

Sph2Pob: Boosting Object Detection on Spherical Images with Planar Oriented Boxes Methods

Xinyuan Liu^{1,2}, Hang Xu³, Bin Chen^{1,2}, Qiang Zhao¹, Yike Ma¹, Chenggang Yan³, Feng Dai^{1*}

¹Institute of Computing Technology, Chinese Academy of Sciences

²University of Chinese Academy of Sciences ³Hangzhou Dianzi University

952 | Wed 23 August 11:45-12:45 (5 / 6)

Introduction

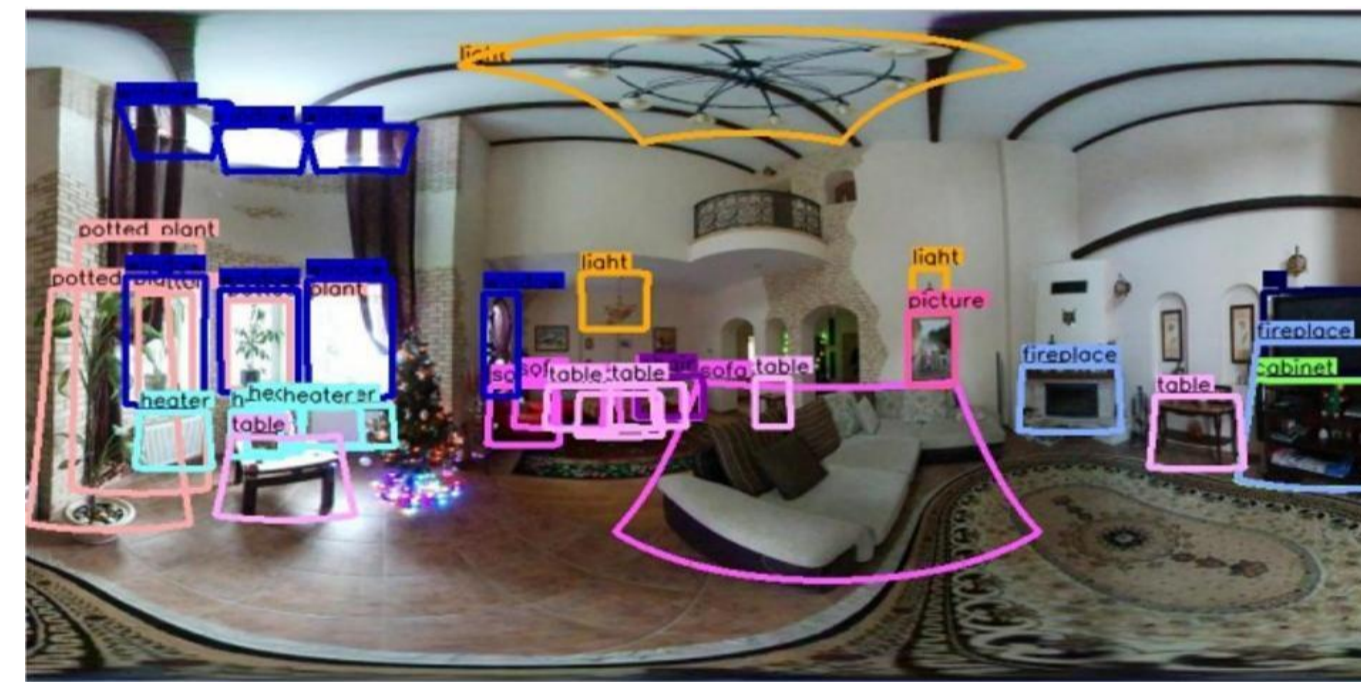
- **Task**
 - detect (locate+classify) objects on panoramic / spherical images.
- **Application**
 - environment perception for robotics and automatic driving.
- **Focus point**
 - IoU calculation and loss design for spherical bounding boxes.
- **Challenges**
 - It's hard to balance differentiability, accuracy and speed in spherical IoU calculation.
 - It's hard to design better spherical loss functions beyond naive L1-Loss.
- **Contributions**
 - convert spherical boxes into planar oriented boxes in pairs, named as Sph2Pob.
 - implement a differentiable, fast, accurate spherical IoU based on Sph2Pob.
 - implement a flexible and extensible spherical loss functions based on Sph2Pob.

