

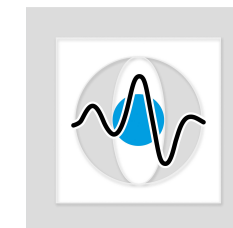
Medical Image Processing for Interventional Applications

Eight Point Algorithm

Online Course – Unit 32

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Pattern Recognition Lab (CS 5)



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The Eight Point Algorithm

Algorithm

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Data Balancing

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The Eight Point Algorithm

read point correspondences $\{(\tilde{\mathbf{p}}_k^i, \tilde{\mathbf{q}}_k^i), k = 1, \dots, N\}$		
rewrite the epipolar constraints in terms of linear equations: $\mathbf{M}\mathbf{e} = 0$		
compute the SVD of $\mathbf{M} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^\top$		
$\sigma_9 > \varepsilon$ (\mathbf{M} has full rank)		
false		true
nullspace of \mathbf{M} defines the components of the essential matrix \mathbf{E}		stop (epipolar constraint not fulfilled)
compute the SVD of $\mathbf{E} = \mathbf{U}'\mathbf{\Sigma}'\mathbf{V}'^\top$		
$\sigma_3 > \varepsilon$ (\mathbf{E} has rank 3)		
false		true
$\sigma_1 - \sigma_2 > \varepsilon$		stop (essential matrix violates rank criterion)
false		
return $\mathbf{E} = \mathbf{U}' \begin{pmatrix} \frac{\sigma_1 + \sigma_2}{2} & 0 & 0 \\ 0 & \frac{\sigma_1 + \sigma_2}{2} & 0 \\ 0 & 0 & 0 \end{pmatrix} \mathbf{V}'^\top$	\emptyset	
stop (essential matrix has different singular values σ_1, σ_2)		

Figure 1: Eight point algorithm for \mathbf{E}

Input Data

For now we ...

- ... have two images of a patient with one camera,
- ... use point features,
- ... assume the correspondence problem to be already solved:

$$\{(\tilde{\mathbf{p}}_k^i, \tilde{\mathbf{q}}_k^i), k = 1, \dots, N\},$$

- i. e., point $\tilde{\mathbf{p}}_k^i$ in image 1 corresponds to point $\tilde{\mathbf{q}}_k^i$ in image 2 [the tilde indicates homogeneous coordinates],
- ... have the points given as normalized homogeneous image coordinates, i. e., the third component is set to 1,
- ... use perspective projection (pinhole camera).

Intrinsic Camera Parameters

- Intrinsic parameters (summarized by the matrix $\mathbf{K} \in \mathbb{R}^{3 \times 3}$) are known.
- Intrinsic parameters do not change when camera moves.
- Origin of image coordinate system does not coincide with the intersection of optical axis and image plane in general.
- Axes of the camera's CCD-Chip are not orthogonal.
- Pixels are non-quadratic.

