

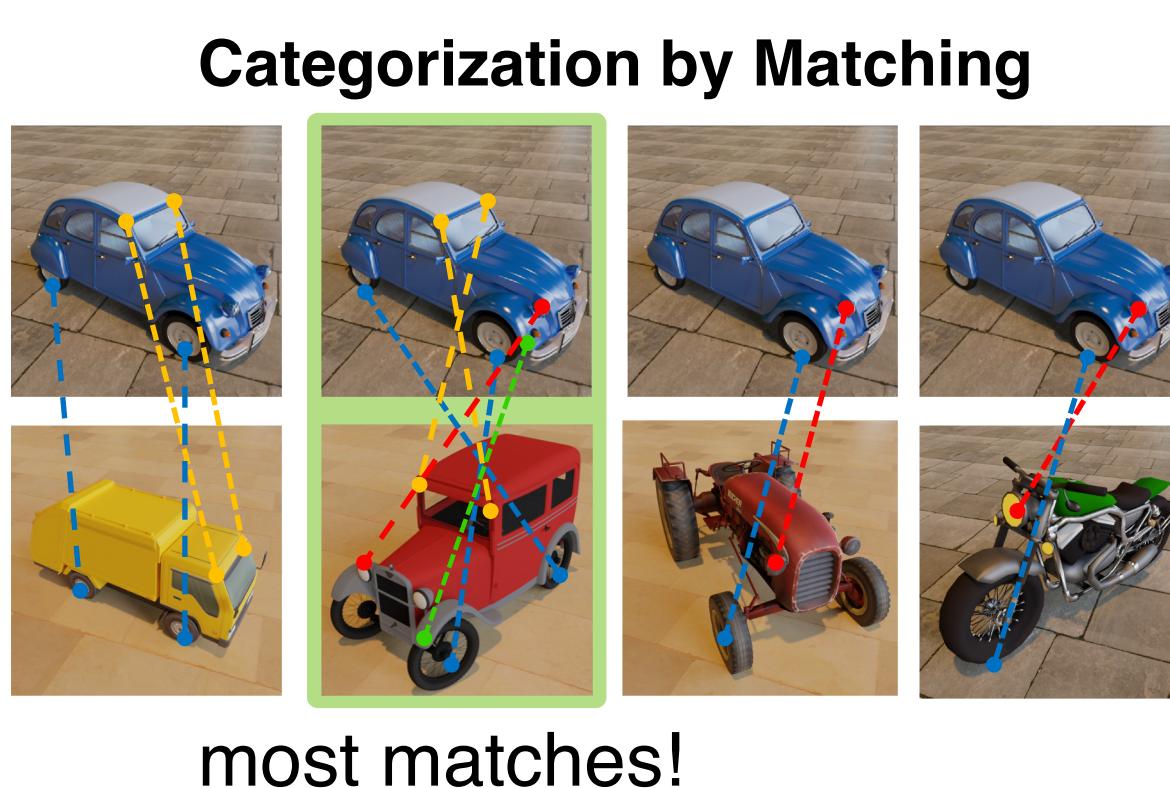
# Learning Dense Object Descriptors from Multiple Views for Low-shot Category Generalization

