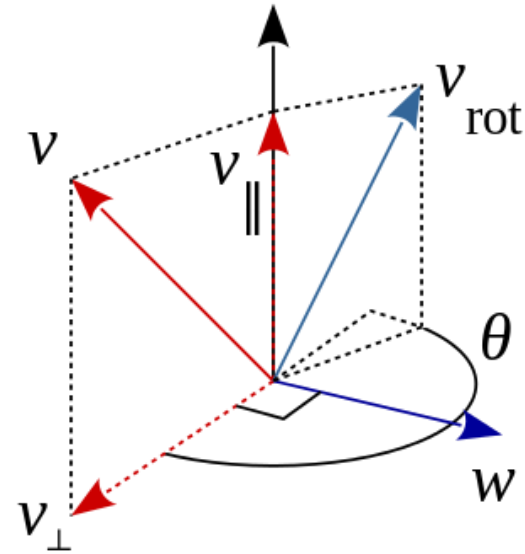


Rodrigues' Rotation Formula k

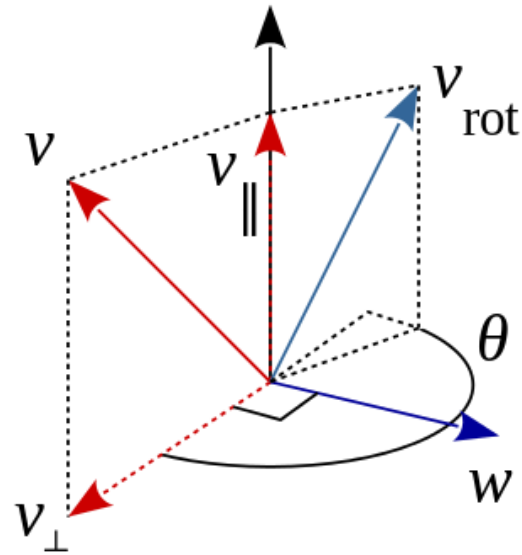
- k – unit vector (rotation axis)
- v – vector to rotate about k by θ
- $v = v_{par} + v_{per}$
- $v_{par} = (v \cdot k)k$ **(1)** [projection of v on k]
- $v_{per} = (v - v_{par})$
 $= v - (k \cdot v)k = (k \cdot k)v - k \cdot v(k) = -k \times (k \times v)$ **(2)**
- $v_{par}^{rot} = v_{par}$ (parallel comp. will not change direction or magnitude Under rotation)
- $|v_{per}^{rot}| = |v_{per}|$ (perpendicular comp. changes its direction but retains its magnitude under rotation)



Rodrigues' Rotation Formula k

- $|v_{per}^{rot}| = |v_{per}|$
 $\Rightarrow v_{per}^{rot} = \cos\theta v_{per} + \sin\theta k \times v_{per}$ **(3)**

- $k \times v_{per} = k \times (v - v_{par})$
 $= k \times v - k \times v_{par}$
 $= k \times v$ **(4)** [as, $k \times v_{par} = 0$]



- Putting in Eq.3, $v_{per}^{rot} = \cos\theta v_{per} + \sin\theta k \times v$

- Finally, the full rotated vector is

$$\begin{aligned}
 v_{rot} &= v_{par}^{rot} + v_{per}^{rot} \\
 &= v_{par} + \cos\theta v_{per} + \sin\theta k \times v \\
 &= v_{par} + \cos\theta (v - v_{par}) + \sin\theta k \times v \\
 &= \cos\theta v + (1 - \cos\theta) v_{par} + \sin\theta k \times v \\
 &= \cos\theta v + (1 - \cos\theta)(k \cdot v)k + \sin\theta k \times v \quad \textbf{(5)}
 \end{aligned}$$

