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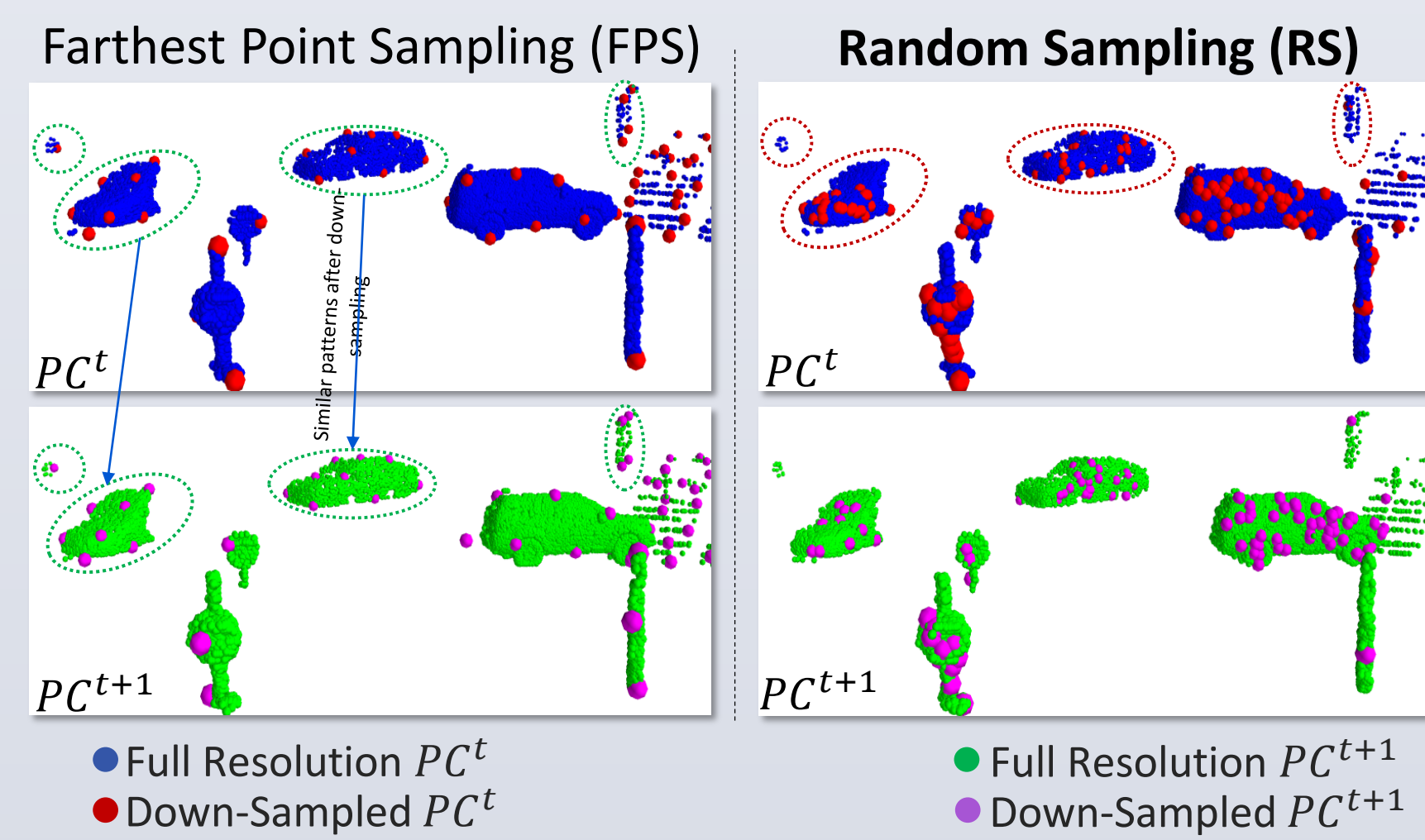
What is Scene Flow?

- 3D Motion field estimation.
- Perception of dynamic changes.
- LiDAR-based / Point-based Solutions.
- Advantages:
 - Impressive Results.
 - Strong Generalization.
- Disadvantages:
 - Low Efficiency.
 - Less Density.



Challenges:

- RS is much efficient, however:
 - Over-representation for locally dense regions.
 - Unevenly distribution.
 - Dissimilar patterns.

