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Optimierung

http://www.math.uni-konstanz.de/numerik/personen/rogg/de/teaching/

Program 3 (6 Points)

Submission by E-Mail: 2015/06/29, 10:00 h

Implementation of a globalized (Quasi-)Newton method

Write Part 1 and Part 2 together in a main file main.m.

Part 1: Implement the local Newton method for optimization known from the lecture (Algorithm 5.6 with $F = \nabla f$). Write a function

```
function [X] = newtonmethod(fhandle, x0, epsilon, nmax)
```

with input arguments

- fhandle: function handle to a function of form [f,g,H] = functionname(x) (the output values are the function value, the gradient and the Hessian matrix corresponding to the input argument x).
- x0: initial point
- epsilon: tolerance for the termination condition $\|\nabla f(x^k)\| < \epsilon$
- nmax : maximum number of iterations

The program should return a matrix X = [x0, x1, x2, ...] containing the whole iterations.

Test your program by using the negative cosine function. Write herefore a function file ncosH.m which is of the above form. Use the parameters epsilon = 1e-5 and nmax = 50. As initial points choose x0 = 1.1655, 1.1656, 1.9, atan(-pi). Explain the results you get and use suitable plots for showing X. For comparison plot also the iterates you obtain by applying the function gradmethod from Program 1 with t0 = 1, alpha = 1e-2, beta = 0.5 and amax = 30.

Part 2: In this part we modify the local Newton method such that it is globally convergent. In addition, we add a switch to a globalized BFGS method if the Hessian matrix of the considered function is not given. The resulting algorithm is defined in Algorithm 1 and will be implemented in the function globalnewtonmethod. Use the Matlab function

nargout to identify if the Hessian is provided or not. Note that the inequality in Line 9 of Algorithm 1 can be interpreted as a generalized ankle condition.

Algorithm 1

Require: Initial point x^0 , stopping tolerance $\varepsilon > 0$, maximal iteration number n_{\max} , $\alpha_1, \alpha_2 > 0, p > 0$, and (for Armijo) an initial step size $t_0^A, \alpha^A \in (0, 1), \beta^A \in (0, 1)$, maximal iteration number a_{max} 1: n = 0;2: if $\nabla^2 f$ is given then $H_n = \nabla^2 f(x^0)$ 3: 4: **else** $H_n = I$ 5: 6: end if 7: while $\|\nabla f(x^n)\| > \varepsilon$ and $n < n_{\max}$ do Compute d^n by solving $H_n d^n = -\nabla f(x^n)$; 8: if $\nabla^2 f$ is given and $-\nabla f(x^n)^\top d^n < \min\{\alpha_1, \alpha_2 \| d^n \|^p \} \| d^n \|^2$ then 9: $d^n = -\nabla f(x^n)$ 10: end if 11: Compute a stepsize t_n using Armijo rule (see Program 1); 12:Set $x^{n+1} = x^n + t_n d^n$; 13:if $\nabla^2 f$ is given then 14: $H_{n+1} = \nabla^2 f(x^{n+1})$ 15:else 16: $s^{n} = x^{n+1} - x^{n}, y^{n} = \nabla f(x^{n+1}) - \nabla f(x^{n})$ 17:if $(y^n)^{\top} s^n > 0$ then 18:Set $H_{n+1} = H_n + \frac{y^n (y^n)^\top}{(y^n)^\top s^n} - \frac{H_n s^n (H_n s^n)^\top}{(s^n)^\top H_n s^n}$ 19:else 20: Set $H_{n+1} = I$ 21: end if 22: 23: end if Set n = n + 1; 24: 25: end while

Write the function in the form

```
[X] = globalnewtonmethod(fhandle, x0, epsilon, alpha1, alpha2, p, ...
```

, t0, alpha, beta, nmax, amax)

Test your program as follows:

- 1. Use the negative cosine function as in Part 1 with additional parameters p = 1/10and alpha1 = alpha2 = 1e-6. Write herefore an additional function file ncos.m which only returns the function- and gradient value.
- 2. Use the Rosenbrock function $f(x) = 100(x_2 x_1^2)^2 + (1 x_1)^2$, $x = (x_1, x_2)^\top \in \mathbb{R}^2$, with the parameter setting from Program 1 but with nmax=100, epsilon = 1e-5 and starting points [1;-0.5] and [-1.5; -1]. Set alpha1 = alpha2 = 1e-6 and

p = 1/10. Use the function file rosenbrock.m from Program 1 and write a new one, rosenbrockH.m, which additionally returns the Hessian matrix computed in x.

Compare the two methods under consideration and take a look at the following: Does the Armijo algorithm have to reduce the (initial) step size 1? In case the exact Hessian is used: When is the algorithm forced to set $d^n = -\nabla f(x^n)$ (Line 10)? In case of BFGS: Is the algorithm forced to reset $H_{n+1} = I$ (Line 21)?

Comment on your observations in the written report and visualize your results in suitable plots.