

Solutions Sheet

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November 22, 2021

Exercise 1

Given: Let $m \leq n \leq k, y \in \mathbb{R}^m, b \in \mathbb{R}^k$ and $A \in \mathbb{R}^{m \times n}, B \in \mathbb{R}^{k \times n}$. We are considering the following optimization problem:

$$\begin{aligned} \min_{x \in \mathbb{R}^n} & \|Ax - y\|_2^2 \\ \text{s.t.} & Bx = b \end{aligned}$$

Task: Find a matrix $P \in \mathbb{R}^{(n+k) \times (n+k)}$ and a vector $p \in \mathbb{R}^{n+k}$ such that solving :

$$P \begin{bmatrix} x \\ \lambda \end{bmatrix} = p$$

gives a critical point for the optimization problem.

Proof. **Solution:** We will start by defining the Lagrangian function associated to this problem:

$$L(\lambda) = \|Ax - y\|_2^2 + \lambda^T \cdot (Bx - b)$$

We will now search for the derivatives with respect to x and λ by using Matrix differential calculus:

- We will start by calculating $D_x L$

$$\begin{aligned} dL &= d\|Ax - y\|_2^2 + d\lambda^T (Bx - b) \\ &= d(Ax - y)^T (Ax - y) + \lambda^T B dx \\ &= 2(Ax - y)^T d(Ax - y) + \lambda^T B dx \\ &= 2(Ax - y)^T A dx + \lambda^T B dx \end{aligned}$$

So :

$$D_x L = 2(Ax - y)^T A + \lambda^T B$$

- We will now calculate $D_\lambda L$:

$$\begin{aligned} dL &= d\|Ax - y\|_2^2 + d\lambda^T (Bx - b) \\ &= (Bx - b)^T d\lambda \end{aligned}$$

So we have :

$$\nabla L = (2(Ax - y)^T A dx + \lambda^T B, (Bx - b)^T)$$

- For $\nabla L = 0$ we have :

$$\begin{aligned} \begin{bmatrix} 0 \\ 0 \end{bmatrix} &= \begin{bmatrix} (Bx - b)^T \\ 2(Ax - y)^T A + \lambda^T B \end{bmatrix} \\ \Leftrightarrow \begin{bmatrix} b \\ 2y^T A \end{bmatrix} &= \begin{bmatrix} Bx \\ 2(Ax)^T A^T A + \lambda^T B \end{bmatrix} \\ \Leftrightarrow \begin{bmatrix} b \\ 2(y^T A)^T \end{bmatrix} &= \begin{bmatrix} Bx \\ 2A^T Ax + B^T \lambda \end{bmatrix} \\ \Leftrightarrow \begin{bmatrix} b \\ 2(y^T A)^T \end{bmatrix} &= \begin{bmatrix} B & 0^{k \times n} \\ 2A^T A & B^T \end{bmatrix} \begin{bmatrix} x \\ \lambda \end{bmatrix} \end{aligned}$$

So with $p = \begin{bmatrix} b \\ 2(y^T A)^T \end{bmatrix}$ and $P = \begin{bmatrix} B & 0^{k \times n} \\ 2A^T A & B^T \end{bmatrix}$ solving

$$P \begin{bmatrix} x \\ \lambda \end{bmatrix} = p$$

will give a critical point to the optimization problem.

□

Exercise 2

Given: Let $v \in [0, 1]$. We then have the following optimization problem:

$$\begin{aligned} \min_{w, b, \xi, p} & \frac{1}{2} \|w\|^2 - vp + \frac{1}{n} \sum_{i=1}^n \xi_i \\ \text{s.t.} & y_i(w^T x_i + b) \geq p - \forall i \\ & \xi \geq 0 \forall i \\ & p \geq 0 \end{aligned}$$

- (a) The Lagrangian is given by

Exercise 3

Exercise 4