

Geometric In-Hand Regrasp Planning- Alternating Optimization of Finger Gaits and In-Grasp Manipulation

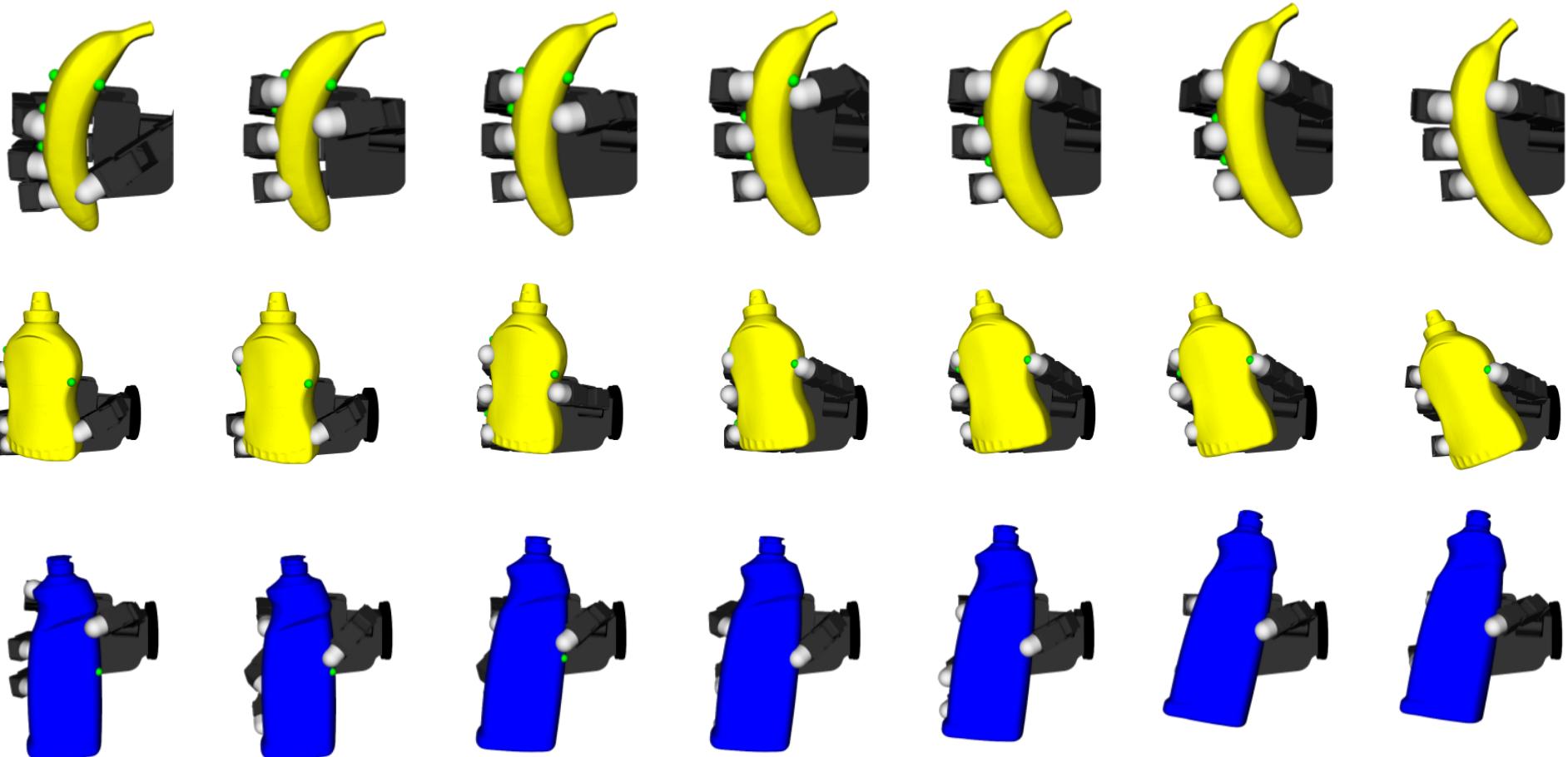
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[2020.5.12]

