

**Exercise 3.18 [Boyd & Vandenberghe, 2004]**

Adapt the proof of concavity of the log-determinant function in §3.1.5 to show the following.

- (a)  $f(X) = \mathbf{tr}(X^{-1})$  is convex on  $\mathbf{dom} f = \mathbf{S}_{++}^n$ .
- (b)  $f(X) = (\det X)^{1/n}$  is concave on  $\mathbf{dom} f = \mathbf{S}_{++}^n$ .

*Ans:* Here only does part (a).

- (a) Let  $Z \in \mathbf{S}_{++}^n$ ,  $V \in \mathbf{S}^n$ , and  $(a, b) \subseteq \mathbb{R}$ , an interval such that  $\forall t \in (a, b), Z + tV \in \mathbf{S}_{++}^n$ . Define  $g(t) = f(Z + tV) = \mathbf{tr}((Z + tV)^{-1})$ .

**Exercise 3.19 [Boyd & Vandenberghe, 2004]**

*Nonnegative weighted sums and integrals.*

(a) Show that  $f(x) = \sum_{i=1}^r \alpha_i x_{[i]}$  is a convex function of  $x$ , where  $\alpha_1 \geq \alpha_2 \geq \dots \geq \alpha_r \geq 0$ , and  $x_{[i]}$  denotes the  $i$ th largest component of  $x$ . (You can use the fact that  $f(x) = \sum_{i=1}^k x_{[i]}$  is convex on  $\mathbb{R}^n$ .)

(b) Let  $T(x, \omega)$  denote the trigonometric polynomial

$$T(x, \omega) = x_1 + x_2 \cos \omega + x_3 \cos 2\omega + \dots + x_n \cos(n-1)\omega.$$

Show that the function

$$f(x) = - \int_0^{2\pi} \log T(x, \omega) d\omega$$

is convex on  $\{x \in \mathbb{R}^n \mid T(x, \omega) > 0, 0 \leq \omega \leq 2\pi\}$ .

*Ans:* Here only does part (a).

**Exercise 3.22 [Boyd & Vandenberghe, 2004]**

*Composition rules.* Show that the following functions are convex.

- (a)  $f(x) = -\log(-\log(\sum_{i=1}^m e^{a_i^T x + b_i}))$  on  $\text{dom } f = \{x \mid \sum_{i=1}^m e^{a_i^T x + b_i} < 1\}$ . You can use the fact that  $\log(\sum_{i=1}^n e^{y_i})$  is convex.
- (b)  $f(x, u, v) = -\sqrt{uv - x^T x}$  on  $\text{dom } f = \{(x, u, v) \mid uv > x^T x, u, v > 0\}$ . Use the fact that  $x^T x/u$  is convex in  $(x, u)$  for  $u > 0$ , and that  $-\sqrt{x_1 x_2}$  is convex on  $\mathbb{R}_{++}^2$ .
- (c)  $f(x, u, v) = -\log(uv - x^T x)$  on  $\text{dom } f = \{(x, u, v) \mid uv > x^T x, u, v > 0\}$ .
- (d)  $f(x, t) = -(t^p - \|x\|_p^p)^{1/p}$  where  $p > 1$  and  $\text{dom } f = \{(x, t) \mid t \geq \|x\|_p\}$ . You can use the fact that  $\|x\|_p^p/u^{p-1}$  is convex in  $(x, u)$  for  $u > 0$  (see exercise 3.23), and that  $-x^{1/p}y^{1-1/p}$  is convex on  $\mathbb{R}_+^2$  (see exercise 3.16).
- (e)  $f(x, t) = -\log(t^p - \|x\|_p^p)$  where  $p > 1$  and  $\text{dom } f = \{(x, t) \mid t > \|x\|_p\}$ . You can use the fact that  $\|x\|_p^p/u^{p-1}$  is convex in  $(x, u)$  for  $u > 0$  (see exercise 3.23).

*Ans:* Here only does part (c).

**Additional Exercise 2.5 [Boyd & Vandenberghe, 2017]**

A *perspective composition rule* [Marèchal]. Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be a convex function with  $f(0) \leq 0$ .

(a) Show that the perspective  $tf(x/t)$ , with domain  $\{(x, t) \mid t > 0, x/t \in \mathbf{dom} f\}$ , is nonincreasing as a function of  $t$ .

(b) Let  $g$  be concave and positive on its domain. Show that the function

$$h(x) = g(x)f(x/g(x)), \quad \mathbf{dom} h = \{x \in \mathbf{dom} g \mid x/g(x) \in \mathbf{dom} f\}$$

is convex.

(c) As an example, show that

$$h(x) = \frac{x^T x}{\left(\prod_{k=1}^n x_k\right)^{1/n}}, \quad \mathbf{dom} h = \mathbb{R}_{++}^n$$

is convex. *Ans:* Here only does part (a) and (b).

**Additional Exercise 2.30 [Boyd & Vandenberghe, 2017]**

*Huber penalty.* The infimal convolution of two functions  $f$  and  $g$  on  $\mathbb{R}^n$  is defined as

$$h(x) = \inf_y (f(y) + g(x - y))$$

(see exercise 2.17). Show that the infimal convolution of  $f(x) = \|x\|_1$  and  $g(x) = (1/2)\|x\|_2^2$ , *i.e.*, the function

$$h(x) = \inf_y (f(y) + g(x - y)) = \inf_y \left( \|y\|_1 + \frac{1}{2}\|x - y\|_2^2 \right)$$

is the *Huber penalty*

$$h(x) = \sum_{i=1}^n \phi(x_i), \quad \phi(u) = \begin{cases} u^2/2, & |u| \leq 1 \\ |u| - 1/2, & |u| > 1 \end{cases}$$

*Ans:*

**Additional Exercise 2.31 [Boyd & Vandenberghe, 2017]**

Suppose the function  $h : \mathbb{R} \rightarrow \mathbb{R}$  is convex, nondecreasing, with  $\text{dom } h = \mathbb{R}$ , and  $h(t) = h(0)$  for  $t \leq 0$ .

(a) Show that the function  $f(x) = h(\|x\|_2)$  is convex on  $\mathbb{R}^n$ .

(b) Show that the conjugate of  $f$  is  $f^*(y) = h^*(\|y\|_2)$ .

(c) As an example, derive the conjugate of  $f(x) = (1/p)\|x\|_2^p$  for  $p > 1$ , by applying the result of part (b) with the function

$$h(t) = \frac{1}{p} \max\{0, t\}^p = \begin{cases} \frac{1}{p} t^p, & t \geq 0 \\ 0, & t < 0 \end{cases}$$

. *Ans:*

**Additional Exercise 3.17 [Boyd & Vandenberghe, 2017]**

*Minimum fuel optimal control.* Solve the minimum fuel optimal control problem described in exercise 4.16 of *Convex Optimization*, for the instance with problem data

$$A = \begin{bmatrix} -1 & 0.4 & 0.8 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}, b = \begin{bmatrix} 1 \\ 0 \\ 0.3 \end{bmatrix}, x_{des} = \begin{bmatrix} 7 \\ 2 \\ -6 \end{bmatrix}, N = 30.$$

You can do this by forming the LP you found in your solution of exercise 4.16, or more directly using CVX. Plot the actuator signal  $u(t)$  as a function of time  $t$ .

*Ans:*