

Loss function: distance to the side of an orthohedron (object)

March 8, 2025

In our 3D object detection problem, the cuboid object is modeled by a 9-dimensional parameter vector

$$\mathbf{v} = \begin{bmatrix} x \\ y \\ z \\ \alpha \\ \beta \\ \gamma \\ w \\ d \\ h \end{bmatrix}, \quad (1)$$

where x, y, z define the position of the object in the room frame, α, β, γ define the rotation of the object in the room frame, and w, d, h define the width, depth, and height of the table respectively. The objective is to define a differentiable distance from a point p with coordinates in the object frame, to one side (top) of the object.

Let the width of the orthohedron be w , the depth d and the height h . We define the transformation from the center of the object to the top side with an Pose3 object:

$$T_{t2s} = \text{Pose3}\left(\mathbf{R} = \mathbf{I}, \mathbf{t} = \begin{bmatrix} 0 \\ 0 \\ \frac{h}{2} \end{bmatrix}\right), \quad (2)$$

and the transformed point ps ,

$$ps = \text{Pose3}::\text{transformTo}(p, H_{ps}(3 \times 6), H_{pt}(3 \times 3)) = \begin{bmatrix} x \\ y \\ z \end{bmatrix} \in \mathbb{R}^3, \quad (3)$$

The method provides the Jacobians with respect to the pose parameters and the input point p .

We follow the same convention for all the sides of the orthohedron, where the normal vector points outwards in the positive Z direction, the $X+$ direction is the width, and the $Y+$ direction is the depth.

Distance computation

Let the top side of the object defined in the side coordinate system be given by the plane,

$$z = 0, \tag{4}$$

with a rectangular boundary given by

$$\mathcal{R} = \{(x, y, 0) \mid -L_x \leq x \leq L_x, -L_y \leq y \leq L_y\}. \tag{5}$$

where $L_x = \frac{w}{2}$ and $L_y = \frac{d}{2}$.

The orthogonal projection of Point ps onto the plane is

$$p_{\text{proj}} = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix}. \tag{6}$$

We now analyse the two possible cases for the projection of p onto the face of the object: falling inside and falling outside the rectangle \mathcal{R} .

Case 1: Projection Inside the Face

If the projection lies inside the rectangle, i.e.,

$$|x| \leq L_x \quad \text{and} \quad |y| \leq L_y, \tag{7}$$

then the closest point on the face is p_{proj} itself and the (squared) distance from p to the face is simply the squared perpendicular distance:

$$D^2(p) = z^2. \tag{8}$$

Case 2: Projection Outside the Face

If the projection lies outside \mathcal{R} , then the closest point q on the rectangle is the one that minimizes the Euclidean distance to p_{proj} . Define the distances along the x and y directions by

$$d_x = \max\{0, |x| - L_x\}, \quad d_y = \max\{0, |y| - L_y\}. \quad (9)$$

Then, the squared distance from p to the face is given by

$$D^2(p) = z^2 + d_x^2 + d_y^2. \quad (10)$$

Unified Formulation

One can combine the two cases into a single expression:

$$D^2(p) = z^2 + [\max\{0, |x| - L_x\}]^2 + [\max\{0, |y| - L_y\}]^2. \quad (11)$$

Smooth Approximation

Since the max function is not differentiable at zero, a smooth alternative is to use the softplus function to approximate the max (or the ReLU). For example, define:

$$\text{softplus}(x) = \ln(1 + \exp(x)), \quad (12)$$

which approximates $\max\{0, x\}$. Then, one can define smooth approximations of d_x and d_y by:

$$d_x \approx \text{softplus}(|x| - L_x), \quad d_y \approx \text{softplus}(|y| - L_y), \quad (13)$$

The derivative of the softplus function is given by:

$$\text{softplus}'(x) = \frac{\exp(x)}{1 + \exp(x)} = \frac{1}{1 + \exp(-x)}. \quad (14)$$

For the absolute value, we can use a *softabs*() function defined as,

$$\text{softabs}(x) = \sqrt{x^2 + \epsilon}, \quad \text{where } \epsilon \text{ is a small constant.} \quad (15)$$

The derivative of the softabs function is given by:

$$\text{softabs}'(x) = \frac{x}{\sqrt{x^2 + \epsilon}}. \quad (16)$$

The smooth squared distance becomes,

$$D(p, v) \approx z^2 + \underbrace{\text{softplus}(\text{softabs}(x) - L_x)^2}_{dx^2} + \underbrace{\text{softplus}(\text{softabs}(y) - L_y)^2}_{dy^2} \quad (17)$$

Computation of the Jacobian of the distance wrt to the parameter vector v

The derivative of the squared distance with respect to the parameter vector \mathbf{v} is given by:

$$\frac{\partial D}{\partial v} = \left[0, 0, 0, 0, 0, 0, \frac{\partial D}{\partial w}, \frac{\partial D}{\partial d}, \frac{\partial D}{\partial h} \right] \quad (18)$$

The derivatives with respect to the parameters w, d, h are computed using the chain rule and the transformation matrix H_{pose} .

