Convex Optimization for Computer Vision

Lecture: M. Möller Exercises: J. Geiping Summer Semester 2019 Universität Siegen Department ETI Computer Vision

Weekly Exercises 11

Room: HA-7116

Submission deadline: Wednesday, 17.07.2019, 18:00

Exercise 1 (4 points). The Huber penalty $h_{\varepsilon}: \mathbb{R} \to \mathbb{R}$ is given as

$$h_{\varepsilon}(x) = \begin{cases} \frac{x^2}{2\varepsilon} & \text{if } |x| \leq \varepsilon, \\ |x| - \frac{\varepsilon}{2} & \text{otherwise.} \end{cases}$$

Show that the huber penalty can be expressed as the so-called *infimal convolution* of the functions $f: \mathbb{R} \to \mathbb{R}$ with $f(x) := \frac{x^2}{2\varepsilon}$ and $g: \mathbb{R} \to \mathbb{R}$ with g(x) := |x|:

$$h_{\varepsilon}(x) = \min_{y} f(x - y) + g(y)$$

Exercise 2 (2 points). Let an energy E be the *infimal convolution* of two other energies, i.e.,

$$E(x) = \inf_{x_1, x_2, x = x_1 + x_2} E_1(x_1) + E_2(x_2).$$

Show that

$$E^*(p) = E_1^*(p) + E_2^*(p).$$

Exercise 3 (4 points). Compute the convex conjugate F^* of the function $F: \mathbb{R}^{2N} \to \mathbb{R}$ defined as

$$F(x) := \sum_{i=1}^{2N} h_{\varepsilon}(x_i).$$

What is the proximity operator of F^* ?

Programming: Huber-TV denoising

Exercise 4 (6 Points). Implement a solution of

$$\min_{u} \frac{1}{2} \|u - f\|_{2}^{2} + \alpha \sum_{i=1}^{2N} h_{\varepsilon}((Du)_{i})$$

using the PDHG method, with fixed step sizes τ and σ (PDHG), with adaptive step sizes for one function being strongly convex (PDHG2), and for functions with a strongly convex and an L-smooth part (PDHG3). In the above, D is a finite difference approximation of the gradient of an image.