State of the Art This chapter will discuss the 3D LiDAR datasets available and classify them based on the type of

Static datasets are advantageous because they can produce highly dense point clouds leading to rich geometric repr The last type of 3D LiDAR dataset is the synthetic dataset. As the name suggests, these datasets are generated from The datasets belonging to each acquisition type are summed up in Table??. Most of the datasets from Table?? and the datasets from Table??

	acquisition mode	e dataset	frames points (in million) classes scene type				
	7*static	Oakland[?]	17	1.6	44	outdoor	
		Paris-lille-3D[?]	3	143	50	outdoor	3D LiDAR datasets cla
		Paris-rue-Madame[?]	2	20	17	outdoor	
[h!]		S3DIS[?]	5	215	12	indoor	
		ScanObjectNN[?]	-	-	15	indoor	
		Semantic3D[?]	30	4009	8	outdoor	
		TerraMobilita/IQmulus[?	] 10	12	15	outdoor	
		TUM City Campus[?]	631	41	8	outdoor	
		DALES[?]	40 (tiles)	492	8	$\operatorname{outdoor}$	
		A2D2[?]	41277	1238	38	$\operatorname{outdoor}$	
		AIO Drive[?]	100	-	23	$\operatorname{outdoor}$	
		KITTI-360[?]	100K	18000	19	outdoor	
		nuScenes-lidarseg[?]	40000	1400	32	$\operatorname{outdoor}$	
		PandaSet[?]	16000	1844	37	$\operatorname{outdoor}$	
		SemanticKITTI[?]	43552	4549	28	$\operatorname{outdoor}$	
		SemanticPOSS[?]	2988	216	14	$\operatorname{outdoor}$	
		Sydney Urban[?]	631	-	26	$\operatorname{outdoor}$	
		Toronto-3D[?]	4	78.3	8	$\operatorname{outdoor}$	
	1*synthetic	SynthCity[?]	75000	367.9	9	outdoor	

We chose the Semantic3D dataset as the in-distribution (ID) training dataset from the available datasets. S3DIS is 3D semantic segmentation models

This section will discuss the methods available for 3D semantic segmentation. The discussion includes a brief peek Traditional approach Traditional methods involve a complex features extraction and passing these features to a class Deep learning approach Deep learning based models are efficient at segmentation and can be divided into three typ Let us look briefly into the point-based methods. Point-based methods usually utilize fully connected layers or trad The other flavours of point-based methods include projection onto a d-dimensional lattice or making the point clou In projection-based models, let us discuss the models that project the data onto a 2D range image. The first of this We now discuss the projection-based models which involve bird eye view projection. These models are relatively ne Table ?? illustrates the detailed summary of each model and the number of parameters. Finally, graph neural netw Uncertainty estimation methods This section will discuss existing methods to estimate uncertainty in deep neural networks are trained in the epistemic and aleatoric uncertainty. Epistemic uncertaints Because of the higher computational complexity and resource intensiveness, there exist multiple flavours of the deep Other methods include snapshot ensembles [?] which iterate over the multiple local optima in the optimization land Bayesian methods Existing neural networks are trained in maximum likelihood, resulting in point estimates for the

Here  $\theta$  represents network parameters (weights),  $p(\theta)$  represents prior distribution over  $\theta$ , and x and y represent the inp

Here  $\theta$  represents trained network parameters, x and y represent the training set, and  $x_t$  and  $y_t$  represent the test set. VI is an approximation method where the posterior probability  $p(\theta|x,y)$  is approximated by specific distributions r Another widely known example for VI is Monte Carlo Dropout (MC-Dropout), in which the dropout layers are referenced out-of-distribution (OOD) detection methods

OOD/Anomaly/Distributional shift [h!] 0.333 [height=0.15width=0.95]images/intro<sub>o</sub>od<sub>a</sub>nomaly/old<sub>s</sub>hip.jpg 0.333 Let us time travel back to 18<sup>th</sup> century and assume that we had implemented a DNN model to detect ships. The d Where as an anomaly can be defined as the patterns that do not conform to the expected training behaviour as pro In the case of OOD, the input is drawn from an unknown distribution of unknown data, which is not near to the training section will discuss the existing OOD detection methods for 2D classification and 2D semantic segmentation to The most widely used benchmark datasets for 2D classification are CIFAR-10 vs SVHN [?], CIFAR-10 vs LSUN [?]

<sup>[?]</sup> has proposed a threshold-based OOD detection method using the Mahalanobis distance as a confidence score. The Mahalanobis distance as a confidence score.

where h(x) is final layer activations, c is the constant and  $h^-(x)$  is ReAct output of h(x). The score from this ReAct fu The OOD data samples have higher uncertainty than ID samples, allowing us to use the uncertainty estimation me In recent years, OOD detection for the task of 2D semantic segmentation has been getting into the limelight. [?] pr