

# NRI: FND: Robust Inverse Learning for Human-Robot Collaboration

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Project URL: <http://thinc.cs.uga.edu>

## Challenges

- **Co-bots** working with humans trained using passive demonstrations in real-world scenarios
- The demonstrations by human experts have **confounding elements, occlusion, and occur in uncontrolled settings**
- The human experts may interleave different ways of solving a given task exhibiting differing preferences

## Approach

- **Hierarchical Bayes model** for IRL in partially-controlled settings
- Develop **online learning, incremental learning** IRL algorithms
- **Maximum entropy multi-task** IRL algorithm

## Scientific Impacts

- Relax knowing the expert's true trajectory due to perception noise. Exploit indirect data sources toward IRL
- A framework and new method for online IRL
- Generalizes well-known MaxEnt IRL to multi-task environments with unknown task labels

## Broader Impacts

- Reduce the deployment time of apprentice robots
- Expand IRL applicability to more **natural and continuous task demonstrations in environments that cannot be controlled**

## Inverse RL in Partially-Controlled Settings

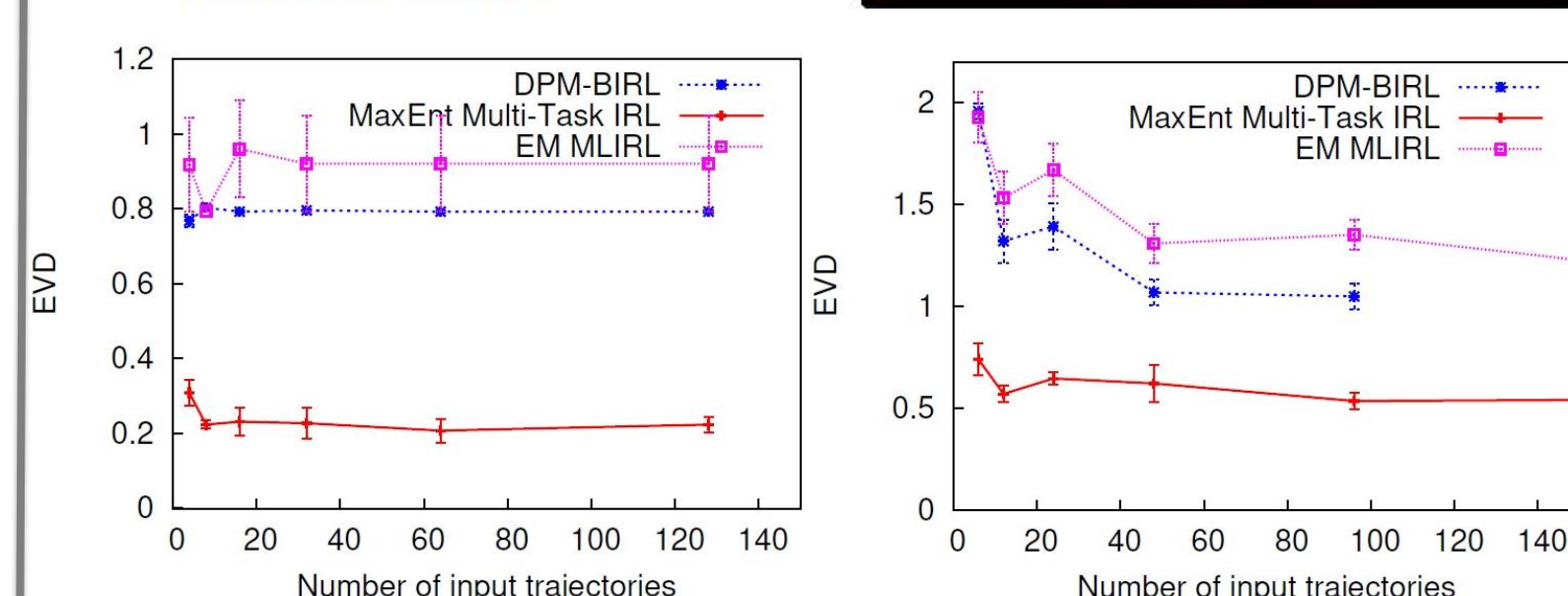
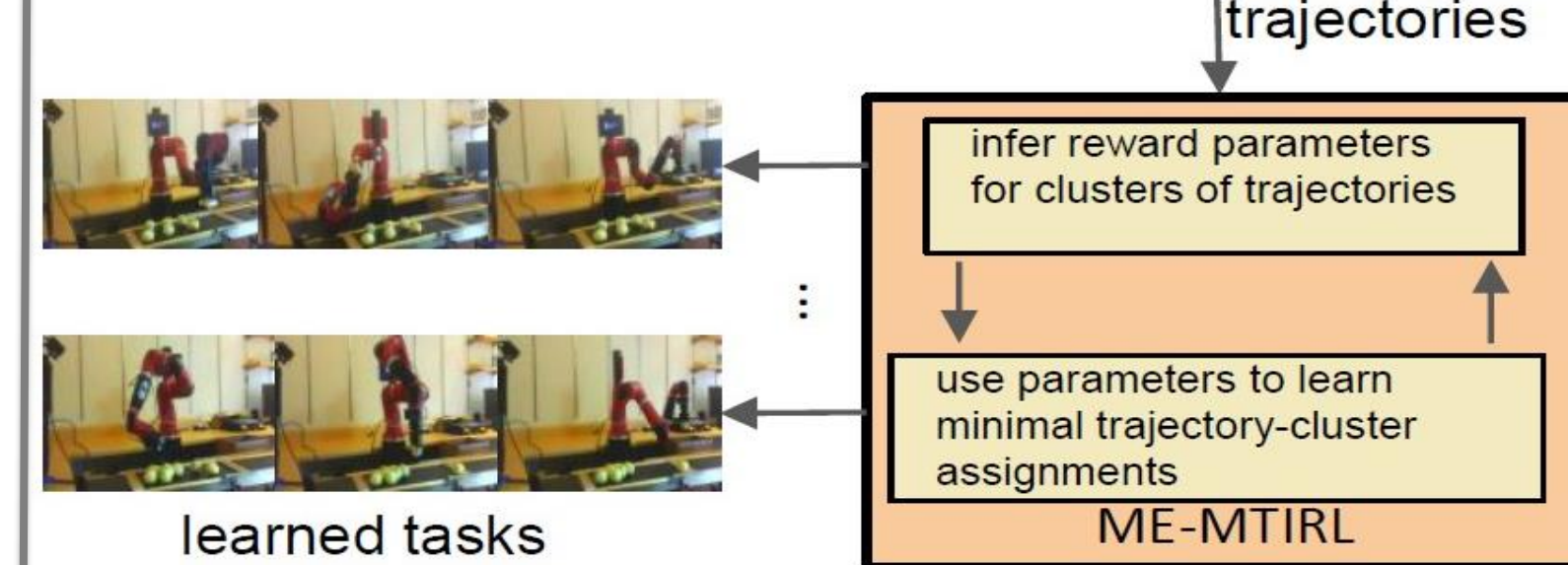
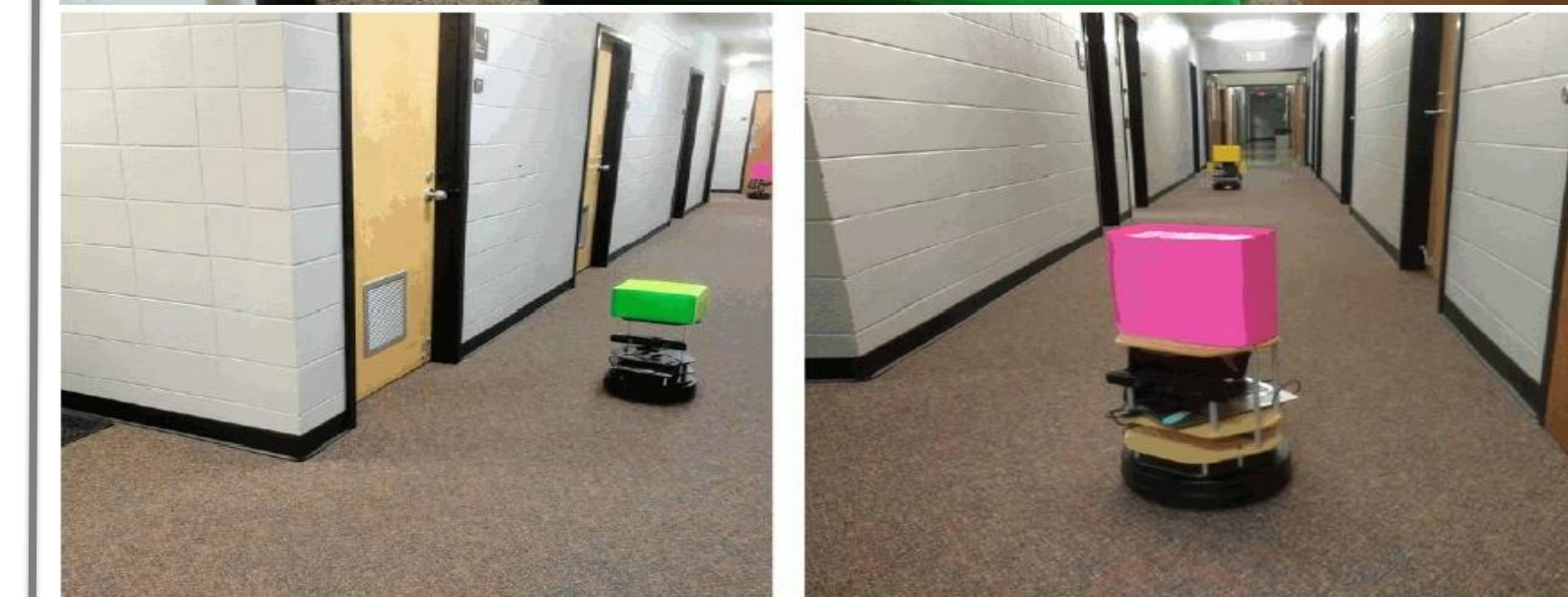
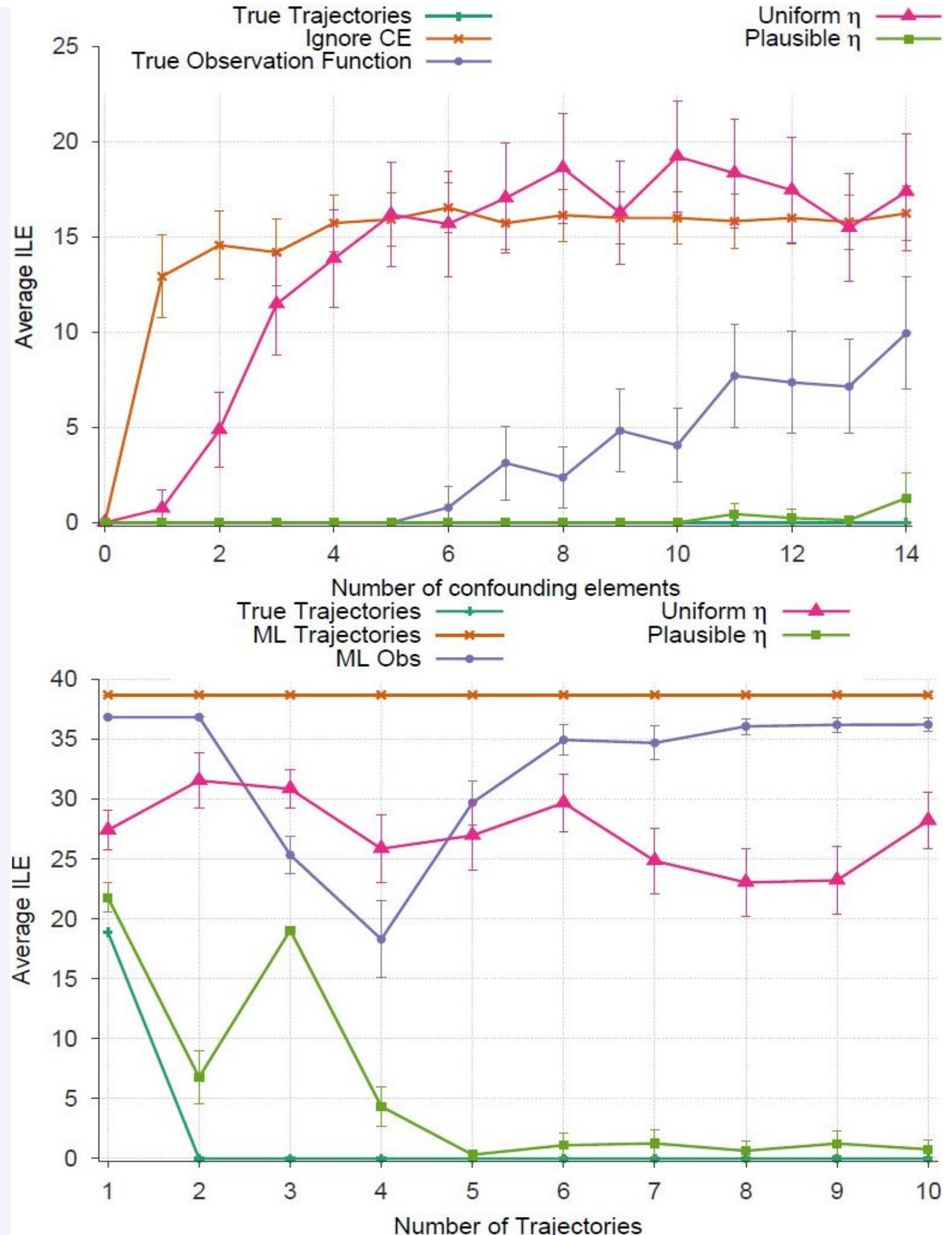
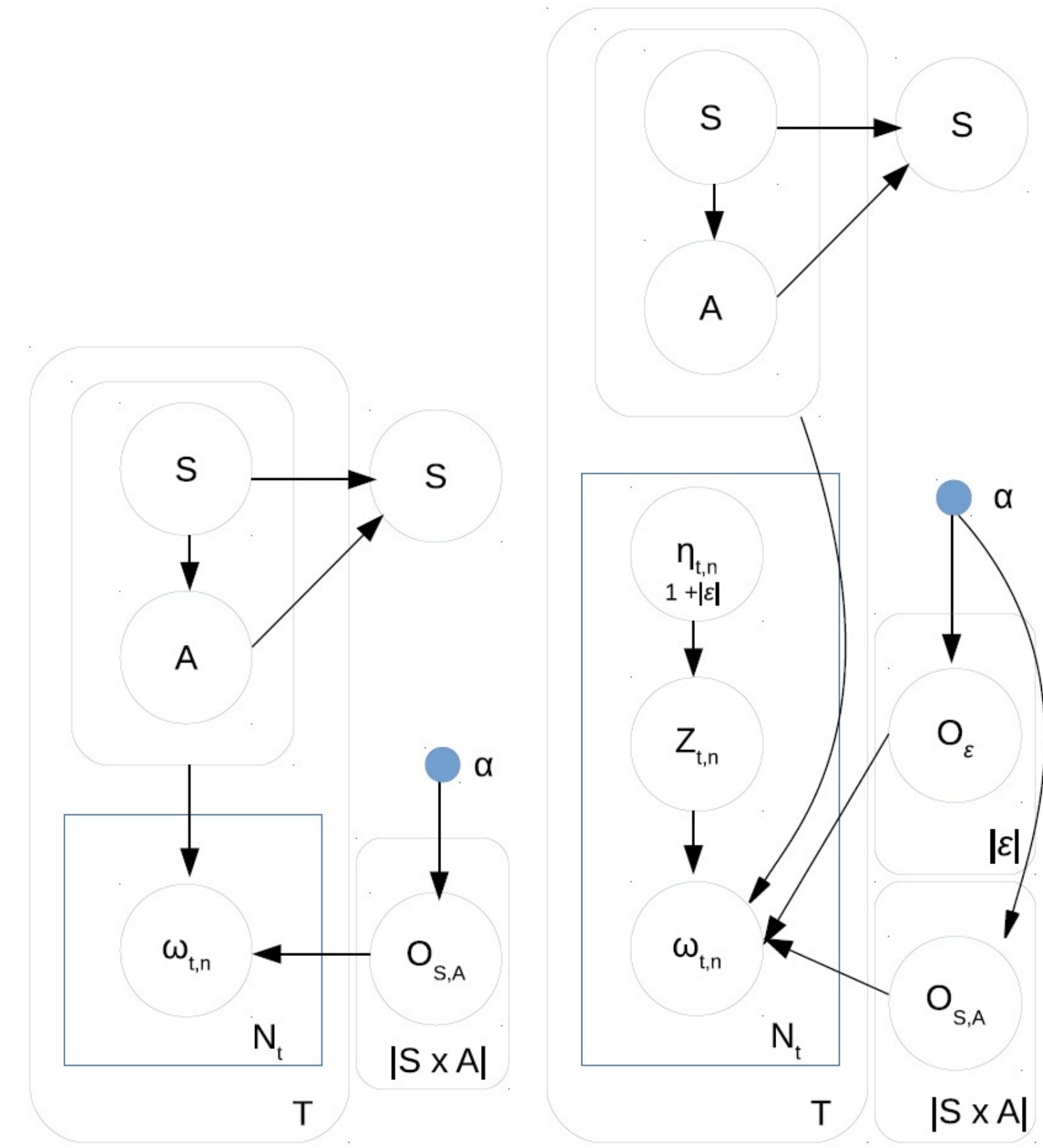
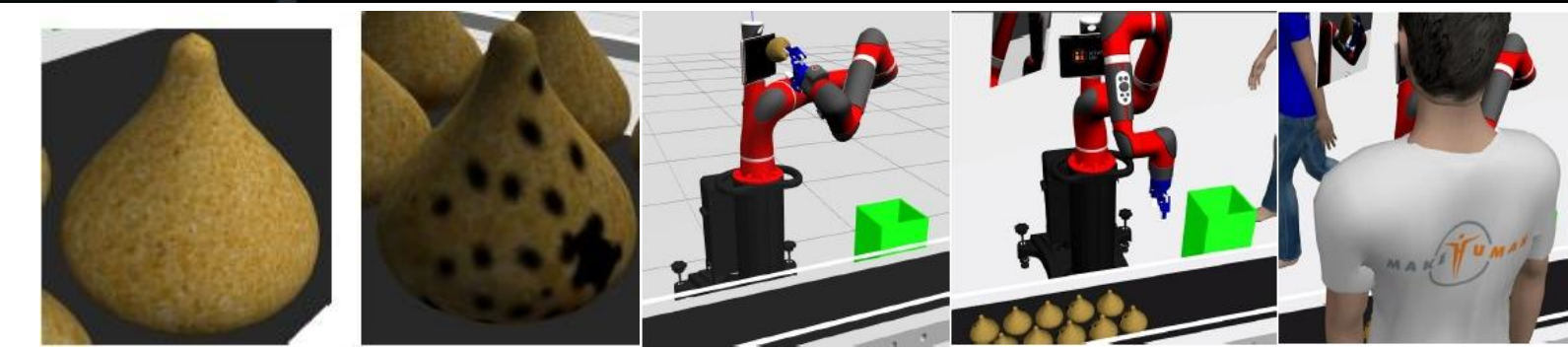
- A **general hierarchical Bayesian model** that captures the uncertainty present in observations during expert agent demonstrations in **partially controlled, real-world situations** to perform apprenticeship learning.
- Generalize the **E-step of HiddenDataEM** to environmental and sensing noise to compute a distribution over possible demonstrations  $X$  given the observations
- Deal with partially controlled environments by **finding the expert's trajectory distribution and exploiting indirect observations.**

