Midterm solutions

Problem 1. Suppose A is an $m \times p$ matrix with linearly independent columns and B is a $p \times n$ matrix with linearly independent rows. We are not assuming that A or B are square. Define

$$X = AB, \qquad Y = B^{\dagger}A^{\dagger}$$

where A^{\dagger} and B^{\dagger} are the pseudo-inverses of A and B. Show the following properties.

- 1. YX is symmetric.
- 2. XY is symmetric.
- 3. YXY = Y.
- 4. XYX = X.

Carefully explain your answers.

Solution. We use the definitions and properties

$$A^{\dagger} = (A^T A)^{-1} A^T, \qquad B^{\dagger} = B^T (B B^T)^{-1}, \qquad A^{\dagger} A = I, \qquad B B^{\dagger} = I.$$

1. Follows from

$$YX = (B^{\dagger}A^{\dagger})(AB) = B^{\dagger}(A^{\dagger}A)B = B^{\dagger}B = B^{T}(BB^{T})^{-1}B$$

2. Follows from

$$XY = (AB)(B^{\dagger}A^{\dagger}) = A(BB^{\dagger})A^{\dagger} = AA^{\dagger} = A(A^{T}A)^{-1}A^{T}.$$

3. In part 2 we have shown that $XY = AA^{\dagger}$. The result follows from

$$Y(XY) = (B^{\dagger}A^{\dagger})(AA^{\dagger}) = B^{\dagger}(A^{\dagger}A)A^{\dagger} = B^{\dagger}A^{\dagger} = Y.$$

4. Similarly,

$$(XY)X = (AA^{\dagger})(AB) = A(A^{\dagger}A)B) = AB = X.$$

Problem 2.

1. Formulate the following problem as a set of linear equations. Find a point $x \in \mathbf{R}^n$ at equal distance to n + 1 given points $y_1, y_2, \ldots, y_{n+1} \in \mathbf{R}^n$:

$$||x - y_1|| = ||x - y_2|| = \dots = ||x - y_{n+1}||.$$

Write the equations in matrix form Ax = b.

2. Show that the solution x in part 1 is unique if the $(n+1) \times (n+1)$ matrix

$$\left[\begin{array}{rrrr} y_1 & y_2 & \cdots & y_{n+1} \\ 1 & 1 & \cdots & 1 \end{array}\right]$$

is nonsingular.

Solution.

1. Squaring the equations and canceling the terms $x^T x$ we obtain n linear equations

$$2(y_2 - y_1)^T x = ||y_2||^2 - ||y_1||^2$$

$$2(y_3 - y_2)^T x = ||y_3||^2 - ||y_2||^2$$

$$\vdots$$

$$2(y_{n+1} - y_n)^T x = ||y_{n+1}||^2 - ||y_n||^2.$$

In matrix form, Ax = b with

$$A = \begin{bmatrix} (y_2 - y_1)^T \\ (y_3 - y_2)^T \\ \vdots \\ (y_{n+1} - y_n)^T \end{bmatrix}, \qquad b = \frac{1}{2} \begin{bmatrix} \|y_2\|^2 - \|y_1\|^2 \\ \|y_3\|^2 - \|y_2\|^2 \\ \vdots \\ \|y_{n+1}\|^2 - \|y_n\|^2 \end{bmatrix}.$$

2. We show that the matrix in part 1 is nonsingular. Suppose it is singular, so it has linearly dependent rows. This means there exists a nonzero z with

$$\begin{bmatrix} y_2 - y_1 & y_3 - y_2 & \cdots & y_{n+1} - y_n \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{bmatrix} = 0.$$

Then

$$\begin{bmatrix} y_1 & y_2 & \cdots & y_n & y_{n+1} \\ 1 & 1 & \cdots & 1 & 1 \end{bmatrix} \begin{bmatrix} -z_1 \\ z_1 - z_2 \\ \vdots \\ z_{n-1} - z_n \\ z_n \end{bmatrix} = 0,$$

which contradicts the fact that the matrix on the left is nonsingular.

Problem 3. We use the notation I_n for the identity matrix of size $n \times n$ and J_n for the reversal matrix of size $n \times n$. (The reversal matrix is the identity matrix with the column order reversed.)

1. Verify that the $2n \times 2n$ reversal matrix J_{2n} can be written as

$$J_{2n} = Q \begin{bmatrix} I_n & 0\\ 0 & -I_n \end{bmatrix} Q^T \quad \text{where} \quad Q = \frac{1}{\sqrt{2}} \begin{bmatrix} I_n & I_n\\ J_n & -J_n \end{bmatrix}.$$

Also show that Q is orthogonal.

2. Let A be a $2n \times 2n$ matrix with the property that

$$J_{2n}A = AJ_{2n}. (1)$$

An example is the 4×4 matrix

$$\begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 8 & 7 & 6 & 5 \\ 4 & 3 & 2 & 1 \end{bmatrix}$$

Use the factorization of J_{2n} in part 1 to show that if A satisfies (1) then the matrix $Q^T A Q$ is block-diagonal:

$$Q^T A Q = \left[\begin{array}{cc} B & 0 \\ 0 & C \end{array} \right],$$

where B and C are $n \times n$ matrices.

3. The complexity of solving a general linear equation Ax = b of size $2n \times 2n$ is $(2/3)(2n)^3 = (16/3)n^3$. Suppose A has the property defined in part 2. By how much can the dominant term in the complexity of solving Ax = b be reduced if we take advantage of the factorization property in part 2? Explain your answer.

Solution.

