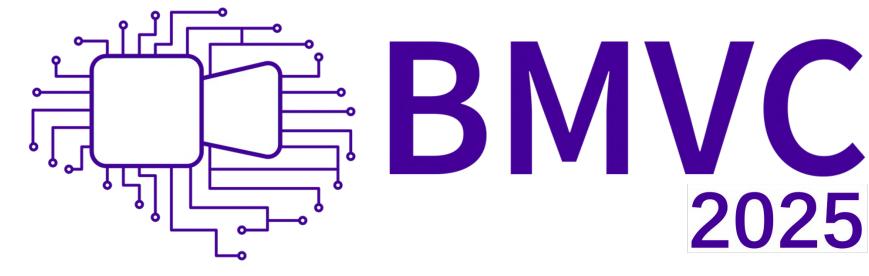
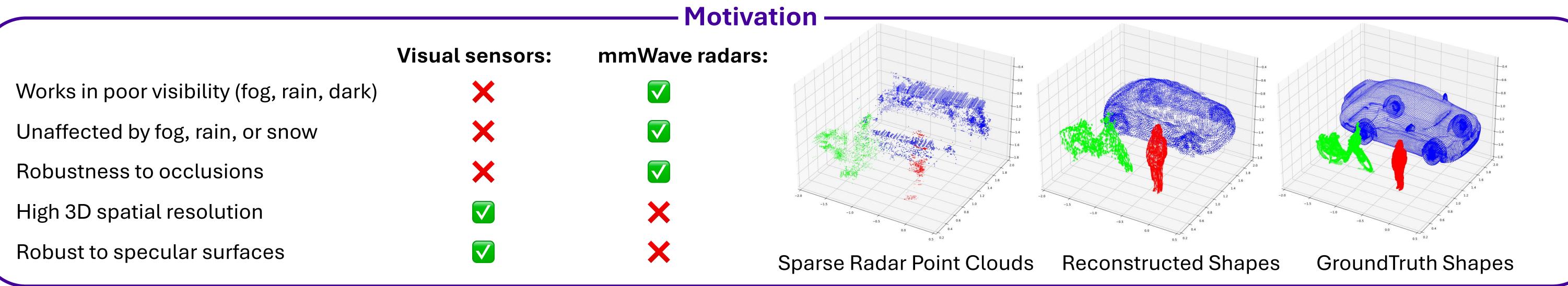
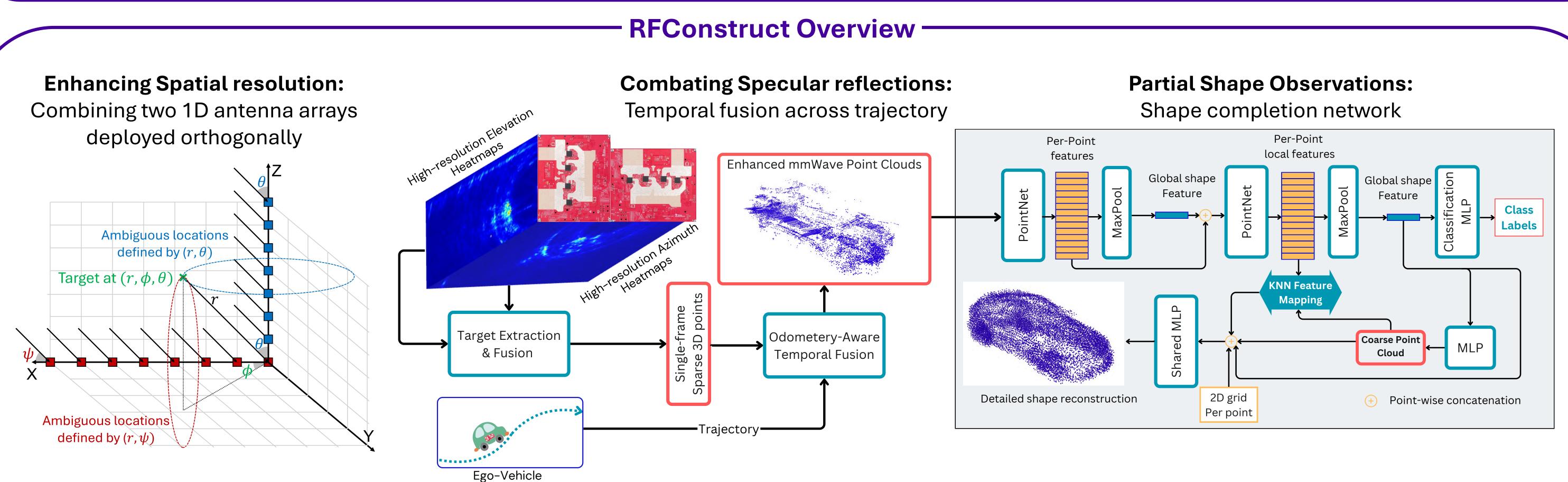


