SS 2022 Submission: 24.05.21, until 12:15

Numerical Optimization - Sheet 7

If you are a student in mathematics please solve the exercises with no tag and the ones with the tag Mathematics. If you are a data science student please solve the problems with no tag and those with the tag Data Science. Submissions with tags other than your subject count as bonus points. The tag Programming marks programming exercises.

Show that for the problem

$$\min_{x \in \mathbb{R}^n} \frac{1}{2} x^\top Q x - x^\top b$$

for $Q \in \mathbb{R}^{n \times n}$ positive definite and symmetric and $b \in \mathbb{R}^n$, the Newton method converges after one step.

Apply Theorem 3.9 and Corollary 3.10 to show that a Newton method with Wolfe line search converges if the spectra of the Hessians satisfy $\sigma(\text{Hess}f(x^k)) \subset [m,L]$ with $0 < m \le L$ independent of k.

Implement the following quasi Newton and the Newton method. Your function has to provide the interface minimization_alg(f, gf, Hessf, x0, callback) where the first three arguments are callable objects which evaluate a function, its gradient and Hessian with respect to the Euclidean scalar product. The value x0 is the starting value and callback is a callback function (see below).

- (i) SR1: Broydens symmetric rank-1-update
 Implement broydensymm1(f, gf, Hessf, x0, callback) and run test(broydensymm1).
- (ii) BFGS (Broyden-Fletcher-Goldfarb-Shanno)Implement bfgs(f, gf, Hessf, x0, callback) and run test(bfgs).
- (iii) Newton method Implement newton(f, gf, Hessf, x0, callback) and run test(newton).

Hints:

- You might want to set an error tolerance and a maximal number of iterations as optional parameters
 of your algorithm.
- Please call the function callback in your algorithm in the form callback(xk) where xk is the current iterate. This allows the function test in the module newton_cg_test to plot your iterates and the errors.
- You find the function armijo in newton_cg_test.
- The broyden and bfgs update fail under certain conditions (see lecture). If that happens simply restart the current Hessian approximation by B = np.eye(n).