



# **Grasping and Manipulating Moving Objects in the Presence of Moving Obstacles**

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**ABSTRACT:** In this paper, we proposed a motion planner with robot controller of the 7 degree-of-freedom Barrett WAM to grasp a moving object in the presence of moving obstacles without any prior knowledge of the surrounding environments. The motion planner is based on a randomized planning algorithm capable of planning. This algorithm is an extension of RRT-JT which uses the Jacobian-based gradient descent to instruct a 7-DoF WAM robotic arm, in order to grasp a moving target, while avoiding moving obstacles. The robot's trajectory is generated by the distances between the target, obstacle and the end-effector. We present a simulation of a scenario that starts with tracking a moving mug while avoiding moving obstacles then grasping it and finally placing the mug in a determined position. The proposed algorithm successfully grasped a moving object in a reasonable time and it is assuring a maximum rate of success.

**KEYWORDS:** Grasping, trajectory, moving obstacles, moving object, planning, robot hand.

## **I. INTRODUCTION**

Nowadays, techniques in robotics are getting mature and people expect robots to do more complex tasks in human daily life. The problem of grasping a moving object in the presence of obstacles with a robotic manipulator has been reported in different works. There have been many studies on grasping motion planning for a manipulator to avoid obstacles [1]. In high-level services, the ability of the robot to grasp objects is essential and the motion planning of controlling the manipulator to grasp the objects without collision has been studied for a long time. The sampling-based planning algorithms such as Rapidly-exploring trees (RRTs) [2] and Probability Roadmap (PRM) [3] algorithms are popular in recent years because of their ability to rapidly discover the connectivity of high-dimensional configuration spaces. However, the goal of the manipulator during human-robot interaction is not stationary but dynamic. In this paper, to increase the robot's ability of interaction with human and environments, we proposed an approach for the Barrett WAM to grasp a moving object in the presence of movable obstacle in the environment.

The whole system is composed of three parts: (1) Perception (2) Motion planner (3) Robot control. In the part of perception, we use the object's position to detect and track the target. Motion planner is based on the RRT-JT algorithm which is used to generate the trajectory which controls the robot motion in joint with considering the motion of the obstacle.

The problem of grasping a moving object in the presence of obstacles with a robotic manipulator has been reported in different works. There have been many studies on grasping motion planning for a manipulator to avoid obstacles [4], [5]. There are many researchers who are interested in motion planning under unknown dynamic environments [6] [7] [8]. Many works calculate the trajectories and represent the trajectories as the parametrized collision-free path. The path is updated as the environment changes at runtime. The elastic strip method [7] builds a collision-free "tunnel" which connects the initial and goal of the robot for avoid moving obstacles by deforming a trajectory. [8] introduces an approach which is based on PRM but considering prior-information of static and dynamic obstacles for dealing with



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unknown moving obstacles for mobile robots. In [9] we presented a simulation of a scenario that starts with tracking a moving mug, then grasping it while avoiding static obstacles.

In this work, we present both cases of grasping static and moving objects in the presence of moving obstacles in the environments by using the RRT-JT. The environment doesn't to be known or built models in advance. All we require to know are the initial position of the moving object and the initial position of the moving obstacle. This paper presents a motion planning and controlling an arm of a humanoid robot for grasping and manipulating of a moving object without cameras. We used an algorithm to control the end effector pose (position and orientation) with respect to the pose of objects which can be moved in the workspace of the robot.

The paper is organized as follows. A description of the Rapidly-Exploring Random Trees (RRT) is detailed in section 2. Then, in section 3, we give a brief overview of the environment used in this paper. The next section contains a description of the WAM™ arm. In Section 5 contains Target and obstacle velocities. In Section 6, we present the robot motion control. Finally, Section 7 some results are given, then the conclusions and future works.

## II. RAPIDLY-EXPLORING RANDOM TREES (RRT)

In previous work [10],[11], approaching the motion planning problem was based on placing the end effector at pre-configured locations, computed using the inverse kinematics(IK) applied to some initial samples taken from the goal region. These locations are then set as goals for a randomized planner, such as an RRT or BiRRT [12], [13]. The solution presented by this approach remains unfinished because of the miss considered probabilistic aspect. The issue is that the planner is forced to use numbers priori chosen from the goal regions.

