

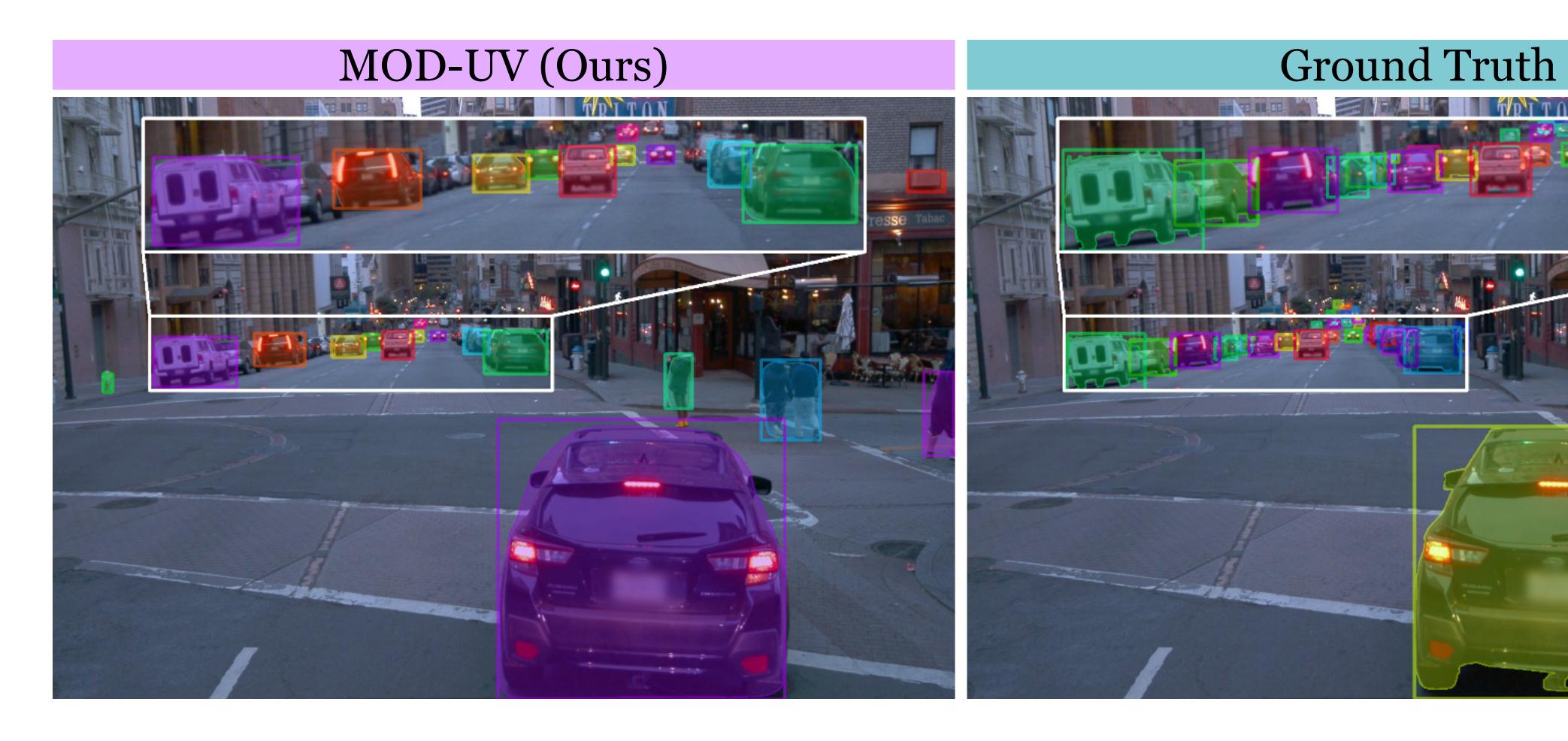
Introduction

Motivation

- Embodied agents must detect and localize objects of interest.
- To alleviate the burden of box-level annotations, prior works proposed unsupervised instance detection and segmentation via self-supervised features (e.g. DINO).
- However, it is unclear how pixels must be grouped into objects and which objects are of interest, which results in over-/under-segmentation and irrelevant objects.

Insight

 A key missing cue is motion: objects of interest are typically mobile objects that frequently move and their motions can specify separate instances.

