

Vision 3D artificielle - Final exam

(duration: 1:30h)

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You can choose to answer in French or English, at your convenience.

1 Computation of homography

In the computation of a homography from (finite) point matches, we can write the vector equation $X'_i \times (HX_i) = 0$ as two independent linear equations in the parameters $h = (H_{11} \ H_{12} \ \cdots \ H_{33})$ of H :

$$\begin{pmatrix} x_i & y_i & 1 & 0 & 0 & 0 & -x'_i x_i & -x'_i y_i & -x'_i \\ 0 & 0 & 0 & x_i & y_i & 1 & -y'_i x_i & -y'_i y_i & -y'_i \end{pmatrix} h = 0.$$

From 4 point matches, we get an 8×9 system in h , so h is a non-null vector of the kernel of the stacked matrix A . There seems to be always a unique solution but obviously there are degenerate situations. The following questions examine these degeneracies.

1. Explain why there is no solution if $X_1 = X_2$ and points X'_1 and X'_2 are distinct.
2. By examining the structure of the matrix A , explain why there is no unique solution if the points X_1 , X_2 and X_3 are distinct but aligned.
3. Show the form of the homography matrix H keeping fixed all points of line $y = 0$ and also the point $(0, 1)$ of Euclidean plane. Explain why it is a 1-parameter family of homographies.

2 Sampson error for fundamental matrix

We recall that the Sampson error $\epsilon_S(X)$ for a given geometric model, an approximation of the geometric error based on the algebraic error $\epsilon(X)$, can be written:

$$\epsilon_S = \epsilon^T (JJ^T)^{-1} \epsilon$$

with J the matrix of partial derivatives (Jacobian matrix) of ϵ as a function of data point X . This exercise computes the formula of Sampson error for fundamental matrix F .

The algebraic error is $P'^T F P = 0$ with $P = (x \ y \ 1)^T$ and $P' = (x' \ y' \ 1)^T$.

1. Compute J at point $X = (x \ y \ x' \ y')$.
2. Write formula for Sampson error associated to F .
3. Discuss its relation to geometric error.

3 Recovery of projection matrices from F

From P and P' , 3×4 projection matrices in left and right images, we can compute F , the fundamental matrix. The goal of this exercise is to recover all possible pairs of matrices (P, P') from a given F . We note e' the epipole in right image ($F^T e' = 0$).

1. Show that if $P = (I|0)$ and $P' = ([e']_{\times} F | e')$, then $S = P'^T F P$ is skew-symmetric ($X^T S X = 0$ for all X).
2. Deduce that the above pair (P, P') gives rise to F . Why is the right camera of this particular solution not a pinhole camera?
3. Show that if $P = (I|0)$ and $P' = (A|a)$ (with A a 3×3 matrix and a a 3-vector), then $F = [a]_{\times} A$.
4. Suppose $F = [a]_{\times} A = [a']_{\times} A'$. Prove that there is some $\lambda \neq 0$ and some 3-vector v such that

$$a' = \lambda a \text{ and } \lambda A' = A + av^T.$$

5. Prove that if $P = (I|0)$, then $P' = ([e']_{\times} F + e'v^T | \lambda e')$ with $\lambda \neq 0$ and v a 3-vector.
6. Explain why if the pair (P, P') yields F then so does $(PH, P'H)$ for any 4×4 matrix H , then show that the general solution is such that $P = (P_{3 \times 3} | P_4)$ with $P_{3 \times 3}$ invertible, and

$$P' = ([e']_{\times} F + e'v^T)P + e'w^T$$

with w a 4-vector such that

$$\begin{pmatrix} P \\ w^T \end{pmatrix}$$

be invertible.

4 Feature detection and description

1. Between a 90° corner and a 45° corner, which one has the strongest response according to the Harris-Stephens criterion, and why?

To answer this question, you may simply compare the case of the central point in images I and I' defined by:

$$I = \begin{array}{|c|c|c|c|c|} \hline 0 & 0 & 1 & 1 & 1 \\ \hline 0 & 0 & 1 & 1 & 1 \\ \hline 0 & 0 & 1 & 1 & 1 \\ \hline 0 & 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 \\ \hline \end{array} \quad I' = \begin{array}{|c|c|c|c|c|} \hline 0 & 0 & 0 & 0 & 1 \\ \hline 0 & 0 & 0 & 1 & 1 \\ \hline 0 & 0 & 1 & 1 & 1 \\ \hline 0 & 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 \\ \hline \end{array}$$

As indicated in the lecture on the Harris corner detector, directional derivatives can be computed convoluting respectively with $[-1 \ 0 \ 1]$ and $[-1 \ 0 \ 1]^T$.

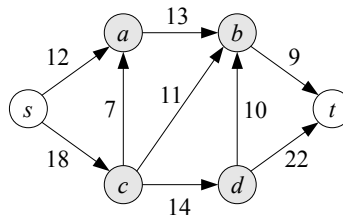
You may restrict yourself to a *derivative* Gaussian, and forget about the *integration* Gaussian. To simplify calculations, you may also use an approximate Gaussian kernel defined as:

$$g = \frac{1}{16} \begin{array}{|c|c|c|} \hline 1 & 2 & 1 \\ \hline 2 & 4 & 2 \\ \hline 1 & 2 & 1 \\ \hline \end{array}$$

2. Does the answer differ if you consider the corner strength as defined by Brown et al. (2005), i.e., the harmonic mean of the eigenvalues?
3. Let $\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3$ be detected SIFT points in image I , \mathbf{x}'_1 be a detected SIFT points in image I' and $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}'_1$ be the corresponding SIFT descriptors. Assume that $\|\mathbf{v}_1 - \mathbf{v}_2\| = 0.1$ and $\|\mathbf{v}_1 - \mathbf{v}_3\| = 0.11$. What can be said of a possible match between \mathbf{x}_1 and \mathbf{x}'_1 ?

5 Graph cuts

1. What is the minimal cut of the following graph, and why?



2. When using a graph cut for estimating a disparity map as explained in the course, suppose you use a zero-mean normalized sum of square differences (E_{ZNSSD}) rather than a zero-mean normalized cross correlation (E_{ZNCC}). Has anything to be changed in edge weight definition?