Outline

- Introduction
- Related work
- Method
- Experiments
- Conclusion

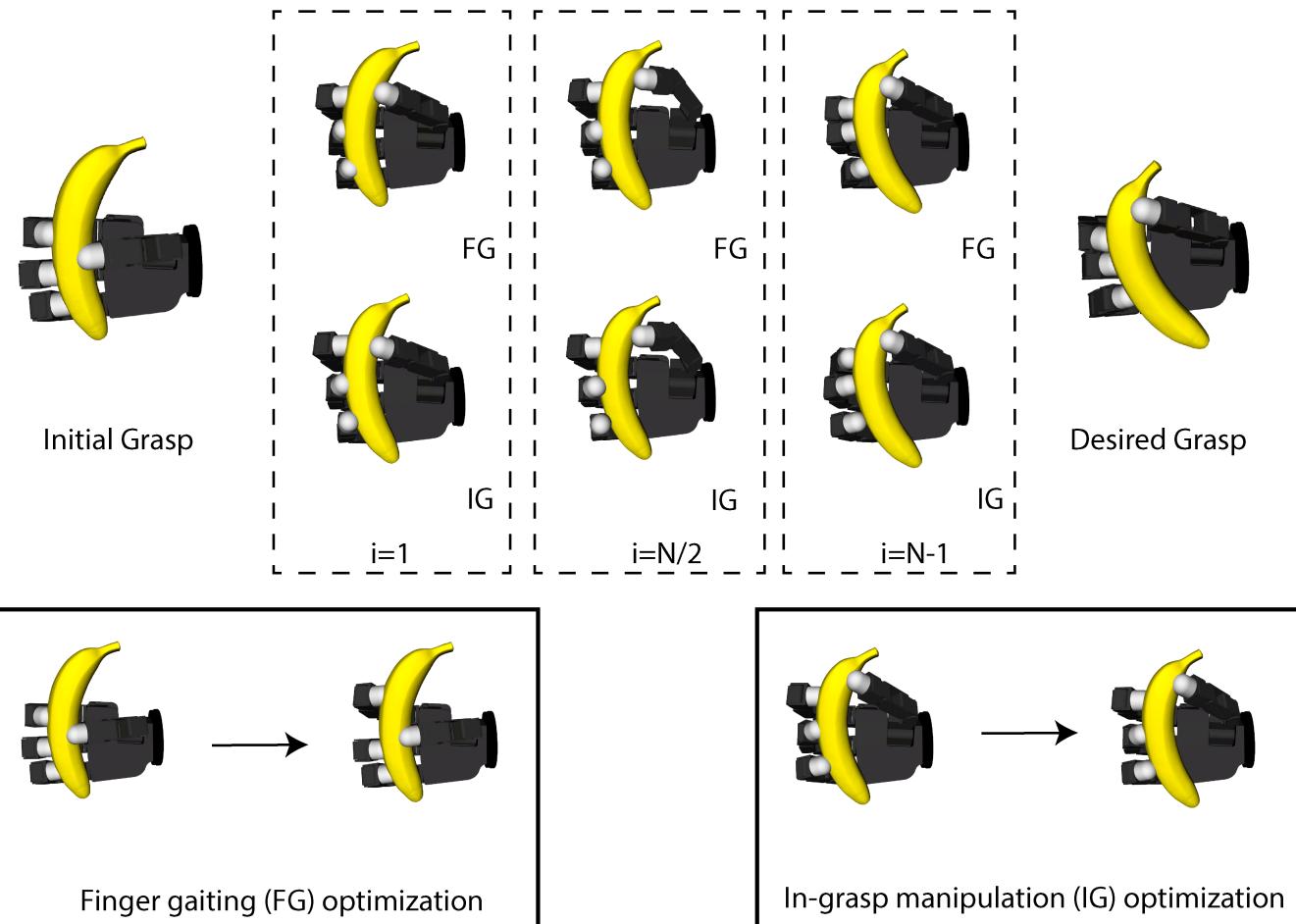
Background

- In-hand regrasping



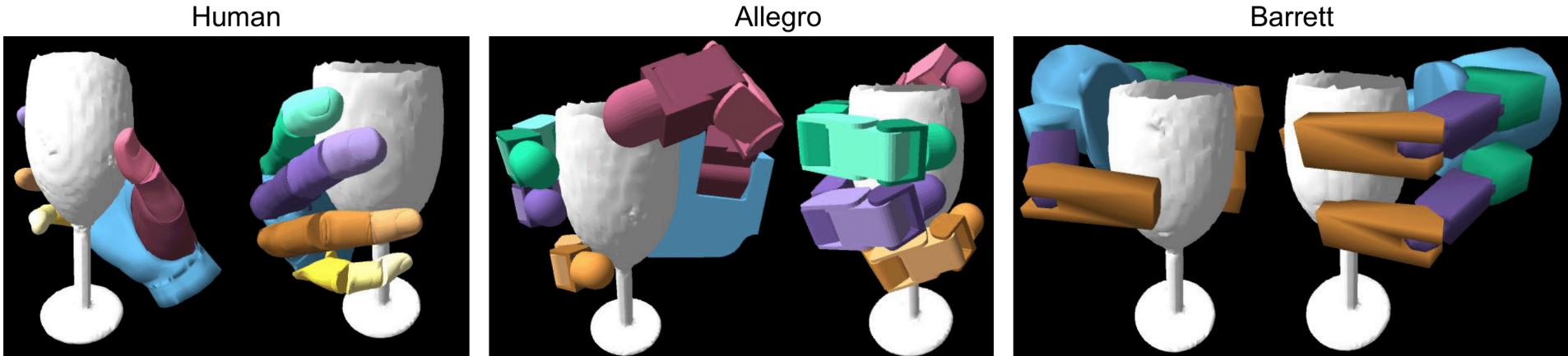
Overview

- Moving from an initial grasp on an object to a desired grasp, while maintaining contact locations



Introduction

- Allegro Hand

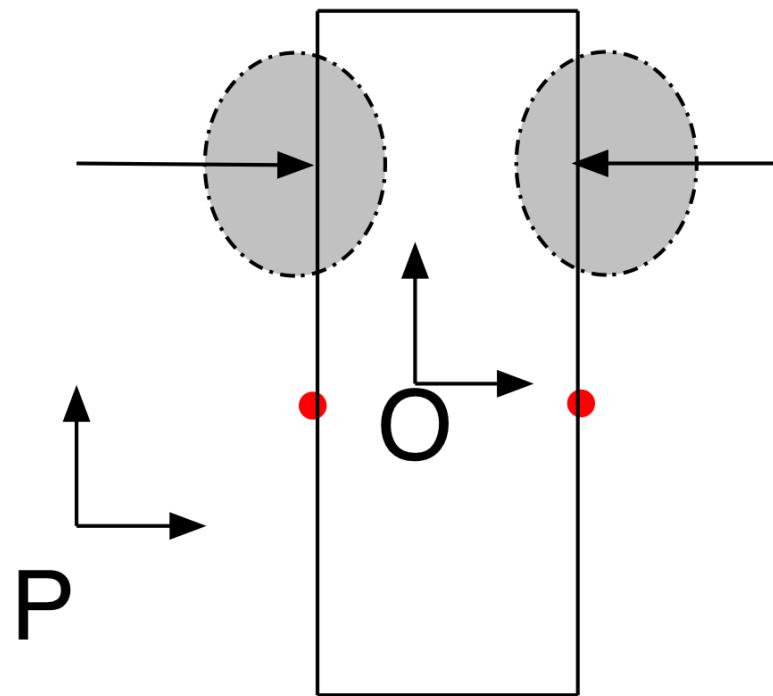


Terminology

- Contact point in R^3
- Joint configuration angle representation Θ
- Forward kinematics (FK), mapping from joint angles to poses

Terminology

- Finger reachable workspace (triangle mesh model by sampling the enclosed volume)



Terminology

- Object Mesh



Open Problems

- 1) Moving to a set of desired grasp contact points from the current grasp
- 2) Avoiding unwanted collisions between the object and the hand during manipulation
- 3) Moving to a desired object pose after reaching the desired grasp contact points
- 4) Ensuring stability of the object during manipulation
- 5) Choosing the correct sequence of fingers in performing finger gaiting
- 6) Planning an initial grasp, which can achieve the desired object pose through in-grasp re-planning

Contributions

- An optimization based framework for planning finger gaits on **arbitrary object meshes**, which directly solves for **collision-free** joint angle trajectories