Rodrigues' Rotation Formula

- $k \times v$ can be expressed as a matrix product

$$\begin{bmatrix} (k \times v)_x \\ (k \times v)_y \\ (k \times v)_z \end{bmatrix} = \begin{bmatrix} 0 & -k_z & k_y \\ k_z & 0 & -k_x \\ -k_y & k_x & 0 \end{bmatrix} \begin{bmatrix} v_x \\ v_y \\ v_z \end{bmatrix} = [k]_X v$$

- So, $[k]_X v = k \times v$
- $(k \cdot v)k = v + k \times (k \times v)$
 $= v + [k]_X([k]_X v) = v + [k]_X^2 v$

- Putting in Eq.5

$$\begin{aligned} v_{rot} &= \cos\theta v + (1 - \cos\theta)(v + [k]_X^2 v) + \sin\theta k \times v \\ &= v + (\sin\theta)[k]_X v + (1 - \cos\theta)[k]_X^2 v \\ &= Rv \end{aligned}$$

where, $\mathbf{R} = \mathbf{I} + (\sin\theta)[k]_X + (1 - \cos\theta)[k]_X^2$

Monasse 3-step Rectification

- INPUT : Fundamental Matrix, F by DLT.

$$e = (e_x, e_y, 1)^T \quad \text{Applying, } Fe = 0, \text{ find } e.$$

$$e' = (e'_x, e'_y, 1)^T \quad \text{Applying, } e'^T F = 0, \text{ find } e'.$$

- Orientation of a camera can be adjusted by,

$$H = K R K^{-1}$$

- Since the image is not calibrated,

$$K = \begin{bmatrix} f & 0 & w/2 \\ 0 & f & h/2 \\ 0 & 0 & 1 \end{bmatrix} \quad \text{where, } w = \text{width of the image,}$$

$h = \text{height of the image.}$

Monasse 3-step Rectification

- Step 1:

$$H_1 e = (e_x, e_y, 0)^T = e_1 \quad \text{where, } H_1 = KR_1K^{-1}$$

$$H'_1 e' = (e'_x, e'_y, 0) = e'_1 \quad \text{where, } H'_1 = KR'_1K^{-1}$$

According to Rodrigues' formulae,

$$R_1(\theta, t) = I + \sin \theta [t]_x + (1 - \cos \theta) [t]_x^2$$

where,

$$\cos \theta = \frac{\mathbf{a} \cdot \mathbf{b}}{|\mathbf{a}| |\mathbf{b}|} \quad \text{and}$$

$$\text{rotation axis, } t = \frac{\mathbf{a} \times \mathbf{b}}{|\mathbf{a} \times \mathbf{b}|}$$

$$H_1 e = (e_x, e_y, 0)^T$$

$$KR_1K^{-1} e = (e_x, e_y, 0)^T$$

$$R_1K^{-1} e = K^{-1} (e_x, e_y, 0)^T$$

$$R_1 \mathbf{a} = \mathbf{b}$$

Monasse 3-step Rectification

- **Step 2:**

$$H_2 e_1 = (1, 0, 0)^T = e_2 \quad \text{where, } H_2 = KR_2K^{-1}$$

$$H'_2 e'_1 = (1, 0, 0)^T = e'_2 \quad \text{where, } H'_2 = KR'_2K^{-1}$$

$\therefore H_1, H'_1, H_2, H'_2$ are all parameterized by f .

- **Step 3:**

The remaining relationship between the two cameras of the rectified image is characterized by a rotation, \hat{R} around the baseline.

Finding the Essential Matrix

- According to Zisserman and Hartley,

\hat{F} of a rectified image is given by

$$\hat{F} = K^{-T} [i]_{\times} \hat{R} K^{-1} = K^{-T} \hat{E} K^{-1}$$

$$\therefore \hat{E} = [i]_{\times} \hat{R}$$

\hat{E} is also parameterized by f .

