

Homework 2

Jordan Saethre

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1. Let $a \in \mathbb{R}^n$ be a nonzero vector. Show that the maximum of $a^t x$ over $B[0, 1] = \{x \in \mathbb{R}^n : \|x\| \leq 1\}$ is attained at $x = \frac{1}{\|a\|}a$ and that the maximal value is $\|a\|$.

Proof. First note that $a^t x = \langle a, x \rangle$. By the Cauchy-Schwarz inequality we have $\langle a, x \rangle \leq \|a\| \cdot \|x\|$. Since $\|x\| \leq 1$ this implies that $\langle a, x \rangle \leq \|a\|$. Therefore the maximal value of $a^t x$ over $B[0, 1]$ is bounded above by $\|a\|$. It is easy to see that this upper bound is attained at $x = \frac{1}{\|a\|}a$ since $\|a\| \cdot \|x\| = \|a\| \cdot \|\frac{1}{\|a\|}a\| = \|a\|$.

□

2. Let $B \in \mathbb{R}^{n \times k}$ and let $A = BB^t$.

(a) Prove that A is positive semidefinite.

Proof. A symmetric matrix $A \in \mathbb{R}^{n \times n}$ is positive semidefinite if $x^t A x \geq 0$ for all $x \in \mathbb{R}^n$. Now consider $A = BB^t$

$$x^t BB^t x = (B^t x)^t (B^t x) = \|B^t x\|^2 \geq 0$$

□

(b) Prove that A is positive definite if and only if B has a full row rank.

Proof. For A to be positive definite it must be that $\|B^t x\|^2 > 0$ for all real non-zero vectors x . Suppose B has full row rank and that $\|B^t x\|^2 = 0$ where $x \neq 0$. The matrix vector product inside the norm is a linear combination of the rows of B and since it is assumed that $x \neq 0$ this implies that at least one of the rows of B is linearly dependent on the rest which means that B cannot have full row rank. Contradiction! Therefore if B has full row rank it must be that $\|B^t x\|^2 > 0$, i.e. A is positive definite.

Going the other way assume that $\|B^t x\|^2 > 0$ (A is positive definite) and that B doesn't have full row rank. This means that the rows of B are linearly dependent and that $\|B^t x\|^2 = 0$ for some nonzero x . This implies that A is not positive definite, a contradiction. Therefore if A is positive definite then B must have full row rank.

□

3. Let A^α be the $n \times n$ matrix ($n > 1$) defined by

$$A_{i,j}^\alpha = \begin{cases} \alpha, & i = j \\ 1, & i \neq j \end{cases}$$

Show that A^α is positive semidefinite if and only if $\alpha \geq 1$.

Proof. Rewrite A^α as $ee^t + (\alpha - 1)I$ where e is a vector comprised of all 1's and I is the $n \times n$ identity matrix. Then

$$x^t A^\alpha x = x^t (ee^t + (\alpha - 1)I)x = (e^t x)^2 + (\alpha - 1)||x||^2$$

For A^α to be positive semidefinite this must hold for all x . Consider what happens when $x = [1, -1, 0, \dots, 0]^t$. The first term in this final expression would vanish leaving us with $(\alpha - 1)||x||^2$ which is non-negative as long as $\alpha \geq 1$.

Going the other direction assume $\alpha \geq 1$ then $(e^t x)^2 + (\alpha - 1)||x||^2$ is a positive number plus a positive number, therefore A^α is positive semidefinite. \square

4. For each of the following functions, find all the stationary points and classify them according to whether they are saddle points, strict/nonstrict local/global minimum/-maximum points:

(a) $f(x_1, x_2) = 2x_2^3 - 6x_2^2 + 3x_1^2x_2$

If $\nabla f(x^*) = 0$ then x^* is a stationary point.

$$\nabla f = \begin{bmatrix} f_{x_1} \\ f_{x_2} \end{bmatrix} = \begin{bmatrix} 6x_1x_2 \\ 6x_2^2 - 12x_2 + 3x_1^2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

From $6x_1x_2 = 0$ either $x_1 = 0$, $x_2 = 0$, or both. This gives $(0, 0)$ as a stationary point. Substituting $x_2 = 0$ into the second equation we get

$$6x_2^2 - 12x_2 + 3x_1^2 = 6(0)^2 - 12(0) + 3x_1^2 = 0 \implies x_2 = 0$$

Substituting $x_1 = 0$ into the second equation we get

$$\begin{aligned} 6x_2^2 - 12x_2 + 3x_1^2 &= 6x_2^2 - 12x_2 + 3(0)^2 \\ &= 6x_2(x_2 - 2) = 0 \\ &\implies x_2 = 2 \end{aligned}$$

which gives a second stationary point $(0, 2)$.

To classify each of these points we consider the matrix

$$\begin{aligned}\nabla^2 f(x_1, x_2) &= \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} \end{bmatrix} \\ &= 6 \begin{bmatrix} x_2 & x_1 \\ x_1 & 6x_2 - 2 \end{bmatrix}\end{aligned}$$

$$\nabla^2 f(0, 0) = 6 \begin{bmatrix} 0 & 0 \\ 0 & -2 \end{bmatrix}$$

$$y^t \nabla^2 f(0, 0) y = -12y_2^2 < 0 \quad \forall y \in \mathbb{R}^2$$

y can be chosen such that $\|y\| \neq 0$ but $-12y_2^2 = 0$ which proves that this matrix is *not negative definite*. Therefore this point cannot be a strict local maximum but could be a non-strict local maximum or a saddle point. To decide we look at points in the neighborhood of $(0, 0)$. Consider the following

$$f(\alpha, \alpha) = 2\alpha^3 - 6\alpha^2 + 3\alpha^3 = \alpha^2(5\alpha - 6)$$

If α is small enough this value is negative.

$$f(-\alpha, \alpha) = 2\alpha^3 - 6\alpha^2 + 3\alpha^2(-\alpha) = -\alpha^2(\alpha - 6)$$

If α is small enough here the value is positive. Therefore $(0, 0)$ is a saddle point.

$$\nabla^2 f(0, 2) = 6 \begin{bmatrix} 2 & 0 \\ 0 & 10 \end{bmatrix}$$

The trace of this matrix and its determinant are both strictly positive therefore $\nabla^2 f(0, 2) \succ 0$ which implies this point is a strict local minimum.

(b) $f(x_1, x_2) = x_1^4 + 2x_1^2x_2 + x_2^2 - 4x_1^2 - 8x_1 - 8x_2$

Solving the system of equations $\nabla f(x_1, x_2) = 0$ yields one stationary point $(1, 3)$.

$$\begin{aligned}\nabla f(x_1, x_2) &= \begin{bmatrix} 4x_1^3 + 4x_1x_2 - 8x_1 - 8 \\ 2x_1^2 + 2x_2 - 8 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ &\implies (x_1, x_2) = (1, 3)\end{aligned}$$

$$\nabla^2 f(x_1, x_2) = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 x_2} \\ \frac{\partial^2 f}{\partial x_2 x_1} & \frac{\partial^2 f}{\partial x_2^2} \end{bmatrix}$$

$$= 2 \begin{bmatrix} 6x_1 + 2x_2 - 4 & 2x_1 \\ 2x_1 & 1 \end{bmatrix}$$

$$\nabla^2 f(1, 3) = 2 \begin{bmatrix} 8 & 2 \\ 2 & 1 \end{bmatrix}$$

The trace and determinant of this matrix are both strictly positive therefore $\nabla^2 f(1, 3) \succ 0$ which implies $(1, 3)$ is a strict local minimum.