

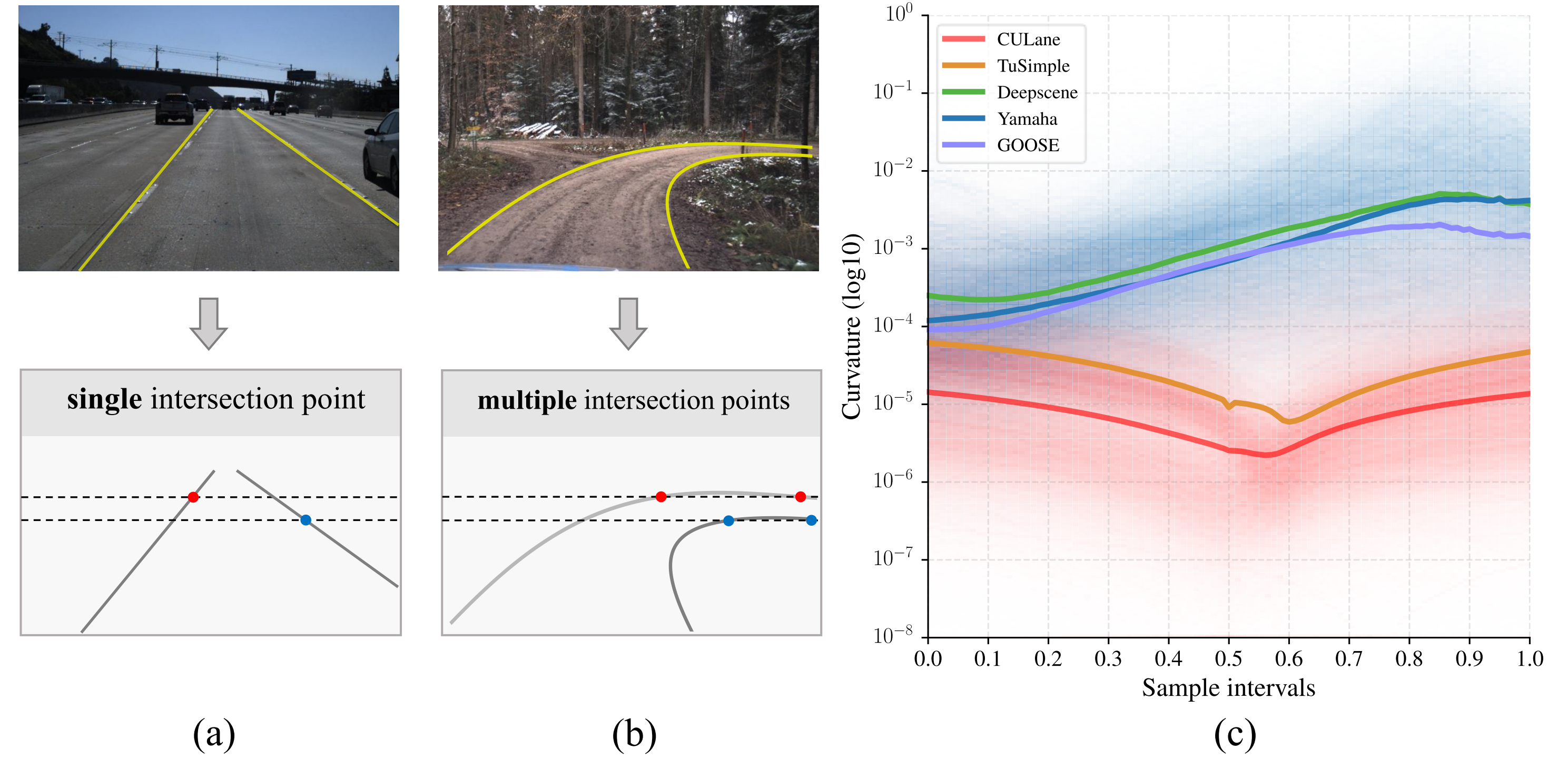
TEAR of the SUNSET: A Benchmark for Road Detection in Off-Road Environment