Stefan Stojanov, Anh Thai, Zixuan Huang, James M. Rehg

## Local Part Matching → Low-Shot Learning

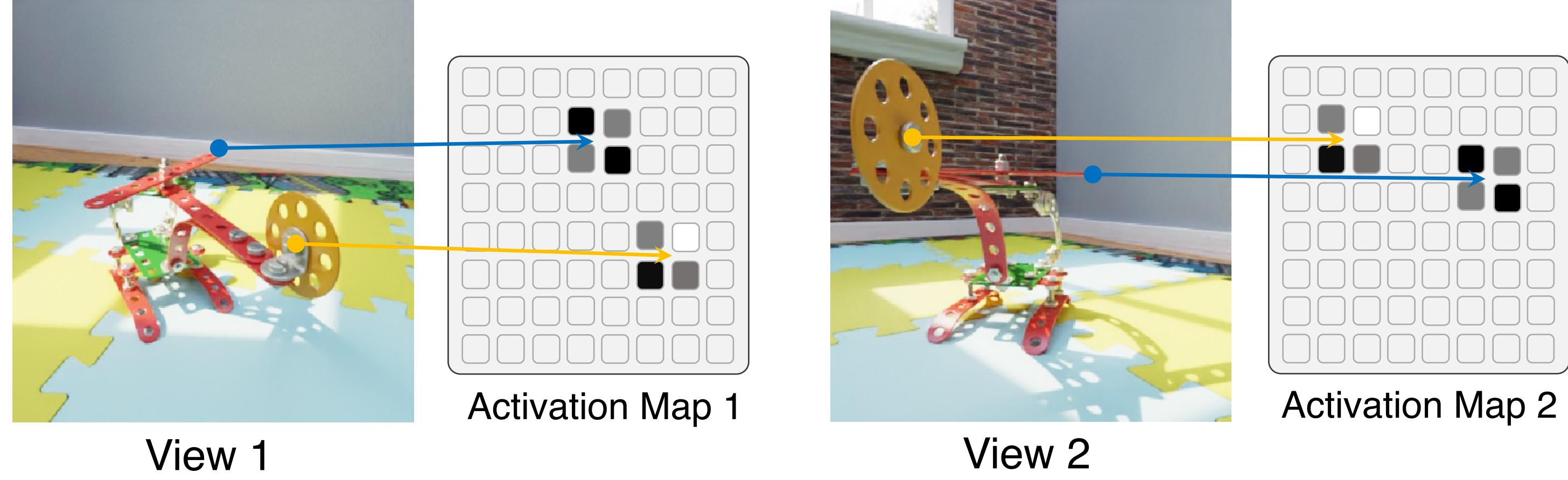
**Object Parts – Constituent Elements**

- Consistently occur across instances
- Removal significantly alters object
- Mutually distinct
- Defined by non-accidental properties
  - Do not depend on viewpoint or object pose



**Key Idea:** Learning to match local object parts across views → low-shot categorization by matching parts across instances

