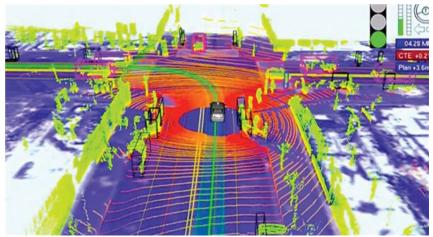
3D SEMANTIC SEGMENTATION IN DENSE URBAN ENVIRONMENTS



Motivation of the project

- Complex and highly important task in various fields, such as computer vision, robotics and virtual reality.
- Better spatial understanding.
- Lidar sensors brings real distance data. On the other hand, with cameras you could only get estimates.
- Independence from ambient lighting.
- Robustness against adverse environmental conditions.





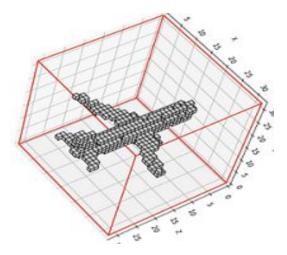
3D data representation

Raw



unordered points scattered

Volumetric

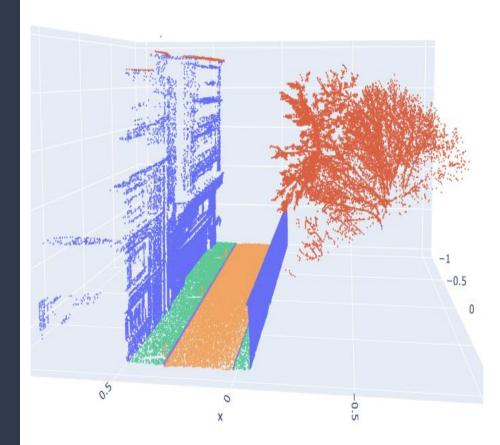


structured points in voxels

Our proposal

PointCloud

- Extends torch.Tensor
- DataFrame behaviour
- Get separate info (coords, feats, labels)
- Random sampling
- Shuffle of points
- Rotation (axis 'z')
- Plotting 3D



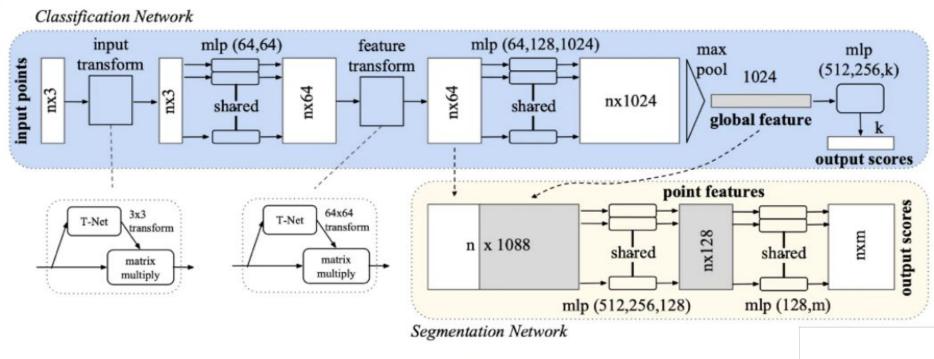
PointNet

INVARIANT TO PERMUTATIONS OF POINTS CAPTURE LOCAL STRUCTURES FROM NEARBY POINTS INVARIANCE UNDER TRANSFORMATIONS EUCLIDEAN SPACE SEMANTIC SPACE RELATIONSHIPS BETWEEN A POINT AND ITS NEIGHBORS CAPTURE GEOMETRY OF THE OBJECTS

Dynamic Graph CNN

- ✓ INVARIANT TO PERMUTATIONS OF POINTS
- ✓ CAPTURE LOCAL STRUCTURES FROM NEARBY POINTS
- ✓ INVARIANCE UNDER TRANSFORMATIONS
- ✓ EUCLIDEAN SPACE
- ✓ SEMANTIC SPACE
- ▼ RELATIONSHIPS BETWEEN A POINT AND ITS NEIGHBORS
- ✓ CAPTURE GEOMETRY OF THE OBJECTS