Limitations

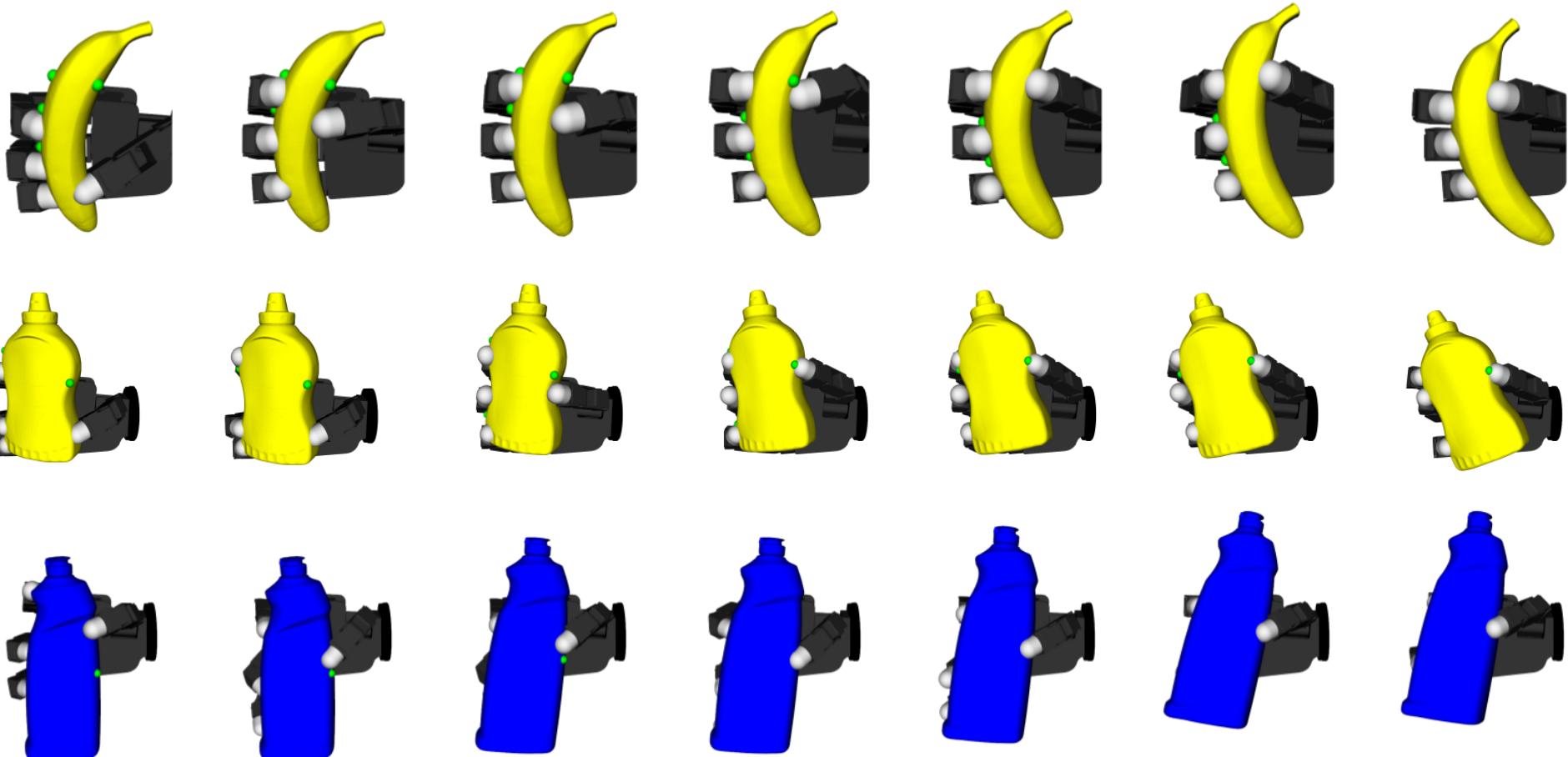
- Not include dynamics, so it would be insufficient for execution on a physical robot
- Contacts are only made at the fingertips

Assumptions

- The object is rigid
- The desired grasp is a stable grasp and the desired object pose is reachable
- Assume that all the fingers can repose their contact on the object. In the case of the thumb, we assume that it can slide to the new contact point
- Generate initial and desired grasps manually

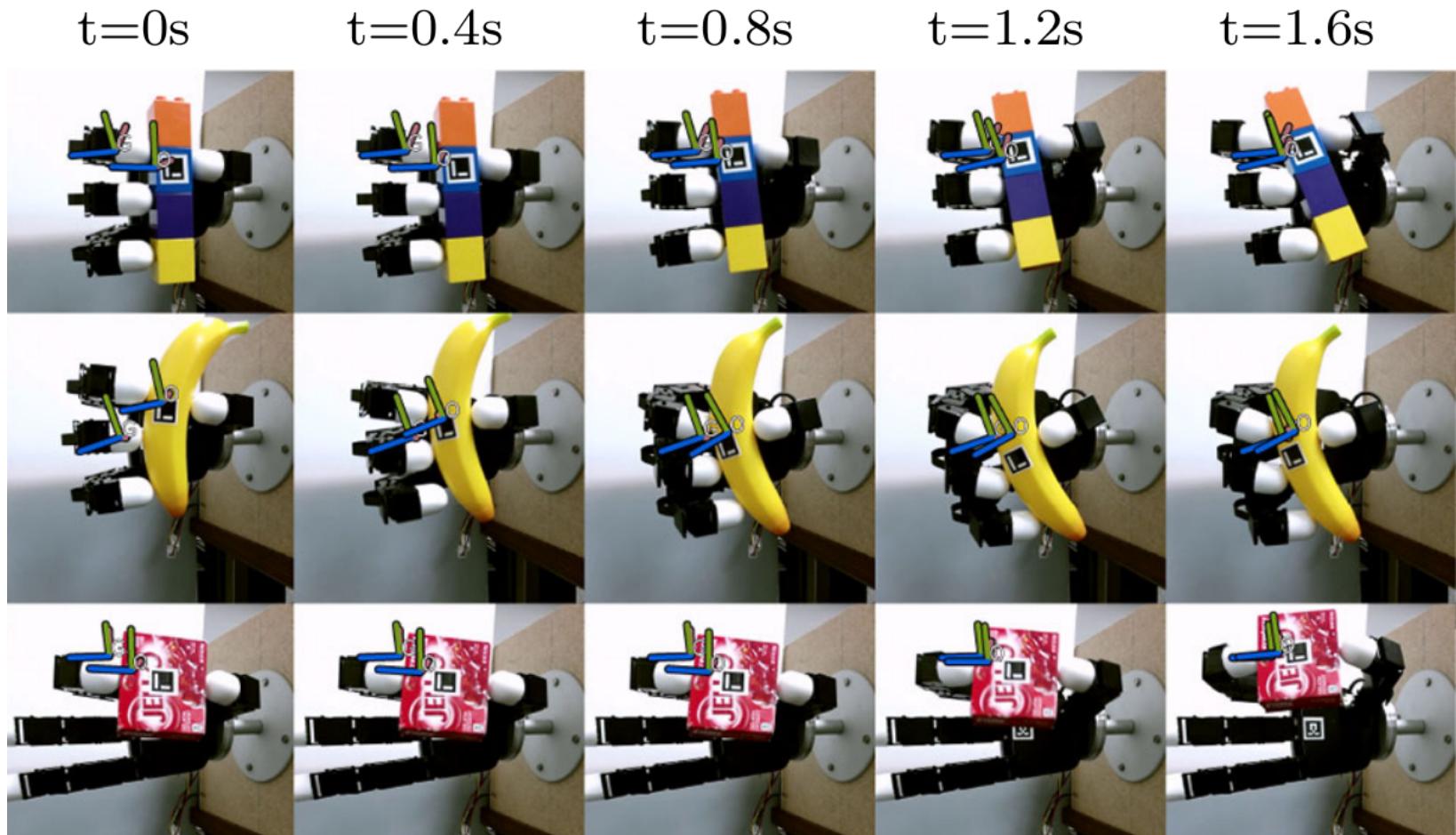
Related Work

- Moving to a set of desired grasp contact points from the current grasp



Related work

- Moving to a desired object pose after reaching the desired grasp contact points



Related Work

- Two categories
 - Fingertips always remain in contact
 - Fingers break and make new contacts
-
- Solutions
 - Enumerate all possible motions and build a search tree
 - Find a continuous path in the joint configuration space to a grasp
-
- Achieve Object pose
 - Achieve goal grasp

Related Work

- How to obtain the ground-truth
- Contact Measure



Overview

- Goal: finding a sequence of grasps which moves the fingertip to the the goal finger contact point in N step
- Sequence of grasps $G = [G_0, G_1 \dots G_N]$
- Each grasp $G_i = [X_i, O_i]$ contains contact points X_i and object pose $O_i \in SE(3)$
- Each contact point $X_i = [f_1, f_2, f_3, f_4]$ where $f_i \in R^3$

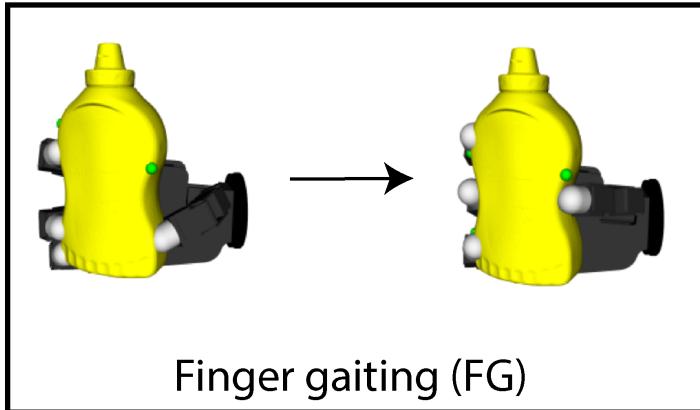
Overview

- At each time step
- Finger-gaiting
- Finding a new location for a fingertip within its reachable workspace.
- In-grasp manipulation
- Moving the object to shift the reachable workspace of the fingertips relative to the object surface.

