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Master Thesis Update

October 2025



Cartography
GIS
Remote Sensing

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1 SR Tests

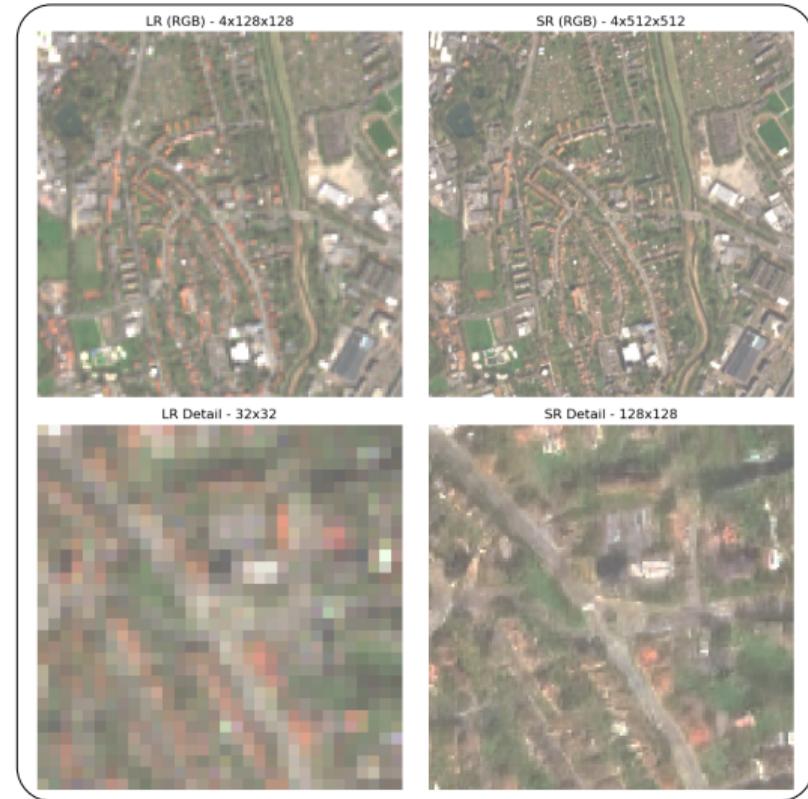
2 Dinov3 Features Experiments

3 SAM2

4 U-Nets

LDSR-S2

- Open Source
- Upscale: 10m -> 2.5m
- Pro: Tested Spectral fidelity
- Con: Low structural fidelity



S2DR3

- Upscale: 10m -> 1m
- Pro: Spectral fidelity is reliable but not perfect
- Con: not clear what is happening under the hood
- Con: Closed Source



Satlas-SR

- 2.5m resolution
- Pro: high structural quality
- Pro: Well-documented paper
- Con: Outdated dependencies
- Con: needs plumbing to make it work
- Con: Computationally expensive

