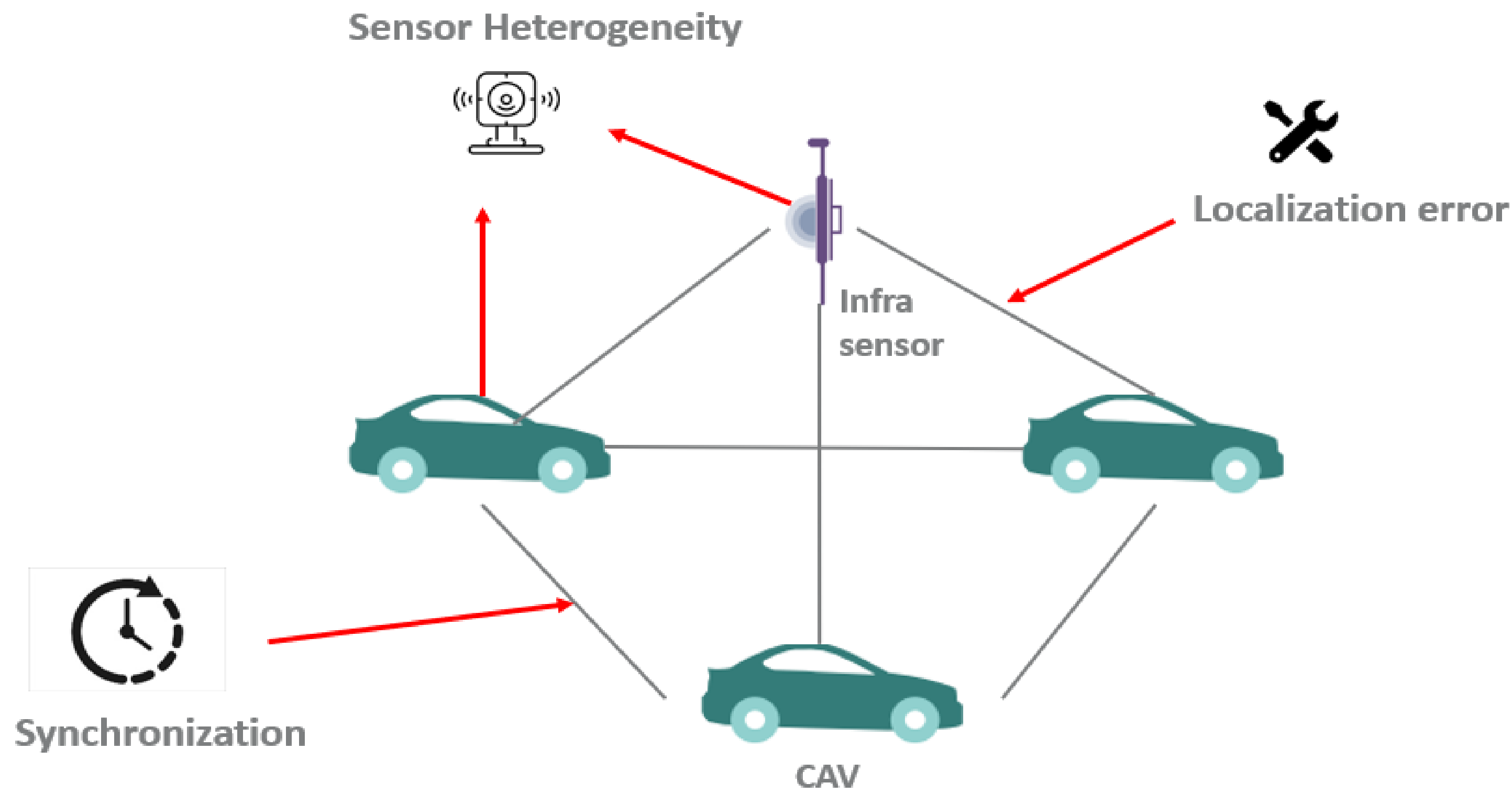


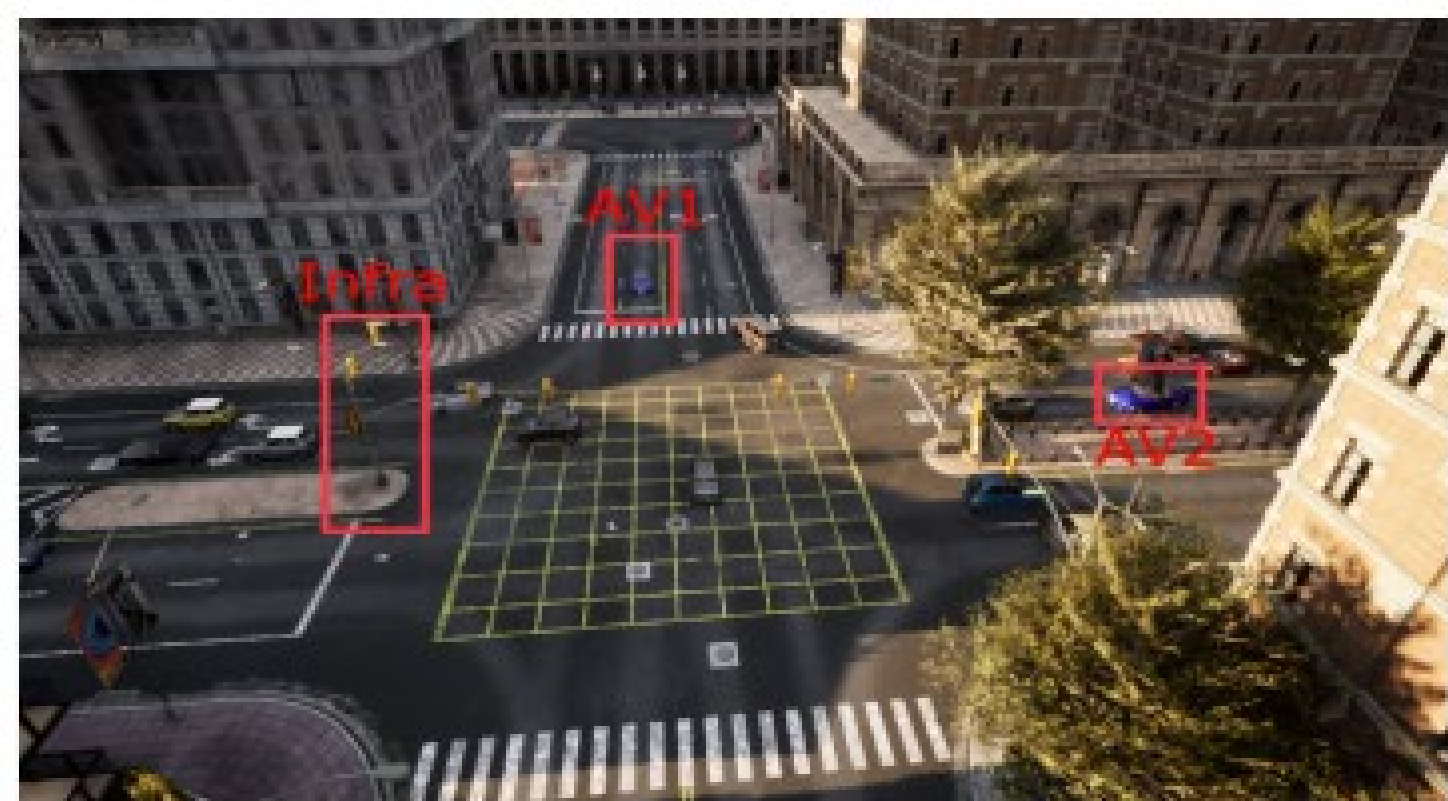
V2X Cooperative Challenge



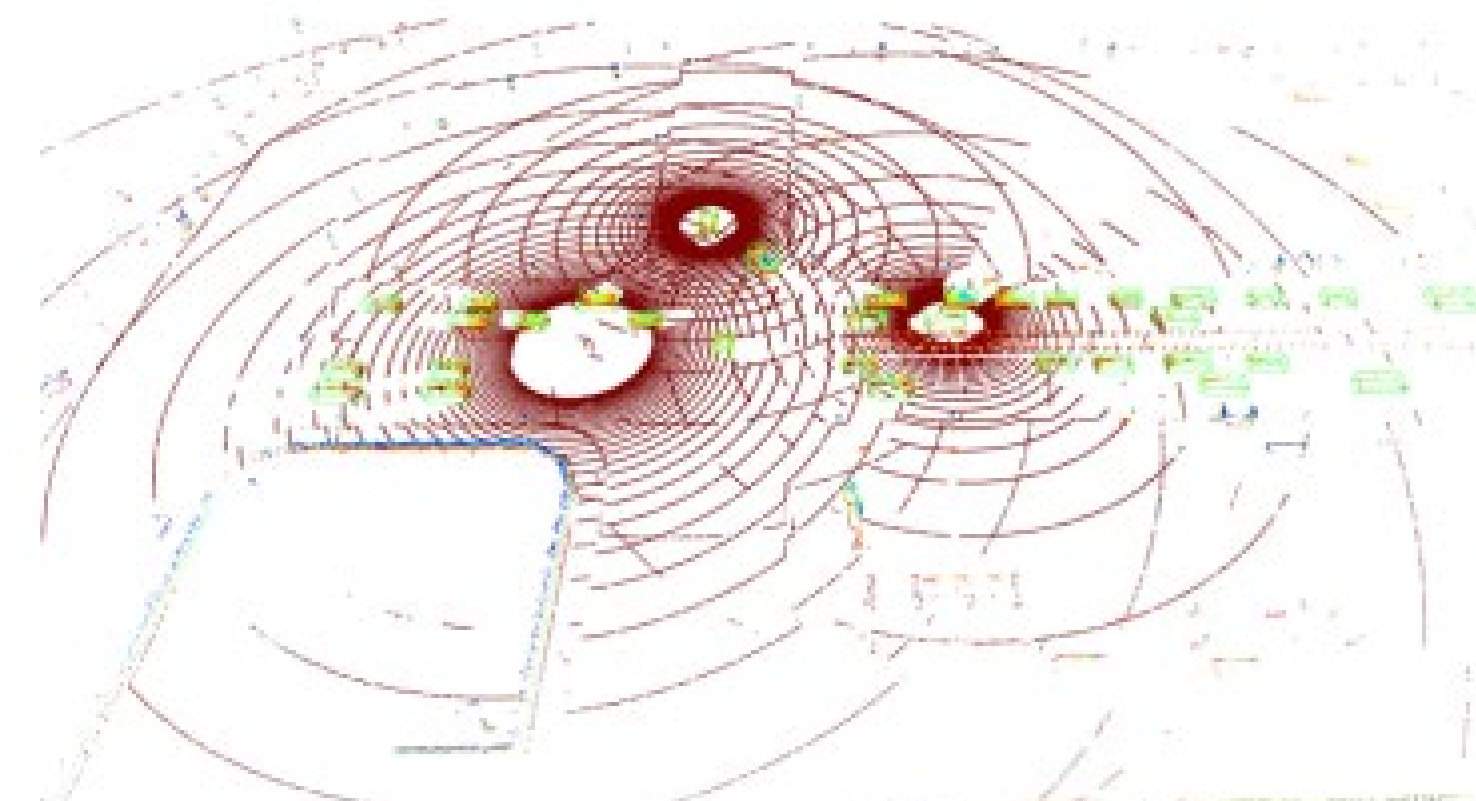
Our contributions:

- We present the first unified transformer architecture (V2X-ViT) for V2X perception, which can capture the heterogeneity nature of V2X systems with strong robustness against various noises.
- We propose a novel heterogeneous multi-agent attention module (HMSA) tailored for adaptive information fusion between heterogeneous agents.
- We present a new multi-scale window attention module (MSwin) that simultaneously captures local and global spatial feature interactions in parallel.
- We construct V2XSet, a new large-scale open simulation dataset for V2X perception, which explicitly accounts for imperfect real-world conditions.

