

# Lecture-15

## Discussion of Program-2

### Lucas & Kanade (Least Squares)

- Optical flow eq

$$f_x u + f_y v = -f_t$$

- Consider 3 by 3 window

$$f_{x1} u + f_{y1} v = -f_{t1}$$

⋮

$$f_{x9} u + f_{y9} v = -f_{t9}$$

$$\begin{bmatrix} f_{x1} & f_{y1} \\ \vdots & \vdots \\ f_{x9} & f_{y9} \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} -f_{t1} \\ \vdots \\ -f_{t9} \end{bmatrix}$$

$$\mathbf{A} \mathbf{u} = \mathbf{f}_t$$

## Lucas & Kanade

$$\mathbf{A}\mathbf{u} = \mathbf{f}_t$$

$$\mathbf{A}^T \mathbf{A}\mathbf{u} = \mathbf{A}^T \mathbf{f}_t$$

$$\mathbf{u} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{f}_t$$



$$\min \sum_{i=-2}^2 \sum_{j=-2}^2 (f_{xi}u + f_{yi}v + f_{ti})^2$$

## Lucas & Kanade

$$\min \sum_{i=-2}^2 \sum_{j=-2}^2 (f_{xi}u + f_{yi}v + f_{ti})^2$$



$$\min \quad \Downarrow$$

$$\sum (f_{xi}u + f_{yi}v + f_{ti})f_{xi} = 0$$

$$\sum f_{xi}^2 u + \sum f_{xi}f_{yi}v + \sum f_{xi}f_{ti} = 0$$

$$\sum f_{yi}f_{yi}u + \sum f_{yi}^2 v + \sum f_{yi}f_{ti} = 0$$

$$\sum (f_{xi}u + f_{yi}v + f_{ti})f_{yi} = 0$$



$$\sum (f_{xi}f_{xi}u + f_{xi}f_{yi}v + f_{xi}f_{ti}) = 0$$

$$\begin{bmatrix} \sum f_{xi}^2 & \sum f_{xi}f_{yi} \\ \sum f_{xi}f_{yi} & \sum f_{yi}^2 \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} -\sum f_{xi}f_{ti} \\ -\sum f_{yi}f_{ti} \end{bmatrix}$$

$$\sum (f_{yi}f_{xi}u + f_{yi}f_{yi}v + f_{yi}f_{ti}) = 0$$

## Lucas & Kanade

$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} \sum f_{xi}^2 & \sum f_{xi} f_{yi} \\ \sum f_{xi} f_{yi} & \sum f_{yi}^2 \end{bmatrix}^{-1} \begin{bmatrix} -\sum f_{xi} f_{ti} \\ -\sum f_{yi} f_{ti} \end{bmatrix}$$

## Least Squares Fit


$$a_j \cdot x - b_j = 0$$

$$\sum_j (a_j \cdot x - b_j)^2$$

$$\sum_j (a_j a_j^T) x = \sum_j b_j a_j$$

$$a_j^T x - b_j = 0$$

$$\min \sum_j (a_j^T x - b_j)^2$$

 differentiate

$$\sum_j 2(a_j^T x - b_j) a_j = 0$$

$$\sum_j a_j (a_j^T x - b_j) = 0$$

$$\sum_j (a_j a_j^T) x = \sum_j b_j a_j$$

## Anandan

$$\mathbf{u}(\mathbf{x}) = \mathbf{X}(\mathbf{x})\mathbf{a}$$

$$E(d\mathbf{a}) = \sum_x (f_t + f_x^T d\mathbf{u})^2$$

$$E(d\mathbf{a}) = \sum_x (f_t + f_x^T \mathbf{X} d\mathbf{a})^2$$

min



$$\left[ \sum \mathbf{X}^T (\mathbf{f}_x) (\mathbf{f}_x)^T \mathbf{X} \right] d\mathbf{a} = - \sum \mathbf{X}^T \mathbf{f}_x f_t$$

