



PoseRBPF: A Rao-Blackwellized Particle Filter for 6D Object Pose Tracking

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6D OBJECT POSE ESTIMATION

3D Model



Input image



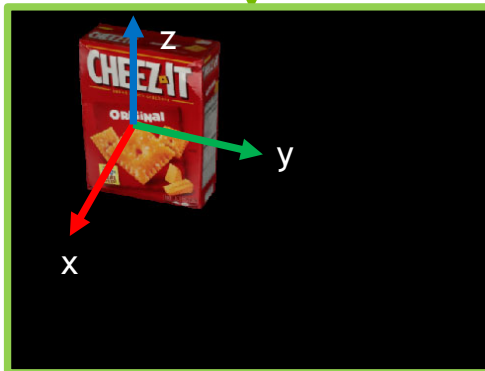
Pose information useful for

- Object manipulation
- Semantic navigation
- Human robot interaction

6D Object Pose

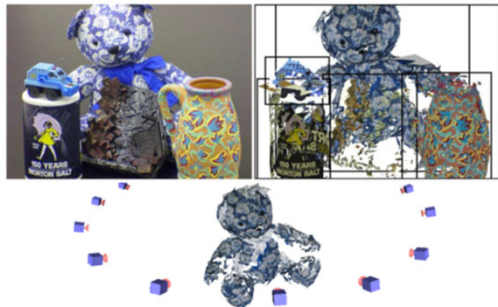
3D
Translation

3D
Orientation

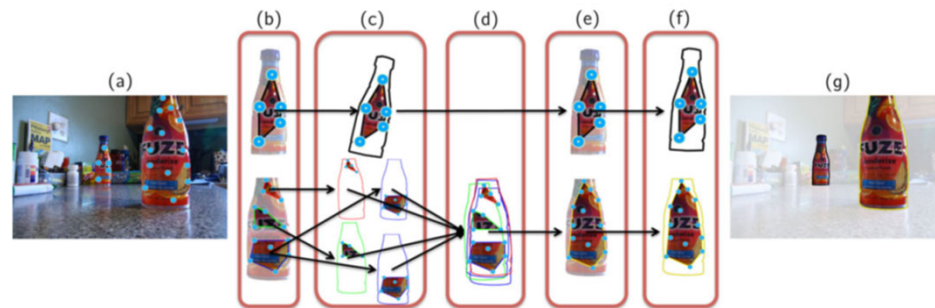


TRADITIONALLY

- Feature matching

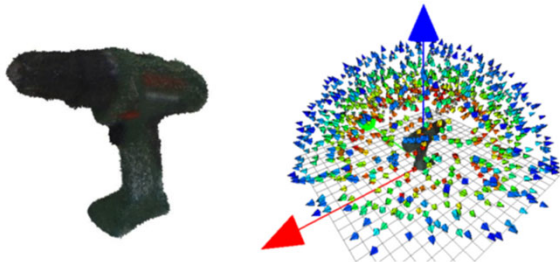


Rothganger et al. IJCV, 2006

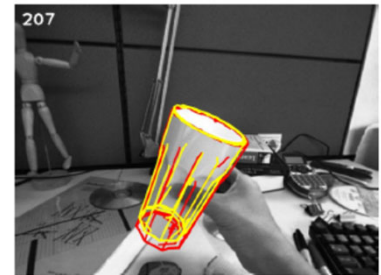
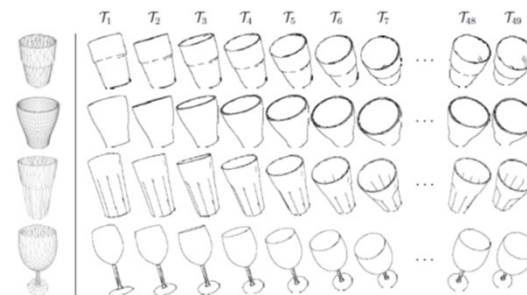


