



Hochschule  
Bonn-Rhein-Sieg  
University of Applied Sciences



Master's Thesis

# Out-of-distribution detection in 3D semantic segmentation

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Submitted to Hochschule Bonn-Rhein-Sieg,  
Department of Computer Science  
in partial fulfillment of the requirements for the degree  
of Master of Science in Autonomous Systems

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April 2022







I, the undersigned below, declare that this work has not previously been submitted to this or any other university and that it is, unless otherwise stated, entirely my own work.

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Date

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Lokesh Veeramacheneni



# Abstract

Your abstract





# Acknowledgements

Thanks to ....



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# 1

## Introduction

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## 1.1 Motivation

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## 1.2 Challenges and Difficulties

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### 1.2.3 ...

## 1.3 Problem Statement

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1.3.2 ...

1.3.3 ...

# 2

## State of the Art

### **2.1 ....**

Use as many sections as you need in your related work to group content into logical groups  
Don't forget to correctly cite your sources [? ].

### **2.2 Limitations of previous work**





# Methodology

Dataset benchmarking is important any task because it allows future researchers to compare and validate their methods. In this thesis, we tried to formulate a benchmark for out of distribution (OOD) detection in 3D datasets and evaluated the benchmarked datasets over the baseline models. In this chapter, we discuss about the benchmarking of datasets particularly how is it done and experimental setup which includes models used and also a description about the process of OOD detection.

## 3.1 Dataset benchmark formulation

In this era, development of novel architectures in deep learning is made easy by improvement in frameworks such as Pytorch and Tensorflow. These rapidly developed architectures requires a standard benchmarked datasets to compare performance with existing architectures. The process of creating benchmarked datasets with high quality are tedious and requires Herculean effort. Moreover the benchmarking for OOD detection task in 2D is already available in [cite]. The benchmarked datasets in OOD detection for 2D classification setting are MNIST vs Fashion MNIST [cite] or CIFAR vs SUN datasets [cite]. Since this thesis deals with OOD detection in 3D segmentation task and it is first of its kind no such benchmarking is available as best of our knowledge. As discussed in [cite], we chose the datasets for in and out distributions based on the criteria of *relevance*, *representativeness*, *experimentally verified case*, *scalability* and *resuability*. As argued in [cite] one more criteria is *non-redundancy* was tried to maintain and its only possible with hard OOD scenarios where classes doesn't overlap between datasets. In case of soft OOD there are class overlaps but it should not be a problem in OOD detection task.

### 3.1.1 SemanticKITTI

The first benchmarked datasets are of LiDAR point annotation datasets for 3D semantic segmentation. This thesis also include the study of the available LiDAR datasets, a detailed descripton of datasets and their classes are availbale in appendix [cite]. The discussion here is only confined to the datasets used for benchmarking. For this study we chose the SemanticKITTI dataset [?] as an in distribution dataset. We adopted SemanticKITTI becuase of its wide usage in evaluation of 3D semantic segmentation model performance. Moreover the dataset consits of high qualitative and quantative scans. Also the sensor used is Velodyne HDL-64E which is widely used sensor for LiDAR scans as its used in other datasets such

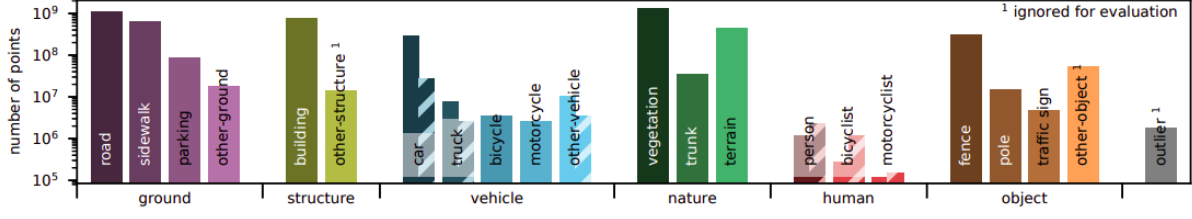


Figure 3.1: Classes in semanticKITTI dataset and their distribution in dataset. The hatched bars means a moving object where as solid bar means a non movable object. Image taken from [? ].

