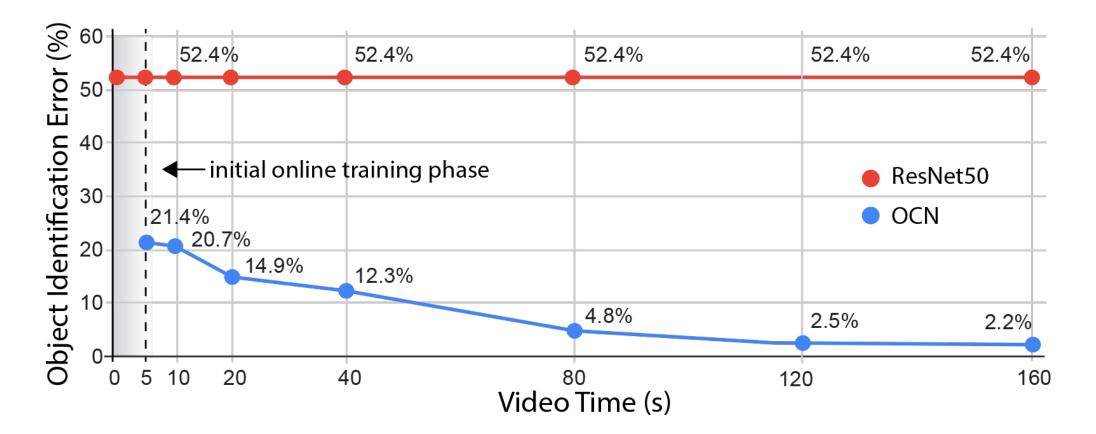


## **Objective**

- Self-teach to **discover** and disentangle **object attributes** from videos **without** using any **labels**.
- Use of **online adaptation**: the longer our online model looks at objects in a video, the lower the object identification error.
- Explore system **free of human supervision** for robotics applications. A **robot** collects its own data, trains on it, and then **identifies objects**.





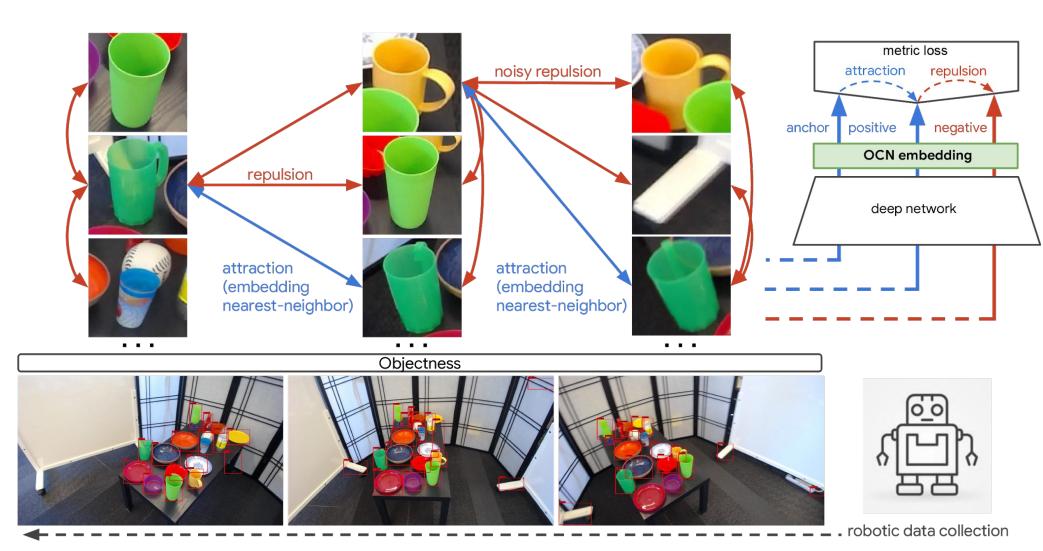
Frame m

## Approach

- Detect and **embed objects** to **extract** their **features**.
- Use metric loss to contrast similar and dissimilar **objects** in embedding space.
- **Observing objects** across different views **facilitates learning invariance** to scene-specific properties, such as scale, occlusion, lighting, or background.

