

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

Xinyuan Liu<sup>1,2</sup>, Hang Xu<sup>3</sup>, Bin Chen<sup>1,2</sup>, Qiang Zhao<sup>1</sup>, Yike Ma<sup>1</sup>, Chenggang Yan<sup>3</sup>, Feng Dai<sup>1\*</sup>

<sup>1</sup>Institute of Computing Technology, Chinese Academy of Sciences, Beijing, China

<sup>2</sup>University of Chinese Academy of Sciences, Beijing, China

<sup>3</sup>Hangzhou Dianzi University, Hangzhou, China

Macao, S.A.R. August 19-25, 2023.



IJCAI/2023 MACAO

# Overview



## ➤ Topic

- **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

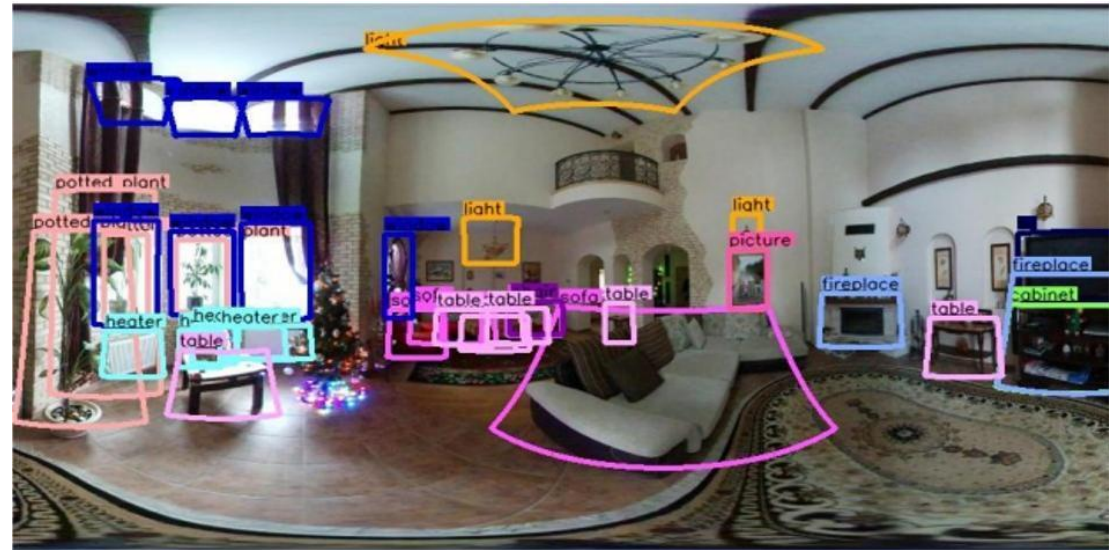
- convert spherical boxes into planar oriented boxes in pairs, named as Sph2Pob.
- implement a differentiable, fast, accurate spherical IoU based on Sph2Pob.
- implement flexible and extensible spherical loss functions based on Sph2Pob.

# Spherical/Panoramic Image

- Spherical/Panoramic image is a natural extend of comon planar image.
- It has the whole 360° view with richer information and higher practice value.
- Spherical image has two display mode, i.e., sphere and ERP.



Sphere



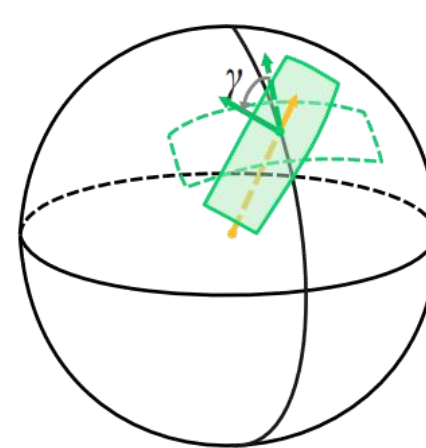
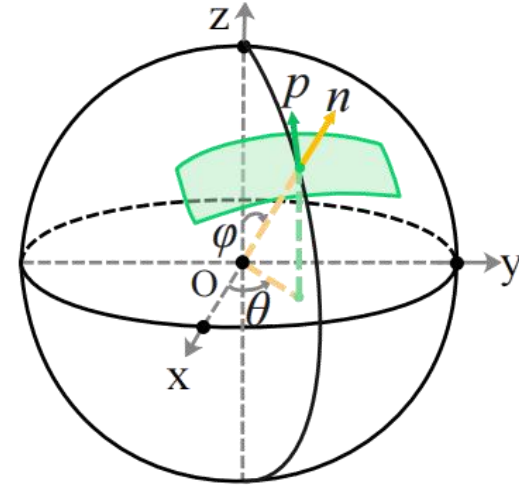
Equal Rectangular Projection (ERP)