V2XSet: A new V2X Perception dataset

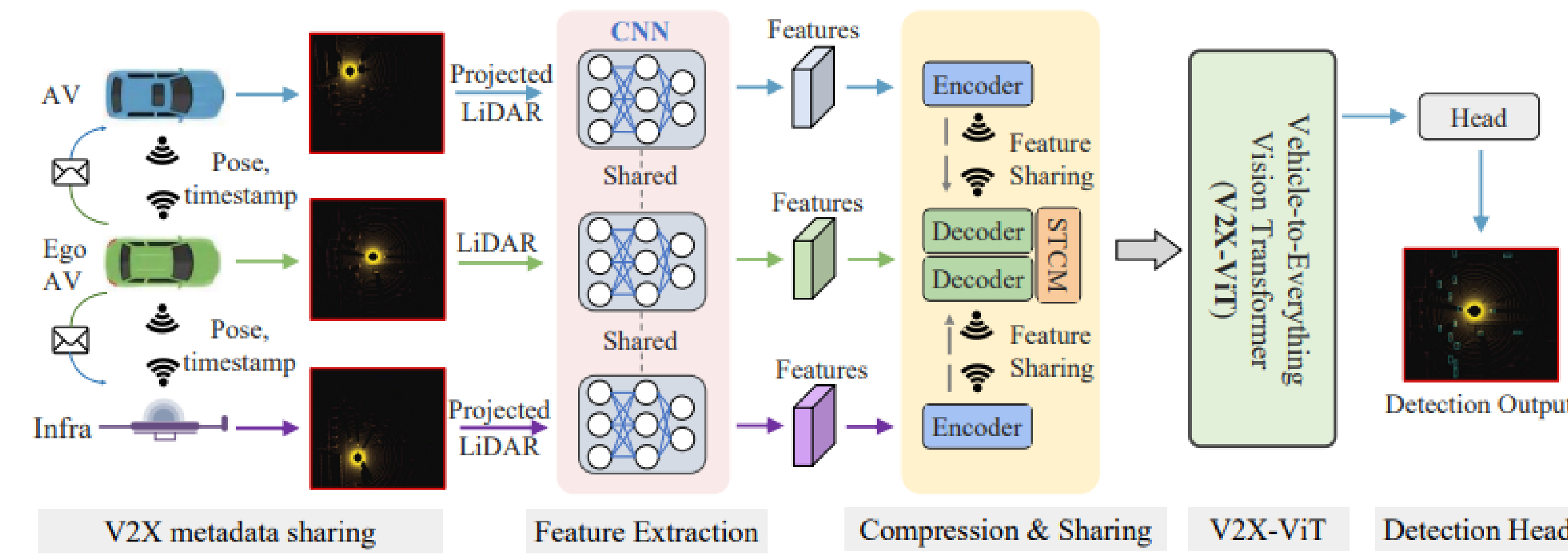


(a) Snapshot of Simulation



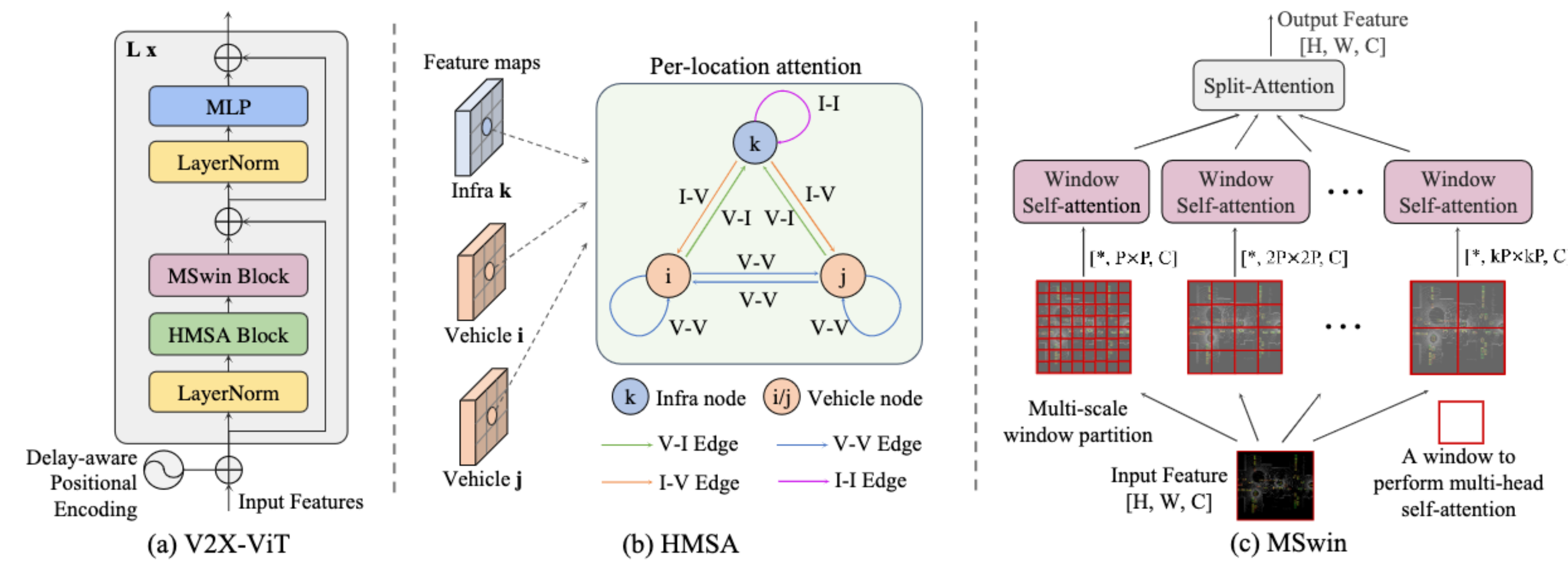
(b) Aggregated LiDAR point cloud

V2X-ViT Overall Framework



V2X-ViT Architecture

- Learn inter-agent interaction and per-agent spatial attention.
- HMSA captures heterogeneity between infra and vehicle.
- MSwin improves the robustness against localization error
- DPE encodes the temporal information

