



## **MACHINE VISION**

Assignment 1: Feature points and image matching



## **DUE DATE**

This assignment should be submitted to Canvas before 11:59pm on Friday 20/03/2020.

Please submit a single ZIP file with your student number and name in the filename. Your submission should contain:

- A <u>detailed documentation</u> of all code you developed, including the tests and evaluations you carried out. Please make sure that you <u>include a document with every result image</u> you produce <u>referencing the exact subtask and lines of code</u>.
- All Python code you developed in a single .py file that can be executed and that generates the outputs you are referring to in your evaluation. Please make sure that you clearly <u>indicate in your comments the exact subtask</u> every piece of code is referring to.

You can achieve a total of 50 points as indicated in the tasks.

## TASK 1 (Feature points, 33 points)

In this task you will implement a scale and rotation invariant point feature extraction algorithm inspired by SIFT to identify a set of interest points in an image together with their respective scales and rotations.

- A. Download the input image file **Assignment\_MV\_01\_image\_1.jpg** from Canvas. Load the file and convert it into a single channel grey value image [2 points]. Make sure the data type is float32 to avoid any rounding errors [1 point]. Determine the size of the image and resize the image to double its size [2 points].
- B. Create twelve Gaussian smoothing kernels with increasing  $\sigma=2^{k/2}, k=0,...,11$ , and plot each of these kernels as image [4 points]. Make sure that the window size is large enough to sufficiently capture the characteristic of the Gaussian. Apply these kernels to the resized input image from subtask A to create a scale-space representation and display all resulting scale-space images [2 points].
- C. Use the scale-space representation from subtask B to calculate difference of Gaussian images at all scales. Display all resulting DoG images [3 points].

- D. Find key-points by thresholding all DoGs from subtask C using a threshold of T=10. Suppress non- maxima in scale-space by making sure that the key-points have no neighbours, both in space as well as in scale, with higher values [3 points]. The resulting key-points should comprise three coordinates  $(x, y, \sigma)$ , two spatial and the scale at which they were detected.
- E. Calculate derivatives of all scale-space images from subtask B using the kernels  $d_x=(1 \ 0 \ -1)$  and  $d_y=(1 \ 0 \ -1)^T$ . Display the resulting derivative images at all scales [4 points].
- F. Calculate the gradient length  $m_{qr}$  and gradient direction  $\theta_{qr}$  for the 7 × 7 grid of points  $(q,r) \in \left\{x + \frac{3}{2}k\sigma \mid k = -3,...,3\right\} \times \left\{y + \frac{3}{2}k\sigma \mid k = -3,...,3\right\}$  sampled around each key-point (x,y) and using the appropriate scale  $\sigma$  determined in subtask D and the correct gradient images from subtask E  $[4\ points]$ . Also calculate a Gaussian weighting function  $w_{qr} = e^{-(q^2+r^2)/(9\sigma^2/2)}/(9\pi\sigma^2/2)$  for each of the grid points  $[1\ point]$ . Now create a 36-bin orientation histogram vector h and accumulate the weighted gradient lengths  $w_{qr}m_{qr}$  for each grid point (q,r) where the gradient direction  $\theta_{qr}$  falls into this particular bin  $[3\ points]$ . Use the maximum of this orientation histogram to determine the orientation of the key-point  $[1\ point]$ .
- G. Draw all the key-points into the input image using a circle with  $3\sigma$  radius and a line from the key-point centre to the circle radius to indicate the orientation (see example for a single key-point on the right). Display the resulting output image with all the key-points [3 points].



## TASK 2 (Image matching, 17 points)

In this task you will implement a correlation-based area matching algorithm to find a patch extracted from one image in a second image.

- A. Download the input image files **Assignment\_MV\_01\_image\_1.jpg** (the same as in the previous task) and **Assignment\_MV\_01\_image\_2.jpg** from Canvas. Load both files and convert them into a single channel grey value image [2 points]. Make sure the data type is float32 to avoid any rounding errors [1 point].
- B. The window on the  $1^{st}$  floor above the arch on the left wing is in a rectangle with the image coordinates ((360,210), (430,300)) in the first input image. Draw a rectangle around this window in the input image and display it [1 point]. Now cut out the image patch only containing the window and display it as image [2 points].
- C. Calculate the mean and standard deviation of the cut-out patch from subtask B [2 points]. Go through all positions in the second input image and cut out a patch of equal size [2 points]. Also calculate mean and standard deviation and from this the cross-correlation between the two patches [3 points]. Create and display an image of all cross-correlations for all potential positions in the second image [2 points]. Find the position with maximum cross-correlation and draw a rectangle around this position in the second input image. Display the result [2 points].