Bayesian Active Learning For Sim-to-Real Robotic Perception

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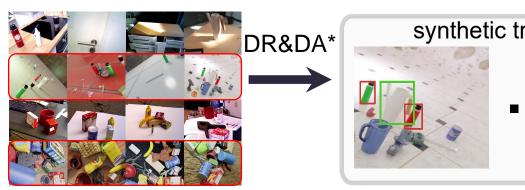










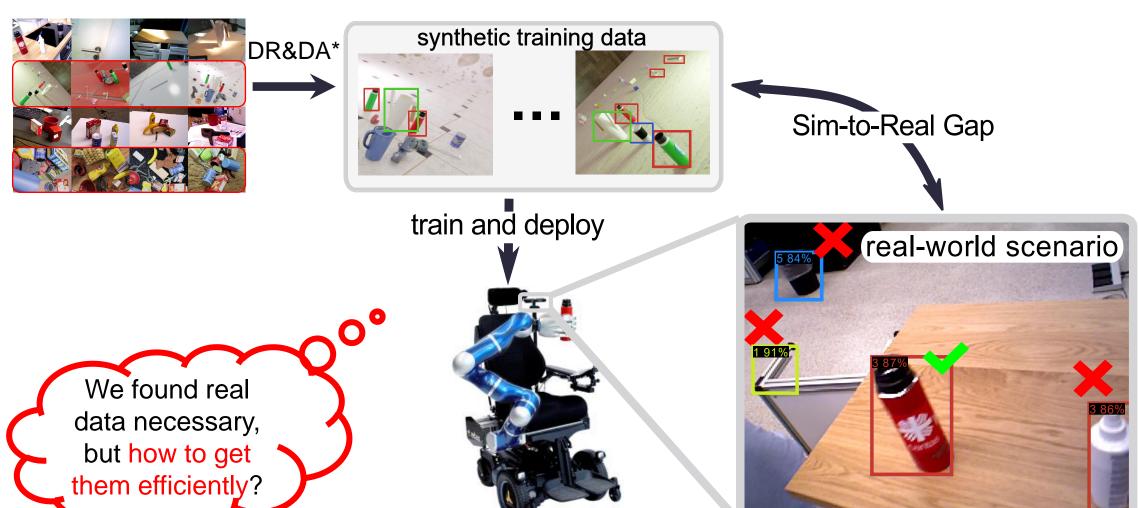








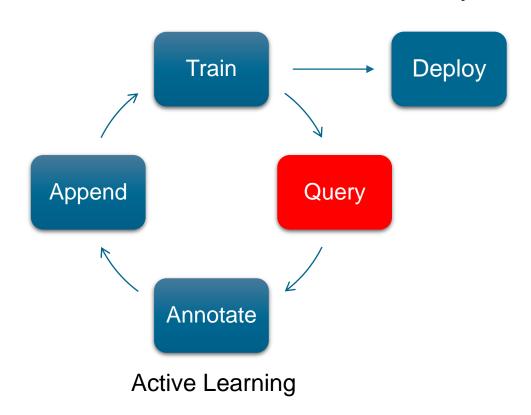


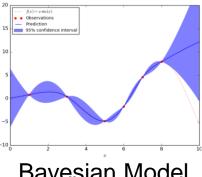


Idea: Bridging the Gap with Bayesian Active Learning

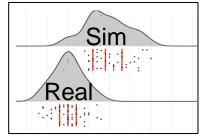


- Active learning for efficient data collection in Sim-to-Real
 - 1. Model predictive posterior for data query
 - Bayesian Neural Networks/Bayesian Object Detectors
 - 2. Mitigating label distribution shift
 - Combination of uncertainty and sub-sampling strategies





