# **Computer Vision Exercise 2**

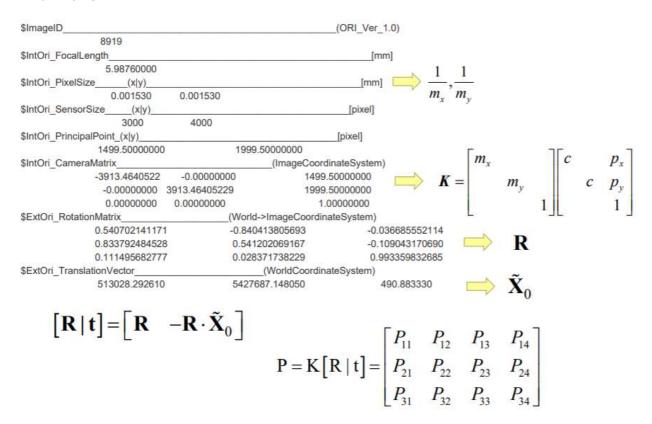
## **Spatial Intersection and Resection**

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# **I. Processing Steps**

#### 1.

Compute projection matrix:



Compute pixel coordinates:

$$\mathbf{x} = \mathbf{P} \cdot \mathbf{X}$$

Plot points:



### 2. Measure one object



# 3. Spatial intersection

For unknown object coordinate X at least two pixel measures x and x' from two cameras with known projection matrix P and P' are available:

$$\mathbf{x} = \mathbf{P} \cdot \mathbf{X}$$
  $\mathbf{x}' = \mathbf{P}' \cdot \mathbf{X}$ 

From which we can build identity equation:

$$\mathbf{x} \times (\mathbf{P}\mathbf{X}) = \mathbf{0}$$
  $\mathbf{x}' \times (\mathbf{P}'\mathbf{X}) = \mathbf{0}$ 

With

$$\mathbf{P} = \mathbf{K} \begin{bmatrix} \mathbf{R} \mid \mathbf{t} \end{bmatrix} = \begin{bmatrix} P_{11} & P_{12} & P_{13} & P_{14} \\ P_{21} & P_{22} & P_{23} & P_{24} \\ P_{31} & P_{32} & P_{33} & P_{34} \end{bmatrix}$$

Then we get

$$x(\mathbf{p}^{3T}\mathbf{X}) - (\mathbf{p}^{1T}\mathbf{X}) = 0$$
  $x'(\mathbf{p}'^{3T}\mathbf{X}) - (\mathbf{p}'^{1T}\mathbf{X}) = 0$   
 $y(\mathbf{p}^{3T}\mathbf{X}) - (\mathbf{p}^{2T}\mathbf{X}) = 0$   $y'(\mathbf{p}'^{3T}\mathbf{X}) - (\mathbf{p}'^{2T}\mathbf{X}) = 0$   
 $x(\mathbf{p}^{2T}\mathbf{X}) - y(\mathbf{p}^{1T}\mathbf{X}) = 0$   $x'(\mathbf{p}'^{2T}\mathbf{X}) - y'(\mathbf{p}'^{1T}\mathbf{X}) = 0$ 

For both x and x', the third equation is linearly dependent on the other two, therefore we eliminate it and get:

$$AX = 0$$

where

$$\mathbf{A} = egin{pmatrix} x\mathbf{p}^{3\mathrm{T}} - \mathbf{p}^{\mathrm{IT}} \ y\mathbf{p}^{3\mathrm{T}} - \mathbf{p}^{2\mathrm{T}} \ x'\mathbf{p}'^{3\mathrm{T}} - \mathbf{p}'^{\mathrm{T}} \ y'\mathbf{p}'^{3\mathrm{T}} - \mathbf{p}'^{2\mathrm{T}} \end{pmatrix} = egin{pmatrix} x\mathbf{p}(3;\mathrm{i}) - \mathbf{p}(1,:) \ y\mathbf{p}(3,:) - \mathbf{p}(2,:) \ x'\mathbf{p}'(3,\mathrm{i}) - \mathbf{p}'(1;) \ y'\mathbf{p}'(3,:) - \mathbf{p}'(2,\mathrm{i}) \end{pmatrix}$$

And

$$\mathbf{X} = \begin{pmatrix} X & Y & Z & W \end{pmatrix}^T$$

Solve this equation using singular vector decomposition, we can get the final object coordinates.

#### 4. Back transformation and errors

We apply back transformation with

$$\mathbf{x} = \mathbf{P} \cdot \mathbf{X} \qquad \qquad \mathbf{x}' = \mathbf{P}' \cdot \mathbf{X}$$

And calculate the error with

$$\mathbf{v'v} = \sum (\mathbf{x}_{meas} - \mathbf{x}_{trafo})^2$$
  $\sigma_0 = \sqrt{\frac{\mathbf{v'v}}{2 \cdot n_{images} - 3}}$ 

#### 5. Direct Linear Transformation

For the direct linear transformation we use the following equation, where P matrix is what we need.

