

## Quiz #1 Solutions

1. Consider the set in  $\mathbb{R}^{2n}$

$$\mathcal{E} := \left\{ (x, y) \in \mathbb{R}^n \times \mathbb{R}^n : \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}^T \begin{pmatrix} A_{xx} & A_{xy} & a_x \\ A_{xy}^T & A_{yy} & a_y \\ a_x^T & a_y^T & \alpha \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} \leq 1 \right\},$$

where  $a_x, a_y \in \mathbb{R}^n$ , and the symmetric matrix

$$A := \begin{pmatrix} A_{xx} & A_{xy} \\ A_{xy}^T & A_{yy} \end{pmatrix}$$

is positive-definite.

- (a) What is the shape of  $\mathcal{E}$ ? Express your answer in geometric terms, assuming that some relevant eigenvalue decomposition is available.

**Solution:** in the space of  $z = (x, y) \in \mathbb{R}^{2n}$ , the set  $\mathcal{E}$  has the form

$$\mathcal{E} = \{z : z^T A z + 2a^T z + \alpha \leq 1\},$$

where  $a = (a_x, a_y) \in \mathbb{R}^{2n}$ . We can write the condition

$$z^T A z + 2a^T z + \alpha \leq 1$$

equivalently as

$$(z - z_0)^T A (z - z_0) \leq \gamma := 1 - \alpha + z_0^T A z_0$$

where  $z_0 := A^{-1}a$ . Thus, if  $\alpha > 1 + z_0^T A z_0 = 1 + a^T A^{-1}a$ , the above set is empty; if  $\alpha = 1 + a^T A^{-1}a$ , the set is the singleton  $\{z_0\}$ . Otherwise, the set is an ellipse having center at  $z_0$ , and with semi-axis lengths given by  $\sqrt{\gamma/\lambda_i}$ ,  $i = 1, \dots, n$ , with  $\lambda_i$  the inverse eigenvalues of  $A$ , and principal directions given by the eigenvectors.

- (b) Likewise, determine the projection of  $\mathcal{E}$  onto the space of  $x$ -variables.

**Solution:** The projection  $\mathcal{P}$  is the set of points  $x \in \mathbb{R}^n$  such that

$$\begin{aligned} 1 &\geq \min_y \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}^T \begin{pmatrix} A_{xx} & A_{xy} & a_x \\ A_{xy}^T & A_{yy} & a_y \\ a_x^T & a_y^T & \alpha \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} \\ &= \min_y x^T A_{xx} x + y^T A_{yy} y + 2x^T A_{xy} y + 2a_x^T x + 2a_y^T y + \alpha \end{aligned} \quad (1)$$

Solving for the minimum simply entails taking the derivative with respect to  $y$  and setting it to zero. This leads to the optimality condition

$$A_{yy} y + A_{xy}^T x + a_y = 0.$$

Solving for  $y$ , we obtain the optimal solution

$$y^*(x) = -A_{yy}^{-1}(A_{xy}^T x + a_y)$$

Replacing  $y$  by its optimal value in (1), we conclude that the condition  $x \in \mathcal{P}$  is equivalent to

$$x^T(A_{xx} - A_{xy}A_{yy}^{-1}A_{xy}^T)x + 2(a_x - A_{yy}^{-1}a_y)^T x + \beta \leq 1,$$

where

$$\beta := \alpha - a_y^T A_{yy}^{-1} a_y.$$

The above condition can be analyzed as before.

2. Find the Fenchel conjugate  $f^*(y) = \max_x x^T y - f(x)$  for the following functions.

(a)  $f(x) = \frac{1}{2}(a^T x)^2 + b^T x$  on  $\mathbb{R}^n$ .

(b)  $f(x) = \max(0, 1 - c^T x)$  on  $\mathbb{R}^n$ .

**Solution:** both cases are special cases of functions of the form

$$f(x) = h(a^T x) + b^T x,$$

where  $h$  is a univariate function. The conjugate of a function  $f(x) = g(x) + b^T x$  is simply  $g^*(y - b)$ , hence we first examine the case  $b = 0$ .

Assume  $a \neq 0$ . Decompose  $x$  as  $x = ta + u$ , with  $a^T u = 0$ , and work with variables  $(t, u)$  and the constraint  $a^T u = 0$ . We obtain

$$f^*(y) = \max_{t, u: a^T u = 0} (ta + u)^T y - h((ta + u)^T a) = \max_t t(y^T a) - h(t(a^T a)) + \max_{u: a^T u = 0} y^T u.$$

The first term is nothing else than  $h^*((y^T a)/(a^T a))$ , and the second term is finite (and zero) if and only if  $y = \alpha a$  for some  $\alpha \in \mathbb{R}$ . We obtain

$$f^*(y) = \begin{cases} h^*(\alpha) & \text{if } y = \alpha a \text{ for some } \alpha \in \mathbb{R}, \\ +\infty & \text{otherwise.} \end{cases}$$

When  $b \neq 0$ , we have

$$f^*(y) = \begin{cases} h^*(\alpha) & \text{if } y = b + \alpha a \text{ for some } \alpha \in \mathbb{R}, \\ +\infty & \text{otherwise.} \end{cases}$$

This allows to solve the two problems at once, by using  $h(t) = t^2/2$  for the first part, and  $h(t) = \max(0, 1 - t)$  for the second part.

The conjugate of  $h(t) = t^2/2$  is simply itself; for the second function, we have

$$\begin{aligned} h^*(\alpha) &= \max_t \alpha t - \max(0, 1 - t) \\ &= \max \left( \max_{t \geq 1} \alpha t, \max_{t \leq 1} (\alpha + 1)t - 1 \right). \end{aligned}$$

The first term is  $+\infty$  when  $\alpha > 0$ ,  $\alpha$  otherwise; the second is  $+\infty$  when  $\alpha < -1$ , and  $\alpha$  otherwise. Hence

$$h^*(\alpha) = \begin{cases} \alpha & \text{if } \alpha \in [-1, 0], \\ +\infty & \text{otherwise.} \end{cases}$$