

# 3D reconstruction

- Given a set of point correspondences  $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$  we wish to reconstruct the world points  $\mathbf{X}_i$  and the cameras  $P, P'$  such that

$$\mathbf{x}_i = P\mathbf{X}_i, \quad \mathbf{x}'_i = P'\mathbf{X}_i, \quad \forall i$$

- Without any additional information it is possible to
  - Estimate the fundamental matrix from point correspondences.
  - Calculate camera pairs from the fundamental matrix.
  - Estimate world coordinates  $\mathbf{X}_i$  corresponding to each point pair  $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$ .
- The reconstructed cameras and points will be unique up to a projective homography of  $\mathcal{P}^3$ .
- Depending on which information we have about the world and/or the cameras (point coordinates, orthogonal line pairs, parallel line pairs, calibrated cameras, etc.) we may either
  - perform a stratified reconstruction — first projectively, then affinely, the metrically, or
  - directly perform a metric reconstruction.

— p. 1

# Calculation of the fundamental matrix

- The defining equation for the fundamental matrix is

$$\mathbf{x}'^\top \mathbf{F} \mathbf{x} = 0$$

for each pair of corresponding points  $\mathbf{x} \leftrightarrow \mathbf{x}'$ . Given enough corresponding points we may calculate  $\mathbf{F}$ .

- Each point pair produces one equation linear in the elements in  $\mathbf{F}$ . If  $\mathbf{x} = (x, y, 1)^\top$  and  $\mathbf{x}' = (x', y', 1)^\top$  then

$$x'x f_{11} + x'y f_{12} + x' f_{13} + y'x f_{21} + y'y f_{22} + y' f_{23} + x f_{31} + y f_{32} + f_{33} = 0,$$

that may be written as

$$\begin{bmatrix} x'x & x'y & x' & y'x & y'y & y' & x & y & 1 \end{bmatrix} \mathbf{f} = 0,$$

where  $\mathbf{f}$  is the 9-vector with elements  $\mathbf{F}$  row-wise.

## Calculation of the fundamental matrix F

- Given  $n$  corresponding points we get a linear equation system on the form

$$\mathbf{A} \mathbf{f} = \begin{bmatrix} x'_1 x_1 & x'_1 y_1 & x'_1 & y'_1 x_1 & y'_1 y_1 & y'_1 & x_1 & y_1 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x'_n x_n & x'_n y_n & x'_n & y'_n x_n & y'_n y_n & y'_n & x_n & y_n & 1 \end{bmatrix} \mathbf{f} = \mathbf{0}.$$

- The equation is homogenous, i.e.  $\mathbf{f}$  can only be determined up to scale.
- In order for a homogenous solution to exist, the rank of  $\mathbf{A}$  cannot be larger than 8. In that case a solution  $\mathbf{f}$  exists in the null-space of  $\mathbf{A}$ .
- A solution may always be determined by solving

$$\begin{aligned} \min_{\mathbf{f}} \|\mathbf{A} \mathbf{f}\| \\ \text{s.t. } \|\mathbf{f}\| = 1, \end{aligned}$$

with solution  $\mathbf{f} = \mathbf{v}_9$  where  $\mathbf{A} = \mathbf{U} \mathbf{D} \mathbf{V}^\top$ .

— p. 3

## Minimal correspondence — the 7-point algorithm

- If the matrix  $\mathbf{A}$  is constructed from  $n = 7$  correspondences,  $\mathbf{A}$  will have a two-dimensional null-space.
- Let the vectors  $\mathbf{f}_1$  and  $\mathbf{f}_2$  be basis vectors for the null-space of  $\mathbf{A}$ . Then each vector

$$\mathbf{f} = \alpha \mathbf{f}_1 + (1 - \alpha) \mathbf{f}_2$$

is a solution of  $\mathbf{A} \mathbf{f} = \mathbf{0}$ .