Prerequisites

- **Spherical Images**
 - Spherical image is a natural extend (360° view) of common planar image.
 - Spherical image has two display mode, i.e., sphere and ERP.



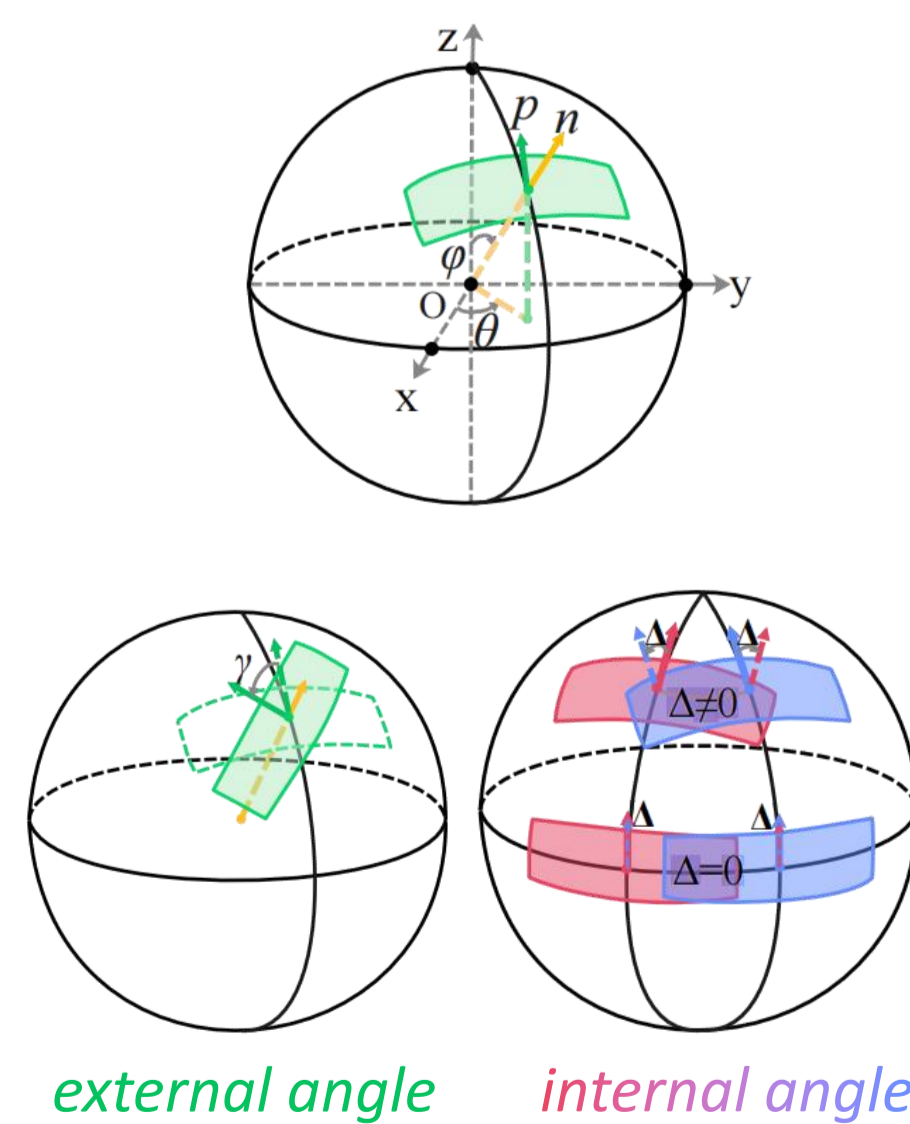
Sphere



Equal Rectangular Projection (ERP)

Spherical Boxes

- Spherical bounding box is defined as $(\theta, \phi, \alpha, \beta, \gamma)$.
- $n(\theta, \phi)$ is the tangent point of the sphere and rectangular tangent plane.
- α and β are the horizontal and vertical fields of view of the spherical bounding box.
- γ is rotated angle around center-axis $p(\theta, \phi)$.
- Apart from γ , another rotated angle Δ coupled with box-pair exists on sphere. [our insight]
- call γ as *external angle*, while Δ as *internal angle*.



