Regrasping by Fixtureless Fixturing

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Abstract—This paper presents a fixturing strategy for regrasping that does not require a physical fixture. To regrasp an object in a gripper, a robot pushes the object against external contact/s in the environment such that the external contact keeps the object stationary while the fingers slide over the object. We call this manipulation technique fixtureless fixturing. Exploiting the mechanics of pushing, we characterize a convex polyhedral set of pushes that results in fixtureless fixturing. These pushes are robust against uncertainty in the object inertia, grasping force, and the friction at the contacts. We propose a sampling-based planner that uses the sets of robust pushes to rapidly build a tree of reachable grasps. A path in this tree is a pushing strategy, possibly involving pushes from different sides, to regrasp the object. We demonstrate the experimental validity and robustness of the proposed manipulation technique with different regrasp examples on a manipulation platform. Such fast and flexible regrasp planner facilitates versatile and flexible automation solutions.

I. INTRODUCTION

The need for regrasp is ubiquitous in assembly operations. In-hand manipulation skills allow one to pick up a part, reorient it in the hand, and then use/insert/assemble it. Robots, especially those with parallel-jaw grippers, lack the dexterity to autonomously adjust the grasp on an object. They are often assisted by systems such as part feeders and fixtures. Part feeders take care of reorienting the parts, while customized fixtures allow robots to regrasp through place-and-pick maneuvers. This approach leads to highly-customized, non-flexible, and expensive automation solutions.

This paper presents an in-hand manipulation technique that allows a robot with a simple parallel-jaw gripper to regrasp an object without any customized hardware. The robot pushes the object against external contact/s in the environment such that the object sticks to the external contact while the fingers slide over the object. Since the object is held stationary in the environment without any fixtures, we call this approach fixtureless fixturing.

Fixtureless fixturing is a variant of prehensile pushing that is robust against uncertainties in the system parameters. In our prior work on prehensile pushing, we present an idea of manipulating an object with external pushes [1, 2, 3, 4]. However, there we assume precise knowledge of system parameters such as the object inertia, grasping force, and friction coefficients at the fingers and at the external contacts. In practice, there is always some uncertainty in the knowledge of these parameters. Based on the mechanics of pushing, we define in this paper a subspace of robust prehensile pushes and refer to it as a robust motion cone. We show that the robust motion cone is invariant to all the system parameters listed above except friction at the external contact/s.

We present a sampling-based planning framework that uses robust motion cones to build long pushing strategies for regrasping. Exploiting the knowledge of the motion cones for dynamics propagation, the planner rapidly builds a tree of reachable grasps. A path in this tree is a pushing strategy, possibly involving pushes from different sides, to manipulate the object in the grasp. These pushing strategies force the object to the desired pose in the grasp irrespective of any uncertainty in the object inertia, grasping force, friction at the fingers. Fig. 1 shows one such pushing strategy to manipulate a T-shaped object.

For experimental validation, we consider different examples of regrasping performed on a manipulation platform well-instrumented to capture the pose of the object and the
robot. We show that the pushing strategies generated by our planner consistently achieve the desired regrasp and are robust against the uncertainties in the system parameters.

The efficient computation of the motion cones and their application to propagate the dynamics allows our planner to generate pushing strategies in less than a second. Moreover, the planner can use and reuse a simple feature in the environment to manipulate many different objects. Such a fast and robust framework for regrasping by fixtureless fixturing takes us a step closer towards a versatile and flexible method for in-hand manipulation for industrial automation.

II. RELATED WORK

Fixturing is one of the widely used practices in industry for applications such as machining, assembly, and inspection. Researchers have extensively studied the principles and designs for fixturing [5, 6, 7]. However, the role of friction in fixturing is often neglected or considered secondary to that of the kinematics.

One of the very few works that describe the use of force analysis with friction in the context of fixture planning is by Hong and Cutkosky [8]. They point out that fixture planning, similar to grasp planning, is primarily considered from the kinematic perspective only. They advocate the analysis of the mechanics of fixturing and use the limit surface model [9] to capture the force-motion relationship at the contacts between the object and the fixture. Their work focuses on the stability of a fixtured part against the forces exerted by a tool and provides estimates for the clamping force required to prevent the part from sliding. Using similar contact mechanics tools, the work presented in this paper studies the controlled sliding of the object in the grasp. With the fixtureless fixturing, we demonstrate the application of the mechanics of frictional contact for fixturing.