## Part Descriptor Properties: Across Views



**Non-accidental → invariant features**

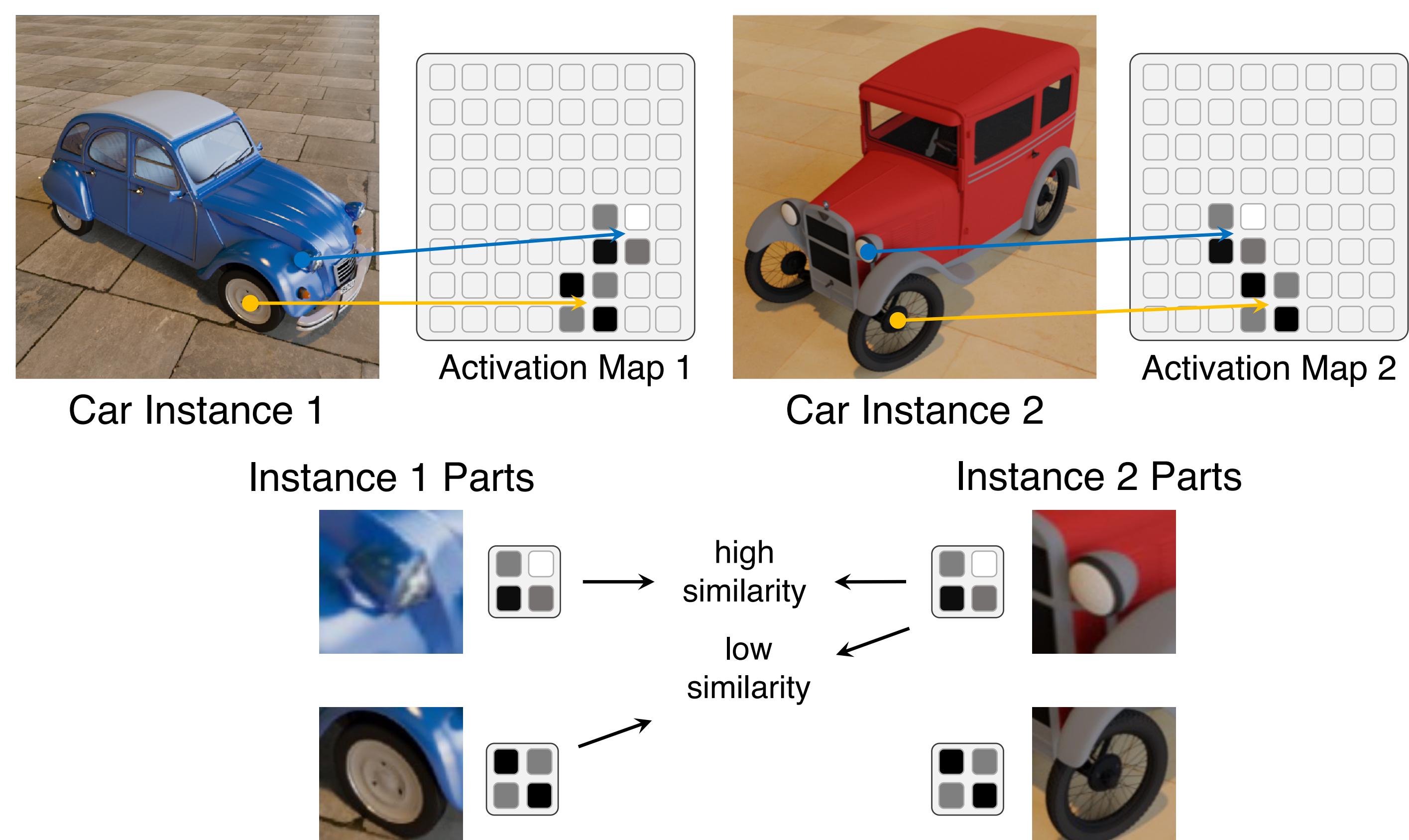
- Same features for whichever pixels the object part is projected to