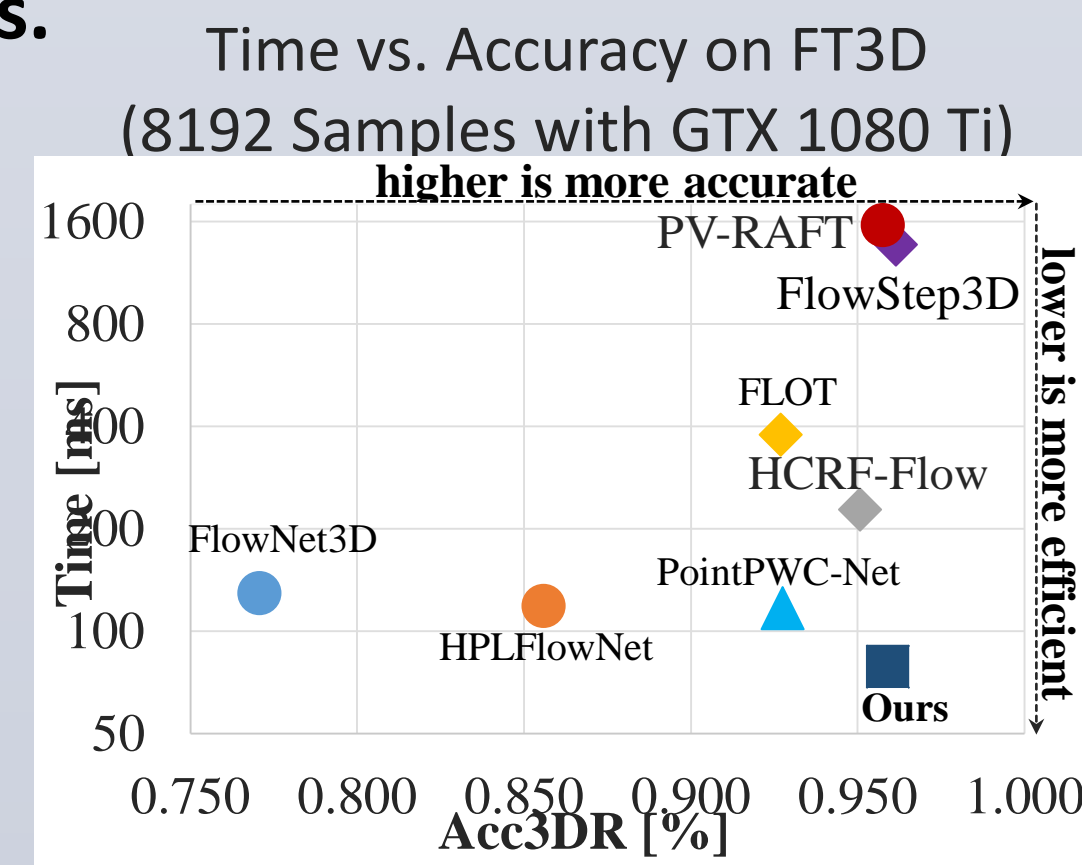


Overview:

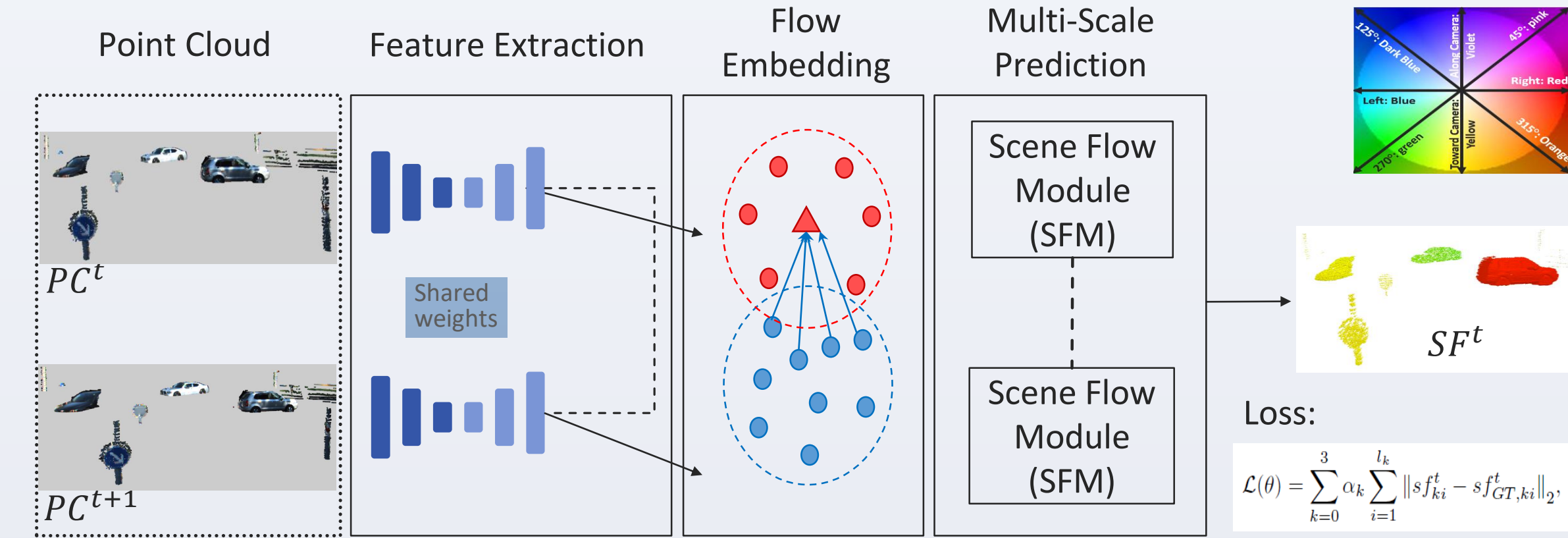
RMS-FlowNet is a fully supervised network for scene flow estimation with high efficiency for large-scale point clouds.

Advantages:

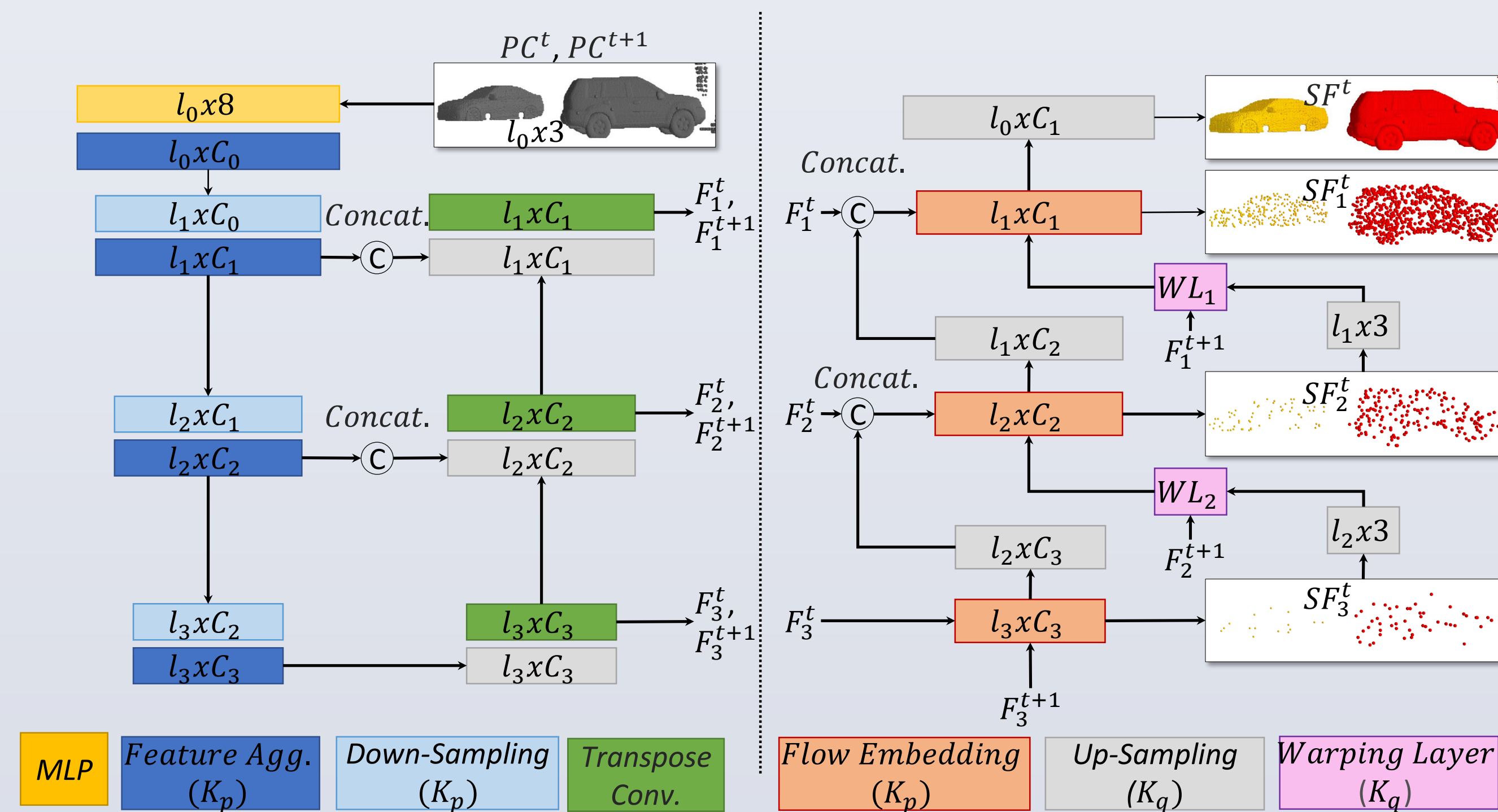
- Hierarchical End-to-End Prediction.
- Efficient Point-based Architecture.
- Random Sampling (RS).
- Attention Technique [RandLA-Net].
- Novel Flow Embedding.



