HVCA - Exercise 2

April 21, 2025

Please submit the exercise in a single pdf file including all the matlab code you wrote and your answers to all questions.

In this exercise you will implement the regularized (or Bayesian) Lucas Kanade algorithm for motion estimation. That is, you will find the velocity vector v that solves the following equation:

$$Av = b$$

with:

$$A = \left(\begin{array}{cc} \sum I_x^2 + \lambda & \sum I_x I_y \\ \sum I_x I_y & \sum I_y^2 + \lambda \end{array}\right)$$

and

$$b = -\left(\begin{array}{c} \sum I_x I_t \\ \sum I_y I_t \end{array}\right)$$

We will implement the algorithm in its iterated form, where in each iteration we use as an initial guess the velocity computed in the previous iteration (this is equivalent to developing the Taylor series around that vector). To initialize the algorithm, we will compute a single iteration of the algorithm on a pair of blurred and downsampled images, where we expect a better guess, especially in complex, natural images.

- 1. Write a function named ImageDerivatives.m that receives two images as inputs (the first and the second images in a sequence) and returns three values: I_x , I_y and I_t , those being the derivatives of the images across the X dimension, the Y dimension, and time. Use the following guidelines:
 - (a) The derivatives are obtained by convolving the two images with a kernel. Use the following kernels:

$$K_y = \frac{1}{4} \begin{pmatrix} -1 & -1 \\ 1 & 1 \end{pmatrix}, K_x = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ 1 & -1 \end{pmatrix}, K_t = \frac{1}{4} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}.$$

- (b) I_x is the sum of the convolution of each image with the appropriate kernel $(K_x)^1$
- (c) I_y is the sum of the convolution of each image with the appropriate kernel (K_y) .
- (d) I_t is the subtraction of the convolution of the first image with K_t from the convolution of the second image with K_t .
- (e) You may use Matlab's conv2 function. If you do, use the 'same' option that returns a matrix the same size as the original image, discarding the borders which we would have eliminated anyway to avoid edge artifacts.
- 2. Write a function named LK_alg.m which implements the iterative algorithm. The function should have six arguments: I_1 , I_2 , λ , mask, $v_iinitial$, $num_iiterations$, and a single output v. I_1 , I_2 are the images, λ is the algorithm's parameter which in the Bayesian formulation is the ratio between the noise variance and the prior variance, mask is a binary matrix the same size as the image that defines the pixels over which the summation should be made, $v_iinitial$ is the initial guess for the velocity, and $num_iiterations$ is the number of iterations to be computed. Use the following guidelines:
 - (a) In each iteration we use the previous velocity estimate v_prev to compute I_2^w , a warped version of I_2 that has been translated by that velocity. We then calculate the velocity v between I_1 and I_2^w , and we sum that velocity to v_prev to obtain the new estimate.
 - (b) Use the provided function warp that receives as arguments I_2 and v and returns as outputs I_2^w and warpMask. warpMask defines which pixels of I_2^w are valid, since the translation needs some pixels that are not available. Multiply term-by-term warpMask with the current mask to obtain the new mask.
- 3. Write a function named blur_dowsample.m which receives as argument an image I and returns an image that has been blurred and downsampled. To blur an image, convolve it with a Gaussian kernel. You may use the provided kernel in GaussKernel.mat or create your own. If you use the conv2 function remember to use the 'same' option. Downsample the image by taking every second pixel in both dimensions. In Matlab, you may use ':' (the colon operator).
- 4. Write a function named Full_LK_alg.m which receives as arguments I_1 , I_2 , λ , mask, $num_iterations$, and returns the estimated velocity v. You should first blur and downsample both images, get a $v_blurred$ estimate using a single iteration of the algorithm (with a (0) initial velocity guess), and the use $2*v_blurred$ as the initial estimate for the iterative

¹The reason why we use the sum of derivatives of *both* images is that this provides a better approximation of the images (by using the average of derivatives of I_1 and I_2).

- algorithm (this time with the original images). The 2 factor is required because of the downsampling.
- 5. Load the two frames from the flower garden sequence. View these frames using the provided function mymovie.m. What is moving faster, the tree or the flowers? Now run your function on subimages that contain the flowers and on subimages that contain the tree with $\lambda=0$. What are the velocities calculated by your algorithm? In Matlab, use: im=double(imread('imagename')); to read an image.
- 6. Use the provided function edgeFrames.m to create pairs of frames of a 128x128 vertical edge that moves with a given velocity v. Run your algorithm for the following values of v: v=(1,0), v=(1,1), v=(1,-1). What is the output of your algorithm (use $\lambda=0.1$)? Does it agree with the correct velocity v? Show the matrix A that your algorithm computes with $\lambda=0.1$ and $\lambda=0$. Is the matrix invertible? Use the provided function mymovie.m to view the frames. Does your percept agree with the output of the algorithm? Discuss.
- 7. Use the provided function rhombusMovie.m to generate a thin rhombus and a fat rhombus. Plot them. Calculate and plot the IOC and VA predictions for these two stimuli assuming both lines are moving to the right with velocity 1. Use the function warp to warp I_2 according to the VA and IOC velocities. Do either of these images agree with I_1 ? Run your algorithm with $\lambda = 0.01$ for these two stimuli with varying degrees of contrast (between 0 and 1). What happens to the algorithm output as the contrast is lowered? Why? Try with different numbers of iterations. Compare these results to your own percept by looking at the rhombus demo in the course's moodle.