Variational Methods for Computer Vision: Exercise Sheet 7

Exercise: December 11, 2014

Part I: Theory

The following exercises should be **solved at home**. You do not have to hand in your solutions, however, writing it down will help you present your answer during the tutorials.

- 1. Let $Q := [0,1] \times [0,1]$ be a rectangular area and let $v : Q \to \mathbb{R}^2$ be a differentiable vector field defined on Q with $v(x,y) = (a(x,y), b(x,y)) \in \mathbb{R}^2$.
 - (a) Prove Green's theorem:

$$\int_{Q} b_x(x, y) - a_y(x, y) \, \mathrm{d}x \, \mathrm{d}y = \oint_{\partial Q} (a \, \mathrm{d}x + b \, \mathrm{d}y)$$

Assume the boundary curve ∂Q to be be oriented counter clockwise. Hint: Use the fundamental theorem of calculus.

- (b) Let $\Omega \subset \mathbb{R}^2$ be an area that can be represented as a disjoint union of a finite number of squares $Q_1, ..., Q_n$. Why does Green's theorem also hold for this set Q?
- 2. Let $\Omega = [-5; 5] \times [-5; 5]$ be a rectangular area and let $I : \Omega \to [0; 1]$ be an image given by:

$$I(x,y) = \begin{cases} 1 & \text{if } x^2 + y^2 \le 1\\ 0 & \text{else.} \end{cases}$$

Furthermore let $C : [0,1] \to \Omega$ be a curve represented by a circle centered at the origin having radius r and local curvature κ_C .

- (a) Write down the two-region piecewise constant Mumford-Shah functional with length regularity weighted by ν .
- (b) Write down the corresponding gradient descent for the two cases r > 1 and $r \le 1$.
- (c) Show that the Gâteaux-Derivative at r = 1 is not continuous.
- (d) In which range should ν be in order to obtain good segmentation results?

Part II: Practical Exercises

This exercise is to be solved during the tutorial.

Super-Resolution from Video.

In the lecture we encountered the concept of super resolution from video. The key idea of super resolution is to exploit redundancy available in multiple frames of a video. Assuming that each input frame is a blurred and downsampled version of a higher resolved image u, the high-resolution image can be recovered as the minimum of the following energy functional:

$$E(u) = \sum_{i=1}^{n} \int_{\Omega} ((ABS_{i}u)(x) - (Uf_{i})(x))^{2} dx + \lambda \int_{\Omega} |\nabla u(x)| dx.$$
(1)

The Linear Operator B denotes a Gaussian Blurring. The upsampling operator U simply replaces every pixel with four pixels of the same intensity. In order to be able to compare image u with the upsampled version of f_i which is constant blockwise, we apply the linear averaging operator A on uwhich assigns every block of pixels the mean values of the pixels in that block. The linear operator S_i accounts for the coordinate shift by motion s_i hence:

$$(S_i u)(x) = u(x + s_i(x)).$$

- 1. In the following we are going to construct a toy example for super resolution by executing the following steps:
 - (a) Download the archive vmcv_ex07.zip and unzip it on your home folder. In there should be a file named Boat.png.
 - (b) Create from the unzipped image 6 versions shifted in x direction by exactly one pixel hence:

$$f_i(x,y) = f(x+i,y),$$

for i = 1...6. In order to account for the boundary, consider taking cropped images from the interior of the original image.

- (c) In order to simulate blurring convolve the shifted images with a gaussian kernel. Next downsample the images f_i by factor 2 by using the immessize function in Matlab with nearest neighbor interpolation.
- 2. In what follows we are going to minimize the above functional in order to obtain a super resolved image from our input images f_i .
 - (a) Derive the Euler-Lagrange equation of E and the corresponding gradient descent scheme.
 - (b) Compute the matrix representations of the linear operators A, B, S_i and U. Since these matrices are huge, again use sparse data structures in Matlab (spdiags speye) in order to obtain a sparse representation.
 - (c) Compute $u^* = \operatorname{argmin}_u E(u)$ by means of gradient descent using matrix vector representation after stacking the function u in a vector using the matlab command reshape.