# **Online Object Representations with Contrastive Learning in Videos**

Soren Pirk, Mohi Khansari, Yunfei Bai, Corey Lynch, Pierre Sermanet



By **attracting** nearest neighbors in embedding space and **repulsing** others using **metric learning**, continuous object representations naturally emerge.

## Datasets

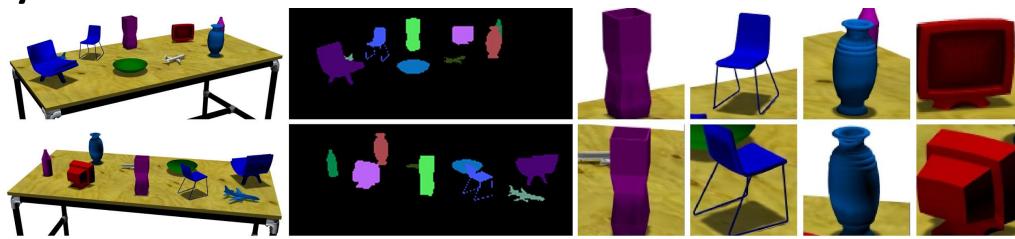
### **Real Data for Online Training** (Complex Scenes, Epic Kitchens)



### Automatic Real Data Collection with Robot

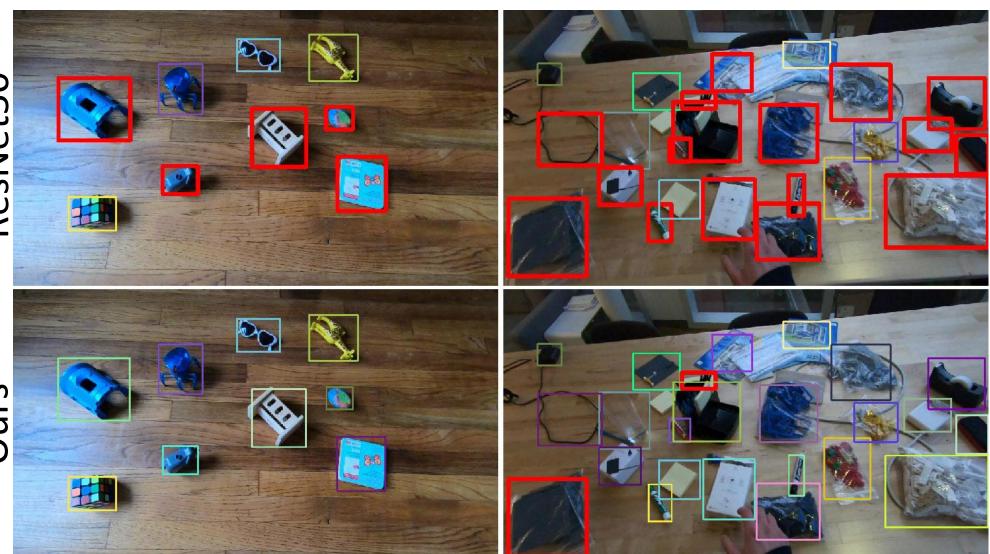


### **Synthetic Data for Evaluation**

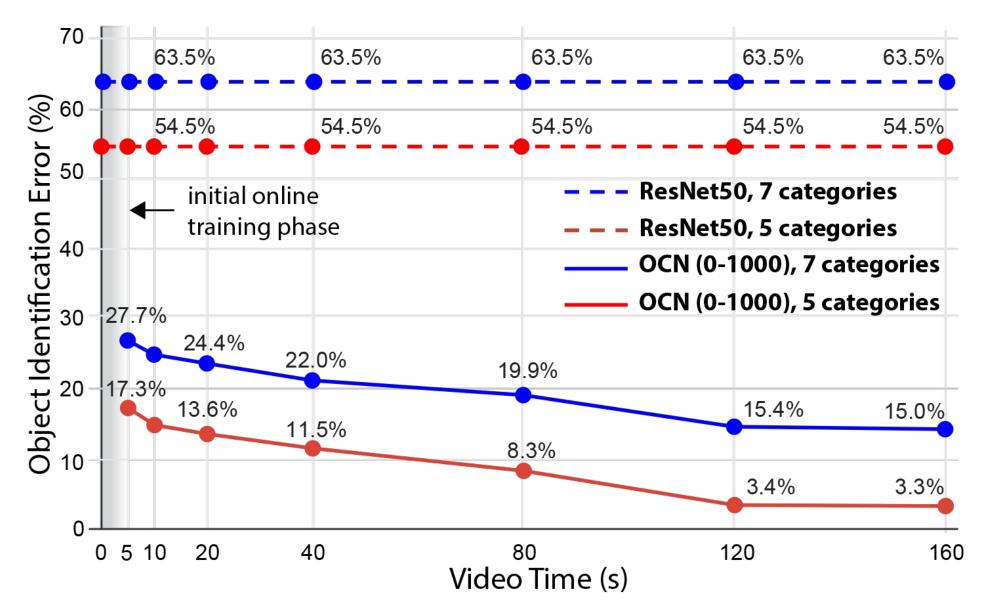


## **Experiments**

**Online Object Identification** (red boxes indicate mismatches)



Self-supervised **online training** enables adapting to unseen objects, important for robotic agents.

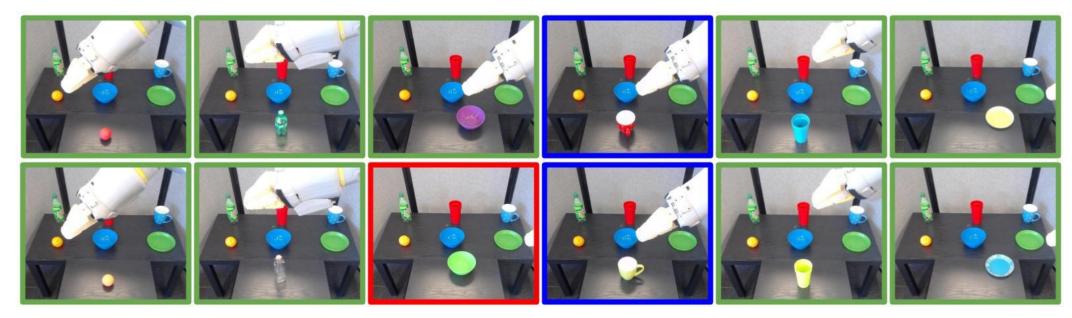


#### **Object Attribute Classification:** Comparison to Baselines

|                            | Class (12)<br>Attribute | Color (8)<br>Attribute | Binary<br>Attributes | Embedding |
|----------------------------|-------------------------|------------------------|----------------------|-----------|
| Method                     | Error                   | Error                  | Error                | Size      |
| [BL] Softmax               | 2.98%                   | 0.80%                  | 7.18%                | -         |
| [BL] OCN sup (linear)      | 7.49%                   | 3.01%                  | 12.77%               | 32        |
| [BL] OCN sup (NN)          | 9.59%                   | 3.66%                  | 12.75%               | 32        |
| [ours] OCN unsup. (linear) | 10.70%                  | 5.84%                  | 13.76%               | 24        |
| [ours] OCN unsup. (NN)     | 12.35%                  | 8.21%                  | 13.75%               | 24        |
| [BL] ResNet50 embed. (NN)  | 14.82%                  | 64.01%                 | 13.33%               | 2048      |
| [BL] Random Chance         | 91.68%                  | 87.50%                 | 50.00%               | -         |









Paper and Videos available here: https://online-objects.github.io/



#### View to View Correspondence (nearest neighbors, same scene)

Anchors Positives Distances Negatives  $\longrightarrow$ Objects of View 2

**Feature Alignment** (nearest neighbors, dataset)



Our approach allows to **organize objects** along their visual and semantic properties.

**Robotic Pointing** 

Point at **object** that is **most similar** to the one shown.

## Conclusion

Self-supervised online learning of object representations, particularly useful for **robotics** to increase robustness and adaptability to **unseen objects**.

