Supplementary Document

Defining action parameters for the nine common atomic actions – an explanation

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This is the supplementary document for human-robot shared assembly taxonomy (HR-SAT). More details of the taxonomy can be found at: <u>https://iai-hrc.github.io/hr-sat</u>.

In this document, we explain how to define action parameters for nine common atomic actions (see Fig. 1) defined in our paper. The action parameters are summarized in Table 1. When we define the action parameters, we follow three principles: (a) provide action parameters that a complete description of the atomic action required for both human and robot; (b) provide action parameters that common robot programming methods need; (c) consider the needs of various applications and provide comprehensive action parameters as much as possible; and (d) avoid repetitive parameter definitions.



Fig. 1. Illustrations and definitions of nine atomic actions.

Table 1. Action parameters for the nine atomic actions

| Action | Parameters | | | | | | Reference | | |
|---|--|----------|-------|----|-----|----------|-----------|--|--|
| Approach | Potential | approach | poses | of | the | subject: | [1–3] | | |
| | $[(T,R)^1,\ldots,$ | | | | | | | | |
| | - The <i>i</i> -th approach direction: $T^i = (t^i_x, t^i_y, t^i_z)$ | | | | | | | | |
| - The <i>i</i> -th approach posture: $R^i = (q_1^i, q_2^i, q_3^i, q_4^i)$ | | | | | | | | | |

| | - Reference system: MCS | |
|--------|---|---------|
| Grasp | Potential grasp poses of the subject: $[(T, R)^1,, (T, R)^n]$ | [4–11] |
| | - The <i>i</i> -th grasp direction: $T^i = (t_x^i, t_y^i, t_z^i)$ | |
| | - The <i>i</i> -th grasp posture: $R^i = (q_1^i, q_2^i, q_3^i, q_4^i)$ | |
| | - Reference system: MCS | |
| | Grasp positions on MO: $(p_1, p_2,, p_n)$ | |
| | - The <i>i</i> -th Grasp point: $p_i = (x_i, y_i, z_i)$ | |
| | - Reference system: MCS | |
| | Grasp force range: $[f_{min}, f_{max}]$ | |
| | - minimum force: f_{min} N | |
| | - maximum force: <i>f_{max}</i> N | |
| Move | Goal pose of MO: (<i>T</i> , <i>R</i>) | [1,12] |
| | - Direction: $T = (t_x, t_y, t_z)$ | |
| | - Posture: $R = (q_1, q_2, q_3, q_4)$ | |
| | - Reference system: WCS, TCS | |
| Hold | Goal pose of MO: (<i>T</i> , <i>R</i>) | [13] |
| | - Direction: $T = (t_x, t_y, t_z)$ | |
| | - Posture: $R = (q_1, q_2, q_3, q_4)$ | |
| | - Reference system: WCS, TCS | |
| Place | Goal pose of MO: (T, R) | [14,15] |
| | - Direction: $T = (t_x, t_y, t_z)$ | |
| | - Posture: $R = (q_1, q_2, q_3, q_4)$ | |
| | - Reference system: WCS, TCS | |
| Insert | Goal pose of MO: (<i>T</i> . <i>R</i>) | [16–18] |
| | - Subassembly/part pose: | [] |
| | - Direction: $T = (t_x, t_y, t_z)$ | |
| | - Posture: $R = (a_1, a_2, a_3, a_4)$ | |
| | - Reference system: TCS | |
| | - Insert-feature (e.g., pin) poses: $[(T_1, R_1), \dots, (T_n, R_n)]$ | |
| | - Direction of the <i>i</i> -th insert-feature: $T_i = (t_i, t_i, t_i)$ | |
| | - Posture of the <i>i</i> -th insert-feature: $R_i = (a_i, a_j, a_j)$ | |
| | - Reference system: TCS | |
| | Force | |
| | - Subassembly/part level force range: [f., f.,] N | |
| | - Insert-feature level force range: $[f, f]$ | |
| | -The force range of <i>i</i> -th insert-area: $f^i = [f^i, f^i]$ N | |
| Screw | Goal pose of MO: (T, R) | [19 20] |
| Serew | - Direction: $T = (t, t, t)$ | [17,20] |
| | Posturo: $P = (a, a, a, a)$ | |
| | Potoronco system: TCS | |
| | Torque range: [m m] Nm | |
| | Scrow in direction, clockwise or counterclockwise | |
| Pototo | Contact points on MQ: [n n] | [01 02] |
| Kotate | The <i>i</i> th contact points $m = (x_1, y_2, y_3)$ | [21-23] |
| | - The <i>t</i> -th contact point. $p_i = (x_i, y_i, z_i)$ | |
| | - Reference system: MC5 | |
| | Rotate angle: r | |
| | - Reference system: ICS | |
| | Goal pose of MO: (I, K) | |
| | - Direction: $I = (t_x, t_y, t_z)$ | |
| | - Posture: $R = (q_1, q_2, q_3, q_4)$ | |
| | | |

