Optimization 1 — Homework 3

November 5, 2020

Problem 1

The principal minors criterion states that a symmetric matrix \mathbf{A} is PD if and only if all the leading principal minors of \mathbf{A} are PD. The purpose of this problem is to give a proof of this result using optimization techniques.

Let $\mathbf{A} \in \mathbb{R}^{(n+1)\times(n+1)}$ and $\mathbf{B} \in \mathbb{R}^{n\times n}$ be two symmetric matrices such that

$$\mathbf{A} = \begin{bmatrix} \mathbf{B} & \mathbf{b} \\ \mathbf{b}^T & c \end{bmatrix},$$

where $\mathbf{B} \succ 0$, $\mathbf{b} \in \mathbb{R}^n$ and $c \in \mathbb{R}$.

(a) Consider the function $p: \mathbb{R}^n \to \mathbb{R}$ defined by

$$p\left(\mathbf{x}\right) = \begin{pmatrix} \mathbf{x}^T & 1 \end{pmatrix} \mathbf{A} \begin{pmatrix} \mathbf{x} \\ 1 \end{pmatrix}.$$

Show that $\mathbf{x}^* = -\mathbf{B}^{-1}\mathbf{b}$ is the unique global minimizer of p over \mathbb{R}^n and that p is positive over its domain if and only if $c > \mathbf{b}^T \mathbf{B}^{-1} \mathbf{b}$.

(b) Show that $\det(\mathbf{A}) = \det(\mathbf{B}) \cdot (c - \mathbf{b}^T \mathbf{B}^{-1} \mathbf{b})$.

Hint: Find a vector $\mathbf{d} \in \mathbb{R}^n$ such that

$$\begin{bmatrix} \mathbf{I} & \mathbf{d} \\ \mathbf{0} & 1 \end{bmatrix}^T \mathbf{A} \begin{bmatrix} \mathbf{I} & \mathbf{d} \\ \mathbf{0} & 1 \end{bmatrix} = \begin{bmatrix} \mathbf{B} & \mathbf{0} \\ \mathbf{0} & c - \mathbf{b}^T \mathbf{B}^{-1} \mathbf{b} \end{bmatrix}.$$

(c) Prove the principal minors criterion by induction on the dimension of A.

Problem 2

Assume that each point in the set $\mathcal{X} = \{(x_i, y_i)\}_{i=1}^m \subset \mathbb{R}^2$ can be approximated by the relation

$$y = \frac{c_0 + c_1 x + c_2 x^2}{d_0 + d_1 x + d_2 x^2},$$

where $\mathbf{u} = (c_0, c_1, c_2, d_0, d_1, d_2)^T \in \mathbb{R}^6$ is an unknown vector of coefficients. Since for all scalar $\alpha \neq 0$ the coefficients \mathbf{u} and $\alpha \mathbf{u}$ represent the same curve, the above relation has redundancy (meaning, there are infinitely many curves).

- (a) One approach to tackle the redundancy is to assume that $d_0 = 1$. Write a least squares problem that finds an approximating vector \mathbf{u} .
- (b) Write a MATLAB function that is called by

$$[c0,c1,c2,d1,d2] = fit rational(X)$$

where X is a matrix with m rows and two columns that contains the set \mathcal{X} . The function solves your least squares problem from section (a). Use the MATLAB operator \setminus . Run the function on the file .curve that is in the moodle site. Write the output coefficients and plot the resulting curve with the given points.

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- (c) Another approach to tackle the redundancy is to assume that $\|(c_0, c_1, c_2, d_0, d_1, d_2)^T\| = 1$. Write an optimization problem with a quadratic objective function that finds an approximating vector \mathbf{u} . Explain how you would solve this problem (you can use the Rayleigh quotient as seen in class in your explanation).
- (d) Write a MATLAB function that is called by

that solves your optimization problem from section (c). Which approach yields a better approximation?

Problem 3

For each of the following functions, determine whether it is coercive or not:

(a)
$$f(x,y) = (x-2y)^4 + 64xy$$
.

(b)
$$f(x,y) = \begin{cases} x^2 - 2xy + y^2, & x \neq y, \\ x^2 + y^2, & x = y. \end{cases}$$

(c)
$$f(\mathbf{x}) = \frac{\mathbf{x}^T \mathbf{A} \mathbf{x}}{\|\mathbf{x}\| + 1}$$
 where $\mathbf{A} \in \mathbb{R}^{n \times n}$ is PD.

Problem 4

Let $f: \mathbb{R}^n \to \mathbb{R}$ be defined as $f(\mathbf{x}) = \mathbf{x}^T \mathbf{A} \mathbf{x} + 2 \mathbf{b}^T \mathbf{x} + c$, where $\mathbf{A} \in \mathbb{R}^{n \times n}$ is symmetric, $\mathbf{b} \in \mathbb{R}^n$ and $c \in \mathbb{R}$. Suppose that $\mathbf{A} \succeq 0$. Show that f is bounded from below over \mathbb{R}^n if and only if $\mathbf{b} \in \text{Image}(\mathbf{A})$.

Problem 5

Let $\mathbf{A} \in \mathbb{R}^{m \times n}$, $\mathbf{b} \in \mathbb{R}^m$, $\mathbf{L} \in \mathbb{R}^{p \times n}$ and $\lambda \in \mathbb{R}_{++}$. Consider the function

$$f(\mathbf{x}) = \|\mathbf{A}\mathbf{x} - \mathbf{b}\|^2 + \lambda \|\mathbf{L}\mathbf{x}\|^2.$$

Show that f has a unique minimum if and only if $Ker(\mathbf{A}) \cap Ker(\mathbf{L}) = \{\mathbf{0}_n\}$.