3D Shape Reconstruction from Autonomous Driving Radars

Junfeng Guan^{1,3} Samah Hussein¹ Swathi Shree Narashiman¹ Saurabh Gupta² Haitham Hassanieh¹ ¹EPFL ³Bosch Research ²UIUC







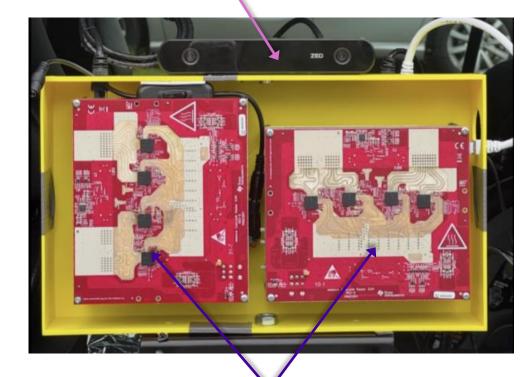
Dataset ·

RFConstruct is **trained on a synthetic dataset** extracted from ShapeNet[2] and CAESAR[3] to resemble radar data via two methods:

- Augmenting parital shape point clouds with noise and cropped portions to emulate signal specularity
- Simlating radar signals using a mmwave radar simulator introduced in [1]

RFConstruct is then fine-tuned using real data collected with our setup for the classes of objects most prominent in self-driving scenarios: Cars, Humans, Bikes, Motorbikes. This data includes 162 cars, 91 bikes, and 52 humans as point clouds.