**Mutually Distinct**

- Different parts encoded using different features

Desired properties can be learned from multiple views of objects!

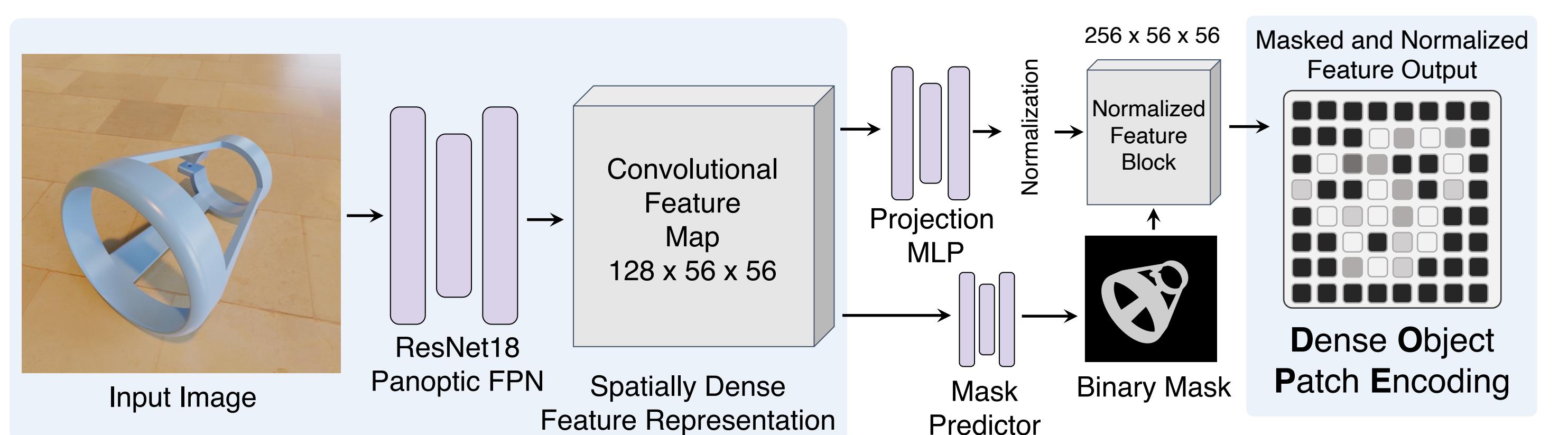
## Part Descriptor Properties: Across Instances