The work presented in this paper is motivated by our recent work on stable prehensile pushing [4]. In that work, we present a planner to regrasp an object in a gripper using the external pushes for which the object sticks to the pusher. To find feasible stable prehensile pushes, the planner evaluates a dynamics check for multiple random push directions. This dynamics check depends on the knowledge of the object inertia, grasping force, and coefficient of friction at the fingers and at the features in the environment. Consequently, the stable prehensile pushing is sensitive to these parameters. In contrast, the fixtureless fixturing presented in this paper uses only the stable prehensile pushes whose dynamics is invariant to these parameters except for the friction at the features in the environment. Moreover, we demonstrate a method to compute bounds on the valid robust stable pushes in the form of robust motion cones. The knowledge of the motion cones avoids the need of dynamics check and speeds up the regrasp planning.

The idea of robust motion cones is inspired by the original concept of the motion cone presented by Mason [10]. Mason studies the mechanics of pushing a planar object on a horizontal plane with a point pusher. He proposes the motion cone as a set of pushes that lead to sticking contact between the pusher and the object. Our recent work [11] extends the idea of motion cones to prehensile tasks in a more general setting and demonstrates their application for fast in-hand manipulation planning. The application of robust motion cones for regrasp planning brings robustness to in-hand manipulation planning.

III. PROBLEM FORMULATION

We present a robust in-hand manipulation technique using a simple parallel-jaw gripper. An object in a gripper is forced to a desired pose in the grasp using prehensile pushes. We implement the prehensile pushes by pushing the object against features in the environment. If provided with a dual-arm robot or a multi-finger gripper, the second arm or the extra fingers of the gripper can be used for the pushes.

We propose a manipulation planning framework that is compatible with several of the practical implementations for the prehensile pushes. We consider that an object is grasped in a gripper that is fixed in the world. A moving pusher pushes the object to a new pose in the grasp. In our implementation, the pusher motion is a reflection of the robot motion against the fixed environment. Now, planning the robot motion to regrasp an object is equivalent to finding a pushing strategy that forces the object to the desired pose.

We assume the following knowledge about the physical properties of the manipulation system:
• Object geometry.
• Initial and goal pose of an object in a grasp, specified by the locations and geometries of each fingers contacts.
• Discrete set of pusher contacts, specified by their locations and geometries.
• Coefficients of friction at the pusher contacts (an approximate value is sufficient).

Our proposed planning framework works at two levels. At the high-level, a T-RRT* based planning architecture samples the configuration space of different grasps, similar to our work in [4]. At the low level, the planning tree is grown in the direction of the sampled poses using the knowledge of local reachable poses in the form of robust motion cones. Exploiting the efficient dynamics propagation via motions cones, the planner rapidly explores the configuration space and generates feasible pushing strategies to move the object in the grasp.

IV. MECHANICS OF FIXTURELESS FIXTURING

In this section, we start with a brief review on fundamental contact modelling techniques that allow us to model the force-motion interaction during prehensile pushing. Based on the contact modelling in Section [IV-A] and dynamics of stable prehensile pushing in Section [IV-B] we discuss the mechanics of fixtureless fixturing and characterize the set of fixtureless fixturing pushes in Section [IV-C] and [IV-D]

A. Contact Modelling

In this paper, we assume that all bodies are rigid and all contacts have dry isotropic friction.

1) Limit Surface: Goyal [9] presented the limit surface as the boundary of the friction wrenches (force-torque pairs) that a contact with finite area can provide. Howe and Cutkosky [12] and Xydas and Kao [13] showed that the limit surface geometry can be approximated as an ellipsoid. We consider circular patch contacts at the fingers. To model the force-motion interaction at the finger contacts, we assume the ellipsoidal approximation to the limit surface, which has been shown to be computationally efficient for pushing motions [14] [15] [16] [17]

