

## Assignment2

Course: Computer Vision

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Q1. In the augmented Euclidean plane, there is a line x - 3y + 4 = 0, what is the homogeneous coordinate of the infinity point of this line?

## Ans.

In the augmented Euclidean plane, the point at infinity on a line exists both on the line itself and on the line at infinity. Therefore, the point at infinity of a line is the intersection point of the line and the line at infinity.

Now the homogeneous coordinates of two lines are:  $\mathbf{l} = (1, -3, 4)$  and  $\mathbf{l'} = (0, 0, 1)$ .

Therefore, the homogeneous coordinates of the intersection point x is  $\mathbf{l} \times \mathbf{l'} = (-3, -1, 0)$ .

From the definition of a point's homogeneous coordinates, if (-3, -1, 0) is a homogeneous coordinate of a point, then (k(-3, -1, 0)) is also a homogeneous coordinate of the same point, where  $k \neq 0$ .

So the homogeneous coordinates of the infinity point of the line is  $k(-3, -1, 0) (k \neq 0)$ .

**Q2.** On the normalized retinal plane, suppose that  $p_n$  is an ideal point of projection without considering distortion. If distortion is considered,  $p_n = (x, y)^T$  is mapped to  $p_d = (x_d, y_d)^T$  which is also on the normalized retinal plane. Their relationship is

$$x_d = x(1 + k_1r^2 + k_2r^4) + 2\rho_1xy + \rho_2(r^2 + 2x^2) + xk_3r^6$$
  
$$y_d = y(1 + k_1r^2 + k_2r^4) + 2\rho_2xy + \rho_1(r^2 + 2y^2) + yk_3r^6$$
  
where  $r^2 = x^2 + y^2$ 

For performing nonlinear optimization in the pipeline of camera calibration, we need to compute the Jacobian matrix of  $p_d$  w.r.t  $p_n$ , i.e.,  $\frac{dp_d}{dp_n^T}$ 

It should be noted that in this question  $p_d$  is the function of  $p_n$  and all the other parameters can be regarded as constants.

Ans.

$$\frac{dp_d}{dp_n^T} = \begin{bmatrix} \frac{\partial x_d}{\partial x} & \frac{\partial x_d}{\partial y} \\ \frac{\partial y_d}{\partial x} & \frac{\partial y_d}{\partial y} \end{bmatrix}$$
$$r^2 = x^2 + y^2 \Rightarrow \frac{\partial r}{\partial x} = \frac{x}{\sqrt{x^2 + y^2}}, \frac{\partial r}{\partial y} = \frac{y}{\sqrt{x^2 + y^2}}$$

$$\frac{\partial x_d}{\partial x} = (1 + k_1 r^2 + k_2 r^4) + x(k_1 2r \frac{\partial r}{\partial x} + k_2 4r^3 \frac{\partial r}{\partial x}) + 2\rho_1 y + \rho_2 (2r \frac{\partial r}{\partial x} + 4x) + k_3 (r^6 + x6r^5 \frac{\partial r}{\partial x}) 
= (1 + k_1 r^2 + k_2 r^4) + (2k_1 x^2 + 4k_2 r^2 x^2) + 2\rho_1 y + 6\rho_2 x + (k_3 r^6 + 6k_3 r^4 x^2) 
\frac{\partial x_d}{\partial y} = x(k_1 2r \frac{\partial r}{\partial y} + k_2 4r^3 \frac{\partial r}{\partial y}) + 2\rho_1 x + \rho_1 (2r \frac{\partial r}{\partial y}) + xk_3 (6r^5 \frac{\partial r}{\partial y}) 
= (2k_1 xy + 4k_2 r^2 xy) + 2\rho_1 x + 2\rho_1 y + 6k_3 r^4 xy 
\frac{\partial y_d}{\partial x} = y(k_1 2r \frac{\partial r}{\partial x} + k_2 4r^3 \frac{\partial r}{\partial x}) + 2\rho_2 y + \rho_1 (2r \frac{\partial r}{\partial x}) + yk_3 6r^5 \frac{\partial r}{\partial x} 
= (2k_1 xy + 4k_2 r^2 xy) + 2\rho_2 y + 2\rho_1 x + 6k_3 r^4 xy 
\frac{\partial y_d}{\partial y} = (1 + k_1 r^2 + k_2 r^4) + y(k_1 2r \frac{\partial r}{\partial y} + k_2 4r^3 \frac{\partial r}{\partial y}) + 2\rho_2 x + \rho_1 (2r \frac{\partial r}{\partial y} + 4y) + k_3 (r^6 + y6r^5 \frac{\partial r}{\partial y}) 
= (1 + k_1 r^2 + k_2 r^4) + (2k_1 y^2 + 4k_2 r^2 y^2) + 2\rho_2 x + 6\rho_1 y + (k_3 r^6 + 6k_3 r^4 y^2)$$