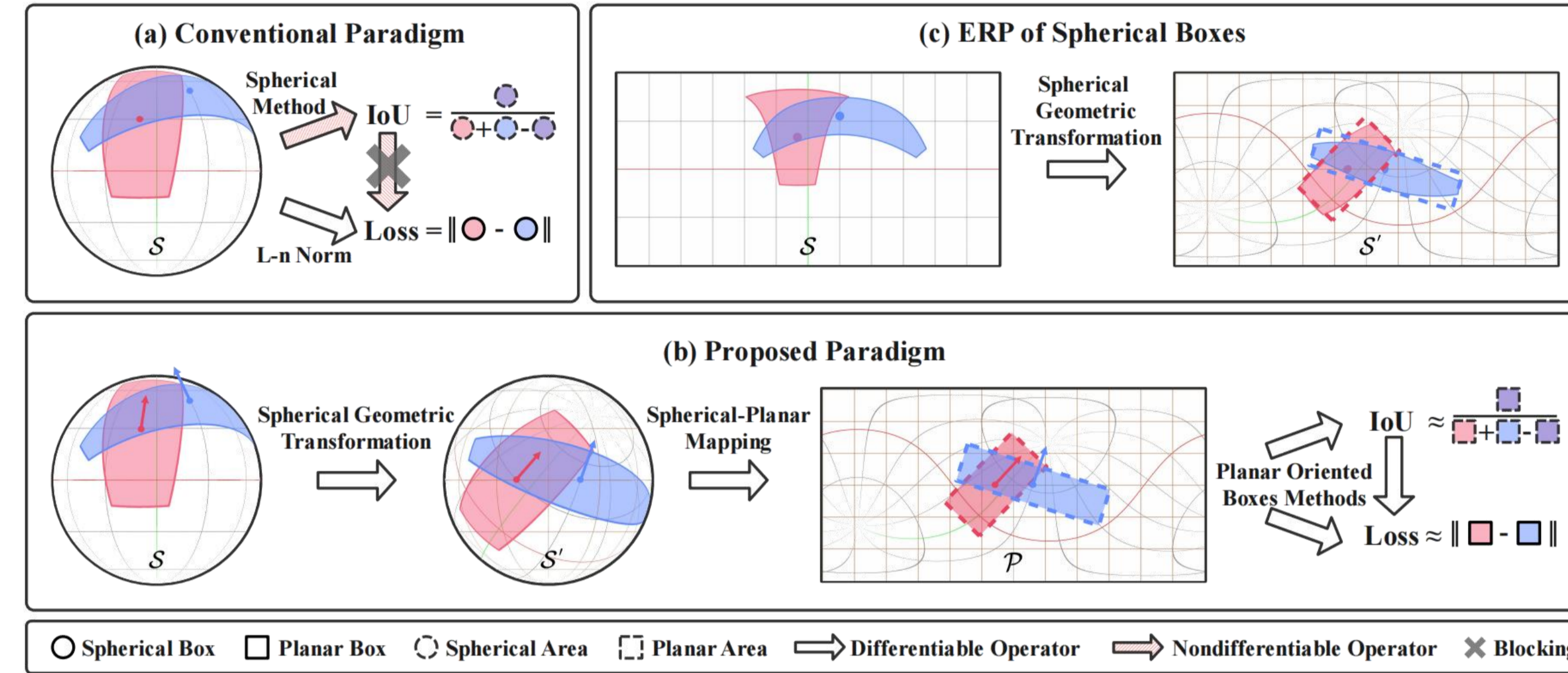
external angle internal angle

Welcome to communicate with me through Email!

luxinyuan21s@ict.ac.cn fdai@ict.ac.cn

Methodology

Overview



Comparison

	Exact Method	Approximate Method	Approximate Method	Approximate Method
Method	Unbiased-IoU	Sph-Io2U	FoV-IoU	Sph2Pob-IoU(Our)
differentiability	☆☆☆☆☆	★★★★★	★★★★★	★★★★★
speed	☆☆☆☆☆	★★★★★	★★★★★	★★★★☆
accuracy	★★★★★	☆☆☆☆☆	★★☆☆☆	★★★★☆