MOPED, Collet, Martinez, Srinivasa, IJRR, 2011

- Template matching



Hinterstoisser et al., ACCV, 2012



Choi et al., IROS, 2012

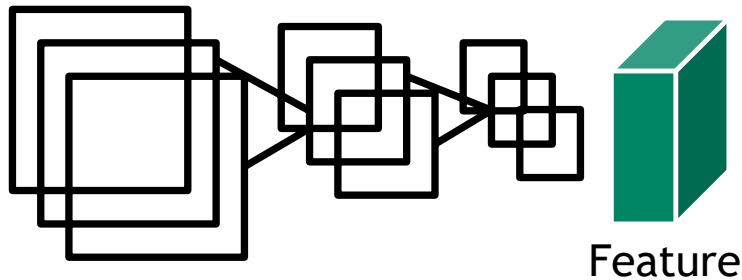
CHALLENGES

- Model capability
 - Texture, texture-less objects
 - Symmetry objects
 - Clutter scenes
- Accuracy and Robustness
 - Lighting change
 - Different background
 - Uncertainty
 - Speed



DEEP LEARNING

- Better image features and stronger model capacity



3D Rotation



3D Translation
3D Rotation

- Rad & Lepetit, ICCV 2017
- Kehl et al. ICCV 2017
- Tremblay et al. CoRL 2018
- Tekin et al. CVPR 2018
- Xiang et al. RSS 2018
- Sundermeyer et al. ECCV 2018
- Li et al. ECCV 2018
- Wang et al. CVPR 2019

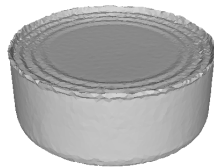
Our goal:

- ✓ Symmetry objects
- ✓ Pose Tracking
- ✓ Pose uncertainty

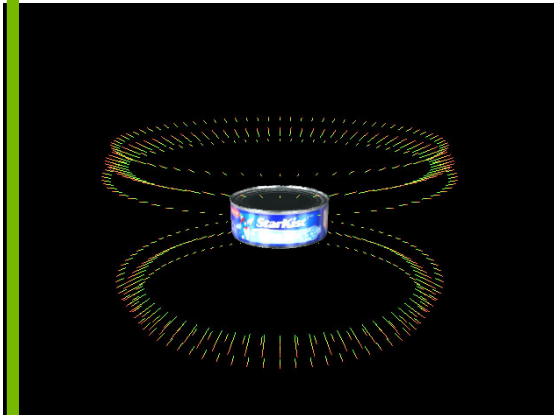
ORIENTATION UNCERTAINTY

Depends on context, shape, sensor

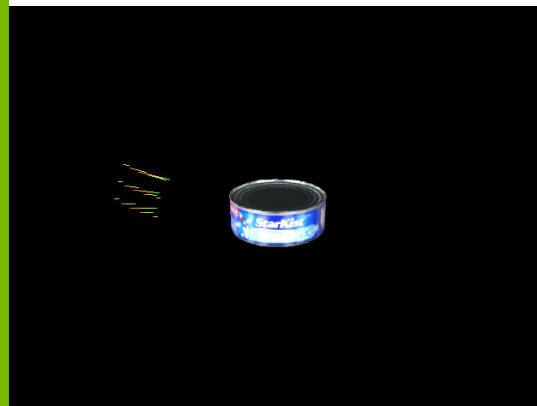
Observation



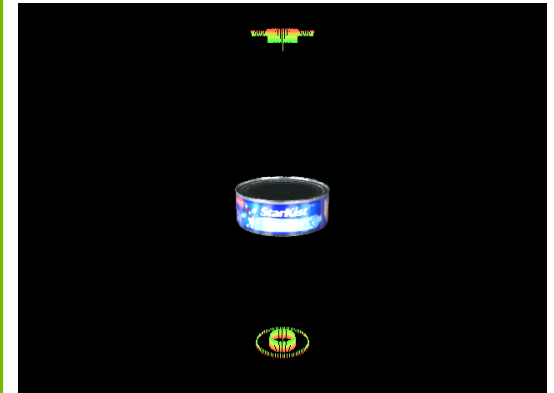
Orientation
uncertainty



Shape symmetry



Texture breaks
symmetry

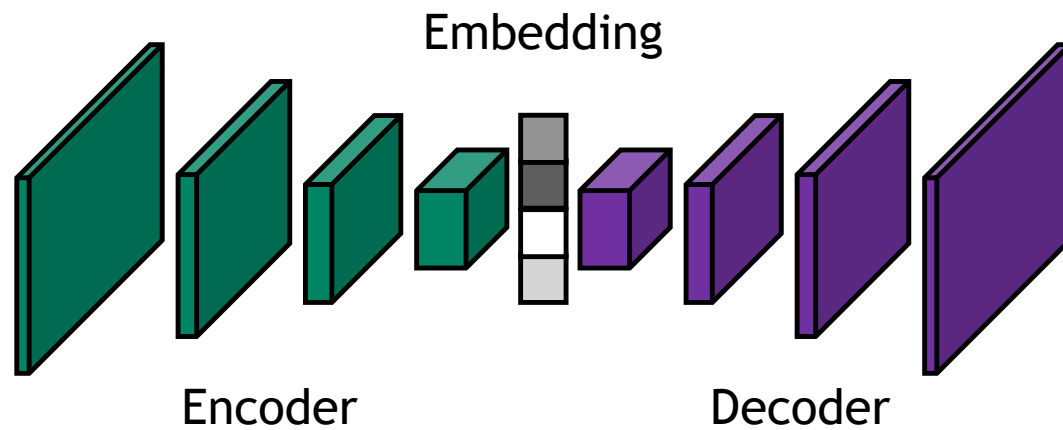


View-based
uncertainty

IMPLICIT ROTATION LEARNING



Input



Reconstruction

Sundermeyer et al. Implicit 3D orientation learning for 6D object detection from RGB images. In ECCV, 2018.

ROTATION ESTIMATION WITH CODEBOOK MATCHING

[Sundermeyer et al. ECCV 2018]

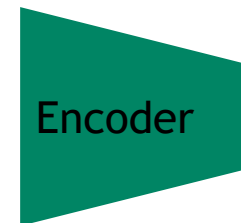
- Inherently handles symmetric views
- Only orientation, no translation
- Single image, single estimate



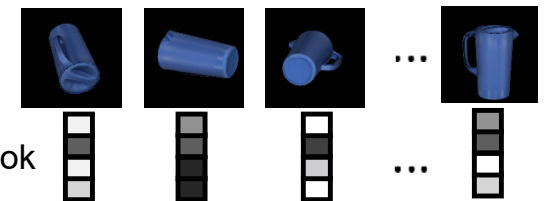
Input



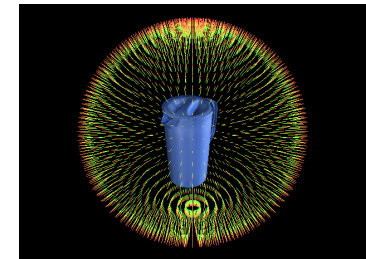
Detection



Codebook



Similarity scores

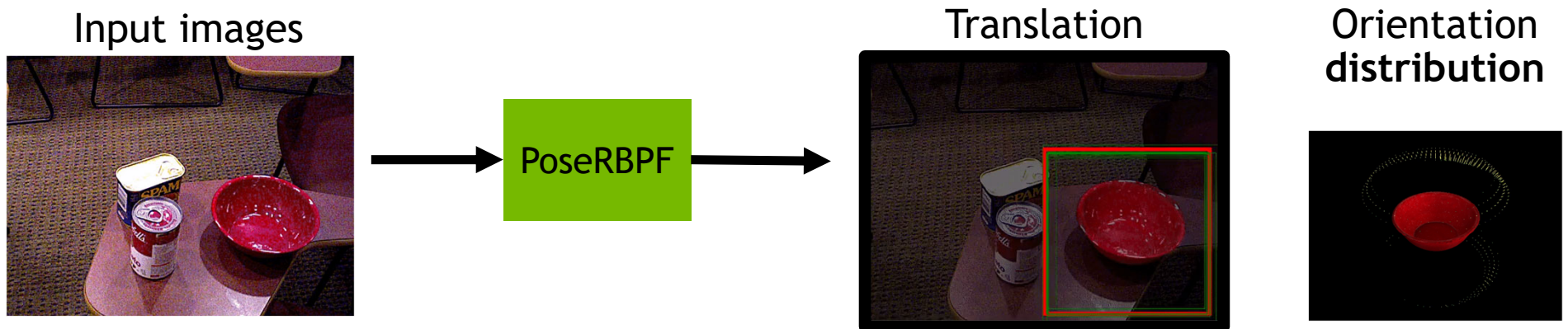


191,808 discrete rotations

PoseRBPF

Generic and Efficient Framework for 6D Object Pose Tracking

- Main idea: Instead of sampling all state dimensions, sample some of the dimensions and solve remaining ones analytically
- Successfully applied to SLAM, tracking, activity recognition, ...
- Here: Sample translation and estimate discrete orientation distribution over orientation



