

EE/M 520, Spring 2007

Exam 2: Due April 19 (9:30am at the ECE front office)

Solutions (version: April 19, 2007, 16:7)

Total 50 pts.

1. (8 pts.) Suppose you are given two different types of concrete. The first type contains 30% cement, 40% gravel, and 30% sand (all percentages of weight). The second type contains 10% cement, 20% gravel, and 70% sand.

How many pounds of each type of concrete should you mix together so that you get a concrete mixture that has as close as possible to a total of 5 pounds of cement, 3 pounds of gravel, and 4 pounds of sand? Formulate and solve the problem using a linear least-squares method.

Ans.: The problem can be formulated as a least-squares problem with

$$\mathbf{A} = \begin{bmatrix} 0.3 & 0.1 \\ 0.4 & 0.2 \\ 0.3 & 0.7 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 5 \\ 3 \\ 4 \end{bmatrix},$$

where the decision variable is $\mathbf{x} = [x_1, x_2]^T$, and x_1 and x_2 are the amounts of concrete of the first and second types, respectively. After some algebra, we obtain the solution:

$$\mathbf{x}^* = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b} = \frac{1}{(0.34)(0.54) - (0.32)^2} \begin{bmatrix} 0.54 & -0.32 \\ -0.32 & 0.34 \end{bmatrix} \begin{bmatrix} 3.9 \\ 3.9 \end{bmatrix} = \begin{bmatrix} 10.6 \\ 0.961 \end{bmatrix}.$$

2. (16 pts.) Consider the problem

$$\begin{aligned} &\text{minimize} && \frac{1}{2} \mathbf{x}^T \mathbf{Q} \mathbf{x} \\ &\text{subject to} && \|\mathbf{x}\|^2 = 1, \end{aligned}$$

where $\mathbf{Q} = \mathbf{Q}^T > 0$. Suppose we apply a fixed step size projected gradient algorithm to this problem.

a. Derive a formula for the update equation for the algorithm (i.e., write down an explicit formula for $\mathbf{x}^{(k+1)}$ as a function of $\mathbf{x}^{(k)}$, \mathbf{Q} , and the fixed step size α). You may assume that the argument in the projection operator to obtain $\mathbf{x}^{(k)}$ is never zero.

b. Is it possible for the algorithm not to converge to an optimal solution, even if the step size $\alpha > 0$ is taken to be sufficiently small? Explain fully.

Hint: Any solution to this optimization problem is an eigenvector of \mathbf{Q} with smallest eigenvalue.

c. Show that for $0 < \alpha < 1/\lambda_{\max}$ (where λ_{\max} is the largest eigenvalue of \mathbf{Q}), the fixed step size projected gradient algorithm (with step size α) converges to an optimal solution, provided $\mathbf{x}^{(0)}$ is not orthogonal to the eigenvectors of \mathbf{Q} corresponding to the smallest eigenvalue.

Hint: Let $\mathbf{v}_1, \dots, \mathbf{v}_n$ be orthonormal eigenvectors of \mathbf{Q} ordered according to ascending eigenvalues (which you may assume to be distinct). Note that \mathbf{v}_1 is the optimal solution. Then write $\mathbf{x}^{(k)} = y_1^{(k)} \mathbf{v}_1 + \dots + y_n^{(k)} \mathbf{v}_n$, and assume that $y_1^{(0)} \neq 0$.

Ans.: a. The projection operator in this case simply maps any vector to the closest point on the unit circle. Therefore, the projection operator is given by $\Pi[\mathbf{x}] = \mathbf{x}/\|\mathbf{x}\|$, provided $\mathbf{x} \neq \mathbf{0}$. The update equation is

$$\mathbf{x}^{(k+1)} = \beta_k(\mathbf{x}^{(k)} - \alpha \mathbf{Q} \mathbf{x}^{(k)}) = \beta_k(\mathbf{I} - \alpha \mathbf{Q}) \mathbf{x}^{(k)},$$

where $\beta_k = 1/\|(\mathbf{I} - \alpha \mathbf{Q}) \mathbf{x}^{(k)}\|$ (i.e., it is whatever constant scaling is needed to make $\mathbf{x}^{(k+1)}$ have unit norm).

b. If we start with $\mathbf{x}^{(0)}$ being an eigenvector of \mathbf{Q} , then $\mathbf{x}^{(k)} = \mathbf{x}^{(0)}$ for all k . Therefore, if the corresponding eigenvalue is not the smallest, then clearly the algorithm is stuck at a point that is not optimal.

c. We have

$$\begin{aligned} \mathbf{x}^{(k+1)} &= \beta_k(\mathbf{I} - \alpha \mathbf{Q}) \mathbf{x}^{(k)} \\ &= \beta_k(\mathbf{I} - \alpha \mathbf{Q})(y_1^{(k)} \mathbf{v}_1 + \cdots + y_n^{(k)} \mathbf{v}_n) \\ &= \beta_k(y_1^{(k)}(\mathbf{I} - \alpha \mathbf{Q}) \mathbf{v}_1 + \cdots + y_n^{(k)}(\mathbf{I} - \alpha \mathbf{Q}) \mathbf{v}_n). \end{aligned}$$

