

EECS 442 Discussion

Arash Ushani

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Projective Geometry

- For more detail, see HZ Chapter 2

Representing Lines

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- Is this a unique representation?

$$(ka)x + (kb)y + kc = 0$$

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- Equivalence Classes

Points on Lines

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$$ax + by + c = 0$$

$$\mathbf{x} = (x, y, 1)^T$$

$$\mathbf{x}^T \mathbf{l} = 0$$

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Intersection of Lines

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HW Hints: Lines and transformations

- Helpful identity:

$$(\mathbf{H}\mathbf{x}) \times (\mathbf{H}\mathbf{y}) = (\det \mathbf{H})\mathbf{H}^{-\top}(\mathbf{x} \times \mathbf{y})$$

HW Hints: Lines and transformations

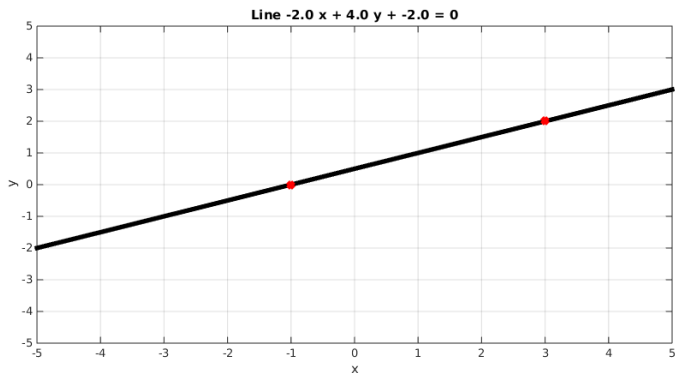
- Helpful identity:

$$(\mathbf{H}\mathbf{x}) \times (\mathbf{H}\mathbf{y}) = (\det \mathbf{H})\mathbf{H}^{-\top}(\mathbf{x} \times \mathbf{y})$$

- Alternatively, if $\mathbf{l}^\top \mathbf{x} = 0$, what can you say about $\mathbf{l}^\top \mathbf{H}^{-1} \mathbf{H}\mathbf{x}$ if \mathbf{H} is a projective transformation?

MATLAB Exercise

- Go to CTools → Resources → Discussion → 09_23_matlab.zip



Direct Linear Transform

- For more detail, see chapter 4 section 1 in HZ

Direct Linear Transform

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- We want to find the transformation \mathbf{H} such that $\mathbf{x}'_i = \mathbf{H}\mathbf{x}_i$

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- If $\mathbf{x}'_i = (x'_i, y'_i, w'_i)^\top$, then:

$$\mathbf{x}'_i \times (\mathbf{H}\mathbf{x}_i) = \begin{bmatrix} y'_i \mathbf{h}^{3\top} \mathbf{x}_i - w'_i \mathbf{h}^{2\top} \mathbf{x}_i \\ w'_i \mathbf{h}^{1\top} \mathbf{x}_i - x'_i \mathbf{h}^{3\top} \mathbf{x}_i \\ x'_i \mathbf{h}^{2\top} \mathbf{x}_i - y'_i \mathbf{h}^{1\top} \mathbf{x}_i \end{bmatrix} = \mathbf{0}$$

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$$\begin{bmatrix} \mathbf{0}^\top & -w'_i \mathbf{x}_i^\top & y'_i \mathbf{x}_i^\top \\ w'_i \mathbf{x}_i^\top & \mathbf{0}^\top & -x'_i \mathbf{x}_i^\top \\ -y'_i \mathbf{x}_i^\top & x'_i \mathbf{x}_i^\top & \mathbf{0}^\top \end{bmatrix} \begin{bmatrix} \mathbf{h}^1 \\ \mathbf{h}^2 \\ \mathbf{h}^3 \end{bmatrix} = \mathbf{0}$$

Direct Linear Transform

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- With 4 correspondences, we have $\mathbf{A}\mathbf{h} = \mathbf{0}$ where \mathbf{A} is rank 8

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- Overdetermined solution, find \mathbf{h} that minimizes error (with $\|\mathbf{h}\| = 1$)
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- SVD Decomposition:

$$\mathbf{A} = \mathbf{U}\mathbf{D}\mathbf{V}^T$$

- \mathbf{D} is a diagonal matrix of the singular values of \mathbf{A}
- \mathbf{V} contains the singular vectors of \mathbf{A} as column vectors

Next Week

- Guest Speaker: Steven Parkison, Calibration Expert
- No Office Hours