We define the intermediate variables:

$$dx^2 = [\text{softplus}(\text{softabs}(ps_x) - L_x)]^2, \quad (19)$$

$$u^2 = [\text{softplus}(v)]^2 \quad (20)$$

$$s = \text{softabs}(ps_x) = \sqrt{ps_x^2 + \epsilon}, \quad (21)$$

$$v = s - L_x, \quad (22)$$

$$ps_x = H_{point}(0, :) p, \quad (23)$$

$$L_x = \frac{\text{width}}{2} \quad (24)$$

$$(25)$$

Derivative of D with respect to w

The derivative of D with respect to w is given by:

$$\frac{\partial D}{\partial w} = \frac{\partial D}{\partial dx^2} \cdot \frac{\partial dx^2}{\partial w} \quad (26)$$

Since only the dx^2 term depends on w , we can compute the derivative of D with respect to dx^2 first:

$$\frac{\partial D}{\partial w} = \frac{\partial dx^2}{\partial w} \quad (27)$$

Using the chain rule, we can express the derivative of dx^2 with respect to w as:

$$\frac{\partial dx^2}{\partial w} = \frac{\partial dx^2}{\partial u} \cdot \frac{\partial u}{\partial v} \cdot \frac{\partial v}{\partial s} \cdot \frac{\partial s}{\partial ps_x} \cdot \frac{\partial ps_x}{\partial p} \cdot \frac{\partial p}{\partial w} \quad (28)$$

where

$$\frac{\partial p}{\partial w} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \quad (29)$$

$$\frac{\partial ps_x}{\partial p_x} = H_{point}(0, :) \quad (30)$$

$$\frac{\partial s}{\partial ps_x} = \frac{ps_x}{\sqrt{ps_x^2 + \epsilon}} \quad (31)$$

$$\frac{\partial v}{\partial s} = 1 \quad (32)$$

$$\frac{\partial u}{\partial v} = \frac{1}{1 + e^{-(s-L_x)}} \quad (33)$$

$$\frac{\partial dx^2}{\partial u} = 2 \text{softplus}(s - L_x) \quad (34)$$

$$(35)$$

Multiplying all the factors, we have:

$$\frac{\partial D}{\partial w} = \frac{\partial dx^2}{\partial w} = 2 \text{softplus}(s - L_x) \cdot \frac{1}{1 + e^{-(s-L_x)}} \cdot \frac{ps_x}{\sqrt{ps_x^2 + \epsilon}} \cdot H_{point}(0, :) \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \quad (36)$$

Derivative of D with respect to d

The derivative of D with respect to d is given by:

$$\frac{\partial D}{\partial d} = \frac{\partial D}{\partial dy^2} \cdot \frac{\partial dy^2}{\partial d} \quad (37)$$

Since only the dy^2 term depends on d , we can compute the derivative of D with respect to dy^2 first:

$$\frac{\partial D}{\partial d} = \frac{\partial dy^2}{\partial d} \quad (38)$$

Using the chain rule, we can express the derivative of dy^2 with respect to d as:

$$\frac{\partial dy^2}{\partial d} = \frac{\partial dy^2}{\partial u_y} \cdot \frac{\partial u_y}{\partial v_y} \cdot \frac{\partial v_y}{\partial s_y} \cdot \frac{\partial s_y}{\partial ps_y} \cdot \frac{\partial ps_y}{\partial p_y} \cdot \frac{\partial p_y}{\partial d} \quad (39)$$

where

$$\frac{\partial p}{\partial d} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \quad (40)$$

$$\frac{\partial ps_y}{\partial p_y} = H_{point}(1, :) \quad (41)$$

$$\frac{\partial s_y}{\partial ps_y} = \frac{ps_y}{\sqrt{ps_y^2 + \epsilon}} \quad (42)$$