- the corresponding  $\mathbf{F}$ -matrices are

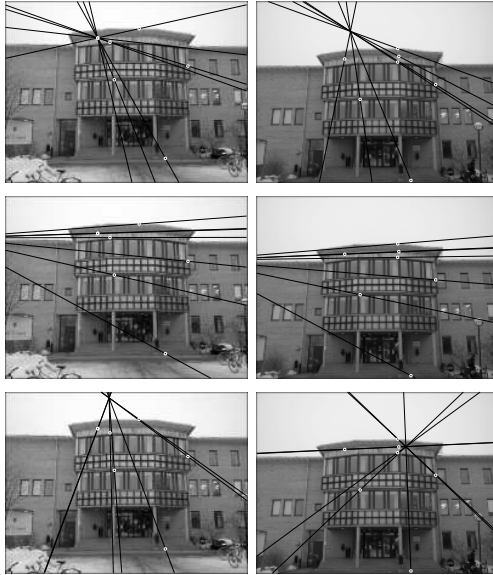
$$\mathbf{F} = \alpha \mathbf{F}_1 + (1 - \alpha) \mathbf{F}_2.$$

- The condition  $\det(\mathbf{F}) = 0$  leads to the equation

$$\det(\alpha \mathbf{F}_1 + (1 - \alpha) \mathbf{F}_2) = 0,$$

that is a third order equation in  $\alpha$  with one or three real roots, i.e. one or three solutions are possible.

# The 7-point algorithm, 3 solutions



- p. 5

# The 8-point algorithm

- The easiest way to calculate the fundamental matrix is to use  $n \geq 8$  points and

1. Calculate  $F$  that minimizes the algebraic error, i.e. solves

$$\min_{\mathbf{f}} \|\mathbf{A}\mathbf{f}\|$$

$$\text{s.t. } \|\mathbf{f}\| = 1.$$

2. Find the “closest” matrix  $F'$  of rank 2, i.e. solve

$$\min_{F'} \|\mathbf{F} - \mathbf{F}'\|_F$$

$$\text{s.t. } \text{rank}(F') = 2,$$

where  $\|\cdot\|_F$  is the Frobenius norm.

- If  $\mathbf{A} = \mathbf{U}\mathbf{D}\mathbf{V}^T$  is the singular value decomposition of  $\mathbf{A}$ , then the solution of problem 1 is

$$\mathbf{f} = \mathbf{v}_9.$$

- If  $\mathbf{F} = \mathbf{U}\mathbf{D}\mathbf{V}^T$  is the singular value decomposition of  $\mathbf{F}$ , where

$$\mathbf{D} = \text{diag}(r, s, t), r \geq s \geq t,$$

then

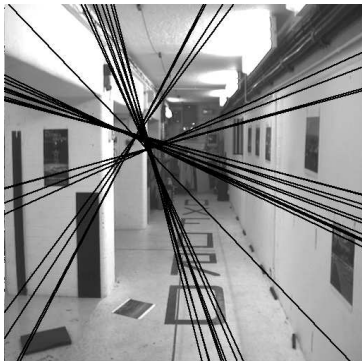
$$\mathbf{F}' = \mathbf{U} \text{diag}(r, s, 0) \mathbf{V}^T$$

is a solution to problem 2.

## Non rank-deficient $F$ matrix

- If  $F$  has full rank it will have an empty null-space, i.e. not have any point that is on all lines, i.e. no epipole.

$\text{rank}(F) = 3$



$\text{rank}(F) = 2$



- p. 7

## The normalized 8-point algorithm

- In order for the 8-point algorithm to work in practice, it has to be normalized. The algorithm then becomes:

1. Determine a transformation  $T$  and  $T'$  such that  $\hat{\mathbf{x}}_i = T\mathbf{x}_i$  and  $\hat{\mathbf{x}}'_i = T'\mathbf{x}'_i$  has center of gravity at the origin and a mean squared distance of 2 from the origin.
2. Calculate a fundamental matrix  $\hat{F}'$  corresponding to the point pairs  $\hat{\mathbf{x}}_i \leftrightarrow \hat{\mathbf{x}}'_i$ :
  - (a) Calculate  $\hat{F}$  from the solution  $\hat{\mathbf{f}}$  that solves

$$\min_{\hat{\mathbf{f}}} \|\hat{\mathbf{A}}\hat{\mathbf{f}}\|$$

$$\text{s.t. } \|\hat{\mathbf{f}}\| = 1,$$

where  $\hat{\mathbf{A}}$  is calculated from the point pairs  $\hat{\mathbf{x}}_i \leftrightarrow \hat{\mathbf{x}}'_i$ .