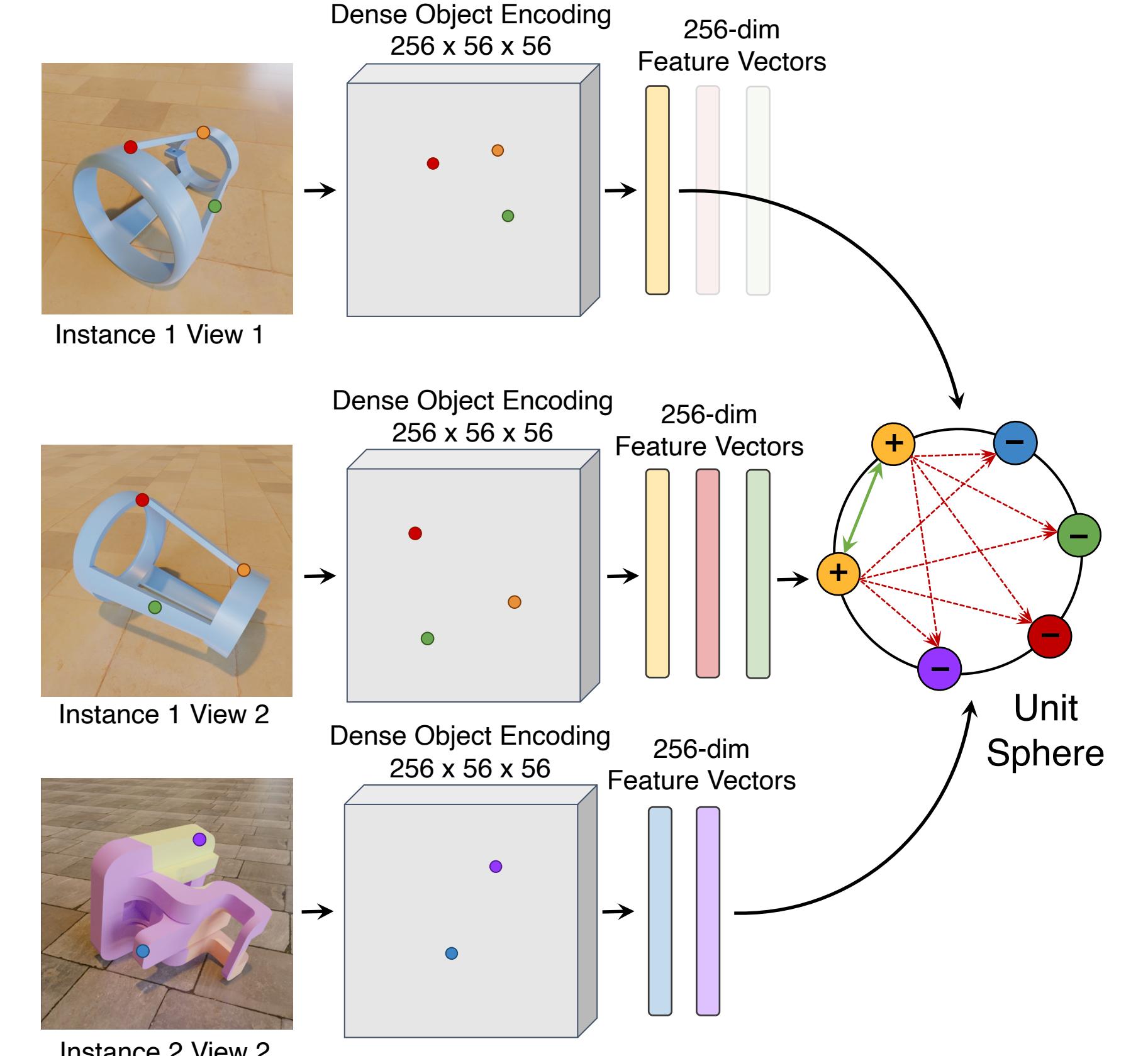
**Hypothesis:** Category-level descriptors can emerge by training across views of single instances