Let \( \mathbf{w}_c = [f_x, f_z, n_y] \) be a frictional wrench at a finger contact in the contact frame. The ellipsoidal limit surface constraint can be written as: \( \mathbf{w}_c^T A \mathbf{w}_c = 1 \), where \( A = \text{Diag}(a_1^2, a_2^2, a_3^2) \). For isotropic friction, the maximum friction force is \( a_1 = a_2 = \mu N \), where \( \mu \) is the friction coefficient, and \( N \) is the normal force at the contact. The maximum friction torque about the contact normal is \( a_3 = r c \mu N \), where \( r \) is the radius of the finger contact and \( c \in [0,1] \) is the constant from numerical integration. For a uniform pressure distribution, \( c \) is about 0.6 [13] [16]. When the object slides on the contact, the friction wrench lies on the limit surface such that the normal to the limit surface is in the direction of the object motion at the contact. If the object twist (linear and angular velocity pair), \( \mathbf{v}_{\text{obj}} = [v_x, v_z, \omega_y]^T \), is known, we can find the friction wrench at the contact as:

\[
\mathbf{w}_c = \frac{A^{-1} \mathbf{v}_{\text{obj,c}}}{\sqrt{\mathbf{v}_{\text{obj,c}}^T A^{-1} \mathbf{v}_{\text{obj,c}}}} = \mu c N \mathbf{w}_c
\]

where \( \mathbf{v}_{\text{obj,c}} \) is the twist of the object in the contact frame. Here, \( \mathbf{v}_{\text{obj,c}} = [v_x, v_z, \omega_y]^T = J_c \cdot \mathbf{v}_{\text{obj}} \), where \( J_c \) is the Jacobian that maps the velocity from the object frame to the contact frame. \( \mathbf{w}_c = [f_x, f_z, m_y]^T \) is a unit wrench on the limit surface, which defines the motion direction, sets the frictional force at the contact (\( \mu N \)) to estimate the net frictional wrench.

For the ellipsoidal limit surface assumption, the linear velocity \( [v_x, v_z]^T \) of the object in the contact frame is parallel and opposite to the frictional force \( f_x, f_z \) applied by the contact in the contact frame [18]. Moreover, the relationship between the friction wrench and the normal to the limit surface, which defines the motion direction, sets the following constraint between the angular velocity at the contact and the linear velocity:

\[
\frac{v_x}{\omega_y} = \frac{(rc)^2 f_x}{m_y} \quad \text{and} \quad \frac{v_z}{\omega_y} = \frac{(rc)^2 f_z}{m_y}
\]

Given the friction wrench on the object from the contact (\( \mathbf{w}_c \)), we can find the object velocity as:

\[
\mathbf{v}_{\text{obj}} = k J_c \cdot \mathbf{w}_c \quad \text{and} \quad \mathbf{B} = \text{Diag}(1, 1, (rc)^{-2}) \quad k \in \mathbb{R}^+ \quad (2)
\]

where \( J_c \) maps the object velocity from the contact frame to the object frame.

2) Generalized Friction Cone: Erdmann [19] introduced the concept of the generalized friction cone \( (\mathbf{W}) \) as a representation of the local Coulomb friction cone at a contact in the wrench-space and in the object frame. The generalized friction cone for a patch contact modelled with multiple point contacts is the convex hull of the generalized friction cones for each constituent contact [20]. We model the friction at a pusher contact with its generalized friction cone:

\[
\mathbf{W}_{\text{pusher}} = \{ \mathbf{w}_{\text{pusher}} = J_p^T \cdot \mathbf{f}_p \mid \mathbf{f}_p \in FC_{\text{pusher}} \} \quad (3)
\]

Here, \( f_p \) is the collection of forces at the constituent point contact/s at the pusher, and \( FC_{\text{pusher}} \) is the Coulomb friction cone/s at those constituent contact/s. The Jacobian \( J_p^T \) maps the local contact forces at the pusher \( (f_p) \) to the wrenches in the object frame; \( \mathbf{w}_{\text{pusher}} \) is the unit wrench corresponding to unit force/s \( \mathbf{f}_p \) inside the friction cone/s at the constituent contact/s of the pusher.

Fig. 2 shows the generalized friction cone for a line pusher modeled with point contacts at the ends. The convex hull of the wrenches exerted by the forces in the friction cones at the two points contacts in fact takes the form of a polyhedral cone.
B. Dynamics of Stable Prehensile Pushing

An object in a grasp moves following the net wrench acting on the object. Under the quasi-static assumption, which is valid for slow pushing operations, the inertial forces on the object are negligible and there is a force balance:

\[ w_{\text{grasp}} + w_{\text{pusher}} + mg = 0 \]  

(4)

Equation (4) is written in the object frame located at the center of gravity of the object, where \( w_{\text{grasp}} \) is the friction wrench provided by the grasp; \( w_{\text{pusher}} \) is the wrench exerted by the pusher; \( m \) is the mass of the object; and \( g \) is the gravitational wrench.

