

# Homework (19)

1. [Nonsingularity of the KKT matrix.] Consider the KKT matrix

$$\begin{bmatrix} \mathbf{P} & \mathbf{A}^T \\ \mathbf{A}^0 & \end{bmatrix},$$

where  $\mathbf{P} \in \mathbb{S}_+^n$ ,  $\mathbf{A} \in \mathbb{R}^{p \times n}$ , and  $\text{rank } \mathbf{A} = p < n$ . Show that each of the following statements is equivalent to nonsingularity of the KKT matrix:

- $\mathcal{N}(\mathbf{P}) \cap \mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$ .
- $\mathbf{Ax} = \mathbf{0}, \mathbf{x} \neq \mathbf{0} \implies \mathbf{x}^T \mathbf{Px} > 0$ .
- $\mathbf{F}^T \mathbf{PF} \succ \mathbf{0}$ , where  $\mathbf{F} \in \mathbb{R}^{n \times (n-p)}$  is a matrix for which  $\mathcal{R}(\mathbf{F}) = \mathcal{N}(\mathbf{A})$ .
- $\mathbf{P} + \mathbf{A}^T \mathbf{QA} \succ \mathbf{0}$  for some  $\mathbf{Q} \succeq \mathbf{0}$ .

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2. [Projected gradient method.]

- Let  $(\mathbf{v}, \mathbf{w})$  be the unique solution of

$$\begin{bmatrix} \mathbf{I} & \mathbf{A}^T \\ \mathbf{A} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{v} \\ \mathbf{w} \end{bmatrix} = \begin{bmatrix} -\nabla f(\mathbf{x}) \\ \mathbf{0} \end{bmatrix}.$$

Show that

$$\mathbf{v} = \operatorname{argmin}_{\mathbf{A}\mathbf{u}=\mathbf{0}} \| -\nabla f(\mathbf{x}) - \mathbf{u} \|_2, \quad \mathbf{w} = \operatorname{argmin}_{\mathbf{y}} \| \nabla f(\mathbf{x}) + \mathbf{A}^T \mathbf{y} \|_2.$$

- What is the relationship between the projected negative gradient  $\mathbf{v}$  and the negative gradient of the reduced unconstrained problem  $f(\mathbf{F}\mathbf{z} + \tilde{\mathbf{x}})$ , assuming  $\mathbf{F}^T \mathbf{F} = \mathbf{I}$ ?

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3. [Equality constrained entropy maximization.] We consider the equality constrained entropy maximization problem

$$\min_{\mathbf{x}} f(\mathbf{x}) = \sum_{i=1}^n x_i \log x_i, \quad s.t. \quad \mathbf{A}\mathbf{x} = \mathbf{b}.$$

Use the following methods to solve it at  $p = 100$  and  $n = 500$  (you may generate  $\mathbf{A}$  and  $\mathbf{b}$  randomly):

- Direct projected gradient, with inexact line search.
- Elimiate equality constraint.
- Dual approach.

Write a report to describe your settings and compare their performance (numerical accuracy vs. iteration number). Codes should also be handed in.