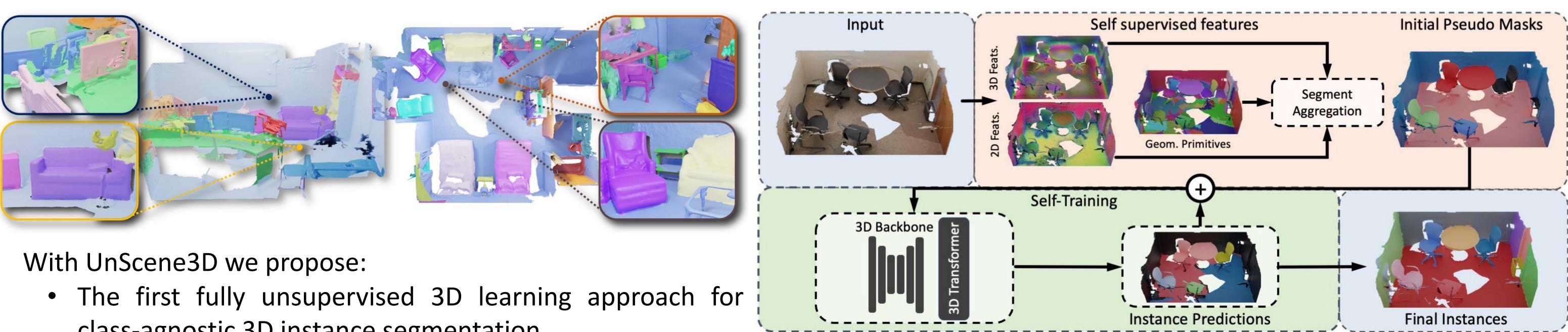
## **Unsupervised Instance Segmentation**

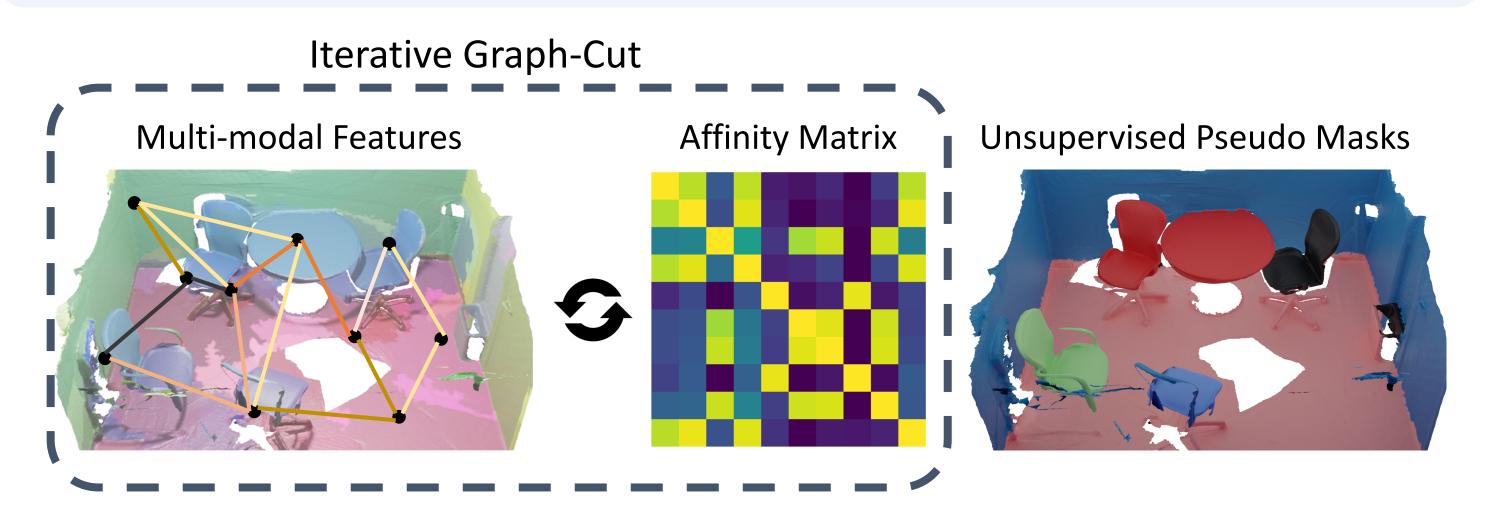
V

of Technology



- class-agnostic 3D instance segmentation.
- on high-resolution 3D data.
- We refine the coarse proposals through a series of selftraining iterations from the model's own predictions.

