

# Example-Driven Model-Based Reinforcement Learning for Solving Long-Horizon Visuomotor Tasks

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## Long-Horizon Vision-Based Manipulation

Demands learning a repertoire of visuomotor skills that are:

**Robust**  
High Success Rates

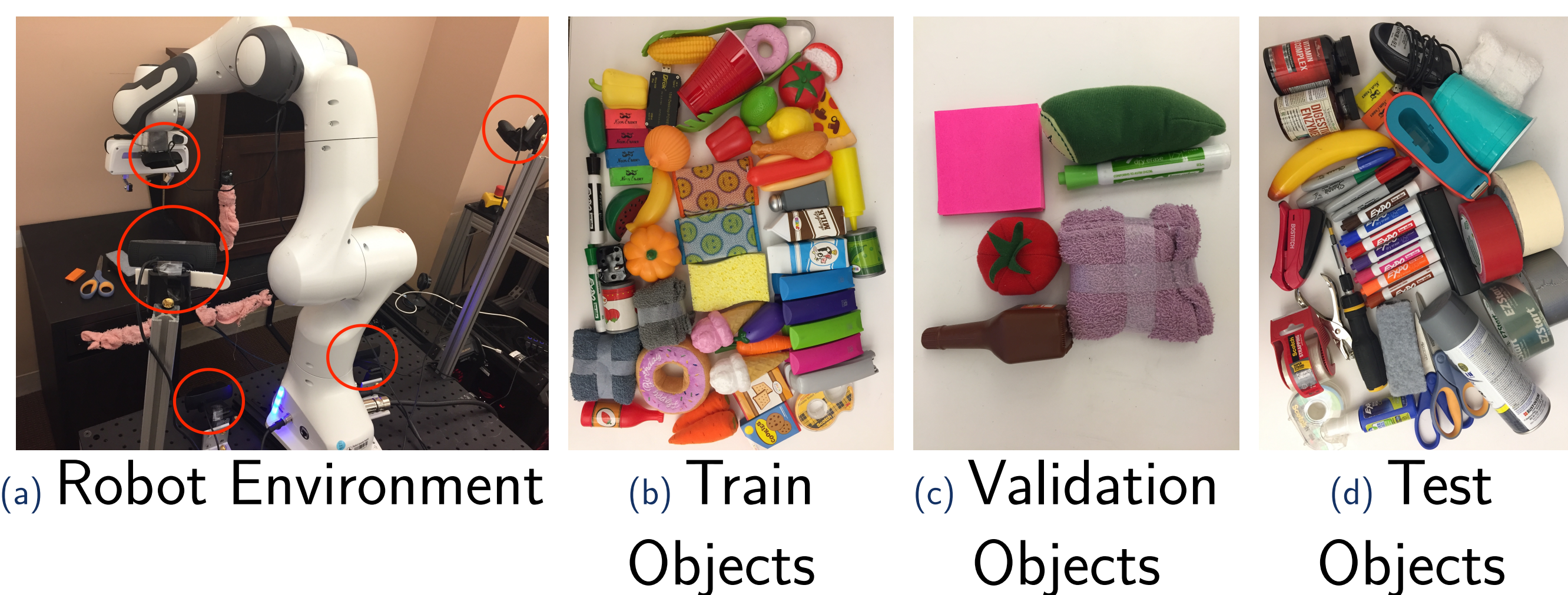
**Persistent**  
Closed-loop and Reactive

### Key Insight

Use human-provided example images as supervision to learn a repertoire of skills, groundings, and success detectors

- Provides supervision signal for reward learning in RL
- Enables skill grounding in long-horizon task planning
- Encourages closed-loop visuomotor control

### Performance Evaluation



Successful trials (out of 20) and success rates (%)

	EMBR (Ours)	EMBR w/o $f_{\mathcal{T}}$	EMBR w/o $f_{vae}$ , $f_{\mathcal{T}}$ (Qt-Opt)
Avg. Skill Success	<b>96.5%</b>	89.1%	85.9%
Avg. Long-Horizon Task Success	<b>85.0%</b>	70.0%	58.3%

\* Kalashnikov et al. QT-Opt: Scalable Deep Reinforcement Learning for Vision-Based Robotic Manipulation, CoRL'18

### Key Takeaways

- Q-function is important for low-level skill performance
- Robustness of replanning w/ model is critical for long-horizon performance

### Future Work

- Reduce the amount of human supervision
- Expand task scope and handle partial observation
- Generalization to new environment setup

## Example-Driven Model-Based Reinforcement Learning (EMBR)

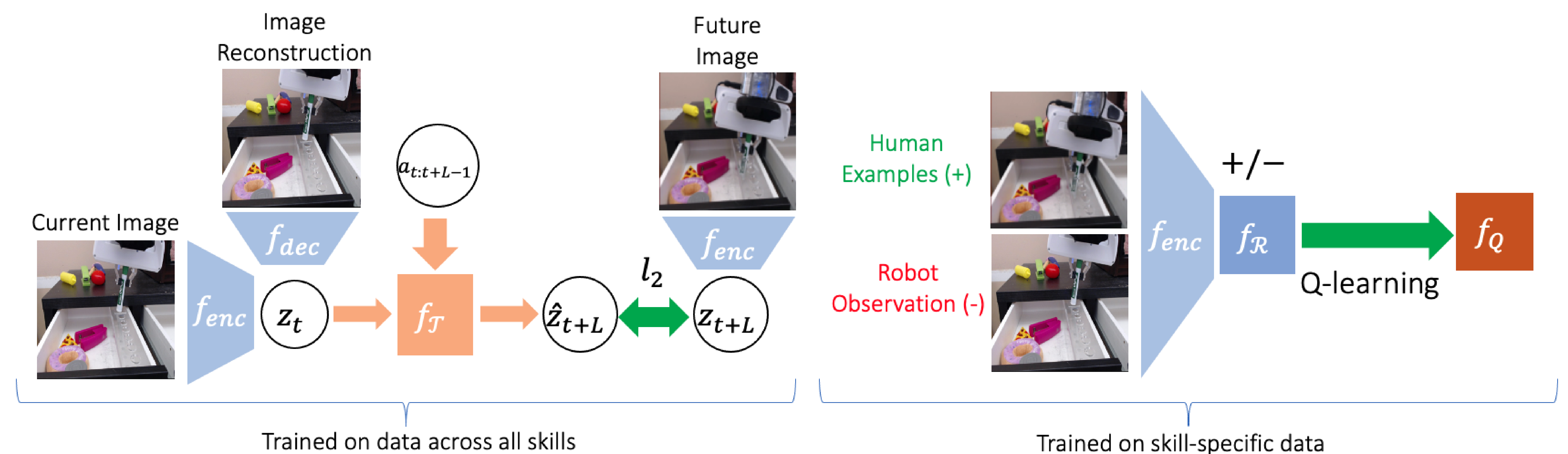
A framework for learning grounded visuomotor skills sequenced by symbolic planners to complete long-horizon tasks.

Step 1: Learn a repertoire of skills with example-driven model-based reinforcement learning (EMBR)

Step 2: Specify model over skills, using image classifiers for representing pre- and post-conditions

Step 3: Run symbolic planner with the repertoire of visuomotor skills learned in Step 1

### Learning a Repertoire of Visuomotor Skills with EMBR



- Learns a low-dimensional latent space with a VAE and a latent dynamics model
- Obtains rewards by learning image classifiers
- Learns Q-functions for model-based control

### Real-Robot Experiments: Long-Horizon Tasks with Novel Objects from Raw Image Observations