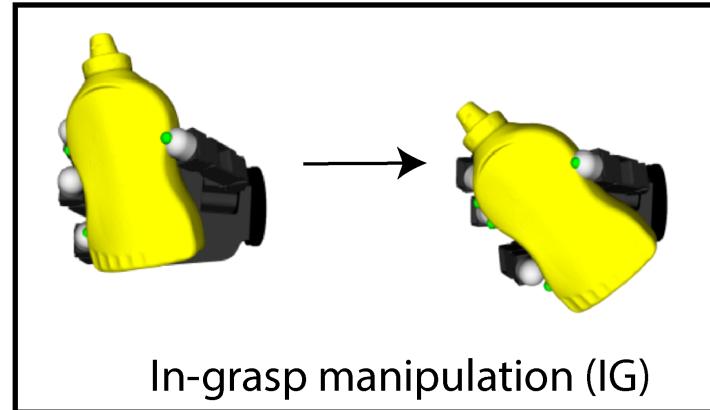
Overview

Optimization Primitives

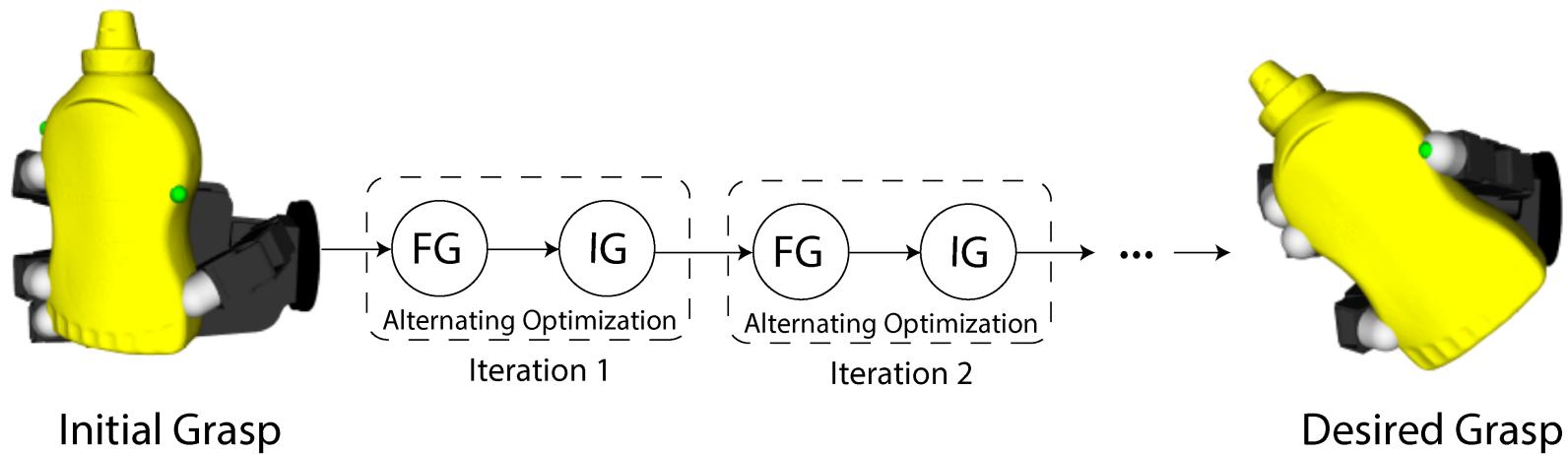
Fingertip Relocation



Object Reposing



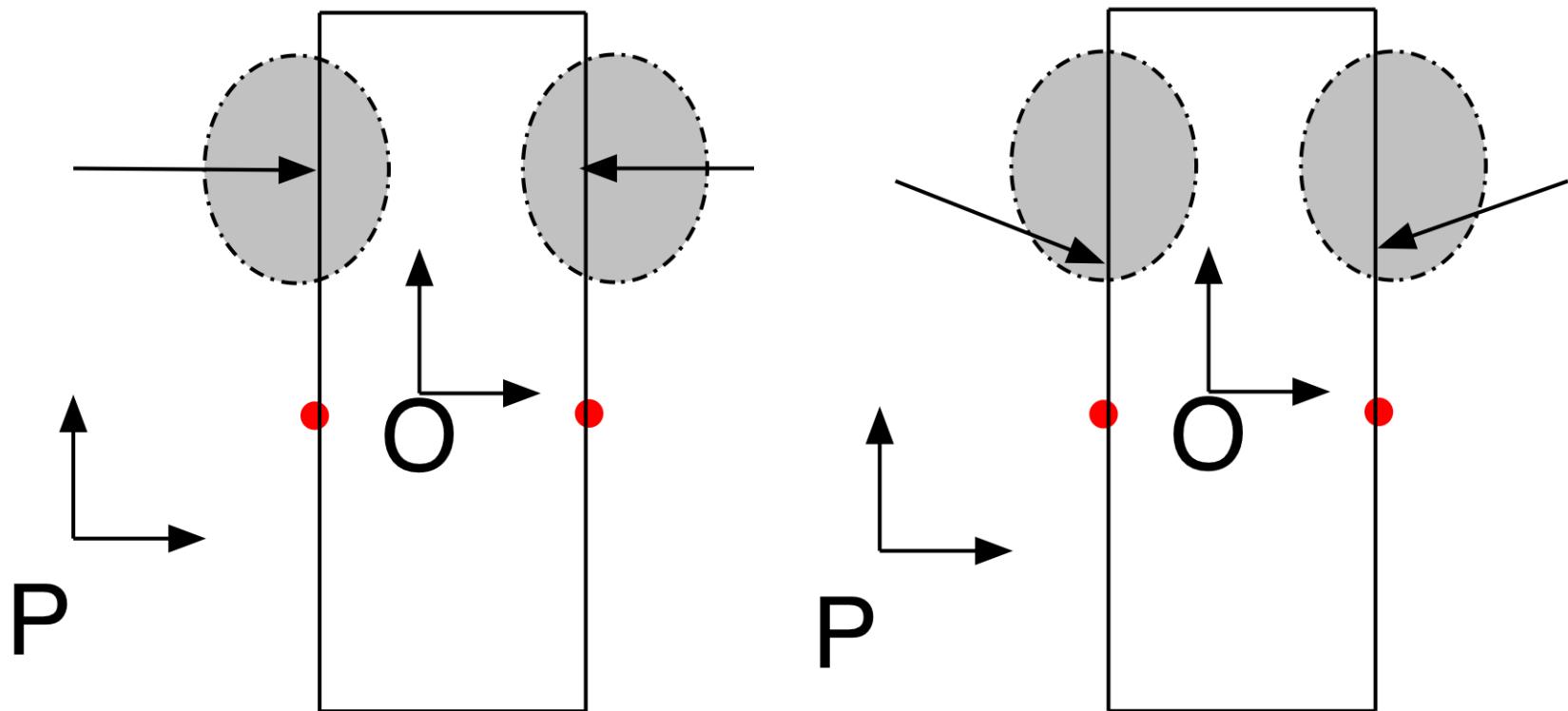
Regrasp Sequence Planning



OPT1

- Optimization for Finger Gaits (OPT1)
- Moves the contact point f_r of fingertip r towards the goal finger contact point $f_{r,g}$ reachable workspace of finger r
- Geometric optimization problem over the joint angles Θ_r of finger r, while the remaining joints in Θ remain fixed
- A fixed gait pattern {index, middle, ring, thumb} or {thumb,ring,middle,index} according to the desired contact location

OPT1



OPT1

- Optimization for Finger Gaits (OPT1)

$$\min_{\Theta^r} D(f_{r,g}, FK_r(\Theta^r))$$

s.t.

$$\Theta_{min}^r \preceq \Theta^r \preceq \Theta_{max}^r$$

$$SD(FK_r(\Theta^r), M) = 0$$

$$C(\Theta^r, M) = 0$$

$$S(FK_r(\Theta^r)) \leq \eta,$$

- Optimize over joint angles Θ_r of finger r

OPT1

- Overall Cost function D

$$D(f_{r,g}, FK_r(\Theta^r)) = \|f_{r,g} - FK_r(\Theta^r)\|_2^2$$

- $f_{r,g}$ is the goal contact point for finger r
- FK_r is the forward kinematics for finger r

OPT1

