

$$pre(A) = \bigcup_{j=1}^m \left(pre(a_j) \setminus \bigcup_{k=1}^j add(a_k) \right)$$

$$add(A) = \bigcup_{j=1}^m (add(a_j) \setminus del(a_j)) - pre(A)$$

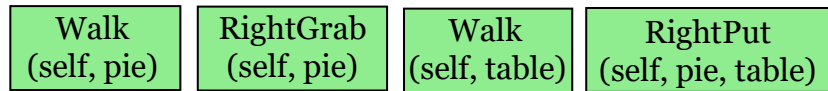
$$del(A) = \bigcup_{j=1}^m (del(a_j) \setminus add(a_j))$$

- Subtree Pre-Planning
 - Iterative one-step expansion to generate the subtree of A

Proposed Algorithm: MRBTP

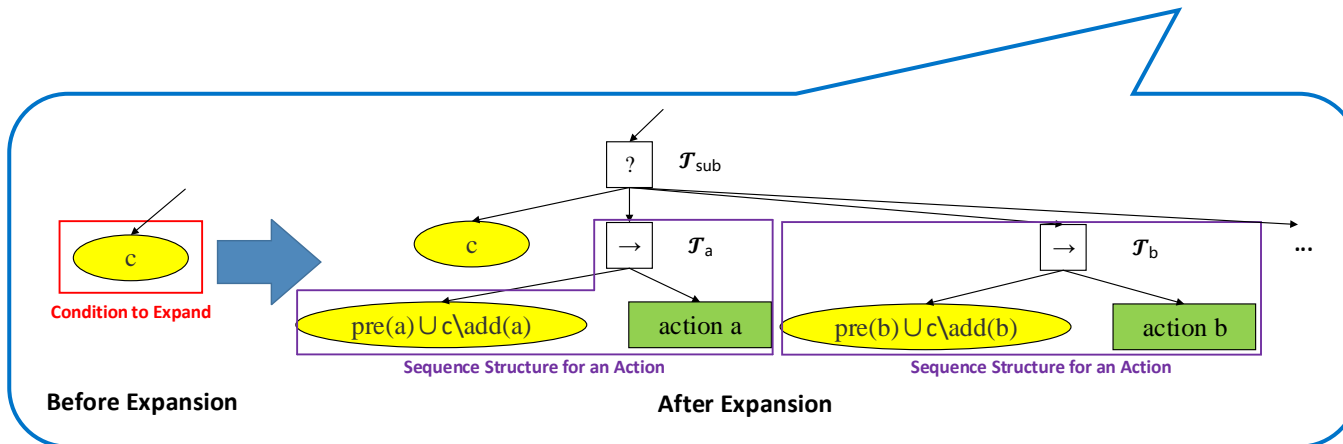
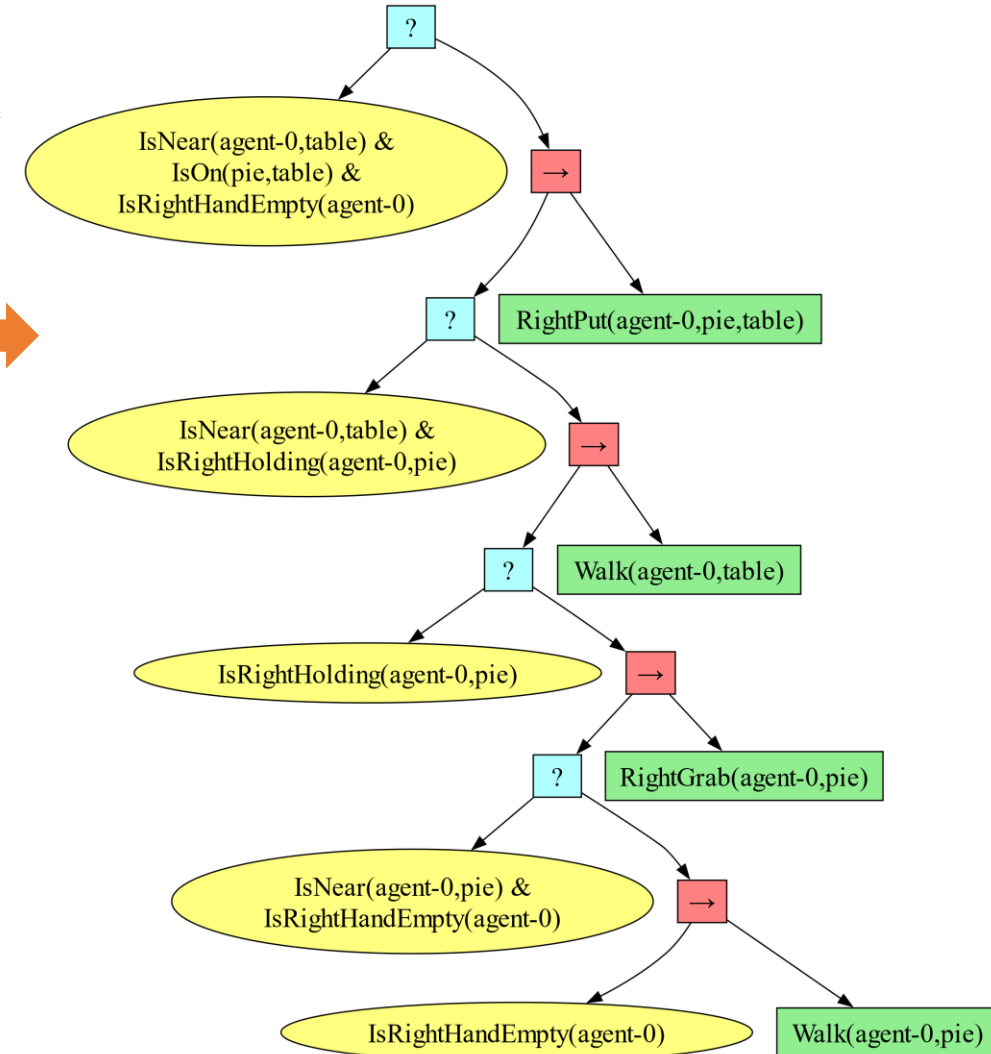
③ Optional Plugin: Subtree Pre-Planning