This paper focuses on manipulations in a parallel-jaw grasp, so the object motions are restricted in the plane of the grasp. We assume uniform pressure distribution at the finger contacts. The normal forces at the finger contacts balance each other. All forces involved in these manipulations (friction forces at the fingers, the pusher forces, and the gravitational force) are in the plane of the grasp. Both the fingers contribute equally to constitute the grasp wrench, i.e., \( w_{\text{grasp}} = 2(J_c^T \cdot w_c) \). Here, \( J_c^T \) maps the contact wrench from the finger contact frame to the object frame. We can rewrite (4) as:

\[ 2(\mu_c N J_c^T \cdot w_c) + J_p^T \cdot f_p + mg = 0 \]  

(5)

A prehensile push for which the pusher contact sticks to the object during the push is called a stable prehensile push [4]. For a stable prehensile push, constraint (5) is balanced by the force/s inside the friction cone/s at the pusher constituent contact/s:

\[ 2(\mu_c N J_c^T \cdot w_c) + J_p^T \cdot f_p + mg = 0 \quad f_p \in FC_{\text{pusher}} \]

Using (3), we can rewrite the previous equation as:

\[ -2(\mu_c N J_c^T \cdot w_c) - mg = k_1 w_{\text{pusher}} \]

\[ w_{\text{pusher}} \in W_{\text{pusher}} \quad k_1 \in \mathbb{R}^+ \]  

(6)

where \( k_1 \) is the magnitude of the pusher normal force. To find if a particular object motion can be achieved with a stable push, we can check if the net required wrench falls inside the generalized friction cone of the pusher:

\[ -2(\mu_c N J_c^T \cdot w_c) - mg \in W_{\text{pusher}} \]  

(7)

C. Dynamics of Fixtureless Fixturing

We define fixtureless fixturing as a subset of stable prehensile pushes that are invariant to the object inertia, grasping force and friction parameters at the finger contacts.

The presence of gravity in (6) makes the dynamics of prehensile pushing sensitive not only to the object inertia \((mg)\), but also to the grasping force \((N)\) and the coefficient of friction at the fingers \((\mu_c)\). If the pusher is aligned such that the pusher contact normal is along the direction of gravity, the gravitational force on the object is entirely balanced by the part of the normal force at the pusher. Then the dynamics checks for stable prehensile pushing (6) becomes:

\[ -2(\mu_c N J_c^T \cdot w_c) - k_2 w_{\text{pusher}a} = k_1 w_{\text{pusher}} \]

\[ w_{\text{pusher}} \in W_{\text{pusher}} \quad k_1, k_2 \in \mathbb{R}^+ \]  

where \( w_{\text{pusher}a} \) is the unit normal wrench at the pusher contact. We know that \( w_{\text{pusher}a} \in W_{\text{pusher}} \), so we can rewrite the above constraint as:

\[ -2(\mu_c N J_c^T \cdot w_c) = k_3 w_p \]

\[ w_p \in W_{\text{pusher}} \quad k_3 \in \mathbb{R}^+ \]  

(8)

As \( \mu_c \) and \( N \) are scalar constants, we further simplify it as:

\[ -J_c^T \cdot w_c = k_4 w_p \]

\[ w_p \in W_{\text{pusher}} \quad k_4 \in \mathbb{R}^+ \]  

(9)

Now, we can write the dynamics condition for stable prehensile pushing with gravity balancing pushers as:

\[ -J_c^T \cdot w_c \in W_{\text{pusher}} \]

D. Computation of Robust Motion Cone

A motion cone abstracts the dynamics of pushing and provides direct bounds on the object motions that can be achieved while keeping the pusher contact sticking to the object. In this section, we will extend the idea of the motion cone for fixtureless fixturing.

**Problem:** Find a set of instantaneous object motions that can be achieved with fixtureless fixturing.

This is equivalent to finding a set of object motions that satisfy constraint (9). To find a motion cone, we find a set of support contact wrenches \((\overline{w}_c)\) that satisfy (9) and then map this wrench-set \((\overline{W}_c)\) to a set of object twists. We denote this object twist-set by \( \overline{V}_{\text{obj}} \) and call it robust motion cone.