- (b) Calculate  $\hat{F}'$  that solves

$$\min_{\hat{F}'} \|\hat{\mathbf{F}} - \hat{\mathbf{F}}'\|_F$$

$$\text{s.t. } \text{rank}(\hat{F}') = 2.$$

3. Calculate  $\mathbf{F} = T'^T \hat{F}' T$  as the fundamental matrix corresponding to the original point pairs  $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$ .

- p. 7

# Optimal F

- In order to determine an optimal F, the reprojection error has to be minimized.
- This may be achieved through e.g. the following problem formulation:

$$\min_{\mathbf{u}} \sum_{i=1}^m d(\mathbf{x}_i, \hat{\mathbf{x}}_i)^2 + d(\mathbf{x}'_i, \hat{\mathbf{x}}'_i)^2$$

$$\text{s.t. } \hat{\mathbf{x}}'_i{}^\top \mathbf{F} \hat{\mathbf{x}}_i = 0, \quad i = 1, \dots, m$$

$$\|\mathbf{F}\|_F = 1, \det(\mathbf{F}) = 0,$$

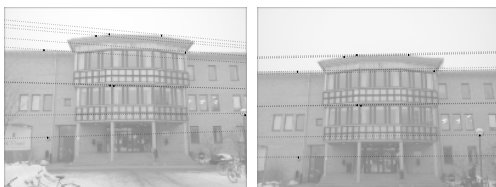
where

$$\mathbf{u} = [f_1, \dots, f_9, \hat{x}_{1x}, \hat{x}_{1y}, \dots, \hat{x}_{mx}, \hat{x}_{my}, \hat{x}'_{1x}, \hat{x}'_{1y}, \dots, \hat{x}'_{mx}, \hat{x}'_{my}]^\top.$$

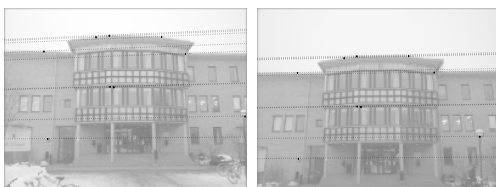
- p. 9

## Optimal F, normal example

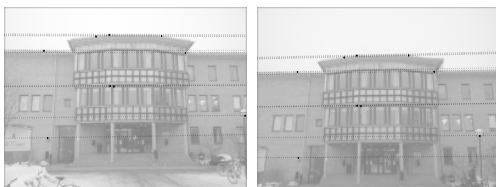
Starting approximation



One iteration



Solution



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# Optimal F

- Given an algorithm to solve

$$\min_{\mathbf{u}} \frac{1}{2} f(\mathbf{u})^T W f(\mathbf{u})$$

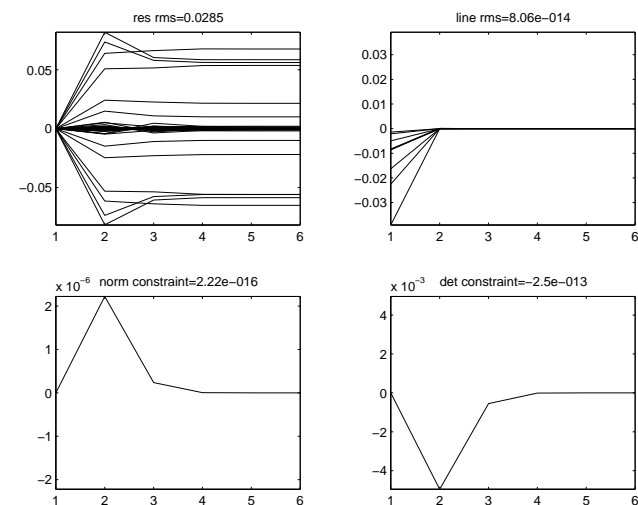
$$\text{s.t. } c(\mathbf{u}) = 0$$

we get

$$f(\mathbf{u}) = \begin{bmatrix} \hat{\mathbf{x}}_1 - \mathbf{x}_1 \\ \vdots \\ \hat{\mathbf{x}}_m - \mathbf{x}_m \\ \hat{\mathbf{x}}'_1 - \mathbf{x}'_1 \\ \vdots \\ \hat{\mathbf{x}}'_m - \mathbf{x}'_m \end{bmatrix} \quad \text{and} \quad c(\mathbf{u}) = \begin{bmatrix} \hat{\mathbf{x}}_1{}^\top \mathbf{F} \hat{\mathbf{x}}_1 \\ \vdots \\ \hat{\mathbf{x}}'_m{}^\top \mathbf{F} \hat{\mathbf{x}}_m \\ \|\mathbf{F}\|_F - 1 \\ \det(\mathbf{F}) \end{bmatrix}$$

- A starting approximation of F can be calculated by the normalized 8-point algorithm. The initial estimates of the line points  $\hat{\mathbf{x}}_i$  and  $\hat{\mathbf{x}}'_i$  are the corresponding measured points  $\mathbf{x}_i$  and  $\mathbf{x}'_i$ .