## Multi-view self-supervised learning allows for low-shot object category recognition

### Architecture

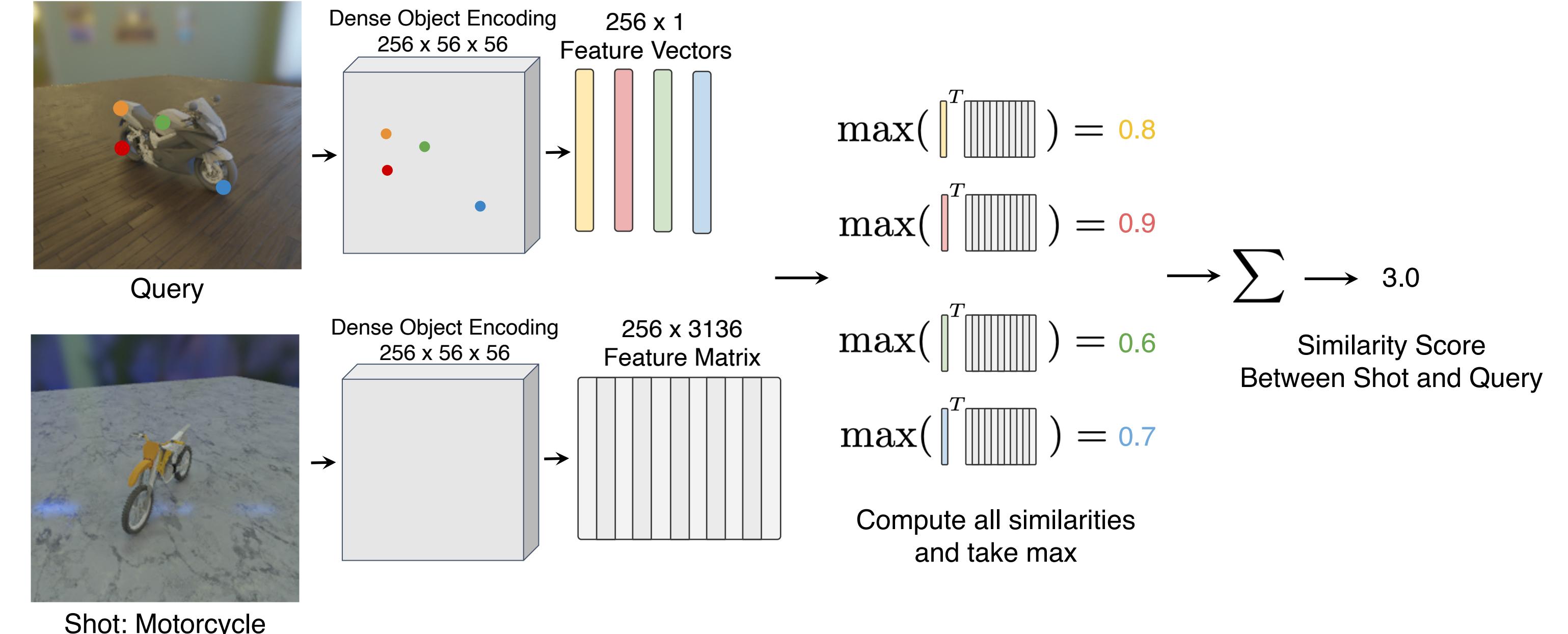


### Multi-View Self Supervised Learning

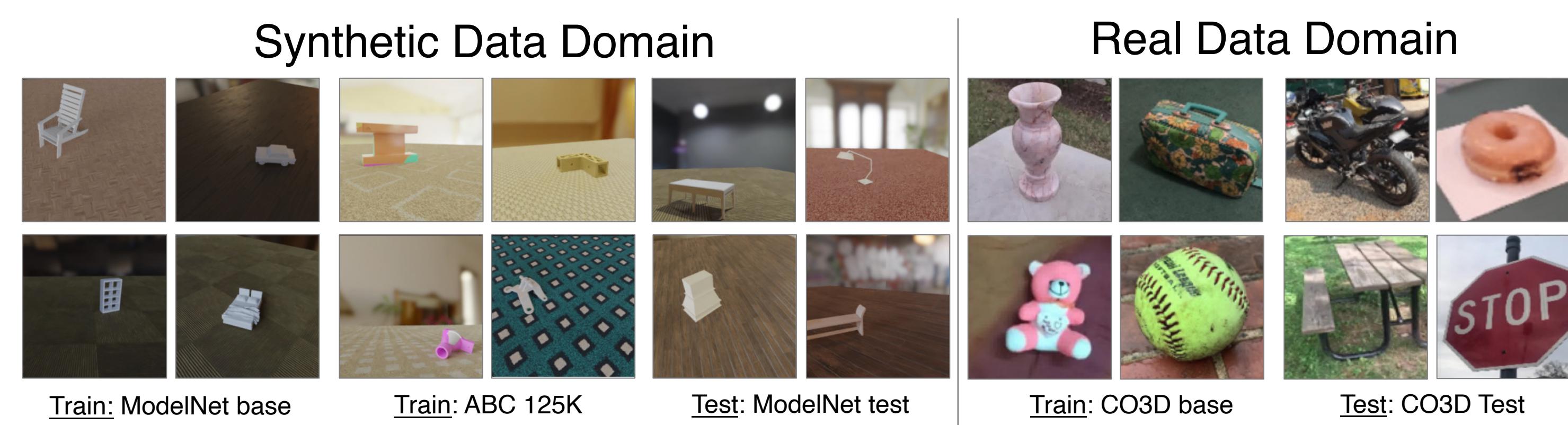


1. Sample points on the object surface using farthest point sampling
2. Find corresponding points in 2<sup>nd</sup> view using known viewpoint geometry
3. Extract local features based on point locations
4. Use corresponding positives and negatives for contrastive training using the InfoNCE loss