Another way to tackle the grasp planning, certain types of workspace goals, is to explore the configuration space of the robot with a heuristic search tree, and try to push the exploration toward one goal region [14]. Nevertheless, the goal regions and heuristics presented in [15] are highly problem specific to generalize and tricky to adjust. Drumwright and Ng-Thow-Hing [16] employ a similar strategy of extending toward a randomly-generated IK solution for a workspace point. In [17], Vande Weghe et al. present the RRT-JT algorithm, which uses a forward-searching tree to explore the C-space and a gradient-descent heuristic based on the Jacobian-transpose to bias the tree toward a work-space goal point. [18] present two probabilistically complete planners: an extension of RRT-JT, and a new algorithm called IKBiRRT. Both algorithms function by interleaving exploration of the robot's C-space (Configuration space) with exploitation of WGRs (Workspace Goal Regions). The extended RRT-JT (Fig. 2) is designed for robots that do not have such algorithms and is able to combine the configuration space exploration of RRTs with a workspace goal bias to produce direct paths through complex environments extremely efficiently, without the need for any inverse kinematics.

## III. SYSTEM OVERVIEW

We use the Open Robotics Automation Virtual Environment (OpenRAVE) which is primarily used for storing the environment representation of the manipulators, sensor positions, and other scene models. We use the OpenRAVE software to control both the WAM arm and Barrett hand.

Our goal is to grasp the moving object, which is our target in the presence of moving obstacle in the environments. The approach in this paper uses a RRT-JT to determinate the object's position and to calculate the kinematics of the robot while avoiding obstacle.

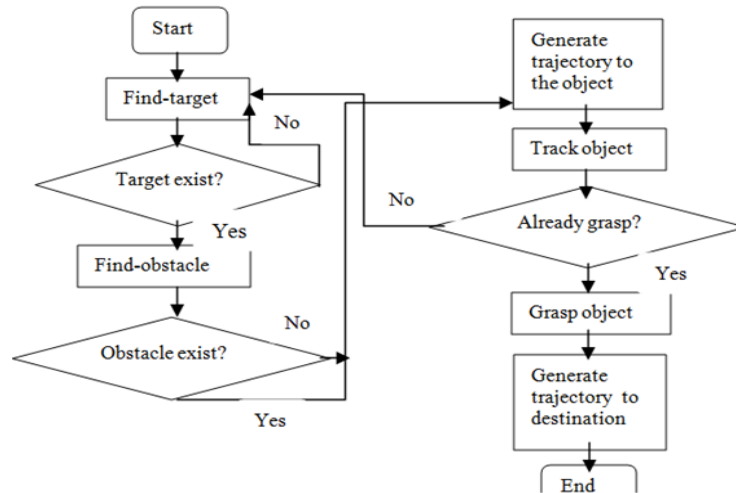


Fig. 1 Flow chart of the local motion planner.

The flow chart of the motion planner is as in Fig. 1. First, we attempt to locate the target, if the target exists, we attempt to locate the obstacle, if it exists, we command the robot to displace to the target while avoiding the existing obstacle. Finally, the robot grasps the target and makes it to the desired position.

#### IV. THE WAM™ ARM

The WAM arm is a robotic back drivable manipulator. It has a stable joint-torque control with a direct-drive capability. It offers a zero backlash and near zero friction to enhance the performance of today’s robots. It comes with three main variants 4-DoF, 7- DoF, both with human-like kinematics, and 4-DoF with 3-DoF Gimbals. Its articulation ranges go beyond those for conventional robotic arms [20].

We use WAM 7-DoF Arm with attached Barrett Hand.

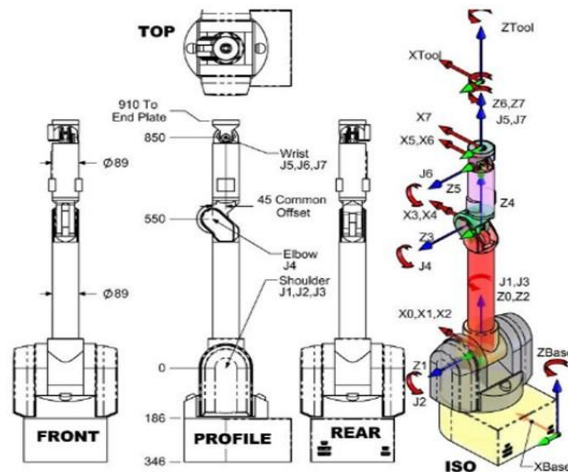


Fig. 2 WAM 7-DoF dimensions and D-H frames, [21].

Fig. 2 presents the whole 7-DoF WAM system in the initial position. A positive joint motion is on the right hand rule, for each axis. The following equation of homogeneous transformation in Fig. 3 is used to determine the transformation between the axes K and K-1.