### Experiment 1:

- Formative evaluation on Gridworld with added confounding elements.

### Experiment 2:

- Summative assessment using Robotic Onion Sorting
- The Sawyer robot arm is tasked with inspecting onions moving down a conveyor belt and sorting good onions from blemished ones.



Expert		(TP,FP,FN,TN)	P%,R%
Expert	P-I-P	(7,0,12,5)	100.00, 58.33
	R-P-P	(9,4,8,3)	69.23, 75.00
Learned (ME-MTIRL)	P-I-P	(7,1,11,5)	87.50, 58.33
	R-P-P	(9,4,8,3)	69.23, 75.00
Learned (DPM-BIRL)	P-I-P	(7,3,9,5)	70.00, 58.33
	R-P-P	(9,4,8,3)	69.23, 75.00
Learned (EM-MLIRL)	P-I-P	(6,3,9,6)	66.67, 50.00
	R-P-P	(6,4,8,6)	60.00, 50.00

## I2RL: Online IRL Under Occlusion

- Introduce **I2RL**, framework for online IRL, which establishes the key components and offers candidate stopping criteria.
- Establish Key theoretical method called **online latent maximum entropy IRL**.
- It offers the capability to perform **online IRL** in contexts where portions of the observed trajectory may be occluded.

### Task:

- Two independent mobile robots patrolling corridors of varying configurations. A third robot observes the patrollers and uses learned reward functions to **predict patrolling trajectories** and **identify a path** to the goal location.

### Experiments:

- Done using physical robots (Turtlebots) to test how well I2RL can be implemented and extended in real world.
- The learner robot can only access to less than 30% observability. Compared to online GAIL, LME I2RL performs significantly better and achieves a reasonable result.

## Maximum Entropy Multi-Task Inverse RL

- An expert may solve a problem in multiple distinct ways each of which optimizes a different reward function while still sharing the features.
- Combine MaxEnt IRL with the **Dirichlet process** based minimum entropy clustering of the observed data.
- Yield a single **nonlinear optimization** problem called **MaxEnt Multi-task IRL**, which can be solved using **Lagrangian relaxation** and **gradient descent methods**.

### Task:

- Human experts will teach robot learner about how to distinguish blemished and unblemished onion and pick up blemished onion by its gripper.

### Experiment:

- Use **expected value difference averaged over the trajectories** and **Precision and Recall** as metrics.
- Use YOLO as objects recognition tools.
- Two distinct sorting methods:
  - **pick-inspect-place** and **roll-pick-place**