

COMP4901L Assignment7 Writeup

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Part 1 Theory Questions

Q1.1: Calculating the Jacobian

Assuming

$$\mathbf{W}(\mathbf{x}; \mathbf{p}) = \begin{bmatrix} \mathbf{W}_x \\ \mathbf{W}_y \\ \mathbf{W}_1 \end{bmatrix} \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} 1 + p_1 & p_3 & p_5 \\ p_2 & 1 + p_4 & p_6 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \\ 1 \end{bmatrix}$$

and,

$$\mathbf{p} = [p_1 \quad p_2 \quad p_3 \quad p_4 \quad p_5 \quad p_6]^T$$

$$\begin{aligned} \frac{\partial \mathbf{W}}{\partial \mathbf{p}} &= \begin{bmatrix} \frac{\partial \mathbf{W}_u}{\partial p_1} & \cdots & \frac{\partial \mathbf{W}_u}{\partial p_6} \\ \frac{\partial \mathbf{W}_v}{\partial p_1} & \cdots & \frac{\partial \mathbf{W}_v}{\partial p_6} \end{bmatrix} \\ &= \begin{bmatrix} u & 0 & v & 0 & 1 & 0 \\ 0 & u & 0 & v & 0 & 1 \end{bmatrix} \end{aligned}$$

$$L = \sum_{\mathbf{x}} [\mathbf{T}(\mathbf{x}) - \mathbf{I}(\mathbf{W})]^2$$

$$\begin{aligned} \mathbf{J} &= \frac{\partial L}{\partial \mathbf{p}} = \frac{\partial L}{\partial \mathbf{I}(\mathbf{W})} \frac{\partial \mathbf{I}(\mathbf{W})}{\partial \mathbf{W}} \frac{\partial \mathbf{W}}{\partial \mathbf{p}} \\ &= -2 \sum_{\mathbf{x}} [\mathbf{T}(\mathbf{x}) - \mathbf{I}(\mathbf{W})] \begin{bmatrix} \frac{\partial \mathbf{I}}{\partial u} & \frac{\partial \mathbf{I}}{\partial v} \end{bmatrix} \begin{bmatrix} u & 0 & v & 0 & 1 & 0 \\ 0 & u & 0 & v & 0 & 1 \end{bmatrix} \end{aligned}$$

Q1.2: Computational complexity

Initialization step

Evaluate $\nabla \mathbf{T}$, which the runtime is $O(n)$

Evaluate $\mathbf{J} = \frac{\partial \mathbf{W}}{\partial \mathbf{p}}$ at $\mathbf{p} = \mathbf{0}$,

To compute $\nabla \mathbf{T} \frac{\partial \mathbf{W}}{\partial \mathbf{p}}$, since $\nabla \mathbf{T}$ is of dimension $(n \times 2)$ and $\frac{\partial \mathbf{W}}{\partial \mathbf{p}}$ is of dimension $(2 \times p)$, this multiplication will have computational cost $O(np)$.

To compute Hessian matrix $\mathbf{H} = \mathbf{J}^T \mathbf{J}$, since \mathbf{J} is of dimension $(n \times p)$, this multiplication will have computational cost $O(np^2)$

Thus the initialization cost will be $O(np + np^2) = O(np^2)$.

Incremental step

First compute $\mathbf{I}(\mathbf{W}(\mathbf{x}; \mathbf{p}))$, which is $O(np)$.

Then compute $[\mathbf{I}(\mathbf{W}(\mathbf{x}; \mathbf{p})) - \mathbf{T}]$, which is $O(n)$.

Then compute $\sum_{\mathbf{x}} (\nabla \mathbf{T} \frac{\partial \mathbf{W}}{\partial \mathbf{p}})^T [\mathbf{I}(\mathbf{W}(\mathbf{x}; \mathbf{p})) - \mathbf{T}(\mathbf{x})]$, which is $O(pn)$.

Then multiply previous result with inverse of Hessian to get $\Delta \mathbf{p}$, which is $O(p^3)$.

Then update the warp function, which is $O(p^2)$

Total $O(pn + p^3)$

Part 2 Lucas-Kanade Tracker

Q2.1

video in `lk_results/`. initial rects:

car: `[180 180 36 36]`

landing: `[552 126 20 20]`

Q2.3

video in `mb_results`. initial rects:

car: `[126 102 208 177]`

landing: `[440 79 118 58]`

Q2.4

Small templates on spots or corners generally work best for Lucas-Kanade. Anything that would not exhibit the barber-pole problem. The tracker tends to breakdown when large motions occur or when the brightness levels suddenly changed because there would be big different everywhere inside the tracker so it difficult for the Jacobian to tell which direction is the steepest

Part 3

Q3.1X Robust LK

Video in `ilk_results`, initial rect:

car: `[158 144 92 66]`

landing: `[440 79 118 58]`

use `ec/ilk_demo.m` to run

Q3.2X Pyramid LK

Video in `plk_results`, initial rect:

car: `[188 175 85 34]`

landing: `[440 82 52 44]`

use `ec/plk_demo.m` to run

Best Videos

For the best outputs, the following methods are used for the clips in our `results/` folder.

- Car: Robust LK

- Landing: Pyramid LK