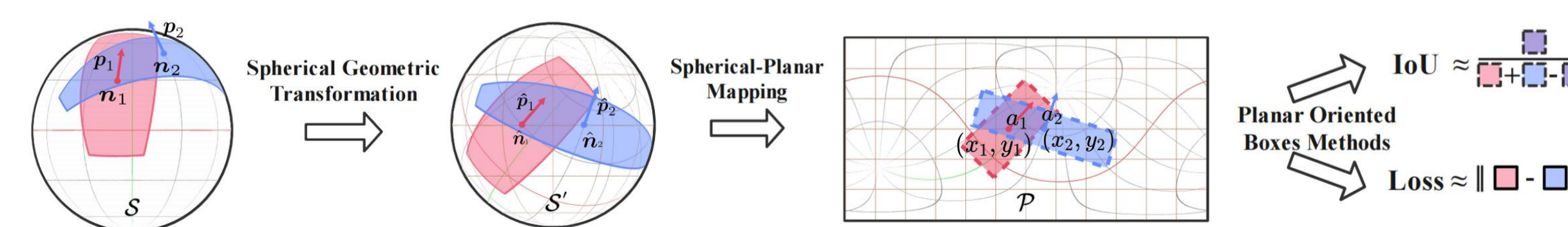
Mathematical Details

1. Compute position and pose.

$$\begin{aligned} \hat{n} &= n(\theta, \phi) = [nx, ny, nz]^T \\ &= [\sin(\phi) \cos(\theta), \sin(\phi) \sin(\theta), \cos(\phi)]^T \\ \hat{p} &= p(\theta, \phi) = \frac{\partial n(\theta, \phi)}{\partial \phi} = [px, py, pz]^T \\ &= [\cos(\phi) \cos(\theta), \cos(\phi) \sin(\theta), -\sin(\phi)]^T \end{aligned}$$

2. Construct spherical transformation.

$$R = [v_x, v_y, v_z]^T = \left[\frac{n_1 + n_2}{\|n_1 + n_2\|}, \frac{n_1 - n_2}{\|n_1 - n_2\|}, v_x \times v_y \right]^T$$

