

1.

Let $(x^1, y^1), (x^2, y^2)$ be two points in the epigraph.

Let $\lambda \in (0, 1)$ and let $\bar{x} = \lambda x^1 + (1-\lambda)x^2,$

$$\bar{y} = \lambda y^1 + (1-\lambda)y^2.$$

$$\text{Now, } f(\bar{x}) = c^T \bar{x} + \frac{1}{2} \bar{x}^T Q \bar{x}$$

$$= c^T \bar{x} + \frac{1}{2} (\lambda x^1 + (1-\lambda)x^2)^T Q (\lambda x^1 + (1-\lambda)x^2)$$

$$= c^T \bar{x} + \frac{1}{2} \lambda^2 x^{1T} Q x^1 + \frac{1}{2} (1-\lambda)^2 x^{2T} Q x^2$$

$$+ \lambda(1-\lambda) x^{1T} Q x^2$$

$$= c^T \bar{x} + \frac{1}{2} \lambda x^{1T} Q x^1 + \frac{1}{2} (1-\lambda) x^{2T} Q x^2$$

$$+ \frac{1}{2} (\lambda^2 - \lambda) x^{1T} Q x^1$$

$$+ \frac{1}{2} ((1-\lambda)^2 - (1-\lambda)) x^{2T} Q x^2$$

$$+ \lambda(1-\lambda) x^{1T} Q x^2$$

$$= \lambda f(x^1) + (1-\lambda) f(x^2)$$

$$+ \frac{1}{2} (\lambda^2 - \lambda) (x^{1T} Q x^1 + x^{2T} Q x^2 - 2 x^{1T} Q x^2)$$

$$= \lambda f(x^1) + (1-\lambda) f(x^2)$$

$$+ \frac{1}{2} (\lambda^2 - \lambda) (x^1 - x^2)^T Q (x^1 - x^2)$$

If Q is psd then $f(\bar{x}) \leq \lambda f(x^1) + (1-\lambda) f(x^2)$, since $\lambda^2 - \lambda < 0$.

So f is convex.

If Q is not psd then we can pick x^1, x^2 with $(x^1 - x^2)^T Q (x^1 - x^2) > 0$,

so $f(\bar{x}) > \lambda f(x^1) + (1-\lambda) f(x^2)$, so f is not convex.

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For integer x , we need $x_1 + x_2 + x_3 \leq 2$,

since $x_1 = x_2 = x_3 = 1 \Rightarrow 4x_1 + 5x_2 + 7x_3 > 13$.

This constraint is violated by \bar{x} .

3.

Let $x^1, x^2 \in \mathbb{R}^n$, let $\lambda \in (0, 1)$, and let

$$\bar{x} = \lambda x^1 + (1-\lambda)x^2.$$

Now, $f(b_i - a_i^T x) = \max \{0, b_i - a_i^T x, a_i^T x - b_i\}$

So $f(\bar{x}) = \max \{0, b_i - a_i^T \bar{x}, a_i^T \bar{x} - b_i\}$

$$= \max \left\{ 0, b_i - \lambda a_i^T x^1 - (1-\lambda)a_i^T x^2, \right. \\ \left. (1-\lambda)a_i^T x^1 + \lambda a_i^T x^2 - b_i \right\}$$

$$= \max \left\{ 0, \lambda(b_i - a_i^T x^1) + (1-\lambda)(a_i^T x^2 + b_i), \right. \\ \left. \lambda(a_i^T x^1 - b_i) + (1-\lambda)(a_i^T x^2 - b_i) \right\}$$

$$\leq \lambda \max \{0, b_i - a_i^T x^1, a_i^T x^1 - b_i\}$$

$$+ (1-\lambda) \max \{0, b_i - a_i^T x^2, a_i^T x^2 - b_i\}$$

$$= \lambda f(\bar{x}^1) + (1-\lambda)f(\bar{x}^2).$$

A sum of convex functions is convex, so $g(x)$ is convex.

