Object Instance Retrieval in Assistive Robotics: Leveraging Fine-Tuned SimSiam with Multi-View Images Based on 3D Semantic Map

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Task & Motivation

Task

Instance Specific Image Goal Navigation (InstanceImageNav)_[1] Target Instance

Motibation

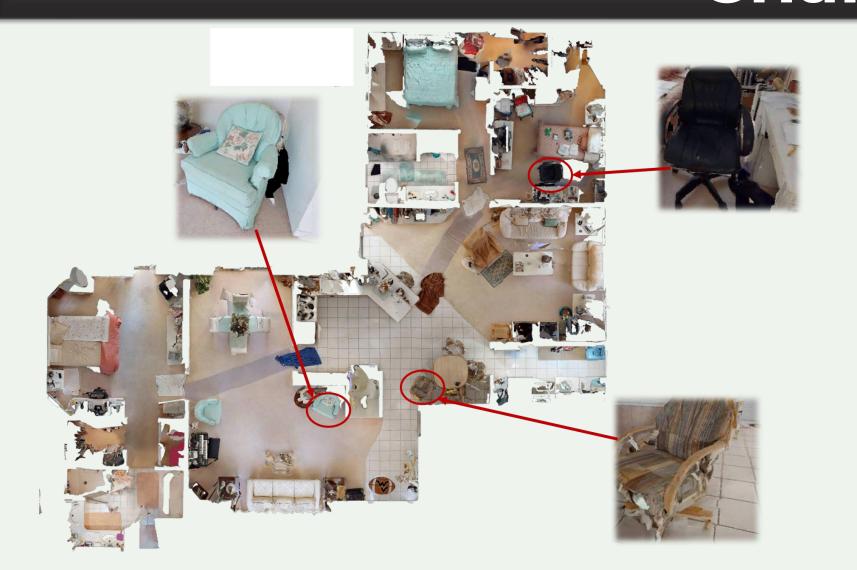
Searching Object is fundamental capability for human support robots In an environment, there are multiple instances of same class

In such environment, Robot must classify between instances of the same class of objects

Important

Previous works

Challenges





In domestic environment, there are multiple instances of same class

Challenge1

Query Image

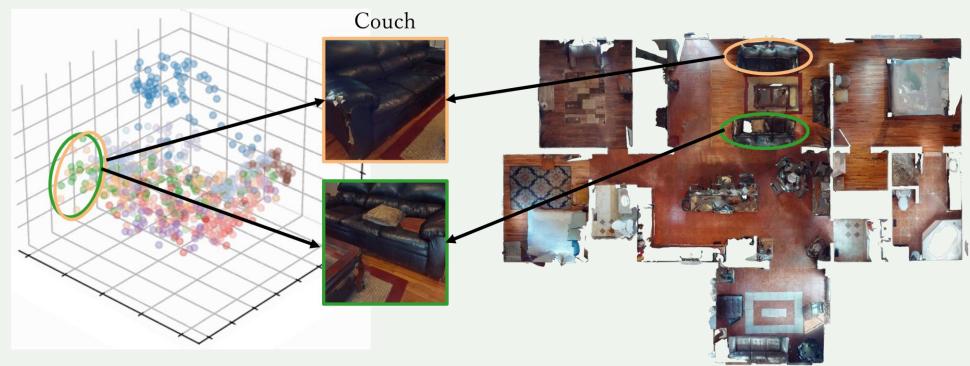
In such scenario, the robot needs to classify objects fine-grained

Given query image may be taken from a different viewpoint than when the robot observes the target object

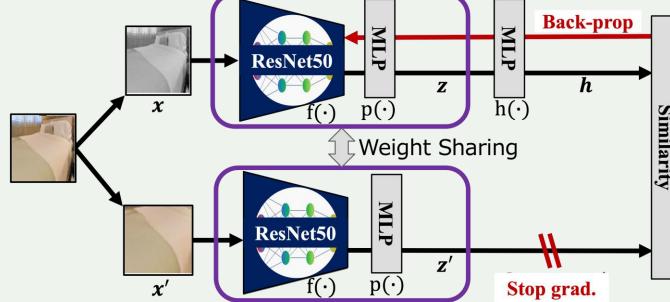
Challenge2

It is important to learn viewpoint invariance of instance images

CLIP_[2] attracts attention in object searching task_[3] However, is CLIP suitable for InstanceImageNav?



