## CSci 4270 and 6270 Computational Vision, Spring Semester, 2025 Lecture 04 - 06 Practice Exercises

## Overview

- 1. Please see the lecture 01 and 02 practice problems for an overview.
- 2. Some of these questions were on Quiz 1 last spring.

## **Problems**

- 1. Convert a line in slope-intercept (y = mx + b) form to the implicit form and use this to compute the minimum Euclidean distance of point  $(x_0, y_0)$  to the line, expressed in terms of m and b.
- 2. Given an  $N \times 2$  array of points, write code to find and print the values of a, b and c for the orthogonal least squares best fitting line. Resolve the sign ambiguity of (a, b) by making sure that a is positive. (For the sake of time, don't worry about the possibilty that a is 0). Please, no for loops.
- 3. If **H** is a 3x3 transformation matrix representing a 2d transformation and  $\tilde{\mathbf{x}} = (u, v, 1)^{\top}$  is the homogeneous representation of a point, show that  $s\mathbf{H}$  maps  $\tilde{\mathbf{x}}$  to the same point location for all  $s \neq 0$ .
- 4. Why does smoothing require the sum of the weights of the convolution kernel to be 1.0?
- 5. Consider boundary effects during convolution. When computing a derivative for example with the Sobel kernels as we did in HW 2 what happens when we consider all values outside the image as 0? Why might it be better to just assign the value of the convolution to be 0 at boundary pixels? (Note, if I asked this on the quiz I would give you the Sobel kernels. You don't need to memorize them.)
- 6. Given a 3x3 array how can you determine the simplest type of 2d transformation it describes? Is it rigid? Is it a similarity? Is it affine? Or, is it projective? Write a short python script that takes a 3x3 array and outputs one of five different words: rigid, similarity, affine, homography or none. You may assume that the last entry (lower right corner) of the array is 1. Ignore numerical issues such as round-off errors. Note that if the matrix is not full rank then none is the correct answer. You can use the determinant (np.linalg.det) to check rank.

As examples, consider for these five example arrays

0 0 114 0 0 255 0.1 0.4 0.82533561 0.56464247 163 -0.56464247 0.82533561 -450 0 1 1 0 114 0 0 255

the outputs should be, in order,

similarity affine none rigid homography

0.01 0.04

- 7. Intuitively, why do the Sobel kernels smooth the image in addition to computing the derivative?
- 8. A camera is in the world at location  $X=0,\,Y=1,\,Z=2$  (in meters), its optical axis is aligned with the Z axis, and its focal length is 50 mm (0.05 meters). Consider a point p at location  $X=-1,\,Y=3,\,Z=7$  (in meters). Showing your work on this problem is very important.
  - (a) Find the projection of p onto the image plane in millimeters.
  - (b) If the optical axis intersects the image plane in the center of a  $4000 \times 6000$  (rows x columns) pixel image and the pixels are 20 microns (0.02 millimeters) on a side, what is the pixel location of p in the image?
- 9. Given are the following binary image (left) and structuring element (right), where blank pixels and pixels outside the shown grid have value 0.

	1	1						
1					1	1		
1	1					1		
1		1				1	1	
1	1	1						
		1	1	1				

- (a) How many 8-connected components are in the binary image?
- (b) Show the result of morphological closing (dilation followed by erosion) with the given structuring element. To do this, simply put an **X** through the pixels (in the figure above!) that change from 0 to 1 or 1 to 0 as a result of the closing (you don't need show the dilation result).
- (c) How many 8-connected components are there after the closing is applied? As part of your answer, circle each of the connected components in the above figure.
- (d) Can a closing **ever** increase the number of connected components? Why (briefly)?
- 10. Give two reasons why the change detection example from the end of Lecture 6 is too naive for non-artificial change detection problems.
- 11. Write NumPy code that, given a gray scale image im, computes the following:
  - (a) Sorts the intensities. Call the resulting array val short for "values". (Note that np has a sort function.)
  - (b) Finds the grayscale values g0 and g1 such that 1/3 the pixels have gray scale values less than or equal to g0 and 2/3 of the pixels have gray scale values less than or equal to g1.
  - (c) Creates a new image called im\_n where all intensities less than or equal to g0 are mapped to 0, all intensities greater than g1 are mapped to 255, and every other intensity is unchanged.

To illustrate, in the example below, the image im is shown on the left, g0 == 21 and g1 == 41, and the new image  $im_n$  is shown on the right.

15	21	30	17	11	12
45	44	23	60	22	33
98	41	61	10	39	42

	0	0	30	0	0	0
	255	255	23	255	22	33
Ī	255	41	255	0	39	255

12. The RANSAC objective function has been criticized for being a simple count of points whose distance is less than a threshold  $\tau$  from a line. Instead, some researchers use a weighted objective function that gives more weight for smaller distances:

$$\begin{cases} 1 - d^2/\tau^2 & \text{if } ||d|| \le \tau \\ 0 & \text{otherwise.} \end{cases}$$

Write code to evaluate this new objective function on the N data points represented as a  $2 \times N$  NumPy array called pts. Assume a hypothesized line is given to you, represented in implicit form by variables a, b, c. Also assume variable tau has been assigned to store the value of  $\tau$ . Assign the final result to a variable called obj.

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