Exercise #4 - Continuous Transformations for Scene Understanding

DUE: AS INDICATED on Canvas

Please thoroughly read through chapter 5 in <u>Computer and Machine Vision by E.R. Davies</u> and notes on Segmentation; please also note the scanned special handout from E.R. Davies Computer and Machine Vision Chapter 9 on methods to skeletonize an image (machine vision process) uploaded on Canvas. The goal of this lab is to introduce concepts and theory related to continuous scene segmentation, basic shape recognition (with skeletal thinning), depth estimation, and behavioral analysis, which feed into recognition methods. It includes both top-down OpenCV application development and understanding as well as bottom-up algorithm implementation and understanding.

Exercise #4 Requirements:

- 1) [5 points] Read the paper <u>Use of the Hough Transformation to Detect Lines and Curves in Pictures</u>, Richard Duda & Peter Hart (also available on Canvas) and summarize the papers key points and contributions to computer vision.
- 2) [5 points] Build and run the basic <u>Hough linear example</u> provided and paste a transformed image of your choice into your report.
- 3) [20 points] Using a top-down OpenCV approach, adapt the example code found in capture_transformer to use the skeletal.cpp transform on continuous frames (like captureskel.cpp) from your camera, but use the much simpler approach of simpler-capture rather than V4L2 camera capture. Gesture in front or your camera and see if you can get a reasonable continuous skeletal transform of your arm and hand. Record example frames (up to 3000 for 100 seconds of video JPEG frames are fine) and encode your results to an MPEG video. Upload the modified code with your report.
- 4) [20 points] Use the methods presented in E.R. Davies Chapter 9 handout on Canvas to write your own algorithms from the ground up to transform video frames of basic arm gestures to create a skeletal model of the arm. First, eliminate background, then convert the image to a binary bit map (as we discussed in class), then apply the skeletal thinning algorithms from E.R. Davies so that your frames track only the movements of the arm skeleton over time. Comment on whether your bottom-up algorithm implementation is better than the OpenCV top-down in terms of quality and efficiency. [Use ffmpeg to decode and save a single frame from the MPEG video]

- 5) [10 points] Read the paper <u>Distinctive Image Features from Scale-Invariant Keypoints</u>, by David Lowe (also available on Canvas) and summarize the papers key points and contributions to computer vision.
- 6) [10 points] Build and run the basic OpenCV keypoint comparison detector code provided in sift as detector_extractor_matcher.cpp and paste a transformed image of your choice into your report. It should provide a comparis on like this between two images:



Take two snapshots of the same scene (as was done above) from a left/right offset and run the detector comparison code and paste the result into your report.

7) [10 points] Build and run the disparity depth estimator example in <u>example stereo</u> using two cameras and describe how it works based on your reading and understanding of the code and just in terms of performance and accuracy. Paste in an example left-eye, right-eye, depth map set of three images that correspond which you captured with your cameras. Are keypoints used for passive depth estimation, and if not, could they be?

Upload all video as encoded MPEG-4 at a reasonable bit-rate and quality.

[20 points] Overall, provide a well-documented professional report of your findings, output, and tests so that it is easy for a colleague (or instructor) to understand what you've done. Include any C/C++ source code you write (or modify) and Makefiles needed to build your code and make sure your code is well commented, documented and <u>follows coding style guidelines</u>. I will look at your report first, so it must be well written and clearly address each problem providing clear and concise responses to receive credit.

In this class, you'll be expected to consult the Linux and OpenCV manual pages and to do some reading and research on your own, so practice this in this first lab and try to answer as many of your own questions as possible, but do come to office hours and ask for help if you get stuck.

Upload all code and your report completed using MS Word or as a PDF to Canvas and include all source code (ideally example output should be integrated into the report directly, but if not, clearly label in the report and by filename if test and example output is not pasted directly into

the report). Your code must include a Makefile so I can build your solution on an embedded Linux system (R-Pi 3b+ or Jetson). Please zip or tar.gz your solution with your first and last name embedded in the directory name and/or provide a GitHub public or private repository link. Note that I may ask you or SA graders may ask you to walk-through and explain your code. Any code that you present as your own that is "re-used" and not cited with the original source is plagiarism. So, be sure to cite code you did not author and be sure you can explain it in good detail if you do re-use, you must provide a proper citation and prove that you understand the code you are using.

Grading Rubric

[5 points] Read and summarize the main points of Duda and Hart paper on Hough transform	
[1 pts] main point #1	
[1 pts] main point #2	
[1 pts] main point #3	
[2 pts] summary overall	
[5 points] Build and run Hough linear example	
[2 pts] Build and run completion evidence	_
[3 pts] Application of Hough lines to an image of interest	
[20 points] Continuous skeletal transform of you hand	
[10 pts] OpenCV code, build, run, test for continuous skeletal transformation	
[10 pts] Recording of 100 seconds of video showing that your continuous skeletal transformation works	
[20 points] Ground up development of skeletal transformation of hand and arm gestures	
[5 pts] Background elimination design and implementation	
[5 pts] Conversion to bitmap and application of skeletal thinning	
[10 pts] MPEG video and comparison to top-down approach for skeletonization	_
[10 points] Read and summarize the main points of David Lowe paper on SIFT	
[2 pts] main point #1	
[2 pts] main point #2	
[2 pts] main point #3	
[4 pts] summary overall	
[10 points] Build and run disparity depth estimator example	
[5 pts] Build and run completion evidence	_
[5 pts] Disparity correspondence analysis for left and right eye image	

[20 poi	nts] Quality of reporting and code quality and originality:
	[10 pts] Professional quality of reporting, testing and analysis (06 is below average, 7 is average, 8 is good, 9 excellent, and 10 is best overall.)
	[10 pts] Code quality including style, commenting, originality, proper citation for re-used code, modified code, etc