So the answer is following

$$\frac{dp_d}{dp_n^2} = \begin{bmatrix} (1+k_1r^2+k_2r^4)+(2k_1x^2+4k_2r^2x^2)+2\rho_1y+6\rho_2x+(k_3r^6+6k_3r^4x^2) & (2k_1xy+4k_2r^2xy)+2\rho_1x+2\rho_1y+6k_3r^4xy \\ (2k_1xy+4k_2r^2xy)+2\rho_2x+2\rho_1x+6k_3r^4xy & (1+k_1r^2+k_2r^4)+(2k_1y^2+4k_2r^2y^2)+2\rho_2x+6\rho_1y+(k_3r^6+6k_3r^4y^2) \end{bmatrix}$$

Q3. In our lecture, we mentioned that for performing nonlinear optimization in the pipeline of camera calibration, we need to compute the Jacobian of the rotation matrix (represented in a vector) w.r.t its axis-angle representation. In this question, your task is to derive the concrete formula of this Jacobian matrix. Suppose that

$$\mathbf{d} = \theta \mathbf{n} \in \mathbb{R}^{3 \times 1}$$

where  $\mathbf{n} = \begin{bmatrix} n_1 \\ n_2 \\ n_3 \end{bmatrix}$  is a 3D unit vector and  $\theta$  is a real number denoting the rotation angle.

With Rodrigues formula, **d** can be converted to its rotation matrix form:

$$\mathbf{R} = \cos \theta \mathbf{I} + (1 - \cos \theta) \mathbf{n} \mathbf{n}^T + \sin \theta \mathbf{n}^{\hat{}}$$
 and obviously  $\mathbf{R} \triangleq \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}$  is a 3 × 3 matrix.

Denote **r** by the vectorized form of **R**, i.e.  $\mathbf{r} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & r_{21} & r_{22} & r_{23} & r_{31} & r_{32} & r_{33} \end{bmatrix}^T$ . Please give the concrete form of Jacobian matrix of **r** w.r.t **d**, i.e.,  $\frac{d\mathbf{r}}{d\mathbf{d}^T} \in \mathbb{R}^{9\times 3}$ .

In order to make it easy to check your result, please follow the following notation requirements,  $\alpha \triangleq \sin \theta$ ,  $\beta \triangleq \cos \theta$ ,  $\gamma = 1 - \cos \theta$ .

In other words, the ingredients appearing in your formula are restricted to  $\alpha$ ,  $\beta$ ,  $\gamma$ ,  $\theta$ ,  $n_1$ ,  $n_2$ , and  $n_3$ .

Ans.

$$R = \begin{bmatrix} \cos \theta & 0 & 0 \\ 0 & \cos \theta & 0 \\ 0 & 0 & \cos \theta \end{bmatrix} + (1 - \cos \theta) \begin{bmatrix} n_1^2 & n_1 n_2 & n_1 n_3 \\ n_1 n_2 & n_2^2 & n_2 n_2 \\ n_1 n_3 & n_2 n_3 & n_3^2 \end{bmatrix} + \sin \theta \begin{bmatrix} 0 & -n_3 & n_2 \\ n_3 & 0 & -n_1 \\ -n_2 & n_1 & 0 \end{bmatrix}$$

$$\gamma = \begin{bmatrix} r_{11} & r_{12} & r_{13} & r_{21} & r_{22} & r_{23} & r_{31} & r_{32} & r_{33} \end{bmatrix}^{T}$$