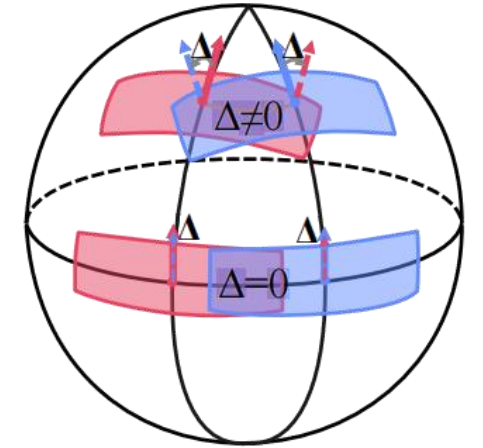
# Spherical Bounding Box



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



external angle



internal angle

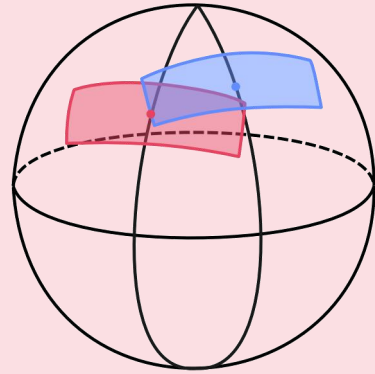
# Overlap(IoU) Calculation

**Exact Method**

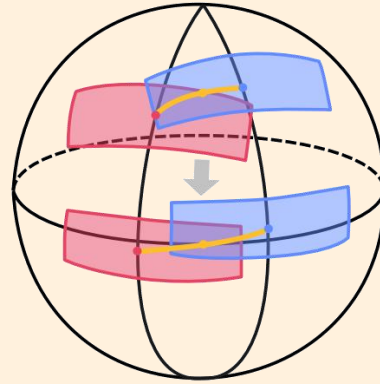
**Approximate Method**

**Approximate Method**

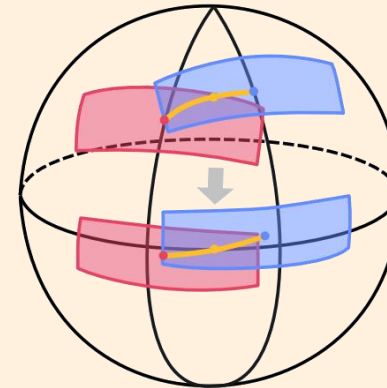
**Method**



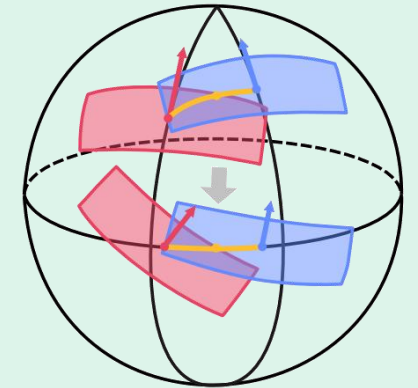
Unbiased-IoU



Sph-IoU



FoV-IoU



Sph2Pob-IoU(Our)

