

LEGO: Learning Edge with Geometry all at Once by Watching Videos

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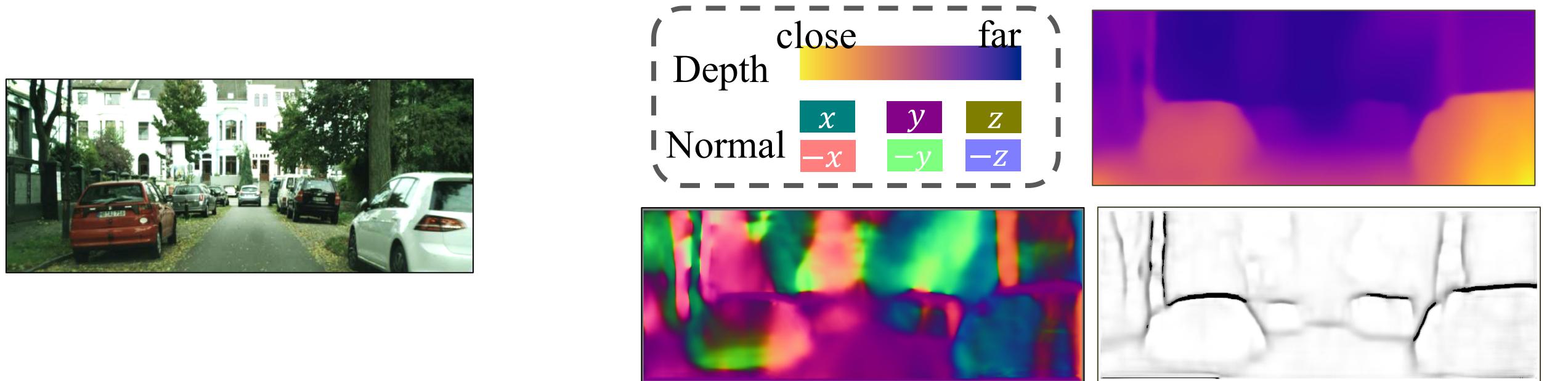
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Task

- Infer depth, surface normal, geometrical from single RGB image



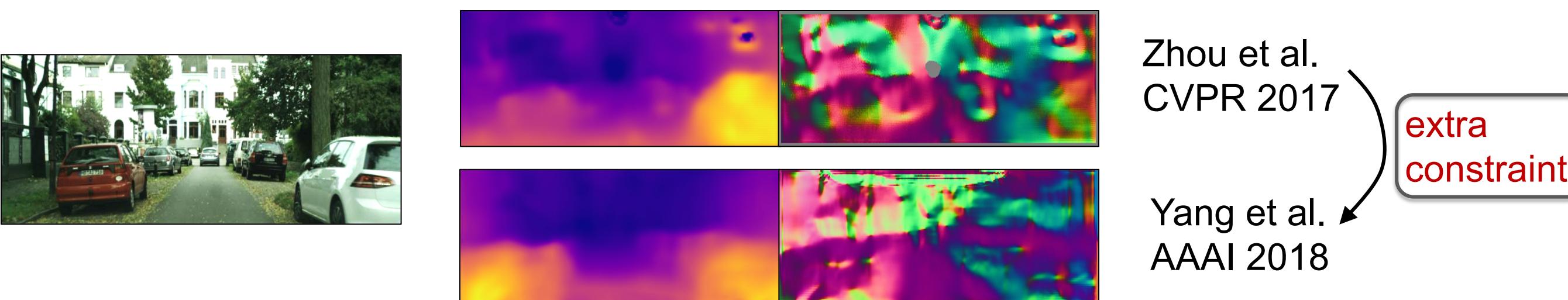
- Model trained on unlabeled monocular videos

