



# Uncovering the Inner Workings of STEGO for Safe Unsupervised Semantic Segmentation

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Merantix Momentum

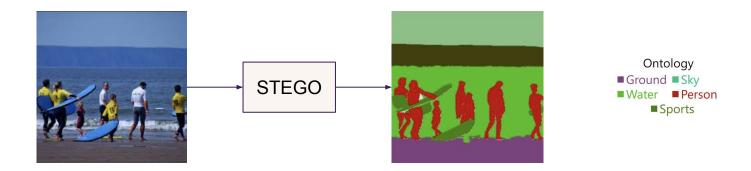
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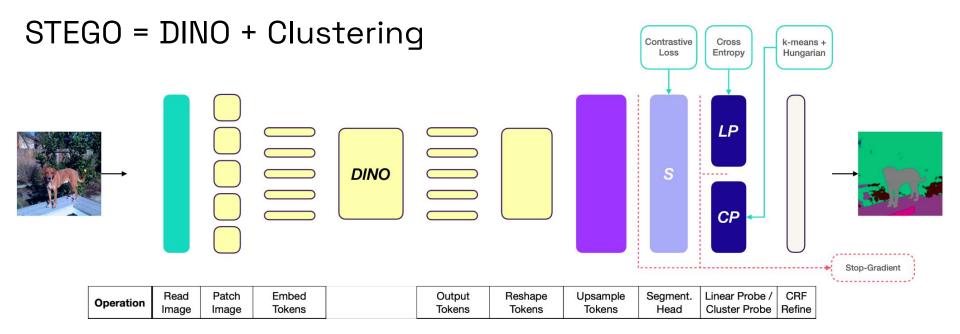
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#### Motivation

- Problem: labeled data is scarce, but unlabeled data is abundant
- Self-supervised learning recently demonstrated impressive results on unlabeled datasets
- STEGO (Hamilton et al., ICLR 2022) is an algorithm for unsupervised semantic segmentation
- To apply STEGO safely in real-world, it's crucial to understand its working mechanisms



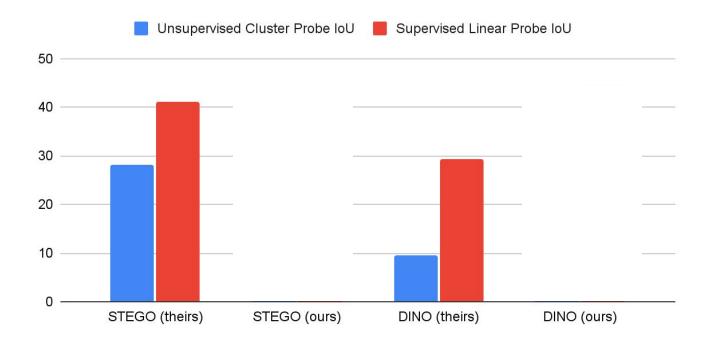


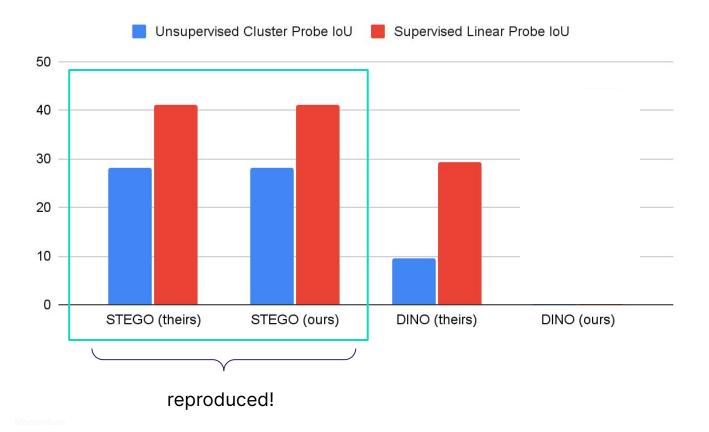
- STEGO builds on DINO (Caron et al., ICCV 2021) pre-trained Vision Transformer
- Segmentation head S projects DINO feats into lower-dimensional space, "distilling" DINO feature correspondences
- CP maps STEGO features to ontologies using k-Means