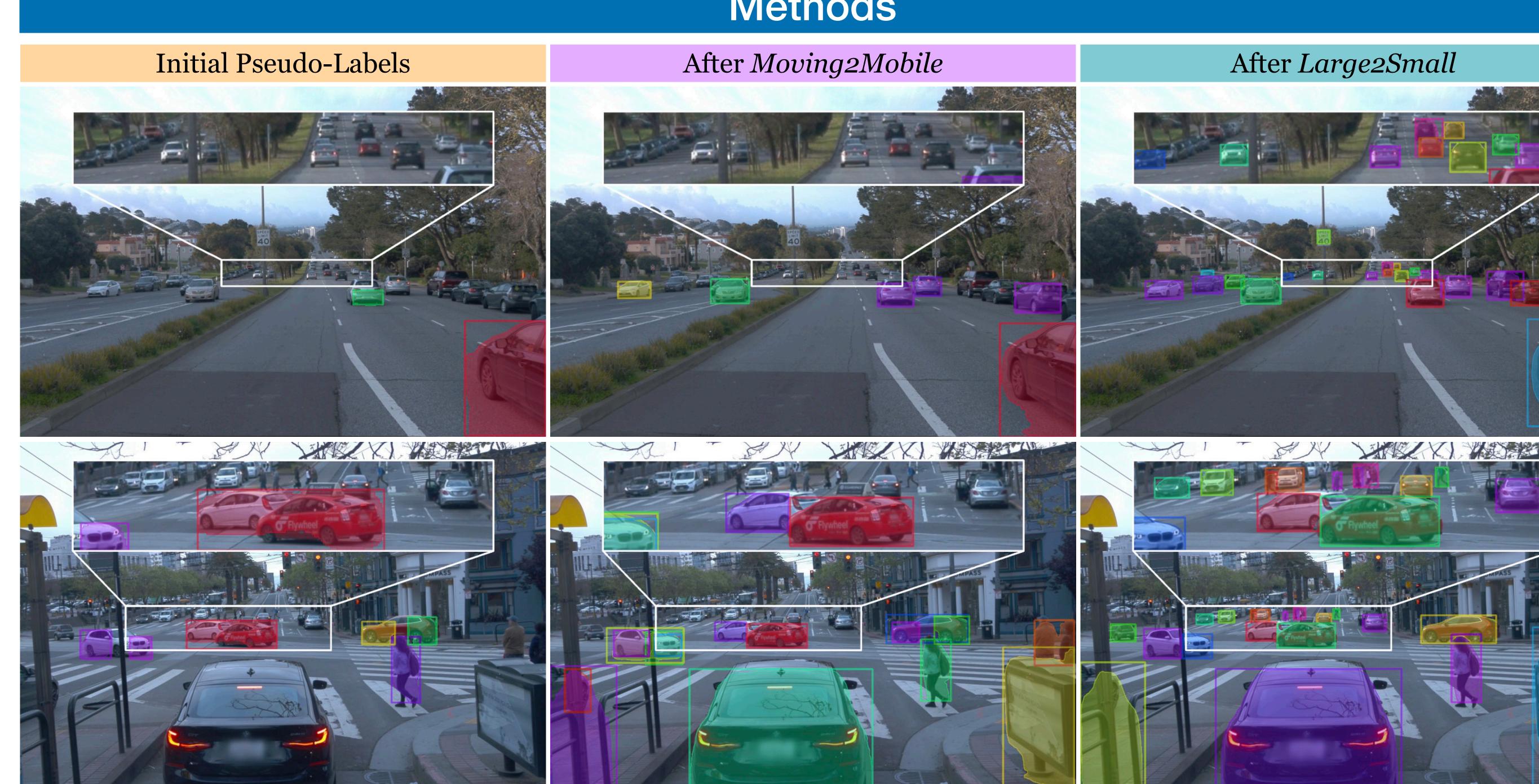


Contributions

- We propose MOD-UV, a Mobile Object Detector learned from Unlabeled Videos only.
- We argue that motion can serve as an effective cue for unsupervised training of instance-level object detectors.
- We propose a new training scheme that trains on unlabeled videos to produce a mobile object detector that can run on static images.
- We demonstrate marked improvements over unsupervised object detection baselines across a range of datasets and metrics.