But $(\mathbf{I} - \alpha \mathbf{Q}) \mathbf{v}_i = (1 - \alpha \lambda_i) \mathbf{v}_i$, where λ_i is the eigenvalue corresponding to \mathbf{v}_i . Hence,

$$\mathbf{x}^{(k+1)} = \beta_k(y_1^{(k)}(1 - \alpha \lambda_1) \mathbf{v}_1 + \cdots + y_n^{(k)}(1 - \alpha \lambda_n) \mathbf{v}_n),$$

which means that $y_i^{(k+1)} = \beta_k y_i^{(k)}(1 - \alpha \lambda_i)$. In other words, $y_i^{(k)} = \beta^{(k)} y_i^{(0)}(1 - \alpha \lambda_i)^k$, where $\beta^{(k)} = \prod_{i=0}^{k-1} \beta_i$. Rewriting $\mathbf{x}^{(k)}$,

$$\begin{aligned} \mathbf{x}^{(k)} &= \sum_{i=1}^n y_i^{(k)} \mathbf{v}_i \\ &= y_1^{(k)} \left(\mathbf{v}_1 + \sum_{i=2}^n \frac{y_i^{(k)}}{y_1^{(k)}} \mathbf{v}_i \right). \end{aligned}$$

But, assuming $y_1^{(0)} \neq 0$,

$$\frac{y_i^{(k)}}{y_1^{(k)}} = \frac{y_i^{(0)}(1 - \alpha \lambda_i)^k}{y_1^{(0)}(1 - \alpha \lambda_1)^k} = \frac{y_i^{(0)}}{y_1^{(0)}} \left(\frac{1 - \alpha \lambda_i}{1 - \alpha \lambda_1} \right)^k.$$

But because $(1 - \alpha \lambda_i)/(1 - \alpha \lambda_1) < 1$ (because the $\lambda_i > \lambda_1$ for $i > 1$ and $\alpha < 1/\lambda_{\max}$), we deduce that

$$\frac{y_i^{(k)}}{y_1^{(k)}} \rightarrow 0,$$

which implies that $\mathbf{x}^{(k)} \rightarrow \mathbf{v}_1$ as required.

3. (10 pts.) Suppose you are given two different types of concrete. The first type contains 30% cement, 40% gravel, and 30% sand (all percentages of weight). The second type contains 10% cement, 20% gravel, and 70% sand. The first type of concrete costs \$5 per pound, while the second type costs \$1 per pound.

How many pounds of each type of concrete should you buy and mix together so that your cost is minimized but you get a concrete mixture that has at least a total of 5 pounds of cement, 3 pounds of gravel, and 4 pounds of sand? Formulate and solve the problem using a linear programming method. (No need to give details on how you solved it; just give the answer.)

Ans.: The problem can be represented as

$$\begin{array}{ll} \text{minimize} & \mathbf{c}^T \mathbf{x} \\ \text{subject to} & \mathbf{Ax} \geq \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0}, \end{array}$$

where

$$\mathbf{A} = \begin{bmatrix} 0.3 & 0.1 \\ 0.4 & 0.2 \\ 0.3 & 0.7 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 5 \\ 3 \\ 4 \end{bmatrix}, \quad \mathbf{c} = \begin{bmatrix} 5 \\ 1 \end{bmatrix},$$

Using a graphical method or converting to standard form using surplus variables and applying the simplex algorithm, we get a solution of $[0, 50]^T$, which means that we should purchase 50 pounds of the second type of concrete.

4. (16 pts.) You are given a linear programming problem in standard form. Suppose you use the two-phase simplex method and arrive at the following canonical tableau in phase I:

$$\begin{bmatrix} ? & 0 & 1 & 1 & ? & ? & 0 & 6 \\ ? & 0 & 0 & ? & ? & ? & 1 & \alpha \\ ? & 1 & 0 & ? & ? & ? & 0 & 5 \\ \gamma & 0 & 0 & \delta & ? & ? & \beta & 0 \end{bmatrix}.$$

The variables α , β , γ , and δ are unknowns to be determined. Those entries marked with “?” are unspecified. The only thing you are told is that the value of γ is either 2 or -1 .

- Determine the value of α . Explain fully.
- Determine the value of β . Explain fully.
- Determine the value of γ . Explain fully.
- Determine the value of δ . Explain fully.
- Does the given linear programming problem have a feasible solution? If yes, find it. If not, explain why.

Ans.: a. The value of α must be 0, because the objective function value is 0 (lower right corner), and α is the value of an artificial variable.

b. The value of β must be 0, because it is the RCC value corresponding to a basic column.

c. The value of γ must be 2, because it must be a positive value. Otherwise, there is a feasible solution to the artificial problem with objective function value smaller than 0, which is impossible.

- d. The value of δ must be 0, because we must be able to bring the fourth column into the basis without changing the objective function value.
- e. The given linear programming problem does indeed have a feasible solution: $[0, 5, 6, 0]^T$. We obtain this by noticing that the right-most column is a linear combination of the second and third columns, with coefficients 5 and 6.