| | Rotate direction: clockwise or counterclockwise | | | | |
|-------|--|---------|--|--|--|
| Slide | Goal pose of MO: (T, R) | [24,25] | | | |
| | - Direction: $T = (t_x, t_y, t_z)$ | | | | |
| | - Posture: $R = (q_1, q_2, q_3, q_4)$ | | | | |
| | - Reference system: TCS | | | | |
| | Sliding trajectory of MO: $(p_1,, p_n)$ | | | | |
| | - The <i>i</i> -th trajectory point: $p_i = (x_i, y_i, z_i)$ | | | | |
| | - Reference system: TCS | | | | |
| | Contact points on TO: $[p_1,, p_n]$ | | | | |
| | - The <i>i</i> -th contact point: $p_i = (x_i, y_i, z_i)$ | | | | |
| | - Reference system: MCS | | | | |
| | Force trajectory: $(f_1,, f_n)$ | | | | |
| | - The <i>i</i> -th force: <i>f</i> _i N | | | | |

*WCS: word coordinate system; MCS: manipulated object coordinate system; TCS: Target object coordinate system; MO: manipulated object; TO: target object.

Approach is the process of the subject moving towards the manipulated object. To

describe this action requires the information of the subject's approaching direction and posture relative to the manipulated object [1–3], which can be described by the translation vector and rotation matrix of the goal 6D pose of the subject relative to the 6D pose of the manipulated object. The relative pose is under the manipulated object coordinate system. However, for a manipulated object, there may be multiple possible approach poses. Therefore we provide a list of potential approach poses: $[(T,R)^1,...,(T,R)^n]$, where $T^i = (t_x^i, t_y^i, t_z^i)$ is the *i*-th approach direction and $R^i = (q_1^i, q_2^i, q_3^i, q_4^i)$ is the *i*-th approach posture.

Grasp is the process of securely holding the manipulated object by the subject to

allow controlled motion of the object [4]. Successful grasps can be force-closure [5] or form-closure [6], and the key parameters are the grasp pose, grasp position and the applied force. Grasp pose includes grasp direction, which is the direction of the subject approaching the manipulated object, and grasp posture, which is the rotation of the subject relative to the manipulated object. They are all independent of the subject and manipulated object properties, and are deterministic in the action description. The grasp direction and posture can be described by the translation vector and rotation matrix of the goal 6D pose of the subject relative to the 6D pose of the manipulated object, as (*R*,*T*). For a manipulated object, there may be multiple feasible grasp poses. Therefore, we provide the potential grasp poses as a list $[(T,R)^1, ..., (T,R)^n]$. However, grasp position and applied force depend on the subject and manipulated object properties. There are various types of robot subjects, including hard or soft grippers with different fingers, suction disk and some special grippers, e.g., ball gripper and O-ring gripper [7-9]. These subjects may require different grasp positions for the same grasp task, thus the grasp positions are often untransferable between different types of robot subjects. However, the grasp position is a necessary

part of a complete grasp action description, which is indispensable in some specific cases. Therefore, the taxonomy should include the grasp position information. The grasp position can be described by the grasp point, which can be *contact point* [4] or grasp center point [10]. To meet the needs of different applications, this taxonomy only provides the data interface of grasp points, and users can flexibly define grasp points according to their own needs. The grasp point is the point on the manipulated object, and it can be described by the three-dimensional point under the manipulated object coordinate system, as $(p_1, p_2, ..., p_n)$, where $p_i = (x_i, y_i, z_i)$ is the *i*-th grasp point. Same with grasp point, we only provide the data interface of grasp force. The grasp force refers to the force that each finger applies on the object to grasp it securely without slipping or causing damage, which corresponds to the contact points [11]. The force has direction and magnitude. Since the direction is the normal vector of the contact point, it has already been constrained. We only provide the magnitude of the force. The force requirement is often a range to ensure that the object does not slip and is not damaged. Therefore, the action parameter of grasp force is a numerical range $[f_{min}, f_{max}]$ N.

Move is the process of moving the manipulated object to a certain pose. Therefore, the key descriptive information is the goal 6D pose of the manipulated object [1,12,26]. However, the pose can be absolute or relative, which depends on the specific application. If the action after Move is Hold, the object's pose is not constrained by another object, so the pose is absolute in the world coordinate. However, if the action after the Move is Insert, Screw or Place, the pose should be constrained to the relative pose of the manipulated object to the target object.

Hold is the process of holding the manipulated object in a certain pose for collaborative assembly [13]. During this period of time, holding the manipulated object may be only for waiting the completion of other assembly actions, or the held manipulated object can be taken as the target object, and other subjects can assemble other manipulated objects with it. No matter what the purpose of the action is, the parameter is the 6D pose of the object, and the pose is absolute in the world coordinate.

