

Problem 1. Prove that, given any set C and a point $x \in C$, the normal cone $N_C(X) = \{g : G^T x \geq g^T y, \forall y \in C\}$ is a convex set (C not necessarily convex)

Solution. Let $g_1, g_2 \in N_C(x)$, and since normal cones are convex cones,

$$(t_1 g_1 + t_2 g_2)^T = t_1 g_1^T x + t_2 g_2^T x \geq t_1 g_1^T y + t_2 g_2^T y = (t_1 g_1 + t_2 g_2)^T y, \quad \forall t_1, t_2 \geq 0$$

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Problem 2. Prove that for any set C (convex or not), its support function $I_C^*(x) = \max_{y \in C} x^T y$ is a convex function.

Solution. For every $y \in C$, $x^T y$ is a linear function of x , so I_C^* is a pointwise supremum of a family of linear functions, therefore convex. ■

Problem 3. Consider a closed set defined by $C = \{(x, y) | y \geq \frac{1}{1+x^2}\}$ where $(x, y) \in \mathbb{R}^2$

1. Is the set C convex? (Hint: you can draw the set C on the plane to explain your answer.)
2. Is the convex hull of set C also a closed set? You can also explain your answer by the plot.

Solution.

1. Consider the function $f(x) = \frac{1}{1+x^2}$, the set $C = \text{epi}(f)$, i.e. $\forall y \in C, f(x) \leq y$
Now consider $\frac{\partial}{\partial^2 x} f(x) = \frac{6x^2-2}{(1+x^2)^3}$, which is not strictly nondecreasing, which means $f(x)$ is not convex, implying $\text{epi}(f)$ is not convex.
2. The closure of a the convex hull H of a closed set C will be H itself, making H closed as well.

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Problem 4. Adapt the proof of log-determinant function to show that $f(x) = \text{tr}(X^{-1})$ is convex on $\text{dom} f = S_{++}^n$

Solution. Let $\lambda_1, \dots, \lambda_n$ be eigenvalues of X and Y be another matrix in S_{++}^n , and its eigenvalues be μ_1, \dots, μ_n . Both X^{-1} and Y^{-1} so their eigenvalues are $\frac{1}{\lambda_1}, \dots, \frac{1}{\lambda_n}$, and $\frac{1}{\mu_1}, \dots, \frac{1}{\mu_n}$, respectively.

Consider the matrix $\theta X + (1 - \theta)Y$ for $0 \leq \theta \leq 1$, which is also positive definite because X and Y are positive definite, so and similar to above, the eigenvalues of $(\theta X + (1 - \theta)Y)^{-1}$ are $\frac{1}{\theta \lambda_i + (1 - \theta) \mu_i}$.

So

$$\text{tr}(\theta X + (1 - \theta)Y) = \sum_{i=1}^n \frac{1}{\theta \lambda_i + (1 - \theta) \mu_i},$$

and

$$\theta \text{tr}(X^{-1}) + (1 - \theta) \text{tr}(Y^{-1}) = \sum_{i=1}^n (\theta \frac{1}{\lambda_i} + (1 - \theta) \frac{1}{\mu_i})$$

And given that the function $g(x) = \frac{1}{x}$ is convex in \mathbb{R}_+ , then for some $a, b \in \mathbb{R}_+$, $0 \leq \theta \leq 1$,

$$\frac{1}{\theta a + (1 - \theta)b} \leq \theta(\frac{1}{a}) + (1 - \theta)(\frac{1}{b})$$

Applying this to the above functions, it follows that

$$\sum_{i=1}^n \frac{1}{\theta \lambda_i + (1 - \theta) \mu_i} \leq \sum_{i=1}^n (\theta \frac{1}{\lambda_i} + (1 - \theta) \frac{1}{\mu_i})$$

Problem 5. 1. Prove that $f(x_1, x_2) = \frac{1}{x_1 x_2}$ on \mathbb{R}_{++}^2 is convex.

2. Prove that $f(x_1, x_2) = \frac{x_1}{x_2}$ on $\mathbb{R} \times \mathbb{R}_{++}^2$ is convex.

Solution. ■

Problem 6. Prove that $\log(\frac{e^x}{1+e^x})$ on \mathbb{R} is concave.

Solution. ■

Problem 7. Prove that, by applying Jensen's inequality, an arithmetic mean is greater than or equal to its geometric mean, i.e., $(\prod_{i=1}^n x_i)^{1/n} \leq \frac{1}{n} \sum_{i=1}^n x_i$ where $\mathbf{x} \in \mathbb{R}_{++}^n$

Solution. ■

Problem 8. Prove that $\frac{1}{\sqrt{n}} \|x\|_1 \leq \|x\|_2 \leq \|x\|_1$

Solution. ■

Problem 9. Show that the following function $f(x)$ is convex:

$$f(x) = x^T (A(x))^{-1} x, \text{ dom } f = \{x | A(x) \succ 0\}$$

Solution. ■

Problem 10. Prove that the maximum eigenvalue of a symmetric matrix is convex.

Solution. ■

Problem 11. Let $f(x) = \sum_{i=1}^r |x|_{[i]}$ on \mathbb{R}^n , where $|x|_{[i]}$ is the i -th largest component of the absolute value of $|x|$, and $r \leq n$ is a positive number.

1. Prove that $f(x)$ is convex.
2. Show that $f(x) = \min_t(rt) + \sum_{i=1}^n \max(0, |x_i - t|)$ is convex.

Solution. ■

Problem 12. Let $f(\lambda) = \mathbf{x}^T D(\lambda)(D(\lambda) + \mathbf{I})^{-1} \mathbf{x} - \mathbf{1}^T \lambda$ on \mathbb{R}_+^n , where $D(\lambda)$ is a diagonal matrix and $\mathbf{1}^T = [1, 1, \dots, 1]$.

1. Prove that $f(\lambda)$ is concave.
2. Find the optimal value of $f(\lambda)$.

Solution. ■

Problem 13. Is the function $f(x) = \ln(1 + \frac{x^2}{2\tau_0^2})$, where $\tau_0 > 0$, convex?

Solution. ■

Problem 14. 1. Determine whether or not $\sqrt{x_1 x_2}$ is convex on \mathbb{R}_{++}^2 .

2. Determine whether or not $g(\mathbf{y}) = \max_{x \in \mathbb{R}_{++}^2} x_1 y_1 + x_2 y_2 - \sqrt{x_1 x_2}$ is convex on \mathbb{R}^2

Solution. ■