

Exercises Nr.6: due, December 5

19-11-2011

Ex.11.n1 [*'trust region method'*]

Consider the quadratic Taylor approximation of f near \mathbf{x}_k :

$$\begin{aligned} q(\mathbf{x}) &:= f(\mathbf{x}_k) + \nabla f(\mathbf{x}_k)^T (\mathbf{x} - \mathbf{x}_k) \\ &\quad + \frac{1}{2} (\mathbf{x} - \mathbf{x}_k)^T \nabla^2 f(\mathbf{x}_k) (\mathbf{x} - \mathbf{x}_k) \end{aligned}$$

Compute the descent step \mathbf{d}_k according to (*)

(*Levenberg-Marquardt*) and put $\mathbf{x}_{k+1} = \mathbf{x}_k + \mathbf{d}_k$, $\tau := \|\mathbf{d}_k\|$.

Show that \mathbf{x}_{k+1} is a local minimizer of the *trust region problem*:

$$\min q(\mathbf{x}) \quad \text{s.t.} \quad \|\mathbf{x} - \mathbf{x}_k\| \leq \tau$$

HINT: Show that the sufficient optimality conditions of Th.12.6 are satisfied.

Ex. Solve the program:

$$\min x^2 \quad \text{s.t.} \quad g(x) := 1 - x \leq 0$$

by applying the penalty method with $p(x) = (g^+(x))^2$ and by applying the exact penalty method, with $p(x) = g^+(x)$.

Ex.6.2 Show $X - xx^T \succeq 0 \Leftrightarrow \begin{pmatrix} 1 & x^T \\ x & X \end{pmatrix} \succeq 0$
and this (latter) constraint is convex in (x, X) .

Hint: Consider the condition

$$(\alpha, y)^T \begin{pmatrix} 1 & x^T \\ x & X \end{pmatrix} \begin{pmatrix} \alpha \\ y \end{pmatrix} \geq 0 \quad \forall \alpha \in \mathbb{R}, y \in \mathbb{R}^n$$