- Joint limit constraints

$$\Theta_{min}^r \leq \Theta^r \leq \Theta_{max}^r$$

- Signed distance of fingertip lying on the surface of the object mesh M

$$SD(FK_r(\Theta^r), M) = 0$$

- Collision cost for links of finger r

$$C(\Theta^r, M) = 0$$

- Stability

$$S(FK_r(\Theta^r)) \leq \eta$$

OPT1

- Collision cost C for links L_r of finger r

$$C(\Theta^r, M) = 0$$

- All links on the moving finger maintain at least distance of β from the object mesh M

$$C(\Theta^r, M) = \sum_{l \in L_r} (\beta - \min(\beta, SD(FK_l(\Theta^r), M)))$$

OPT1

- Stability Cost S
- Limit the finger gait distance to be within a threshold

$$S(FK_r(\Theta^r)) \leq \eta$$

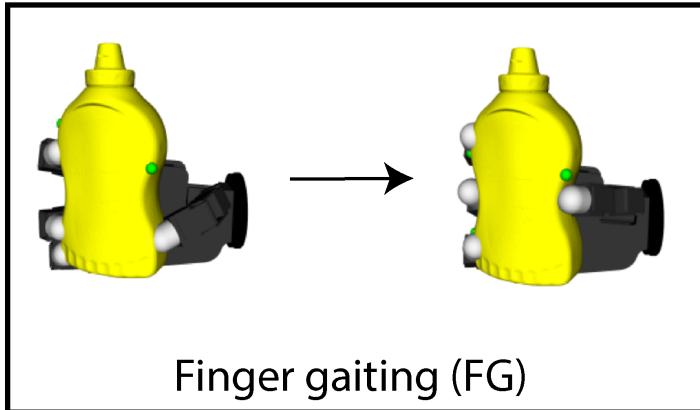
$$S(FK(\Theta^r)) = \|FK_r(\Theta_0^r) - FK_r(\Theta^r)\|_2^2$$

- small steps are taken when η is small

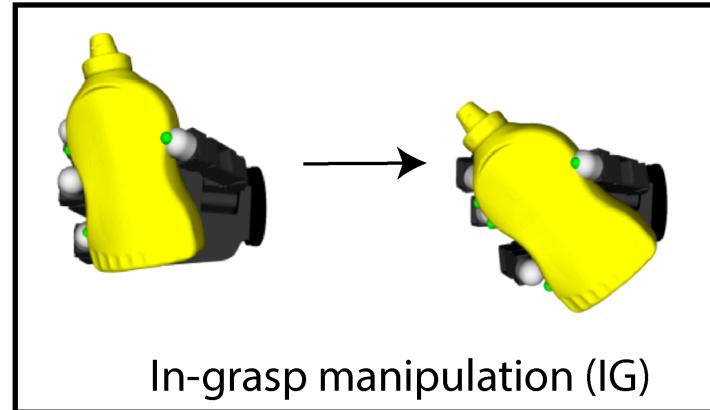
Overview

Optimization Primitives

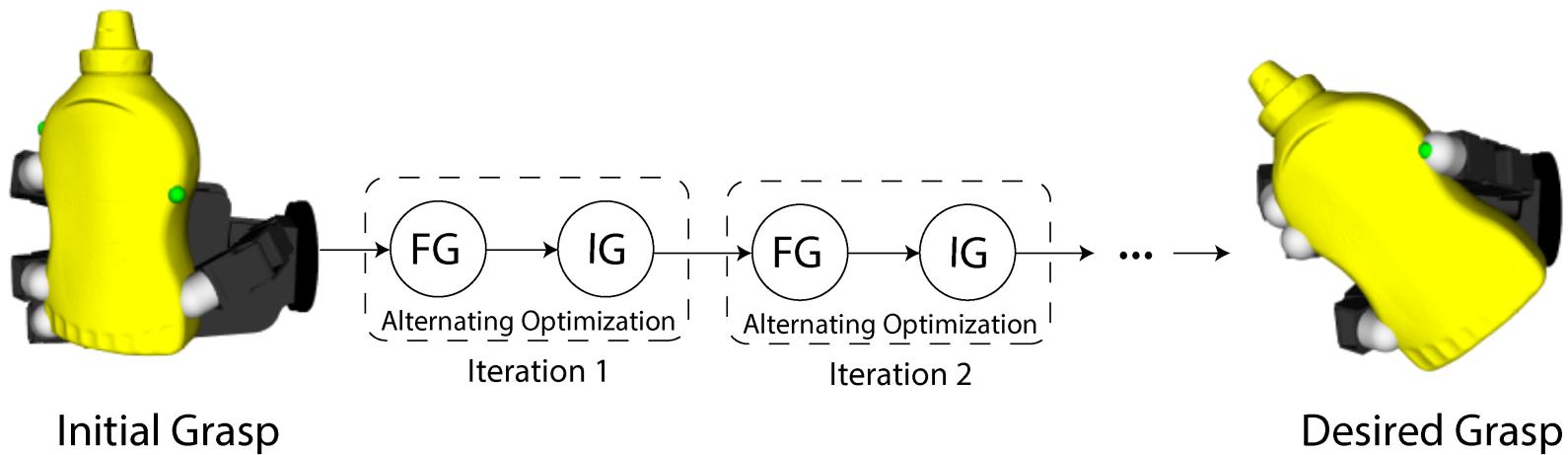
Fingertip Relocation



Object Reposing



Regrasp Sequence Planning



OPT2

- Optimization for Object Reposing (OPT2)
- Change object pose in order to shift the reachable workspace of the fingertips relative to the object