RMS-FlowNet Network:

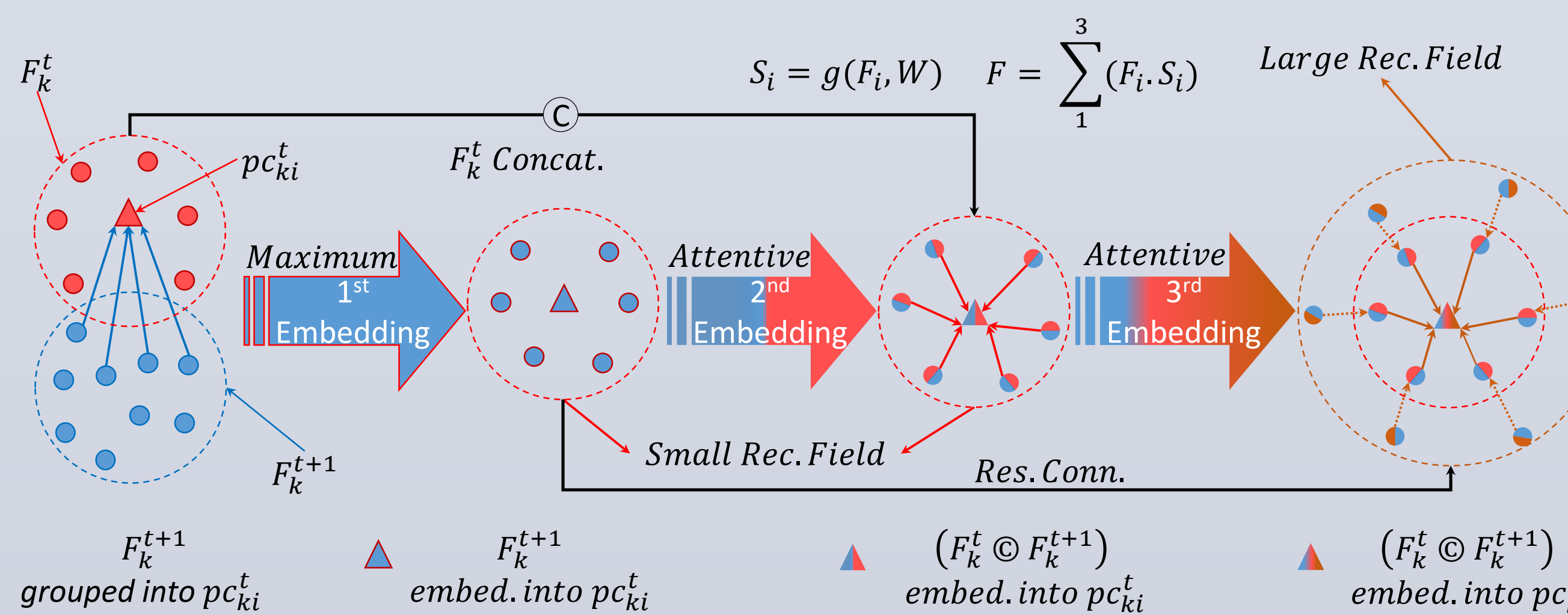


Multi-Scale Predictions:



Hierarchical Feature Extraction and Hierarchical Scene Flow Prediction.