From the constraint (9) we can observe that \(-J_c^T \cdot \overline{W}_c = W_{\text{pusher}} \) or \( \overline{W}_c = -J_c^T \cdot W_{\text{pusher}} \) where \( J_c^T \) maps wrenches...
from the object frame to the finger contact frame. From definition \( W_{\text{pusher}} \), \( W_{\text{pusher}} \) is convex and polyhedral. Therefore, \( W_C \) and \( V_{\text{obj}} \) are also convex and polyhedral. The generators of the robust motion cone \( V_{\text{obj}} \) can be computed by a linear mapping of the generators of the generalized friction cone of the pusher \( W_{\text{pusher}} \). Fig. 3 shows a graphical illustration of the motion cone computed for a pushing scenario.

**Procedure to compute a robust motion cone:**

1. Solve \( \mathbf{w}_c = -\mathbf{J}_c^T \mathbf{w}_p \) for \( \mathbf{w}_p \) equal to every edge of the generalized friction cone \( W_{\text{pusher}} \).
2. Define the set of \( \mathbf{w}_c \) computed in step 1 as the generators/edges of the wrench-set \( W_c \).
3. Map \( W_c \) to the object twist space using \( \mathbf{T}_c \) to obtain the robust motion cone \( V_{\text{obj}} \).

**V. Application to Regrasp Planning**

In this section, we discuss the application of fixtureless fixturing to regrasp planning. We present a sampling-based planning framework that uses the knowledge of robust motion cones and generates a tree of grasps that can be reached with fixtureless fixturing. A path in this tree is a pushing strategy to regrasp an object in the gripper through a series of robust stable prehensile pushes.

The high-level planning framework is similar to that presented in our prior work \([3, 4, 11]\). It is based on T-RRT* – an optimal sampling based method for planning on configuration space cost-maps \([21, 22]\).

**For selective sampling,** the T-RRT* framework uses a transition test and filters the sampled configurations to prefer the exploration in low configuration-cost regions. We define the configuration cost as the distance from the goal. The transition test gives preference to the stochastic exploration towards the goal grasp, while allowing the flexibility to move the object away from the goal if that is necessary to move the object finally to the goal.

**For effective connections,** the T-RRT* exploits the underlying RRT* \([23]\) framework to make and rewire the connections in the tree such that the cost of the nodes is reduced. We define the cost of a node as the sum of the cost of the parent node and the cost to push the object to the sampled node. The cost of a push is 0.1 if the parent node uses the same pusher as the child and 1 otherwise.

In practice, every time a pusher contact is changed, it introduces a possibility to add noise to the system. Reducing the pusher switch-overs can add to the robustness of the manipulations planned. With our node cost definition, the planner tries to reduce the number of pusher switch-overs required to push the object to the desired pose.

Algorithm 1 shows our in-hand manipulation planner. Let \( q \) be a configuration of an object, i.e., the pose of the object in the gripper frame that is fixed in the world. In this paper, we consider manipulations in a parallel-jaw grasp, so the configuration space \( C \) is \( [X, Z, \theta_y] \in \mathbb{R}^3 \). The object can translate in the grasp plane (\( XZ \) plane) and rotate around a perpendicular to the grasp plane (\( Y \) axis).

Let \( q_{\text{init}} \) and \( q_{\text{goal}} \) be an initial and desired pose of the object in the grasp frame, respectively. The planner initiates a tree \( T \) with \( q_{\text{init}} \) and generates robust motion cones at it. While the goal pose is not reached within some cost threshold, a random configuration \( q_{\text{rand}} \) is sampled. A nearest configuration \( q_{\text{parent}} \) to \( q_{\text{rand}} \) in the tree \( T \) is found, and an unit-step object pose \( q_{\text{sample}} \) towards the \( q_{\text{rand}} \) is computed. Using the transition test, the planner evaluates whether a move from the parent node \( q_{\text{parent}} \) to \( q_{\text{sample}} \) is beneficial or not. If it is beneficial, the \textit{robust push} routine computes an object configuration \( q_{\text{new}} \) closest to \( q_{\text{sample}} \) that can be reached using the robust motion cones at \( q_{\text{parent}} \). The node \( q_{\text{new}} \) is added to the tree such that the node costs of \( q_{\text{new}} \) and the nodes near \( q_{\text{new}} \) are lowered if possible. The robust motion cones are generated for every new node added to the tree.

