

# CS5321 Numerical Optimization Homework 5

Due May 9

1. (30%) Use the optimality conditions of constrained optimization problems to verify the following properties.

(a) The optimal solution of the total least square problem is  $A^T A \vec{x} = \lambda \vec{x}$  for some  $\lambda$ .

(b) For the trust region method, the optimal solution  $\vec{p}^*$  of the local model

$$\min_{\vec{p} \in \mathbb{R}^n} m(\vec{p}) = \vec{g}^T \vec{p} + \frac{1}{2} \vec{p}^T A \vec{p} \quad \text{s.t.} \quad \vec{p}^T \vec{p} \leq \Delta^2,$$

satisfies

$$(A + \lambda I) \vec{p}^* = -\vec{g}, \quad \lambda(\Delta - \|\vec{p}^*\|) = 0, \quad \text{and} \quad (A + \lambda I) \text{ is positive definite.}$$

2. (30%) For a quadratic programming,

$$\begin{aligned} \min_{\vec{x}} g(\vec{x}) &= \frac{1}{2} \vec{x}^T G \vec{x} + \vec{x}^T \vec{c} \\ \text{s.t.} \quad A \vec{x} &= \vec{b}, \end{aligned}$$

Prove that if  $A$  has full row-rank and the reduced Hessian  $Z^T G Z$  is positive definite, where  $\text{span}\{Z\}$  is the null space of  $\text{span}\{A^T\}$ , then the KKT matrix

$K = \begin{bmatrix} G & A^T \\ A & 0 \end{bmatrix}$  is nonsingular. (Hint: Prove that every vector  $\begin{bmatrix} \vec{w} \\ \vec{v} \end{bmatrix}$  making  $\begin{bmatrix} G & A^T \\ A & 0 \end{bmatrix} \begin{bmatrix} \vec{w} \\ \vec{v} \end{bmatrix} = 0$  is a zero vector. Using the property that  $\vec{w}^T G \vec{w} > 0$ .)

3. (40%) Consider the quadratic programming problem with bounded constraints

$$\begin{aligned} \min_{x_1, x_2, x_3} & (x_1 - 4)^2 + (x_2 - 3)^2 + (x_3 - 2)^2 \\ \text{s.t.} \quad & 0 \leq x_1, x_2, x_3 \leq 2 \end{aligned}$$

Use gradient projection method to find its optimal solution with  $\vec{x}_0 = 0$ . Write down the trace, like

$$\vec{x}_0 = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \vec{x}_1 = \begin{pmatrix} 2 \\ 3/2 \\ 1 \end{pmatrix} \rightarrow \vec{x}_2 = \dots$$