$$A\mathbf{x} = \mathbf{b}$$

$$f_x = \begin{bmatrix} f_x \\ f_y \end{bmatrix}$$

Optical flow constraint eq

$$f_x u + f_y v = -f_t$$

Homework 2: Show this due 9/21

$$a_j^T x - b_j = 0$$

$$\min \sum_j (a_j^T x - b_j)^2$$

$$\sum_j (a_j a_j^T) x = \sum_j b_j a_j$$

## Rotation Around an Arbitrary Axis (Rodriguez's Formula)

$$V = (V \cdot n)n + (V - (V \cdot n)n)$$

HW 4.1

$$V'_\perp = \cos q (V - (V \cdot n)n) + \sin q (n \times (V - (V \cdot n)n))$$

$$V'_\parallel = (V \cdot n)n$$

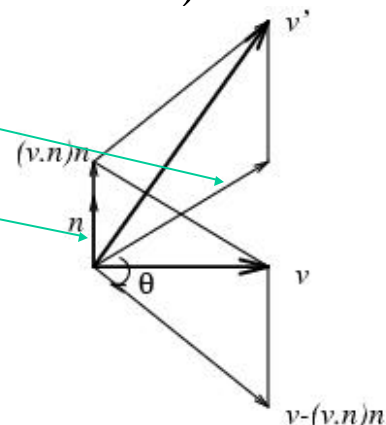
$$V' = V'_\perp + V'_\parallel$$

$$V' = \cos q V + \sin q n \times V + (1 - \cos q)(V \cdot n)n$$

$$V' = V + \sin q n \times V + (1 - \cos q)n \times (n \times V)$$

$$n \times (n \times V) = (V \cdot n)n - V$$

$$n \times (n \times V) + V = (V \cdot n)n$$



## Rotation Around an Arbitrary Axis (Rodriguez's Formula)

$$V' = V + \sin \mathbf{q} \, n \times V + (1 - \cos \mathbf{q}) \, n \times (n \times V)$$

$$V' = R(n, \mathbf{q})V$$

$$R(n, \mathbf{q}) = I + \sin \mathbf{q} \, X(n) + (1 - \cos \mathbf{q}) \, X^2(n) \quad \text{HW 4.2}$$

$$X(n) = \begin{bmatrix} 0 & -n_z & n_x \\ n_z & 0 & -n_x \\ -n_y & n_x & 0 \end{bmatrix}$$

## Rotation Around an Arbitrary Axis (Rodriguez's Formula)

$$r = \|r\| \frac{r}{\|r\|} = \mathbf{q}n$$

$$R(r, \mathbf{q}) = I + \sin \mathbf{q} \frac{X(r)}{\|r\|} + (1 - \cos \mathbf{q}) \frac{X^2(r)}{\|r\|^2}$$

$$X(r) = \begin{bmatrix} 0 & -r_z & r_x \\ r_z & 0 & -r_x \\ -r_y & r_x & 0 \end{bmatrix}$$

$$R(n, \mathbf{q}) = I + \sin \mathbf{q} X(n) + (1 - \cos \mathbf{q}) X^2(n)$$

$$R(m) = I + \sin \mathbf{q} \frac{X(m)}{\mathbf{q}} + \frac{X^2(m)}{\mathbf{q}^2} (1 - \cos \mathbf{q})$$

$$\tilde{R} \approx I + X(m) \quad m = \mathbf{q} = (m_x, m_y, m_z)$$

$$R^{it+1} \leftarrow \tilde{R} R^{it}$$

## 3D Rigid Transformation

$$\begin{bmatrix} X' \\ Y' \\ Z' \end{bmatrix} = R \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + T = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + \begin{bmatrix} T_X \\ T_Y \\ T_Z \end{bmatrix}$$

Camera coordinates

Wireframe coordinates

$$x_i'^k = f^k \frac{X_i'^k}{Z_i'^k}, \quad y_i'^k = f^k \frac{Y_i'^k}{Z_i'^k} \quad \text{perspective}$$

$K$ =camera no.

## 3D Rigid Transformation

$$x_i'^k = f^k \frac{X_i'^k}{Z_i'^k}, y_i'^k = f^k \frac{Y_i'^k}{Z_i'^k}$$

$$x_i'^k = f_k \frac{r_{11}^k X_i + r_{12}^k Y_i + r_{13}^k Z_i + T_X^k}{r_{31}^k X_i + r_{32}^k Y_i + r_{33}^k Z_i + T_Z^k}$$

$$y_i'^k = f_k \frac{r_{21}^k X_i + r_{22}^k Y_i + r_{23}^k Z_i + T_Y^k}{r_{31}^k X_i + r_{32}^k Y_i + r_{33}^k Z_i + T_Z^k}$$

## Model Fitting

$$x_i'^k = f_k \frac{\mathbf{r}_x^k \mathbf{p}_i + T_X^k}{\mathbf{r}_z^k \mathbf{p}_i + T_Z^k}$$

$$y_i'^k = f_k \frac{\mathbf{r}_y^k \mathbf{p}_i + T_Y^k}{\mathbf{r}_z^k \mathbf{p}_i + T_Z^k}$$