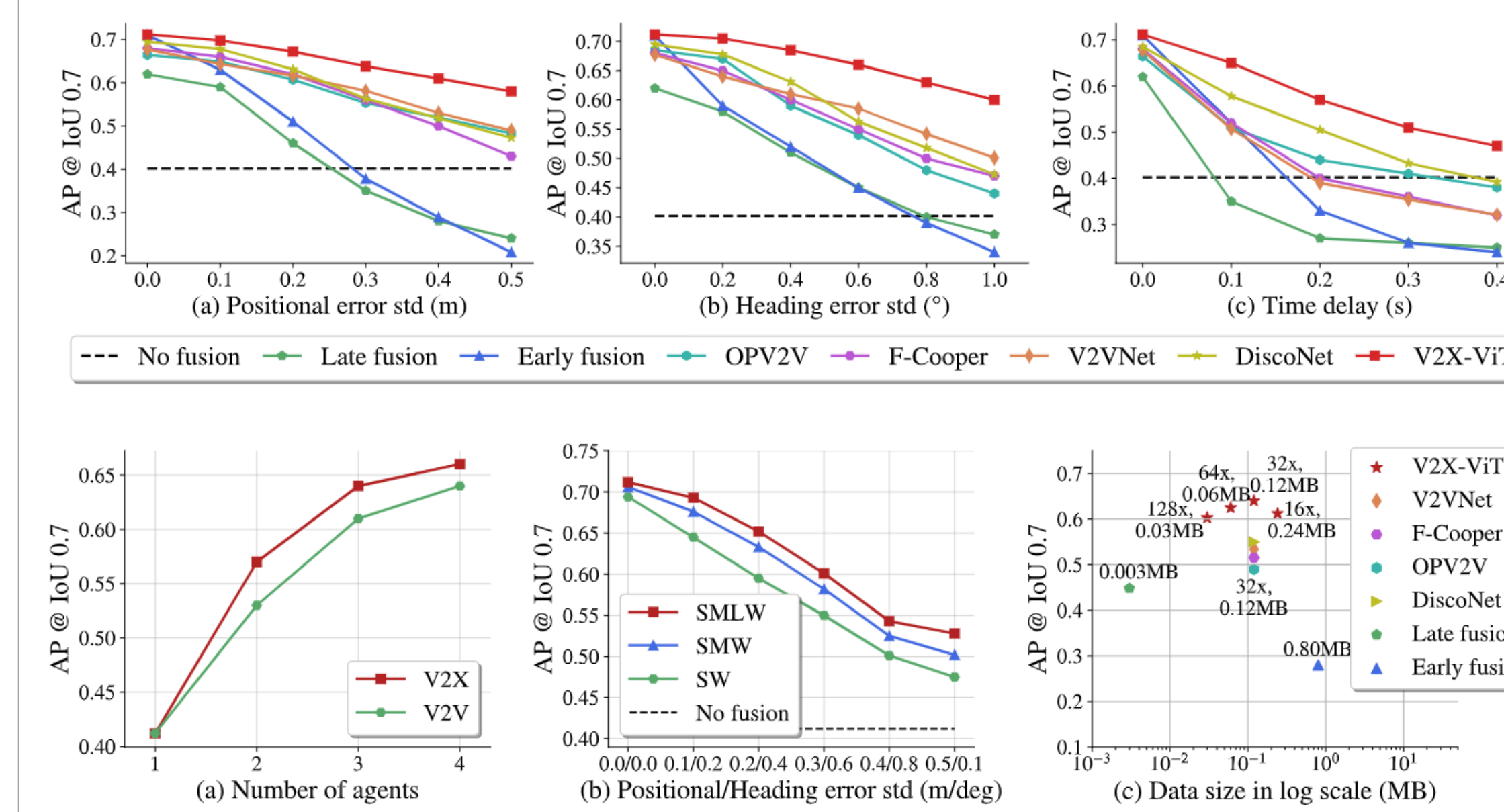


Benchmark results

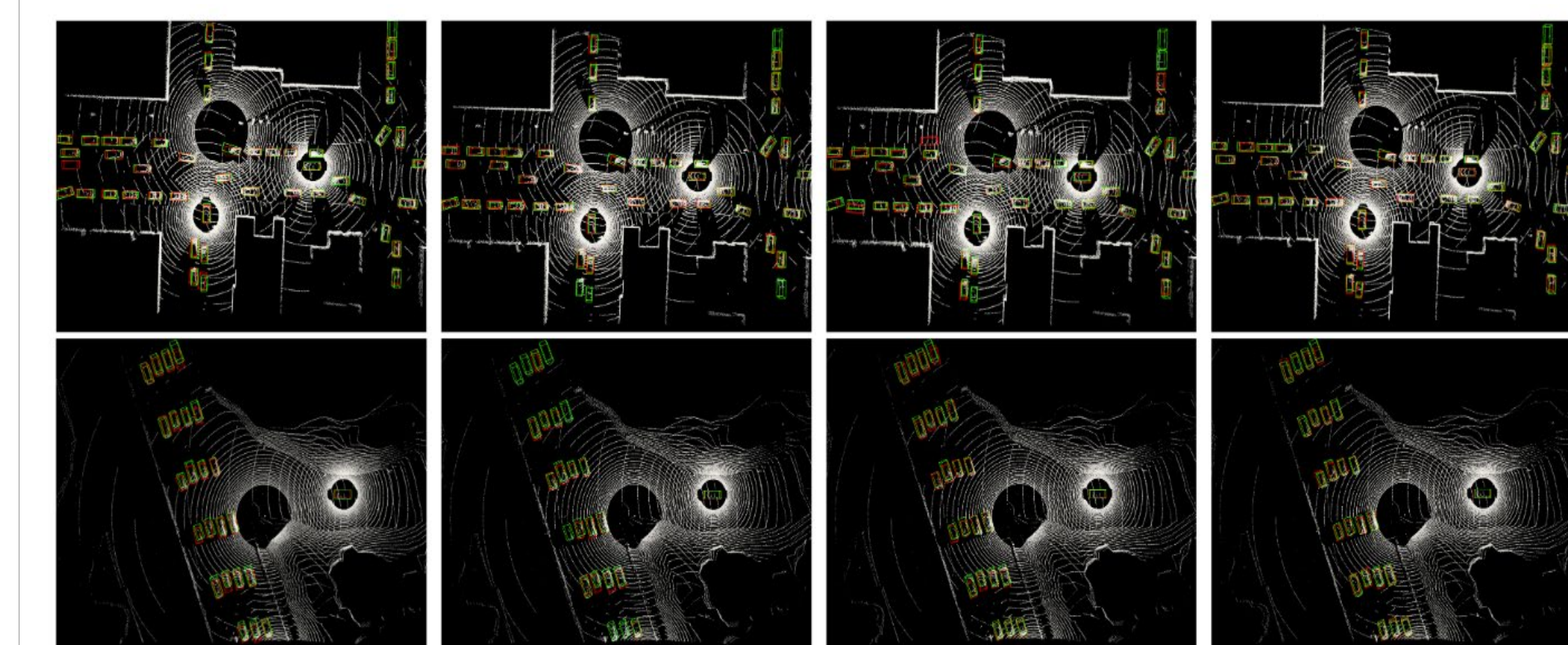
Models	Perfect		Noisy	
	AP0.5	AP0.7	AP0.5	AP0.7
No Fusion	0.606	0.402	0.606	0.402
Late Fusion	0.727	0.620	0.549	0.307
Early Fusion	0.819	0.710	0.720	0.384
F-Cooper [4]	0.840	0.680	0.715	0.469
OPV2V [44]	0.807	0.664	0.709	0.487
V2VNet [39]	0.845	0.677	0.791	0.493
DiscoNet [21]	0.844	0.695	0.798	0.541
V2X-ViT (Ours)	0.882	0.712	0.836	0.614

