

Feature Tracking and Optical Flow

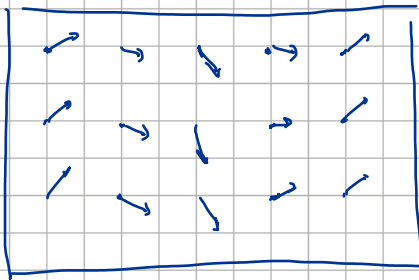
How do we detect and track motion in a scene.

Two basic methods:

1. Optic Flow
2. Feature tracking.

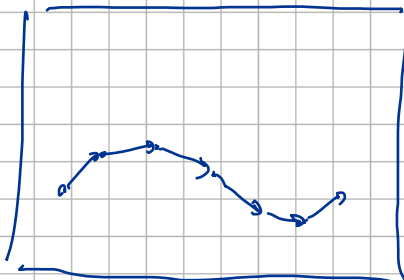
The two methods are intimately related.

Optic Flow:



motion of pixels
is computed at
fixed locations in
the image

Feature tracking



A fixed feature
is tracked over
time.

Both methods require a model for how pixels move from one image to the next.

Small baseline

$$\underbrace{I_1(x)}_{\text{image intensity in image \#1 at pixel } x} = \underbrace{I_2(x + \Delta x)}_{\text{image intensity in image \#2 at } (x + \Delta x) \text{ where } \Delta x \text{ is new location of } x \text{ in image \#2.}}$$

image intensity
in image #1
at pixel x

image intensity
in image #2
at $(x + \Delta x)$ where
 Δx is new location
of x in image #2.

For short period of time we have

$$\Delta x = u \, dt$$

↑
pixel motion (pixels/sec)

Then

$$I(x(t), t) = I(x(t) + u \, dt, t + dt)$$

$$\approx I(x(t), t) + \left(\frac{\partial I}{\partial x} \right)_{x(t), t} u + \left(\frac{\partial I}{\partial t} \right)_{x(t), t} dt + \dots$$

↑
Taylor series

Neglecting the H.O.T and letting

$$\nabla I \approx \begin{pmatrix} I_x \\ I_y \end{pmatrix} = \begin{pmatrix} \frac{\partial I}{\partial x} \\ \frac{\partial I}{\partial y} \end{pmatrix}_{x(t), t} \quad \text{and}$$

$$I_t = \frac{\partial I}{\partial t} \bigg|_{x(t), t} \quad \text{given}$$

$$\boxed{\nabla I^T u + I_t = 0}$$

Note 1 - equation is two unknowns.

∇I is called the image gradient and can be found using, for example, the

Sobel operators:

$$G_x = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix} \quad (\text{Horizontal changes})$$

$$G_y = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix} \quad (\text{Vertical changes})$$

i.e. $I_x(x)$ is found by multiply the 3×3 window around x by G_x

I_t is found as

$$I_t \approx I_2(x, t + \Delta t) - I_1(x, t)$$

Since this is one equation in two unknowns, and is susceptible to noise, the motion vector u is found as the solution of a least squares problem:

$$u = \arg \min \sum_{\tilde{x} \in W(x)} \left| \nabla I(\tilde{x})^T u + I_t(\tilde{x}) \right|^2$$

$\therefore u$ satisfies

$$\underbrace{\begin{bmatrix} \sum I_x^2 & \sum I_x I_y \\ \sum I_x I_y & \sum I_y^2 \end{bmatrix}}_{G \in \mathbb{R}^{2 \times 2}} u + \underbrace{\begin{bmatrix} \sum I_x I_t \\ \sum I_y I_t \end{bmatrix}}_{b \in \mathbb{R}^{2 \times 1}} = 0$$

or

$$\boxed{u = -G^{-1}b}$$

Of course this depends on G being full rank.

If $G(x)$ is invertible then we say that x is a "feature point."

< Algorithm 4.1 Feature Tracking / Optical Flow >