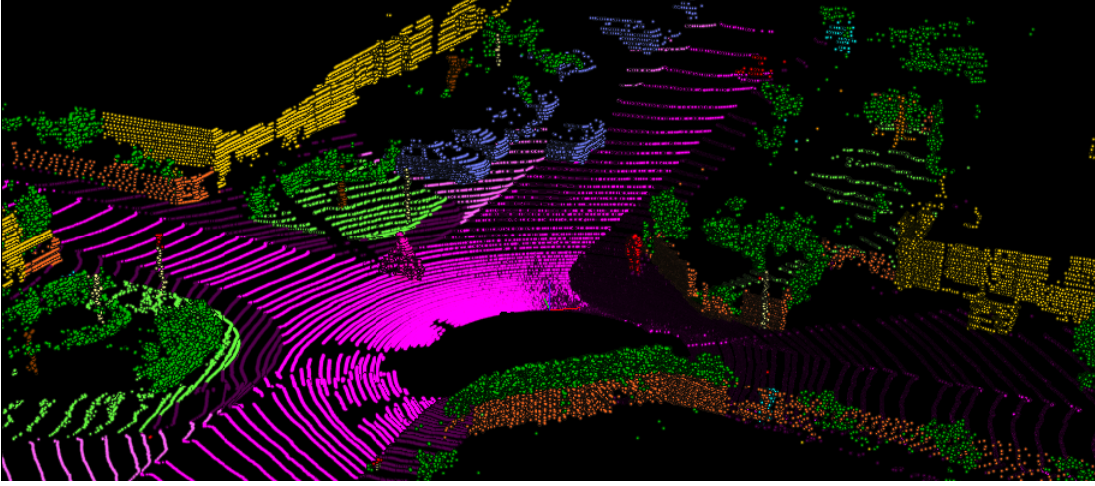


Figure 3.2: Ground truth example of a scan in SemanticKITTI dataset depicting various classes

as [cite]. SemanticKITTI [?] is a large dataset with 23201 and 20351 scans for training and testing respectively. The datasets has a gigantic 4549M number of points which are annotated individually. It has 28 classes annotated but only 25 are used for evaluation. The dataset is also publicly available at [cite] for download and API at [cite]. The available classes and their distribution in dataset is given in Figure 3.1. SemanticKITTI is an outdoor autonomous driving dataset as depicted in Figure 3.2.

### 3.1.2 Stanford 3D Indoor Scene Dataset (S3DIS)

# 4

## Solution

Your main contributions go here

### **4.1 Proposed algorithm**

### **4.2 Implementation details**



# 5

## Evaluation

Implementation and measurements.

---

# 6

## Results

### **6.1 Use case 1**

Describe results and analyse them

### **6.2 Use case 2**

### **6.3 Use case 3**





# 7

## Conclusions

**7.1 Contributions**

**7.2 Lessons learned**

**7.3 Future work**



# 8

## Notes/Remarks

### 8.1 Introduction-OOD/Anomaly/Distributional shift



Figure 8.1: Illustration of distributional shift, anomaly and out of distribution examples using various kind of ships. 8.1a represents the sail ship during 18th century. 8.1b depicts the current training data. 8.1c, 8.1d represents the anomalous ship data and 8.1e, 8.1f represents the OOD data. Images are taken from [25], [14], [17], [26], [8], and [3] respectively in the order they appear.

Let us time travel back to 18<sup>th</sup> century and assume that we had implemented a model to detect ships, the dataset images for the trained model will be similar to Figure 8.1a. 18<sup>th</sup> century ships as in 8.1a can be defined as “*ship contains hull and sails*”. Fast forward to present time, current ships are as shown in Figure 8.1b. Ship as in 8.1b can be defined as “*ship contains hull and passenger decks stacked upon each other*”. Now if we want to deploy the old model trained with old ships to detect the present generation of ships, it is difficult because of the change in definition and properties of ship. This change in data distribution over a period of time is called “*distributional shift*” of the data.

Anomaly can be defined as the patterns that doesn't conform to the expected training behavior. By this definition, Figure 8.1c and Figure 8.1d can be considered as anomalies. This is because Figure 8.1c is a container ship looking similar to Figure 8.1b instead of passenger decks we have containers stacked. Figure 8.1d is also anomaly because the Titanic also has a hull, passenger decks and chimneys. This additional chimneys as a features deviates this image from the definition of the ship and can be considered as “*anomaly*”.

The input for out of distribution (OOD) is drawn from an unknown distribution of unknown data, which is not near to the training distribution. Figures 8.1e and 8.1f are submarine and ariship which are from unknown distribution and they doesn't adhere to the definition of ship by any means. In general, one can argue that OOD can be defined as inputs which doesn't belong to any class in the training data.