**differentiability**

☆☆☆☆☆

★★★★★

★★★★★

★★★★★

**speed**

☆☆☆☆☆

★★★★★

★★★★★

★★★★☆

**accuracy**

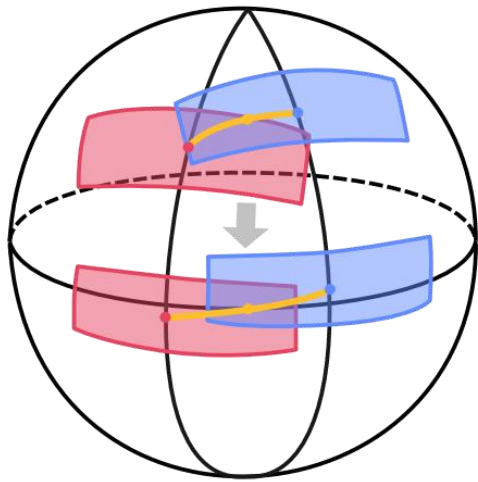
★★★★★

☆☆☆☆☆

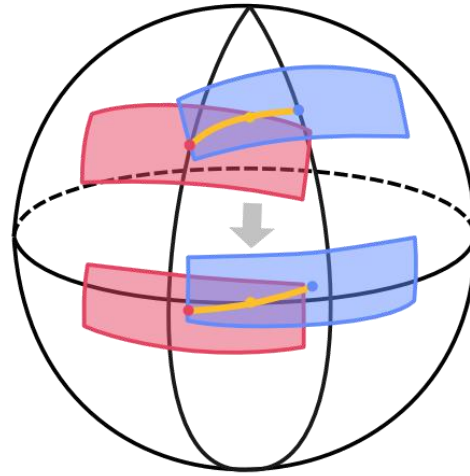
☆☆☆☆☆

★★★★☆

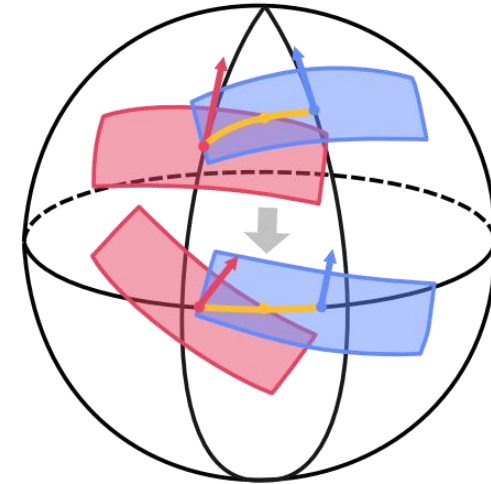
# Overlap(IoU) Calculation



Sph-IoU



FoV-IoU



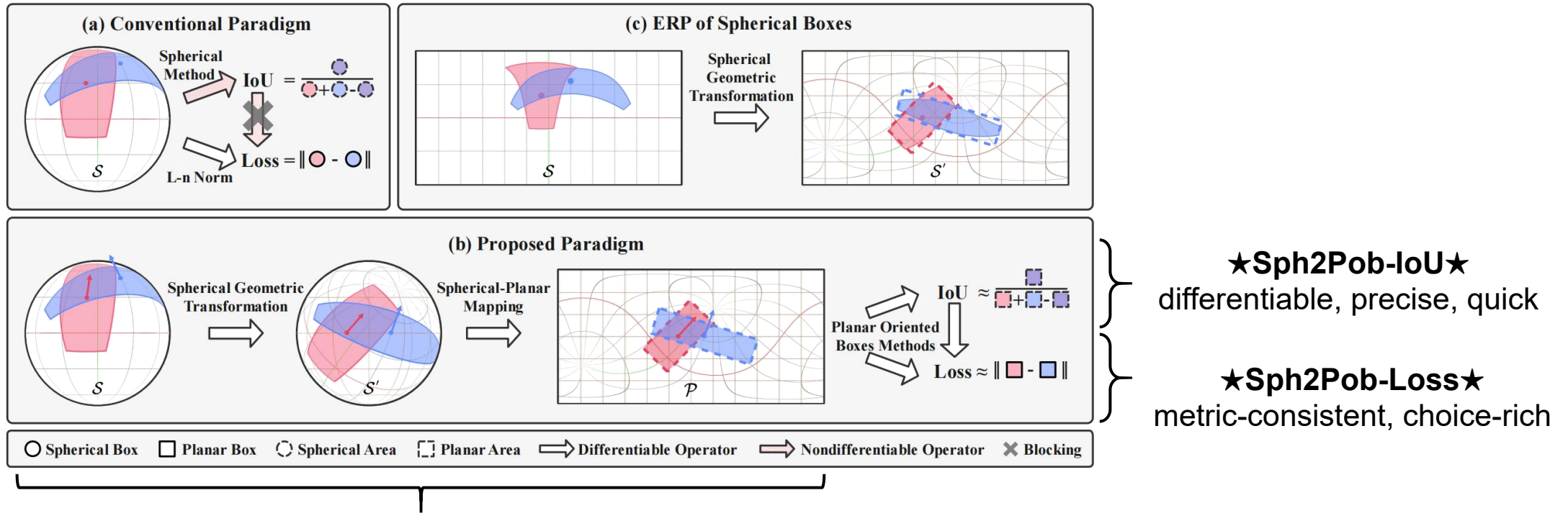
Sph2Pob-IoU(Our)

- ✓ Try to move boxes to the equator, which is the line with no distortion on ERP-images.
- ✗ Fail to ensure the boxes after moving on the equator.
- ✗ Ignore the internal angle, so relative poses change.