Challenge

- Multiple geometric cues are jointly modeled without any ground truth

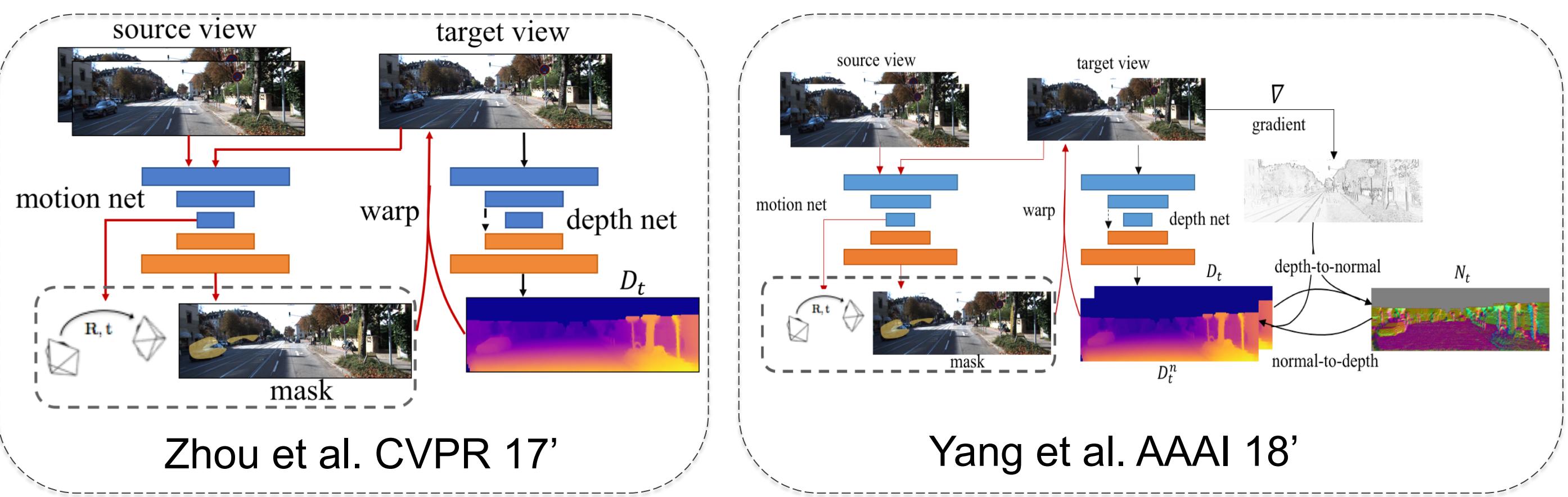
Motivation

- Unsupervised approach is important to avoid expensive annotations
- Joint learning of edges with 3D geometry