Now, \hat{E} is decomposed into $\hat{E} = UDV^T$

Following the definition of Essential Matrix,

$$\hat{\tilde{E}} = U \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} V^T$$

$$\therefore \hat{\tilde{F}} = K^{-T} \hat{\tilde{E}} K^{-1}$$

The optimization step

$$e_2' \hat{\tilde{F}} e_2 = 0$$

$$(H_2' H_1' e')^T \hat{\tilde{F}} (H_2 H_1 e) = 0$$

$$e'^T H_1'^T H_2'^T \hat{\tilde{F}} H_2 H_1 e = 0 \quad \text{and, } e'^T \tilde{F} e = 0$$

$$\therefore \tilde{F} = H_1'^T H_2'^T \hat{\tilde{F}} H_2 H_1$$

Now an optimization function, S is defined as :

$$S(f) = \sum_{i=1}^N d(x_i', \tilde{F} x_i) + d(x_i, \tilde{F}^T x_i') \quad \text{where, } N \text{ is the no. of pixels in the image.}$$

$d(p, q)$ is the Euclidean distance between p and q .

A minimization of $S(f)$ is done to estimate K in terms of f .

From K , P and P' is estimated.

$$\therefore X = P^+ x \quad \text{or} \quad X = P'^+ x'$$

The idea is to transform both images so that the fundamental matrix gets the form $[i]_{\times}$. Unlike the other methods which directly parameterize the homographies from the constraints $H e = i$, $H' e' = i$ and $H'^T [i]_{\times} H = F$ and find an optimal pair by minimizing a measure of distortion, we shall compute the homography by explicitly rotating each camera around its optical center. The algorithm is decomposed into three steps (Fig. 1):