Place is the process of placing the manipulated object on the surface of the target object at a certain pose. Therefore, the key descriptive information is the 6D pose of the manipulated object relative to the target object [14,15,26]. The difference between

Place and Insert is that Place does not involve the mating of two objects. Therefore, there is no force constraint.

Insert is the process of placing the manipulated object onto the target object towards

a certain pose. Describing the action of inserting the manipulated object onto the target object requires the information of the relative pose between the two objects after the insertion, as well as the force range required during the insertion [16–18,26]. However, there may be multiple insert-features (e.g., pins) corresponding to multiple inserted-features (e.g., holes) on one subassembly/part. Therefore, when describing the relative pose, there are two levels: one is subassembly/part level, which describes the 6D pose of manipulated subassembly/part relative to target subassembly/part as (T, R), and the other is assemblable feature level, which describes the 6D pose of each manipulated insert-feature relative to the corresponding target inserted-feature as $[(T_1, R_1), ..., (T_n, R_n)]$. Similarly, the description of the applied force range should also have two levels: the subassembly/part level $[f_{min}, f_{max}]$ N and the assemblable feature level $[f_1, ..., f_n]$, where $f^i = [f^i_{min}, f^i_{max}]$ N. The direction of the force is along the axial of the hole, so it is only necessary to describe the magnitude of the force. The reason why the requirements of pose and force are divided into two levels is that applying the insert force at one time and at one point may not ensure that all insertfeatures are inserted successfully. Therefore, the hierarchical representation can facilitate the inspection of the insertion results, and it is also convenient for users to manage the force distribution according to their own robot capability.

Screw is the process of rotating the manipulated object to fit into the target object by

means of threads to fasten parts together. Threaded object can be categorized into inner-threaded object and outer-threaded object, but both of them can be manipulated object or target object. There may be multiple screw-in/screwed-in features on one subassembly/part. To completely describe a Screw action, it requires the information of the goal 6D pose of the manipulated object relative to target object [19], as (*T*, *R*). The purpose of Screw is to fasten two/more objects together. To ensure the fastening quality, it is usually necessary to specify the preload [20]. The preload acting on the manipulated object and the target object is a pair of forces with the same magnitude and opposite direction. Since the direction is along the central axis of the thread, it has been constrained already. Only the magnitude of the preload needs to be constrained, so the preload requirement here is a constant. Because controlling torque is easier than controlling force when screwing, the preload requirement can be constrained by the tightening torque acting on the manipulated object, and it is also a constant. Because the torque requirement usually is a torque range, we provide the torque range as $[m_{min}, m_{max}]$ Nm. In addition, another important descriptive information is the screw-in direction. Threads can be lefthanded or right-handed, thus the screw-in direction can be clockwise or counterclockwise. The screw-in direction is also the direction of applying the torque.

Rotate is the process of rotating the manipulated object to a certain pose. Unlike screw, the purpose of rotate is not to fasten parts together. The key descriptive information is the goal pose of the manipulated object relative to the target object [19,21,23], as (*T*,*R*). It is sometimes necessary to describe the contact points between the subject and the manipulated object. The contact points are the three-dimensional points in the manipulated object coordinate system, as ($p_1, p_2, ..., p_n$), where $p_i = (x_i, y_i, z_i)$ is the *i*-th contact point. Another important descriptive information is the rotate angle – the angle of rotating the manipulated object. The rotate angle is a constant *r*.

Slide is the process of moving the manipulated object to a certain pose along the surface of the target object. The key descriptive information is the goal pose of the manipulated object, sliding trajectory, contact points and sliding force [24,25]. The goal pose is the 6D pose of the manipulated object relative to the target object, as (T, R). The sliding trajectory is a line on the surface of the target object, which can be represented by a series of three-dimensional points in the target object coordinate system, as $(p_1, p_2, ..., p_n)$, where $p_i = (x_i, y_i, z_i)$ is the *i*-th trajectory point. On this trajectory, the initial position of the object can be anywhere, but the goal position is fixed. The contact points between the subject and the manipulated object can be represented by the three-dimensional points in the manipulated object coordinate system, as $(p_1, p_2, ..., p_n)$, where $p_i = (x_i, y_i, z_i)$ is the *i*-th contact point. The direction of the sliding force applied on the manipulated object should be along the tangent direction of the sliding trajectory. Since the magnitude of the force required at different positions on the sliding trajectory may be different, the data interface we provide is a force trajectory, which means each trajectory point corresponds to a force magnitude. We provide the force trajectory as $(f_1, ..., f_n)$ where f_i N is the force magnitude of the *i*-th force trajectory point.

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