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$${}^{k-1}T_k = \begin{bmatrix} \cos\theta_k & -\sin\theta_k \cos\alpha_k & \sin\theta_k \sin\alpha_k & a_k \cos\theta_k \\ \sin\theta_k & \cos\theta_k \cos\alpha_k & -\cos\theta_k \sin\alpha_k & a_k \sin\theta_k \\ 0 & \sin\alpha_k & \cos\alpha_k & d_k \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Fig. 3 D-H generalized transform matrix.

- $a_{k-1}$  = the distance from  $Z_{k-1}$  to  $Z_k$  measured along  $X_{k-1}$
- $d_k$  = the distance from  $X_{k-1}$  to  $X_k$  measured along  $Z_k$
- $\alpha_{k-1}$  = angle between  $Z_{k-1}$  to  $Z_k$  was approximately  $X_{k-1}$
- $\theta_k$  = angle between  $X_{k-1}$  to  $X_k$  was approximately  $Z_k$

The Table I contains the parameters of the arm with 7-DoF

Table I.7-DoF WAM frame parameters

$K$	$a_k$	$\alpha_k$	$d_k$	$\theta_k$
1	0	$-\pi/2$	0	$\theta_1$
2	0	$\pi/2$	0	$\theta_2$
3	0.045	$-\pi/2$	0.55	$\theta_3$
4	-0.045	$\pi/2$	0	$\theta_4$
5	0	$-\pi/2$	0.3	$\theta_5$
6	0	$\pi/2$	0	$\theta_6$
7	0	0	0.060	$\theta_7$
T	0	0	0	

As with the previous example, we define the frame  ${}^7T_{Tool}$  for our specific end effector. By multiplying all of the transforms up to and including the final frame, we determine the forward kinematics for any frame on the robot. To determine the end tip location and orientation we use the following equation:

$${}^0T_{Tool} = {}^0T_1 {}^1T_2 {}^2T_3 {}^3T_4 {}^4T_5 {}^5T_6 {}^6T_7 {}^7T_{Tool}$$

The transformation equations used to update the manipulator's joints until the distance between the end effector and the moving object almost equal to zero. Once the position of the contact is achieved, the Barrett hand closes its fingers and grasps the object.

## V. TARGET AND OBSTACLE VELOCITIES

From the first position  $a(x_1, y_1, z_1)$  to the second position  $b(x_2, y_2, z_2)$  which  $x_2 = x_1 + \Delta_x$ ,  $y_2 = y_1 + \Delta_y$  and  $z_2 = z_1 + \Delta_z$  and  $t_{a \rightarrow b} = t_{sleep} + \epsilon$  which  $t_{sleep}$  is the time to sleep en  $a$  and  $\epsilon$  is the time from  $a$  to  $b$ )

$V_x = \frac{\Delta_x}{t_{a \rightarrow b}}$ ,  $V_y = \frac{\Delta_y}{t_{a \rightarrow b}}$  and  $V_z = \frac{\Delta_z}{t_{a \rightarrow b}}$  so the target's velocity  $V_T = V_x i + V_y j + V_z z$  and the obstacle's velocity  $V_O = V_x i + V_y j + V_z z$

## VI. ROBOT MOTION CONTROL

### A) Robot Dynamics

The dynamic simulation uses the Lagrange equations to get the angular acceleration from the torque of each joint. First, I computed the body Jacobian of each joint  $J_i$  corresponding to  $M_i$ , where  $M_i$  is the joint's inertia matrix. So the manipulator inertia matrix  $M(\theta)$  can be calculated as



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$$M(\theta) = \sum_{i=1}^n J_i^T(\theta) M_i J_i(\theta)$$

Also calculate potential part,

$$P(\theta) = \sum_{i=1}^n m_i g h_i(\theta)$$

Second, I computed the torques of each joint using Lagrange equation, which is,

$$\frac{d}{dt} \frac{\partial L}{\partial \dot{\theta}} - \frac{\partial L}{\partial \theta} = \tau$$

where  $\tau$  represents the torque of the joint and,

$$L(\theta, \dot{\theta}) = \frac{1}{2} \dot{\theta}^T M(\theta) \dot{\theta} - P(\theta)$$

After expanding the components of the equation, the equation is as following:

$$M(\theta)\ddot{\theta} + H(\theta, \dot{\theta})\dot{\theta} + G(\theta, \dot{\theta}) = \tau$$