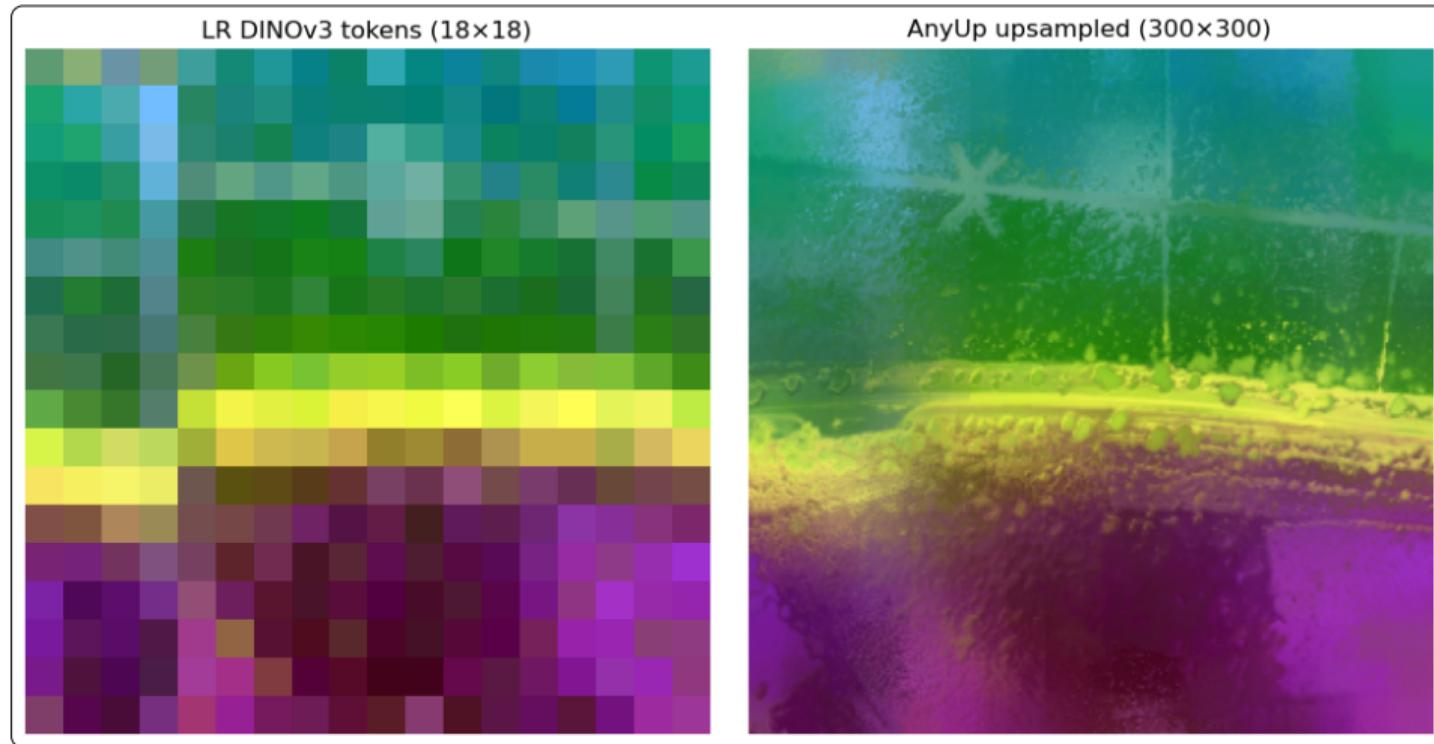
Dinov3-Models

- 10 models pretrained on web data (LVD-1689M dataset)
 - ▶ 1 ViT-7B trained from scratch
 - ▶ 5 ViT-{S, S+, B, L, H+} models distilled from the ViT-7B
 - ▶ 4 ConvNeXt-{T, S, B, L} models distilled from the ViT-7B
- 2 models pretrained on satellite data (SAT-493M dataset)
 - ▶ 1 ViT-7B trained from scratch
 - ▶ 1 ViT-L distilled from the ViT-7B

