



Assume that  $\Omega \subset \mathbb{R}^n$  is a closed and convex set. Recall that we can define in this case for  $x \in \Omega$  the normal cone  $N_\Omega(x)$  as

$$N_\Omega(x) = \{q \in \mathbb{R}^n : q^T(\hat{x} - x) \leq 0 \text{ for all } \hat{x} \in \Omega\},$$

and the tangent cone  $T_\Omega(x)$  as

$$T_\Omega(x) = \{p \in \mathbb{R}^n : p^T q \leq 0 \text{ for all } q \in N_\Omega(x)\}.$$

**1** Assume that  $\Omega \subset \mathbb{R}^n$  is non-empty, convex, and closed, and that  $x \in \Omega$ . Show that  $N_\Omega(x)$  and  $T_\Omega(x)$  are non-empty, closed, and convex cones.<sup>1</sup>

**2** Consider the sets

$$\Omega_1 := \{x \in \mathbb{R}^n : \|x\|_\infty \leq 1\}$$

and

$$\Omega_2 := \{x \in \mathbb{R}^n : \|x\|_2 \leq 1\}.$$

- a) Show that  $\Omega_1$  and  $\Omega_2$  are non-empty, convex, and closed.
- b) In dimension  $n = 2$ , determine the normal and tangent cones to the sets  $\Omega_1$  and  $\Omega_2$  at the point  $x = (1, 0)$ . In addition, determine the normal and tangent cones to  $\Omega_1$  at the point  $x = (1, 1)$ .

*Hint: This exercise will probably become easier if you start with drawing a sketch.*

c) Consider the projection problems

$$\pi_{\Omega_i}(z) = \arg \min_{x \in \Omega_i} \frac{1}{2} \|x - z\|_2^2.$$

Use the variational characterisation of the projection ( $(\pi_{\Omega_i}(z) - z)^T(x - \pi_{\Omega_i}(z)) \geq 0$  for all  $x \in \Omega_i$ ) to verify the formulas

$$\pi_{\Omega_1}(z) = (\max\{-1, \min\{1, z_i\}\})_{i=1, \dots, n}$$

and

$$\pi_{\Omega_2}(z) = \frac{z}{\max\{1, \|z\|_2\}}.$$

<sup>1</sup>A subset  $C \subset \mathbb{R}^n$  is called a cone, if, whenever  $p \in C$ , then also  $\lambda p \in C$  for all  $\lambda > 0$ .

d) Consider now the case  $n = 2$  and let

$$f(x) = x_1^2 + (x_2 + 2)^2.$$

Find the global solution of the problem

$$\min_{x \in \Omega_1} f(x)$$

(you can do this graphically, if you want) and perform one step of the gradient projection method

$$x^{(k+1)} \leftarrow \pi_{\Omega_1}(x^{(k)} - \alpha \nabla f(x^{(k)}))$$

with step length  $\alpha = 1/2$  starting at  $x^{(0)} = (1, 1)$ .

- 3 Assume that  $\Omega \subset \mathbb{R}^n$  is non-empty, convex and closed. Show that the projection mapping  $\pi_{\Omega}: \mathbb{R}^d \rightarrow \Omega$  is a non-expansive map in the sense that

$$\|\pi_{\Omega}(x) - \pi_{\Omega}(y)\|_2 \leq \|x - y\|_2$$

for all  $x, y \in \mathbb{R}^n$ .

*Hint: Show first that*

$$(\pi_{\Omega}(x) - x)^T (\pi_{\Omega}(y) - \pi_{\Omega}(x)) \geq 0 \quad \text{and} \quad (\pi_{\Omega}(y) - y)^T (\pi_{\Omega}(x) - \pi_{\Omega}(y)) \geq 0.$$

*Then consider the function  $g: \mathbb{R} \rightarrow \mathbb{R}$ ,*

$$g(\lambda) := \|\lambda(x - y) + (1 - \lambda)(\pi_{\Omega}(x) - \pi_{\Omega}(y))\|_2^2.$$

*Show that  $g'(0) \geq 0$  and deduce from the fact that  $g$  is quadratic that  $g(1) \geq g(0)$ .*

- 4 Let  $A \in \mathbb{R}^{m \times n}$  with  $m \geq n$  have full rank, let  $b \in \mathbb{R}^m$ , and let  $\Omega \subset \mathbb{R}^n$  be non-empty, convex, and closed. Consider the restricted least squares problem

$$\min_{x \in \Omega} f(x) \quad \text{where } f(x) = \frac{1}{2} \|Ax - b\|_2^2 \quad (1)$$

and the gradient projection algorithm

$$x^{(k+1)} \leftarrow \pi_{\Omega}(x^{(k)} - \alpha \nabla f(x^{(k)})).$$

Show that this algorithm converges to the unique solution of (1) provided that  $0 < \alpha < 2/\sigma_{\max}^2$ , where  $\sigma_{\max}$  denotes the largest singular value of  $A$ .

*Hint: Show that the gradient descent step  $x \mapsto x - \alpha \nabla f(x)$  is a contraction on  $\mathbb{R}^n$ , and then use the result of the previous exercise.*