Novel Flow Embedding:



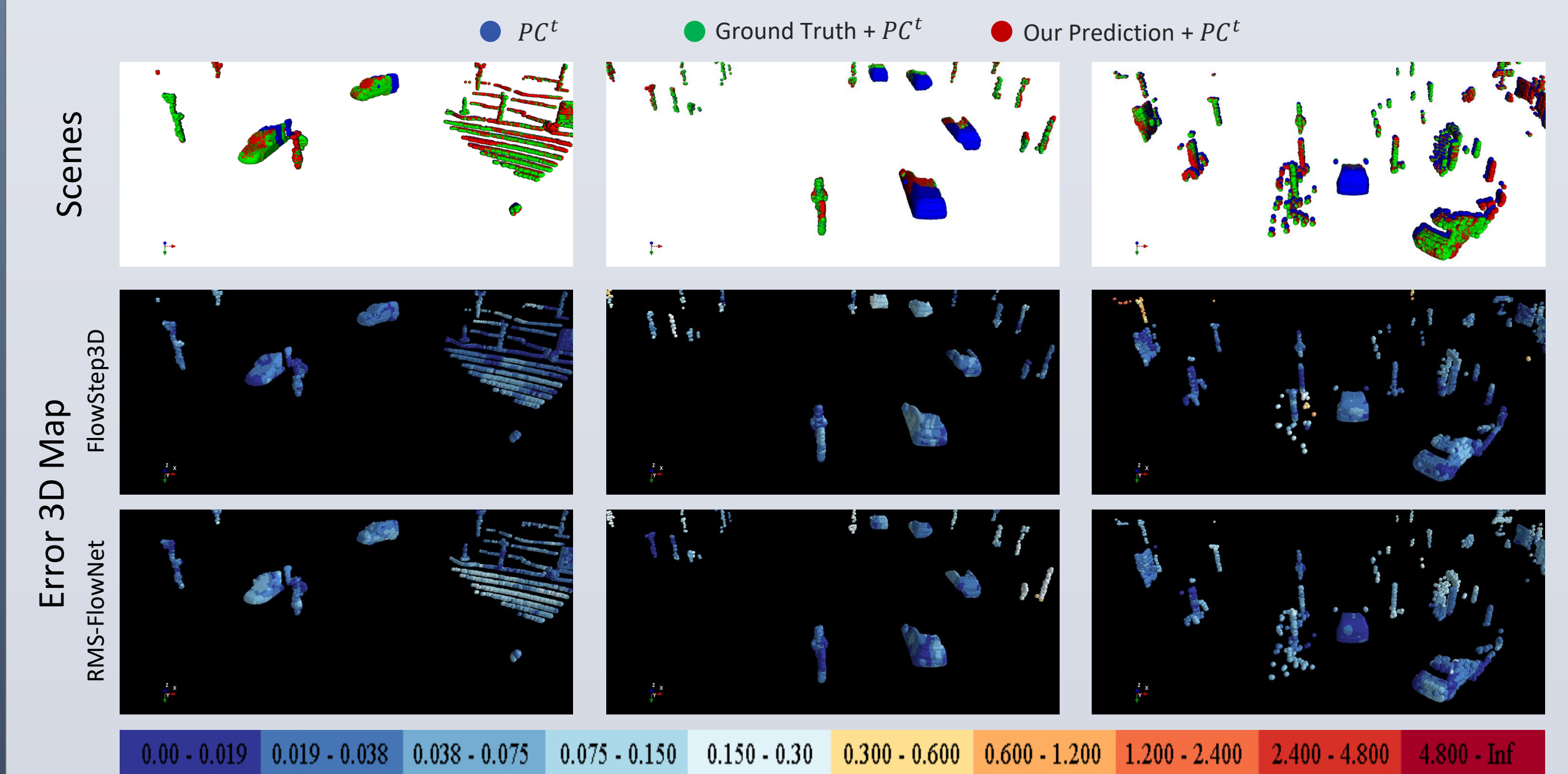
Patch-to-Dilated Patch flow embedding module is inspired by [RandLA-Net]

Results:

Quantitative Results: 8192 Samples with GTX 1080 Ti.