$$\mathbf{x}_{i} \times \mathbf{P} \cdot \mathbf{X}_{i} = \begin{pmatrix} y_{i} \mathbf{p}^{3T} \mathbf{X}_{i} - w_{i} \mathbf{p}^{2T} \mathbf{X}_{i} \\ w_{i} \mathbf{p}^{1T} \mathbf{X}_{i} - x_{i} \mathbf{p}^{3T} \mathbf{X}_{i} \\ x_{i} \mathbf{p}^{3T} \mathbf{X}_{i} - y_{i} \mathbf{p}^{2T} \mathbf{X}_{i} \end{pmatrix} = 0$$

This can be rewritten into

$$\begin{pmatrix} \mathbf{0}^T & -w_i \mathbf{X}_i^T & y_i \mathbf{X}_i^T \\ w_i \mathbf{X}_i^T & \mathbf{0}^T & -x_i \mathbf{X}_i^T \\ -y_i \mathbf{X}_i^T & x_i \mathbf{X}_i^T & \mathbf{0}^T \end{pmatrix} \begin{pmatrix} \mathbf{p}^1 \\ \mathbf{p}^2 \\ \mathbf{p}^3 \end{pmatrix} = 0$$

The third row is linear dependent on the first two rows, therefore we can eliminate it:

$$egin{pmatrix} \mathbf{0}^T & -w_i \mathbf{X}_i^T & y_i \mathbf{X}_i^T \ w_i \mathbf{X}_i^T & \mathbf{0}^T & -x_i \mathbf{X}_i^T \end{pmatrix} egin{pmatrix} \mathbf{p}^1 \ \mathbf{p}^2 \ \mathbf{p}^3 \end{pmatrix} = A_i \mathbf{p} = 0$$

To solve this equation we need more than 6 pairs of points:

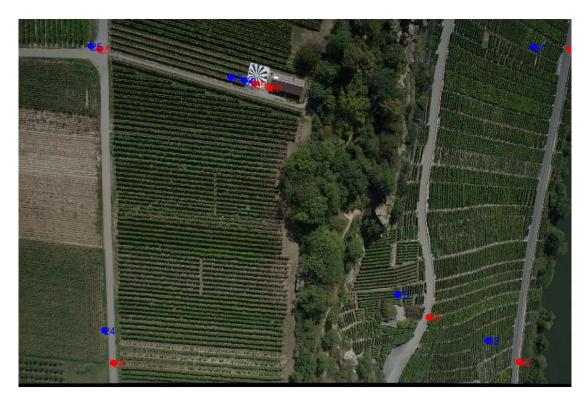
$$\mathbf{A}_{2n\times 12} \cdot \mathbf{p} = 0$$

Similar to before, we use singular vector decomposition to calculate the P matrix.

$$[\cup, \mathrm{D}, \mathrm{V}] = \mathrm{svd}(\mathrm{A}, 0)$$
  
% Extract homography  $P = \mathrm{reshape} \; (V(:, 12), 4, 3)'$ 

#### 6. Re-mapping and comparison

Similar to task 1,  $\mathbf{X} = \mathbf{P} \cdot \mathbf{X}$  is used to re-compute the mapping



### 7. Reconstruct the camera parameters

a) translation vector X0

X0 can be computed from Singular Value Decomposition (SVD) of P:

$$[U, D, V] = svd(P, 0)$$

Where X0 is the last column of V

b) camera matrix K and rotation matrix R

$$P = K[R \mid t] = K[R \mid -R\tilde{X}_{\theta}] = KR[I_{3} \mid -\tilde{X}_{\theta}] = M[I_{3} \mid -\tilde{X}_{\theta}]$$

Where  $\mathbf{M} = \mathbf{K}\mathbf{R}$  is the left 3x3-Sub-Matrix of P, With  $M^{-1} = R^T K^{-1}$  matrix  $\mathbf{M}$  can be decomposed into QR decomposition:

$$[q,r] = qr(M^{-1})$$

And

$$R = q^{-1}$$

$$K = r^{-1}$$

PS: we have to normalize the K and X0 by the scale factor.

### **II. Results**

1. Fundamental matrix & Pixel coordinates

3.4975e+03	2.0830e+03	1.8694e+03	-1.3101e+10
2.3237e+03	-3.3321e+03	997.5443	1.6893e+10
0.1131	-0.0513	0.9923	2.1995e+05



### 2. Measure an object



## 3. Object Coordinates (m)

5.1300e+05 5.4277e+06 327.2105

### 4. Back transformation errors (pixel)

4.5173

$$sigma0_y =$$

3.6680

#### 5. Direct Linear Transformation

$$P_20851_B =$$

0.0000	0.0000	0.0000	-0.6128
0.0000	-0.0000	0.0000	0.7902
0.0000	-0.0000	0.0000	0.0000

### 6. Remapping Difference (pixel) & Error (pixel)

243.8114	69.7921	64.1058	242.0076	282.3202	197.3427	182.8119
176.9169	252.3815	24.1761	165.4313	15.5587	54.0080	50.1713

$$sigma0_T6_x =$$

159.2771

$$sigma0_T6_y =$$

108.1346

#### 7. Reconstruct Camera Parameters

X0=

K=

-3.9334e+03	-0.0087	2.1435e+03
0	3.9334e+03	1.4235e+03
0	0	1

R=

#### Differences:

diff\_X0 =

-10.9532 -1.1280 21.7447

diff\_K =

diff R =

1.0e-05 \*

0.0193 -0.0254 0.0272 -0.0264 0.0388 -0.4424 0.2694 -0.3509 -0.0488