- Subtree Pre-Planning
 - Iterative one-step expansion to generate the subtree of A



Task-Related Action Sequences
Robot1

Iterative
One-Step
Expansion



Evaluation

□ Scenarios

■ Warehouse Management

- Minigrid^[1], 4-8 robots in 4 rooms with random packages

■ Home Service

- VirtualHome^[2], 2-4 robots handle 10+ objects and 100+ actions

□ Settings

■ Homogeneity α

- The proportion of redundant actions assigned to robots
- $\alpha = 1$: complete homogeneity
- $\alpha = 0$: complete heterogeneity

■ Intention Sharing

- Atomic IS: intention sharing among atomic action
- Subtree IS: intention sharing among subtree

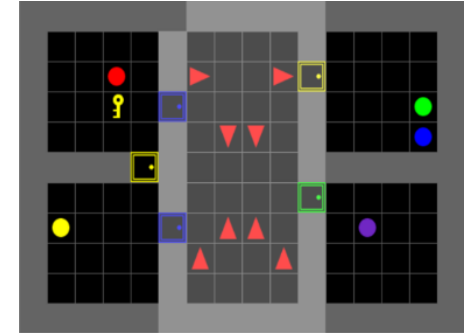


Table 1: Action predicates for different scenarios

Scenarios	Included Action Predicates
Warehouse Mangement	GoToInRoom, GoBtwRoom, PickUp, PutInRoom, PutNearInRoom, Toggle
Home Service	Walk, LeftPut, LeftPutIn, LeftGrab, RightGrab, Open, Close, RightPut, RightPutIn, SwitchOn, SwitchOff

Evaluation

□ Evaluation Metrics

- Success Rate (SR): Task success rate across trials
- Team Steps (TS): Total steps for all robots to complete tasks in parallel
- Total Robot Steps (RS): Sum of steps taken by each robot independently

□ Performance Comparison

- MRBTP achieves a **100% success rate** across all settings due to its **cross-tree expansion**
- MRBTP outperforms BT-Expansion in execution efficiency due to intention sharing

Table 1: Performance comparison with baseline in warehouse management (4 robots, averaged over 500 trials).