Unimodal contrastive method_[4] learn instance discriminative representation from images created by 2D data augmentation

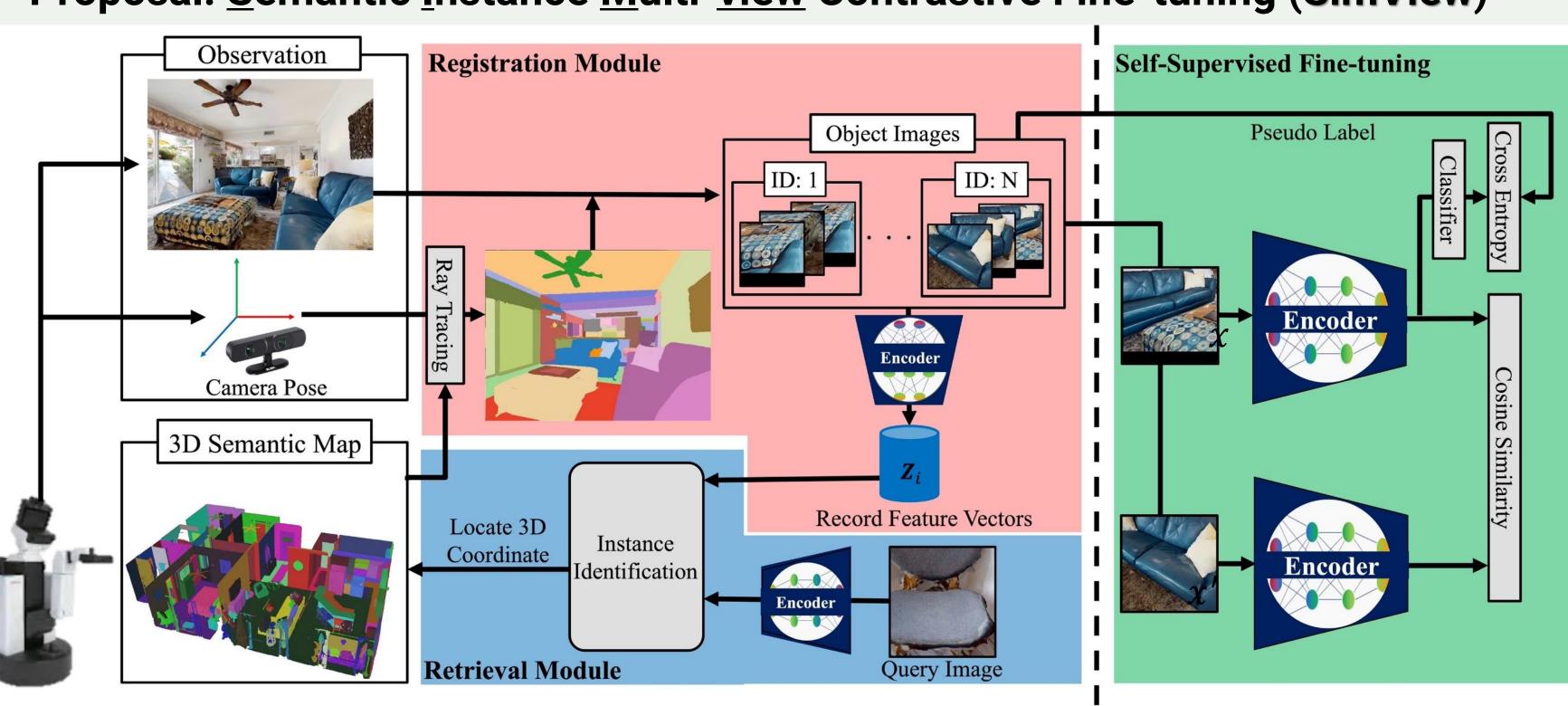


Dose conventional contrastive image pre-training methods learn the similarity of multi-view images of an instance?

Proposed Method

Proposal: <u>Semantic Instance Multi-view</u> Contrastive Fine-tuning (SimView)

- Learning instance discriminative representation by leveraging contrastive learning
- Learning viewpoint invariant instance representations utilizing contrastive learning with multi-view images
- x, x': Same instance images observed by robot from different view-points
- y, y': Instance ID of x, x' labeled by robot utilizing 3D Semantic Map $\hat{y}, \hat{y'}$: Instance ID of x, x' labeled by robot
 - utilizing 3D Semantic Map
- $L = \frac{1}{2} \{ \operatorname{CosSim}(f(x), h(x')) + \operatorname{CosSim}(f(x), h(x')) \}$ $+ \frac{1}{2} \{ CE(\hat{y}, y) + CE(\hat{y'}, y') \}$