Where  $H(\theta, \dot{\theta})$  is the Coriolis and centrifugal force term and  $G(\theta, \dot{\theta})$  is the gravity term and

$$\theta = [\theta_1 \theta_2 \theta_3 \theta_4 \theta_5 \theta_6 \theta_7]^T \text{ and } \dot{\theta} = \frac{d\theta}{dt}, \ddot{\theta} = \frac{d^2\theta}{dt^2}.$$

## B) Using the Jacobian

Given a robot arm configuration  $q \in Q$  (the configuration space) and a desired end effector goal  $x_g \in X$ , where  $X$  is the space of end effector positions  $R^3$ , we are interested in computing an extension in configuration space from  $q$  towards  $x_g$ . Although the mapping from  $Q$  to  $X$  is often nonlinear and hence expensive to deduct, its derivative the Jacobian, is a linear map from the tangent space of  $Q$  to that of  $X$ , that can be computed easily ( $J \dot{q} = \dot{x}$ , where  $x \in X$  is the end effector position (or pose) corresponding to  $q$ ). Ideally, to drive the end effector to a desired configuration  $x_g$ , ( $d x_g / dt \approx 0$ : object moves slowly) we could compute the error  $e(t) = (x_g - x)$  and run a controller of the form  $\dot{q} = KJ^{-1}e$ , where  $K$  is a positive gain. This simple controller is capable to attain the target without considering any possible barriers or articulation limits. However this turn into a complex controller, where the inverse of the Jacobian must be done at each time step. To escape this expensive approach, we use alternatively the transpose of the Jacobian and the control law fall into the form of  $\dot{q} = KJ^T e$ . The controller eliminates the large overhead of computing the inverse by using the easy-to-compute Jacobian instead. It is easy to show that, under the same obstacle-free requirements as the Jacobian inverse controller, the Jacobian transpose ( $JT$ ) controller is also guaranteed to reach the goal. The instantaneous motion of the end effector is given by  $\dot{x} = J\dot{q} = J(KJ^T e)$ . The inner product of this instantaneous motion with the error vector is given by  $e^T \dot{x} = ke^T J J^T e \geq 0$ . As this is always positive, under our assumptions with obstacles, we may ensure that the controller will be able to make onward progress towards the target [17].

## VI. RESULTS AND ANALYSIS

### A) GRASPING OBJECT

To demonstrate and illustrate the proposed procedure, we present an example which the robot is equipped with a 7-DoF arm (see Fig. 3) and a three-fingered Barrett hand. The goal is to follow a moving model mug, stably holding it, pick it up and move it to the desired position while avoiding the existing moving obstacles which is a complex obstacle which contains three coplanar cylinders, the size of the object is smaller than the width of Barrett hand. We assume that the object remains inside the workspace of the manipulator and it moves slow enough to be tracked and approached by the manipulator, the obstacle moves between the manipulator and the object. The mug was moving in a trajectory in the space with velocity range 8-16 cm/s (for handicapped man) slower than the velocity used by normal persons (velocity is 21 cm/s for a normal man as [22]). The obstacle was moving in a trajectory in the space with velocity range 8-16 cm/s. The initial positions of the end effector were (-0.730m, 0.140m, 2.168m) and those of the moving object were (-0.005m, -0.200m, 1.105m). In order to grasp the moving object stably and move it, the robot hand avoids obstacle and reaches the object than it closes its fingers. The end effector would move right or left with the shorter distance.

