

# USC I Lab 3D: A Large-scale, Long-term, Semantically Annotated Outdoor Dataset

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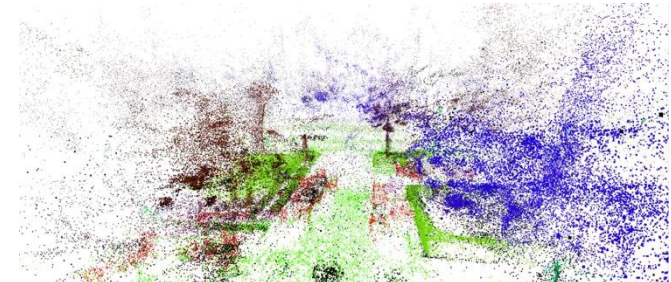
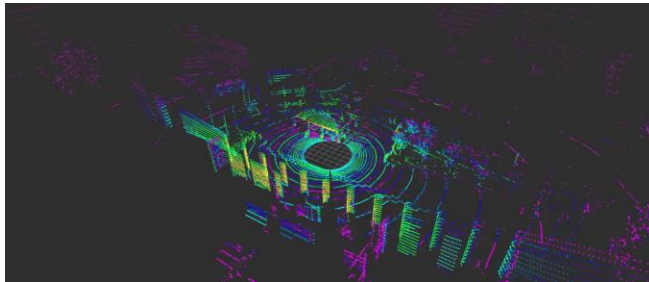
# Introduction

- Why do we need a new semantically annotated 3D dataset?

<b>Dataset</b>	<b>Frames</b>	<b>Indoor</b>	<b>Outdoor</b>	<b>LiDAR Point Cloud</b>	<b>Semantic</b>
LLFF[14]	< 1K images	✓	✓	✗	✗
DTU[11]	30K images	✓	✗	✗	✗
ScanNet[6]	2,500K images	✓	✗	✗	✗
Tanks and Temples[13]	147K images	✓	✓	✗	✗
ETH3D[21]	<1K images	✓	✗	✗	✗
Matterport3D[4]	195K images	✓	✗	✗	✓
Habitat[17]	-	✓	✗	✗	✓
iGibson[24]	-	✓	✗	✓	✓
SemanticKITTI[3]	23K scans	✗	✓	✓	✓
USCILab3d (ours)	10M images 1.4M scans	✗	✓	✓	✓

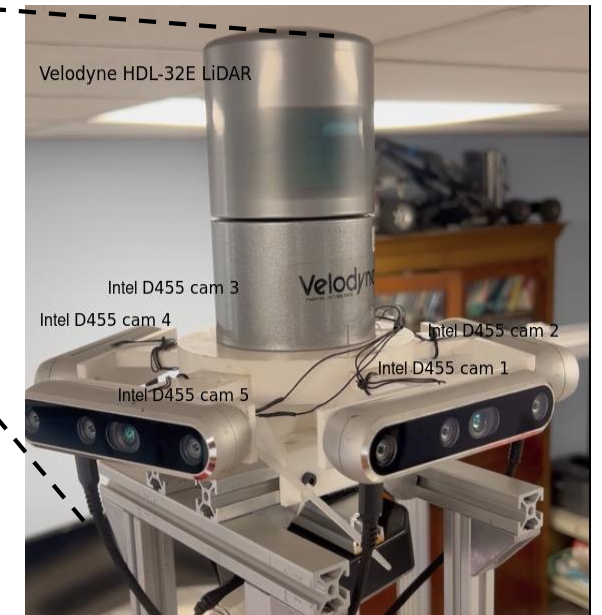
# Dataset features

- **Long-term**: Dataset collected over 12 months
- **Large-scale**: Data across 1.17 sq miles / 3.0 sq km.
- 10M 360 deg Multi-view RGB images
- 1.4M pointcloud scans
- Automated pipeline for semantic annotations of pointcloud