In particular,

$$g(\bar{x}) = \sum_{i=1}^m f(b_i - a_i^T \bar{x})$$

$$\leq \sum_{i=1}^m (\lambda f(b_i - a_i^T x^1) + (1-\lambda)f(b_i - a_i^T x^2))$$

$$= \lambda \sum_{i=1}^m f(b_i - a_i^T x^1) + (1-\lambda) \sum_{i=1}^m f(b_i - a_i^T x^2)$$

$$= \lambda g(x^1) + (1-\lambda)g(x^2).$$

40 (a) $\frac{df}{dx} = 2x - \frac{1}{x+2}$ $\frac{d^2f}{dx^2} = 2 + \frac{1}{(x+2)^2} > 0$ for $x \geq 0$.

(b) $\frac{df}{dx} = 0$ when $2x(x+2) = 1$, or $2x^2 + 4x - 1 = 0$
 $\Rightarrow x^* = \frac{1}{2}(-4 + \sqrt{24}) = \sqrt{6} - 2$

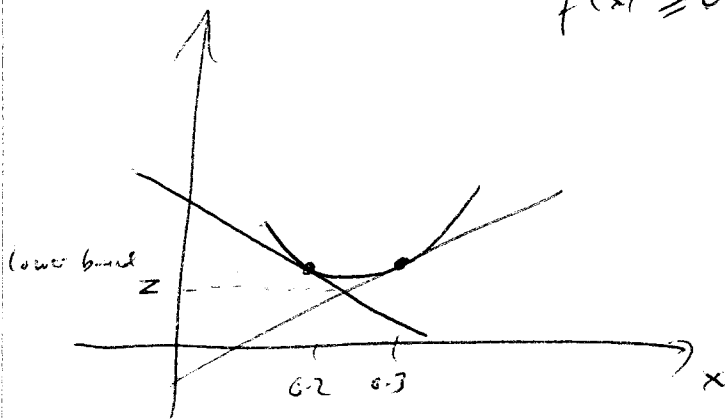
At $x = 0.2$, $\frac{df}{dx} = 0.4 - \frac{1}{2.2} = \frac{-0.12}{2.2} < 0$

At $x = 0.3$, $\frac{df}{dx} = 0.6 - \frac{1}{2.3} = \frac{0.38}{2.3} > 0$

So the minimum is between 0.2 and 0.3.

Get valid inequalities: $f(x) \geq 0.04 - \ln(2.2) + (x-0.2)\left(\frac{-0.12}{2.2}\right)$

$$f(x) \geq 0.09 - \ln(2.3) + (x-0.3)\left(\frac{0.38}{2.3}\right)$$



Find where lines cross
to get lower bound

Solve $z = 0.04 - \ln(2.2) - \frac{0.12}{2.2}(x-0.2)$ ①

$z = 0.09 - \ln(2.3) + \frac{0.38}{2.3}(x-0.3)$ ②

② - ① $\Rightarrow \left(\frac{0.38}{2.3} + \frac{0.12}{2.2}\right)x = \frac{0.3 \times 0.38}{2.3} + \frac{0.2 \times 0.12}{2.2} - 0.05 + \ln\left(\frac{2.3}{2.2}\right)$

Q4 p2

$$\frac{0.38}{2.3} \textcircled{1} + \frac{0.12}{2.2} \textcircled{2} \Rightarrow$$

$$\left(\frac{0.38}{2.3} + \frac{0.12}{2.2} \right) z = \frac{0.38}{2.3} \left(0.04 - \ln(2.2) + 0.2 \left(\frac{0.12}{2.2} \right) \right)$$

$$+ \frac{0.12}{2.2} \left(0.09 - \ln(2.3) - \frac{0.38}{2.3} (0.3) \right)$$

5.

