

Robotic Navigation and Exploration

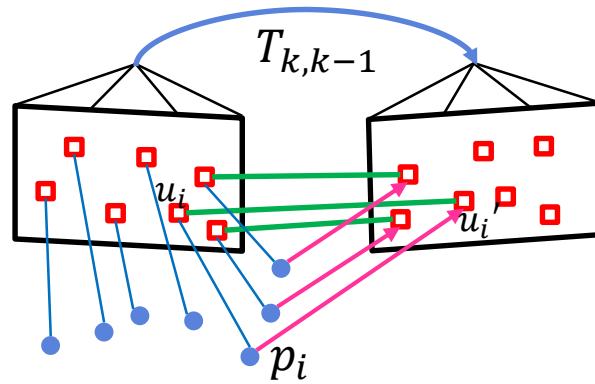
Week 9: SLAM - Direct Method

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CS, NTHU / CSIE, NCKU

Direct Method & Indirect Method

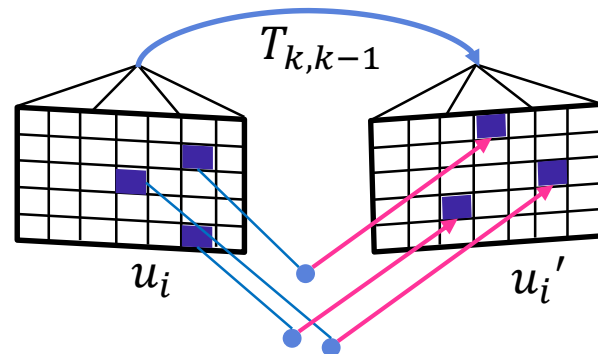
Indirect Method



$$T_{k,k-1} = \underset{T}{\operatorname{argmin}} \sum_i^N ||u_i' - \pi p_i||^2$$

Minimize Geometric Error (Reprojection Error)

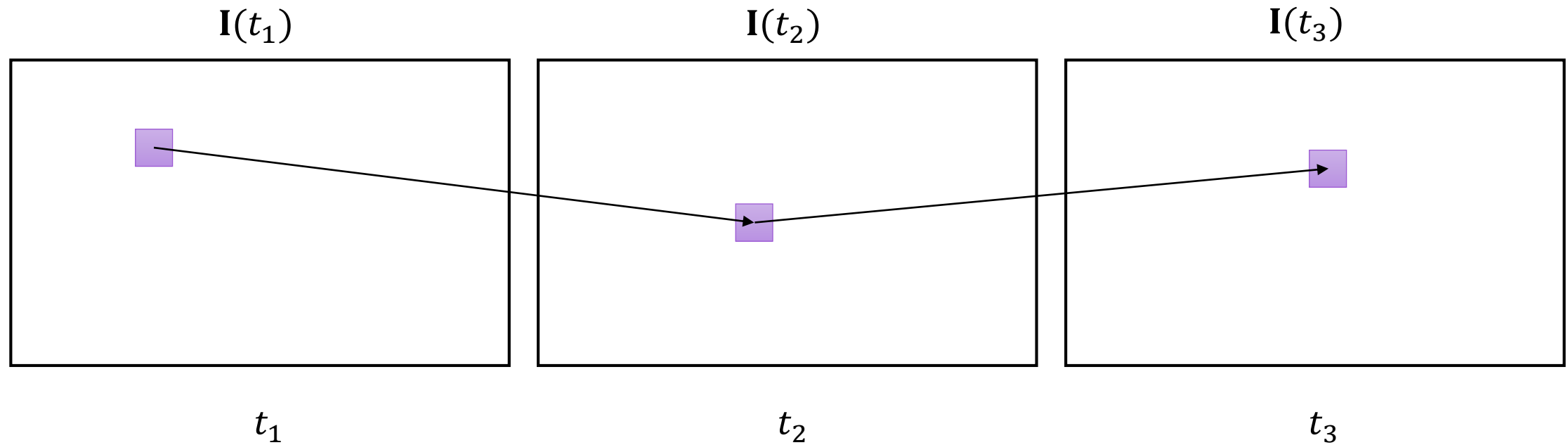
Direct Method



$$T_{k,k-1} = \underset{T}{\operatorname{argmin}} \sum_i^N ||I_k(u_i') - I_{k-1}(u_i)||^2$$