- ✓ Continue to move boxes to the equator.
- ✓ Ensure the boxes on the equator.
- ✓ Preserve the internal angle and relative pose.

# Sph2Pob

- **Spherical Geometric Transformation** can move boxes without any info loss.
- **Spherical-Planar Mapping** is also used to calculate more general loss beyond IoU.



**★Sph2Pob★** : well approximated box transform



# Mathematical Details

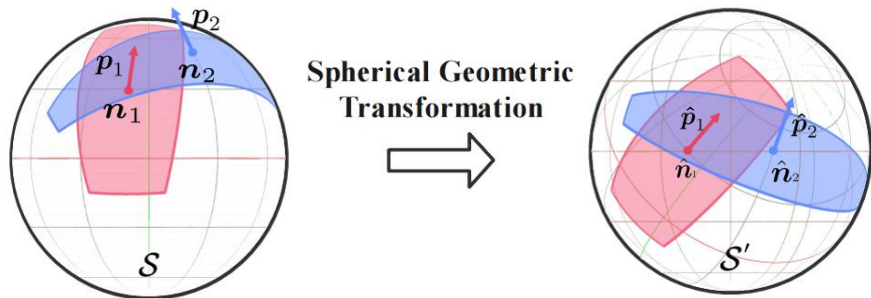


1. Compute position and pose.

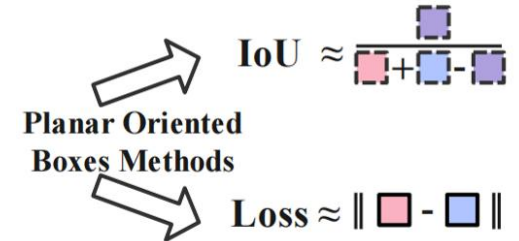
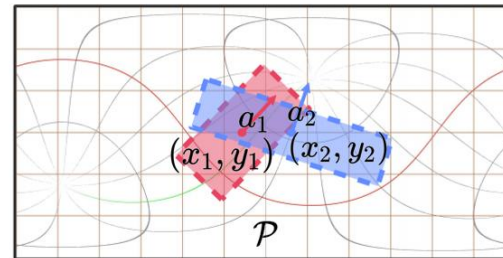
$$\begin{aligned} \mathbf{n} &= \mathbf{n}(\theta, \phi) = [nx, ny, nz]^\top \\ &= [\sin(\phi) \cos(\theta), \sin(\phi) \sin(\theta), \cos(\phi)]^\top \\ \mathbf{p} &= \mathbf{p}(\theta, \phi) = \frac{\partial \mathbf{n}(\theta, \phi)}{\partial \phi} = [px, py, pz]^\top \\ &= [\cos(\phi) \cos(\theta),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

2. Construct spherical transformation.

$$\mathbf{R} = [\mathbf{v}_x, \mathbf{v}_y, \mathbf{v}_z]^\top = \left[ \frac{\mathbf{n}_1 + \mathbf{n}_2}{\|\mathbf{n}_1 + \mathbf{n}_2\|}, \frac{\mathbf{n}_1 - \mathbf{n}_2}{\|\mathbf{n}_1 - \mathbf{n}_2\|}, \mathbf{v}_x \times \mathbf{v}_y \right]^\top$$



Spherical-Planar Mapping



3. Transform position and pose.

$$\hat{\mathbf{n}}(\hat{\theta}, \hat{\phi}) = \mathbf{R} \mathbf{n}(\theta, \phi) \quad [\hat{\mathbf{p}}_1, \hat{\mathbf{p}}_2] = \mathbf{R} [\mathbf{p}_1, \mathbf{p}_2]$$

4. Compute *Internal Angle*.

$$\Delta = \Delta_1 + \Delta_2 = \arccos(\hat{\mathbf{p}}_1 \cdot \hat{\mathbf{p}}_{ref}) + \arccos(\hat{\mathbf{p}}_2 \cdot \hat{\mathbf{p}}_{ref})$$

5. Map spherical boxes to planar boxes.

$$\begin{aligned} \mathcal{B}_i^{\mathcal{P}} &= (x_i, y_i, w_i, h_i, a_i) = (\hat{\theta}_i, \hat{\phi}_i, \hat{\alpha}_i, \hat{\beta}_i, \hat{\Delta}_i) \\ a_i &= \Delta_i + \gamma_i, i = 1, 2 \end{aligned}$$