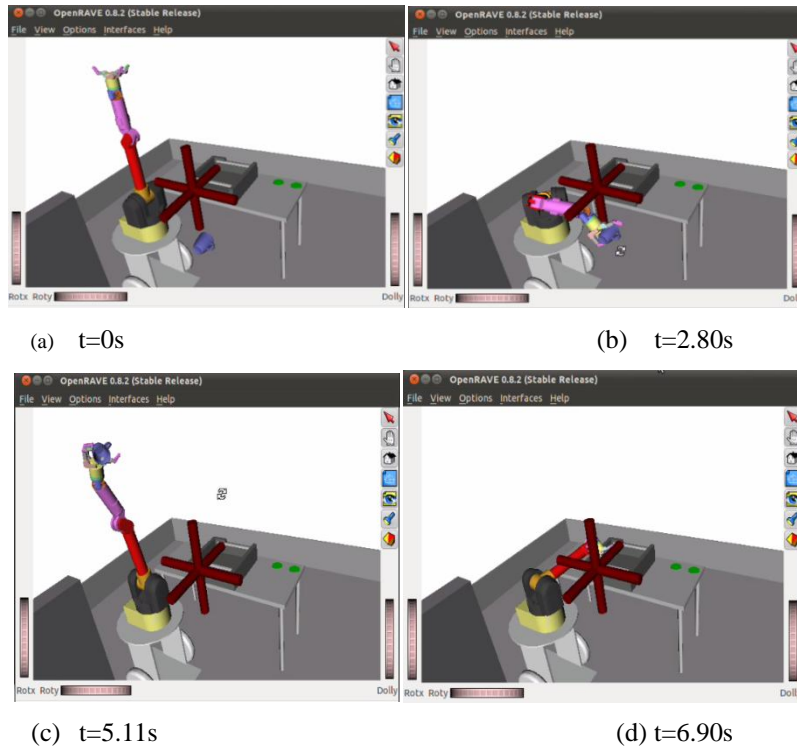


Fig. 4. Example of a successful case of grasping a moving object

As mentioned in Fig. 4, the grasping of the object is done successfully. Fig. 4.a show that the hand of the robot keeps at a distance from the object, the Barrett hand and the object are in the initial position, Fig. 4.b the object moves with the velocity  $V_1 = 8\text{cm/s}$  and the robot moves to the position of the centroid of the object that the center of the Barrett hand should coincide with the position on the object which is closest to the Barrett hand, opens the fingers, closes the fingers and finally grasps the object. In Fig. 4.c the robot picks up the object and moves it to the desired position. In Fig. 4.d the robot puts the target in the desired position and opens his fingers.

To capture the moving object safely and to lift it up stably without slippage, the end effector needs to be as controlled as the relation between their position and the object 'ones'. So they determine the actual position of the moving object, and pick the closest distance between the end effector and the target.

### B) Object and obstacle have the same velocity

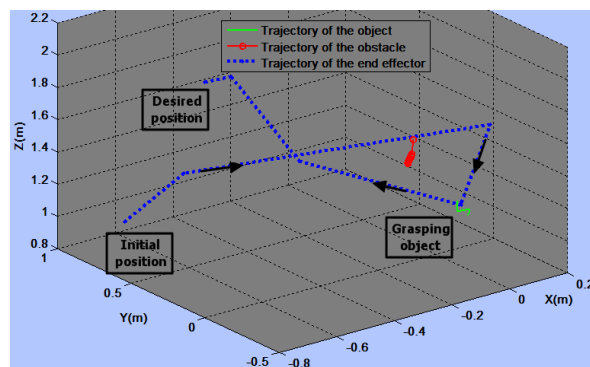


Fig. 5. The trajectory of the object, the obstacle and the end effector.

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Fig. 5 illustrates the curves of the robot, the obstacle and the object: the robot avoids the moving obstacle which moves in the space with a velocity  $V=8\text{cm/s}$  and grasps the object in time  $T_{\text{grasp}}=2.75\text{ s}$ , which moves in the space with velocity  $V=8\text{cm/s}$ . The robot puts the object in the desired position in time  $T_{\text{end}}=6.75\text{ s}$ .

Table II. Object and obstacle move with  $V=8\text{cm/s}$

	$T_{\text{grasp}}(\text{s})$	$T_{\text{end}}(\text{s})$
Trial1	2.75	6.75
Trial2	2.81	6.49
Trial3	3.09	6.92

The medium of grasp time in table II is  $T=2.88\text{ s}$

### C) The object's velocity double of the obstacle's velocity

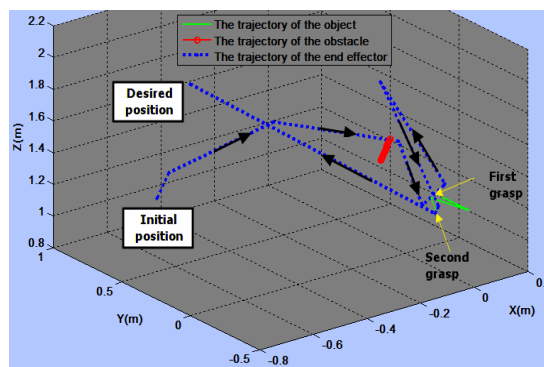


Fig. 6. The trajectory of the object, the obstacle and the end effector.

