

## Assignment 6

This exercise is part of the course assignment. **Deadline for the assignment 30.11.2022 at 23:59**

The topic of this assignment is features matching using SURF descriptors. For this assignment you should return

- The files `surf_nn.m` and `surf_nndr.m`. Each file should contain your name and student number (of both students if you work in pairs).
- Your answers to the questions in the analysis part. At the end of the course, you should return a single pdf containing the answers to all questions of the assignments. The report should contain also your name and student number (of both students if you work in pairs).

### Coding part (5pt)

First run the demo `demo6.p` to get a taste of the assignment. Your first task is to complete the function `[y1, y2] = surf_nn(surf1, surf2, x1, x2)`. The variables  $x_1, x_2$  are  $m \times 2$  and  $n \times 2$  matrices containing the coordinates of the interest points in the first and second image, respectively. The matrices  $surf_1$  and  $surf_2$  contain the surf descriptor (one per row) of each keypoint.

Your first task is to find the coordinates of the *mutually nearest* keypoints and return their coordinates sorted from the most similar pair to the most different. The similarity is measured using the standard Euclidean distance between descriptors. The smaller the distance, the more similar the descriptors are. Particularly you should

1. Compute the  $m \times n$  matrix `distmat`, each element at row  $i$  and column  $j$  should contain the euclidean distance between the  $i$ th descriptor of the first image and the  $j$ th descriptor of the second image.
2. Let  $f_1$  a descriptor from the first image and  $f_2$  a descriptor from the second image. The two descriptors are said to be mutually nearest if  $f_1$  is the most similar descriptor to  $f_2$  among all descriptors in the first image **and**  $f_2$  is the most similar descriptor to  $f_1$  among all descriptors in the second image. Construct a matrix `pairs` which contains the mutually nearest descriptors. The first column should contain the index of the descriptor of the first image, the second column the index of the corresponding closest descriptor in the second image and the third column the distance between the two descriptors.

**Example 1:** The third row from `surf1` and the fifth row from `surf2`

form a mutually nearest pair and their distance is 0.4, the matrix `pairs` should then contain the row `[3, 5, 0.4]`.

**Example 2:** You pick the fourth row from `surf1` and find that the closest descriptor from `surf2` is the second row. However, when you consider the second row of `surf2`, you find that closest row from `surf1` is the fifth, hence the fourth and second descriptor from `surf1` and `surf2` do not form a mutually nearest pair.

**Hint!** Read carefully the documentation of the matlab function `min`. What does `min(A, [], 2)` do? What happens if you call `min` with two outputs? You may want to loop through each descriptor in `surf1`, find its closest neighborhood from `surf2` and check if they form a mutually nearest pair.

3. Sort the rows of `pairs` in increasing order of distances.
4. Extract the coordinates of the sorted mutually nearest pairs from  $x_1$  and  $x_2$ .

Your second task is to perform feature matching using nearest neighborhood distances ratios. Let  $f_1$  and  $f_2$  two mutually nearest descriptors as before and let  $f_3$  the second closest descriptor to  $f_1$  from the second image. Now we define the distances ratio as

$$dr = \frac{\|f_1 - f_2\|}{\|f_1 - f_3\|} \quad (1)$$

and use this distances ratio to measure the similarity between descriptors instead of plain Euclidean distance, i.e. the smaller the distances ratio, the more similar the descriptors are. Here you will need to complete the function `[y1, y2] = surf_nndr(surf1, surf2, x1, x2)`. Particularly,

1. Repeat steps 1-2 from the previous task
2. For each row in `pairs` (i.e. each mutually nearest pair), compute the distances ratio as defined in eq. (1). Store the distances ratios in an array called `nndr`. **Hint:** you may want to create a copy of `distmat` where the columns are sorted in increasing order.
3. Sort the rows in `pairs` in increasing order with respect to `nndr`.
4. Extract the coordinates of the sorted mutually nearest pairs from  $x_1$  and  $x_2$ .

When you are done, run the script `main6.m` to verify your functions work correctly.

## Analysis part (5pt)

Answer the following questions in your report

- Consider the 5 best matches for each algorithm. How many matches are correct using plain distances? How many matches are correct using ratios of distances? Why does using distance ratios lead to a more robust result? I.e. what is the idea behind nearest neighborhood distance ratio? (3pt)
- What are the advantages of using descriptors like SIFT or SURF to perform feature matching over comparing the raw RGB values of the pixels? (2pt)