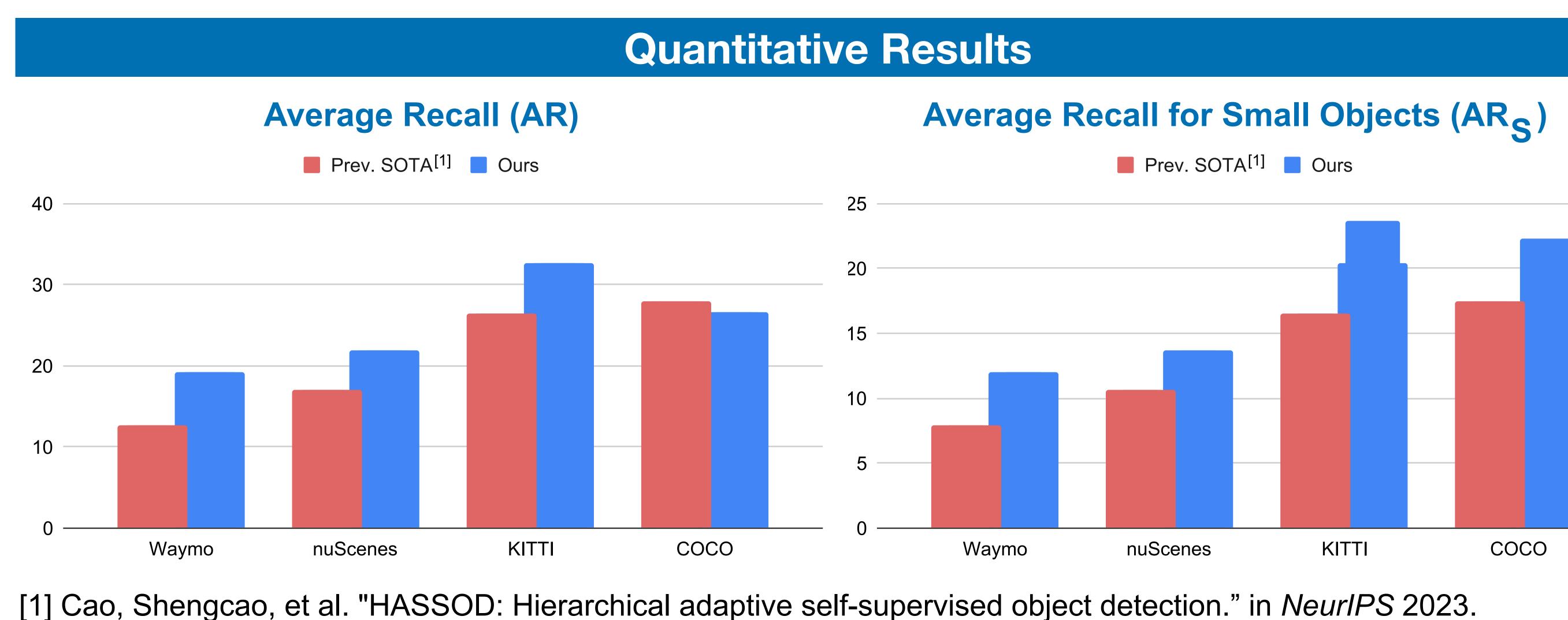
MOD-UV: Learning Mobile Object Detectors from Unlabeled Videos

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Initial Pseudo-Labels: We compute initial object seeds from self-supervised motion segmentation network. • Due to bias in motion segmentation, static and small objects are missing in the initial seeds. Moving2Mobile: We first discover the static objects by training on (single-frame, pseudo-label) pairs. • The per-frame detector cannot distinguish static objects from moving when trained on the initial labels.

Large2Small: We then discover small objects by training detectors at varying image scales.

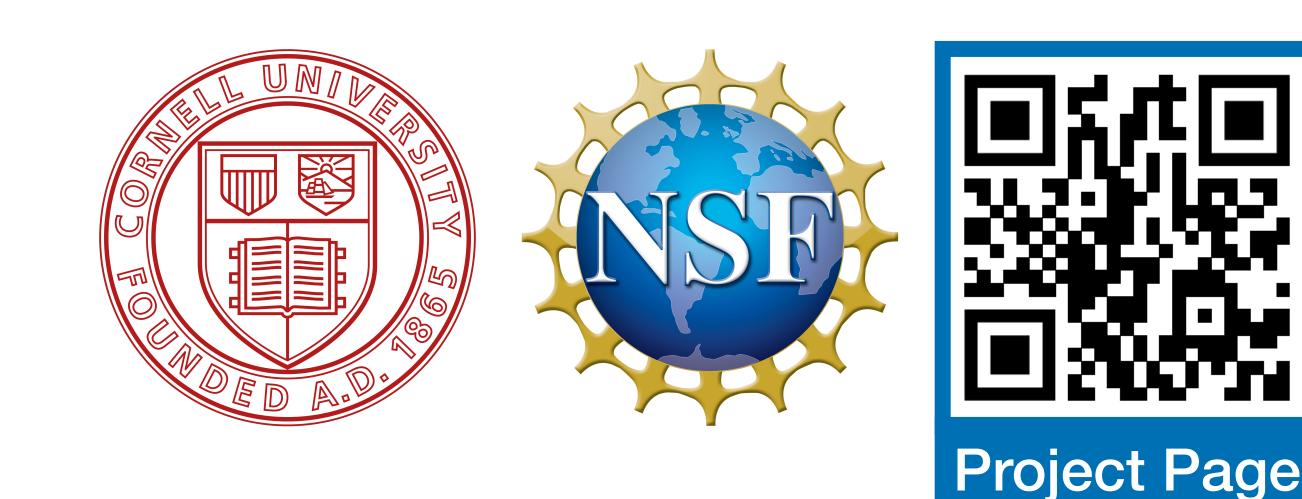


Methods

- After training with the labels after M2M, we merge predictions at varying scales to retrieve small objects.



Acknowledgements



Conclusions

• We argue that motion is an important cue for unsupervised object detection.

 We propose a new training pipeline, MOD-UV, that bootstraps from motion segmentation but removes its bias by discovering static and small objects.

 MOD-UV achieves significant improvement over prior self-supervised detectors on multiple datasets.

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