

SF2822 Applied nonlinear optimization, final exam Thursday August 21 2014 14.00–19.00

Examiner: Anders Forsgren, tel. 08-790 71 27.

Allowed tools: Pen/pencil, ruler and eraser.

Note! Calculator is not allowed.

Solution methods: Unless otherwise stated in the text, the problems should be solved by systematic methods, which do not become unrealistic for large problems. Motivate your conclusions carefully. If you use methods other than what have been taught in the course, you must explain carefully.

Note! Personal number must be written on the title page. Write only one question per sheet. Number the pages and write your name on each page.

22 points are sufficient for a passing grade. For 20-21 points, a completion to a passing grade may be made within three weeks from the date when the results of the exam are announced.

1. Consider the nonlinear optimization problem (NLP) defined as

(NLP) minimize
$$\frac{1}{2}x_1^2 + \frac{1}{2}x_2^2$$

subject to $x_1 + x_2 + x_2^2 + 2 = 0$.

You have obtained a printout from an SQP solver for this problem. The initial point is $x = (0\ 0)^T$ and $\lambda = 0$. Six iterations, without linesearch, have been performed. The printout reads:

It	$ x_1 $	x_2	λ	$\ \nabla f(x) - \nabla g(x)\lambda\ $	$\ g(x)\ $
0	0	0	0	0	2
1	-1	-1	-1	2	1
2	-1.25	-0.25	-1.25	0.3750	0.5625
3	-1.7250	-0.4250	-1.7250	0.1663	0.0306
4	-1.7610	-0.3889	-1.7610	0.0026	0.0013
5	-1.7622	-0.3895	-1.7622	$1.5 \cdot 10^{-6}$	$4.0 \cdot 10^{-7}$
6	-1.7622	-0.3895	-1.7622	$2.8 \cdot 10^{-13}$	$9.2 \cdot 10^{-14}$

- (c) For the original problem (NLP), show that in this case the iterates converge to a global minimizer. (You need not verify the numerical values.) (2p)

Note: According to the convention of the book we define the Lagrangian $\mathcal{L}(x,\lambda)$ as $\mathcal{L}(x,\lambda) = f(x) - \lambda^T g(x)$, where f(x) the objective function and g(x) is the constraint function.

2. Consider the QP-problem (QP) defined as

$$(QP) \qquad \begin{array}{ll} \text{minimize} & \frac{1}{2}x_1^2 + \frac{1}{2}x_2^2 \\ \text{subject to} & x_1 + x_2 \ge 2. \end{array}$$

- (b) Show that $x(\mu)$ and $\lambda(\mu)$ which you obtained in (2a) converge to the optimal solution and Lagrange multiplier respectively of (QP).(3p)
- 3. Consider the semidefinite programming problem (P) defined as

(P) minimize
$$c^T x$$

subject to $G(x) \succeq 0$,

where $G(x) = \sum_{j=1}^{n} A_j x_j - B$ for B and A_j , j = 1, ..., n, are symmetric $m \times m$ -matrices. The corresponding dual problem is given by

maximize
$$\operatorname{trace}(BY)$$

(D) subject to $\operatorname{trace}(A_jY) = c_j, \quad j = 1, \dots, n,$
 $Y = Y^T \succ 0.$

A barrier transformation of (P) for a fixed positive barrier parameter μ gives the problem

$$(P_{\mu})$$
 minimize $c^{T}x - \mu \ln(\det(G(x)))$.

(a) Show that the first-order necessary optimality conditions for (P_{μ}) are equivalent to the system of nonlinear equations

$$c_j - \operatorname{trace}(A_j Y) = 0, \quad j = 1, \dots, n,$$

 $G(x)Y - \mu I = 0,$

- (c) In linear programming, when G(x) and Y are diagonal, it is not an issue how the equation $G(x)Y \mu I = 0$ is written. The linearizations of $G(x)Y \mu I = 0$ and $YG(x) \mu I = 0$ are then identical. Explain why this is in general not the case for semidefinite programming and how it can be handled. (2p)

Remark: For a symmetric matrix M we above use $M \succ 0$ and $M \succeq 0$ to denote that M is positive definite and positive semidefinite respectively. You may use the relations

$$\frac{\partial \ln(\det(G(x)))}{\partial x_j} = \operatorname{trace}(A_j G(x)^{-1}) \quad \text{for} \quad j = 1, \dots, n,$$

without proof.

4. Consider a nonlinear programming problem (NLP) defined by

(NLP) minimize
$$e^{x_1} + x_1x_2 + x_2^2 - 2x_2x_3 + x_3^2 - 2x_1 - x_2 - x_3$$

subject to $-x_1^2 - x_2^2 - x_3^2 + 5 \ge 0$, $a^Tx + 2 \ge 0$,

where $a \in \mathbb{R}^3$ is a given constant vector. Let $\widetilde{x} = (0 \ 1 \ 1)^T$.

- **5.** Consider the optimization problem

$$(NLP) \quad \begin{array}{ll} \underset{x \in \mathbb{R}^n}{\text{minimize}} & \frac{1}{2} \sum_{i \in \mathcal{U}} (p_i^T x - u_i)_+^2 + \frac{1}{2} \sum_{i \in \mathcal{L}} (l_i - p_i^T x)_+^2, \\ \text{subject to} & x \ge 0, \end{array}$$

where \mathcal{L} and \mathcal{U} are nonintersecting index sets such that $\mathcal{L} \bigcup \mathcal{U} = \{1, \ldots, m\}$, and the subscript "+" denotes the positive part, i.e., $x_+ = \max(x, 0)$. The constants u_i , $i \in \mathcal{U}$, and l_i , $i \in \mathcal{L}$, are known as well as the constant vectors p_i , $i = 1, \ldots, m$. This means that we pay a quadratic penalty cost for violating lower bounds l_i , $i \in \mathcal{L}$, and upper bounds u_i , $i \in \mathcal{U}$, respectively.

The formulation (NLP) is straightforward, but a drawback is that the objective function is not twice-continuously differentiable. Your task is to show that we may obtain a smooth problem by introducing additional variables and constraints.

- (b) Show that (NLP) is equivalent to the quadratic programming problem

$$(QP) \quad \begin{array}{ll} \underset{x \in \mathbb{R}^n, y \in \mathbb{R}^m}{\text{minimize}} & \frac{1}{2} \sum_{i \in \mathcal{U}} y_i^2 + \frac{1}{2} \sum_{i \in \mathcal{L}} y_i^2, \\ \text{subject to} & y_i \geq p_i^T x - u_i, \ i \in \mathcal{U}, \\ & y_i \geq l_i - p_i^T x, \ i \in \mathcal{L}, \\ & x \geq 0. \end{array}$$

Note: The motivation for considering this reformulation is that we obtain a smooth problem. The increased dimensionality introduced by the y variables can be eliminated in the linear equations that are solved in a primal-dual interior method.