

Semantic Novel View Synthesis with 3D Gaussians

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Advanced Deep Learning for Computer Vision

End-to-end hybrid Gaussian- and voxel-based method with 2D to 3D segmentation lifting

Dense Gaussian Representation

Advantages

- Real-time rasterization on modern GPUs
- Compact form for geometric scene representation
- → Enables reconstruction with low hardware cost even for large scale scenes



- Multi-view RGB images
- Camera intrinsics and extrinsics
- ScanNet++ dataset

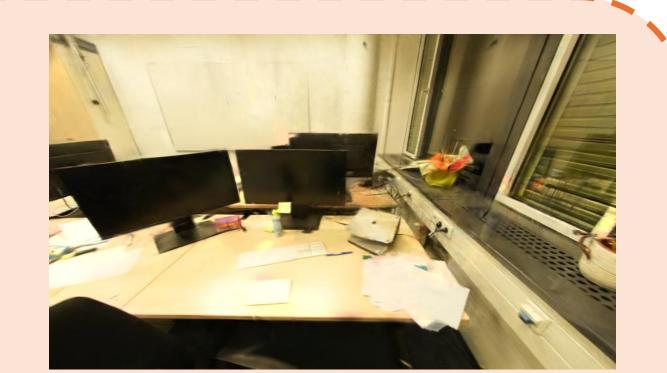
Sparse Voxel Representation

Advantages

- Compact grid structure for efficient data storage
- Significantly fewer voxels than Gaussians
 - → allows storing high-dimensional feature vectors
 - → enables efficient calculations

1. Geometric Scene Reconstruction





Method: 3D Gaussian Splatting

Output: scene reconstruction with ~1-2

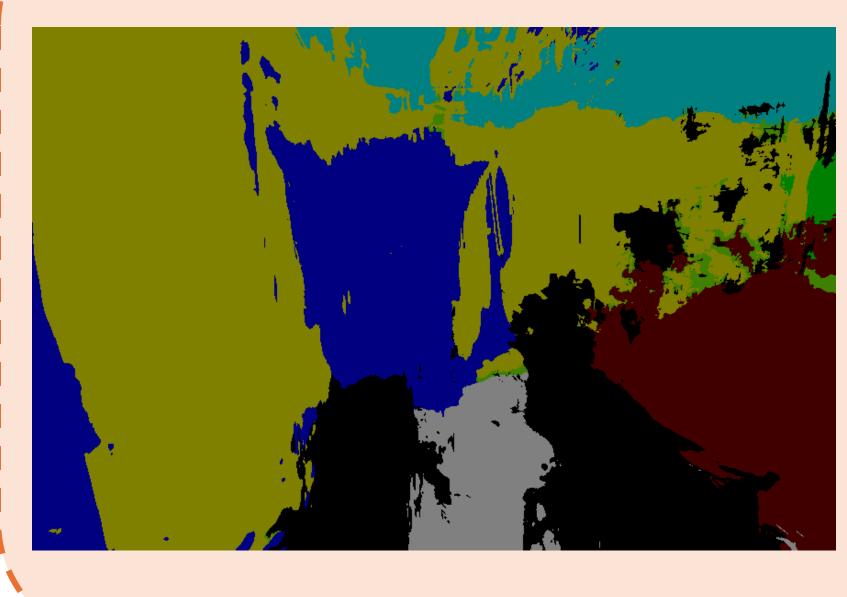
million Gaussians used as geometry prior!

2. Sparse Voxel Grid Initialization



Method: opacity, local density and scale based filtering of Gaussian centers
Output: ~10-100k initialized voxels around the geometry of the scene, saved as a sparse voxel grid

5. Semantic Rasterization



Method:

- Open vocabulary query of the text space embeddings stored in the voxels
- 2. Associating Gaussians with voxels and replacing RGB+SH with the label logits
- 3. Alpha blending over the logits, argmax to get per-pixel semantic label in 2D

Output: Gaussians with semantics logits, enabling semantic novel view synthesis

3. 2D Per-Pixel Semantic Map

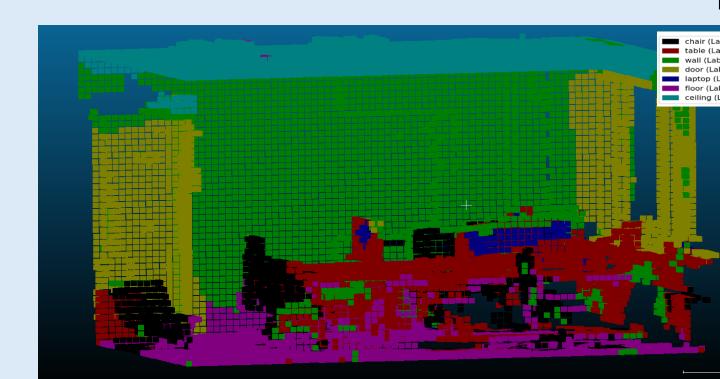


Method: 2D-semantic segmentation with LSeg

Output: 512-D raw per

pixel feature embeddings

4. Semantic Feature Map projection



Method: Lifting 2D segmentation results by projecting per-pixel feature embeddings from 2D to 3D sparse voxels using efficient ray tracing

Output: per-voxel semantic vectors

7. Evaluation: 2D semantic metrics

mloU	0.115
FloU	0.206

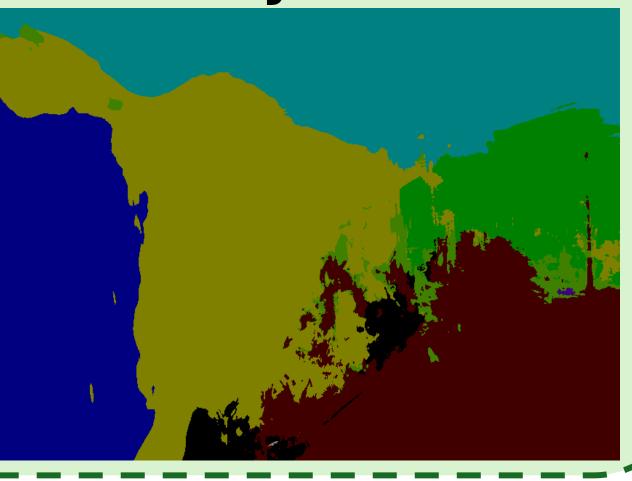
mIoU over handselected query inputs

Large structures (e.g. walls, ceilings) → decent segmentation

Small/fine objects → poor performance
Many classes near 0 mloU

6. Result: Semantic Novel View Synthesis





8. Future work:

- Improving 3DGS scene reconstruction: strongly limitates the segmentation performance
- Incorporating depth values of Gaussians at the rasterization step
- Optimization of the semantic logits stored in the Gaussians by supervision with ground truth 2D semantic maps
- More comprehensive evaluation