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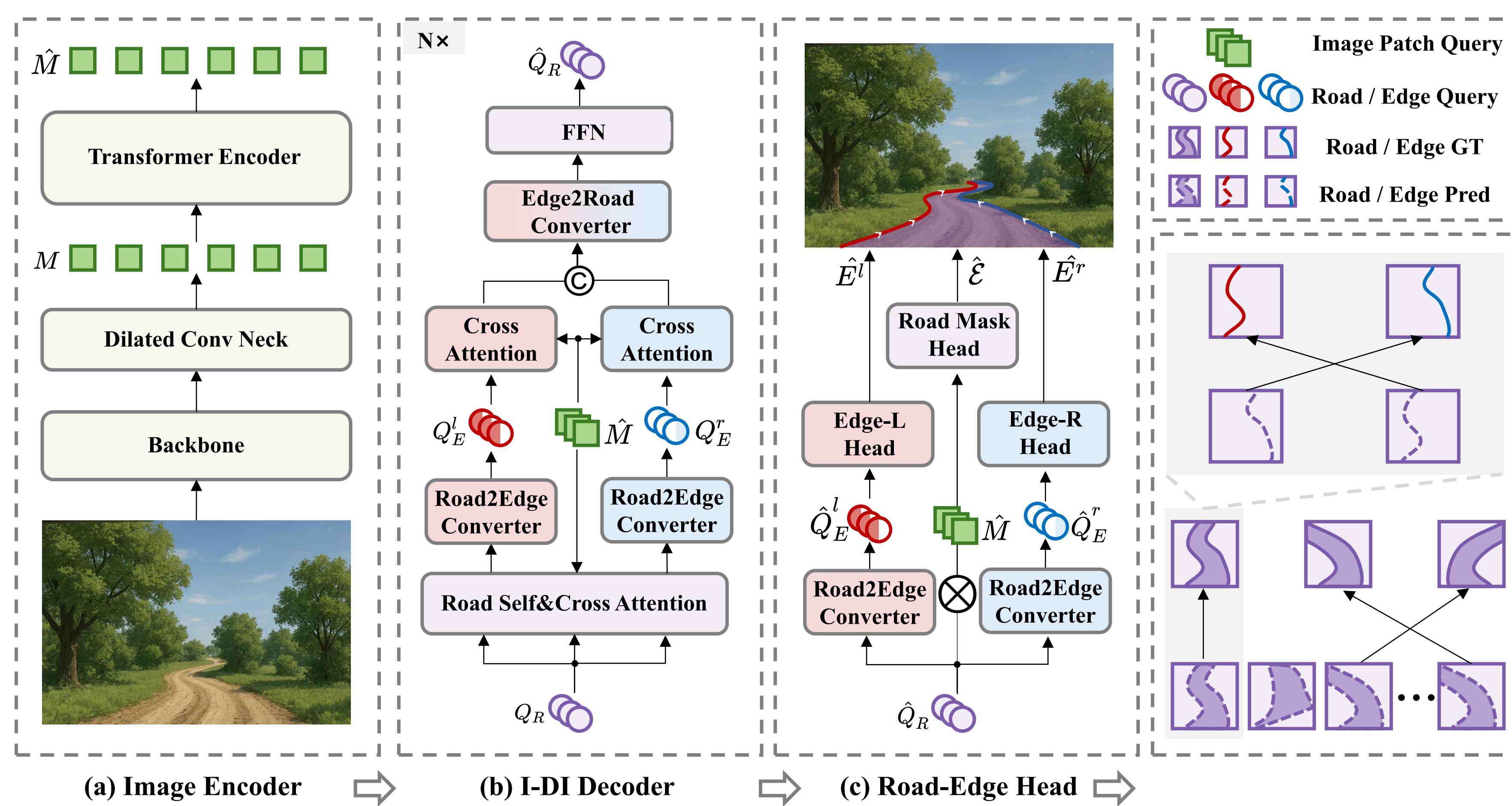
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1. Motivation

(a) and (b) demonstrate significant geometric differences in road morphologies between structured environments and semi-structured environments. For the former, lane markings strictly adhere to a single valued correspondence in the Cartesian coordinate. For the latter, road edge lines with large curvatures exhibit multi-valued correspondences. (c) presents the logarithmic curvature distribution at each line sampling point for traffic environments (CULane and TuSimple) and semi-structured environments (SUNSET) in the form of a heatmap. Solid lines of different colors represent the curvature of each dataset at each sampling point.



