#### **Computer Vision II: Multiple-view Geometry**

Exercise 08

25<sup>th</sup> June, 2025

#### Robust Odometry Estimation for RGB-D Cameras

- Given two images  $I_1$  and  $I_2$ , and the depth maps ( $Z_1$  and  $Z_2$ )
  - The goal is to find a transformation which maps  $I_2 \rightarrow I_1$
- Towards that, we have  $I_1(x) = I_2(\tau(\xi, x))$ ,
  - where,  $\tau$  is a warping function and  $\xi \in se(3)$  is the lie algebra defining the camera motion and x is a 2D pixel location **in the first image**,  $I_1$

## Warping Function

• Step 1. We first unproject the 2D location x (from  $I_1$  as)

• 
$$p = \pi^{-1}(x, Z_1(x)) = Z_1(x) \left(\frac{u+c_x}{f_x}, \frac{v+c_y}{f_y}, 1\right)^T$$
,  
• where,  $K = \begin{pmatrix} f_x & 0 & -c_x \\ 0 & f_y & -c_y \\ 0 & 0 & 1 \end{pmatrix}$  is the intrinsic matrix.

• Question: Is p now in world coordinate system? Of which camera?

## Warping Function

- Step 2. The transformation function T(g, p) transforms the point p to I<sub>2</sub>'s coordinate system
  - T(g,p) = Rp + t, where,  $g(\xi) = \exp(\hat{\xi}) = \begin{bmatrix} R & t \\ 0 & 1 \end{bmatrix}$  is the transformation matrix formed using  $\xi \in se(3)$
  - Question: what is the dimensionality of  $\xi$ ?
- Step 3. Next, we project T(g, p) to the image space of  $I_2$  as  $\pi(T(g, p)) = \left(\frac{f_x x}{z} c_x, \frac{f_y y}{z} c_y\right)$
- Therefore,  $\tau(\xi, x) = \pi(T(g(\xi), p)) = \pi(T(g(\xi), \pi^{-1}(x, z_1(x))))$

## Residual

- Our goal is to find the parameter  $\xi$ , therefore, we optimize to reduce the residual between the two images,  $r_i(\xi) = I_2(\tau(\xi, x_i)) - I_1(x_i)$ , for the  $i^{th}$  pixel.
- We assume that all the pixels are i.i.d. and the likelihood of the  $\xi$  over the entire image is  $p(r \,|\, \xi) = \Pi_i(p(r_i | \xi)$

# Maximum A Posteriori (MAP)

• Using Bayes' rule,  $p(\xi|r) = \frac{p(r|\xi)p(\xi)}{p(r)}$ 

$$\begin{aligned} \bullet \ \xi_{MAP} &= \frac{argmax}{\xi} \ p(\xi|r) \\ &= \frac{argmax}{\xi} \ p(r|\xi)p(\xi) = \frac{argmin}{\xi} - \Sigma_i \log p(r_i|\xi) - \log p(\xi) \end{aligned}$$

• Minimizing the term, we obtain the optimal  $\xi$  , the transformation between the two images