## Low-Shot Learning by Matching

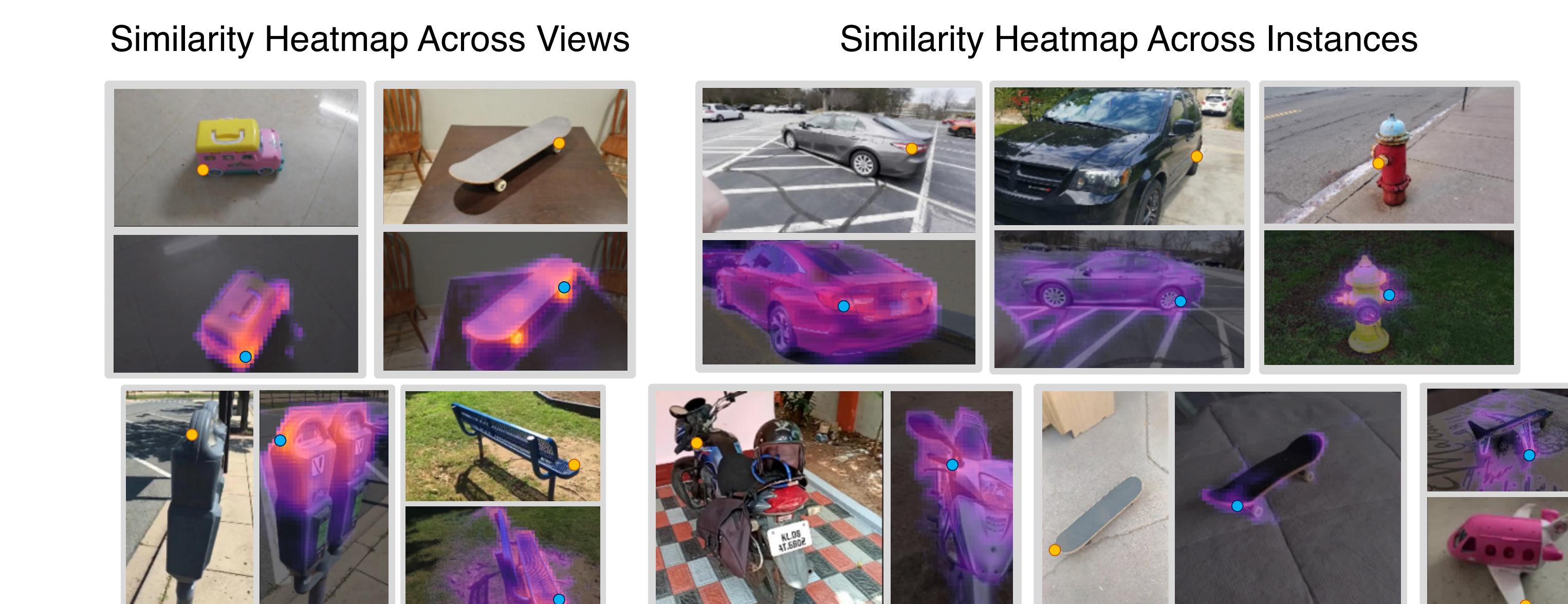
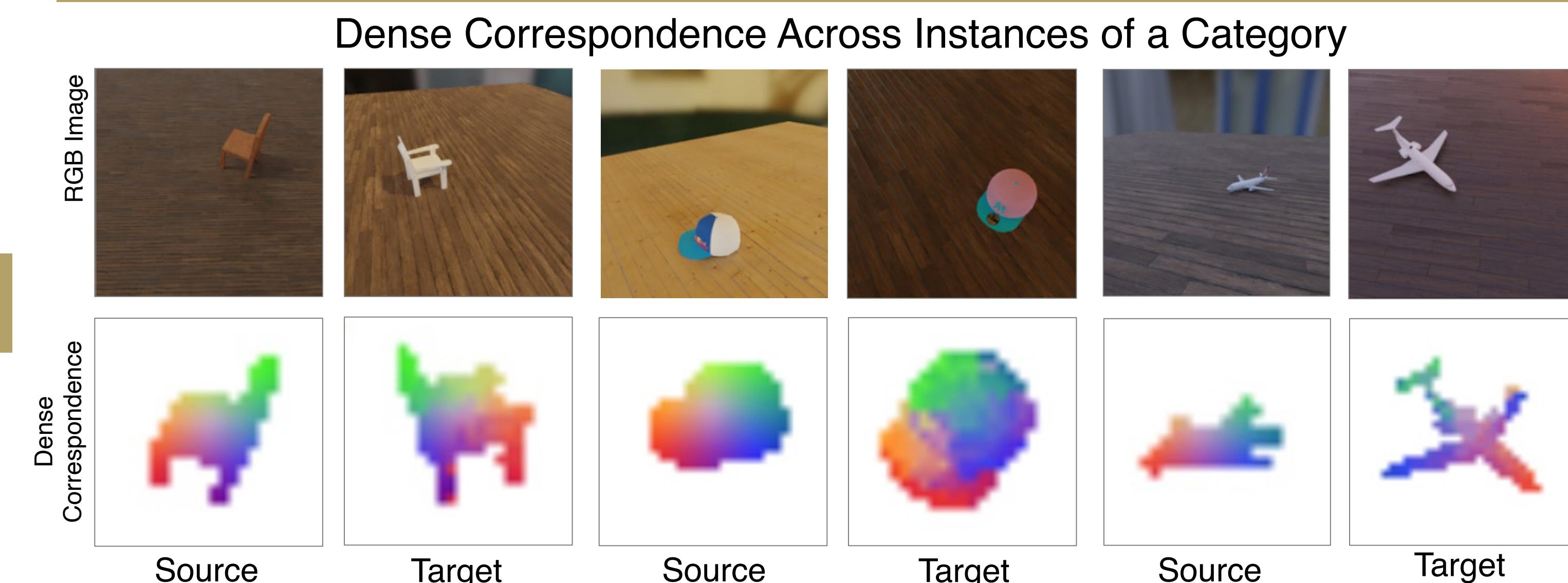