Image Example



Feature Extraction with PCA coloring

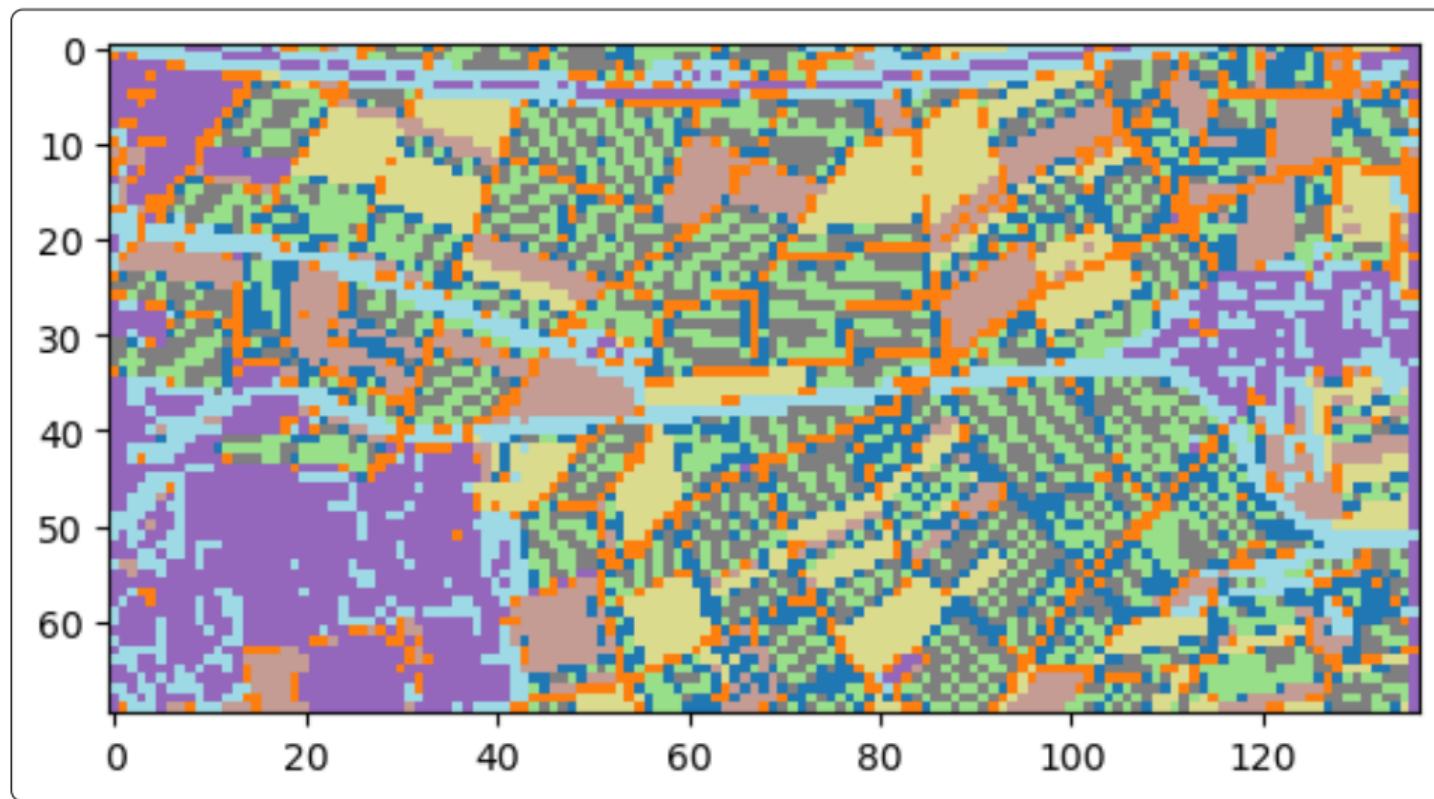