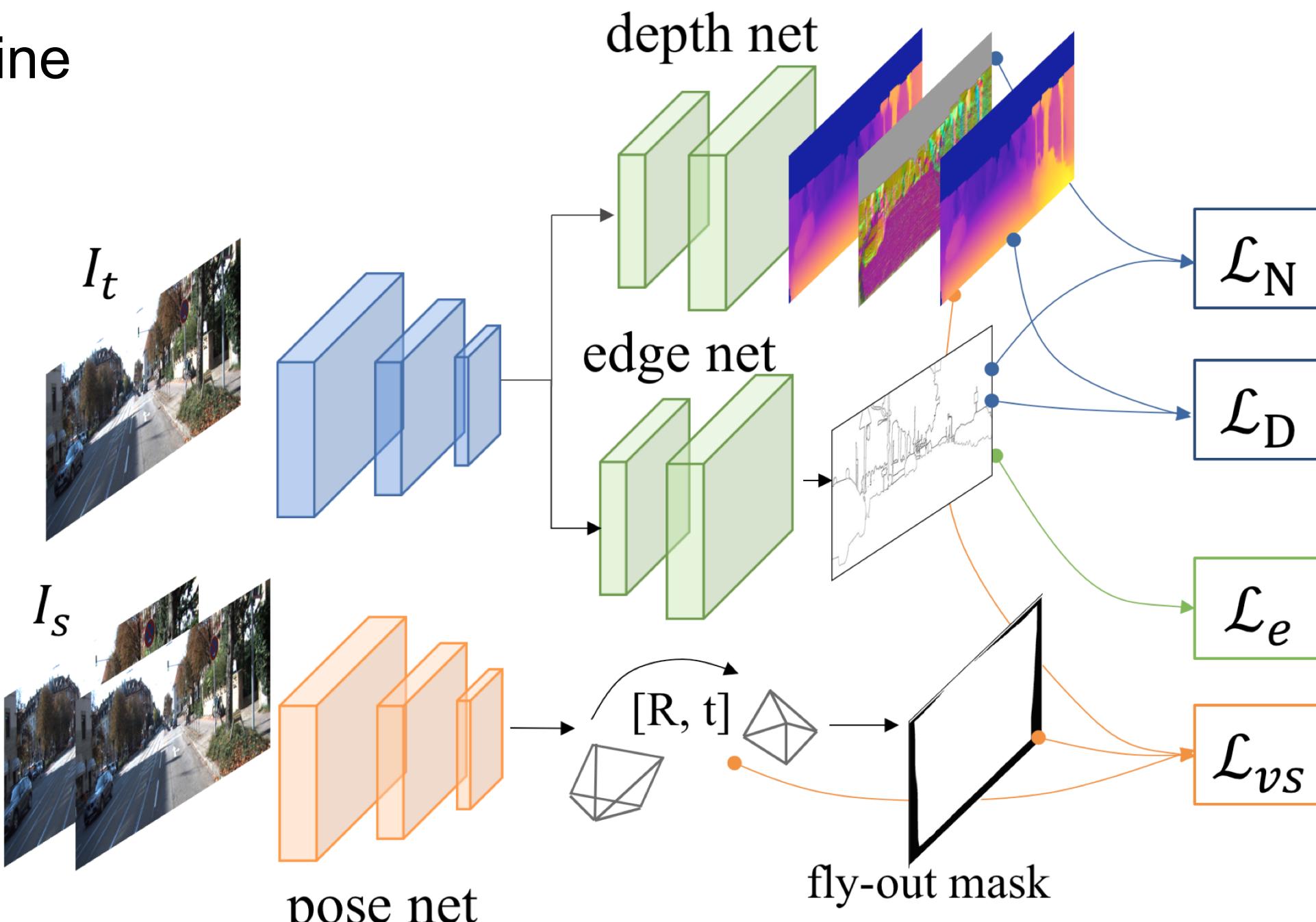
Preliminary

- Zhou et al. proposed to use view synthesis as supervision
- Yang et al. proposed to leverage extra depth-normal consistency



Approach

- Pipeline



- Sub-networks: depth net, edge net, pose net

- Regularization terms:

$$\mathcal{L}_N = \sum_{p_i} \sum_{x,y} \|N_t(p_i) - N_t(p_i(x,y))\|_1 \kappa(p_i, p_i(x,y))$$

Normal directions on same surface are consistent (1st order constraint)

$$\mathcal{L}_D = \sum_{p_i} \sum_x \|\mathbf{g}(p_i, x)\|_1 \kappa(p_i, p_i(x)) \kappa(p_i, p_i(-x))$$

$$\mathbf{g}(p_i, x) = \frac{D_t(p_i(x)) - D_t(p_i)}{\phi(p_i(x)) - \phi(p_i)} - \frac{D_t(p_i) - D_t(p_i(-x))}{\phi(p_i) - \phi(p_i(-x))}$$

Depth changes smoothly in 3D on same surface (2nd order constraint)

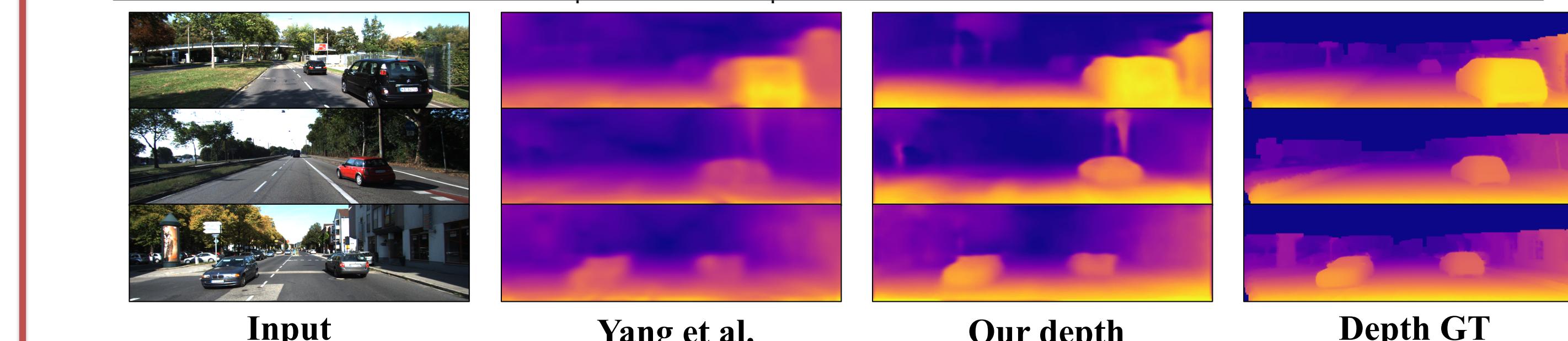
$$\mathcal{L}_e(E) = \sum_{p_i} \|E_t(p_i)\|^2 \quad \text{L2 regularization of edge map}$$