$$\begin{bmatrix} \beta + \gamma n_{1}^{2} \\ \gamma n_{1} n_{2} - \alpha n_{3} \\ \gamma n_{1} n_{3} + \alpha n_{2} \\ \gamma n_{1} n_{2} + \alpha n_{3} \\ \beta + \gamma n_{2}^{2} \\ \gamma n_{2} n_{3} - \alpha n_{1} \\ \gamma n_{1} n_{3} - \alpha n_{2} \\ \gamma n_{2} n_{3} + \alpha n_{1} \\ \beta + \gamma n_{3}^{2} \end{bmatrix}$$

$$\mathbf{d} = \theta \mathbf{n} \Rightarrow \begin{cases} d_1 &= \theta n_1 \\ d_2 &= \theta n_2 \\ d_3 &= \theta n_3 \end{cases}$$

Since **n** is a 3D unit vector, we can know:

$$\left(\frac{d_1}{\theta}\right)^2 + \left(\frac{d_2}{\theta}\right)^2 + \left(\frac{d_3}{\theta}\right)^2 + 1 \Rightarrow \theta = \sqrt{d_1^2 + d_2^2 + d_3^2}$$
$$n_i = \frac{d_1}{\sqrt{d_1^2 + d_2^2 + d_3^2}}, \quad \text{where } i \in \{1, 2, 3\}$$

In this problem, we consider  $d_1$ ,  $d_2$ , and  $d_3$  as independent variables.

$$\frac{\partial \theta}{\partial d_i} = \frac{d_i}{\sqrt{d_1^2 + d_2^2 + d_3^2}}, \quad \text{where } i \in \{1, 2, 3\}$$

$$\begin{cases} \frac{\partial \alpha}{\partial d_i} = \frac{\partial \alpha}{\partial \theta} \frac{\partial \theta}{\partial d_i} = \cos \theta \frac{d_i}{\sqrt{d_1^2 + d_2^2 + d_3^2}} = \beta n_i, & \text{where } i \in \{1, 2, 3\} \\ \frac{\partial \beta}{\partial d_i} = \frac{\partial \beta}{\partial \theta} \frac{\partial \theta}{\partial d_i} = -\sin \theta \frac{d_i}{\sqrt{d_1^2 + d_2^2 + d_3^2}} = -\alpha n_i, & \text{where } i \in \{1, 2, 3\} \\ \frac{\partial \gamma}{\partial d_i} = \frac{\partial \gamma}{\partial \theta} \frac{\partial \theta}{\partial d_i} = \sin \theta \frac{d_i}{\sqrt{d_1^2 + d_2^2 + d_3^2}} = \alpha n_i, & \text{where } i \in \{1, 2, 3\} \end{cases}$$

$$\frac{\partial n_j}{\partial d_i} = \begin{cases}
\frac{\sqrt{d_1^2 + d_2^2 + d_3^2} - d_i \frac{d_i}{\sqrt{d_1^2 + d_2^2 + d_3^2}}}{\frac{d_1^2 + d_2^2 + d_3^2}{2}} & \text{if } i = j \\
\frac{d_j \frac{d_i}{\sqrt{d_1^2 + d_2^2 + d_3^2}}}{-\frac{d_1^2 + d_2^2 + d_3^2}{2}} & \text{if } i \neq j
\end{cases}$$

$$= \begin{cases}
\frac{1 - n_i^2}{\theta} & \text{if } i = j \\
-\frac{n_i n_j}{\theta} & \text{if } i \neq j
\end{cases}$$

The answer is following, and the detail is in the next page:

$$\frac{d\mathbf{r}}{d\mathbf{d}^T} = \begin{bmatrix} \frac{\partial r_{11}}{\partial d_1} & \frac{\partial r_{11}}{\partial d_2} & \frac{\partial r_{11}}{\partial d_3} \\ \frac{\partial r_{12}}{\partial d_1} & \frac{\partial r_{12}}{\partial d_2} & \frac{\partial r_{12}}{\partial d_3} \\ \frac{\partial r_{13}}{\partial d_1} & \frac{\partial r_{13}}{\partial d_2} & \frac{\partial r_{13}}{\partial d_3} \\ \frac{\partial r_{21}}{\partial d_1} & \frac{\partial r_{21}}{\partial d_2} & \frac{\partial r_{21}}{\partial d_3} \\ \frac{\partial r_{22}}{\partial d_1} & \frac{\partial r_{22}}{\partial d_2} & \frac{\partial r_{22}}{\partial d_3} \\ \frac{\partial r_{23}}{\partial d_1} & \frac{\partial r_{23}}{\partial d_2} & \frac{\partial r_{23}}{\partial d_3} \\ \frac{\partial r_{31}}{\partial d_1} & \frac{\partial r_{31}}{\partial d_2} & \frac{\partial r_{31}}{\partial d_3} \\ \frac{\partial r_{32}}{\partial d_1} & \frac{\partial r_{32}}{\partial d_2} & \frac{\partial r_{32}}{\partial d_3} \\ \frac{\partial r_{33}}{\partial d_1} & \frac{\partial r_{33}}{\partial d_2} & \frac{\partial r_{33}}{\partial d_3} \\ \frac{\partial r_{33}}{\partial d_1} & \frac{\partial r_{33}}{\partial d_2} & \frac{\partial r_{33}}{\partial d_3} \end{bmatrix}$$

$$\begin{bmatrix} \alpha n_1(n_1^2-1) + \frac{2\gamma n_1(1-n_1^2)}{\theta} & \alpha n_2(n_1^2-1) - \frac{2\gamma n_1^2 n_2}{\theta} & \alpha n_3(n_1^2-1) - \frac{2\gamma n_1^2 n_3}{\theta} \\ n_1(\alpha n_1 n_2 - \beta n_3) + \frac{\gamma n_2(1-2n_1^2) + \alpha n_1 n_3}{\theta} & n_2(\alpha n_1 n_2 - \beta n_3) + \frac{\gamma n_1(1-2n_2^2) + \alpha n_2 n_3}{\theta} & n_3(\alpha n_1 n_2 - \beta n_3) - \frac{2\gamma n_1 n_2 n_3 + \alpha (1-n_3^2)}{\theta} \\ n_1(\alpha n_1 n_3 + \beta n_2) + \frac{\gamma n_3(1-2n_1^2) - \alpha n_1 n_2}{\theta} & n_2(\alpha n_1 n_3 + \beta n_2) + \frac{\alpha (1-n_2^2) - 2\gamma n_1 n_2 n_3}{\theta} & n_3(\alpha n_1 n_3 + \beta n_2) + \frac{\gamma n_1(1-2n_3^2) - \alpha n_2 n_3}{\theta} \\ n_1(\alpha n_1 n_2 + \beta n_3) + \frac{\gamma n_2(1-2n_1^2) - \alpha n_1 n_3}{\theta} & n_2(\alpha n_1 n_2 + \beta n_3) + \frac{\gamma n_1(1-2n_2^2) - \alpha n_2 n_3}{\theta} & n_3(\alpha n_1 n_2 + \beta n_3^2) + \frac{\alpha (1-n_3^2) - 2\gamma n_1 n_2 n_3}{\theta} \\ n_1(\alpha n_2 n_3 - \beta n_1) - \frac{2\gamma n_1 n_2^2}{\theta} & \alpha n_2(n_2^2 - 1) + \frac{2\gamma n_2(1-n_2^2)}{\theta} & \alpha n_3(\alpha n_2 n_3 - \beta n_1) + \frac{\gamma n_2(1-2n_3^2) + \alpha n_1 n_2}{\theta} \\ n_1(\alpha n_1 n_3 - \beta n_2) + \frac{\gamma n_3(1-2n_1^2) + 2\gamma n_1 n_2 n_3}{\theta} & n_2(\alpha n_1 n_3 - \beta n_2) - \frac{\alpha (1-n_2^2) + 2\gamma n_1 n_2 n_3}{\theta} & n_3(\alpha n_1 n_3 - \beta n_2) + \frac{\gamma n_1(1-2n_3^2) + \alpha n_1 n_3}{\theta} \\ n_1(\alpha n_2 n_3 + \beta n_1) + \frac{\alpha (1-n_1^2) + 2\gamma n_1 n_2 n_3}{\theta} & n_2(\alpha n_2 n_3 + \beta n_1) + \frac{\gamma n_3(1-2n_2^2) - \alpha n_1 n_2}{\theta} & n_3(\alpha n_2 n_3 + \beta n_1) + \frac{\gamma n_2(1-2n_3^2) + \alpha n_1 n_3}{\theta} \\ \alpha n_1(\alpha n_2 n_3 + \beta n_1) + \frac{\alpha (1-n_1^2) - 2\gamma n_1 n_2 n_3}{\theta} & n_2(\alpha n_2 n_3 + \beta n_1) + \frac{\gamma n_3(1-2n_2^2) - \alpha n_1 n_2}{\theta} & n_3(\alpha n_2 n_3 + \beta n_1) + \frac{\gamma n_2(1-2n_3^2) - \alpha n_1 n_3}{\theta} \\ \alpha n_1(\alpha n_2 n_3 + \beta n_1) + \frac{\alpha (1-n_1^2) - 2\gamma n_1 n_2 n_3}{\theta} & n_2(\alpha n_2 n_3 + \beta n_1) + \frac{\gamma n_3(1-2n_2^2) - \alpha n_1 n_2}{\theta} & n_3(\alpha n_2 n_3 + \beta n_1) + \frac{\gamma n_2(1-2n_3^2) - \alpha n_1 n_3}{\theta} \\ \alpha n_2(n_1^2 - 1) - \frac{2\gamma n_2 n_3}{\theta} & \alpha n_2(n_2^2 - 1) - \frac{2\gamma n_2 n_3}{\theta} & \alpha n_3(n_3^2 - 1) + \frac{2\gamma n_3(1-n_3^2)}{\theta} \\ \alpha n_3(n_3^2 - 1) + \frac{2\gamma n_3(1-n_3^2)}{\theta} & \alpha n_3(n_3^2 - 1) + \frac{2\gamma n_3(1-n_3^2)}{\theta} \\ \alpha n_3(n_3^2 - 1) - \frac{2\gamma n_3(1-n_3^2)}{\theta} & \alpha n_3(n_3^2 - 1) + \frac{2\gamma n_3(1-n_3^2)}{\theta} \\ \alpha n_3(n_1^2 - 1) - \frac{2\gamma n_3(1-n_3^2)}{\theta} & \alpha n_3(n_1^2 - 1) + \frac{2\gamma n_3(1-n_3^2)}{\theta} \\ \alpha n_3(n_1^2 - 1) - \frac{2\gamma n_3(1-n_3^2)}{\theta} & \alpha n_3(n_1^2 - 1) - \frac{2\gamma n_3(1-n_3^2)}{\theta} \\ \alpha n_3(n_1^$$