Example Image

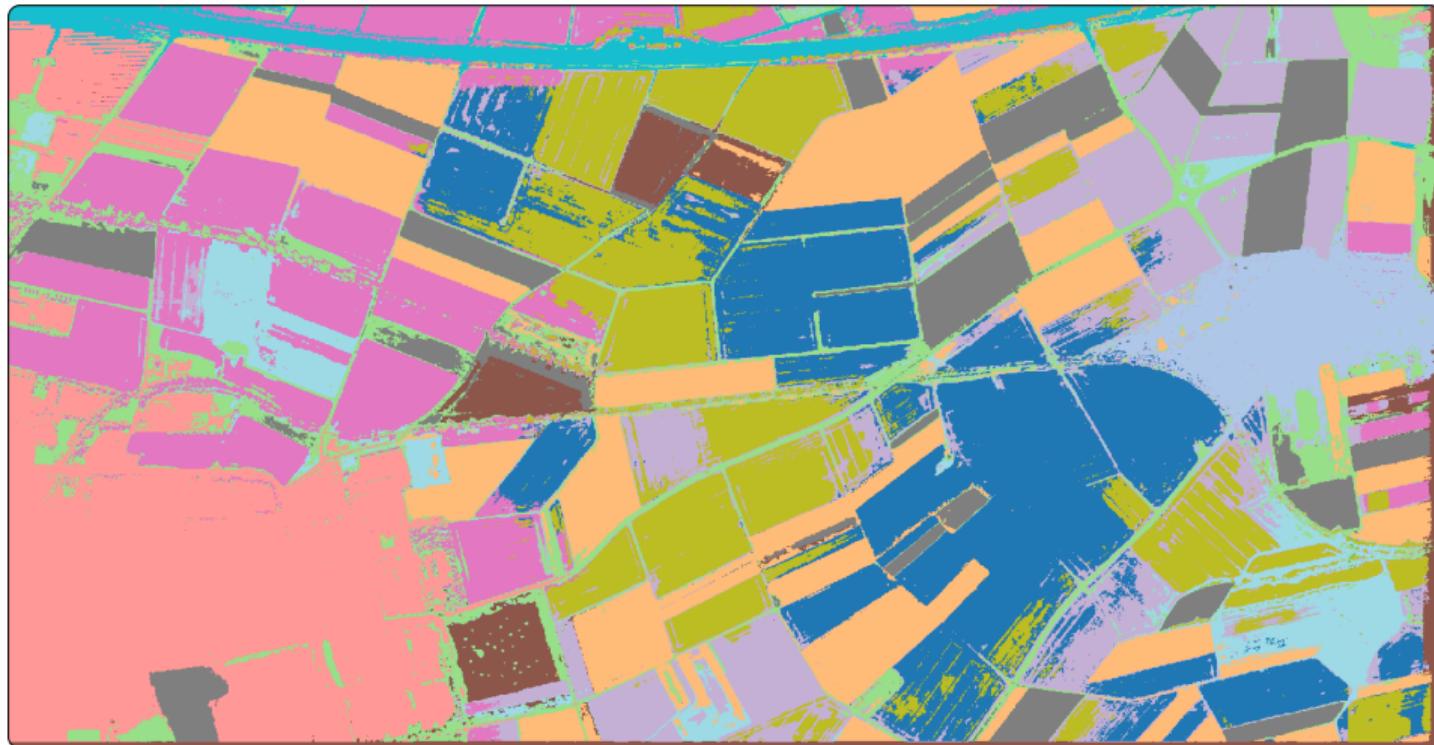
Normalized image (visual check)