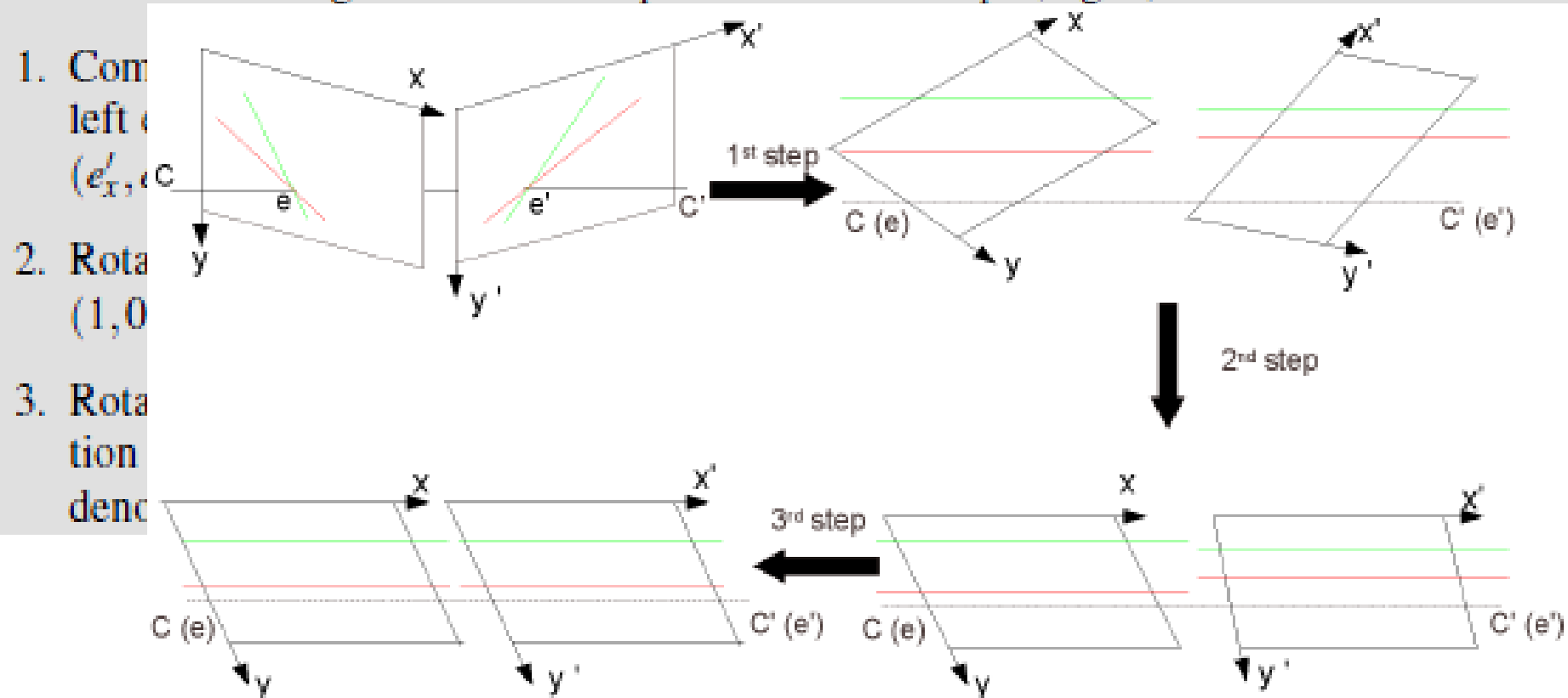


Figure 1: Three-step rectification. First step: the image planes become parallel to CC' . Second step: the images rotate in their own plane to have their epipolar lines also parallel to CC' . Third step: a rotation of one of the image planes around CC' aligns corresponding epipolar lines in both images. Note how the pairs of epipolar lines become aligned.

Input: F , computed using correspondences;
which gives epipoles e and e' ;

Let,

$$\mathbf{x}_1 = K[\mathbf{I} \mid 0]\mathbf{X}; \Rightarrow K^{-1}\mathbf{x}_1 = [\mathbf{I} \mid 0]\mathbf{X}$$

$$\mathbf{x}_2 = K \cdot \mathbf{R}[\mathbf{I} \mid 0]\mathbf{X};$$

$$\Rightarrow \mathbf{x}_2 = K \cdot \mathbf{R}K^{-1}\mathbf{x}_1 = H\mathbf{x}_1;$$

where, **Homography is:**

$$H = K \cdot \mathbf{R}K^{-1}$$

Steps: 1 & 2:

$$\mathbf{H}_1 \mathbf{e} = (e_x, e_y, 0)^T \text{ and } \mathbf{H}'_1 \mathbf{e}' = (e'_x, e'_y, 0)^T$$

$$\mathbf{H}_1 = \mathbf{K}\mathbf{R}\mathbf{K}^{-1} \text{ and } \mathbf{H}'_1 = \mathbf{K}\mathbf{R}'\mathbf{K}^{-1}$$

$$\mathbf{R}\mathbf{K}^{-1}\mathbf{e} = \mathbf{K}^{-1}(e_x, e_y, 0)^T$$

rotates the vector $\mathbf{a} = \mathbf{K}^{-1}\mathbf{e}$ to $\mathbf{b} = \mathbf{K}^{-1}(e_x, e_y, 0)^T$

$$\mathbf{K} = \begin{bmatrix} f & 0 & \frac{w}{2} \\ 0 & f & \frac{h}{2} \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{R}(\theta, \mathbf{t}) = \mathbf{I} + \sin \theta [\mathbf{t}]_{\times} + (1 - \cos \theta) [\mathbf{t}]_{\times}^2$$

minimal angle θ is $\arccos\left(\frac{\mathbf{a} \cdot \mathbf{b}}{|\mathbf{a}||\mathbf{b}|}\right)$ and the rotation axis \mathbf{t} is $\frac{\mathbf{a} \times \mathbf{b}}{|\mathbf{a}||\mathbf{b}|}$

$\mathbf{H}_1, \mathbf{H}'_1, \mathbf{H}_2$ and \mathbf{H}'_2 are all parametrized by f

Step 3: Rotation \hat{R} , of one camera about baseline:

$$\hat{\mathbf{F}} = \mathbf{K}^{-T} [\mathbf{i}]_{\times} \hat{\mathbf{R}}\mathbf{K}^{-1}$$

\mathbf{H}_3 is obtained after obtaining optimal K (or f)

Examples



Image-1

Image-2

Rectified Image