**Algorithm 1:** In-Hand Manipulation Planner

**Input:** \( q_{\text{init}}, q_{\text{goal}} \)

**Output:** tree \( T \)

\[
T \leftarrow \text{initialize tree}(q_{\text{init}}) \\
\text{generate}_r \text{ robustCones}(T, q_{\text{init}}) \\
\text{while } q_{\text{goal}} \notin T \text{ or cost}(q_{\text{goal}}) > \text{cost threshold do} \\
\quad q_{\text{rand}} \leftarrow \text{sample random configuration}(C) \\
\quad q_{\text{parent}} \leftarrow \text{find nearest neighbor}(T, q_{\text{rand}}) \\
\quad q_{\text{sample}} \leftarrow \text{take unit step}(q_{\text{parent}}, q_{\text{rand}}) \\
\quad \text{if } q_{\text{sample}} \notin T \text{ then} \\
\quad \quad \text{if transition test}(q_{\text{parent}}, q_{\text{sample}}, T) \text{ then} \\
\quad \quad \quad q_{\text{new}} \leftarrow \text{robust}_\text{push}(q_{\text{parent}}, q_{\text{sample}}) \\
\quad \quad \quad \text{if transition test}(q_{\text{parent}}, q_{\text{new}}, T) \text{ and} \\
\quad \quad \quad \quad \text{ grasp maintained}(q_{\text{new}}) \text{ then} \\
\quad \quad \quad \quad \quad q_{\text{parent}} \leftarrow \text{optimEdge}(T, q_{\text{new}}, q_{\text{parent}}) \\
\quad \quad \quad \quad \quad \text{add new node}(T, q_{\text{new}}) \\
\quad \quad \quad \quad \quad \text{add new edge}(q_{\text{parent}}, q_{\text{new}}) \\
\quad \quad \quad \quad \quad \text{generate}_r \text{ robustCones}(T, q_{\text{new}}) \\
\quad \quad \text{rewire tree}(T, q_{\text{new}}, q_{\text{parent}}) \\
\quad \quad \text{end if} \\
\text{end if} \\
\text{end while} \\
\text{return } T
\]

\( 1 \)The implementation of the transition test and its adaptive nature is discussed in detail in \([3]\).
the configuration space of different grasps and generate pushing strategies for the desired in-hand manipulation. The manipulation plans generated by the planner are invariant to the object inertia, grasping force and friction at the fingers.

VI. REGRASP EXAMPLES AND EXPERIMENTAL VALIDATION

We validate our manipulation planner with regrasping tasks in simulation and with experiments. The experiments were performed on a manipulation platform instrumented with an industrial robot arm, a parallel-jaw gripper with force control, and a Vicon system for object tracking.

A. Simulation and Experimental Setup

We consider different regrasp tasks for the objects listed in Table I. We use a parallel-jaw gripper with flat circular finger contacts. The planner is initiated with pusher contacts on either sides of the object and under the object. In practice, all these contacts are implemented using a single feature in the environment.

The initial pose of the object is treated as $[0, 0, 0]$, and the goal poses for different regrasp examples are listed in Table I. The plan time for each regrasp, shown in Table II, is the median time in seconds over 10 planning trials. A computer with an Intel Core i7 2.8 GHz processor and MATLAB R2017a was used for all the computations and planning.

B. Regrassp Examples

1) Regrasping an object offset to the center: The goal in this example is to regrasp the square prism 20 mm offset from the center in the horizontal direction. The gripping force, and frictional parameters at the fingers and at the features in the environment, are chosen such that pushing the object horizontally from side will not be a valid solution.

2) General manipulation in $[X, Z, \theta_y]$: This example demonstrates large in-hand manipulation for the rectangular prism, involving both displacement and rotation of the fingers on the object. The planner quickly converges to a strategy which does not require any pusher switch-over. The knowledge of motion cones allows the planner to reason about all feasible motions that can be made instantaneously and to choose the motion that best moves the object towards the goal. Fig. 5 shows the pushing strategy our planner generated.

3) Manipulation with a complex non-convex object: Here we show manipulation of a T-shaped object. Manipulating such a non-convex object is particularly hard, because pushing the object directly towards the goal may lead to losing the grasp on the object. The long horizon planning feature of the planner and the capability to reason about discreetly changing the pushers are essential for regrasping this object.