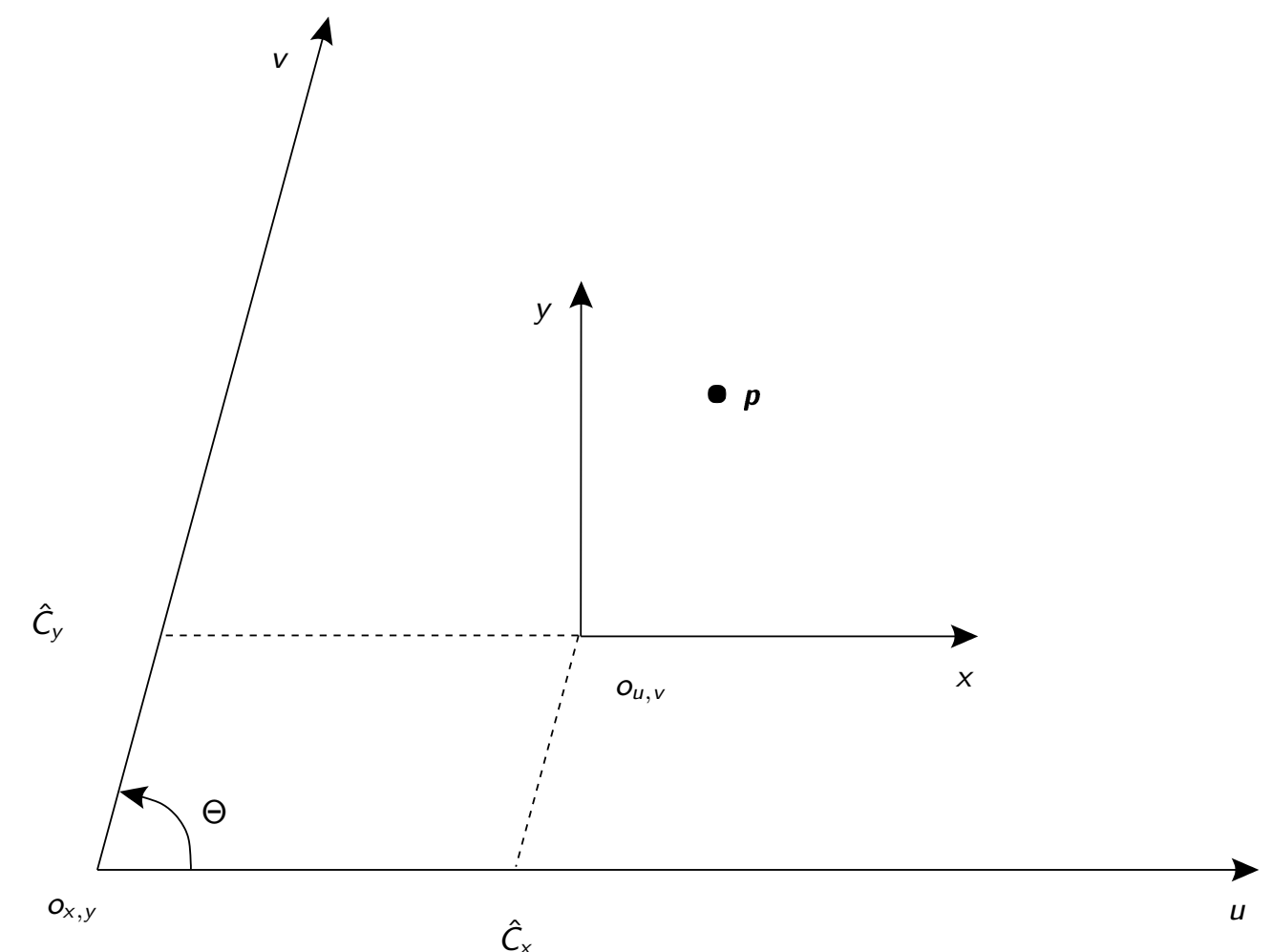


Figure 2: Pixel coordinate system

Coordinate System

(x, y) - coordinate system:

- ideal coordinate system used so far (image coordinate system with origin o^i)

(u, v) - coordinate system:

- real system, in which pixels are addressed, (pixel coordinate system, origin o^p)
- Θ : angle between axes, skew $s = -k_x \tan \Theta$
- k_x, k_y : units of u and v axis, with respect to units in x/y system