★ IoU & Loss ★

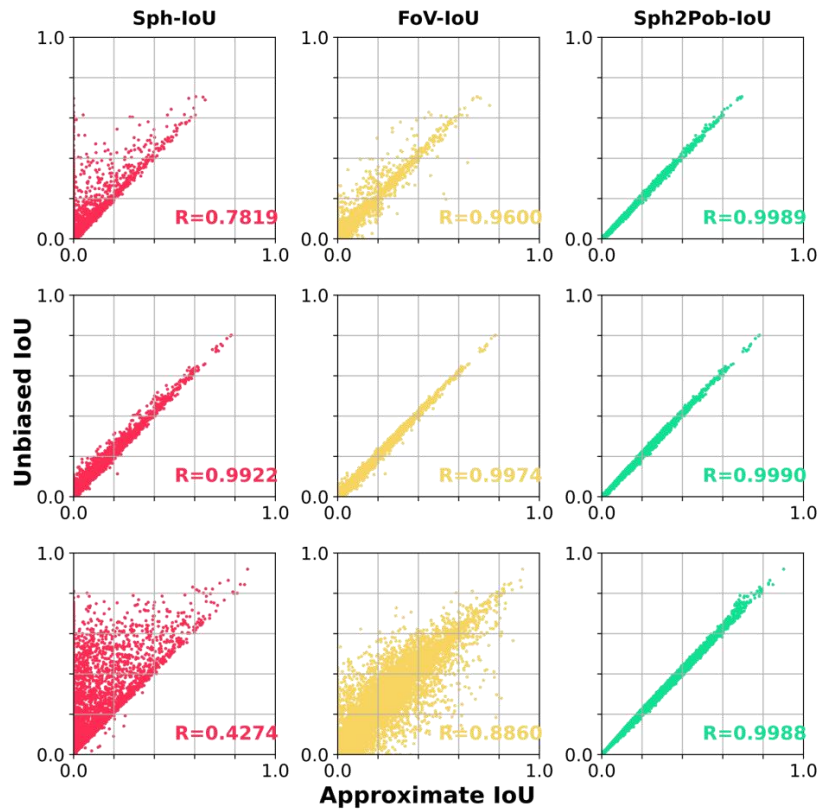
$$\text{IoU}^{\mathcal{S}}(\mathbf{B}_1^{\mathcal{S}}, \mathbf{B}_2^{\mathcal{S}}) \approx \text{IoU}^{\mathcal{P}}(\text{Sph2Pob}(\mathbf{B}_1^{\mathcal{S}}, \mathbf{B}_2^{\mathcal{S}})) \quad \text{Loss}^{\mathcal{S}}(\mathbf{B}_1^{\mathcal{S}}, \mathbf{B}_2^{\mathcal{S}}) \approx \text{Loss}^{\mathcal{P}}(\text{Sph2Pob}(\mathbf{B}_1^{\mathcal{S}}, \mathbf{B}_2^{\mathcal{S}}))$$



# Experiments

- Scatter of different IoU.

- Comprehensive comparison of box transform methods.



Method	Consistency			Time-cost		Detection		
	$R_{all} \uparrow$	$R_{low} \uparrow$	$R_{high} \uparrow$	$T_{cpu} \downarrow$	$T_{cuda} \downarrow$	$AP \uparrow$	$AP_{50} \uparrow$	$AP_{75} \uparrow$
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 $\downarrow$ (mean $\pm$ std)		$R \uparrow$	Angle	Error $\downarrow$ (mean $\pm$ std)		$R \uparrow$
arc	0.0016 $\pm$ 0.0042	0.9989	original	0.0025 $\pm$ 0.0086	0.9946		
chord	0.0023 $\pm$ 0.0063	0.9974	equator	0.0016 $\pm$ 0.0042	0.9989		
tangent	0.0086 $\pm$ 0.0192	0.9681	project	0.0017 $\pm$ 0.0043	0.9987		

# Evaluations

- Evaluation on different Loss.