$$\min_{\Theta} E_{des} + k_1 E_{pos}(\Theta) + k_2 E_{or}(\Theta)$$

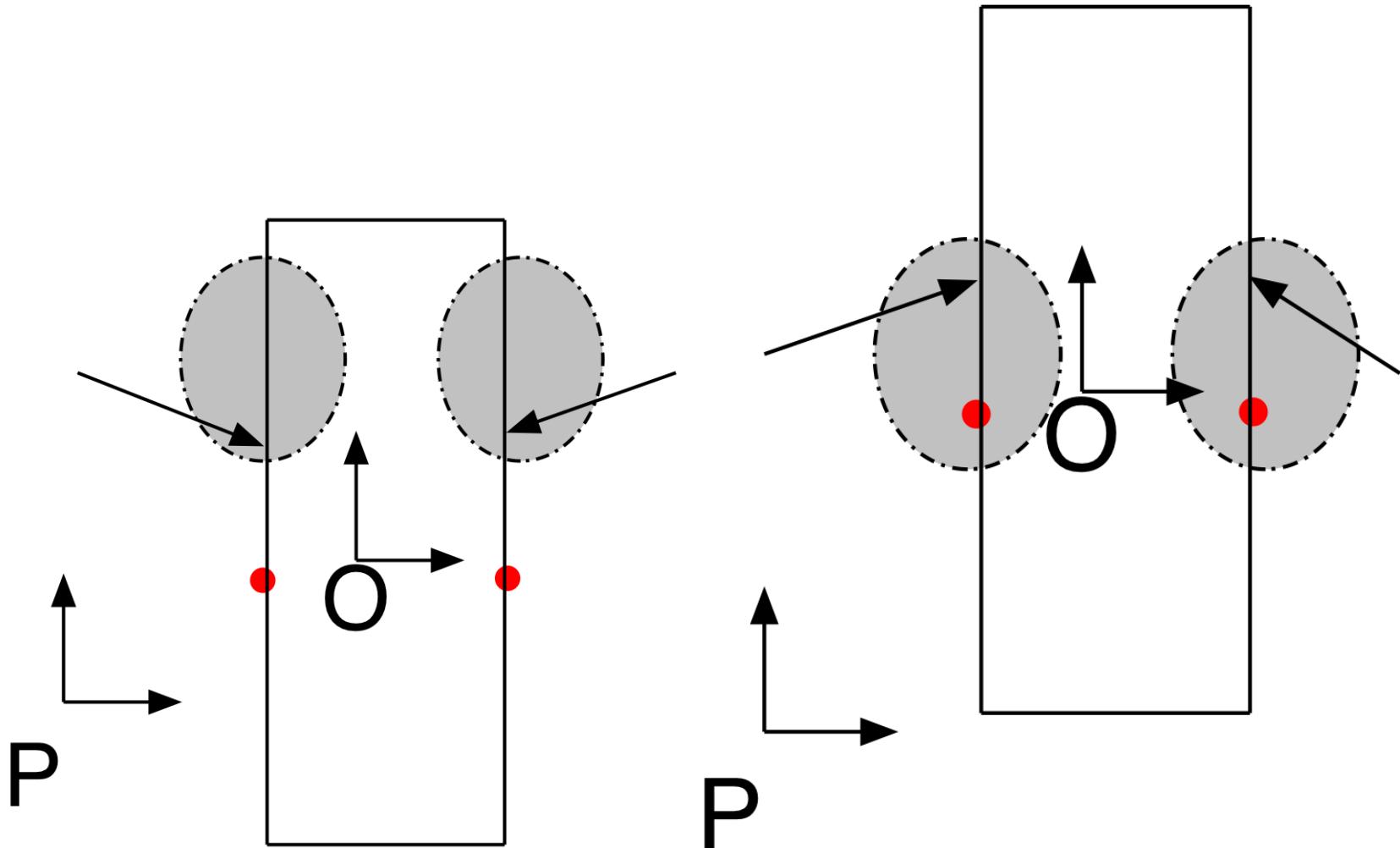
s.t.

$$\Theta_{min} \preceq \Theta \preceq \Theta_{max}$$

$$C(\Theta, M) = 0$$

- Optimize over all joint angles Θ

OPT2



OPT2

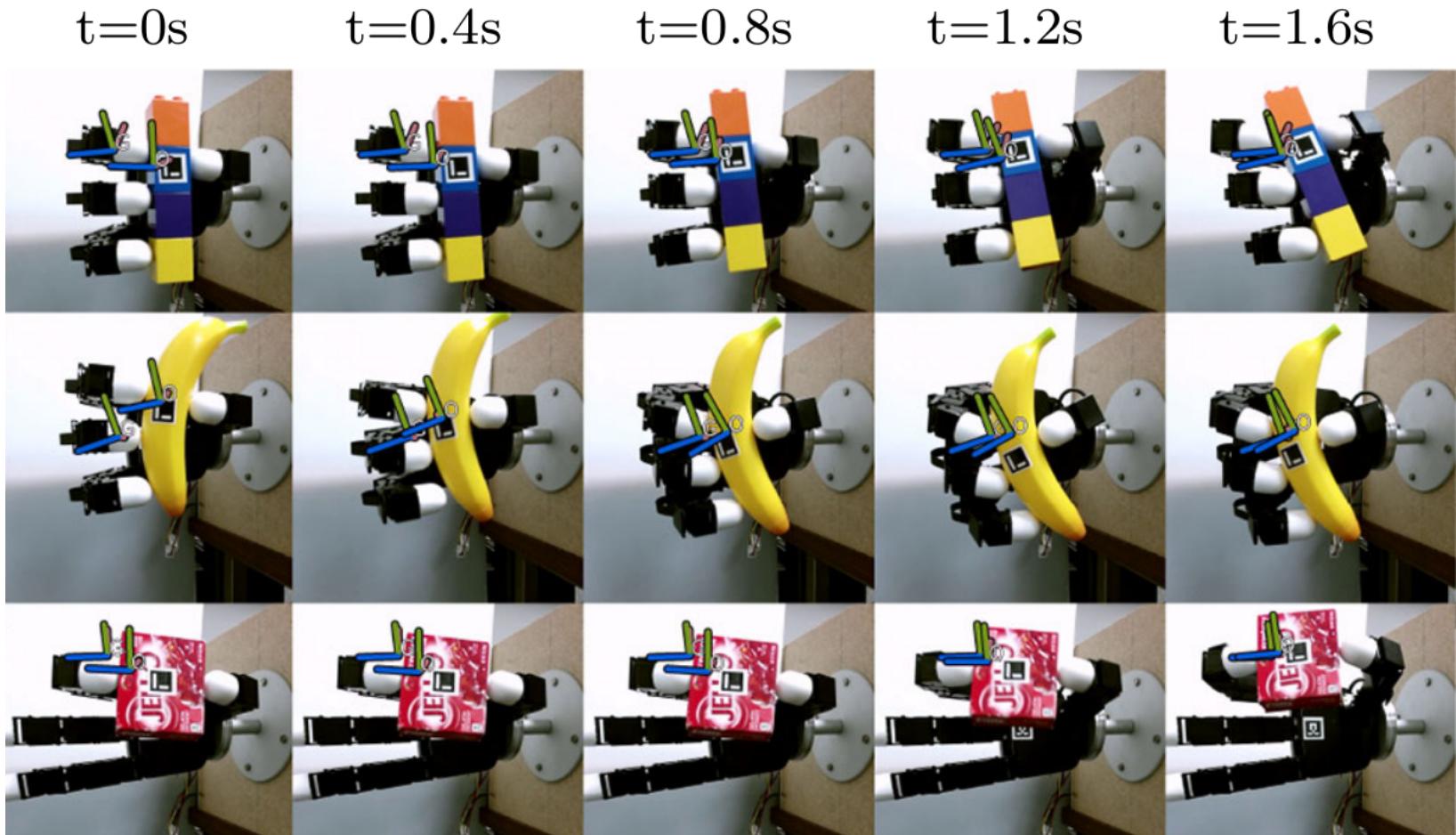
- Adjust Object pose
- Solution-1:
- Reduces the distance between the reachable workspace R_r of the finger r and the desired contact points, where m is the number of fingers