Minimize Photometric Error (Pixel Grayscale)

Optical Flow



Assumption : the displacement of the image content between two nearby frames is small and approximately constant within a neighborhood of the point p under consideration

$$\mathbf{I}(x + dx, y + dy, t + dt) = \mathbf{I}(x, y, t)$$

Lucas –Kanade (L-K) Method

$$\mathbf{I}(x + dx, y + dy, t + dt) \approx \mathbf{I}(x, y, t) + \frac{\partial \mathbf{I}}{\partial x} dx + \frac{\partial \mathbf{I}}{\partial y} dy + \frac{\partial \mathbf{I}}{\partial t} dt$$

Intensity Invariant:

$$\frac{\partial \mathbf{I}}{\partial x} dx + \frac{\partial \mathbf{I}}{\partial y} dy + \frac{\partial \mathbf{I}}{\partial t} dt = 0$$

$$\frac{\partial \mathbf{I}}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial \mathbf{I}}{\partial y} \frac{\partial y}{\partial t} = - \frac{\partial \mathbf{I}}{\partial t}$$

$$\begin{bmatrix} \mathbf{I}_x & \mathbf{I}_y \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = - \mathbf{I}_t$$

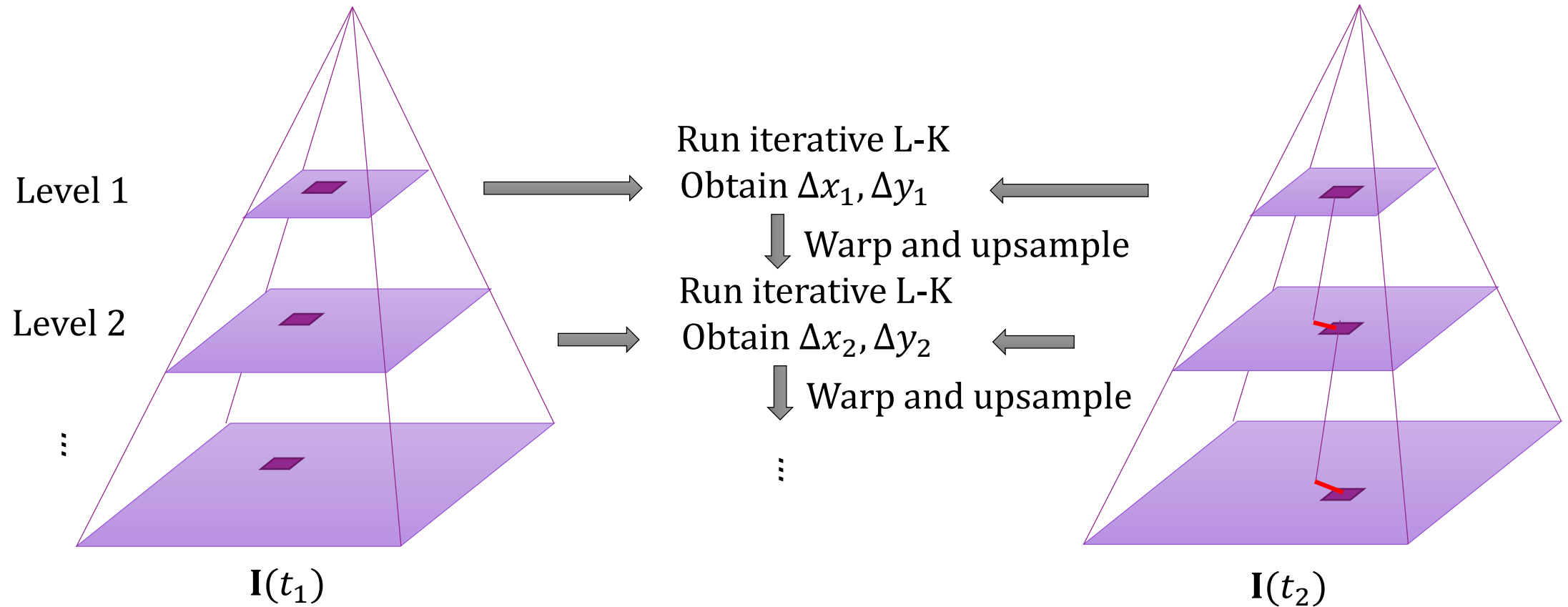
Local consistency:

$$\begin{bmatrix} \mathbf{I}_x & \mathbf{I}_y \end{bmatrix}_k \begin{bmatrix} u \\ v \end{bmatrix} = - \mathbf{I}_{tk}, \quad k = 1, \dots, w^2$$

$$\mathbf{A} = \begin{bmatrix} [\mathbf{I}_x & \mathbf{I}_y]_1 \\ \vdots \\ [\mathbf{I}_x & \mathbf{I}_y]_k \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} \mathbf{I}_{t1} \\ \vdots \\ \mathbf{I}_{tk} \end{bmatrix} \quad \mathbf{A} \begin{bmatrix} u \\ v \end{bmatrix} = -\mathbf{b} \quad \begin{bmatrix} u \\ v \end{bmatrix}^* = -(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b}$$

Coarse-to-fine Optical Flow

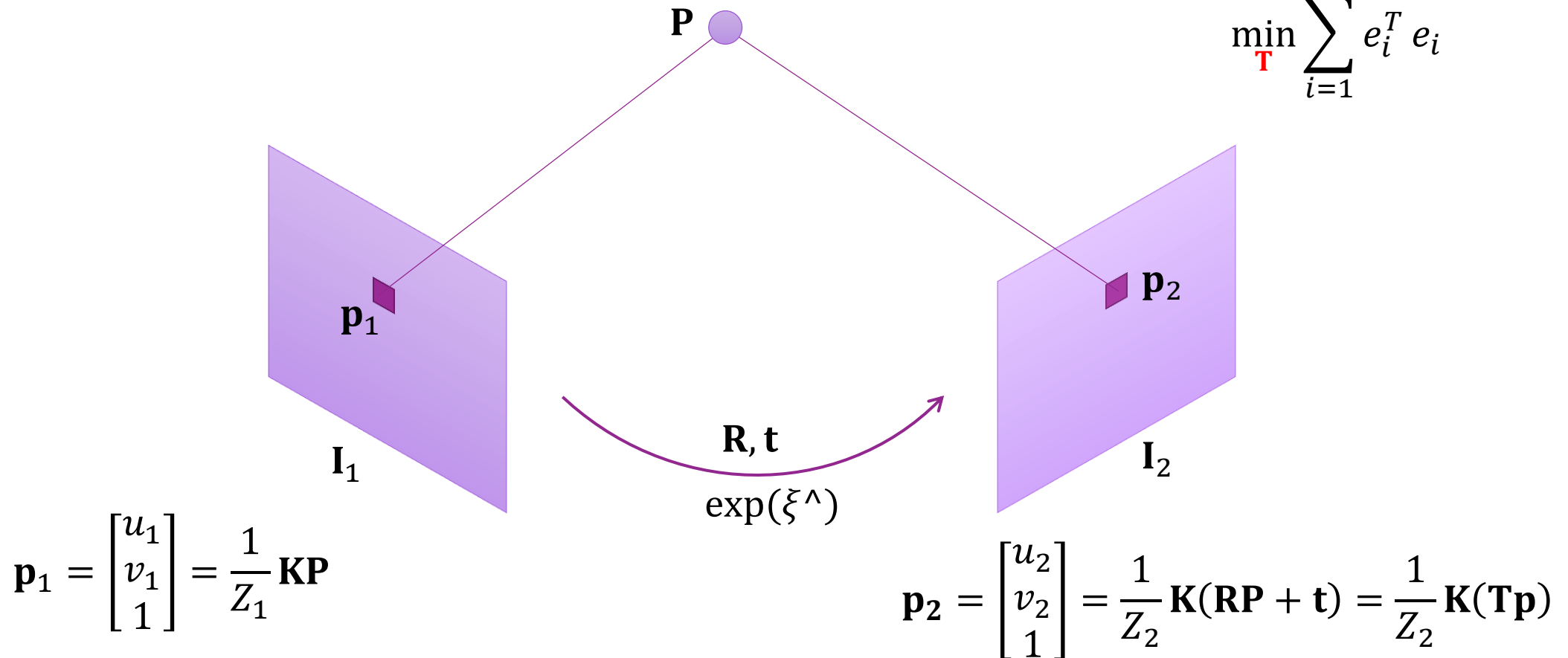
$$\min_{\Delta x, \Delta y} \|I(x, y, t) - I(x + \Delta x, y + \Delta y, t + \Delta t)\|_2^2$$