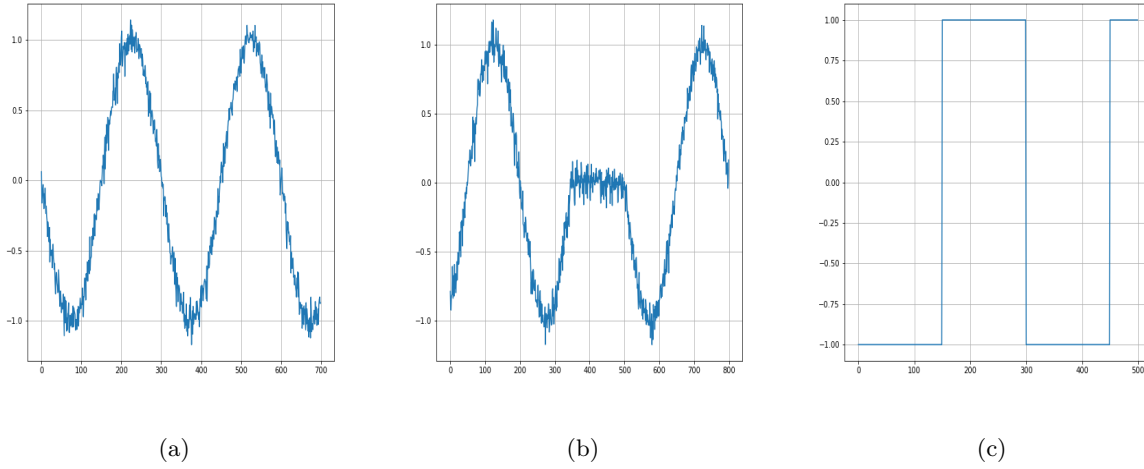


Figure 8.2: Illustration of anomaly and OOD with time series data as example. 8.2a depicts the training data as sinusoidal wave. 8.2b represents the anomaly in the sinusoidal wave and 8.2c represents the square wave as OOD signal

## 8.2 Related work - Datasets

LiDAR is one of the central component in the sensor suite for SLAM system in robotic applications [28], [20], [13] and autonomous driving [16]. 3D LiDAR data is preferred because, it can provide the exact replica of 3D geometry of the real world represented in the form of 3D point clouds. Because of these rich features and widespread use of LiDAR sensors, tasks such as 3D object detection [36], [35] and 3D semantic segmentation [21], [1] are becoming more predominant area for research.

In this section, we will discuss about the available 3D LiDAR datasets for 3D semantic segmentation task and classify the datasets based on acquisition methods as in [7]. [7] classifies the available public datasets into three classes based on the data acquisition process. They are *Sequential*, *Static* and *Synthetic* datasets. The data for sequential datasets are collected as frame sequences where mechanical LiDAR is mounted on top of a autonomous driving platform as in Figure 8.3. Most of the popular autonomous



Figure 8.3: Sequential mounted LiDAR for data collection of Lyft L5 dataset. Image from [15]

driving datasets are of sequential type, but these kind of datasets comes with a drawback of sparse points than other datasets.

Static datasets consists of data collected from a stationary view point by a terrestrial laser scanner. These kind of datasets capture the static information of the realworld whereas the sequential datasets capture the dynamic movements of the surrounding objects. Static datasets find their way in applications such as the urban planning, augmented reality and robotics. Figure 8.4 depicts a terrestrial laser scanner used to capture point cloud of an industrial environment. An advantage with the static datasets, are they



Figure 8.4: Terrestrial laser scanner in an industrial environment with the laser scanner mounted on a yellow tripod in the left corner of the floor. Image taken from [22]

can produce highly dense point clouds leading to rich 3D geometric representations.

Last type of 3D LiDAR datasets are synthetic datasets. As the name suggests these datasets are generated from the computer simulation. Figure 8.5 depicts a simulated point cloud in a synthtic dataset called SynthCity. Eventhough synthetic datasets can be generated in large scale with cheap cost, they lack the accuracy in detail when compared to the point clouds generated from real world.

The datasets belonging to the each acquisition type are summed up in Table 8.1. Most of the datasets from the Table 8.1 are taken from [7] and also as a part of this study, additional new datasets were added