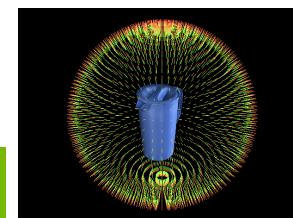
Xinke Deng*, Arsalan Mousavian, Yu Xiang, Fei Xia*, Timothy Bretl and Dieter Fox. PoseRBPF: A Rao-Blackwellized Particle Filter for 6D Object Pose Tracking. In RSS, 2019 (*intern at NVIDIA).

PoseRBPF: Particle Representation

$$\mathcal{X}_i = \{\mathbf{T}_i, P(\mathbf{R}_i | \mathbf{T}_i, \mathbf{Z}_{1:k})\}$$

3D Translation
 T_i

Orientation Distribution
 $P(R_i | T_i, Z_{1:k})$

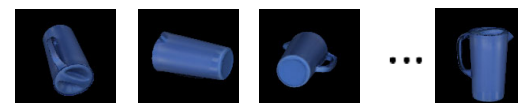


191,808 bins



RoI

Discretized Rotations



Codebook



Encoder



Particle
Code



Rotation Likelihood

PoseRBPF: Observation Update

Compute posterior

$$P(\mathbf{R}_k | \mathbf{T}_k^i, \mathbf{Z}_{1:k}) \propto P(\mathbf{R}_k | \mathbf{T}_k^i, \mathbf{Z}_k) P(\mathbf{R}_k | \mathbf{R}_{k-1})$$

