

Homework - 1 -

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Name: SI SALAH Imane

Exercise 1 which of the following sets are convex

1) A Rectangle, i.e. $S = \{x \in \mathbb{R}^n \mid \alpha_i \leq x_i \leq \beta_i, i=1, \dots, n\}$
→ this set is the intersection of a set of halfspaces characterized by the inequalities $\alpha_i \leq x_i \leq \beta_i$ which forms a polyhedron so it is convex.

2) The hyperbolic set: $S = \{x \in \mathbb{R}_+^2 \mid x_1 x_2 \geq 1\}$

→ it is convex

- we take two points (x_1, x_2) and (y_1, y_2) from S
so $x_1 x_2 \geq 1$ and $y_1 y_2 \geq 1$

- we define the linear combination of x and y as $z = \theta x + (1-\theta)y$ and we evaluate if $z \in S$ in 3 cases:

case ①: $x \geq y \Rightarrow x_1 \geq y_1$ and $x_2 \geq y_2$

$$\text{so } z = \theta x + (1-\theta)y \geq y$$

$$z_1 z_2 \geq y_1 y_2 \geq 1 \Rightarrow z \in S$$

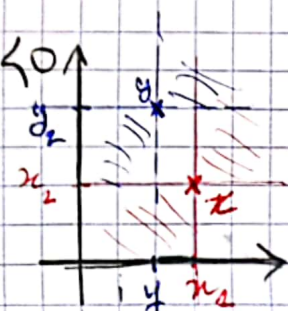
case ②: $x \leq y \Rightarrow$ similarly to case ① we get

$$z_1 z_2 \geq x_1 x_2 \geq 1 \Rightarrow z \in S$$

case ③ $x \neq y$ so $(y_1 - x_1)(y_2 - x_2) < 0$

$$\text{so } z_1 z_2 = (\theta x_1 + (1-\theta)y_1)(\theta x_2 + (1-\theta)y_2)$$

$$= \theta^2 x_1 x_2 + \theta(1-\theta)x_1 y_2 + \theta(1-\theta)x_2 y_1 + (1-\theta)^2 y_1 y_2$$



$$\begin{aligned}
 z_1 z_2 &= \theta x_1 x_2 + (1-\theta) y_1 y_2 - \theta x_1 x_2 - (1-\theta) y_1 y_2 + \theta^2 x_1 x_2 + (1-\theta)^2 y_1 y_2 \\
 &\quad + \theta(1-\theta) x_1 y_2 + \theta(1-\theta) x_2 y_1 \\
 &= \theta x_1 x_2 + (1-\theta) y_1 y_2 - \theta(1-\theta)(x_1 - y_1)(x_2 - y_2) \\
 \Rightarrow z_1 z_2 &\geq 1 \rightarrow \underline{z \in S}
 \end{aligned}$$

3) The set of points closer to a given point than a given set i.e: $\mathcal{S} = \{x \mid \|x - x_0\|_2 \leq \|x - y\|_2 \quad \forall y \in S\} \quad S \subseteq \mathbb{R}^n$

\rightarrow this set is convex

we have: the inequality $\|x - x_0\|_2 \leq \|x - y\|_2$ is equivalent to $\|x - x_0\|_2^2 \leq \|x - y\|_2^2$

$$(x - x_0)^T (x - x_0) \leq (x - y)^T (x - y)$$

$$x^T x - 2x_0^T x + x_0^T x_0 \leq x^T x - 2y^T x + y^T y$$

$$2(y^T - x_0^T)x \leq y^T y - x_0^T x_0$$

$$2(y - x_0)^T x \leq y^T y - x_0^T x_0$$

this can be written as $Ax \leq b \quad (A = 2(y - x_0)^T)$

this is a halfspace \rightarrow convex $b = y^T y - x_0^T x_0$

4) The set of points closer to one set than another

$$\{x \mid \text{dist}(x, S) \leq \text{dist}(x, T)\} \quad S, T \subseteq \mathbb{R}^n$$