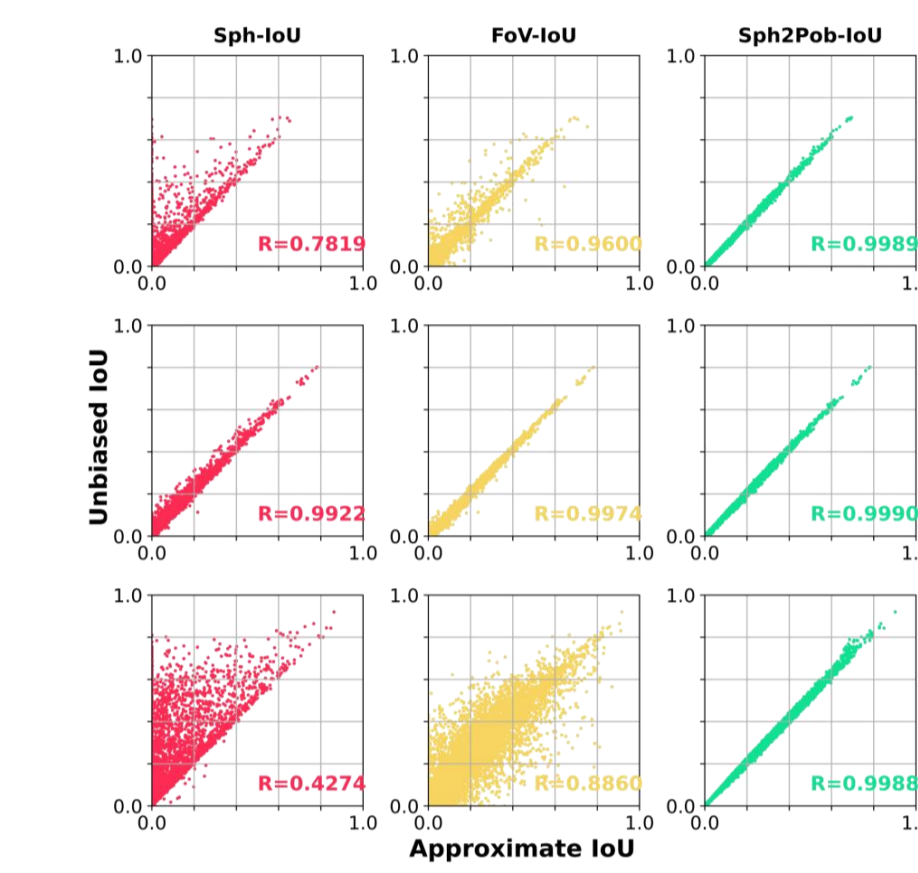


$$\star \text{IoU \& Loss} \star \text{IoU}^S(B_1^S, B_2^S) \approx \text{IoU}^P(\text{Sph2Pob}(B_1^S, B_2^S)) \quad \text{Loss}^S(B_1^S, B_2^S) \approx \text{Loss}^P(\text{Sph2Pob}(B_1^S, B_2^S))$$

Experiments

Main Results

- Scatter of different IoU.



- Comprehensive comparison of box transform methods.

Method	Consistency			Time-cost		Detection		
	R _{all} ↑	R _{low} ↑	R _{high} ↑	T _{cpu} ↓	T _{cuda} ↓	AP↑	AP ₅₀ ↑	AP ₇₅ ↑
Sph	0.7819	0.9922	0.4274	0.0364	0.0033	10.7	24.3	7.8
Fov	0.9600	0.9974	0.8860	0.0372	0.0034	10.9	25.0	7.9
Sph2Pob	0.9989	0.9990	0.9988	2.2275	0.0096	11.5	25.7	8.2
Unbiased	1.0000	1.0000	1.0000	46.4417	-	-	-	-

- Ablation studies about edge & angle calculation.

Edge	Error↓(mean±std)		R↑	Angle		Error↓(mean±std)		R↑
	original	equator		original	project			
arc	0.0016±0.0042	0.9989		0.0025±0.0086	0.9946			
chord	0.0023±0.0063	0.9974		0.0016±0.0042	0.9989			
tangent	0.0086±0.0192	0.9681		0.0017±0.0043	0.9987			

Evaluations

- Evaluation on different Loss.

Loss	360-Indoor			PANDORA		
	AP↑	AP ₅₀ ↑	AP ₇₅ ↑	AP↑	AP ₅₀ ↑	AP ₇₅ ↑
L1	10.2	23.0	7.8	10.3	24.3	6.6
L1 [†]	9.9	21.9	7.7	10.1	23.7	6.8
GWD [†] [Yang et al., 2021b]	6.8	14.5	5.6	5.9	12.3	5.0
KLD [†] [Yang et al., 2021c]	9.5	21.5	6.8	10.3	23.5	7.1
KFloU [†] [Yang et al., 2022b]	8.5	19.7	6.2	9.6	23.2	5.6
IoU [†] [Yu et al., 2016]	9.8	22.1	6.8	10.4	24.8	6.9
CloU [†] [Hosang et al., 2019]	10.5	23.9	7.8	10.3	24.7	6.8
DiIoU [†] [Zheng et al., 2020]	11.0	24.6	8.2	10.4	24.8	7.0
CloU [†] [Zheng et al., 2021]	11.5	25.7	8.2	10.5	25.3	7.0

- Evaluation on different detectors.

Detector	Loss	360-Indoor			PANDORA		
		AP↑	AP ₅₀ ↑	AP ₇₅ ↑	AP↑	AP ₅₀ ↑	AP ₇₅ ↑
Faster R-CNN	L1	12.5	28.1	9.1	11.0	27.8	6.2
	CloU [†]	12.9	29.1	9.4	11.3	28.6	7.1
SSD	L1	10.8	27.6	6.3	9.5	25.8	4.6
	CloU [†]	12.0	28.7	8.0	10.5	26.9	6.0
FCOS	L1	8.8	20.2	6.7	7.7	19.7	4.4
	CloU [†]	9.2	21.0	7.0	8.8	21.2	5.6

Visualization