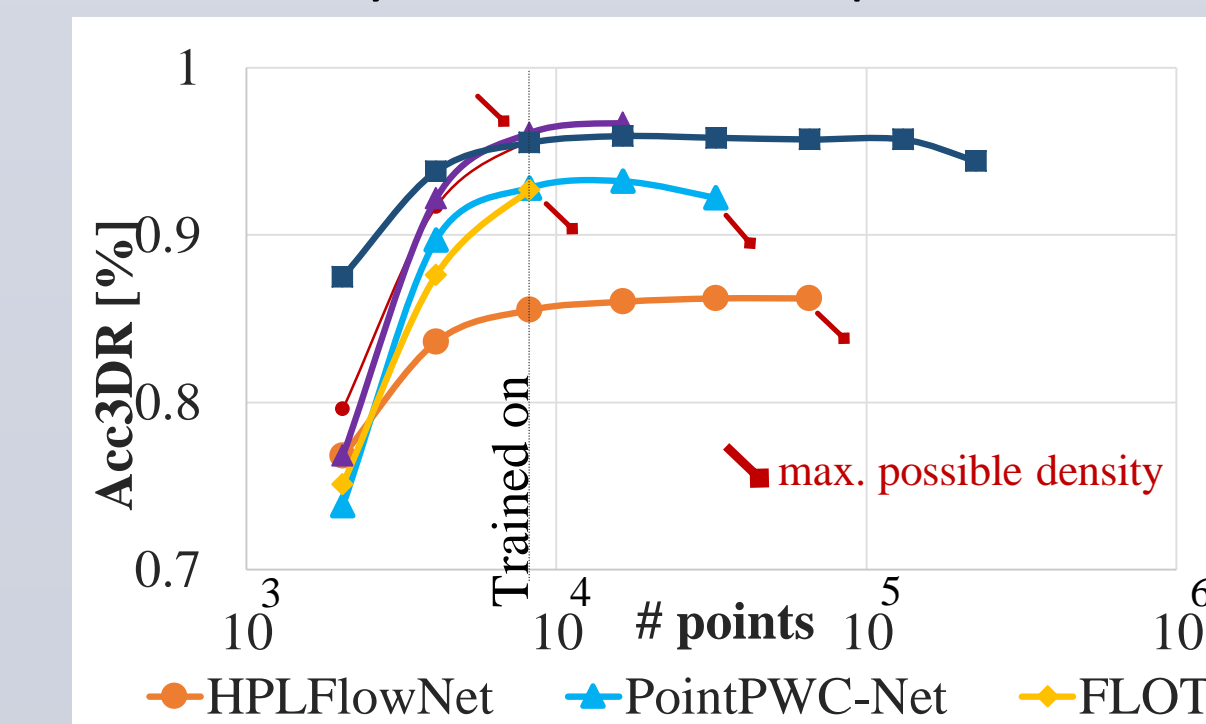
Model	Sampling	EPE3D [m]	KEO3D [%]	ACC3DS [%]	ACC3DR [%]	Time [ms]	GPU [GB]	
FT3D (Trained Domain)	FlowNet3D	FPS	0.114	0.602	0.413	0.771	132	10.85
	HPLFlowNet	-	0.080	0.428	0.616	0.856	119	1.58
	PointPWC-Net	FPS	0.059	0.342	0.738	0.928	117	2.86
	HALFlow	FPS	0.049	0.308	0.785	0.947	-	-
	FLOT	-	0.052	0.357	0.732	0.927	376	3.84
	FlowStep3D	FPS	0.046	0.217	0.816	0.961	1369	1.31
	PV-RAFT	-	0.046	0.292	0.817	0.957	1565	4.03
	WSLR	-	0.052	0.361	0.746	0.936	234	-
	HCRF-Flow	FPS	0.049	0.261	0.834	0.951	228	-
	RMS-FlowNet (Ours)	RS	0.051	0.309	0.800	0.956	77	1.39
RMS-FlowNet (Ours)	FPS	0.051	0.322	0.791	0.955	134	10.60	
KITTI (w/o Fine-tuning)	FlowNet3D	FPS	0.177	0.527	0.374	0.668	132	10.85
	HPLFlowNet	-	0.117	0.410	0.478	0.778	119	1.58
	PointPWC-Net	FPS	0.069	0.265	0.728	0.888	117	2.86
	HALFlow	FPS	0.062	0.249	0.765	0.903	-	-
	FLOT	-	0.054	0.244	0.764	0.915	376	3.84
	FlowStep3D	FPS	0.054	0.233	0.814	0.926	1369	1.31
	PV-RAFT	-	0.045	0.196	0.856	0.956	1565	4.03
	WSLR	-	0.042	0.208	0.849	0.959	234	-
	HCRF-Flow	FPS	0.053	0.180	0.863	0.944	228	-
	RMS-FlowNet (Ours)	RS	0.054	0.211	0.811	0.934	77	1.39
RMS-FlowNet (Ours)	FPS	0.048	0.187	0.876	0.957	134	10.60	

