SoSe 2020 Dr. Joscha Reimer

Optimization and Data Science

6. Homework exercises

Theoretical exercise 1:

Define

$$x_k := \begin{cases} \left(\frac{1}{3}\right)^k & \text{if } k \text{ is odd} \\ \left(\frac{1}{4}\right)^k & \text{if } k \text{ is even} \end{cases} \quad \text{for all } k \in \mathbb{N} \text{ and } x_0 = \frac{1}{3}.$$

Examine the sequence $(x_k)_{k\in\mathbb{N}}$ regarding convergence rate with the Q-factor and the R-factor.

Theoretical exercise 2:

Consider the globalized Newton-Method with an efficient step-size method applied to

$$f: \mathbb{R} \to \mathbb{R}, x \mapsto x^2$$
.

Let $(x_k)_{k\in\mathbb{N}_0}$ be the sequence of iterates. Show that $(x_k)_{k\in\mathbb{N}_0}$ converges Q-superlinear to a global minimizer of f.

Hint: Apply the convergence results from the lecture.

Theoretical exercise 3:

Consider the globalized Newton-Method with $c \in (0,1]$ applied to

$$f: \mathbb{R} \to \mathbb{R}, x \mapsto |x|^p$$

with p > 2. Let $(x_k)_{k \in \mathbb{N}_0}$ be the sequence of iterates. Show that

- a) If $x_k > 0$, then $d_k = -\frac{1}{p-1}x_k$, for all $k \in \mathbb{N}_0$.
- b) If $x_0 > 0$ and $\rho_k = 1$ for all $k \in \mathbb{N}_0$, then $x_k > x_{k+1} > 0$ for all $k \in \mathbb{N}_0$.
- c) If $x_0 > 0$ and $\rho_k = 1$ for all $k \in \mathbb{N}_0$, then $(\rho_k)_{k \in \mathbb{N}_0}$ is an efficient step-sizes sequence.
- d) If $x_0 > 0$ and $\rho_k = 1$ for all $k \in \mathbb{N}_0$, then $(x_k)_{k \in \mathbb{N}_0}$ converges Q-linear but not Q-superlinear to a global minimizer of f.
- e) If $x_0 > 0$ and $\rho_k = 1$ for all $k \in \mathbb{N}_0$, can we apply the convergence result for the globalized Newton method?

Hint: Use the definition of gradient-related and efficient step-sizes to prove a) and c), respectively. A mathematical induction might be useful to prove part b). Calculate the resulting Q-factors to prove d).

Programming exercise 1:

Implement a general descent method using the armijo step-size method and supporting the gradient method and the globalized Newton method. The following inputs should be supported:

- the function (including first and second derivative)
- whether the gradient method or the globalized Newton method should be used
- a stopping bound for the stopping criterion
- a maximal number of iterations
- parameters of the step-size method and the optimization method
- a callback function which is called during the optimization in each iteration step with the current iterate as argument (and thus allows for example printing the progress)

as well as the following outputs:

- the minimizer and the minimum
- the number of needed iterations, function evaluations, gradient evaluations and Hessian matrix evaluations.

If the first derivative of the function is not supplied, your method should approximate it using finite differences.

Try your implementation with the Roosenbrock function and the Bazaraa-Shetty function. Compare the behavior of both methods with different configurations using these two examples.

The solutions of the theoretical exercises will be discussed on 18. Mai 2020.