

ECE/MATH 520, Spring 2008
Exam 2: Due Session 26

Name: _____

75 mins.; Total 50 pts.

1. (16 pts.) The purpose of this question is to derive a recursive least-squares algorithm where we *remove* (instead of add) a data point. To formulate the algorithm, suppose we are given matrices \mathbf{A}_0 and \mathbf{A}_1 such that

$$\mathbf{A}_0 = \begin{bmatrix} \mathbf{A}_1 \\ \mathbf{a}_1^T \end{bmatrix},$$

where $\mathbf{a}_1 \in \mathbb{R}^n$. Similarly, suppose vectors $\mathbf{b}^{(0)}$ and $\mathbf{b}^{(1)}$ satisfy

$$\mathbf{b}^{(0)} = \begin{bmatrix} \mathbf{b}^{(1)} \\ b_1 \end{bmatrix},$$

where $b_1 \in \mathbb{R}$. Let $\mathbf{x}^{(0)}$ be the least-squares solution associated with $(\mathbf{A}_0, \mathbf{b}^{(0)})$, and $\mathbf{x}^{(1)}$ the least-squares solution associated with $(\mathbf{A}_1, \mathbf{b}^{(1)})$. Our goal is to write $\mathbf{x}^{(1)}$ in terms of $\mathbf{x}^{(0)}$ and the “removed” data point (\mathbf{a}_1, b_1) . As usual, let \mathbf{G}_0 and \mathbf{G}_1 be the Grammians associated with $\mathbf{x}^{(0)}$ and $\mathbf{x}^{(1)}$, respectively.

- a. Write down expressions for the least-squares solutions $\mathbf{x}^{(0)}$ and $\mathbf{x}^{(1)}$ in terms of \mathbf{A}_0 , $\mathbf{b}^{(0)}$, \mathbf{A}_1 , and $\mathbf{b}^{(1)}$.
- b. Derive a formula for \mathbf{G}_1 in terms of \mathbf{G}_0 and \mathbf{a}_1 .
- c. Let $\mathbf{P}_0 = \mathbf{G}_0^{-1}$ and $\mathbf{P}_1 = \mathbf{G}_1^{-1}$. Derive a formula for \mathbf{P}_1 in terms of \mathbf{P}_0 and \mathbf{a}_1 . (The formula must not contain any matrix inversions.)
- d. Derive a formula for $\mathbf{A}_0^T \mathbf{b}^{(0)}$ in terms of \mathbf{G}_1 , $\mathbf{x}^{(0)}$, and \mathbf{a}_1 .
- e. Finally, derive a formula for $\mathbf{x}^{(1)}$ in terms of $\mathbf{x}^{(0)}$, \mathbf{P}_1 , \mathbf{a}_1 , and b_1 . Use this and part c to write a recursive algorithm associated with successive removals of rows from $(\mathbf{A}_k, \mathbf{b}^{(k)})$.

2. (10 pts.) Use the penalty method to solve the following problem analytically:

$$\begin{array}{ll}\text{minimize} & x_1^2 + 2x_2^2 \\ \text{subject to} & x_1 + x_2 = 3.\end{array}$$

Hint: Use the penalty function $P(x) = (x_1 + x_2 - 3)^2$. The solution you find must be exact, not approximate.

3. (12 pts.) Consider a standard form LP problem. Suppose we start with an initial basic feasible solution $\mathbf{x}^{(0)}$ and we apply one iteration of the simplex algorithm to obtain $\mathbf{x}^{(1)}$.

As pointed out in class, it turns out that we can express $\mathbf{x}^{(1)}$ in terms of $\mathbf{x}^{(0)}$ as

$$\mathbf{x}^{(1)} = \mathbf{x}^{(0)} + \alpha_0 \mathbf{d}^{(0)},$$

where α_0 minimizes $\phi(\alpha) = f(\mathbf{x}^{(0)} + \alpha \mathbf{d}^{(0)})$ over all $\alpha > 0$ such that $\mathbf{x}^{(0)} + \alpha \mathbf{d}^{(0)}$ is feasible.

- a. Show that $\mathbf{d}^{(0)} \in \mathcal{N}(\mathbf{A})$.
- b. As usual, assume that the initial basis is the first m columns of \mathbf{A} , and the first iteration involves inserting \mathbf{a}_q into the basis, where $q > m$. Let the q th column of the canonical augmented matrix be $\mathbf{y}_q = [y_{1q}, \dots, y_{mq}]^T$.

Express $\mathbf{d}^{(0)}$ in terms of \mathbf{y}_q .

- c. Show that $\mathbf{d}^{(0)}$ is a descent direction if and only if $r_q < 0$.

4. (12 pts.) Suppose we are given a matrix $\mathbf{A} \in \mathbb{R}^{m \times n}$ and a vector $\mathbf{b} \in \mathbb{R}^m$ such that $\mathbf{b} \geq \mathbf{0}$. We are interested in an algorithm that, given this \mathbf{A} and \mathbf{b} , is guaranteed to produce one of following two outputs: (1) If there exists \mathbf{x} such that $\mathbf{Ax} \geq \mathbf{b}$, then the algorithm produces one such \mathbf{x} . (2) If no such \mathbf{x} exists, then the algorithm produces an output to declare so.

Describe in detail how to design this algorithm based on the simplex method.