## Optimal F, normal example



Iteration sequence.

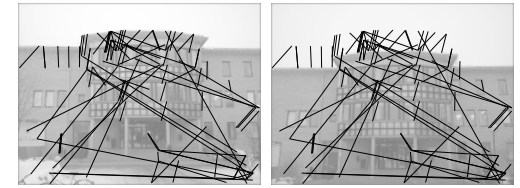
# Automatic calculation of F

- By using RANSAC we may estimate F automatically. Given  $n$  preliminary point matches, a probability  $p$  and a distance threshold  $t$ :
  - Draw 7 point pairs randomly.
  - Calculate a fundamental matrix from the point pairs. We have 1 or 3 solutions. For each of the solutions
    - Calculate the distance  $d_i$  between each point pair and the corresponding epipolar lines.
    - Calculate the number of point pairs  $k_i$  that are inliers, i.e. with distance  $d_i < t$ .
    - If  $k_i > k_{best}$  or  $k_i = k_{best}$  and  $\sigma(d_i) < \sigma_{best}$  then
      - $best = i$ ,
      - $\epsilon = k_{best} / (1 - n)$ ,  $N_{max} = \frac{\log 1-p}{\log(1-(1-\epsilon)^7)}$
    - $i = i + 1$
    - Repeat until  $i \geq N_{max}$ .
  - Then refine the best solution by minimizing the reprojection error.
  - Add correspondences that now satisfies  $d_i < t$ . Optionally repeat the previous step.

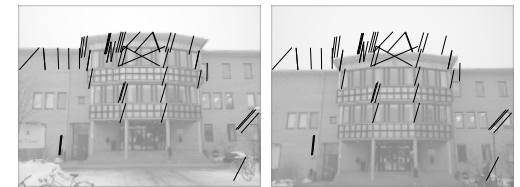
- p. 13

# Example

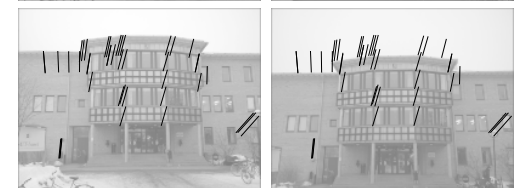
Suggested matches



After calculation of F with RANSAC



After manual cleaning



## Calculation of 3D coordinates, homogenous solution

- Given a corresponding point pair  $x \leftrightarrow x'$  and two cameras  $P, P'$ , calculate the corresponding 3D point  $X$  such that  $x = PX$  and  $x' = P'X$ .
- As previously with homography calculations we may dehomogenize the image points in e.g. image 1 and get  $x \times (PX) = 0$  or

$$x(p^{3T}X) - (p^{1T}X) = 0$$

$$y(p^{3T}X) - (p^{2T}X) = 0$$

$$x(p^{2T}X) - y(p^{1T}X) = 0$$

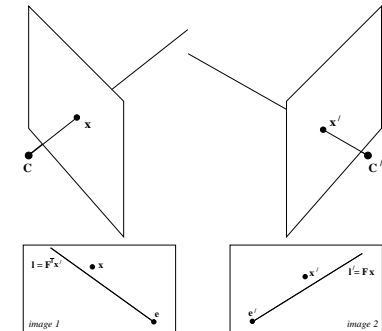
- Of these, only two are linearly independent. If we remove the third and combine with the point in image 2 we get

$$AX = 0, \text{ where } A = \begin{bmatrix} x p^{3T} - p^{1T} \\ y p^{3T} - p^{2T} \\ x' p'^{3T} - p'^{1T} \\ y' p'^{3T} - p'^{2T} \end{bmatrix}$$

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## Calculation of 3D coordinates, homogenous solution

- The equation system  $AX = 0$  is overdetermined in the sense that  $X$  has 3 degrees of freedom, but we have 4 independent equations.
- If the points  $x$  and  $x'$  do not correspond *exactly*,  $X = 0$  will be the only solution.