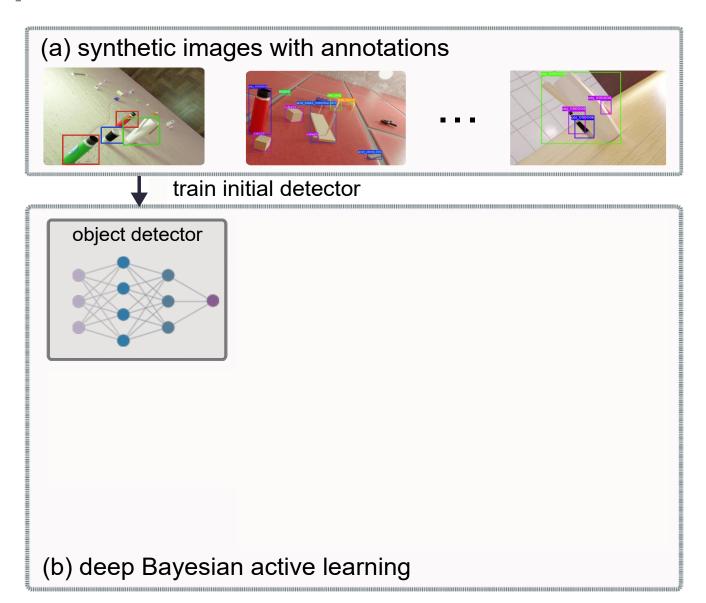
Bayesian Model



Dist. Mismatch

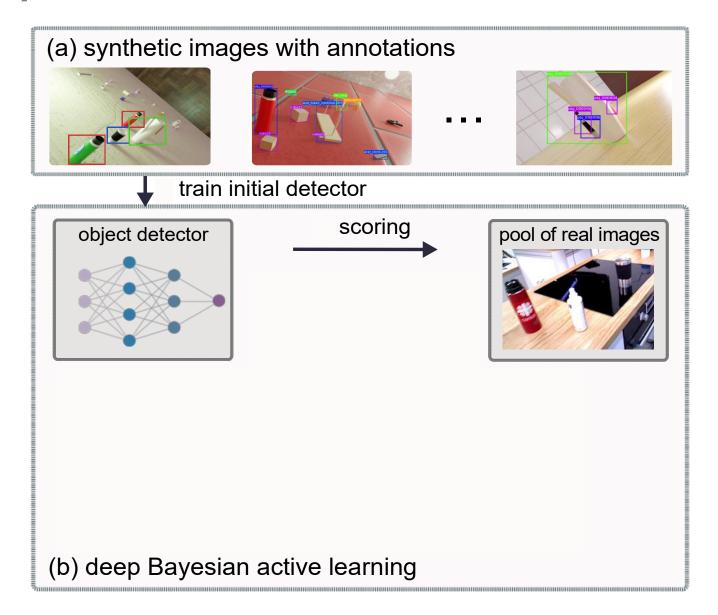




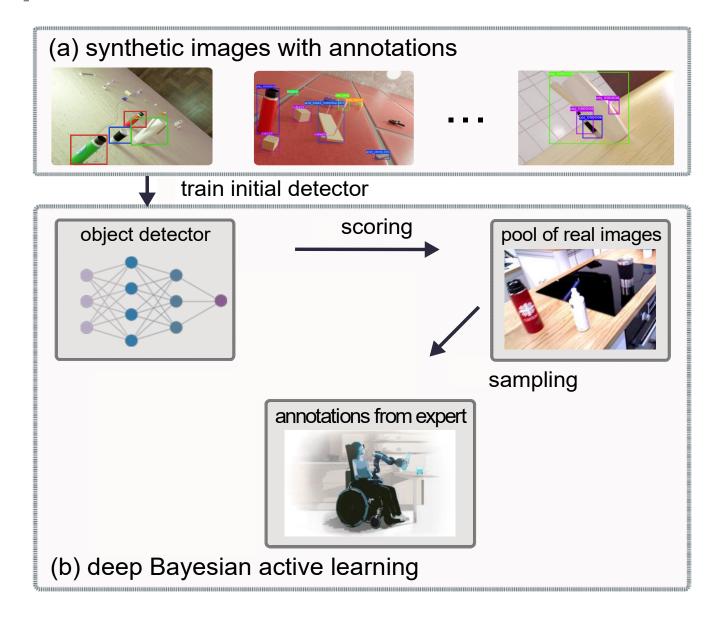




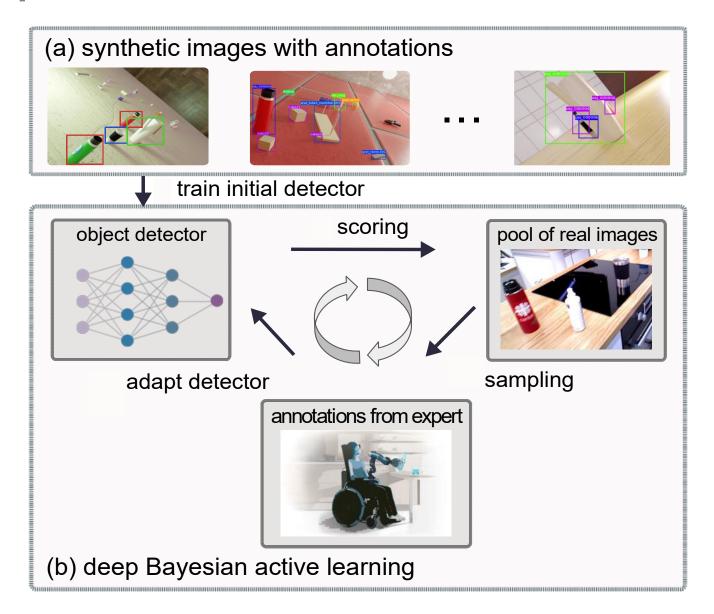








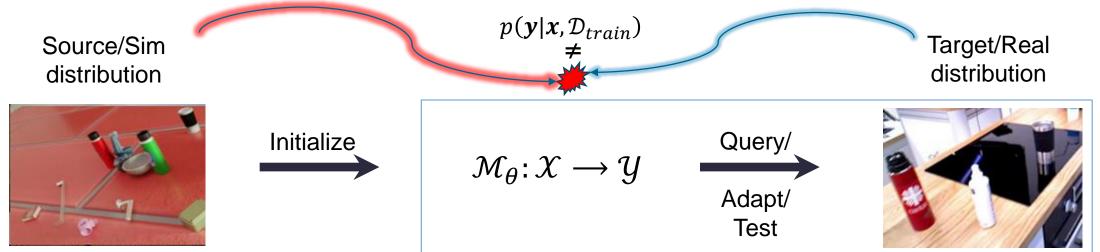






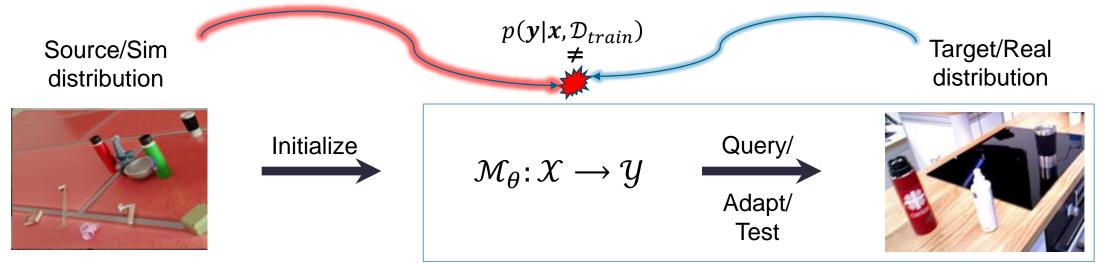
Sampling Strategy to Mitigate Label Shift





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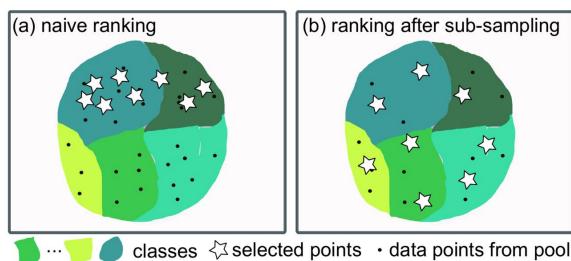


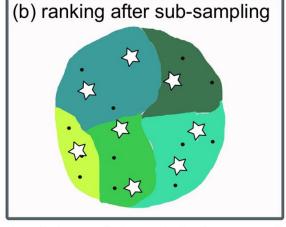


Sampling Strategy

 P_{label} : target label (uniform) distribution P_{unc} : uncertainty sampling distribution P_{SS} : sub-sampling (uniform) distribution \hat{P}_{label} : sampling label distribution

