

## Exercise 1

- Let us consider inactive constraints first

$$\forall i \in I \setminus A(x): \quad a_i^T x - b_i =: \delta_i > 0$$

$$\forall p \in \mathbb{R}^n \quad \text{and} \quad \forall \delta \in \left[0, \frac{\delta_i}{\|a_i\| \|p\|}\right) \quad \leftarrow \begin{array}{l} \text{positive number} \\ \text{(possibly infinity)} \end{array}$$

$$a_i^T(x + \delta p) = a_i^T x + \delta a_i^T p \geq a_i^T x - \delta \|a_i\| \|p\|$$

$$\geq a_i^T x - \delta_i = b_i$$

Thus any direction is feasible with respect to inactive constraints.

- Let us consider equality constraints now.

For any  $\delta > 0$ :

$$a_i^T(x + \delta p) = a_i^T x + \delta a_i^T p$$

$$= b_i + \delta a_i^T p = b_i \quad i \in E$$

if and only if  $a_i^T p = 0, i \in E$

- Similarly, for active inequality constraints:

$$\text{For any } \delta > 0: \quad a_i^T(x + \delta p) = a_i^T x + \delta a_i^T p$$

$$= b_i + \delta a_i^T p \geq b_i, \quad i \in I \cap A(x)$$

if and only if  $a_i^T p \geq 0, i \in I \cap A(x)$

○ Therefore, the direction  $p \in \mathbb{R}^n$   
is feasible for linear constraints  
if and only if

$$a_i^T p = 0, \quad i \in \mathcal{E}$$

○  $a_i^T p \geq 0, \quad i \in \mathcal{I} \cap \mathcal{A}(x)$

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## Exercise 2

$$\text{Put } \Omega = \{ p \in \mathbb{R}^n \mid \|p\| \leq \Delta_k \}$$

$$\text{Clearly } \pi_{\Omega}(x) = \begin{cases} x, & \|x\| \leq \Delta_k \\ \Delta_k \frac{x}{\|x\|}, & \|x\| > \Delta_k \end{cases}$$

Suppose that  $p$  is a locally optimal solution to the trust-region problem.

Then  $\pi_{\Omega}[p - \alpha \nabla m_k(p)] = p, \forall \alpha > 0$   
by (5).

If  $\|p - \alpha \nabla m_k(p)\| \leq \Delta_k$  for some  $\alpha > 0$

$$\Rightarrow \pi_{\Omega}[p - \alpha \nabla m_k(p)] = p - \alpha \nabla m_k(p) = p$$

for this  $\alpha > 0 \Rightarrow \nabla m_k(p) = 0$ .

Therefore, we can take  $\lambda = 0$  and get

$$\nabla m_k(p) = B_k p + g_k = 0$$

$$B_k p = -g_k$$

$$\begin{cases} (B_k + \lambda I)p = -g_k \end{cases}$$

$$\begin{cases} \lambda(\|p\| - \Delta_k) = 0 \quad (\text{because } \lambda \geq 0) \end{cases}$$

○ If  $\|p - \alpha \nabla m_k(p)\| > \Delta_k$ ,  $\forall \alpha > 0$

$$\Rightarrow \frac{p - \alpha \nabla m_k(p)}{\|p - \alpha \nabla m_k(p)\|} = p, \forall \alpha > 0$$

∴  $p = \frac{p - \alpha \nabla m_k(p)}{\|p - \alpha \nabla m_k(p)\|}$

$$\begin{aligned} \nabla m_k(p) &= -\alpha^{-1} \left( \frac{\|p - \alpha \nabla m_k(p)\|}{\Delta_k} - 1 \right) p \\ &= -\lambda_\alpha p \end{aligned}$$

Observe that  $\lambda_\alpha > 0$  (we assumed that

$\|p - \alpha \nabla m_k(p)\| > \Delta_k$ ). In fact, since

$\nabla m_k(p)$  and  $p$  do not depend on  $\alpha$

○  $\Rightarrow \nabla m_k(p) = -\lambda p$ , for some  $\lambda > 0$ .

Therefore

$$B_k p + g_k = -\lambda p$$

$$\left\{ \begin{aligned} (B_k + \lambda I)p &= -g_k, \lambda > 0 \\ \lambda (\|p\| - \Delta_k) &= \\ \lambda \left( \frac{\|p - \alpha \nabla m_k(p)\|}{\Delta_k} - \Delta_k \right) &= 0 \end{aligned} \right.$$

### Exercise 3

$\Omega$  - closed, convex, non-empty

$z \in \mathbb{R}^n \setminus \Omega$

Consider the optimality conditions (6)

for the projection problem:

$x^* \in \Omega$ :

$$(x^* - z)^T (x - x^*) \geq 0 \quad \forall x \in \Omega$$

Put  $a = x^* - z \neq 0$  ( $x^* \in \Omega$ ,  $z \notin \Omega$ ,

$\Omega$  is closed)

$\forall x \in \Omega$ :

$$a^T x \geq a^T x^*$$

$$\text{Also: } \|x^* - z\|_2^2 = (x^* - z)^T (x^* - z) > 0$$

$$\Rightarrow (x^* - z)^T x^* > (x^* - z)^T z$$

$$a^T x^* > a^T z \Rightarrow$$

Thus we can take  $\beta := a^T x^*$  ▣



- separating hyperplane

## Exercise 4

$\Omega$ -convex,  $\bar{x} \in \Omega$

$$\bullet \vec{0}^T \cdot (x - \bar{x}) = 0 \leq 0 \quad \forall x \in \Omega$$

$$\Rightarrow \vec{0} \in N_{\Omega}(\bar{x}) \Rightarrow N_{\Omega}(\bar{x}) \text{ is non-empty}$$

$$\bullet q_1, q_2 \in N_{\Omega}(\bar{x})$$

$$0 \leq \lambda \leq 1$$

$$q = \lambda q_1 + (1-\lambda)q_2$$

$$q^T(x - \bar{x}) = \underbrace{\lambda}_{\geq 0} \underbrace{q_1^T(x - \bar{x})}_{\leq 0} + \underbrace{(1-\lambda)}_{\geq 0} \underbrace{q_2^T(x - \bar{x})}_{\leq 0} \leq 0$$

$$\Rightarrow q \in N_{\Omega}(\bar{x}) \Rightarrow N_{\Omega}(\bar{x}) \text{ is convex}$$

$$\bullet q \in N_{\Omega}(\bar{x}), \alpha > 0:$$

$$\underbrace{\alpha}_{> 0} \underbrace{q^T(x - \bar{x})}_{\leq 0} \leq 0 \Rightarrow \alpha q \in N_{\Omega}(\bar{x})$$

$\Rightarrow N_{\Omega}(\bar{x})$  is a cone.

$$\bullet \text{Suppose that } \{q_i\} \rightarrow \bar{q}, \quad q_i \in N_{\Omega}(\bar{x})$$

$$\text{Then } \bar{q}^T(x - \bar{x}) = \lim_{i \rightarrow \infty} \underbrace{q_i^T(x - \bar{x})}_{\leq 0} \leq 0$$

$$\Rightarrow \bar{q} \in N_{\Omega}(\bar{x}) \Rightarrow N_{\Omega}(\bar{x}) \text{ is closed} \quad \blacksquare$$