Qualitative Results: 8192 Samples with GTX 1080 Ti.

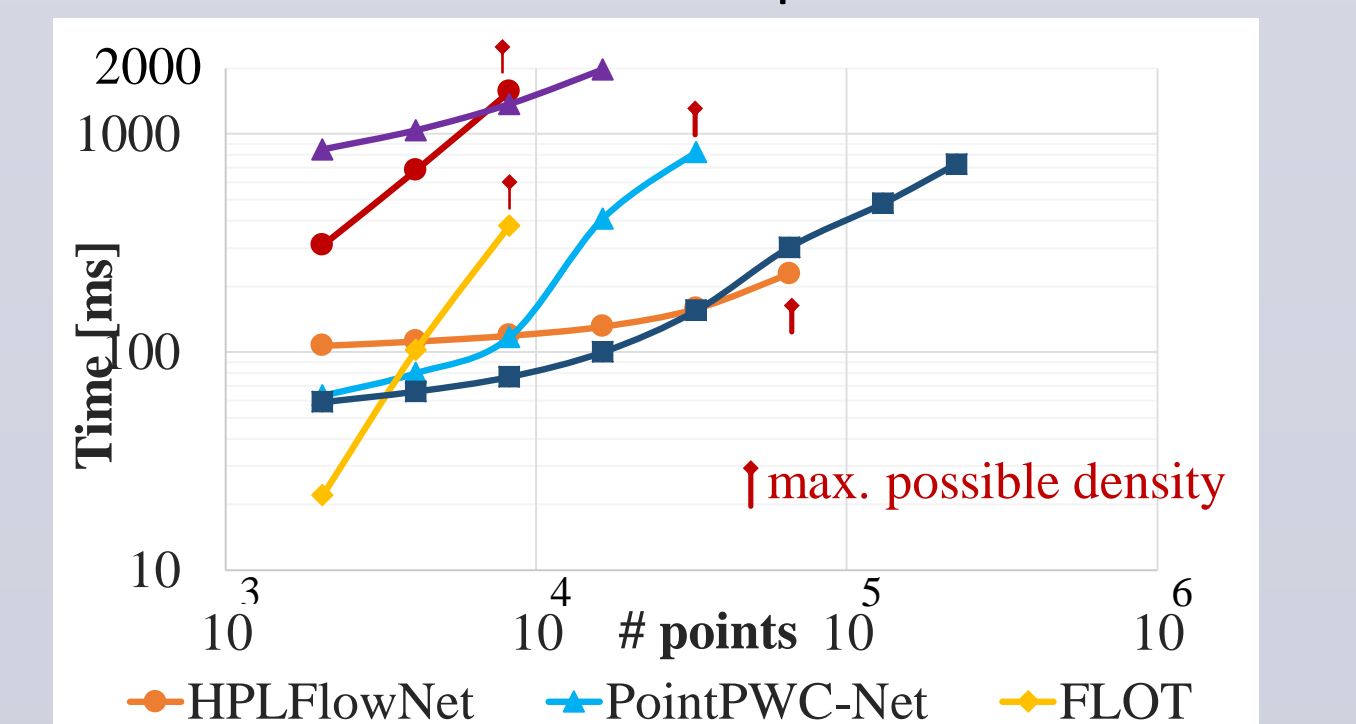


Robustness with High Density: Ours can operate on high dense (> 400K at once).

Accuracy over Dense Samples on FT3D



Time over Dense Samples on FT3D



More Details in our Paper:



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