# Dataset collection

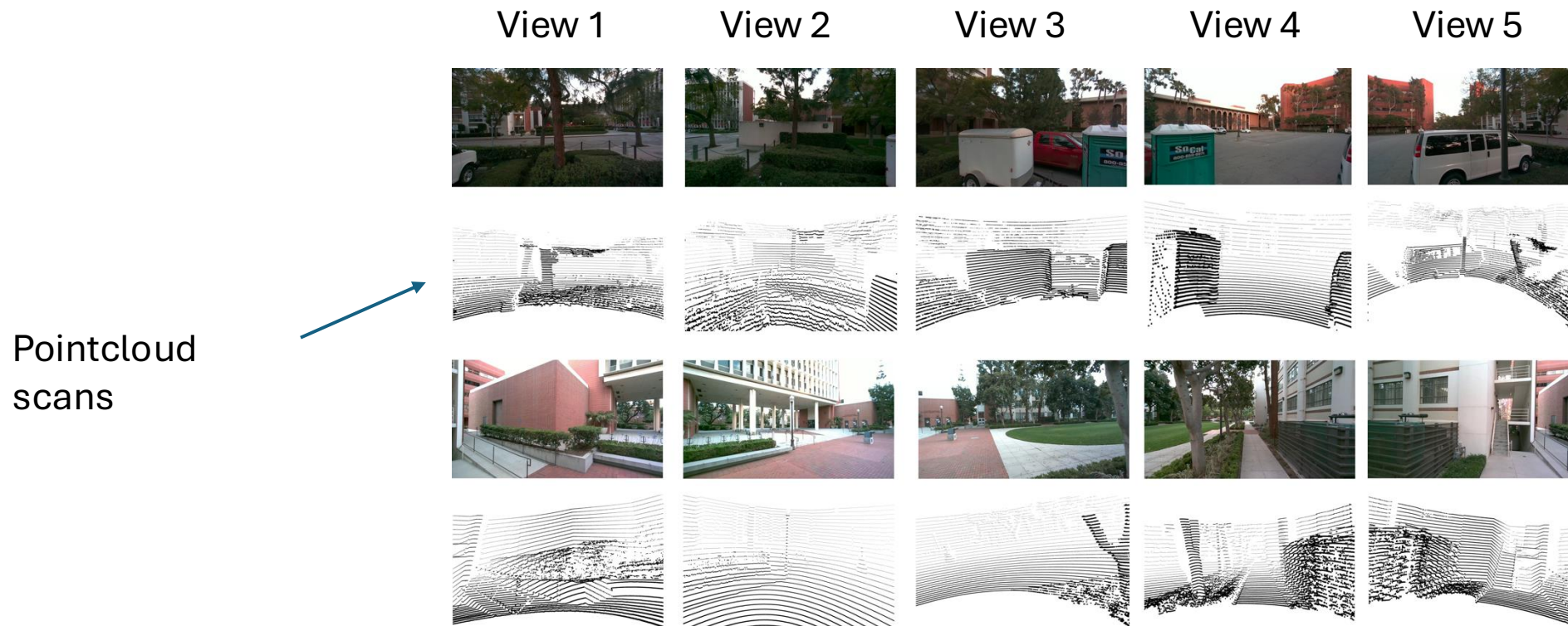
- Data collected in various terrains, lighting conditions





