

## Numerical Optimization exercise sheet

review on 15.01.2024 during the exercise class

### 1. (Active set method)

The active set method is an optimization method for quadratic problems (QP) with (affine) linear equality and inequality constraints of the form

$$\begin{cases} \min_x f(x) := \frac{1}{2}x^T A x + a^T x \\ \text{s.t. } Bx = b, Cx \leq c, \end{cases} \quad (1)$$

where  $A$  is symmetric positive definite on  $\ker B$  and  $B$  has full rank. The idea is to reduce the problem to a sequence of QPs with only equality constraints. At an iterate  $x^{(k)} \in \mathcal{F}$  not being the solution  $x^*$ , we seek a *feasible descent direction*  $d^{(k)}$  by solving (see (3.6.19))

$$\begin{cases} \min_{d^{(k)}} f(x^{(k)} + d^{(k)}) \\ \text{s.t. } B d^{(k)} = 0, C_j d^{(k)} = 0, j \in \mathcal{A}(x^{(k)}) \end{cases}$$

which is equivalent to

$$\begin{cases} \min_{d^{(k)}} \frac{1}{2}(d^{(k)})^T A d^{(k)} + (d^{(k)})^T (A x^{(k)} + a) \\ \text{s.t. } D^{(k)} d^{(k)} = 0. \end{cases} \quad (2)$$

We know from the lecture, that there can now arise three possibilities.

Case 1:  $d^{(k)} = 0, \mu^{(k)} \geq 0$

Case 2:  $d^{(k)} = 0, \mu_j^{(k)} < 0$  for at least on  $j \in \mathcal{A}(x^{(k)})$

Case 3:  $d^{(k)} \neq 0$  is a feasible direction

- a) Derive the KKT system for (1) and (2).
- b) Let (1) be convex and let  $x^{(k)} \in \mathcal{F}$  be the current iterate. Show that, if for the solution  $d^{(k)}$  and the Lagrange multiplier  $\mu^{(k)}$  of (2) holds Case 1, then  $x^{(k)}$  is a solution of (1).
- c) Assume that we are in Case 2 and the Inactivation step has been performed, i.e. we have  $d^{(k)} = 0$  and  $\mu_j^{(k)} < 0$  for at least one  $j \in \mathcal{A}^{(k)}$ . Further, we have  $\tilde{d}^{(k)}, \tilde{\lambda}^{(k)}$  and  $\tilde{\mu}^{(k)}$  as the solution of problem (2) w.r.t. the set  $\tilde{\mathcal{A}}^{(k)} := \mathcal{A}^{(k)} \setminus \{j\}$  as described in the lecture notes. Show that  $\tilde{d}^{(k)}$  is a feasible direction at  $x^{(k)}$  by showing  $\tilde{d}^{(k)} \in \mathcal{L}(\mathcal{F}, x^{(k)})$ .  
**Hint:** Exploit the optimality conditions for  $d^{(k)}$  and  $\tilde{d}^{(k)}$ .
- d) Assume everything from c) holds and further assume that  $D^{(k)}$  has full rank for all  $k \in \mathbb{N}$ . Show that  $\tilde{d}^{(k)} \neq 0$ .

(2 + 4 + 6 + 4 = 16 Points)

2. (*Active set method*, MATLAB)

In this exercise we want to visualize the iterations of the active set method.

- a) Download the MATLAB function `activeset.m` from Moodle and adjust the method such that the `nullspace_method.m` is used to solve the subproblems.
- b) Apply the active set method to the problem

$$\begin{aligned} \min_{x \in \mathbb{R}^2} f(x) &= (x_1 - 1)^2 + (x_2 - 2.5)^2 \\ \text{s.t. } x_1 - 2x_2 + 2 &\geq 0 \\ -x_1 - 2x_2 + 6 &\geq 0 \\ -x_1 + 2x_2 + 2 &\geq 0 \\ x_1 &\geq 0 \\ x_2 &\geq 0 \end{aligned}$$

with the initial value  $x_0 = (2, 0)^T$  and the associated active set  $\mathcal{A}(x_0)$ .

- c) Plot the contours of  $f$  with the constraints and the path of the iterations of the active set method.

(2 + 4 + 6 = 12 Points)