Experimental Setup ZED 2i Depth camera

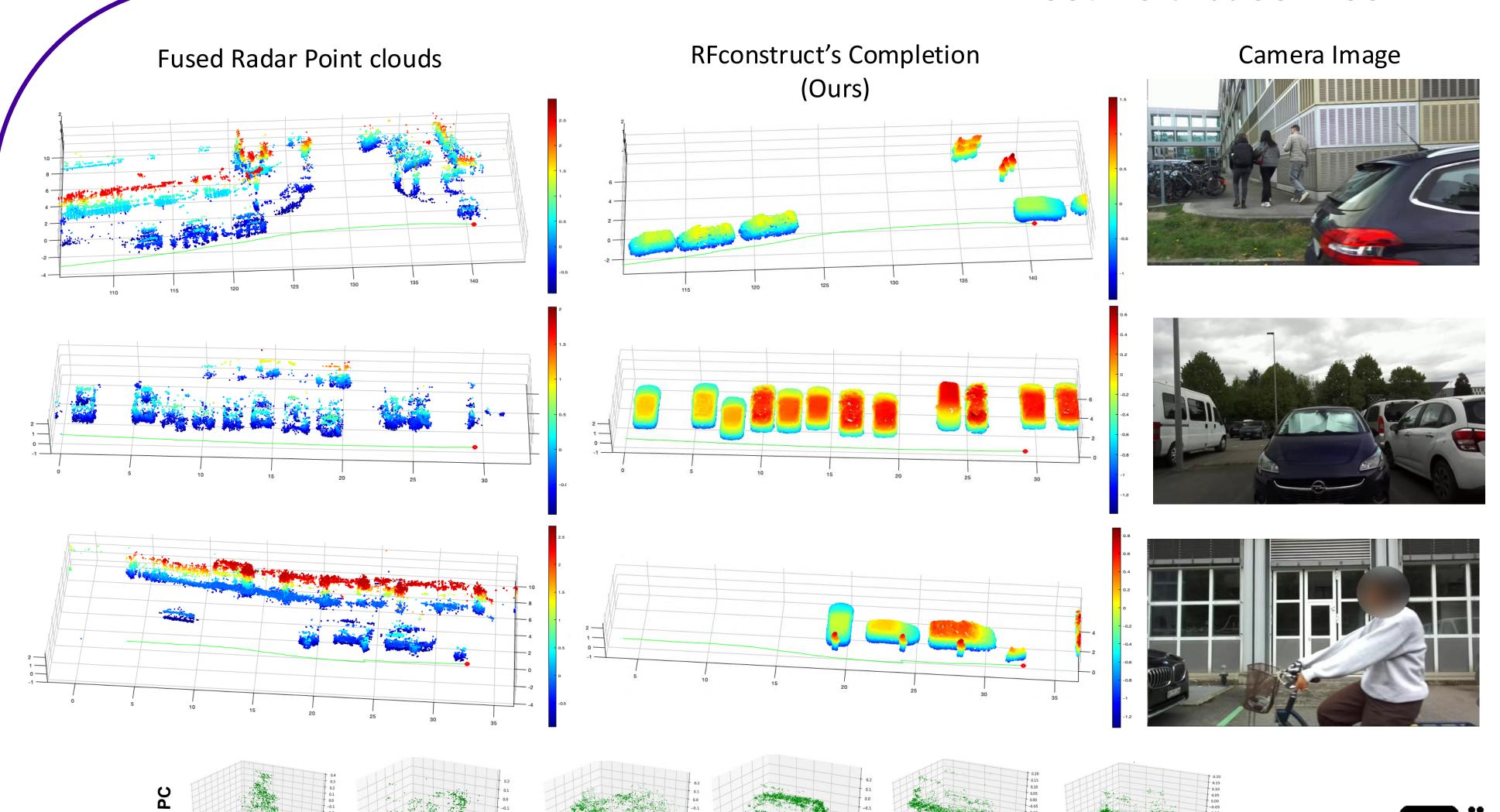


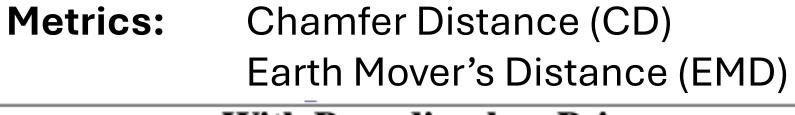


In our simulations, data collection, and evaluation, we ensure that the trajectory resembles natural driving scenarios where mostly one side of an object is briefly captured as the Ego-Vehicle is passing by.