$$\mathcal{L}_{vs}(D, \mathcal{T}) = \sum_s \sum_{p_t \in I_t} |I_t(p_t) - \hat{I}_s(p_t)| \quad \text{View synthesis}$$

Results

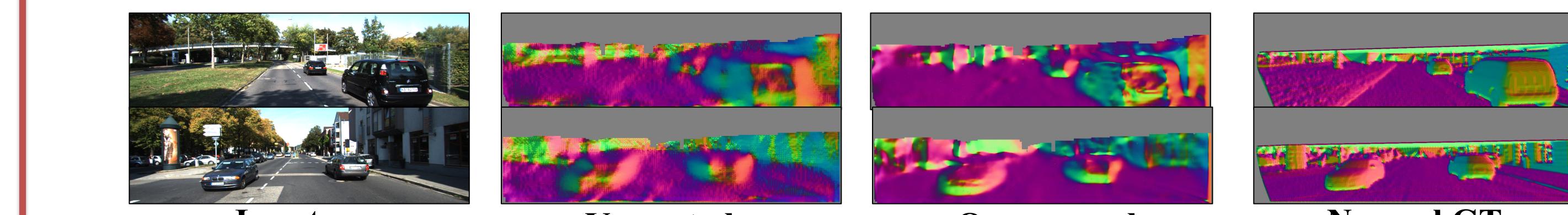
- Depth

Method	Test data	Supervision	Lower the better			
			Abs Rel	Sq Rel	RMSE	RMSE log
Godard <i>et al.</i> [19]		Pose	0.124	1.388	6.125	0.217
Zhou <i>et al.</i> [64]			0.216	2.255	7.422	0.299
Yang <i>et al.</i> [60]			0.165	1.360	6.641	0.267
LEGO (no fly-out)	KITTI split		0.157	1.303	6.223	0.241
LEGO			0.154	1.272	6.012	0.230
LEGO+CS (more data)			0.142	1.237	5.846	0.225

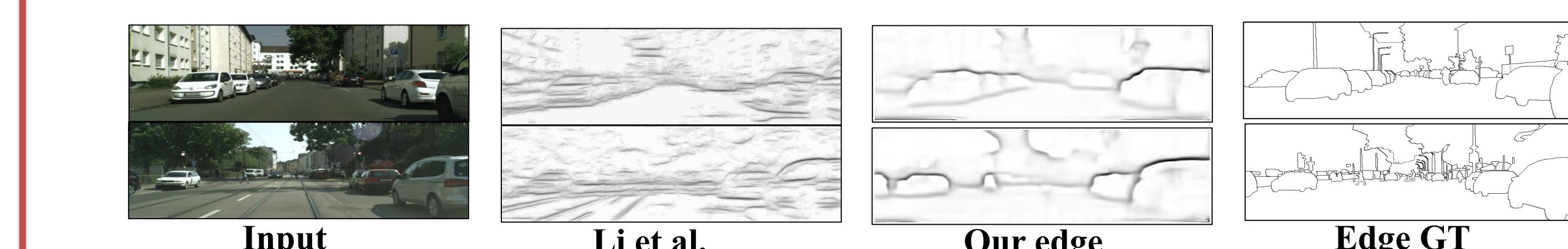


- Normal

Method	Mean	Median	11.25°	22.5°	30°
Godard <i>et al.</i> [19]	39.28	29.37	0.158	0.412	0.496
Yang <i>et al.</i> [60]	47.52	33.98	0.149	0.369	0.473
LEGO	36.13	25.94	0.241	0.473	0.542



- Edge



Methods	ODS	OIS	AP
Li <i>et al.</i> [36]	0.623	0.663	0.557
HED-D	0.457	0.503	0.467
LEGO	0.710	0.731	0.729