TABLE I

| Shape             | Material | Dim. [L, B, H] (mm) | Mass (g) |
|-------------------|----------|---------------------|----------|
| square prism      | Delrin   | 80, 25, 38          | 130      |
| rectangular prism | Delrin   | 100, 25, 23         | 202      |
| T-shaped          | ABS      | 70, 25, 50          | 62       |

TABLE II

| Manipulation | Goal          | Plan Time | Plan Time | Plan Time |
|--------------|---------------|-----------|-----------|-----------|
| [X, Z, \theta_y] Regrasp | 20, 0, 0 | 0.39 | 2.83 | 392.8 |
| [X, Z, \theta_y] | 15, -13, 45 | 0.65 | 2.54 | 17084 |
| T-shaped     | 25, 17.5, 0   | 0.60 | 0.82 | 32657 |
As shown in Fig. 4 and Fig. 6, our planner generates a strategy that respects the object shape and corresponding geometric constraints. The planner utilizes pushes from two different sides to finally move the object to the goal grasp.

C. Experimental Observations

In this section, we highlight some of the key characteristics of fixtureless fixturing that are supported experimentally. We carried out multiple runs of the pushing strategies in Section VI-B for different grasping forces and recorded the object motion in the grasp.

1) Minimal displacement of the object: The experimental data shows that the object moves a minimal amount with respect to the environment during the planned regrasps. From the mechanics of robust stable prehensile pushing, we expect the object to stick to the environment during the regrasps.

Fig. 7 shows the displacement of the object observed during the pushing strategy shown in Fig. 6. The displacement in the X direction is less than ±0.2 mm, whereas that in the Z direction is even smaller, less than ±0.1 mm. These values are close to the precision level of the Vicon object tracking system, so some of the object displacement can be attributed to the noise from the camera system.

Fig. 8 shows the displacement of the object during the pushing strategy shown in Fig. 5. The object moves during the regrasp by a very small amount, less than 3% of the object displacement required for this regrasp.

2) Robustness against variation in the friction at the fingers: We observe a minimal effect on the outcome of the regrasp actions planned with fixtureless fixturing when the friction at the fingers is changed. As the grasping force changes, the friction force at the fingers changes. However, as seen in the plots in Fig. 7 and Fig. 8, the variation in the outcome of the pushing strategy for three different grasping forces is negligible. This supports the invariance we expect theoretically from fixtureless fixturing.

The repeatability and robustness of the pushing actions allowed the robot to place the T-shaped object after every regrasp and restart the next run autonomously twenty times in a row. The T-shaped object does not have Vicon markers on it for tracking. Twenty runs for the manipulation shown in Fig. 1 were performed in a sequence in an open-loop fashion with different grasping forces.

3) Universal Fixture: For all the examples discussed in Section VI-B, a single feature in the environment was sufficient to manipulate different objects. Fixtureless fixturing allows robots to use a simple feature as a “universal” fixture for different objects and different regrasping tasks.

VII. DISCUSSION

We present a manipulation method that acts as a fixturing technique for regrasping without using a conventional fixturing hardware. The object in a gripper is manipulated with a series of pushes against simple features in the environment such that the environment keeps the object stationary while the fingers slide on the object to a desired pose. We refer to this approach as fixtureless fixturing.

Based the mechanics of prehensile pushing, we construct convex polyhedral sets of object motions that can be produced with fixtureless fixturing. We call this subsets of pushes robust motion cones and show that they are invariant to the object inertia, gripping force, and friction at the fingers. The robust motion cones abstract the dynamics of prehensile pushing and provide direct bounds on the object motions that are feasible with fixtureless fixturing.

We demonstrate the application of the robust motion cones for planning in-hand manipulations with a parallel-jaw gripper. Our planner generates pushing strategies in a fraction of a second to force the object to the goal pose. The generated pushing plans are robust against the uncertainty in the object inertia, friction at the fingers, and gripping force. We validate this robustness with controlled experiments on a platform instrumented with a Vicon object tracking system. The experimental results confirm that the object remains fixtured to the environment during the manipulations. Moreover, minimal variation is observed in the outcome of the regrasp experiments performed with different grasping forces.