Fig. 6 illustrates the curves of the robot, the obstacle and the object: the robot avoids moving obstacles which moves in the space with velocity  $V=8\text{cm/s}$  and grasps the object in first grasp in time  $T_{\text{grasp}1}=2.95\text{ s}$ , which moves in the space with velocity  $V=16\text{cm/s}$ , but it's unstably grasp, so the robot opens its fingers and change its trajectory, it repeats a second grasp in  $T_{\text{grasp}2}=5.30\text{ s}$ . The robot puts the object in the desired position in time  $T_{\text{end}}=9.15\text{ s}$ .

Table III. Object moves with  $V=16\text{cm/s}$  and obstacle moves with  $V=8\text{cm/s}$

	$T_{\text{grasp}}(\text{s})$	$T_{\text{end}}(\text{s})$
Trial1	2.95-5.30	9.15
Trial2	3.66	8.64
Trial3	3.35	7.89

The medium of grasp time in table III is  $T=4.10\text{ s}$

If we double the velocity of the object, we remark that the first time to grasp is about the double.

If we increase the velocity of the object, the robot finds some difficulty in grasping the moving object. Therefore, increasing the speed affects on the time of grasping the moving object, even the direction of movement of the object affects on the time of grasping.

As shown in the tables, a failure of grasping in the first trial, but in the second trial the robot learns the trajectory and our robot successfully grasped the objects. We demonstrate that the robot is able to grasp the moving object in a reasonable time. The presence of the moving obstacle idling speed of the robot and the increasing of the velocity of the object complicates the grasping.

**D) The obstacle’s velocity double of the object’s velocity**

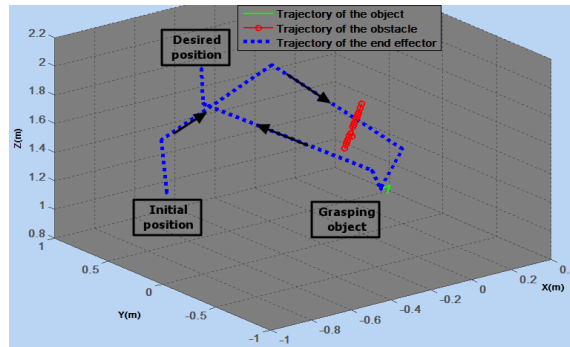


Fig. 1. The trajectory of the object, the obstacle and the end effector.

Fig. 7 illustrates the curves of the robot, the obstacle and the object: the robot avoids moving obstacles which move in the space with velocity  $V= 16\text{cm/s}$  and grasps the object in time  $T_{\text{grasp}}= 3.20 \text{ s}$ , which moves in the space with velocity  $V= 8\text{cm/s}$ . The robot puts the object in the desired position in time  $T_{\text{end}}= 7.13 \text{ s}$ .

Table IV. Object moves with  $V=8\text{cm/s}$  and obstacle moves with  $V= 16\text{cm/s}$

	$T_{\text{grasp}}(\text{s})$	$T_{\text{end}}(\text{s})$
Trial1	3.20	7.13
Trial2	2.93	7.50
Trial3	3.13	7.05

The medium of grasp time is  $T=3.08 \text{ s}$ .

The table IV presents results separately of the time for grasping the moving object while avoiding obstacles, and the time to move the object to the desired position. Times are nigh in the different trial.

If we double the velocity of the obstacle, we see that the results are nigh but slightly higher. Therefore, increasing the speed of the obstacle affects on the time of grasping the moving object, even the direction of movement of the object affects on the time of grasping. We demonstrate that the robot is able to grasp the moving object in a reasonable time.

**VI.CONCLUSIONS**

We have presented a simulation of grasping a moving object while avoiding a moving obstacle with different velocities and move it to a desired position while avoiding obstacles using the 7-DoF robotic arm with the Barrett hand in which we involve the RRT algorithm. In fact, this algorithm allows us overcome the problem of the inverse kinematics by exploiting the nature of the Jacobian as a transformation from a configuration space to workspace.

We set forth separately the time of grasping the moving object shifting with different velocity in the presence of a moving obstacles and the time to put this object in a desired position. Firstly, it moves with velocity  $V_1=8\text{cm/s}$ . Second, it moves with velocity  $V_2=2V_1$ . The proposed algorithm successfully holding the moving object in a rational time putting it in the aim station.

Times are nigh in the different trial. The presence of moving obstacles increases the speed of grasping the object. The times recorded in which the velocity of the object is upper than the obstacle are slightly higher.

In this article, we have studied an algorithm dedicated for grasping a moving target while trying to escape a moving obstacle with Barrett hand. A future work will tend to increase the velocity of the object and the obstacle.

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