\rightarrow this set is Not Convex

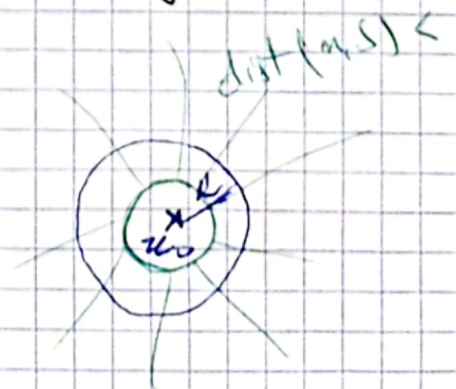
and $\text{dist}(x, S) = \inf\{\|x - z\|_2 \mid z \in S\}$

we take the example of S being a circle

$$S = \{x \mid \|x - x_0\|_2 = R, R \in \mathbb{R}\}$$

$$T = \{x_0\}$$

we clearly see that this set is Not convex.



5) The set $S = \{x \mid x + S_2 \subseteq S_1\}$ where $S_1, S_2 \subseteq \mathbb{R}^n$ (3)

the set S can be expressed as

$$S = \{x \mid x + S_2 \subseteq S_1\} = \{x \mid x + y \in S_1 \text{ for all } y \in S_2\} \\ = \{x \mid x \in S_1 - y \text{ for all } y \in S_2\}$$

this is an intersection of multiple convex sets so S is convex.

Exo 2: Determine whether the functions are convex concave or not.

(1) $f(x_1, x_2) = x_1 x_2$ on \mathbb{R}_{++}^2

we have $\nabla f = \begin{bmatrix} x_2 \\ x_1 \end{bmatrix}$ and $H = \nabla^2 f = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

$$\det(H - \lambda I) = \lambda^2 - 1$$

λ_1 and λ_2 have different signs so f is neither convex nor concave.

then we have the superlevel set defined by $\{x \mid x_1 x_2 \geq \alpha\}$ which is convex so f is quasiconcave!

(2) $f(x_1, x_2) = 1/x_1 x_2$ on \mathbb{R}_{++}^2

$$\nabla f = \begin{bmatrix} -\frac{1}{x_1^2 x_2} \\ -\frac{1}{x_1 x_2^2} \end{bmatrix} \text{ and } H = \nabla^2 f = \frac{1}{x_1 x_2} \begin{bmatrix} \frac{2}{x_1^3} & \frac{1}{x_1^2 x_2} \\ \frac{1}{x_1 x_2^2} & \frac{2}{x_2^3} \end{bmatrix}$$

$$\det(H - \lambda I) = \frac{1}{x_1 x_2} \begin{vmatrix} \frac{2}{x_1^3} - \lambda & \frac{1}{x_1^2 x_2} \\ \frac{1}{x_1 x_2^2} & \frac{2}{x_2^3} - \lambda \end{vmatrix} = \frac{1}{x_1 x_2} \left[\lambda^2 - 2 \frac{x_1^2 + x_2^2}{x_1^2 x_2^2} \lambda + \frac{3}{x_1^2 x_2^2} \right]$$

we have $\lambda_1 \lambda_2 = \frac{3}{x_1^2 x_2^2} > 0$

$\lambda_1 + \lambda_2 = 2 \frac{x_1^2 + x_2^2}{x_1^2 x_2^2} > 0$ } so $\lambda_1, \lambda_2 \in \mathbb{R}_+$
 It is positive semidefinite
 $f \rightarrow$ convex

