

<http://cs.ucf.edu/~bagci/>

[PROGRAMMING ASSIGNMENT] (3)

ROBOT VISION

DR. ULAS BAGCI • (SPRING) 2018 • UNIVERSITY OF CENTRAL FLORIDA (UCF)

Coding Standard and General Requirements

Code for all programming assignments should be **well documented**. A working program with no comments will receive **only partial credit**. Documentation entails writing a description of each function/method, class/structure, as well as comments throughout the code to explain the program flow. Programming language for the assignment is **Python**. You can use standard python built-in IDLE, or other IDLEs such as CANOPY, PyCharm Community Edition, PyScripter, CodeSculptor, Eric Python, Eclipse plus PyDev, etc.

Following libraries (or others if you can find in addition to those listed below) can be used throughout the course:

- PIL (The Python Imaging Library), Matplotlib, NumPy, SciPy, LibSVM, OpenCV, VLFeat, python-graph.

If you are asked to implement “Gaussian Filtering”, you are not allowed to use a Gaussian function from a known library, you need to implement it from scratch.

Along with your well commented code, please submit a “brief” report as doc or pdf format.

Submit by **27th of March 2018**, 11.59pm.

Optical Flow [6 pts]

[3 pts] Implement Lucas-Kanade optical flow estimation, and test it for the two-frame data set provided in the webcourses: basketball.

[3 pts] Implement Lucas-Kanade optical flow estimation algorithm in a multi-resolution Gaussian pyramid framework. After experimentally optimize number of levels for Gaussian pyramid, local window size, and Gaussian width, use the same data set (basketball) to find optical flows, visually compare your results with the previous step where you don't use Gaussian pyramid.

Evaluation: Please create some random colors to show **velocity vectors** on the original images, and save them as outcomes of your program. If you have difficulty showing optical flows with velocity vectors, you can use some built-in functions (color) to illustrate optical flows. Visit http://opencv-python-tutroals.readthedocs.org/en/latest/py_tutorials/py_video/py_lucas_kanade/py_lucas_kanade.html for a sample implementation of Lucas-Kanade using Python and OpenCV. Other Python sources are available too. You are allowed to use built-in functions such as Gaussian smoothing, convolution, gradient operations, corner detections, etc., but not the Lucas-Kanade implementation itself. For features: you are suggested to use OpenCV's **goodFeaturesToTrack**, or your own implementation of corner detection.

Gaussian Mean Shift Clustering [4 pts]

Implement mean-shift clustering algorithm for the segmentation of the following RGB image: "GMS.jpg". Set the parameters (h_s : spatial resolution, h_r : range resolution) according to the properties of the attached image (you need to explore them yourself). The use of RGB (color) image in this assignment is also new to you. Please explore OpenCV libraries about how to use color images. Basically, you have three channels as opposed to one channel gray scale representation of the images. Briefly mention optimality of why and how you decide those parameters along with the output. There is no ground truth available for this image as this question, evaluate the goodness of the segmentation visually. In the pseudo-code, $p(n|x)$ is posterior probability, and x 's update means difference of x 's values between consecutive iterations.

Data: Input image with N data points (pixels)

Result: Output image segmented via Gaussian mean shift

for $n \in \{1, \dots, N\}$ **do**

$x \leftarrow x_n$;

while Update of x 's $>$ tolerance **do**

$\forall n : p(n|x) \leftarrow \frac{\exp(-0.5 \cdot \|(x-x_n)/\sigma^2\|^2)}{\sum_{m=1}^N \exp(-0.5 \cdot \|(x-x_m)/\sigma^2\|^2)}$;

$x \leftarrow \sum_{n=1}^N p(n|x)x_n$;

end

$z_n \leftarrow x$;

end

Use connected component in z_n to identify clusters (built-in function);

Algorithm 1: Pseudo-Code for Gaussian Mean Shift Clustering