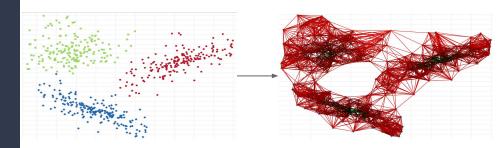
Point Net

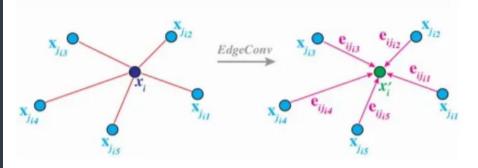


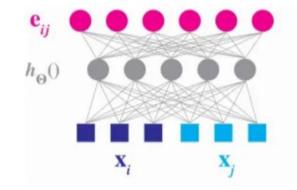
PointNet architecture

Dynamic Graph CNN

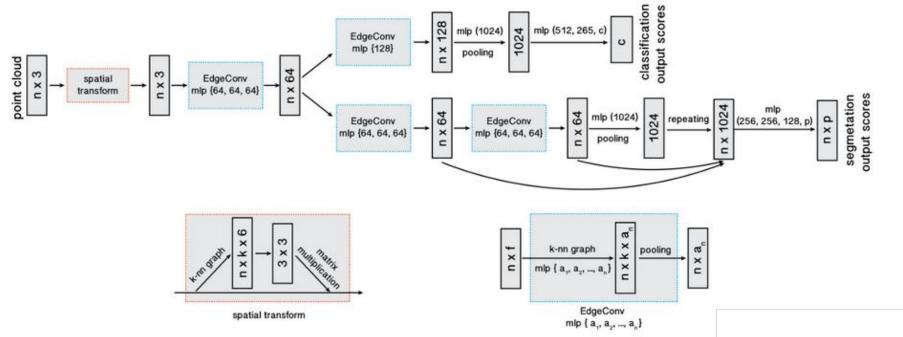
- KNN graph
- Edge Conv
- Embeddings
- Grouping points both in Euclidean space and in semantic space.







Dynamic Graph CNN



Computational Requirements

- Google colab
 - Concat trainings using state_dicts
- Google Cloud
 - Only 8 CPUs were used
- Local PC
 - GPU 3070, 16 GB RAM







Milestones

STATE OF **DATASET** MODEL **PROJECT THE ART PREPARATION TRAINING EXTENSION** Dataset selection Data understanding PointNet implementation **Instance Segmentation** PointNet evaluation Model selection Data preprocessing **DGCNN** implementation **DGCNN** evaluation

Results

	lou	Accuracy
PointNet	0.8423	0.8803
DGCNN	0.7818	0.8573

DGCNN PointNet Training and Validation Loss per Epoch Training and Validation Loss per Epoch Training Loss Validation Loss 3.5 1.2 3.0 2.5 SSO 2.0 0.8 1.5 0.6 1.0 0.4 0.5 20 80 100 10 20 30 60

Epoch

Training Loss

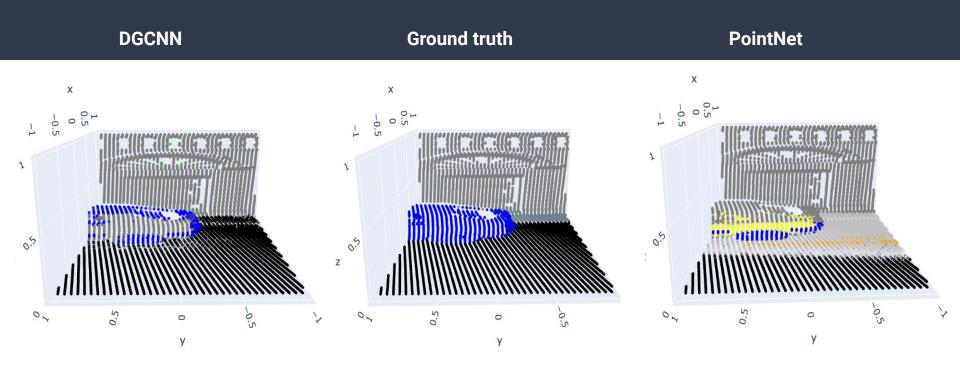
40

Epoch

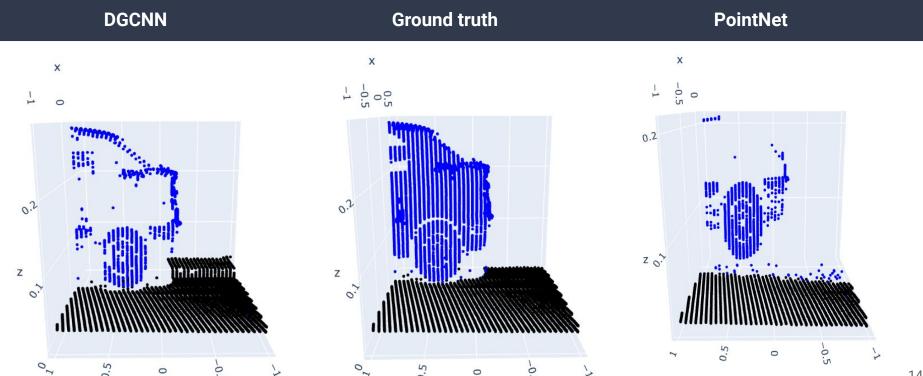
50

Validation Loss

Graphical results



Graphical results



Conclusions

Challenges Faced

- Dataset Selection: limited availability, binary data format, and incomplete annotations.
- Training: computational requirements extremely high for training stage.
- Data: Learn to treat data structure not seen in class. Architectures, batches, transformations, plotting...
- Class imbalance: highly unbalanced dataset

What have we learned?

- Importance of quality data
- Importance of using the right model
- Data preprocessing and normalization
- Hyperparameter optimization
- Evaluation and performance metrics
- How PointCloud architectures works
- Use of Google Cloud
- Use efficiently the available hardware

Thanks for your attention