Image Processing and Computer Vision (IPCV)



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Example Solutions for Classroom Assignment 12 (C12)

Problem 1 (Optic Flow Constraint in 1-D and 3-D)

(a) For a 1-D signal f(x,t), the grey value constancy assumption can be formulated as

$$f(x+u, t+1) - f(x, t) = 0.$$

If u is small and f varies smoothly, we can perform a linearisation by means of first order Taylor expansion:

$$f(x + u, t + 1) \approx f(x, t) + f_x(x, t)u + f_t(x, t).$$

Thus, we obtain the following optic flow constraint:

$$f_x(x,t)u + f_t(x,t) = 0.$$

Since one equation is sufficient to determine one unknown uniquely, no aperture problem exists in the 1-D case. In fact, we can compute u via

$$u = -\frac{f_t}{f_x}.$$

However, there are cases where u cannot be computed: At locations, where $f_x = 0$.

(b) In the case of a 3-D signal f(x, y, z, t), the grey value constancy assumption is given by

$$f(x+u, y+v, z+w, t+1) - f(x, y, z, t) = 0.$$

As in the 1-D case, we can perform a linearisation of this assumptions by means of first order Taylor expansion here. To this end, we assume again that u is small and f only varies smoothly. This give us

$$\begin{split} f(x+u,y+v,z+w,t+1) &\approx f(x,y,z,t) &+ f_x(x,y,z,t)u \\ &+ f_y(x,y,z,t)v \\ &+ f_z(x,y,z,t)w \\ &+ f_t(x,y,z,t). \end{split}$$

The resulting optic flow constraint then reads

$$f_x(x, y, z, t)u + f_y(x, y, z, t)v + f_z(x, y, z, t)w + f_t(x, y, z, t) = 0.$$

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This time the aperture problem is present: We have three unknowns but only one equation. Taking a closer look at the previous constraint, one can see that it is actually (save for the missing normalisation) the Hesse normal form of a plane that contains all possible solutions at a point $(x,y,z,t)^{\top}$. This plane has unit normal vector $\frac{(f_x,f_y,f_z)^{\top}}{\|(f_x,f_y,f_z)^{\top}\|}$ and distance $\frac{f_t(x,y,z,t)}{\|(f_x,f_y,f_z)^{\top}\|}$ from the origin.

Let us now embed this optic flow constraints in a Bigün–like approach. Then, we obtain

$$E(\mathbf{a}) = \int_{B_{\rho}(x_0, y_0, z_0, t_0)} \left(f_x a_1 + f_y a_2 + f_z a_3 + f_t a_4 \right)^2 dx dy dz dt,$$

where the last component of the minimising vector has to be normalised after the computation to 1. This yields the 3-D optic flow

$$u = \frac{a_1}{a_4}, \qquad v = \frac{a_2}{a_4}, \qquad w = \frac{a_3}{a_4}.$$

As in the original method of Bigün *et al.* the approach can be formulated as a quadratic form. In the 3-D case this form is given by

$$E(\mathbf{a}) = \mathbf{a}^{\mathsf{T}} \hat{\mathbf{J}}_{\rho} \, \mathbf{a} \,,$$

with the 3-D spatiotemporal structure tensor

$$\hat{\mathbf{J}}_{\rho} = \begin{pmatrix} K_{\rho} * (f_{x}^{2}) & K_{\rho} * (f_{x}f_{y}) & K_{\rho} * (f_{x}f_{z}) & K_{\rho} * (f_{x}f_{t}) \\ K_{\rho} * (f_{x}f_{y}) & K_{\rho} * (f_{y}^{2}) & K_{\rho} * (f_{y}f_{z}) & K_{\rho} * (f_{y}f_{t}) \\ K_{\rho} * (f_{x}f_{z}) & K_{\rho} * (f_{y}f_{z}) & K_{\rho} * (f_{z}^{2}) & K_{\rho} * (f_{z}f_{t}) \\ K_{\rho} * (f_{x}f_{t}) & K_{\rho} * (f_{y}f_{t}) & K_{\rho} * (f_{z}f_{t}) & K_{\rho} * (f_{z}^{2}) \end{pmatrix}.$$

Let $\mu_1 \geq \mu_2 \geq \mu_3 \geq \mu_4 \geq 0$ be the eigenvalues of $\hat{\mathbf{J}}_{\rho}$. Then, we have five different cases that are processed in the following order

- No eigenvalue is large $(\operatorname{tr} \hat{\mathbf{J}}_{\rho} = \hat{j}_{11} + \hat{j}_{22} + \hat{j}_{33} + \hat{j}_{44} \leq \tau_1)$: There is a homogeneous spatiotemporal volume and thus not sufficient local information to compute the optic flow.
- All eigenvalues are large $(\mu_4 \ge \tau_2)$: Either the grey value constancy assumption or the assumption of a constant flow is violated.
- One eigenvalue is large $(\mu_2 \leq \tau_3)$: There is an *edge*, the solution lies on a plane.

- Two eigenvalues are large $(\mu_3 \le \tau_4)$: There is a 2-D corner, the solution lies on a line.
- Three eigenvalues are large (the remaining case): There is a 3-D corner. Then, we have to compute the eigenvector to the smallest eigenvalue of $\hat{\mathbf{J}}_{\rho}$ and normalise it so that its last component becomes 1.