# Synchronized multimodal pairs

- Pairs of multi-view images and pointcloud scans

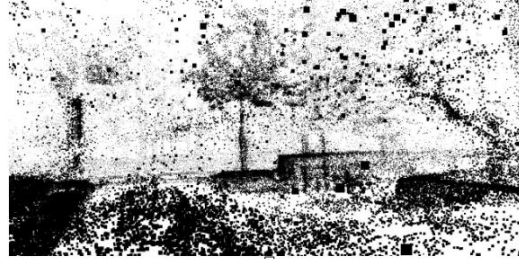


# Annotation pipeline

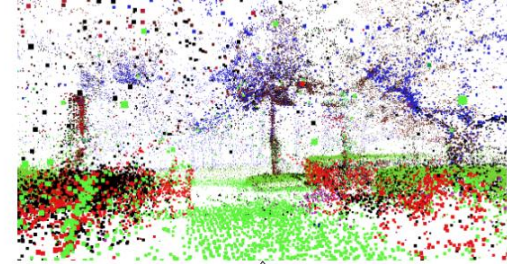
Input image



Raw pointcloud



Semantic pointcloud

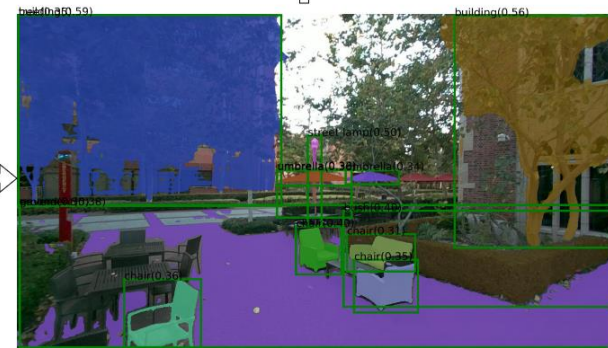


Align 2D pixels with 3D points

GPT4

background, building,  
street lamp, building,  
umbrella, pavement,  
chair, ground, bush,  
umbrella, chair, chair,  
tree, chair, umbrella

Grounded SAM

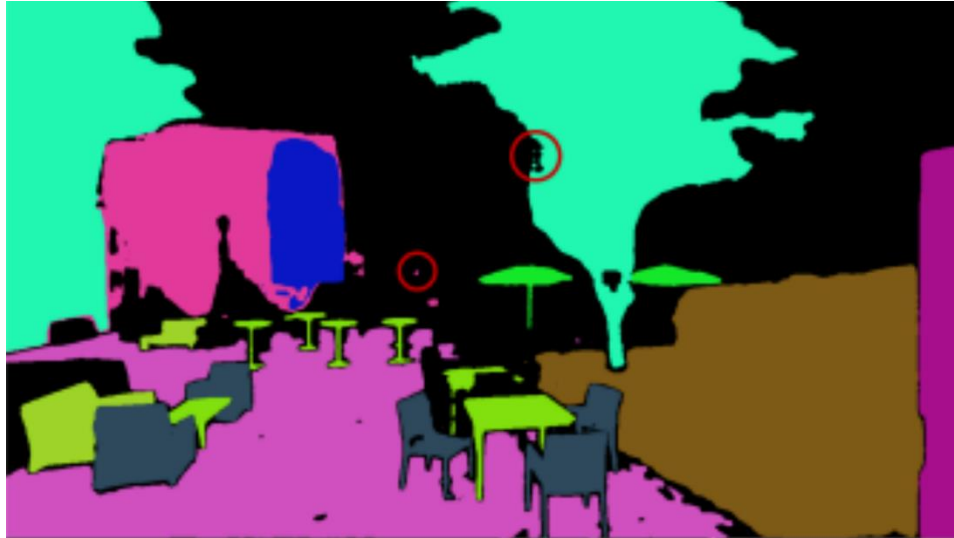


Semantic mask

Input prompt:  
“Generate  
semantic labels..”

# Robustness

Only 0.92% pixel error with 10 additional candidate labels!



Prompt: Trees, Bushes, Benches, Tables, Chairs, Pavement, Buildings, Windows, Doors, Emergency Call Box, Umbrellas, Leaves, Grass



Prompt: Trees, Bushes, Benches, Tables, Chairs, Pavement, Buildings, Windows, Doors, Emergency Call Box, Umbrellas, Leaves, Grass, Fire Hydrant, Person, Car, Parking Lot Lines, Boat, Scooter, Dog, Bear, Cat

Additional  
spurious labels



# Conclusion

- Following are our contributions:
  - Dataset consisting of multi-view images and pointcloud scans
  - Automated semantic annotations using foundation models and VLM
    - Postprocessing methods to avoid and remove outliers
- USCILab3D can be used for training/evaluating:
  - Models related to 3D perception
  - 3D generative models
  - NeRF/Gaussian splatting
  - 3D-LLMs



Thank you!

**Wed, 11 Dec**  
**1PM-4PM CST**