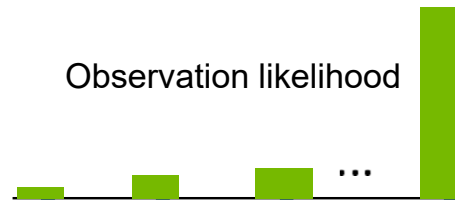
Weights



Encoder



Observation likelihood



Encoder



Normalizer



Encoder



Orientation Distribution



Particle
Rols

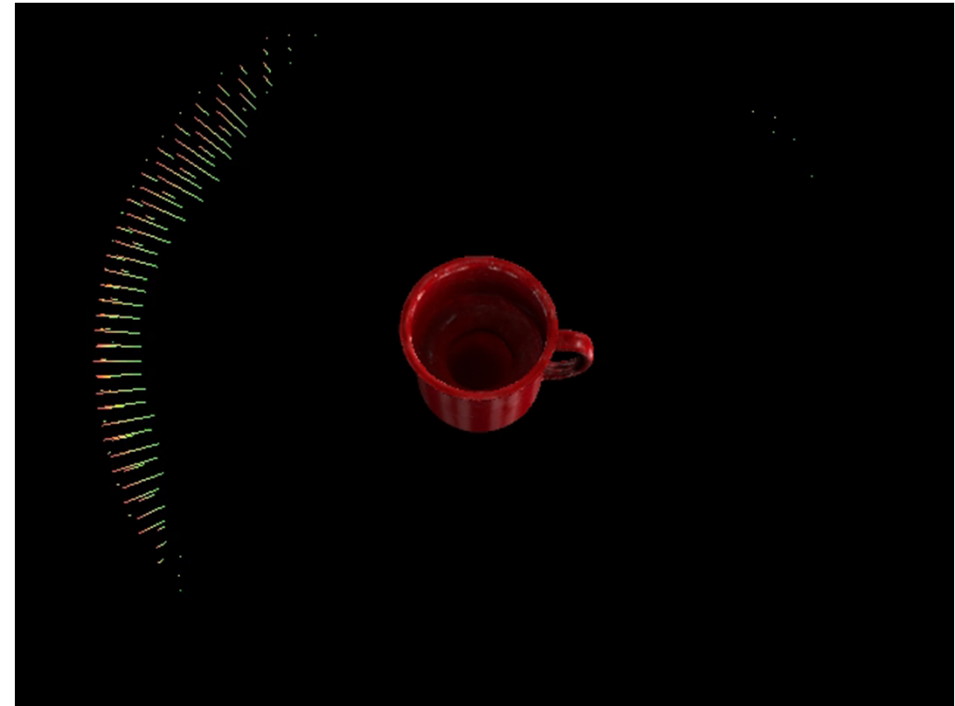
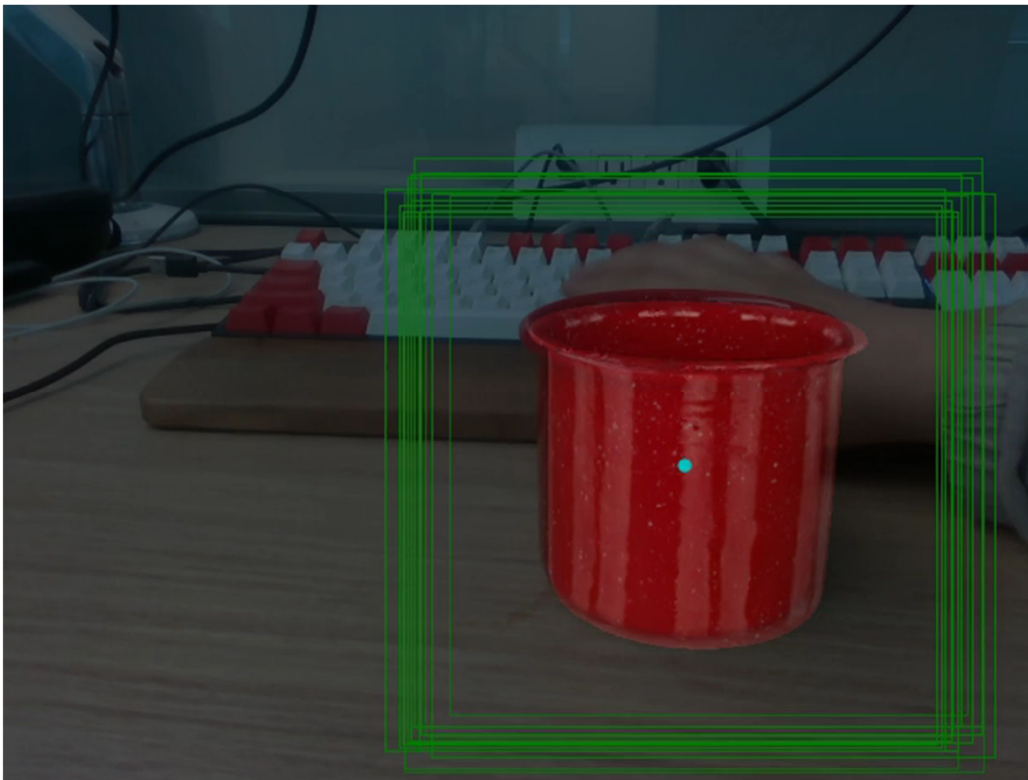
Particle
Code

Results: YCB Objects

Example: YCB mug (50 particles, ~20fps)

