

MAT3

MATHEMATICAL TRIPOS **Part III**

Friday, 9 June, 2023 9:00 am to 11:00 am

PAPER 339

TOPICS IN CONVEX OPTIMISATION

Before you begin please read these instructions carefully

Candidates have **TWO HOURS** to complete the written examination.

Attempt **BOTH** questions.
There are **TWO** questions in total.
The questions carry equal weight.

STATIONERY REQUIREMENTS

Cover sheet
Treasury tag
Script paper
Rough paper

SPECIAL REQUIREMENTS

None

<p>You may not start to read the questions printed on the subsequent pages until instructed to do so by the Invigilator.</p>

1 Let $a_1, \dots, a_m \in \mathbb{R}^n$, $b \in \mathbb{R}^m$, and consider the function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ defined by

$$f(x) = \max_{i=1, \dots, m} (\langle a_i, x \rangle + b_i).$$

(a) Show that f is convex. State the definition of *subgradient*. For $x \in \mathbb{R}^n$ give a subgradient of f at x and justify your answer.

(b) Is the function f Lipschitz (with respect to the Euclidean norm)? Justify your answer, and if yes, give an upper bound on the Lipschitz constant. Write down the subgradient method for minimizing f . State, without proof, an upper bound of the form $O(\epsilon^{-p})$ (where $p > 0$) on the number of iterations needed to find a point x such that $f(x) - \min f \leq \epsilon$.

For $\beta > 0$, let

$$f_\beta(x) = \beta^{-1} \log \sum_{i=1}^m \exp(\beta(\langle a_i, x \rangle + b_i)).$$

(c) Show that f_β is convex, and that for all x ,

$$f(x) \leq f_\beta(x) \leq f(x) + \beta^{-1} \log m.$$

(d) Give an expression for the gradient of f_β . Show that f_β is L -smooth with respect to the Euclidean norm for some constant $L > 0$ that you should specify.

(e) Explain how, using Nesterov's accelerated gradient, one can compute x such that $f(x) - \min f \leq \epsilon$ in at most $O(\epsilon^{-1})$ iterations.

2 A nonlinear map $F : \mathbb{R}^N \rightarrow \mathbb{R}^N$ is *monotone* with respect to an inner product $\langle \cdot, \cdot \rangle$ if $\langle F(v) - F(w), v - w \rangle \geq 0$ for all $v, w \in \mathbb{R}^N$.

(a) Show that if $f : \mathbb{R}^N \rightarrow \mathbb{R}$ is convex and differentiable, then $F(w) = \nabla f(w)$ is a monotone map.

(b) Let $M : \mathbb{R}^N \rightarrow \mathbb{R}^N$ be a linear map which is self-adjoint with respect to $\langle \cdot, \cdot \rangle$ and positive definite. Define the (nonlinear) map $T : \mathbb{R}^N \rightarrow \mathbb{R}^N$ for each $w \in \mathbb{R}^N$ via the implicit equation

$$M(T(w)) + F(T(w)) = M(w) \quad (1)$$

which we assume admits a unique solution $T(w)$ for all $w \in \mathbb{R}^N$, where $F : \mathbb{R}^N \rightarrow \mathbb{R}^N$ is a generic monotone map with respect to $\langle \cdot, \cdot \rangle$. In short, we write $T(w) = (M + F)^{-1}(M(w))$. Verify that the fixed points of T are zeros of F , i.e., $T(w) = w \implies F(w) = 0$. After recalling the definition of a *firmly nonexpansive map*, show that the map T is firmly nonexpansive with respect to the inner product $\langle \cdot, \cdot \rangle_M$ defined by $\langle v, w \rangle_M = \langle v, M(w) \rangle = \langle M(v), w \rangle$.

Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be a convex differentiable function, and consider the problem

$$\min_{x \in \mathbb{R}^n} f(x) \quad \text{subject to} \quad Ax = b \quad (2)$$

where $A \in \mathbb{R}^{m \times n}$ and $b \in \mathbb{R}^m$.

(c) Write down the Lagrangian and the dual optimization problem. Show that (x, z) is a pair of primal-dual optimal points if, and only if, $F \begin{pmatrix} x \\ z \end{pmatrix} = 0$, where $F : \mathbb{R}^{n+m} \rightarrow \mathbb{R}^{n+m}$ is the nonlinear map defined by

$$F \begin{pmatrix} x \\ z \end{pmatrix} = \begin{pmatrix} \nabla f(x) - A^T z \\ Ax - b \end{pmatrix}.$$

Show that the map F is monotone with respect to the Euclidean inner product.

(d) Consider a symmetric matrix M of size $n + m$ of the form

$$M = \begin{bmatrix} \alpha I & A^T \\ A & \beta I \end{bmatrix}. \quad (3)$$

where α, β are scalars, and I represents an identity matrix of suitable size. Show that if $\alpha\beta > \|A\|_2^2$, then M is positive definite. [Here $\|A\|_2$ is the operator norm of A .]

(e) Let $T : \mathbb{R}^{n+m} \rightarrow \mathbb{R}^{n+m}$ be the nonlinear map defined as in (1), i.e., $T(w) = (M + F)^{-1}(M(w))$, with the linear map $M : \mathbb{R}^{n+m} \rightarrow \mathbb{R}^{n+m}$ given in (3). Consider the fixed point iterations for T , namely

$$w_{k+1} = T(w_k).$$

By writing $w_k = \begin{pmatrix} x_k \\ z_k \end{pmatrix} \in \mathbb{R}^{n+m}$, simplify these iterations, and show that they can be expressed solely in terms of the *proximal operator* of $\alpha^{-1}f$ and the linear map A .

END OF PAPER