Loss	360-Indoor			PANDORA		
	AP $\uparrow$	AP $_{50}$ $\uparrow$	AP $_{75}$ $\uparrow$	AP $\uparrow$	AP $_{50}$ $\uparrow$	AP $_{75}$ $\uparrow$
L1	10.2	23.0	7.8	10.3	24.3	6.6
L1 $^\dagger$	9.9	21.9	7.7	10.1	23.7	6.8
GWD $^\dagger$ [Yang <i>et al.</i> , 2021b]	6.8	14.5	5.6	5.9	12.3	5.0
KLD $^\dagger$ [Yang <i>et al.</i> , 2021c]	9.5	21.5	6.8	10.3	23.5	7.1
KFIoU $^\dagger$ [Yang <i>et al.</i> , 2022b]	8.5	19.7	6.2	9.6	23.2	5.6
IoU $^\dagger$ [Yu <i>et al.</i> , 2016]	9.8	22.1	6.8	10.4	24.8	6.9
GIoU $^\dagger$ [Rezatofghi <i>et al.</i> , 2019]	10.5	23.9	7.8	10.3	24.7	6.8
DIoU $^\dagger$ [Zheng <i>et al.</i> , 2020]	11.0	24.6	8.2	10.4	24.8	7.0
CIoU $^\dagger$ [Zheng <i>et al.</i> , 2021]	11.5	25.7	8.2	10.5	25.3	7.0

- Evaluation on different detectors.

Detector	Loss	360-Indoor			PANDORA		
		AP $\uparrow$	AP $_{50}$ $\uparrow$	AP $_{75}$ $\uparrow$	AP $\uparrow$	AP $_{50}$ $\uparrow$	AP $_{75}$ $\uparrow$
Faster R-CNN	L1	12.5	28.1	9.1	11.0	27.8	6.2
	CIoU $^\dagger$	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
	CIoU $^\dagger$	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
	CIoU $^\dagger$	9.2	21.0	7.0	8.8	21.2	5.6

- Evaluation on different components, including Label Assignment, Loss, NMS.

Label Assignment	Loss	NMS	360-Indoor			PANDORA		
			AP $\uparrow$	AP $_{50}$ $\uparrow$	AP $_{75}$ $\uparrow$	AP $\uparrow$	AP $_{50}$ $\uparrow$	AP $_{75}$ $\uparrow$
			9.8	22.2	7.0	10.4	23.8	6.9
✓			10.2	23.0	7.8	10.3	24.3	6.6
	✓		11.0	25.4	7.8	10.6	24.5	6.9
		✓	9.8	22.1	6.8	10.4	23.9	6.9
✓	✓		11.5	25.7	8.2	10.5	25.3	7.0
✓	✓	✓	11.6	26.1	8.4	10.6	25.7	7.1



# Visualized Comparisons (360Indoor)

- Some easily confused bed/sofa/table-like objects are labeled as bed.
- With our CloU<sup>†</sup>, predicted bboxes are more compact even though category-wrong.



Ground Truth



(Ours) CloU<sup>†</sup>

# Visualized Comparisons (PANDORA)

- The cabinet is rotated with some angle, and lights are tiny.
- With L1, predicted box of the cabinet is coarse with wrong angle, and lights are missing.



Ground Truth



L1



# Visualized Comparisons (PANDORA)

- The cabinet is rotated with some angle, and lights are tiny.
- With our CloU<sup>†</sup>, the cabinet is more tight with right angle, and lights are caught.



Ground Truth



(Ours) CloU<sup>†</sup>

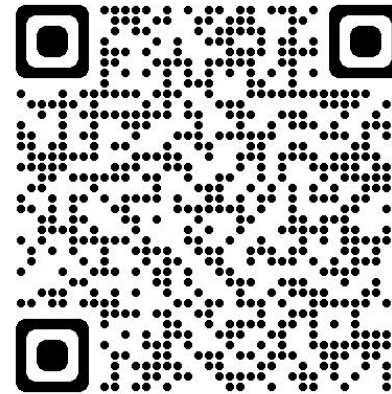


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# Thank You !



- **Paper:** Sph2Pob: Boosting Object Detection on Spherical Images with Planar Oriented Boxes Methods
- **Codes:** <https://github.com/AntXinyuan/sph2pob>
- **Contact:**
  - Xinyuan Liu: [liuxinyuan21s@ict.ac.cn](mailto:liuxinyuan21s@ict.ac.cn)
  - Hang Xu: [hxu@hdu.edu.cn](mailto:hxu@hdu.edu.cn)
  - Feng Dai: [fdai@ict.ac.cn](mailto:fdai@ict.ac.cn)



Homepage

**Tips:** All codes/posters/slides can be available on our github page.