YCB-Video RGB

- PoseRBPf:
ADD: 62.1, ADD-S: 78.4
- PoseCNN:
ADD: 53.7, ADD-S: 75.9



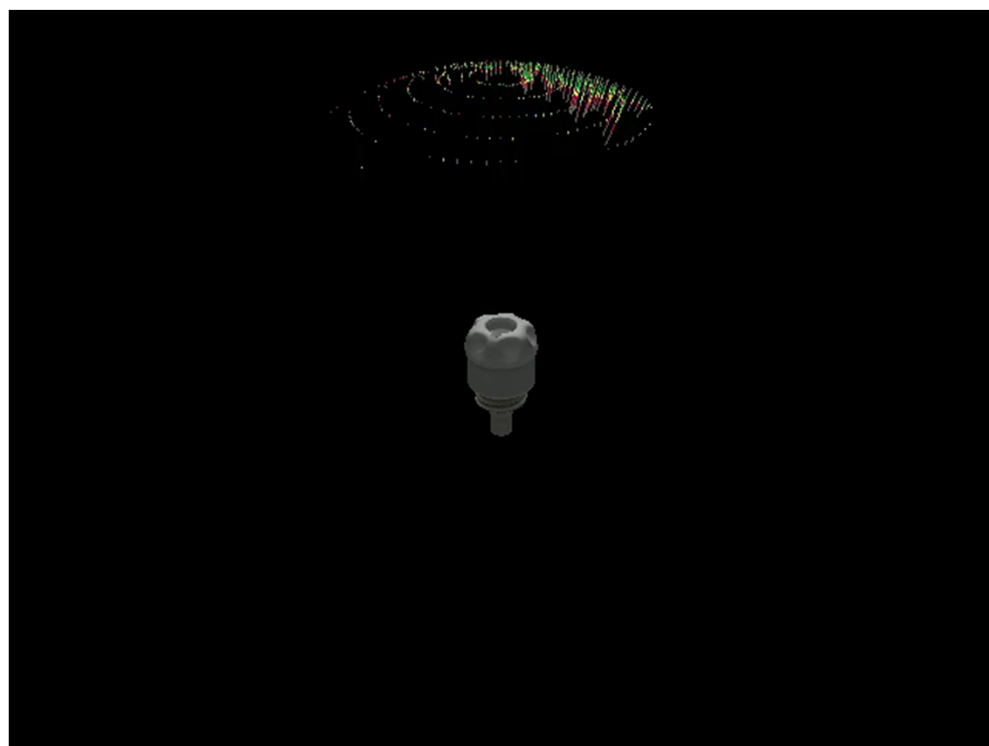
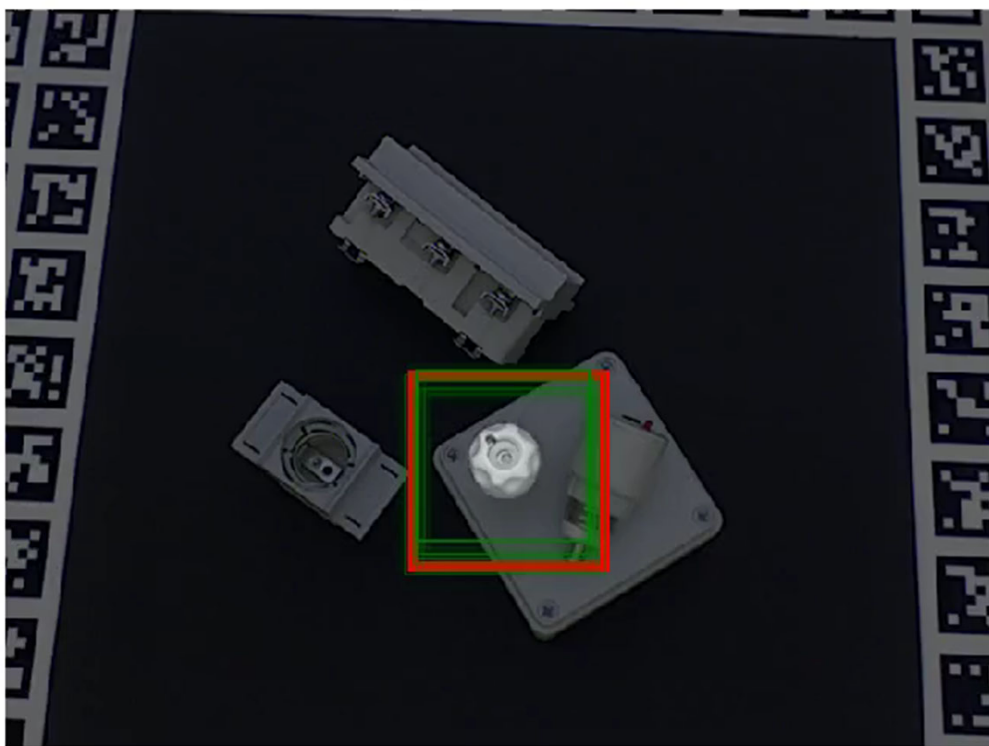
Results: TLess Objects

Example: TLess 01 (100 particles, ~11fps)

TLess RGB

Object recall for $\text{Err_vsd} < 0.3$:

- PoseRBPF: 41.47%
- Sundermeyer et al: 18.35%



ROBUSTNESS?