$$\frac{\partial \hat{g}_{1}^{2} + \partial \hat{g}_{1}^{2} + \partial \hat{g}_{2}^{2} + \partial \hat{g}_{1}^{2} + \gamma 2n_{1} \frac{\partial \hat{g}_{1}}{\partial h}}{\partial \hat{g}_{1}^{2} + \gamma 2n_{1} \frac{\partial \hat{g}_{2}}{\partial h}} + \frac{\partial \hat{g}_{1}}{\partial h} + \gamma 2n_{1} \frac{\partial \hat{g}_{2}}{\partial h} + \frac{\partial \hat{g}_{2}}{\partial h} + \gamma 2n_{1} \frac{\partial \hat{g}_{2}}{\partial h} + \frac{\partial \hat{g}_{2}}{\partial h} + \gamma 2n_{1} \frac{\partial \hat{g}_{2}}{\partial h} + \frac{\partial \hat{g}_{2}}{\partial h} + \gamma 2n_{1} \frac{\partial \hat{g}_{2}}{\partial h} + \frac{\partial \hat{g}_{2}}{\partial h} n_{1} + \gamma 2n_{1} \frac{\partial \hat{g}_{2}}{\partial h} + \frac{\partial \hat{g}_{2}}{\partial h} n_{2} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \gamma n_{1} \frac{\partial \hat{g}_{2}}{\partial h} - \frac{\partial \hat{g}_{2}}{\partial h} n_{3} + \alpha n_{1} n_{2} + \alpha n_{2} n_{3} + \alpha n_{2} n_{3} + \alpha n_{$$