Using the fixtureless fixturing technique, robots can exploit simple features in the environment as versatile fixtured fixtures. The flexibility gained by using fixtureless fixturing facilitates dexterous manipulation even with simple grippers. With such an approach to in-hand manipulation, we move a step closer to the goal of versatile, yet simple, robust, and flexible robotic automation.
REFERENCES

[1] N. Chavan-Dafle, A. Rodriguez, R. Paolini, B. Tang, S. S. Srinivasa, M. Erdmann, M. T. Mason, I. Lundberg, H. Staab, and T. Fuhlbrigge, “Extrinsic dexterity: In-hand manipulation with external forces,” in IEEE International Conference on Robotics and Automation (ICRA), 2014, pp. 1578–1585.

[2] N. Chavan-Dafle and A. Rodriguez, “Prehensile Pushing: In-hand Manipulation with Push-Primitives,” in IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS), 2015, pp. 6215 – 6222.

[3] N. Chavan-Dafle and A. Rodriguez, “Sampling-based Planning of In-Hand Manipulation with External Pushes,” in Int Symp Robot Res, 2017.

[4] N. Chavan-Dafle and A. Rodriguez, “Stable Prehensile Pushing: In-Hand Manipulation with Alternating Sticking Contacts” in IEEE International Conference on Robotics and Automation (ICRA), 2018.

[5] American Society of Tool and Manufacturing Engineers, Handbook of fixture design: a practical reference book of workholding principles and designs for all classes of machining, assembly, and inspection. McGraw-Hill, 1962.

[6] H. Asada and A. By, “Kinematic analysis of workpart fixtureing for flexible assembly with automatically reconfigurable fixtures” IEEE Journal on Robotics and Automation, vol. 1, no. 2, pp. 86–94, 1985.

[7] Y.-C. Chou, V. Chandru, and M. Barash, “A Mathematical Approach to Automatic Configuration of Machining Fixtures; Analysis and Synthesis” ASME Journal of Engineering for Industry, vol. 111, no. 4, pp. 299–306, 1989.

[8] L. S. Hong and M. R. Cutkosky, “Fixture Planning With Friction” Journal of Manufacturing Science and Engineering, vol. 113(3), pp. 320–327, 1991.

[9] S. Goyal, “Planar sliding of a rigid body with dry friction: Limit surfaces and dynamics of motion,” PhD Dissertation, Dept. of Mechanical Engineering, Cornell University, 1989.

[10] M. T. Mason, “Mechanics and Planning of Manipulator Pushing Operations” The International Journal of Robotics Research, vol. 5, no. 3, pp. 53–71, 1986.

[11] N. Chavan-Dafle, R. Holladay, and A. Rodriguez, “In-hand manipulation via motion cones,” in Robotics: Science and Systems (RSS), 2018.

[12] R. D. Howe and M. R. Cutkosky, “Practical Force-Motion Models for Sliding Manipulation” The International Journal of Robotics Research, vol. 15, no. 6, pp. 557–572, 1996.

[13] N. Xydas and I. Kao, “Modeling of Contact Mechanics and Friction Limit Surfaces for Soft Fingers in Robotics, with Experimental Results” The International Journal of Robotics Research, vol. 18, no. 9, pp. 941–950, 1999.

[14] K. M. Lynch and M. T. Mason, “Stable Pushing: Mechanics, Controllability, and Planning” The International Journal of Robotics Research, vol. 15, no. 6, pp. 533–556, 1996.

[15] M. Dogar and S. Srinivasa, “A Framework for Push-grasping in Clutter” in Robotics: Science and Systems. MIT Press, July 2011.

[16] J. Shi, J. Z. Woodruff, and K. M. Lynch, “Dynamic in-hand sliding manipulation,” in 2015 IEEE International Conference on Intelligent Robots and Systems, 2015, pp. 870–877.

[17] J. Zhou and M. T. Mason, “Pushing revisited: Differential flatness, trajectory planning and stabilization” in Int Symp Robot Res, 2017.

[18] K. M. Lynch, H. Maekawa, and K. Tanie, “Manipulation And Active Sensing By Pushing Using Tactile Feedback,” in Proceedings of the IEEE/RSJ International Conference on Intelligent Robots and Systems, vol. 1, 1992, pp. 416–421.

[19] M. Erdmann, “On a Répétition de la Friction en Configuration Space” International Journal of Robotics Research, vol. 13, no. 3, pp. 240–271, 1994.