Method	$\alpha = 1$			$\alpha \approx 0.5$	$\alpha = 0$
	SR(%)	TS	RS	SR(%)	SR(%)
BT-Expansion	100	8.8	33.8	12.4	4.6
MRBTP	100	5.8	15.3	100	100

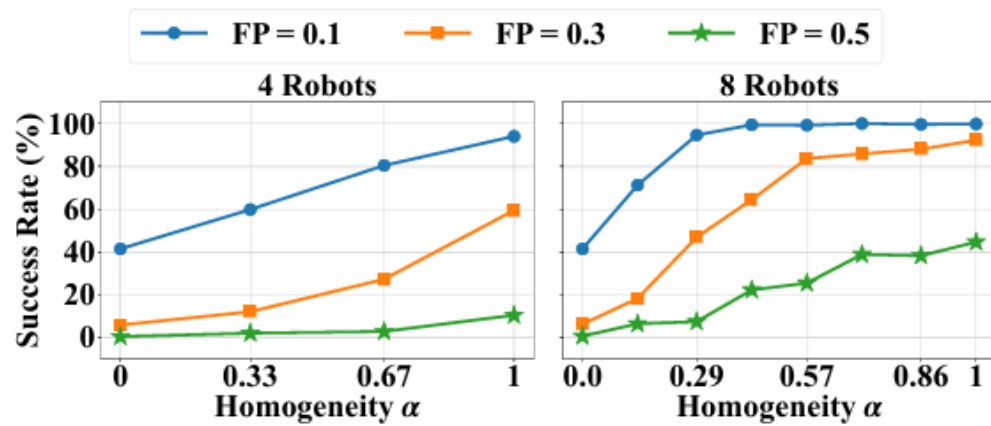
Evaluation

□ Robustness

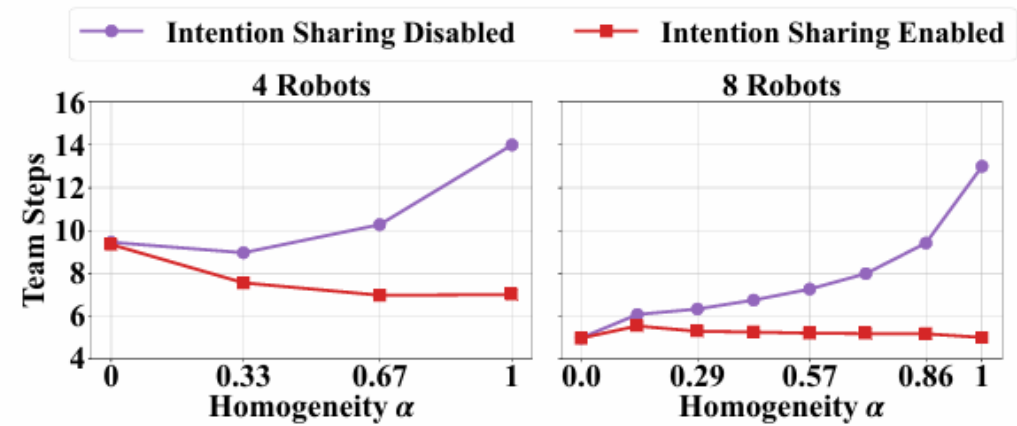
- Action Failure Probability (FP): The probability that a robot fails to execute an action
- Robustness improves with higher homogeneity and more robots

□ Impact of **Intention Sharing** on Execution Efficiency

- Intention sharing **reduces team steps** and maintains superior execution efficiency
- In **fully heterogeneous** scenarios, intention sharing achieves the **greatest efficiency gains**