(3) $f(x_1, x_2) = x_1/x_2$ on \mathbb{R}_{++}^2

$$\nabla f = \begin{bmatrix} \frac{1}{x_2} \\ -\frac{x_1}{x_2^2} \end{bmatrix} \quad \text{and} \quad H = \nabla^2 f = \begin{bmatrix} 0 & -\frac{1}{x_2^2} \\ -\frac{1}{x_2^2} & \frac{2x_1}{x_2^3} \end{bmatrix}$$

$$\det(\lambda I - H) = \begin{vmatrix} \lambda & \frac{1}{x_2^2} \\ \frac{1}{x_2^2} & \lambda - \frac{2x_1}{x_2^3} \end{vmatrix} = \lambda^2 - \frac{2x_1}{x_2^3} \lambda - \frac{1}{x_1^2 x_2^2}$$

$\Rightarrow -\frac{1}{x_1^2 x_2^2} < 0$ so λ_1 and λ_2 don't have the same sign.
therefore H is Not convex nor concave.

(4) $f(x_1, x_2) = x_1^\alpha x_2^{1-\alpha}$ st $0 \leq \alpha \leq 1$ on \mathbb{R}_{++}^2

$$\nabla f = \begin{bmatrix} \alpha x_1^{\alpha-1} x_2^{1-\alpha} \\ (1-\alpha) x_1^\alpha x_2^{-\alpha} \end{bmatrix}$$

$$H = \begin{bmatrix} \alpha(\alpha-1) x_1^{\alpha-2} x_2^{1-\alpha} & \alpha(1-\alpha) x_1^{\alpha-1} x_2^{-\alpha} \\ \alpha(1-\alpha) x_1^{\alpha-1} x_2^{-\alpha} & -\alpha(1-\alpha) x_1^\alpha x_2^{-\alpha-1} \end{bmatrix}$$

$$= \alpha(1-\alpha) x_1^{\alpha-1} x_2^{-\alpha} \begin{bmatrix} -x_1^{-1} & x_1^{-1} x_2^{-1} \\ x_1^{-1} x_2^{-1} & -x_2^{-1} \end{bmatrix}$$

Now $H = -\alpha(1-\alpha) x_1^{\alpha-1} x_2^{-\alpha} \begin{bmatrix} x_2^{-1} \\ x_1^{-1} \end{bmatrix} \begin{bmatrix} x_2^{-1} \\ x_1^{-1} \end{bmatrix}^T \leq 0$
so f is concave

Exo 3: show that the following functions are convex

① $f(x) = \text{tr}(x^{-1})$, $\text{dom } f = S_{++}^n$

to prove that f is convex, we prove that the composition of f with an arbitrary line $z + tv$ is convex

$$\begin{aligned} \text{let } g(t) &= f(z + tv) = \text{tr}((z + tv)^{-1}) \\ &= \text{tr}\left[\left(z^{-\frac{1}{2}}(I + tz^{-\frac{1}{2}}vz^{-\frac{1}{2}})z^{\frac{1}{2}}\right)^{-1}\right] \\ &= \text{tr}\left(z^{-1}(I + tz^{-\frac{1}{2}}vz^{-\frac{1}{2}})^{-1}\right) \end{aligned}$$

Now we introduce $z^{-\frac{1}{2}}vz^{-\frac{1}{2}} = Q\Lambda Q^T$ the eigenvalue decomposition

$$\begin{aligned} \text{so } g(t) &= \text{tr}\left(z^{-1}(I + tQ\Lambda Q^T)^{-1}\right) \\ &= \text{tr}\left(z^{-1}(Q(I + t\Lambda)Q^T)^{-1}\right) \\ &= \text{tr}\left(z^{-1}Q(I + t\Lambda)^{-1}Q^T\right) \\ &= \text{tr}\left(Q^T z^{-1}Q(I + t\Lambda)^{-1}\right) \\ &= \sum_{i=1}^n (Q^T z^{-1}Q)_{ii} (1 + t\lambda_i)^{-1} \end{aligned}$$

this is a linear combination of convex functions of the form $(1 + t\lambda_i)^{-1}$

(2) $f(x, y) = y^T x^{-1} y$ on $\text{dom } f = S_{++}^n \times \mathbb{R}^n$

we consider $z \in \mathbb{R}^n$

we will expand the following expression

$$(x^{-1}y - xz)^T (x^{-1}y - xz) \geq 0$$

$$(x^{-1}y - xz)^T (x^{-1}y - xz) \geq 0$$

$$y^T x^{-1} y + z^T x z \geq 2zy$$

$$y^T x^{-1} y \geq 2zy - z^T x z$$

$$\text{we } y^T x^{-1} y \geq \sup (2zy - z^T x z)$$

so it is convex.