# Reproducibility Cocostuff

Cluster Probe = Linear Probe =

SegHead + K-Means + Hungarian SegHead + Lin. Layer + X Entropy

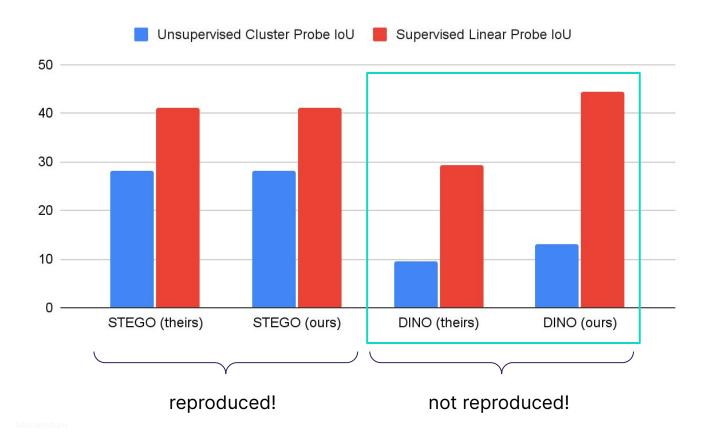




### Reproducibility Cocostuff

Unsupervised = Linear Probe =

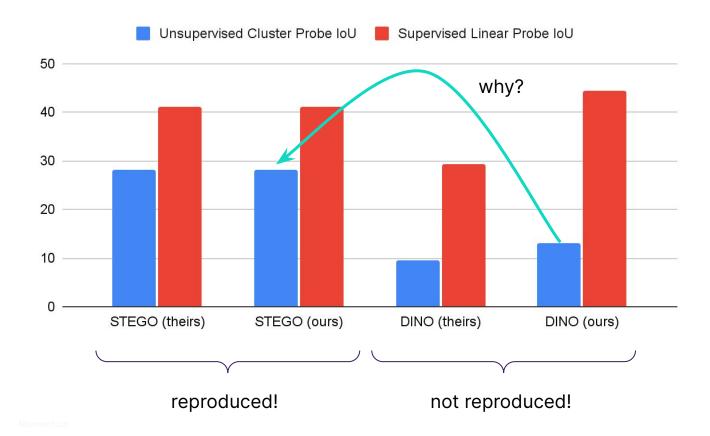
SegHead + K-Means + Hungarian SegHead + Lin. Layer + X Entropy



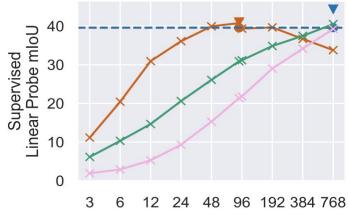
# Reproducibility Cocostuff

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#### STEGO's Working Mechanisms



# Embedding Dimension D

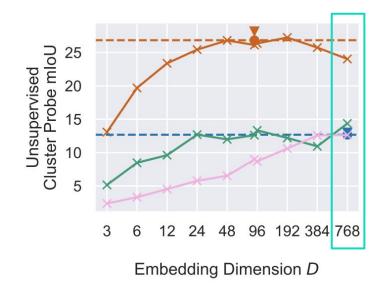
# ▼ STEGO (theirs) ▼ DINO (ours) → PCA -O- STEGO (theirs) w/o CRF → DINO (ours) w/o CRF → RP STEGO

#### Working Mechanism 1:

- STEGO is a dimensionality reduction technique
- k-Means converges better in fewer dimensions

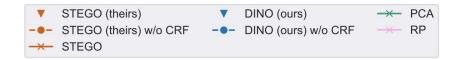


# STEGO's Working Mechanisms



#### Working Mechanism 2:

Segmentation head output forms more distinct clusters





#### Thank You

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