Experiment Result

Purpose

Main Idea

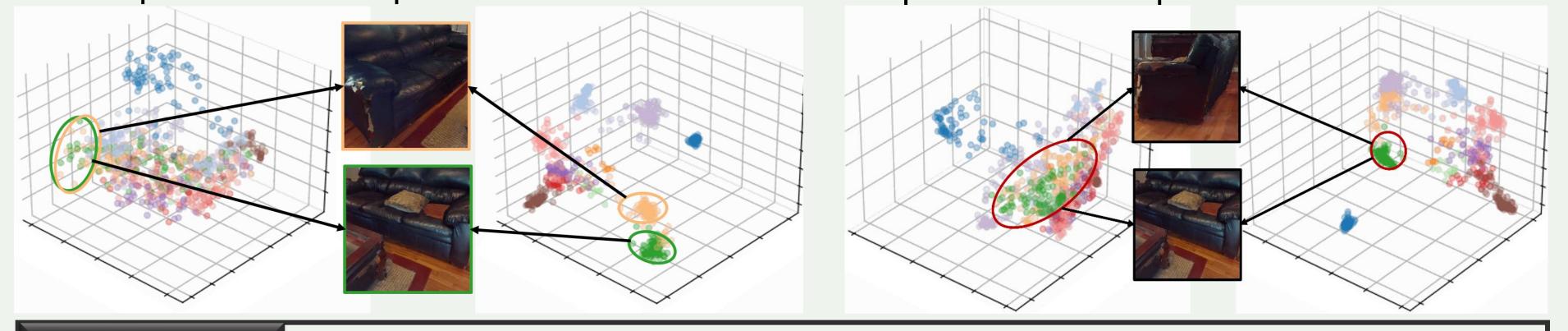
- Evaluating if contrastive learning of only image pairs is more suitable than CLIP for identifying the instance
- Evaluating if SimView learn viewpoint invariance of same instance images compared with prior contrastive methods

| Method | Arch. | Env. 1 | Env. 2 | Env. 3 | Env. 4 | Env. 5 | Env. 6 | Env. 7 | Env. 8 | Env. 9 | Avg. |
|--|----------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|
| SimView (ours) | ResNet50 | <u>0.79</u> | <u>0.67</u> | <u>0.68</u> | <u>0.91</u> | 0.68 | <u>0.45</u> | <u>0.72</u> | <u>0.74</u> | <u>0.8</u> | <u>0.72</u> |
| SimSiam | ResNet50 | <u>0.70</u> | <u>0.58</u> | <u>0.56</u> | 0.80 | 0.65 | 0.41 | 0.63 | 0.68 | <u>0.75</u> | 0.64 |
| DINOv2 | ViT-B/14 | 0.51 | 0.48 | 0.45 | 0.66 | <u>0.71</u> | 0.41 | 0.55 | 0.44 | 0.61 | 0.54 |
| SimCLR | ResNet50 | 0.65 | 0.56 | 0.52 | <u>0.83</u> | 0.64 | <u>0.45</u> | <u>0.66</u> | 0.62 | <u>0.75</u> | 0.63 |
| CLIP | ResNet50 | 0.51 | 0.36 | 0.35 | 0.51 | 0.48 | 0.44 | 0.42 | 0.44 | 0.59 | 0.46 |
| CLIP | ViT-B/16 | 0.56 | 0.38 | 0.37 | 0.47 | 0.48 | 0.37 | 0.38 | 0.35 | 0.48 | 0.43 |
| omparison Latent Space of CLIP and ours Comparison Latent Space of SimSiam and | | | | | | | | | | | nd ours |

- **Dataset :** 9 scenes which are included in Habitat Matterport 3D_[5]
- Task : Image retrieval
- Evaluate whether the robot can find instances
- that are identical to the query image.
- **Metrics :** mean Average Precision

 $mAP = \frac{1}{K} \sum_{k=1}^{K} AP_{k}$ K: Number of trials AP_{k}: Average Precision of k-th trial

- [1] Krantz, Jacob, et al. "Navigating to objects specified by images." *Proceedings of the IEEE/CVF* International Conference on Computer Vision. 2023.
- [2] Radford, Alec, et al. "Learning transferable visual models from natural language supervision." International conference on machine learning. PMLR, 2021.
- [3] Chen, Boyuan, et al. "Open-vocabulary queryable scene representations for real world planning." IEEE International conference on Robotics and Automation. 2023.
- [4] Chen, Xinlei, et al. "Exploring simple siamese representation learning." *Proceedings of the IEEE/CVF* conference on computer vision and pattern recognition. 2021.
- [5] Yadav, Karmesh, et al. "Habitat-matterport 3d semantics dataset." Proceedings of the IEEE/CVF conference on Computer Vision and Pattern Recognition. 2023.



Suggestion

. Unimodal contrastive learning methods better at identifying instances than CLIP 2. SimView learn more invariant feature representations than SimSiam