(3) $f(x) = \sum_{i=1}^n \sigma_i(x)$ sum of singular values of x

we have $f(x) = 0$ iff $x = 0$ — (I)

and $f(tx) = t f(x)$ — (II)

then we set $x = U \Sigma V^T = \sum_i \sigma_i u_i v_i^T$ singular value decomposition.

set $Q = U V^T = U I V^T$ we $\sigma_1(Q) = \sigma_{\max}(Q) = 1$

then $\langle Q, x \rangle = \langle U V^T, U \Sigma V^T \rangle$

$$= \text{tr}(V U^T U \Sigma V^T)$$

$$= \text{tr}(V^T V U^T U \Sigma) = \text{tr}(\Sigma)$$

$$= \sum_{i=1}^n \sigma_i = f(x)$$

we $\sum_{i=1}^n \sigma_i \leq \sup_{\sigma_1(Q) < 1} \langle Q, x \rangle$ — (D)

on the other hand we have

$$\begin{aligned}\langle Q, X \rangle &= \text{tr}(Q^T U \Sigma V^T) = \text{tr}(V^T Q^T U \Sigma) \\ &= \langle U^T Q V, \Sigma \rangle = \sum_{i=1}^n \sigma_i (U^T Q V)_{ii} \\ &= \sum_{i=1}^n \sigma_i U_i^T Q V_i\end{aligned}$$

$$\text{Now } \sup \sum_{i=1}^n \sigma_i U_i^T Q V_i < \sup \sum \sigma_i \frac{\rho_{\max}(Q)}{1} = \sum_{i=1}^n \sigma_i$$

$$\text{So } \sup \langle Q, X \rangle \leq \sum_{i=1}^n \sigma_i \quad \text{--- (2)}$$

$$\text{from (1) and (2) } \quad \underline{\sum \sigma_i = \sup \langle Q, X \rangle}$$

$$\begin{aligned}\text{then } f(X+Y) &= \sup \langle Q, X+Y \rangle \leq \sup \langle Q, X \rangle + \sup \langle Q, Y \rangle \\ &= -f(X) + f(Y)\end{aligned}$$

$$f(X+Y) \leq -f(X) + f(Y) \quad \text{--- (III)}$$

cauchy inequality.

So f is Norm since it satisfies all three conditions of Norms \rightarrow therefore it is convex.

Ex 4: $K_{m+} = \{x \in \mathbb{R}^n \mid x_1 \geq x_2 \dots x_n \geq 0\}$

① proper cone

- we have K_{m+} is composed of the intersection of half planes delimited by the inequalities (polyhedron) so it is closed
- And, K_{m+} is non empty, and it is pointed iff $x \in K_{m+}$ then $-x \in K_{m+}$ only if $x=0$
 we have $x_1 \geq x_2 \geq \dots \geq 0$ so for $-x \in K_{m+}$ we need to have $x=0$

② The dual cone K_{m+}^*

we find the set $y \in \mathbb{R}^n$ s.t. $y^T x \geq 0$ for

$$y^T x = \sum y_i x_i \geq 0 \text{ iff } y_1, y_1+y_2, \dots, y_1+y_2+\dots+y_n \geq 0$$

$$\text{so } K_{m+}^* = \{y \mid \sum_{i=1}^k y_i \geq 0, k=1, \dots, n\}$$