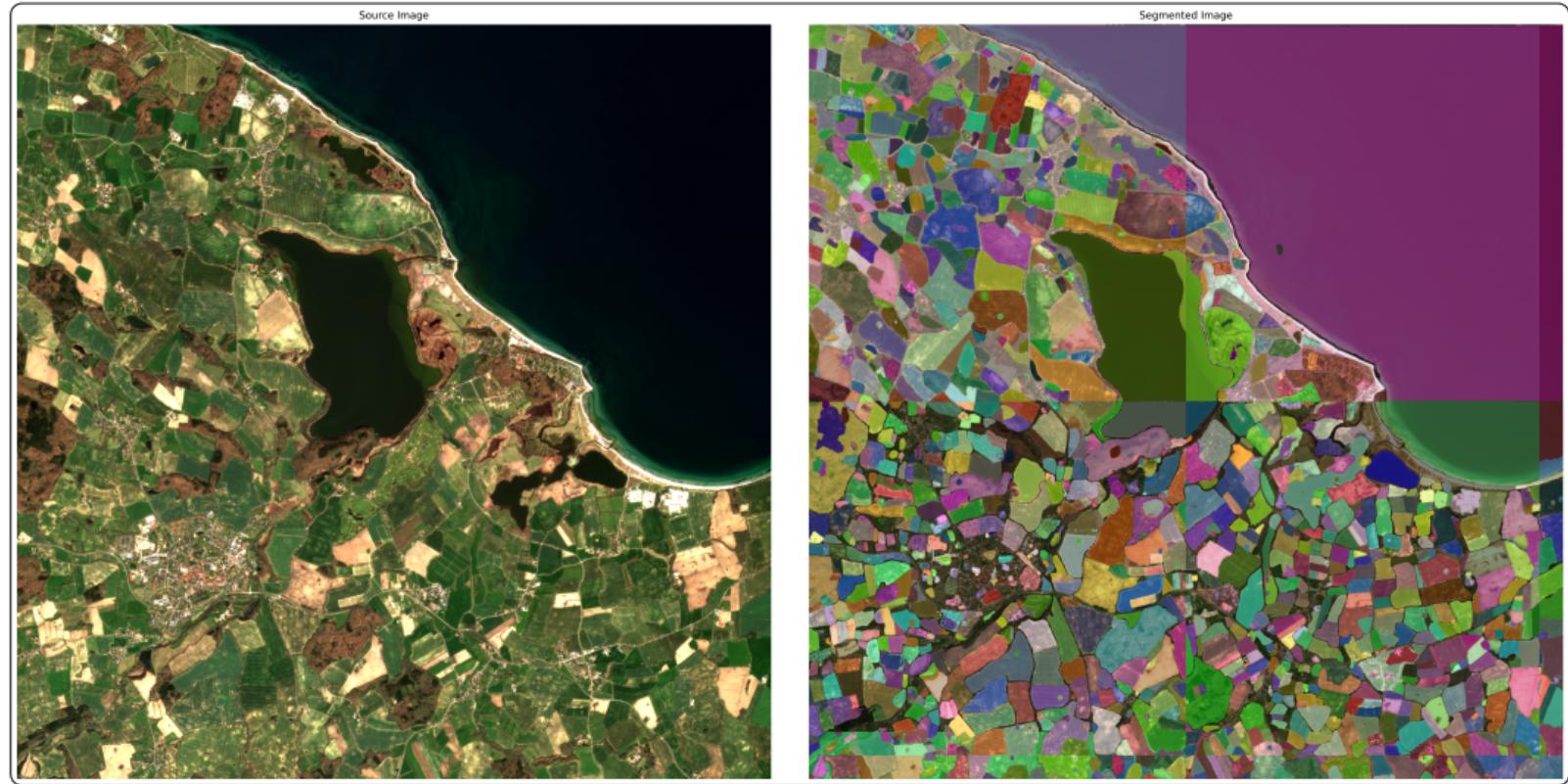
Feature Extraction with KNN



Feature Extraction with KNN Upscaled



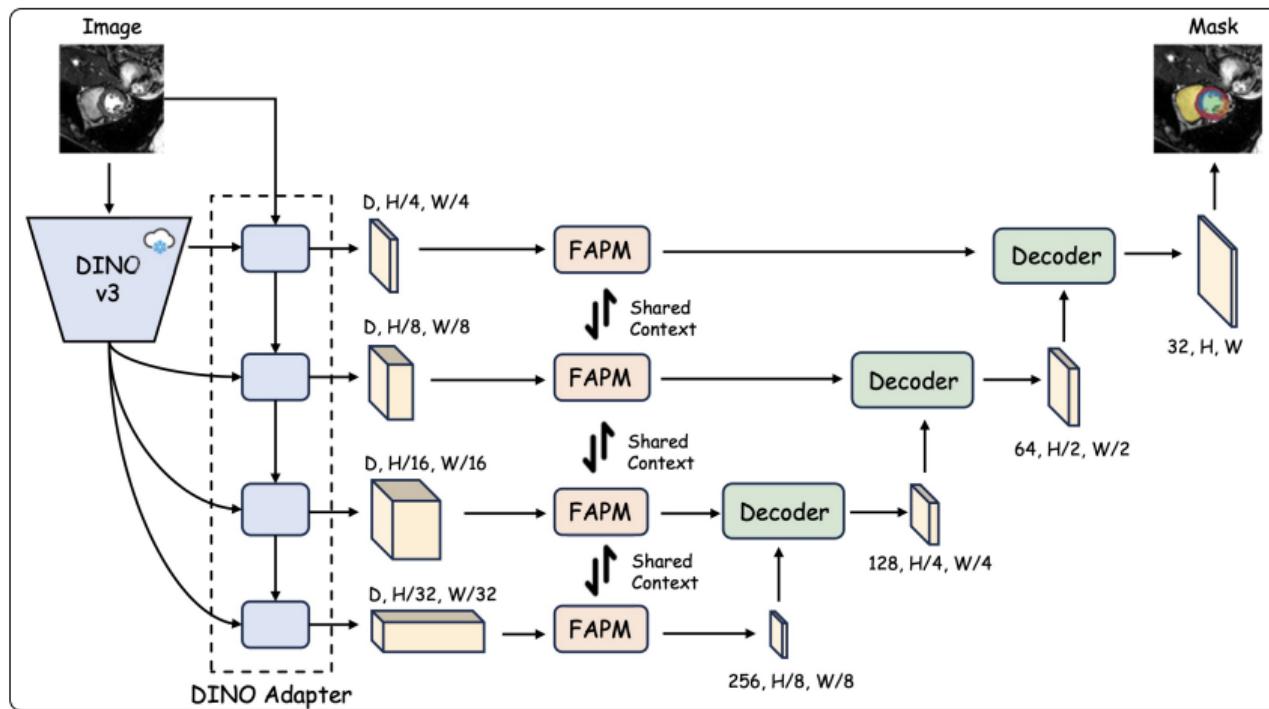
Sam2 Output



Sam2 Status

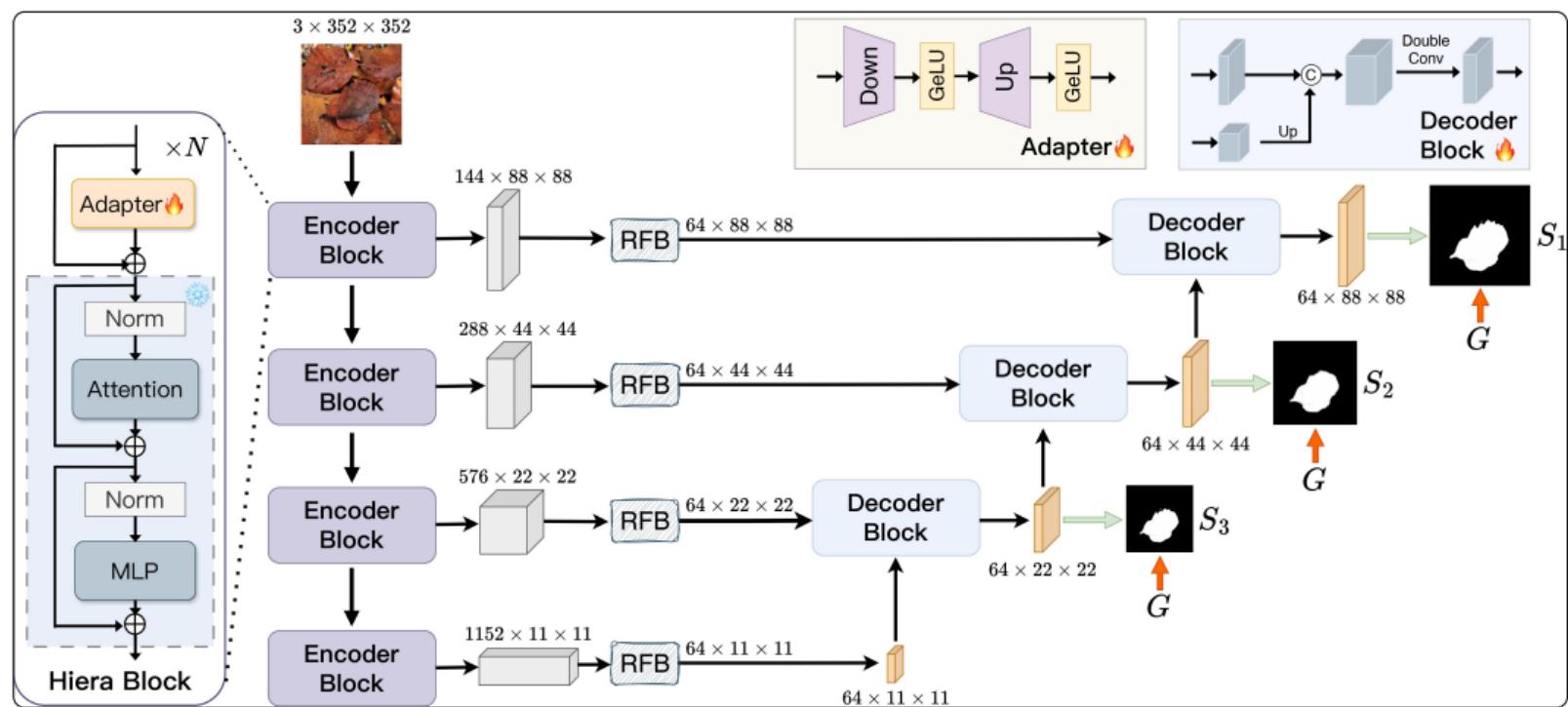
- Export SAM2-generated masks for all image tiles
- Build a classifier on top of SAM2 masks to distinguish linear woody features (LWF)
- Add an auxiliary mask covering areas not segmented by SAM2
- Validate classifier predictions against ground-truth LWF annotations