## Model Fitting

$$\begin{aligned}
 x_i'^k &= f_k \frac{\mathbf{r}_x^k \mathbf{p}_i + T_X^k}{\mathbf{r}_z^k \mathbf{p}_i + T_Z^k} \\
 y_i'^k &= f_k \frac{\mathbf{r}_y^k \mathbf{p}_i + T_Y^k}{\mathbf{r}_z^k \mathbf{p}_i + T_Z^k} \quad \mathbf{h}^k = \frac{1}{T_Z^k}, s^k = f^k \mathbf{h}^k \\
 x_i'^k &= s_k \frac{\mathbf{r}_x^k \mathbf{p}_i + T_X^k}{1 + \mathbf{h}^k \mathbf{r}_z^k \mathbf{p}_i} \\
 y_i'^k &= s_k \frac{\mathbf{r}_y^k \mathbf{p}_i + T_Y^k}{1 + \mathbf{h}^k \mathbf{r}_z^k \mathbf{p}_i}
 \end{aligned}$$

## Model Fitting

$$\begin{aligned}
 x_i'^k &= s_k \frac{\mathbf{r}_x^k \mathbf{p}_i + T_X^k}{1 + \mathbf{h}^k \mathbf{r}_z^k \mathbf{p}_i} \\
 y_i'^k &= s_k \frac{\mathbf{r}_y^k \mathbf{p}_i + T_Y^k}{1 + \mathbf{h}^k \mathbf{r}_z^k \mathbf{p}_i} \quad w_i^k = (1 + \mathbf{h}^k (\mathbf{r}_z^k \cdot \mathbf{p}_i))^{-1}
 \end{aligned}$$

$$w_i^k (x_i'^k + x_i'^k \mathbf{h}^k (\mathbf{r}_z^k \cdot \mathbf{p}_i) - s^k (\mathbf{r}_X^k \cdot \mathbf{p}_i + T_X^k)) = 0$$

$$w_i^k (y_i'^k + y_i'^k \mathbf{h}^k (\mathbf{r}_z^k \cdot \mathbf{p}_i) - s^k (\mathbf{r}_Y^k \cdot \mathbf{p}_i + T_Y^k)) = 0$$



## Model Fitting

- Solve for unknowns in five steps:

$$s^k; \mathbf{p}_i; \mathbf{R}^k; T_X^k, T_Y^k; \mathbf{h}^k$$

- Use linear least squares fit.
- When solving for an unknown, assume other parameters are known.

$$a_j \cdot x - b_j = 0$$

$$\sum_j (a_j \cdot x - b_j)^2 \quad w_i^k (x_i^k + x_i^k \mathbf{h}^k (\mathbf{r}_z^k \cdot \mathbf{p}_i) - s^k (\mathbf{r}_x^k \cdot \mathbf{p}_i + T_X^k)) = 0$$

$$w_i^k (y_i^k + y_i^k \mathbf{h}^k (\mathbf{r}_z^k \cdot \mathbf{p}_i) - s^k (\mathbf{r}_y^k \cdot \mathbf{p}_i + T_Y^k)) = 0$$

$$\sum_j (a_j a_j^T) x = \sum_j b_j a_j$$

Update for  $s^k$

$$a_{2k+0} = w_i^k (r_x^k \cdot p_i + t_x^k) \quad \mathbf{b}_{2k+0} = w_i^k (x_i^k + x_i^k \mathbf{h}^k (r_z^k \cdot p_i))$$

$$a_{2k+1} = w_i^k (r_y^k \cdot p_i + t_y^k) \quad \mathbf{b}_{2k+1} = w_i^k (y_i^k + y_i^k \mathbf{h}^k (r_z^k \cdot p_i))$$

$$x'_i + x'_i \mathbf{h}(\mathbf{r}_z \cdot \mathbf{p}_i) - s(\mathbf{r}_x \cdot \mathbf{p}_i + T_x) = 0$$

$$y'_i + y'_i \mathbf{h}(\mathbf{r}_z \cdot \mathbf{p}_i) - s(\mathbf{r}_y \cdot \mathbf{p}_i + T_y) = 0$$

**Solving for  $T_x$  and  $T_y$**

$$sT_x = x'_i + x'_i \mathbf{h}(\mathbf{r}_z \cdot \mathbf{p}_i) - s(\mathbf{r}_x \cdot \mathbf{p}_i)$$

$$a_0 = s, b_0 = x'_i + x'_i \mathbf{h}(\mathbf{r}_z \cdot \mathbf{p}_i) - s(\mathbf{r}_x \cdot \mathbf{p}_i)$$

$$sT_y = y'_i + y'_i \mathbf{h}(\mathbf{r}_z \cdot \mathbf{p}_i) - s(\mathbf{r}_y \cdot \mathbf{p}_i)$$

$$a_0 = s, b_0 = y'_i + y'_i \mathbf{h}(\mathbf{r}_z \cdot \mathbf{p}_i) - s(\mathbf{r}_y \cdot \mathbf{p}_i)$$

$$x'_i + x'_i \mathbf{h}(\mathbf{r}_z \cdot \mathbf{p}_i) - s(\mathbf{r}_x \cdot \mathbf{p}_i + T_x) = 0$$

$$y'_i + y'_i \mathbf{h}(\mathbf{r}_z \cdot \mathbf{p}_i) - s(\mathbf{r}_y \cdot \mathbf{p}_i + T_y) = 0$$