- We thus have to solve  $AX = 0$  approximately, either via SVD  $A = UDV^T = A$ ,  $X = v_4$  or by specifying  $X = (X, Y, Z, 1)^T$ .

# Calculation of 3D coordinates, Euclidian solution

- We may also choose to minimize the Euclidian distance in  $\mathcal{R}^3$ . If camera 1 has center in  $\tilde{\mathbf{C}}$  and we know another point  $\tilde{\mathbf{U}}$  on the line  $\mathbf{U} = \mathbf{P}^+ \mathbf{x}$ , then a point  $\tilde{\mathbf{X}}$  on the line satisfies

$$\tilde{\mathbf{X}} = \tilde{\mathbf{C}} + \alpha(\tilde{\mathbf{U}} - \tilde{\mathbf{C}}).$$

- Similarly in image 2

$$\tilde{\mathbf{X}} = \tilde{\mathbf{C}}' + \alpha'(\tilde{\mathbf{U}}' - \tilde{\mathbf{C}}'), \text{ where } \mathbf{U}' = \mathbf{P}'^+ \mathbf{x}'.$$

- If the lines do not intersect there is no point  $\tilde{\mathbf{X}}$  that satisfies both equations. However, then we may minimize the distance to the lines

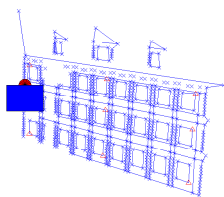
$$\min_{\alpha, \alpha', \tilde{\mathbf{X}}} \left\| \begin{bmatrix} \tilde{\mathbf{U}} - \tilde{\mathbf{C}} & \mathbf{0} & -\mathbf{I}_3 \\ \mathbf{0} & \tilde{\mathbf{U}}' - \tilde{\mathbf{C}}' & -\mathbf{I}_3 \end{bmatrix} \begin{bmatrix} \alpha \\ \alpha' \\ \tilde{\mathbf{X}} \end{bmatrix} - \begin{bmatrix} -\tilde{\mathbf{C}} \\ -\tilde{\mathbf{C}}' \end{bmatrix} \right\|^2.$$

- This method is called *forward intersection* in the photogrammetric community.

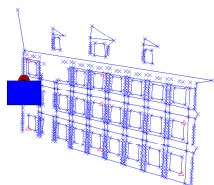
- p. 17

## Example

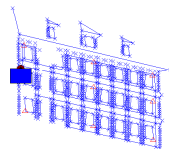
Starting approximate from the normalized 8-point algorithm, via the essential matrix and forward intersection.



One iteration



Solution



- p. 19

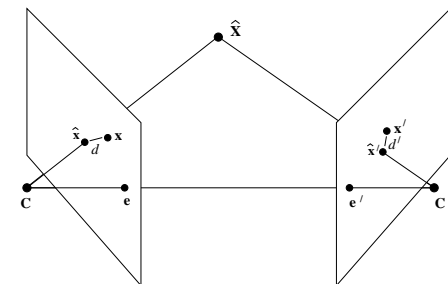
# Minimization of the reprojection error

- The optimal solution is obtained if we minimize the reprojection error, i.e. solves

$$\begin{aligned} \min_{\hat{\mathbf{x}}, \hat{\mathbf{x}}', \mathbf{X}} & d(\mathbf{x}, \hat{\mathbf{x}})^2 + d(\mathbf{x}', \hat{\mathbf{x}}')^2 \\ \text{s.t. } & \hat{\mathbf{x}} = \mathbf{P}\mathbf{X} \\ & \hat{\mathbf{x}}' = \mathbf{P}'\mathbf{X} \end{aligned}$$

which exactly corresponds to

$$\begin{aligned} \min_{\hat{\mathbf{x}}, \hat{\mathbf{x}}'} & d(\mathbf{x}, \hat{\mathbf{x}})^2 + d(\mathbf{x}', \hat{\mathbf{x}}')^2 \\ \text{s.t. } & \hat{\mathbf{x}}'^T \mathbf{F} \hat{\mathbf{x}} = 0 \end{aligned}$$



- If we have determined  $\mathbf{F}$  by minimizing the reprojection error, the corresponding points have already been determined and the SVD solution of  $\mathbf{A}\mathbf{X} = 0$  gives the null-space solution.