#### **Pseudo GT Generation**



We utilize self-supervised features from the 2D/3D domains, and aggregate these on a geometric oversegmentation of the scenes. This enables us to efficiently represent both low- and high-level properties of the scene. From here, we iteratively foreground-background partitions separate the Normalized Cut algorithm. For every iteration we mask out the already predicted nodes and continue the process until there are no more segments left in the scene.



Check out our website for more results, full paper & code!

# **UnScene3D: Unsupervised 3D Instance Segmentation for Indoor Scenes** David Rozenberszki, Or Litany, Angela Dai

# Self-training Approach

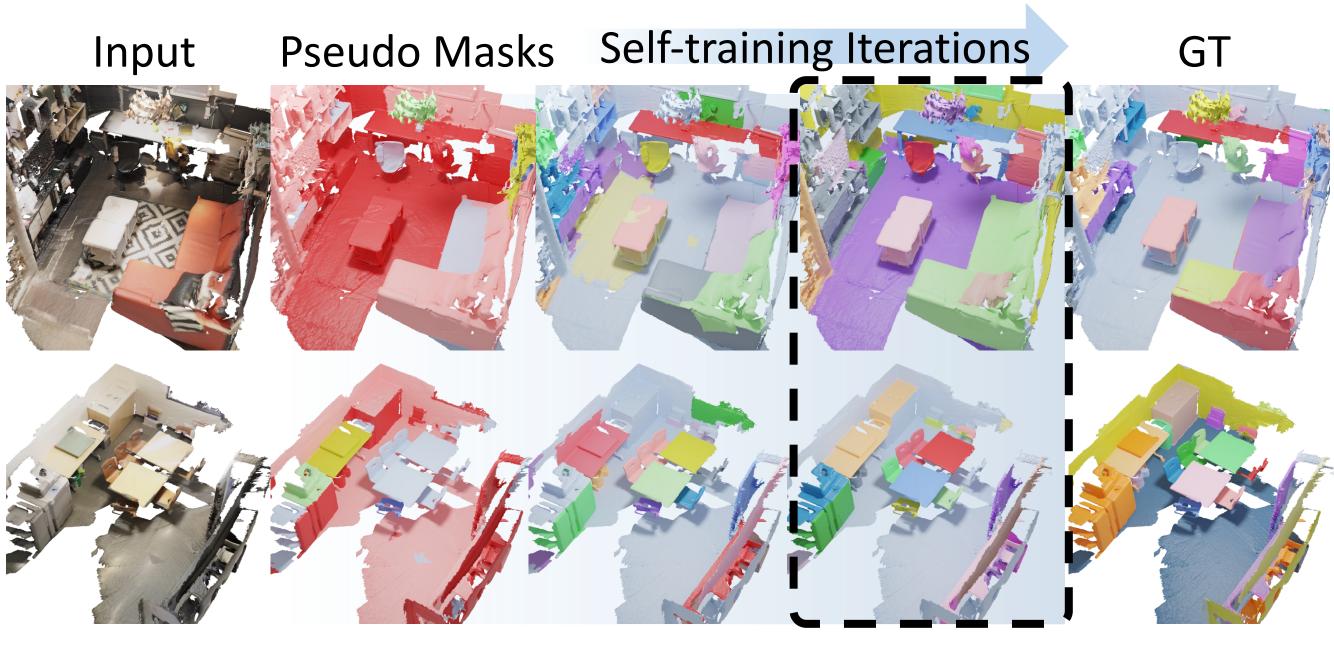
using

For pseudo ground truth we leverage self-supervised While pseudo masks provide a good set of initial proposals, we color/geometry features aggregated on geometric further refine them by training our model on its predictions. This selfsegments, enabling efficient representation and learning training process can effectively densify and clean the originally sparse pseudo masks, resulting in a more accurate instance segmentation.