Dinov3 U-Net



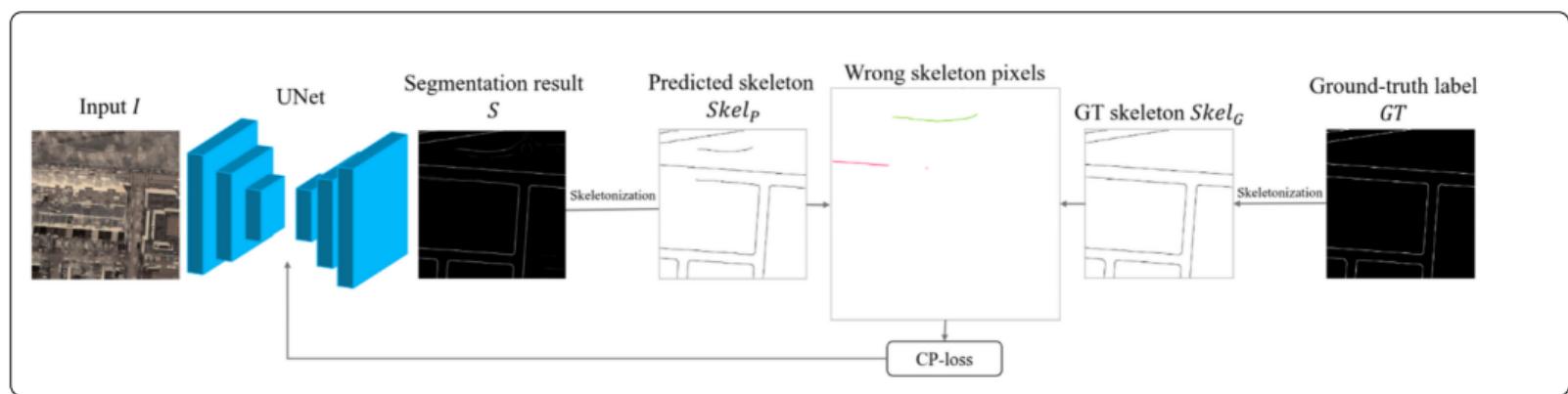
Gao et al., "Dino U-Net: Exploiting High-Fidelity Dense Features from Foundation Models for Medical Image Segmentation"

Sam2 U-Net



Xiong et al., "SAM2-UNet: Segment Anything 2 Makes Strong Encoder for Natural and Medical Image Segmentation"

CP-loss Architecture



Xu et al., “CP-Loss: Connectivity-Preserving Loss for Road Curb Detection in Autonomous Driving with Aerial Images”

CP-loss



(a) Ground-truth

(b) BCE

(c) Focal loss [13]

(d) Distance CE [29] (e) Balance CE [30]

(f) Dice loss [14]

(g) Ours

Xu et al., "CP-Loss: Connectivity-Preserving Loss for Road Curb Detection in Autonomous Driving with Aerial Images"

References

- Gao, Yifan et al. "Dino U-Net: Exploiting High-Fidelity Dense Features from Foundation Models for Medical Image Segmentation". In: *arXiv preprint arXiv:2508.20909*. 2025. DOI: [10.48550/arXiv.2508.20909](https://doi.org/10.48550/arXiv.2508.20909). URL: <https://arxiv.org/abs/2508.20909>.
- Xiong, Xinyu et al. "SAM2-UNet: Segment Anything 2 Makes Strong Encoder for Natural and Medical Image Segmentation". In: *arXiv preprint arXiv:2408.08909* (2024). Technical Report. URL: <https://arxiv.org/abs/2408.08909>.
- Xu, Zhenhua et al. "CP-Loss: Connectivity-Preserving Loss for Road Curb Detection in Autonomous Driving with Aerial Images". In: *arXiv preprint arXiv:2107.11920* (2021). Accepted at IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS) 2021. DOI: [10.48550/arXiv.2107.11920](https://doi.org/10.48550/arXiv.2107.11920). URL: <https://arxiv.org/abs/2107.11920>.