$$\frac{\partial v_y}{\partial s_y} = 1 \quad (43)$$

$$\frac{\partial u_y}{\partial v_y} = \frac{1}{1 + e^{-(s-L_y)}} \quad (44)$$

$$\frac{\partial dy^2}{\partial u_y} = 2 \cdot \text{softplus}(s - L_y) \quad (45)$$

$$(46)$$

Multiplying all the factors, we have:

$$\frac{\partial D}{\partial d} = \frac{\partial D}{\partial dy^2} = 2 \cdot \text{softplus}(s - L_y) \cdot \frac{1}{1 + e^{-(s-L_y)}} \cdot \frac{ps_y}{\sqrt{ps_y^2 + \epsilon}} \cdot H_{point}(1, :) \cdot \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \quad (47)$$

Derivative of D with respect to h

The derivative of D with respect to h is given by:

$$\frac{\partial D}{\partial h} = \frac{\partial D}{\partial ps_z} \cdot \frac{\partial ps_z}{\partial h} \quad (48)$$

Since $D = ps_z^2$ plus two other terms that not depend on h ,

$$\frac{\partial D}{\partial h} = \frac{\partial ps_z^2}{\partial ps_z} \cdot \frac{\partial ps_z}{\partial z} \cdot \frac{\partial p_z}{\partial h} = 2 \cdot ps_z \cdot H_{point}(2, :) \cdot \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad (49)$$

Final form of the Jacobian of \mathbf{D} wrt \mathbf{v}

The final Jacobian of the distance with respect to v is given by:

$$\frac{\partial D}{\partial v} = \left[0, 0, 0, 0, 0, 0, \frac{\partial D}{\partial w}, \frac{\partial D}{\partial d}, \frac{\partial D}{\partial h} \right] \quad (50)$$

where

$$\begin{aligned} \frac{\partial D}{\partial w} &= 2 \text{softplus}(s_x - L_x) \cdot \frac{1}{1 + e^{-(s_x - L_x)}} \cdot \frac{ps_x}{\sqrt{ps_x^2 + \epsilon}} \cdot H_{point}(0, :) \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \\ \frac{\partial D}{\partial d} &= 2 \text{softplus}(s_y - L_y) \cdot \frac{1}{1 + e^{-(s_y - L_y)}} \cdot \frac{ps_y}{\sqrt{ps_y^2 + \epsilon}} \cdot H_{point}(1, :) \cdot \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \\ \frac{\partial D}{\partial h} &= 2 \cdot ps_z \cdot H_{point}(2, :) \cdot \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \end{aligned} \quad (51)$$

Derivatives with respect to the input point \mathbf{p}

The derivative of the squared distance with respect to the input point \mathbf{p} is given by:

$$\frac{\partial D}{\partial \mathbf{p}} = \left[\frac{\partial D}{\partial p_x}, \frac{\partial D}{\partial p_y}, \frac{\partial D}{\partial p_z} \right] \quad (52)$$

and D is defined as:

$$D = ps_z^2 + dx^2 + dy^2 \quad (53)$$

The intermediate variables are defined as:

$$dx^2 = [\text{softplus}(\text{softabs}(ps_x) - L_x)]^2, \quad (54)$$

$$u^2 = [\text{softplus}(v)]^2 \quad (55)$$

$$s = \text{softabs}(ps_x) = \sqrt{ps_x^2 + \epsilon}, \quad (56)$$

$$v = s - L_x, \quad (57)$$

$$ps_x = H_{\text{point}}(0, :) p, \quad (58)$$

$$L_x = \frac{\text{width}}{2}. \quad (59)$$

Derivative of D with respect to p_x

The derivative of D with respect to p_x is given by:

$$\frac{\partial D}{\partial p_x} = \frac{\partial D}{\partial dx^2} \cdot \frac{\partial dx^2}{\partial p_x} \quad (60)$$

and since only the dx^2 term depends on p_x , we can compute the derivative of D with respect to dx^2 first:

$$\frac{\partial D}{\partial p_x} = \frac{\partial dx^2}{\partial p_x} \quad (61)$$

Apply the chain rule:

$$\frac{\partial dx^2}{\partial p_x} = \frac{\partial dx^2}{\partial u} \cdot \frac{\partial u}{\partial v} \cdot \frac{\partial v}{\partial s} \cdot \frac{\partial s}{\partial ps_x} \cdot \frac{\partial ps_x}{\partial p_x} \quad (62)$$