2. Methodology



2.1 Task Definition

Line Instance: $E(t) = \sum_{i=0}^n b_{i,n}(t) \mathcal{P}_i, 0 \leq t \leq 1, b_{i,n} = C_n^i t^i (1-t)^{n-i}, i = 0, \dots, n$

Road Instance: $\Omega(u, t) = uE^l(t) + (1-u)E^r(t), t \in [0, 1], u \in [0, 1]$

2.2 Interconvertible Dual-Instance Decoder

$Q_E^l = \mathcal{T}_{RE}^l(Q_R); Q_E^r = \mathcal{T}_{RE}^r(Q_R) \mid \bar{Q}_R = \mathcal{T}_{ER}(\text{Concat}(Q_E^l, Q_E^r))$

2.3 Hierarchical Bipartite Match

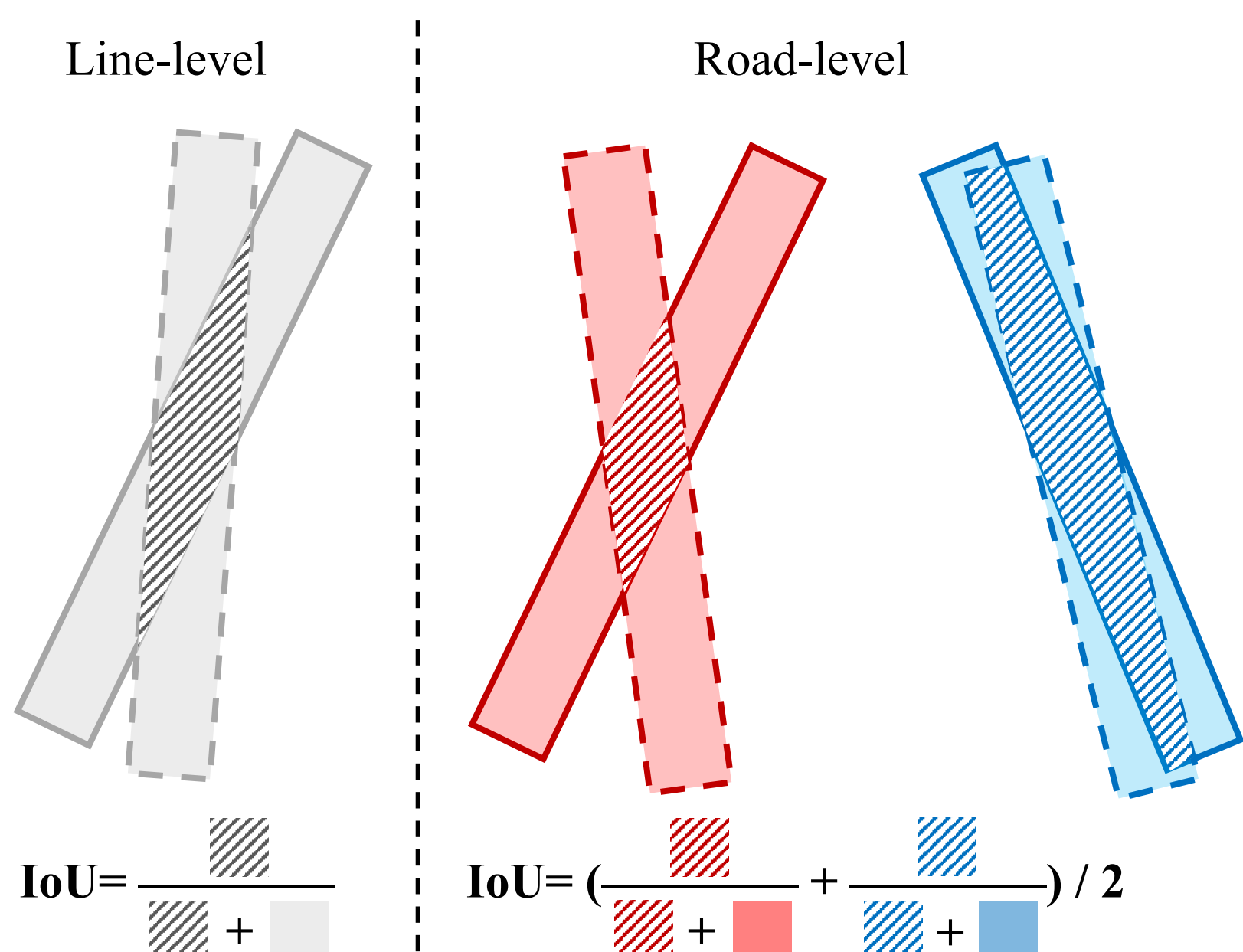
$\hat{\sigma}_R = \arg \min_{\sigma_R \in \mathcal{G}_{N_R}} \sum_{i=1}^{N_R} \mathcal{C}_{Rmatch}(z_i, \hat{z}_{\sigma_R(i)}) \quad \mathcal{G}_2^{(i)} = \left\{ \begin{array}{l} \{y_1^{(i)} \leftrightarrow \hat{y}_1^{\sigma_R(i)}, y_2^{(i)} \leftrightarrow \hat{y}_2^{\sigma_R(i)}\} \\ \{y_1^{(i)} \leftrightarrow \hat{y}_2^{\sigma_R(i)}, y_2^{(i)} \leftrightarrow \hat{y}_1^{\sigma_R(i)}\} \end{array} \right\}$