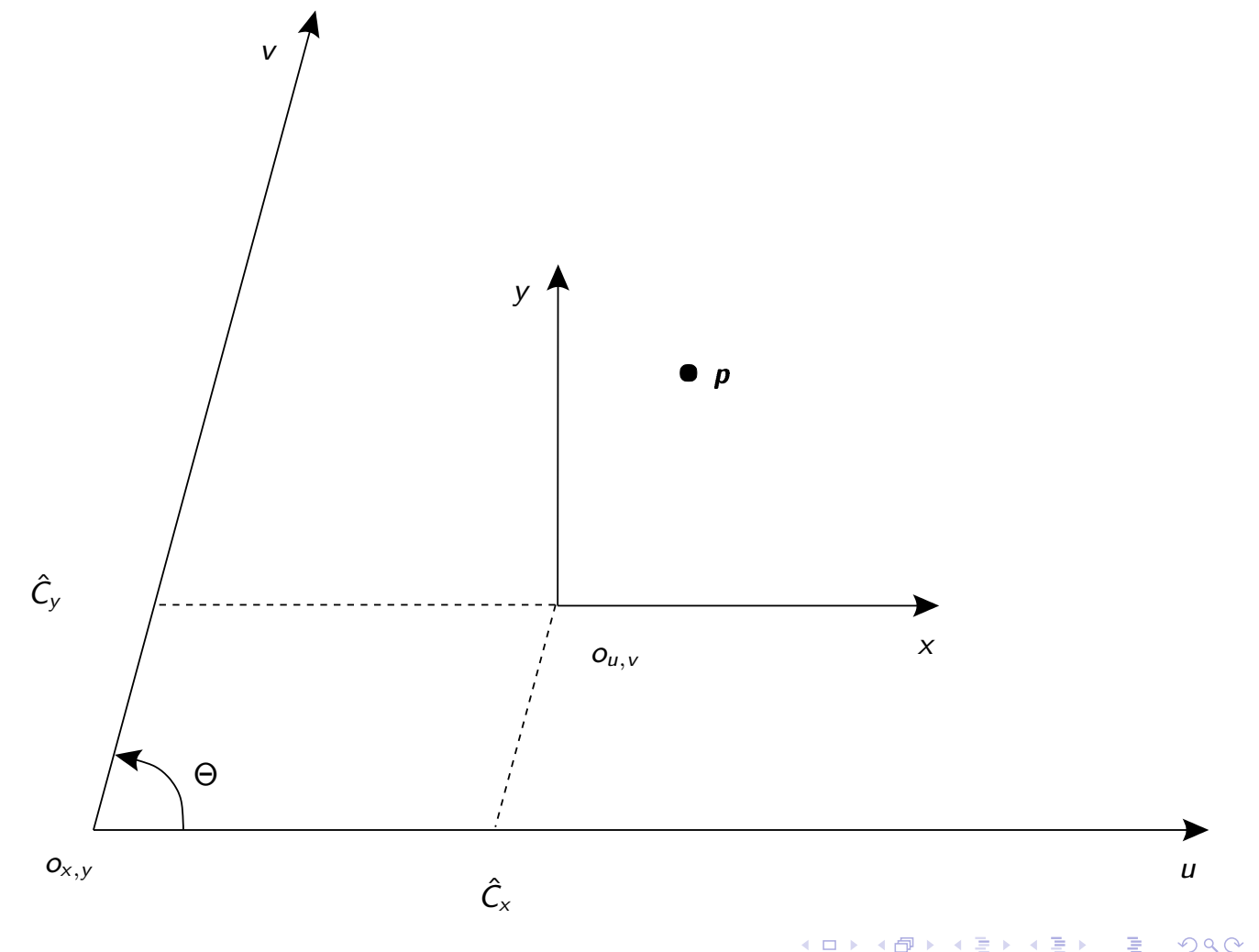


Figure 3: Pixel coordinate system

Fundamental Matrix

If pixel coordinates are used instead of ideal image coordinates

$$\tilde{\mathbf{p}}^p = \mathbf{K} \tilde{\mathbf{p}}^i,$$

we substitute

$$\tilde{\mathbf{p}}^i = \mathbf{K}^{-1} \tilde{\mathbf{p}}^p, \quad \tilde{\mathbf{q}}^i = \mathbf{K}^{-1} \tilde{\mathbf{q}}^p,$$

and get:

$$(\tilde{\mathbf{q}}^p)^T \cdot \underbrace{\left((\mathbf{K}^{-1})^T \cdot \mathbf{E} \cdot \mathbf{K}^{-1} \right)}_{\mathbf{F}} \cdot \tilde{\mathbf{p}}^p = 0.$$

\mathbf{F} is called the ***fundamental matrix***.

Properties of Fundamental Matrix

- \mathbf{F} has rank 2.
- \mathbf{F} encodes intrinsic and extrinsic parameters.
- \mathbf{F} maps a point $\tilde{\mathbf{p}}^p$ to its epipolar line \mathbf{l} in pixel coordinates by $\mathbf{l}^T = \mathbf{F} \cdot \tilde{\mathbf{p}}^p$:

$$\mathbf{l}_2^T = \mathbf{F} \tilde{\mathbf{p}}^p, \quad \mathbf{l}_1^T = \mathbf{F}^T \tilde{\mathbf{q}}^p.$$

- All epipolar lines intersect in the epipole (left epipole $\tilde{\mathbf{e}}_l^p$, right epipole $\tilde{\mathbf{e}}_r^p$):

- computation of the left null space:

$$(\tilde{\mathbf{e}}_r^p)^T \mathbf{l}_1^T = (\tilde{\mathbf{e}}_r^p)^T \mathbf{F} \tilde{\mathbf{p}}^p = 0$$

implies $(\tilde{\mathbf{e}}_r^p)^T \mathbf{F} = 0$,

- computation of the right null space:

$$\mathbf{l}_2 \tilde{\mathbf{e}}_r^p = (\tilde{\mathbf{e}}_r^p \mathbf{F})^T \tilde{\mathbf{q}}^p = 0$$

implies $\mathbf{F}^T \tilde{\mathbf{e}}_l^p = 0$.