We compute each term individually:

$$\frac{\partial ps_x}{\partial p_x} = H_{point}(0, 0) \quad (63)$$

$$\frac{\partial s}{\partial ps_x} = \frac{ps_x}{\sqrt{ps_x^2 + \epsilon}} \quad (64)$$

$$\frac{\partial v}{\partial s} = 1 \quad (65)$$

$$\frac{\partial u}{\partial v} = \frac{1}{1 + e^{-(s-L_x)}} \quad (66)$$

$$\frac{\partial dx^2}{\partial u} = 2 \text{softplus}(s - L_x) \quad (67)$$

$$(68)$$

Combining all derivatives

Combining these terms, the complete derivative of dx with respect to p_x is:

$$\frac{\partial dx^2}{\partial p_x} = 2 \cdot \text{softplus}(s - L_x) \cdot \frac{1}{1 + e^{-(s-L_x)}} \cdot 1 \cdot \frac{ps_x}{\sqrt{ps_x^2 + \epsilon}} \cdot H_{point}(0, 0) \quad (69)$$

Derivative of D with respect to p_y

The derivative of D with respect to p_y is given by:

$$\frac{\partial D}{\partial p_y} = \frac{\partial D}{\partial dy^2} \cdot \frac{\partial dy^2}{\partial p_y} \quad (70)$$

and since only the dy^2 term depends on p_y , we can compute the derivative of D with respect to dy^2 first:

$$\frac{\partial D}{\partial p_y} = \frac{\partial dy^2}{\partial p_y} \quad (71)$$

Apply the chain rule:

$$\frac{\partial dy^2}{\partial p_y} = \frac{\partial dy^2}{\partial u} \cdot \frac{\partial u}{\partial v} \cdot \frac{\partial v}{\partial s} \cdot \frac{\partial s}{\partial ps_y} \cdot \frac{\partial ps_y}{\partial p_y} \quad (72)$$

We compute each term individually:

$$\frac{\partial ps_y}{\partial p} = H_{point}(1, 1) \quad (73)$$

$$\frac{\partial s}{\partial ps_y} = \frac{ps_y}{\sqrt{ps_y^2 + \epsilon}} \quad (74)$$

$$\frac{\partial v}{\partial s} = 1 \quad (75)$$

$$\frac{\partial u}{\partial v} = \frac{1}{1 + e^{-(s-L_y)}} \quad (76)$$

$$\frac{\partial dy^2}{\partial u} = 2 \text{softplus}(s - L_y) \quad (77)$$

$$(78)$$

Combining all derivatives

Combining these terms, the complete derivative of dy with respect to p_y is:

$$\frac{\partial dy^2}{\partial p_y} = 2 \cdot \text{softplus}(s - L_y) \cdot \frac{1}{1 + e^{-(s-L_y)}} \cdot 1 \cdot \frac{ps_y}{\sqrt{ps_y^2 + \epsilon}} \cdot H_{point}(1, 1). \quad (79)$$

Derivative of D with respect to p_z

The derivative of D with respect to p_z is given by:

$$\frac{\partial D}{\partial p_z} = \frac{\partial D}{\partial ps_z} \cdot \frac{\partial ps_z}{\partial p_z} \quad (80)$$

and since only the ps_z term depends on p_z , we can compute the derivative of D with the second factor only:

$$\frac{\partial D}{\partial p_z} = \frac{\partial ps_z}{\partial p_z} = 2 ps_z \cdot H_{pose}(2, 2) \quad (81)$$

Final Jacobian of the distance with respect to p

The final Jacobian is given by:

$$\begin{aligned}\frac{\partial D}{\partial p} &= \left[\frac{\partial D}{\partial p_x}, \frac{\partial D}{\partial p_y}, \frac{\partial D}{\partial p_z} \right] \\ &= \left[2 \text{softplus}(s_x - L_x) \cdot \frac{1}{1 + e^{-(s_x - L_x)}} \cdot \frac{ps_x}{\sqrt{ps_x^2 + \epsilon}} \cdot H_{\text{point}}(0, 0), \right. \\ &\quad 2 \text{softplus}(s_y - L_y) \cdot \frac{1}{1 + e^{-(s_y - L_y)}} \cdot \frac{ps_y}{\sqrt{ps_y^2 + \epsilon}} \cdot H_{\text{point}}(1, 1), \\ &\quad \left. 2ps_z \cdot H_{\text{point}}(2, 2) \right]\end{aligned}\tag{82}$$