#### **Unsupervised Instance Segmentation Results**

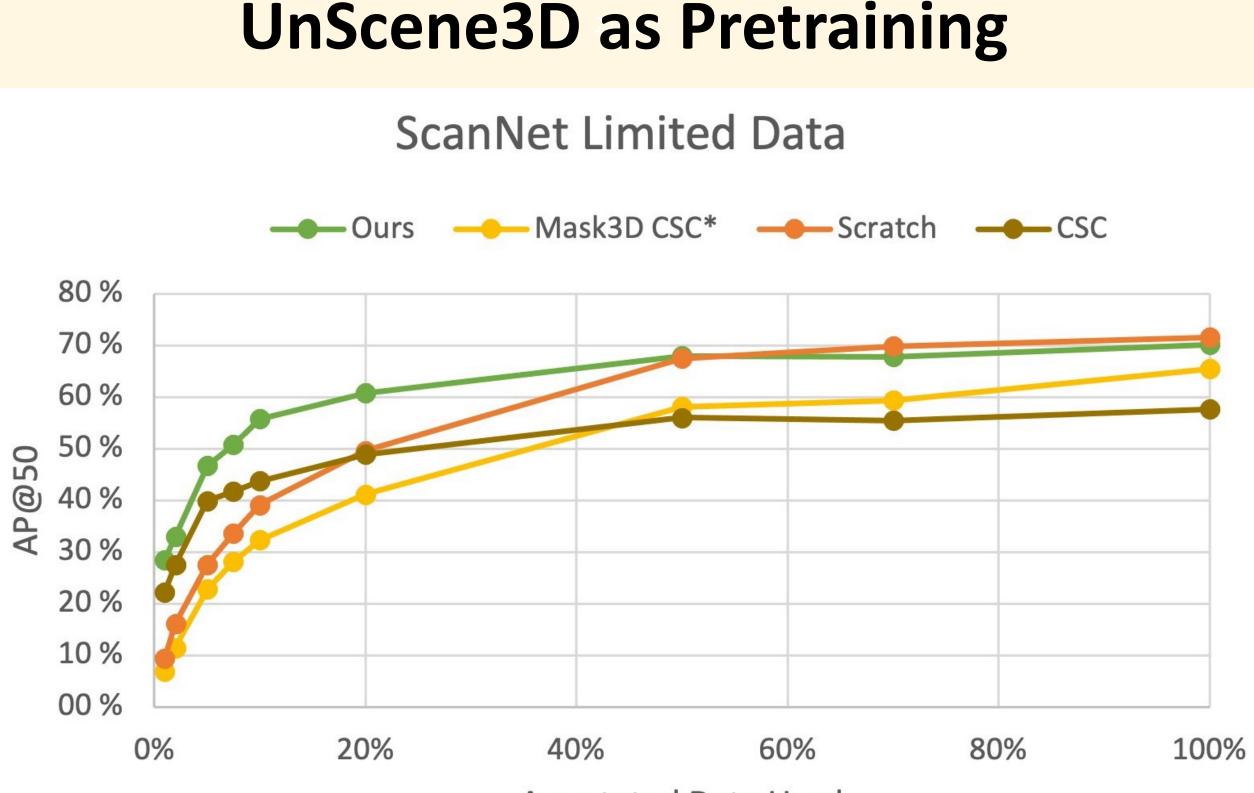
	Images	AP@25	AP@50		mAP
Nunes et al. [1]	X	30.5	7.3		2.3
CutLER (2D) [2]	$\checkmark$	7.0	0.2		0.3
SAM3D [3]	$\checkmark$	19.9	10		5.9
Ours	×	58.5	32.2		15.9
CutL Input (use 2		Mask3D [4] (full sup.)	Nunes et al. (2022)	Ours	GT

UnScene3D improves over unsupervised 3D instance segmentation methods by more than 300% Average Precision score, demonstrating effective instance segmentation in challenging, cluttered 3D scenes.



After every training iteration we generate a new set of predictions and combine them with the initial mask to create a new, enriched version of pseudo training masks. We show that the initial sparse set of masks get better with every self-training iteration.





UnScene3D can learn powerful class-agnostic object properties, which can be used for downstream tasks such as data-efficient dense 3D instance segmentation. Our approach improves over the previous state-of-the-art self-supervised pretraining methods for limited data scenarios.

[1] Nunes, Lucas, et al. "Unsupervised class-agnostic instance segmentation of 3d lidar data for autonomous vehicles." IEEE Robotics and Automation Letters 7.4 (2022): 8713-8720. [2] Wang, Xudong, et al. "Cut and learn for unsupervised object detection and instance segmentation." Proceedings of the IEEE/CVF conference on computer vision and pattern recognition. 2023. [3] Yang, Yunhan, et al. "Sam3d: Segment anything in 3d scenes." arXiv preprint arXiv:2306.03908 (2023). [4] Schult, Jonas, et al. "Mask3d: Mask transformer for 3d semantic instance segmentation." 2023 IEEE International Conference on Robotics and Automation (ICRA). IEEE, 2023.

### **The Effect of Self-Training**





Annotated Data Used