$$E_{des} = \sum_{r=0}^m \max(0, SD(f_{r,g}, R_r))$$

- Denoted as “SD”

OPT2

- Adjust Object pose



OPT2

- Adjust Object pose
- Solution-2:
- Finding a target object pose \widehat{O}_d and set as target, then solve for the joint angles at the desired object pose
- The target object pose is obtained by ICP algorithm using singular value decomposition

$$E_{des} = E_{obj}(\Theta, \hat{O}_d)$$

- Euclidean distance in world frame
- Denoted as “SVD”

OPT2

- Joint position limits constraint

$$\Theta_{min} \preceq \Theta \preceq \Theta_{max}$$

- Relaxed-rigidity constraint
- $E_{pos}(\Theta)$ encouraging fingertips to keep the same contact locations on the object as in the initial grasp
- $E_{or}(\Theta)$ encourages the other fingers to maintain their relative orientation to the thumb to be the same as that in the initial grasp
- Collision cost C as in OPT1

Pipeline

- Regrasp Planner

Algorithm 1: In-hand Regrasping Planner

Data: M, K_0, X_g, o_g

Result: $\mathbf{K} = [G, \Theta]$

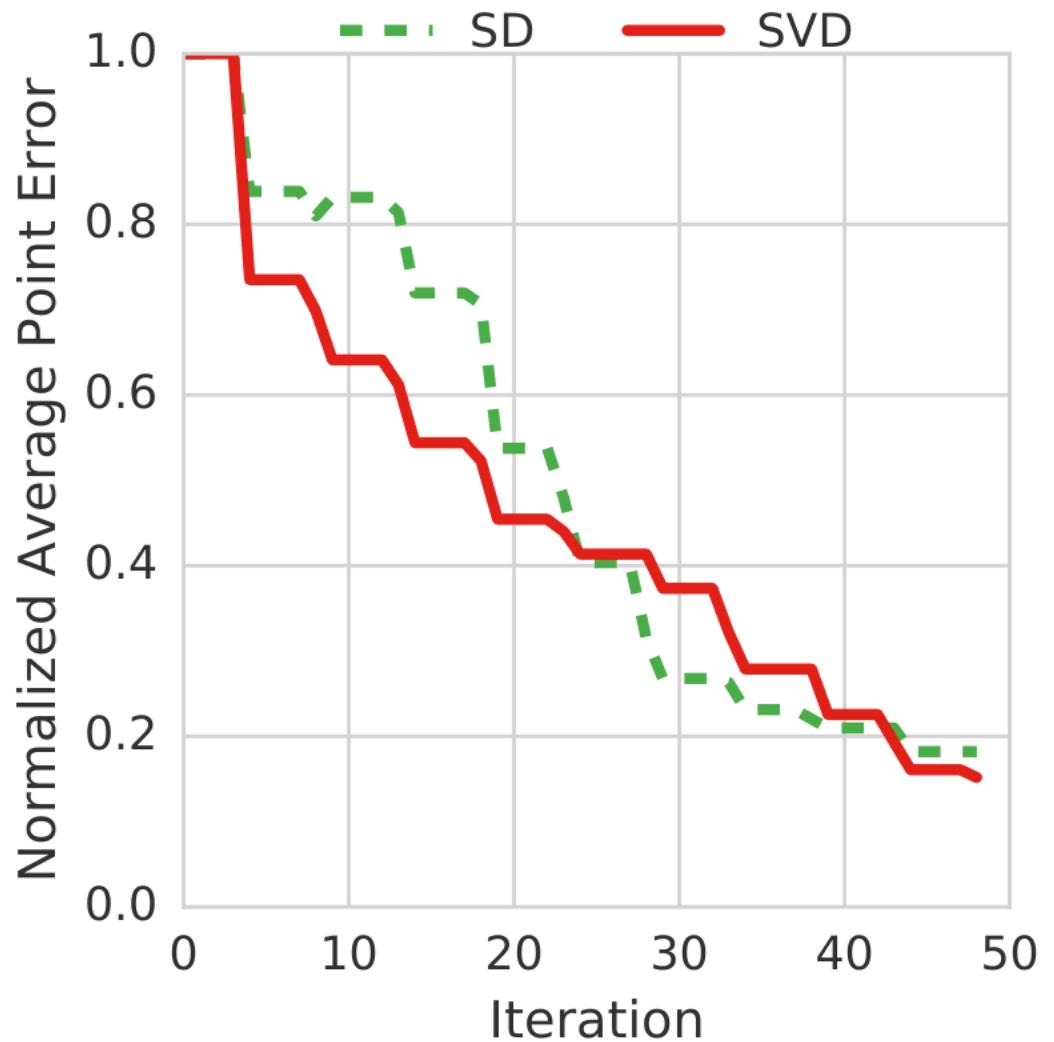
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1 K = [];
2 K.append( $K_0$ );
3  $\text{err} \leftarrow \max_{r \in [0, m]} (f_{r,0} - f_{r,g})$ ;
4  $n = 0$ ;
5 while  $\text{err} > \zeta$  and  $n < 50$  do
6   for  $i \in P$  do
7      $K_t \leftarrow \text{OPT1}(\mathbf{K}.last, X_g, i)$ ;
8     K.append( $K_t$ );
9   end
10   $\text{err} \leftarrow \max_{r \in [0, m]} (f_{r,t} - f_{r,g})$ ;
