

Computer Vision II - Homework Assignment 4

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This homework is due on July 12, 2017 at 9:00.

Please note that a late handin is not possible for this assignment!

Problem 1 - Horn-Schunck optical flow - 32 points

In this problem, you will implement a global optical flow algorithm. The test data for this assignment is from the [Middlebury optical flow database](#), and you can evaluate your implementations with the corresponding ground truth data and the provided functions `read_flow_file` and `flow_to_color`.

In a first step, you implement several basic functions that will be needed for the optical flow estimation. For a quantitative evaluation of optical flow, we use the average endpoint (EP) error metric. The endpoint error for a pixel is defined as follows:

$$\text{EP error} = \sqrt{(u - u_{\text{gt}})^2 + (v - v_{\text{gt}})^2}, \quad (1)$$

where u, v are the horizontal and vertical components of the flow vector at a certain pixel and $u_{\text{gt}}, v_{\text{gt}}$ represents the corresponding ground truth.

- Write a function `evaluate_flow(uv, uv_gt)` that computes the average endpoint error (AEPE) by averaging the EP error of all pixels.

1 point

- The ground truth data uses the value $2e9$ for invalid pixels. Make sure to not include these pixels in your AEPE calculation.

1 point

To evaluate the brightness constancy given an initial flow estimate uv_0 , it is necessary to warp the second image I_2 . Here, a so-called backward warping is performed, i.e. for each pixel (i, j) the corresponding brightness value is obtained at the position $I_2(i + u_0, j + v_0)$.

- Write a function `warp_image(im2, uv0)` that returns the second image warped by the initial estimate uv_0 . Use bilinear interpolation.

2 points

- Load images `frame10.png`, `frame11.png` and convert the data to grayscale. Load the ground truth data `flow10.flo` using the provided function `read_flow_file`. Apply the ground truth flow to obtain the warped second image I_2^w . Display I_1 , I_2^w and the difference between both images. What do you observe?

3 points

To obtain an efficient optimization scheme for the optical flow estimation, it is necessary to linearize the brightness constancy assumption around an initial flow estimate uv_0 such that

$$I_2(i + u(i, j), j + v(i, j)) - I_1(i, j) \approx I_x \cdot (u(i, j) - u_0(i, j)) + I_y \cdot (v(i, j) - v_0(i, j)) + I_t. \quad (2)$$

Here, I_x and I_y denote the spatial derivatives of the warped second image I_2^w and $I_t = I_2^w - I_1$.

- Calculate I_x , I_y and I_t in the function `compute_grad_images(im1, im2, uv0)`. Use central differences to approximate the derivatives I_x and I_y and apply an appropriate boundary handling.

2 points

Now, you implement a simplified form of dense Horn-Schunck optical flow without the coarse-to-fine estimation.

- Implement the logarithm of an optical flow posterior assuming a pairwise MRF prior as well as a Gaussian distribution with parameter σ for likelihood and prior. In particular, write the function

```
logposterior_HS(uv, uv0, Ix, Iy, It, lambda, sigma)
```

to compute the log-posterior itself and

```
grad_logposterior_HS(uv, uv0, Ix, Iy, It, lambda, sigma)
```

to calculate its gradient. The parameter λ scales the smoothness or prior term and allows for a trade-off between the two components of the posterior.

6 points

- Implement a flow estimation algorithm

```
flow_HS(im1, im2, uv0, lambda, sigma)
```

performing an iterative minimization of the negative log posterior. Use Julia's `Optim` package to perform the optimization with `method = LBFGS()` and `iterations = 200`.

3 points

- Use the grayscale images `frame10.png`, `frame11.png` and estimate Horn-Schunck optical flow with parameter $\sigma = 1$. Write a function

```
find_lambda(im1, im2, uv0, uv_gt, sigma)
```

to determine an appropriate trade-off parameter λ , e.g. with grid search. Here, an evaluation of the AEPE w.r.t. the ground truth `flow10.flo` might be beneficial.

3 points

- Display your flow result using the provided function `flow_to_color` and calculate its AEPE.

1 point

- Estimate Horn-Schunck flow for a significantly smaller trade-off parameter λ_2 and for a significantly larger parameter λ_3 . Display the results. What do you observe?

2 points

- Compare the AEPE of your results for $\lambda_1, \dots, \lambda_3$. Do the visual results match the quantitative evaluation?

1 points

- Perform 5 iterations of the optical flow estimation using the previously obtained flow field $uv_{k-1}, k = 1, \dots, 5$ as the initial flow estimate for the next iteration k . Find an appropriate parameter λ , evaluate the AEPE and visualize the results for each iteration. What do you observe?

6 points