$$h(x) = g(f(x))$$

$$\text{Let } x^1, x^2 \in \mathbb{R}^n, \lambda \in (0, 1), \bar{x} = \lambda x^1 + (1-\lambda)x^2$$

$$\text{Then } h(\bar{x}) = g(f(\bar{x}))$$

$$\leq g(\lambda f(x^1) + (1-\lambda)f(x^2))$$

since f is convex
and g is nondecreasing

$$\leq \lambda g(f(x^1)) + (1-\lambda)g(f(x^2))$$

since g is convex

$$= \lambda h(x^1) + (1-\lambda)h(x^2).$$

Thus, h is convex.

$$f(x) = x^2 - 4xy + 5y^2 \text{ is convex, since } D^2 f(x) = \begin{bmatrix} 2 & -4 \\ -4 & 10 \end{bmatrix}$$

which is positive definite.

$$g(y) = e^y \text{ is convex and nondecreasing: } \frac{d^2 g}{dy^2} = e^y > 0 \forall y.$$

So, by the first part of the question,

$$e^{x^2 - 4xy + 5y^2} \text{ is a convex function on } \mathbb{R}^2.$$

6.

Let $x^1, x^2 \in K$.

Since K is convex, need to show $\lambda x^1 + \mu x^2 \in K$

for any $\lambda, \mu \geq 0$.

$$\begin{aligned} \text{Now, } \sum_{i=1}^{n-1} (\lambda x_i^1 + \mu x_i^2)^2 - (\lambda x_n^1 + \mu x_n^2)^2 \\ = \lambda^2 \left(\sum_{i=1}^{n-1} (x_i^1)^2 - (x_n^1)^2 \right) \\ + \mu^2 \left(\sum_{i=1}^{n-1} (x_i^2)^2 - (x_n^2)^2 \right) \\ + \lambda \mu \left(\sum_{i=1}^{n-1} x_i^1 x_i^2 - x_n^1 x_n^2 \right) \end{aligned}$$

Now, let $\bar{x}^1 = (x_1^1, \dots, x_{n-1}^1) \in \mathbb{R}^{n-1}$

$\bar{x}^2 = (x_1^2, \dots, x_{n-1}^2) \in \mathbb{R}^{n-1}$

We have $x_n^1 \geq \|\bar{x}^1\|$ and $x_n^2 \geq \|\bar{x}^2\|$,

$$\text{so } x_n^1 x_n^2 \geq \bar{x}^1 \cdot \bar{x}^2$$

$$\text{Thus } \sum_{i=1}^{n-1} (\lambda x_i^1 + \mu x_i^2)^2 - (\lambda x_n^1 + \mu x_n^2)^2 \leq 0,$$

so we have a convex cone.

Let $V = \{s \in \mathbb{R}^n; \sum_{i=1}^{n-1} s_i^2 \leq s_n^2, s_n \leq 0\}$.

Claim

$$K^0 = V.$$

Proof

~~Let~~

For any $x \in \mathbb{R}^n$, let $\bar{x} = (x_1, \dots, x_{n-1}) \in \mathbb{R}^{n-1}$.

Define \bar{s} similarly.

Show $K^0 \subseteq V$:

Assume $s \notin V$, show $s \notin K^0$.

$$\begin{aligned} \text{Define } x_i &= s_i, \quad i=1, \dots, n-1 \\ x_n &= \|\bar{x}\|. \end{aligned}$$

$$\begin{aligned} \text{Then } s^T x &= \bar{s}^T \bar{x} + s_n \|\bar{x}\| = \|\bar{s}\|^2 + \|\bar{s}\| s_n \\ &= \|\bar{s}\| (\|\bar{s}\| + s_n) \end{aligned}$$

$$\text{So } s_n \leq -\|\bar{s}\| \text{ for } s \in K^0.$$

But since $s \notin V$, must have $s_n > -\|\bar{s}\|$. So $K^0 \subseteq V$.

Show $V \subseteq K^0$:

Let $s \in V$

$$\begin{aligned} \text{Then } s^T x &= \bar{s}^T \bar{x} + s_n x_n \leq 0 \quad \text{since } s_n \leq -\|\bar{s}\|, \\ & \quad x_n \geq \|\bar{x}\|. \end{aligned}$$

$$\text{So } s \in K^0.$$

