Numerical Methods for Visual Computing M. Möller, V. Gandikota, University of Siegen Winter Semester 19/20

## Weekly Exercises 10

To be discussed on Friday, 20.12.2019, 10:15-11:45, in room H-C 6336 Submission deadline: Tuesday, 17.12.2019, in the lecture

## Theory

Exercise 1 (2 points). Compute the gradient and the Hessian of the energy

$$E(a_1, a_2) = \frac{1}{2} \sum_{i=1}^{n} (a_1 \cos(a_2 x_i) - y_i)^2.$$

## **Programming**

In our programming exercise we will test our optimization methods on the task of parameter estimation. We have some model given, here for example, we know that our data behaves like a cosine wave, but we want to compute the parameters of this wave from given measurements. We view this task of finding the 'optimal' parameters as a minimization problem.

**Exercise 2** (2 points). You first generate measurements  $y_i = a_1^* \cos(a_2^* x_i)$  for points  $x_i \in [-1, 1], i = 1, ..., 20$  for the ground truth parameters  $a_1^* = 0.5$  and  $a_2^* = 2\pi$ . Now add some Gaussian noise (with  $\sigma = 0.02$ ) to the measurements  $y_i$ . Plot the ground truth cosine wave and the distorted cosine wave.

Next generate a random starting point  $(a_1, a_2)$  close to the true values via  $a_1 = a_1^* + 0.2n_1$  and  $a_2 = a_2^* + 2n_2$  for two random numbers  $n_1, n_2$  from a unit Gaussian distribution. Plot the curve that you would get for these initial parameters.

**Exercise 3** (6 points). Now try both methods from the lecture for this problem of finding the optimal parameters  $a_1, a_2$ . You already computed the gradient and Hessian of the energy in exercise 1.

- Implement the Newton method to solve the minimization of  $E(a_1, a_2)$ . How close to the true solution do you have to initialize for the Newton method to converge?
- Implement the gradient descent method to solve the minimization of  $E(a_1, a_2)$ . Pick a reasonably small step size  $\tau$ , e.g.  $\tau = 0.01$ . Does the method always converge? Does it always converge to a good solution?