

Introduction to Computer Vision (ECSE 415)

Assignment 3

Due: November 23rd, 11:59PM

Please submit your assignment solutions electronically via the myCourses assignment dropbox. Attempt all parts of this assignment. The assignment will be graded out of total of **26 points**. Students are expected to write their own code. (Academic integrity guidelines can be found at <https://www.mcgill.ca/students/srr/academicrights/integrity>). Assignments received up to 24 hours late will be penalized by 30%. Assignments received more than 24 hours late will not be graded.

Submission Instructions

1. Prepare and submit a single Google Colab notebook containing answers to all three questions.
2. Comment your code appropriately.
3. Do not submit input/output images. The output images should be displayed in the notebook. Assume the input image folders are kept in a same directory as the codes.
4. Make sure that the submitted code is running without error. Add a README file if required.
5. Answers to reasoning questions should be comprehensive but concise.
6. Submissions that do not follow the format will be penalized 10%.

1 Image Segmentation using K-means

Implement K-means algorithm using only the numpy library. You can use opencv and matplotlib libraries only to read and display images. Apply K-means to the images 'home' and 'flower' shown in Figure 1. Try K=2 and K=3. Run the algorithm for 10 iterations and display the resulting segmented images in each case. (10 points)



Figure 1: Segment above images using K-means algorithm.

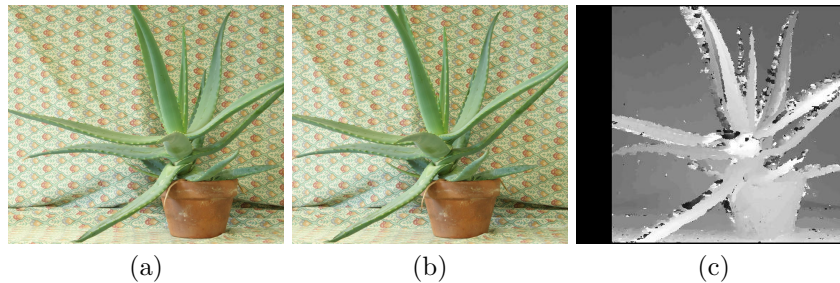


Figure 2: A pair of stereo images (a) left image (b) right image (c) disparity map (expected output).

2 Disparity

In this section, we will compute disparity map D from a pair of stereo images captured using parallel cameras. The images are shown in Figure 2(a) and 2(b). We will solve correspondence problem with the **window search algorithm**. Refer to slides 58-59 in *Lecture 18 - Stereo Vision*. Instead of searching for a matching window on the entire scanline, we will restrict the search on a small region on the scanline.

1. Extract a **5×5 window centered at each pixel-location $(i, j)_L$ in the left image. Let's call these windows reference windows. (2 points)**
2. For each reference window in the left image do the following.
 - (a) On the right scanline, create a search region bounded by pixel-locations $(i, j - 47)_R$ and $(i, j)_R$. Extract 5×5 windows centered at every pixel-location in this search region. **(2 points)**
 - (b) For few boarder pixel-locations either the reference window or the search region lie outside the boundary of the image. Set disparity

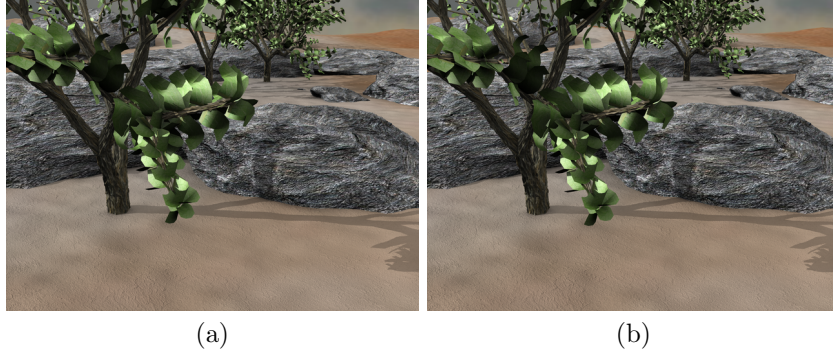


Figure 3: Input frames for optical flow computation. (a) frame1 (b) frame2.

- $D(i, j) = 48$ for these pixel-locations. For the remaining locations do the following. **(2 points)**
- (c) Compute sum-of-square-difference(SSD) between the windows in the search region and the reference window. **(2 points)**
 - (d) Find a location $(i', j')_R$ with minimum SSD and compute disparity $D(i, j) = j_L - j'_R$. (Note that $0 \leq D(i, j) \leq 47$ as the search region contains 48 pixel-locations.) **(1 point)**
3. Display the final disparity map D with the cmap argument of `plt.imshow` set to `'gray_r'`. The expected output is shown in Figure 2(c). **(1 point)**

3 Optical Flow

In this section, we will observe the effect of the window-size on the prediction accuracy of optical flow. The input frames are shown in Figure 3(a-b) and the ground-truth flow is given in 'flow10.npz' file. Read ground truth flow as follows: `gt = np.load('flow10.npz')['flow']`

1. Use `calcOpticalFlowFarneback` from OpenCV to compute optical flow between the input frames with the arguments set as follows. **(2 points)**
 - `flow=None`, `pyr_scale=0.5`, `levels=3`, `iterations=3`, `poly_n=5`, `poly_sigma=1.2` and `flags=0`.
 - Very winsize from 5 to 21 in the steps of 2.
2. For each setting of winsize, measure mean squared error(MSE) between estimated optical flow and the ground truth optical flow. Plot MSE (y-axis) vs winsize (x-axis). **(2 points)**
3. Do you observe any trend in the plot above? Does the error increase or decrease with increasing window-size? Explain the effect of window-size on the prediction error. **(2 points)**