

A MIP-Based Approach for Multi-Robot Geometric Task-and-Motion Planning

Goal : Move Objects M to Region R

Initial setup : n_R robots, fixed objects F , n_M movable objects, n_{Rg} regions, $Gr = \bigcup_{R \in Rg} Gr_{M,R}$

Grounded joint action : at time j , $s_j = \langle (\alpha_{R_1}^j, \xi_{R_1}^j) \dots (\alpha_{R_n}^j, \xi_{R_n}^j) \rangle$ α : action (act/place/move)
 ξ : trajectory

Pick&Place action : $a = \langle M, R_e, R_p^{pick}, R_p^{place}, g_p^{pick}, g_p^{place}, P_m^{place} \rangle$
object region robot pick place pose

Partially grounded joint action : $\langle \bar{\alpha}_{R_1}, \dots, \bar{\alpha}_{R_n} \rangle \cdot \bar{\alpha}_e \setminus P_m^{place}$

Task skeleton \bar{S} : a sequence of partially grounded joint action $\xrightarrow{goal} S$ (grounded)

Swept volume : $V_{pick}(M, g, R, \xi)$, $V_{place}(M, g, R, P_m^{place}, \xi)$
object sweep robot trajectory pose

Two-Phase Method:

Phase I: Computing Collaborative Manipulation Information.

Goal : Determine all occlusion (collider-free action) and reachability information.

5 Predicates:

1. OccludePick(M_1, M_2, g, R) iff $M_1 \in V_{pick}$
2. OccludeGoalPlace(M_1, M_2, R_e, g, R) iff $M_1 \in V_{place}$
3. ReachablePick(M, g, R) R can reach M
4. ReachablePlace(M, R_e, g, R) R can reach R_e
5. EnableGoalHandler(M, g, g_s, R, R_e) iff $R \rightarrow \leftarrow R_e$ and $M \in Goal$.

Method : 1. Use inverse-kinematics solvers and motion planners to generate ξ_j .

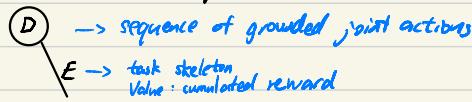
2. Minimize collision for ξ_j (minimum constraint removal). (Costly)

3. Instead, First find ξ_j for movable + fixed. If failed, only ξ_j for fixed.

Phase II. Searching for Task-and-Motion Plans

Goal: find high quality task-and-motion plans.

Search Tree:



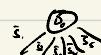
Case 1: Grounding is successful

Case 2: not successful → generate new task skeleton for objects (key component I)

Monte-Carlo Tree Search:

1. Initialize a set of task skeletons for $M \in G$

Add root node D_0



2. Selection: select E_i with highest $UCB = \left(\frac{E_i.value}{E_i.visits+1} + C \times E_i.prior \times \sqrt{\frac{D_j.visit}{E_i.visits+1}} \right)$

$$UCB = \frac{\text{cumulated reward}}{\text{# of visits}} + bias$$

3. Expansion: create a new node $D_{j,i} \rightarrow$ head node of E_i



4. Evaluation: Ground $\begin{cases} E_0 \\ D_1 \end{cases}$

i. failed, set $r=0$.

ii. found, $r = 1 + \alpha \frac{1}{|S^*|} \rightarrow$ # of objects to move

iii. found $S^* + M^*$, need to generate skeleton to move M^*

If no plan found, $r=0$

else $r = \frac{S^*.length}{S^*.length + S^*.length} + \alpha \frac{1}{|S^*| + |M^*|}$
stat time step
lets to move M^*

5. Backpropagation

Update $E_{selected}.value += r$, $num_visits++$.

Key Component I: Generating Promising Task Skeletons.

- Building the collaborative manipulation task graph

Two types of nodes: object node $M \in M$, action node \bar{a} partially grounded.

Three types of edges: action edge $\xrightarrow{(M)} \bar{a}$, block-pick edge $\xrightarrow{(M)} \bar{a}_{\text{pick}}$ block-place edge $\xrightarrow{(M)} \bar{a}_{\text{place}}$

Add Object Algorithm

For each R and grasp, construct \bar{a} and check for reachability
If M is in G , then compute potential handover action s'

for each \bar{a} , add \bar{a} to $C.actionnodes$, and $M \rightarrow \bar{a}$ to $C.actionedges$.

for every M

If $Occludes(\bar{a}, M)$, AddObject(M, c) and $C.back_pick.edges.add \bar{a} \rightarrow M$
 $H \in G$,

for every M ,

If $occupies(GoalPlace, AddObject(M, c))$, $C.back_place.edges.add \bar{a} \rightarrow M$

Result: Construct a full CMTG for later tasks.

MIP formulation and Solving

Goal: find the best plan (minimum objects to be moved) from CMTG

Formulation: minimize $\sum X_{m,a}$

subject to

Key Component 2: Task-Skeleton Grounding

• Reverse search

(not needed) (moved in first)

1. Start at time T , sample collision free placements with respect to M_{out} $U_{M_{out}}$ UV_{out}

2. Plan pick & place trajectories collision free w.r.t FUM_{out} UV_{out}

UV_{out}

If succeed \rightarrow expand V_{out} , M_{out} , and S_{out}

3. Repeat for $T-1$

4. If failed \rightarrow relax constraints M_{out} ok to collide \rightarrow near S
after all

5. Return S^* and M^* ($M \in G \cup M_{out}$)

6. If $|M^*| \neq 0$, generate new $\overset{\text{not yet}}{\overset{\text{move}}{S^*}}$, else found

7. If failed overall, return failure.