Ablation study

MSwin	SpAttn	HMSA	DPE	AP0.5 / AP0.7
✓				0.719 / 0.478
✓	✓			0.748 / 0.519
✓	✓	✓		0.786 / 0.548
✓	✓	✓	✓	0.836 / 0.614

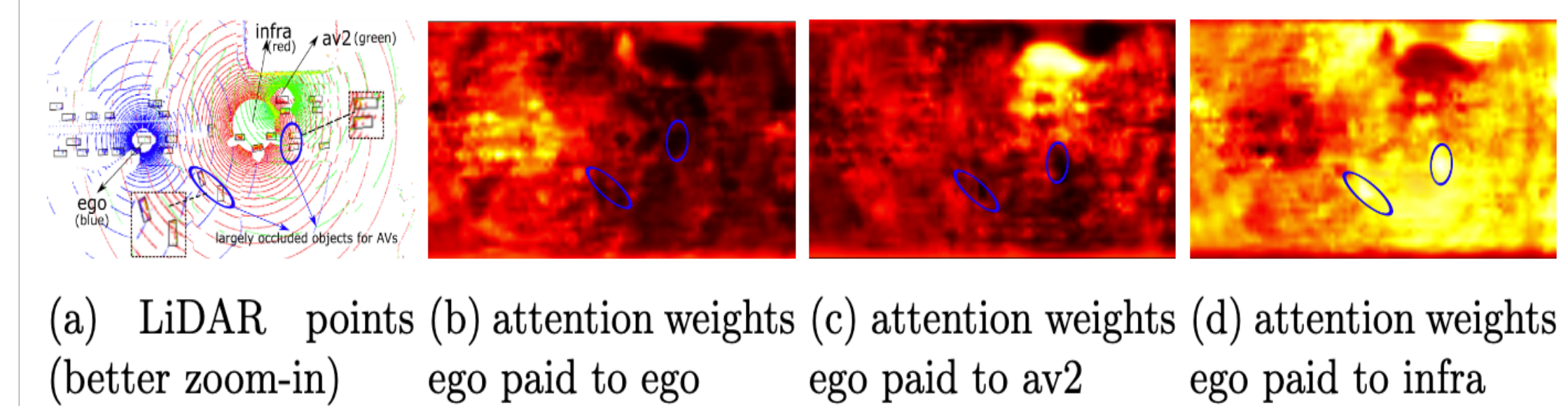


Detection results



(a) OPV2V [44] (b) V2VNet [39] (c) DiscoNet [21] (d) V2X-ViT (ours)

Attention map visualization



(a) LiDAR points (b) attention weights (c) attention weights (better zoom-in) (d) attention weights (ego paid to ego) (e) attention weights (ego paid to av2) (f) attention weights (ego paid to infra)