## Evaluation Setting



In the synthetic domain correspondences are obtained from known cameras and viewpoints, on real video data like CO3D they are obtained with COLMAP

## Qualitative Results



## Quantitative Results on ModelNet

	5-classes		10-classes	
	1-shot	5-shot	1-shot	5-shot
SimpleShot	56.55 ( $\pm 0.42$ )	69.87 ( $\pm 0.32$ )	41.27 ( $\pm 0.24$ )	54.84 ( $\pm 0.18$ )
RFS	57.31 ( $\pm 0.30$ )	73.77 ( $\pm 0.33$ )	42.22 ( $\pm 0.34$ )	59.97 ( $\pm 0.18$ )
FEAT	57.46 ( $\pm 0.39$ )	71.73 ( $\pm 0.32$ )	41.72 ( $\pm 0.24$ )	57.84 ( $\pm 0.18$ )
SupMoCo	55.32 ( $\pm 0.40$ )	71.82 ( $\pm 0.33$ )	39.87 ( $\pm 0.23$ )	57.15 ( $\pm 0.17$ )
VISPE	56.27 ( $\pm 0.44$ )	67.76 ( $\pm 0.35$ )	40.41 ( $\pm 0.26$ )	51.97 ( $\pm 0.18$ )
VISPE++ SimSiam	53.83 ( $\pm 0.29$ )	68.75 ( $\pm 0.25$ )	39.84 ( $\pm 0.17$ )	54.34 ( $\pm 0.13$ )
VISPE++ MoCoV2	57.05 ( $\pm 0.42$ )	71.81 ( $\pm 0.35$ )	43.23 ( $\pm 0.25$ )	58.68 ( $\pm 0.18$ )
DOPE (Ours)	57.51 ( $\pm 0.44$ )	70.44 ( $\pm 0.36$ )	42.73 ( $\pm 0.26$ )	55.52 ( $\pm 0.19$ )
VISPE++ SimSiam - ABC	60.24 ( $\pm 0.28$ )	<b>76.55 (<math>\pm 0.22</math>)</b>	47.02 ( $\pm 0.18$ )	<b>64.51 (<math>\pm 0.13</math>)</b>
VISPE++ MoCoV2 - ABC	61.07 ( $\pm 0.41$ )	75.96 ( $\pm 0.32$ )	47.67 ( $\pm 0.25$ )	63.27 ( $\pm 0.18$ )
DOPE (Ours)	<b>62.76 (<math>\pm 0.43</math>)</b>	<b>76.86 (<math>\pm 0.31</math>)</b>	<b>49.39 (<math>\pm 0.26</math>)</b>	<b>64.77 (<math>\pm 0.18</math>)</b>