 $\hat{P}_{label} \cong P_{label}$ with $\hat{P}_{label} \propto P_{ss}P_{unc}$

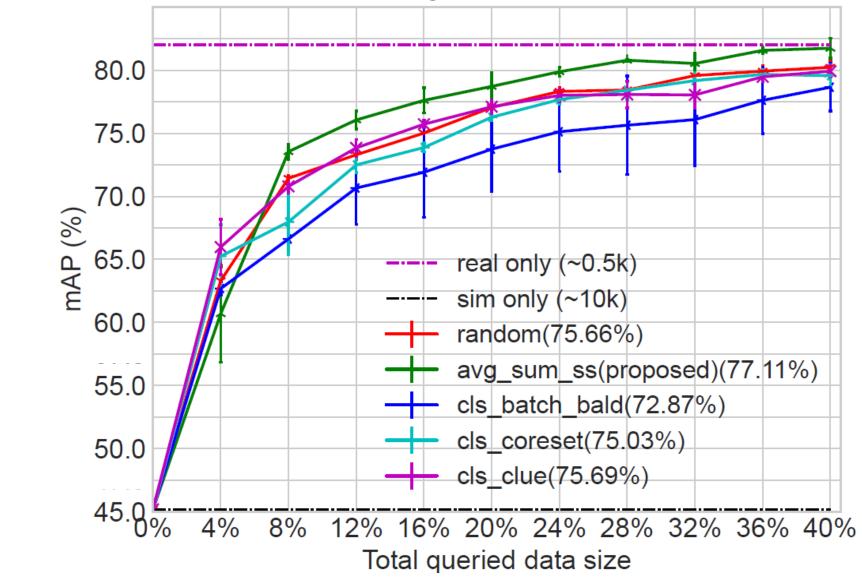




Object Detection Experiment







Real Robot Deployment



Task 1&2:

Grasping a bottle & Opening a drawer

Take-away



- An active learning based pipeline to bridge Sim-to-Real gap for robotic perception.
- A simple and effective sampling strategy to mitigate label shift for active learning in Sim-to-Real.
- Extensive empirical experiments and real robot deployment along with a failure case analysis.





Thank you! & Questions?