Two orthogonal mmWave Cascaded radars from Texas Instruments

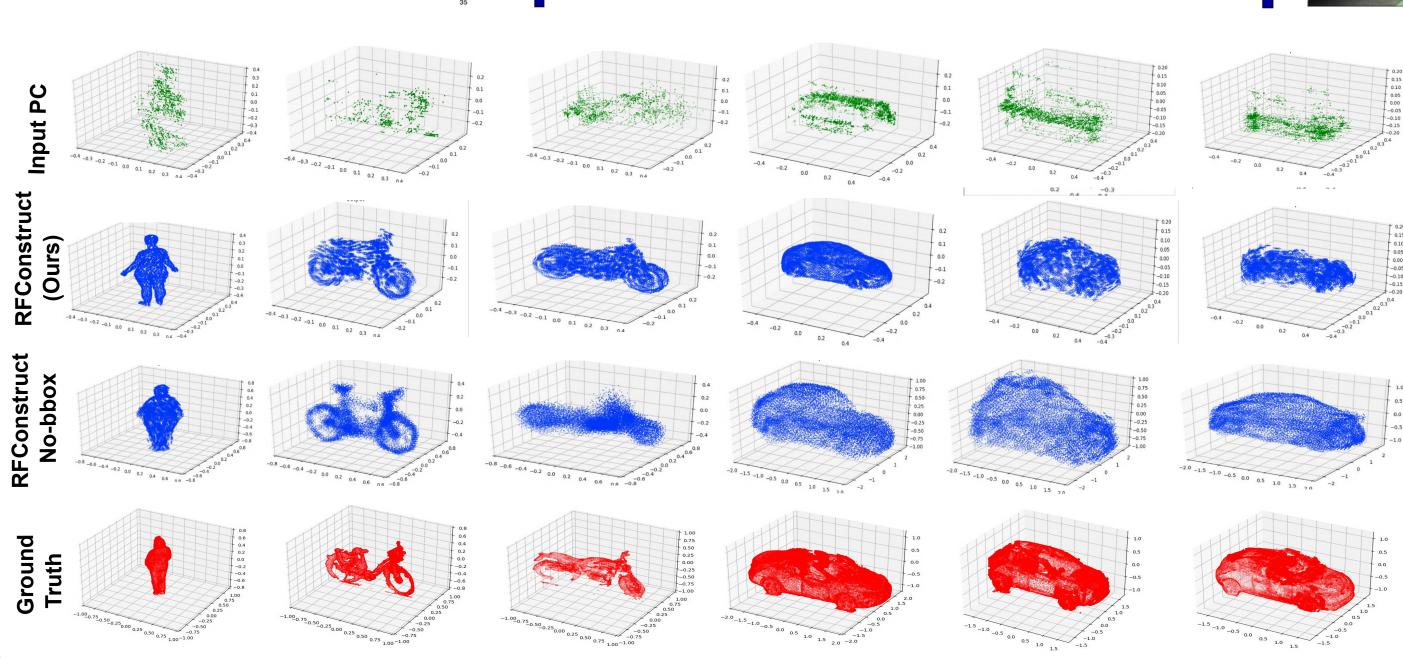
Results & baselines

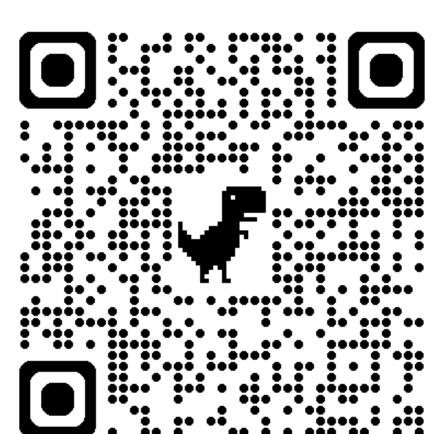




With Bounding-box Priors		
	Mean CD	Mean EMD
ODGNet	3.89	16.07
AdaPointTr	3.42	14.94
PCN	4.58	13.03
Medoid Baseline	2.52	9.16
RFconstruct	3.52	11.7
Fine-tuned RFconstruct	2.01	6.25

Without Bounding-box Priors			
	Mean CD	Mean EMD	
ODGNet	52.2	108	
AdaPointTr	43	91.46	
PCN	42.1	90.8	
Medoid Baseline	N/A	N/A	
RFconstruct	28.8	82.7	
Fine-tuned RFconstruct	15.3	49.9	





Check out the project Website!



Check out **RFConstruct Demo!**