11  if  $\text{err} > \zeta$  then
12     $K_t \leftarrow \text{OPT2}(\mathbf{K}.last, X_g)$ ;
13    K.append( $K_t$ );
14  end
15   $n++$ ;
16 end
17  $K_t \leftarrow \text{in\_grasp}(\mathbf{K}.last, o_g)$ ;
18 K.append( $K_t$ );
19 return  $\mathbf{K}$ ;
```

Methods

- Optimization
 - sequential quadratic programs (SQP)
 - Maximum number of iterations to 50
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- Evaluation
 - Average Point Error: error between the planned and desired final grasp contact grasp contact points using Euclidean distance over all fingers
 - Convergence rate
 - Planning Time

Experiments

- Convergence Rate



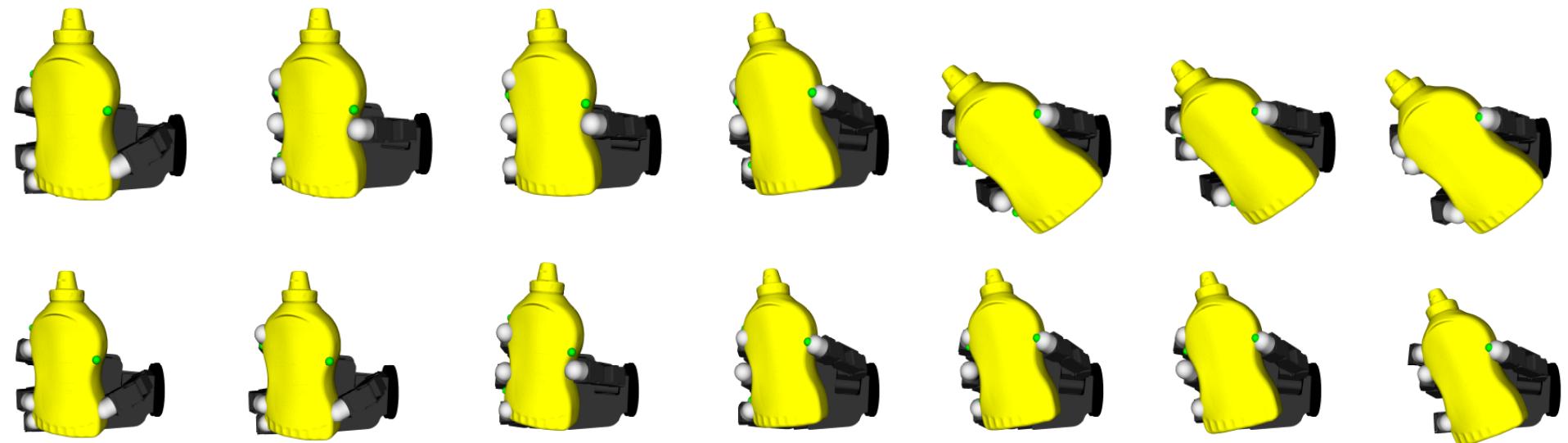
Experiments

- Planning time

Object	Method	Maximum(s)	Median(s)
Banana	SD	1025.35	902.65
	SVD	84	45
Sugar-box	SD	2283.37	1917.41
	SVD	112.59	99.38
Mustard	SD	1133.504	649.69
	SVD	135	92.83
Soft-scrub	SD	884.927	388.317
	SVD	95	74.97
Pringles	SD	3513.96	796.93
	SVD	134.275	16.0052

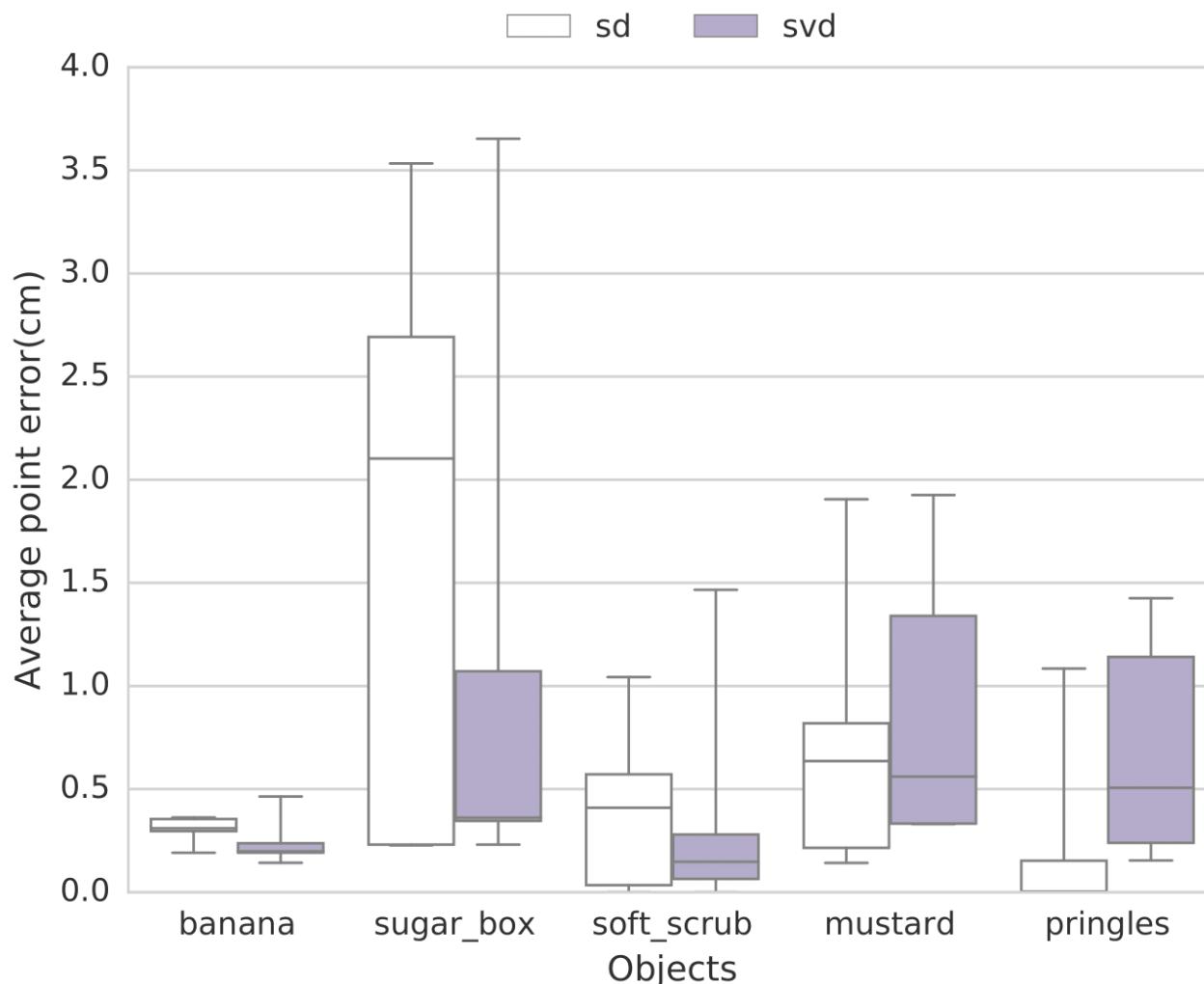
Experiments

- Different strategy



Experiments

- Average Point Error across the objects



Conclusion

- An optimization based planner that can generate collision-free plans to regrasp objects in-hand without dropping the object
- Planner builds on two in-hand manipulation primitives— finger gaiting and in-grasp manipulation— both formulated as optimization problems

Future Work

- Sensitivity to hyper-parameters
- Distance Metric
- Learning based approaches