(a) The impact of homogeneity on robustness



(b) The impact of intention sharing on execution efficiency

Evaluation

□ Impact of Task-Specific Subtree Pre-Planning

- Subtree pre-planning and reuse **significantly reduces BTs planning time**
- Intention sharing within subtrees **improves execution efficiency while minimizing communication overhead**

□ LLMs and Feedback

- Feedback (F) and No Feedback (NF)
- Feedback: grammar errors, Insufficient/incoherent action sequences,
- Feedback effectively **enhances both planning and execution efficiency**
- It becomes more effective as the reasoning ability of LLMs improves

Table 3: Planning efficiency with pre-planned subtrees. The average response time per LLM invocation is 4.2 seconds.

Homogeneity	Subtree	Feedback	EC	PT (s)
$\alpha = 1$	-	-	8033.3	Timeout
	✓	-	998.1	12.4
	✓	✓	384.3	3.7
$\alpha \approx 0.5$	-	-	7882.5	Timeout
	✓	-	623.8	7.2
	✓	✓	267.9	2.6
$\alpha = 0$	-	-	2695.5	20.2
	✓	-	576.6	5.6
	✓	✓	146.8	1.4

Table 4: Execution efficiency across different LLMs under $\alpha = 1$ with subtree and subtree intention sharing.

Models	No Feedback			Feedback		
	TS	RS	Comm.	TS	RS	Comm.
GPT-3.5-turbo	81.6	223.6	5.1	80.0	219.0	5.1
GPT-4o-mini	78.9	217.4	6.7	77.1	205.2	6.6
GPT-4o	77.4	200.9	6.3	74.9	190.7	6.3

Table 2: Execution efficiency with subtree pre-planning and intention sharing.

Homogeneity		$\alpha = 1$						$\alpha \approx 0.5$						$\alpha = 0$					
Subtree		-	-	✓	✓	✓	✓	-	-	✓	✓	✓	✓	-	-	✓	✓	✓	✓
Subtree IS		-	-	-	-	✓	✓	-	-	-	-	✓	✓	-	-	-	-	✓	✓
Atomic IS		-	✓	-	✓	-	✓	-	✓	-	✓	-	✓	-	✓	-	✓	-	✓
TS	NF	161	159.4	114.4	109.6	78.9	79.8	139.7	137.5	126.2	119.6	86.9	102.5	73.7	68.5	96.1	94.8	75.6	78.0
	F	-	-	116.7	114.2	77.1	79.16	-	-	124.6	126.0	80.8	96.2	-	-	107.1	106.4	70.4	76.8
RS	NF	570.8	557.3	374	359.4	217.4	219.1	385.2	380.1	345.8	326.6	209.6	222	128.6	128.6	128.6	128.6	128.6	128.6
	F	-	-	377	370.9	205.2	208	-	-	380.7	348.2	192.2	209	-	-	128.6	128.6	128.6	128.6
Comm.	NF	0.0	63.8	0.0	7.1	6.7	14.1	0.0	43.5	0.0	8.0	7.1	20.9	0.0	15.2	0.0	2.8	6.5	9.3
	F	-	-	0.0	4.8	6.6	12.2	-	-	0.0	4.4	6.4	13.2	-	-	0.0	0.7	5.2	6.0

Summary

□ Conclusion

- MRBTP, the first **sound and complete** algorithm for solving the multi-robot BT planning problem
- MRBTP uses **cross-tree expansion** to coordinate heterogeneous actions and enhances robustness and execution efficiency with backup structures and **intention sharing**
- **Optional LLMs plugin** speeds up planning and improves collaboration with pre-planned long-horizon subtrees

□ Future Work

- Real-time deadlock detection and resolution in BT execution
- Efficient large-scale BT communication, coordination, and emergent intelligence
- Planning under uncertainty and periodic tasks



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MRBTP: Efficient Multi-Robot Behavior Tree Planning and Collaboration

Thank you!

Q & A

Email: dids_ei@163.com

Code: <https://github.com/DIDS-EI/MRBTP>