$\hat{\pi}_E = \arg \min_{\pi_E \in \mathcal{G}_2} \sum_{j=1}^2 \mathcal{C}_{Ematch}(y_j, \hat{y}_{\pi_E(j)})$

$\mathcal{C}_{Rmatch}(z_i, \hat{z}_{\sigma_R(i)})$

3. Experiments

3.1 Metrics



3.2 Main Results

Methods	Publication	Backbone	DeepScene		YCOR		GOOSE		SUNSET		Para(M)	FLOPs(G)
			$F1_E$	$F1_R$	$F1_E$	$F1_R$	$F1_E$	$F1_R$	$F1_E$	$F1_R$		
Segmentation	▼ Mask based											
	SCNN [11]	AAAI'18	VGG16	52.26	-	31.75	-	30.67	-	43.72	-	183.54
	RESA [9]	AAAI'21	R-34	56.29	-	33.89	-	34.41	-	48.51	-	114.01
	▼ Keypoints based											
	GANet [12]	CVPR'22	R-18	58.44	-	37.87	-	34.33	-	50.69	-	30.92
CondLSTR [13]	ICCV'23	R-34	54.10	-	31.92	-	36.41	-	47.18	-	24.40	
Detection	▼ Line-Anchor based											
	SRLane [14]	AAAI'24	R-18	60.75	-	37.57	-	34.67	-	52.16	-	18.97
	LaneATT [15]	CVPR'21	R-18	51.31	-	38.48	-	32.51	-	45.57	-	42.73
	▼ Curve based											
	BezierLaneNet [4]	CVPR'22	R-34	29.53	-	23.60	-	19.36	-	29.03	-	32.31
	LSTR [16]	WACV'21	R-18	48.15	-	31.07	-	30.52	-	40.56	-	38.18
	▼ Ours											
TEAR-S (ResNet18+E2/D2)	-	R-18	68.43	69.57	38.31	33.85	35.89	33.66	59.60	59.81	5.64	19.12
TEAR-M (ResNet34+E3/D3)	-	R-34	70.51	73.07	39.63	40.29	36.64	34.74	61.27	62.56	11.88	38.09

3.3 Visualized Results