Direct Method

$$e_i = \mathbf{I}_1(\mathbf{p}_{1,i}) - \mathbf{I}_2(\mathbf{p}_{2,i})$$

$$\min_{\mathbf{T}} \sum_{i=1}^N e_i^T e_i$$



Direct Method

$$e_i = \mathbf{I}_1(\mathbf{p}_{1,i}) - \mathbf{I}_2(\mathbf{p}_{2,i}) \quad \min_{\mathbf{T}} \sum_{i=1}^N e_i^T e_i$$

$$\mathbf{q} = \mathbf{T}\mathbf{p} \quad \mathbf{u} = \frac{1}{Z_2} \mathbf{K}\mathbf{q}$$

$$e(\mathbf{T}) = \mathbf{I}_1(\mathbf{p}_1) - \mathbf{I}_2(\mathbf{u}) \quad \frac{\partial e}{\partial \mathbf{T}} = \frac{\partial \mathbf{I}_2}{\partial \mathbf{u}} \frac{\partial \mathbf{u}}{\partial \mathbf{q}} \frac{\partial \mathbf{q}}{\partial \delta \xi} \delta \xi$$

$$\frac{\partial \mathbf{u}}{\partial \mathbf{q}} = \begin{bmatrix} \frac{\partial u}{\partial X} & \frac{\partial u}{\partial Y} & \frac{\partial u}{\partial Z} \\ \frac{\partial v}{\partial X} & \frac{\partial v}{\partial Y} & \frac{\partial v}{\partial Z} \end{bmatrix} = \begin{bmatrix} \frac{f_x}{Z} & 0 & -\frac{f_x X}{Z^2} \\ 0 & \frac{f_y}{Z} & -\frac{f_y Y}{Z^2} \end{bmatrix}$$

$$\frac{\partial \mathbf{q}}{\partial \delta \xi} = [\mathbf{I}, -\mathbf{q}^{\wedge}]$$

$$\mathbf{p}_2 = \begin{bmatrix} u_2 \\ v_2 \\ 1 \end{bmatrix} = \frac{1}{Z_2} \mathbf{K}(\mathbf{R}\mathbf{p} + \mathbf{t}) = \frac{1}{Z_2} \mathbf{K}(\mathbf{T}\mathbf{p})$$

$$u = \frac{f_x X}{Z}, \quad v = \frac{f_y X}{Z}$$

$$\begin{aligned} \frac{\partial(Rp)}{\partial \psi} &= \lim_{\psi \rightarrow 0} \frac{\exp(\psi^{\wedge}) \exp(\phi^{\wedge}) p - \exp(\phi^{\wedge}) p}{\psi} \\ &= \lim_{\psi \rightarrow 0} \frac{(I + \psi^{\wedge}) \exp(\phi^{\wedge}) p - \exp(\phi^{\wedge}) p}{\psi} \\ &= \lim_{\psi \rightarrow 0} \frac{\psi^{\wedge} R p}{\psi} = \lim_{\psi \rightarrow 0} \frac{-(R p)^{\wedge} \psi}{\psi} = -(R p)^{\wedge} \end{aligned}$$

Direct Method

$$\frac{\partial e}{\partial \mathbf{T}} = \frac{\partial \mathbf{I}_2}{\partial \mathbf{u}} \frac{\partial \mathbf{u}}{\partial \mathbf{q}} \frac{\partial \mathbf{q}}{\partial \delta \xi} \delta \xi$$

$$\frac{\partial \mathbf{u}}{\partial \delta \xi} = \begin{bmatrix} \frac{f_x}{Z} & 0 & -\frac{f_x X}{Z^2} & -\frac{f_x XY}{Z^2} & f_x + \frac{f_x X^2}{Z^2} & -\frac{f_x Y}{Z} \\ 0 & \frac{f_y}{Z} & -\frac{f_y Y}{Z^2} & -f_y - \frac{f_y Y^2}{Z^2} & \frac{f_y XY}{Z^2} & \frac{f_y X}{Z} \end{bmatrix}$$

$$\mathbf{J} = -\frac{\partial \mathbf{I}_2}{\partial \mathbf{u}} \frac{\partial \mathbf{u}}{\partial \delta \xi}$$

Apply Gauss Newton or Levenberg–Marquardt algorithm to solve the optimization problem.