**Solving for  $\mathbf{h}$**

$$a_0 = x'_i(\mathbf{r}_z \cdot \mathbf{p}_i), b_0 = s(\mathbf{r}_x \cdot \mathbf{p}_i + T_x) - x'_i$$

$$a_1 = y'_i(\mathbf{r}_z \cdot \mathbf{p}_i), b_1 = s(\mathbf{r}_y \cdot \mathbf{p}_i + T_y) - y'_i$$

Solving for rotation:

$$w_i^k (x_i'^k + x_i'^k \mathbf{h}^k (\mathbf{r}_z^k \cdot \mathbf{p}_i) - s^k (\mathbf{r}_x^k \cdot \mathbf{p}_i + T_x^k)) = 0$$

$$w_i^k (y_i'^k + y_i'^k \mathbf{h}^k (\mathbf{r}_z^k \cdot \mathbf{p}_i) - s^k (\mathbf{r}_y^k \cdot \mathbf{p}_i + T_y^k)) = 0$$



$$w_i^k (x_i'^k + x_i'^k \mathbf{h} (\tilde{\mathbf{r}}_z^k \cdot q_i) - s^k (\tilde{\mathbf{r}}_x^k \cdot q_i + T_x^k)) = 0$$

$$w_i^k (y_i'^k + y_i'^k (\tilde{\mathbf{r}}_z^k \cdot q_i) - s^k (\tilde{\mathbf{r}}_y^k \cdot q_i + T_y^k)) = 0$$



$$x_i' + x_i' \mathbf{h} (-m_y q_x + m_x q_y + q_z) - s (q_x - m_z q_y + m_y q_z + T_x) = 0$$

$$y_i' + y_i' \mathbf{h} (-m_y q_x + m_x q_y + q_z) - s (m_z q_x + q_y - m_x q_z + T_y) = 0$$

$$\tilde{\mathbf{r}}_z^k = (-m_y, m_x, 1)$$

$$\tilde{\mathbf{r}}_x^k = (1, -m_z, m_y)$$

$$\tilde{\mathbf{r}}_y^k = (m_z, 1, -m_x)$$

$$q_i = R^{it} p_i$$

$$R^{it+1} \leftarrow \tilde{\mathbf{R}} R^{it}$$

$$x_i' + x_i' \mathbf{h} (-m_y q_x + m_x q_y + q_z) - s (q_x - m_z q_y + m_y q_z + T_x) = 0$$

$$y_i' + y_i' \mathbf{h} (-m_y q_x + m_x q_y + q_z) - s (m_z q_x + q_y - m_x q_z + T_y) = 0$$

$$\begin{bmatrix} x' \mathbf{h} q_y & -x' \mathbf{h} q_x + s q_z & s q_y \end{bmatrix} \begin{bmatrix} m_x \\ m_y \\ m_z \end{bmatrix} = \begin{bmatrix} s T_x - x' - x' \mathbf{h} q_z + s q_x \end{bmatrix}$$

$$\begin{bmatrix} y' \mathbf{h} q_y - s q_z & -y' \mathbf{h} q_x & s q_x \end{bmatrix} \begin{bmatrix} m_x \\ m_y \\ m_z \end{bmatrix} = \begin{bmatrix} s T_y - y' - y' \mathbf{h} q_z + s q_y \end{bmatrix}$$

$$a_0 = \begin{bmatrix} x' \mathbf{h} q_y & -x' \mathbf{h} q_x + s q_z & s q_y \end{bmatrix}, b_0 = \begin{bmatrix} s T_x - x' - x' \mathbf{h} q_z + s q_x \end{bmatrix}$$

$$a_1 = \begin{bmatrix} y' \mathbf{h} q_y - s q_z & -y' \mathbf{h} q_x & s q_x \end{bmatrix}, b_1 = \begin{bmatrix} s T_y - y' - y' \mathbf{h} q_z + s q_y \end{bmatrix}$$

$$\sum_j (a_j a_j^T) x = \sum_j b_j a_j$$