Self-supervised Learning



Lift-long Learning

SELF-SUPERVISED 6D POSE ESTIMATION

Interactive data collection (5x)

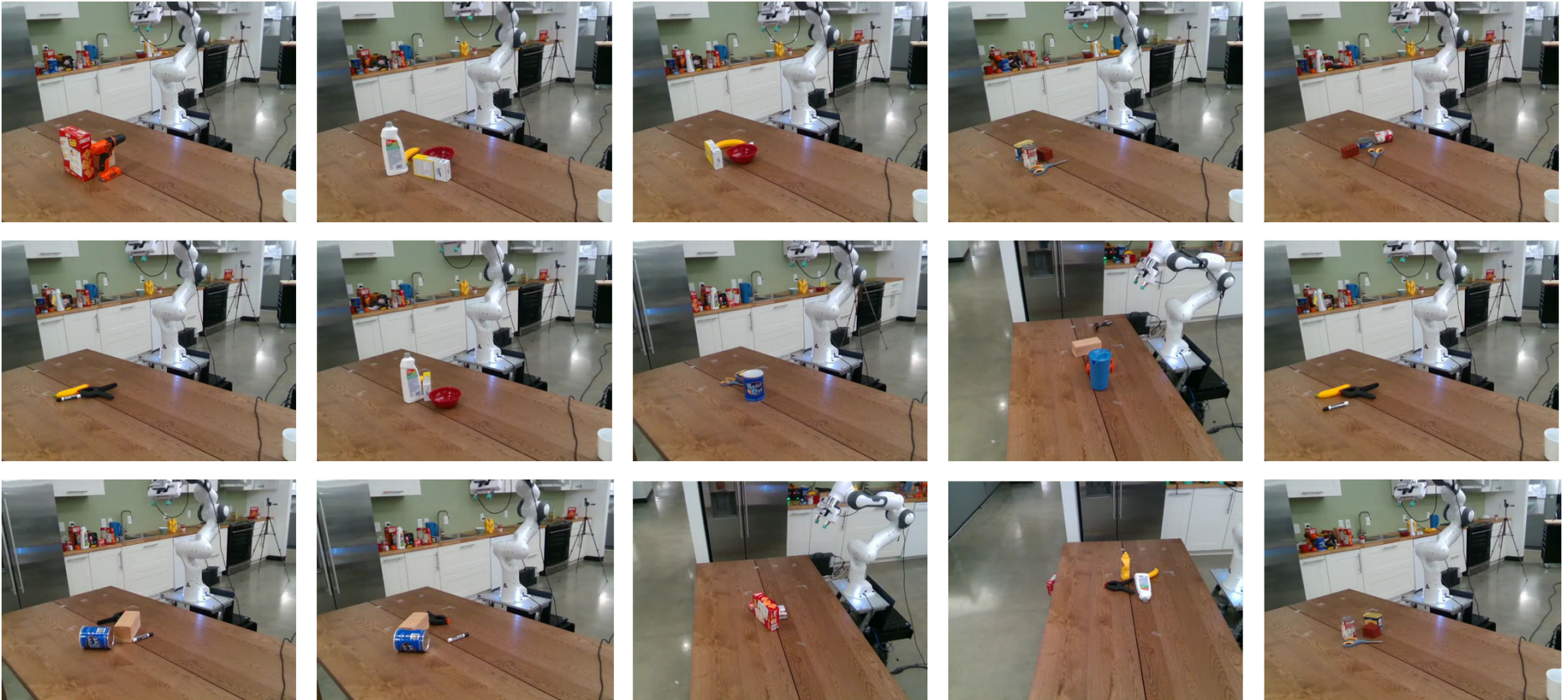


Generated pose annotations



The robot can automatically interact with the objects to create new scenes by grasping and pushing.

DATA COLLECTION



The system can automatically collect large scale datasets.

GRASPING RESULTS

Success Rate (30 grasps)

- Synthetic: 46.7%
- Fine-tuned: 86.7%

trained with only synthetic data



fine-tuned with self-annotated data



Better pose estimates lead to higher grasp success.

GRASPING RESULTS

Scene 1



Scene 2



Here, we show the performance of our system on pick-and-place tasks.

POSERBPF

- Estimates full 6D object pose distributions
- Combines Bayesian filtering with deep learning for embeddings
- Handles symmetric objects and pose uncertainty
- Fully trained in simulation, state-of-the-art results on RGB only datasets
- Enables us to build a self-supervised 6D pose estimation system for manipulation

Questions?