Eight Point Algorithm for F

Computation of F :

- We get N equations of the form $(\tilde{\mathbf{q}}_i^p)^T \cdot \mathbf{F} \cdot \tilde{\mathbf{p}}_i^p = 0$.
- This system of equations is **linear** in the components of $\mathbf{F} = [f_{ij}]_{i,j \in \{1,2,3\}}$:

$$\mathbf{M} \cdot \mathbf{f} = 0, \quad \mathbf{f} = \begin{pmatrix} f_{11} \\ f_{12} \\ \vdots \\ f_{33} \end{pmatrix}, \quad \mathbf{M} \in \mathbb{R}^{N \times 9},$$

where $\text{rank}(\mathbf{M}) = 8$.

- Solve this system using singular value decomposition.
- Make sure that $\text{rank}(\mathbf{F}) = 2$.

Eight Point Algorithm

Starting point:

- Over-determined system of equations $\mathbf{M} \cdot \mathbf{f} = 0$
- \mathbf{f} lies in the null space of \mathbf{M} . The null space is non-trivial, since $\mathbf{M} \in \mathbb{R}^{N \times 9}$ and $\text{rank}(\mathbf{M}) = 8$.

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1. SVD of $\mathbf{M} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$:

- $\sigma_9 \approx 0 \Rightarrow \mathbf{f} = \lambda \cdot \mathbf{v}_9$, and since $\|\mathbf{f}\|_F = \|\mathbf{f}\|_2 = 1 \Rightarrow \mathbf{f} = \mathbf{v}_9$
- If $\sigma_9 > \varepsilon \rightarrow \text{error}$

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2. Enforce $\text{rank}(\mathbf{F}) = 2$ using SVD of $\mathbf{F} = \mathbf{U}_F \mathbf{\Sigma}_F \mathbf{V}_F^T$:

- For the fundamental matrix it is: $\sigma_1 \geq \sigma_2 > 0$, $\sigma_3 = 0$.
- If $\sigma_3 > \varepsilon \rightarrow \text{error}$
- Set $\sigma_3 = 0$, and compute \mathbf{F} using $\mathbf{\Sigma}'_F$ anew:

$$\mathbf{F} = \mathbf{U}_F \begin{pmatrix} \sigma_1 & 0 & 0 \\ 0 & \sigma_2 & 0 \\ 0 & 0 & 0 \end{pmatrix} \mathbf{V}_F^T.$$

Numerical Instabilities

- Image coordinates are usually defined with respect to the top left corner of the image.
- Thus coordinates vary from 0 to a few hundred.
- The third (homogeneous) coordinate is usually set to 1.

Numerical Instabilities

Normalize the coordinates $\tilde{\mathbf{p}}_i^p = (\mathbf{p}_{1,i}, \mathbf{p}_{2,i}, 1)^T$ and $\tilde{\mathbf{q}}_i^p = (\mathbf{q}_{1,i}, \mathbf{q}_{2,i}, 1)^T$ such that the entries of \mathbf{M} are of comparable size:

- Translate the origin of the image coordinate system to the centroid of the feature points, that is:
 - to $(\frac{1}{N} \sum_i^N \mathbf{p}_{1,i}, \frac{1}{N} \sum_i^N \mathbf{p}_{2,i}, 1)^T$ for the left side,
 - and $(\frac{1}{N} \sum_i^N \mathbf{q}_{1,i}, \frac{1}{N} \sum_i^N \mathbf{q}_{2,i}, 1)^T$ for the right image.
- Scale the feature points such that the mean homogeneous point looks like $\frac{1}{\sqrt{2}}(1, 1, 1)^T$, i. e., the mean norm of a 2-D point is $\sqrt{2}$.

This is called ***balancing***.

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- We have gone over the eight point algorithm using the fundamental matrix.
- The fundamental matrix maps a point to its epipolar line.
- The epipolar constraint is linear in components of \mathbf{E} and \mathbf{F} .
- Balancing can be used to make an estimation of an essential matrix numerically robust.

Further Readings

Epipolar geometry is nicely introduced in:

Emanuele Trucco and Alessandro Verri. *Introductory Techniques for 3-D Computer Vision.* Upper Saddle River, NJ, USA: Prentice Hall, 1998

All the math regarding epipolar geometry can be found in:

Richard Hartley and Andrew Zisserman. *Multiple View Geometry in Computer Vision.* 2nd ed. Cambridge: Cambridge University Press, 2004. DOI: 10.1017/CB09780511811685

Magnetic navigation:

Michelle P. Armacost et al. “Accurate and Reproducible Target Navigation with the Stereotaxis Niobe® Magnetic Navigation System”. In: *Journal of Cardiovascular Electrophysiology* 18 (Jan. 2007), S26–S31. DOI: 10.1111/j.1540-8167.2007.00708.x