



Multiple View Geometry: Exercise 4

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Wednesdays 16:00–18:15 at Hörsaal 2, "Interims I"
(5620.01.102), and on RBG Live

Exercise: May 31, 2023

The Lucas-Kanade method

The weighted Lucas-Kanade energy $E(\mathbf{v})$ is defined as

$$E(\mathbf{v}) = \int_{W(\mathbf{x})} G(\mathbf{x} - \mathbf{x}') \left\| \nabla I(\mathbf{x}', t)^\top \mathbf{v} + \partial_t I(\mathbf{x}', t) \right\|^2 d\mathbf{x}'.$$

Assume that the weighting function G is chosen such that $G(\mathbf{x} - \mathbf{x}') = 0$ for any $\mathbf{x}' \notin W(\mathbf{x})$. In the following, we note $I_t = \partial_t I$ and $(I_{x_1}, I_{x_2})^\top = \nabla I$.

1. Prove that the minimizer \mathbf{b} of $E(\mathbf{v})$ can be written as

$$\mathbf{b} = -M^{-1}\mathbf{q}$$

where the entries of M and \mathbf{q} are given by

$$m_{ij} = G * (I_{x_i} \cdot I_{x_j}) \quad \text{and} \quad q_i = G * (I_{x_i} \cdot I_t)$$

2. Show that if the gradient direction is constant in $W(\mathbf{x})$, i.e. $\nabla I(\mathbf{x}', t) = \alpha(\mathbf{x}', t)\mathbf{u}$ for a scalar function α and a 2D vector \mathbf{u} , M is not invertible.

Explain how this observation is related to the aperture problem.

Note: In the formalism of Lucas and Kanade, one cannot always estimate a translational motion. This problem is often referred to as the aperture problem. It arises for example, if the region in the window $W(x)$ around the point x has entirely constant intensity (for example a white wall), because then $\delta I(x) = 0$ and $I_t(x) = 0$ for all points in the window.

3. Write down explicit expressions for the two components b_1 and b_2 of the minimizer in terms of m_{ij} and q_i .

Note: $G * A$ denotes the convolution of image A with a kernel $G : \mathbb{R}^2 \rightarrow \mathbb{R}$ and is defined as

$$G * A = \int_{\mathbb{R}^2} G(\mathbf{x} - \mathbf{x}') A(\mathbf{x}') d\mathbf{x}'.$$

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$$\begin{aligned} E(\mathbf{v}) &= \int_{W(\mathbf{x})} G(\mathbf{x} - \mathbf{x}') \left(\nabla I(\mathbf{x}', t)^\top \mathbf{v} \right)^2 d\mathbf{x}' \\ &\quad + \int_{W(\mathbf{x})} G(\mathbf{x} - \mathbf{x}') \left(2 \nabla I(\mathbf{x}', t)^\top \mathbf{v} \cdot \partial_t I(\mathbf{x}', t) \right) d\mathbf{x}' \\ &\quad + \int_{W(\mathbf{x})} G(\mathbf{x} - \mathbf{x}') \left(\partial_t I(\mathbf{x}', t) \right)^2 d\mathbf{x}' \end{aligned}$$

$$\begin{aligned} \frac{\partial E(\mathbf{v})}{\partial \mathbf{v}} &= \int_{W(\mathbf{x})} G(\mathbf{x} - \mathbf{x}') \left[2 \nabla I(\mathbf{x}', t)^\top \mathbf{v} \right] \left[\nabla I(\mathbf{x}', t) \right] d\mathbf{x}' \\ &\quad + \int_{W(\mathbf{x})} G(\mathbf{x} - \mathbf{x}') \left[2 \nabla I(\mathbf{x}', t)^\top \partial_t I(\mathbf{x}', t) \right] d\mathbf{x}' \\ &= 2 G \nabla I^\top \mathbf{v} \nabla I + 2 G \nabla I^\top \partial_t I \\ &= 2 G * (I_{x_1}, I_{x_2})^\top \begin{pmatrix} I_{x_1} \\ I_{x_2} \end{pmatrix}^\top \mathbf{v} + 2 G * (I_{x_1}, I_{x_2}) \cdot I_t \\ &= 2 G * \begin{pmatrix} I_{x_1} I_{x_1} & I_{x_1} I_{x_2} \\ I_{x_2} I_{x_1} & I_{x_2} I_{x_2} \end{pmatrix} \mathbf{v} + 2 G * (I_{x_1} I_t, I_{x_2} I_t) \\ &= 2 M \cdot \mathbf{v} + 2 \mathbf{q} = 0 \\ &\quad \mathbf{b} = -M^{-1} \mathbf{q} \end{aligned}$$

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$$\begin{aligned}
 &= 2\gamma \nabla I^T \nabla I + 2\gamma \nabla I^T \partial_t I \\
 &= 2\gamma \alpha^2 \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} \begin{pmatrix} u_1 & u_2 \end{pmatrix} + 2\gamma \alpha (u_1 I_t, u_2 I_t) \\
 &= 2\gamma \alpha^2 \begin{pmatrix} u_1^2 & u_1 u_2 \\ u_1 u_2 & u_2^2 \end{pmatrix} = 0
 \end{aligned}$$

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$$b = -M^{-1}q$$

$$= - \frac{1}{\det M} \begin{bmatrix} m_{22} & -m_{21} \\ -m_{12} & m_{11} \end{bmatrix} \begin{bmatrix} q_1 \\ q_2 \end{bmatrix}$$