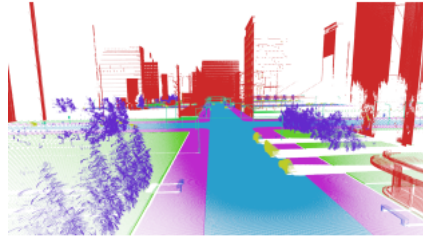


Figure 8.5: Illustration of a scene in synthetic dataset called SynthCity. Image taken from [11]

to the list. The newly added datasets include DALES [31], ScanObjectNN [29] in static acquisition mode and AIO Drive [32], Toronto3D [27] are additions in the sequential mode. [7] also classifies GTAV (#cite) dataset as synthetic 3D LiDAR but the corresponding paper doesn't report any LiDAR dataset and proposed only 2D dataset for segmentation. The limited number of datasets in 3D LiDAR allowed us to study the characteristics of each individual datasets such as each class, data distribution and features of each point in point cloud. It is summarized in Table (#ref) in Appendix (#chapter number)

acquisition mode	dataset	frames	points (in million)	classes	scene type
static	Oakland[18]	17	1.6	44	outdoor
	Paris-lille-3D[23]	3	143	50	outdoor
	Paris-rue-Madame[24]	2	20	17	outdoor
	S3DIS[2]	5	215	12	indoor
	ScanObjectNN[29]	-	-	15	indoor
	Semantic3D[12]	30	4009	8	outdoor
	TerraMobilita/IQmulus[30]	10	12	15	outdoor
	TUM City Campus[9]	631	41	8	outdoor
	DALES[31]	40 (tiles)	492	8	outdoor
sequential	A2D2[10]	41277	1238	38	outdoor
	AIO Drive[32]	100	-	23	outdoor
	KITTI-360[34]	100K	18000	19	outdoor
	nuScenes-lidarseg[5]	40000	1400	32	outdoor
	PandaSet[33]	16000	1844	37	outdoor
	SemanticKITTI[4]	43552	4549	28	outdoor
	SemanticPOSS[19]	2988	216	14	outdoor
	Sydney Urban[6]	631	-	26	outdoor
	Toronto-3D[27]	4	78.3	8	outdoor
synthetic	SynthCity[11]	75000	367.9	9	outdoor

Table 8.1: 3D LiDAR datasets classified based on the acquisition type. Table updated from [7]

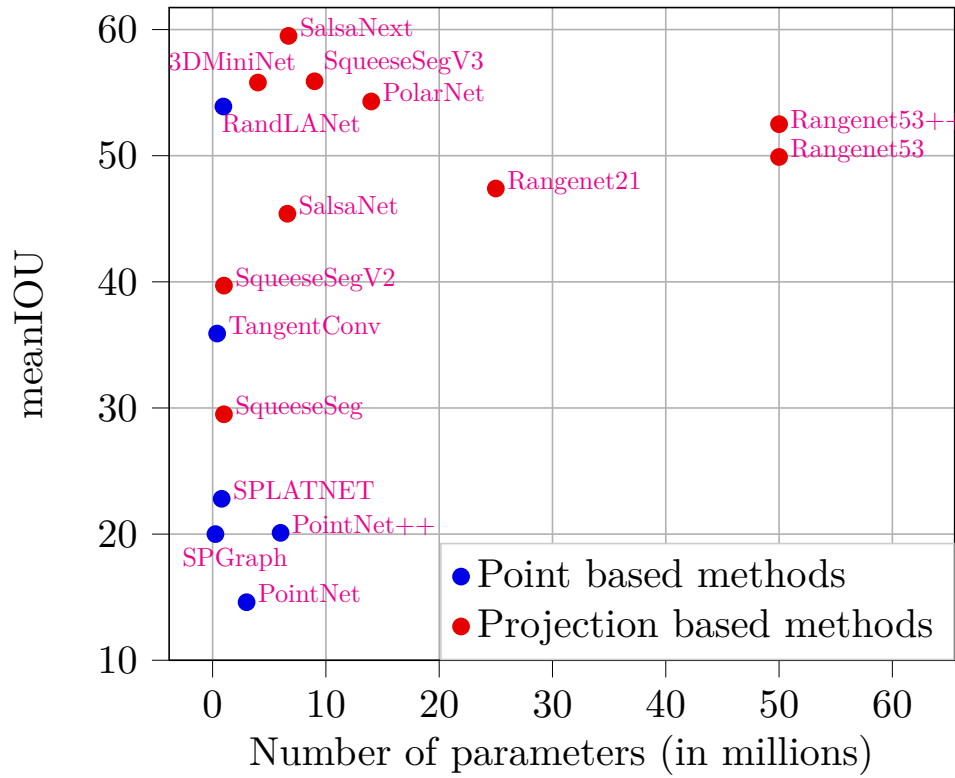


Figure 8.6: Comparison of 3D semantic segmentation methods performance on SemanticKITTI dataset against the number of parameters. Blue points represent point based methods and red represented projection based methods.

### 8.3 Related work - Models







## Design Details

Your first appendix

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# B

## Parameters

Your second chapter appendix

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