

Write your name: _____

Write your student number: _____

Exam

Write your answers (A, B, C, D, E, or F) to problems 1 to 3 in this box

Your answer to problem 1: _____

Your answer to problem 2: _____

Your answer to problem 3: _____

1. *Simple convex function.* (3 points) One of the following six functions $\mathbf{R} \rightarrow \mathbf{R}$ is convex:

- (A) $(1 - (x - 1)_+)_+$
- (B) $|(x - 1)_+ - 1|$
- (C) $-(1 - (x - 1)_+)_+$
- (D) $((x - 1)_+ - 1)_+$
- (E) $-((x - 1)_+ - 1)_+$
- (F) $-|(x - 1)_+ - 1|$

Which one?

Write your answer (A, B, C, D, E, or F) in the box at the top of page 1

2. *Least-squares.* (2 points) Consider the following six optimization problems:

- (A)
$$\underset{x \in \mathbf{R}^n}{\text{minimize}} \quad \|Ax - b\|_1^2 + \rho \|x\|_2^2$$
- (B)
$$\underset{x \in \mathbf{R}^n}{\text{minimize}} \quad \|Ax - b\|_2^2 + \rho \|x\|_2^2$$
- (C)
$$\underset{x \in \mathbf{R}^n}{\text{minimize}} \quad \|Ax - b\|_2^2 + \rho \|x\|_1^2$$
- (D)
$$\underset{x \in \mathbf{R}^n}{\text{minimize}} \quad \|Ax - b\|_2 + \rho \|x\|_2^2$$

(E)

$$\underset{x \in \mathbf{R}^n}{\text{minimize}} \quad \|Ax - b\|_1^2 + \rho \|x\|_1^2$$

(F)

$$\underset{x \in \mathbf{R}^n}{\text{minimize}} \quad \|Ax - b\|_1 + \rho \|x\|_2$$

In each of the six problems above, the variable to optimize is $x \in \mathbf{R}^n$. The matrix A and the vector b are given. The scalar $\rho > 0$ is also given.

One of the optimization problems above is a least-squares problem.

Which one?

Write your answer (A, B, C, D, E, or F) in the box at the top of page 1

3. *Convex function.* (3 points) Let $f: \mathbf{R}^n \rightarrow \mathbf{R}$ be a convex function. One of the following functions is guaranteed to be convex:

(A) $|f(x)|$

(B) $f(x) + (f(x))^2$

(C) $(f(x))^2$

(D) $f(x)(f(x))^2$

(E) $|f(x)| + (f(x))^2$

(F) $f(x) + |f(x)|$

Which one?

Write your answer (A, B, C, D, E, or F) in the box at the top of page 1

4. *Robust portfolio selection.* (4 points) A problem that often occurs in finance has the following form

$$\begin{aligned} &\underset{x \in \mathbf{R}^n}{\text{minimize}} && \max\{x^T V_1^T D V_1 x, x^T V_2^T D V_2 x\} \\ &\text{subject to} && \min\{\mu_1^T x, \mu_2^T x\} \geq \alpha \\ &&& \mathbf{1}^T x = 1, \end{aligned} \tag{1}$$

where the variable to optimize is $x \in \mathbf{R}^n$.

The matrices $V_1 \in \mathbf{R}^{p \times n}$, $V_2 \in \mathbf{R}^{p \times n}$, and $D \in \mathbf{R}^{p \times p}$ are given, the matrix D being diagonal with positive entries in the diagonal:

$$D = \begin{bmatrix} d_1 & & & \\ & d_2 & & \\ & & \ddots & \\ & & & d_p \end{bmatrix},$$

with $d_i > 0$ for $i = 1, \dots, p$.

The vectors $\mu_1 \in \mathbf{R}^n$, $\mu_2 \in \mathbf{R}^n$ and the scalar $\alpha \in \mathbf{R}$ are given. Finally, recall that the symbol $\mathbf{1}$ stands for the vector of dimension n with all components equal to one:

$$\mathbf{1} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}.$$

Show that the optimization problem (1) is convex.

5. *Mahalanobis projection.* (4 points) Consider the optimization problem

$$\begin{aligned} & \underset{x}{\text{minimize}} && (x - \mu)^T \Sigma^{-1} (x - \mu) \\ & \text{subject to} && a^T x = b, \end{aligned} \tag{2}$$

where the variable to optimize is $x \in \mathbf{R}^n$. The vector $\mu \in \mathbf{R}^n$ and the matrix $\Sigma \in \mathbf{R}^{n \times n}$ are given, with Σ being symmetric and positive definite.

Show that the optimal value of problem (2) is

$$\frac{(a^T \mu - b)^2}{a^T \Sigma a}.$$

6. *Strictly convex functions.* (4 points) Suppose that the functions $f_1: \mathbf{R}^n \rightarrow \mathbf{R}$ and $f_2: \mathbf{R}^n \rightarrow \mathbf{R}$ are both convex, and let $f: \mathbf{R}^n \rightarrow \mathbf{R}$ be defined as $f(x) = \max\{f_1(x), f_2(x)\}$ for each $x \in \mathbf{R}^n$. Is the function f strictly convex? If you think the answer is ‘yes’, then prove it; if you think the answer is ‘no’, then give a counter-example.