1. The first statement follows from

$$Q\begin{bmatrix}I_n & 0\\0 & -I_n\end{bmatrix}Q^T = \frac{1}{2}\begin{bmatrix}I_n & I_n\\J_n & -J_n\end{bmatrix}\begin{bmatrix}I_n & 0\\0 & -I_n\end{bmatrix}\begin{bmatrix}I_n & J_n\\I_n & -J_n\end{bmatrix} = \begin{bmatrix}0 & J_n\\J_n & 0\end{bmatrix} = J_{2n}.$$

Q is orthogonal because it is square and

$$Q^{T}Q = \frac{1}{2} \begin{bmatrix} I_{n} & J_{n} \\ I_{n} & -J_{n} \end{bmatrix} \begin{bmatrix} I_{n} & I_{n} \\ J_{n} & -J_{n} \end{bmatrix} = \begin{bmatrix} I_{n} & 0 \\ 0 & I_{n} \end{bmatrix}.$$

2. Substituting the factorization of J_{2n} we find that

$$Q\begin{bmatrix}I_n & 0\\0 & -I_n\end{bmatrix}Q^T A = AQ\begin{bmatrix}I_n & 0\\0 & -I_n\end{bmatrix}Q^T.$$

Multiplying on the left with Q^T and on the right with Q gives

$$\begin{bmatrix} I_n & 0\\ 0 & -I_n \end{bmatrix} Q^T A Q = Q^T A Q \begin{bmatrix} I_n & 0\\ 0 & -I_n \end{bmatrix}.$$

If we partition $Q^T A Q$ as

$$Q^T A Q = \left[\begin{array}{cc} B & D \\ E & C \end{array} \right]$$

then this means that

$$\begin{bmatrix} B & D \\ -E & -C \end{bmatrix} = \begin{bmatrix} I_n & 0 \\ 0 & -I_n \end{bmatrix} \begin{bmatrix} B & D \\ E & C \end{bmatrix} = \begin{bmatrix} B & D \\ E & C \end{bmatrix} \begin{bmatrix} I_n & 0 \\ 0 & -I_n \end{bmatrix} = \begin{bmatrix} B & -D \\ E & -C \end{bmatrix}$$

Therefore D = -D = 0 and E = -E = 0.

3. The cost is reduced to $2 \times (2/3)n^3 = (4/3)n^3$, *i.e.*, by a factor of four. It is four times less expensive to solve two equations of size n than one equation of size 2n.

The details are as follows. Partition A, x, and b as

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}, \qquad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \qquad b = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}.$$

Note from part 2 that

$$\begin{bmatrix} B & 0 \\ 0 & C \end{bmatrix} = Q^{T}AQ$$

$$= \frac{1}{2} \begin{bmatrix} I_{n} & J_{n} \\ I_{n} & -J_{n} \end{bmatrix} \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} I_{n} & I_{n} \\ J_{n} & -J_{n} \end{bmatrix}$$

$$= \frac{1}{2} \begin{bmatrix} A_{11} + A_{12}J_{n} + J_{n}A_{21} + J_{n}A_{22}J_{n} & 0 \\ 0 & A_{11} - A_{12}J_{n} - J_{n}A_{21} + J_{n}A_{22}J_{n} \end{bmatrix}.$$

Computing B and C costs $6n^2$ flops $(n^2 \text{ for } A_{11} + J_n A_{22} J_n, n^2 \text{ for } A_{12} J_n + J_n A_{21}, n^2 \text{ for the sum of these two matrices, } n^2 \text{ for the difference, and } 2n^2 \text{ for the scalar multiplication with } 1/2).$ To solve Ax = b we use the factorization

$$A = Q \left[\begin{array}{cc} B & 0 \\ 0 & C \end{array} \right] Q^T.$$

• Solve Qu = b. Since Q is orthogonal, the solution is

$$u = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = Q^T b = \frac{1}{\sqrt{2}} \begin{bmatrix} b_1 + J_n b_2 \\ b_1 - J_n b_2 \end{bmatrix}.$$

This requires 4n flops.

• Compute B and C ($6n^2$ flops) and solve

$$\left[\begin{array}{cc} B & 0 \\ 0 & C \end{array}\right] \left[\begin{array}{c} y_1 \\ y_2 \end{array}\right] = \left[\begin{array}{c} u_1 \\ u_2 \end{array}\right].$$

This is equivalent to two independent linear equations $By_1 = u_1$ and $Cy_2 = u_2$. The complexity is $(4/3)n^3$.

• Solve $Q^T x = y$. Since Q is orthogonal, the solution is

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = Qy = \frac{1}{\sqrt{2}} \begin{bmatrix} y_1 + y_2 \\ J_n y_1 - J_n y_2 \end{bmatrix}$$

This requires 4n flops.

The dominant term is $(4/3)n^3$.

Problem 4. Let A be an $m \times n$ matrix with linearly independent columns. Suppose $A_{ij} = 0$ for i > j + 1. In other words, the elements of A below the first subdiagonal are zero:

$$A = \begin{bmatrix} A_{11} & A_{12} & \cdots & A_{1,n-1} & A_{1n} \\ A_{21} & A_{22} & \cdots & A_{2,n-1} & A_{2n} \\ 0 & A_{32} & \cdots & A_{3,n-1} & A_{3n} \\ 0 & 0 & \cdots & A_{4,n-1} & A_{4n} \\ \vdots & \vdots & & \vdots & \vdots \\ 0 & 0 & \cdots & 0 & A_{n+1,n} \\ 0 & 0 & \cdots & 0 & 0 \\ \vdots & \vdots & & \vdots & \vdots \\ 0 & 0 & \cdots & 0 & 0 \end{bmatrix}$$

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Show that the Q-factor in the QR factorization of A has the same property: $Q_{ij} = 0$ for i > j + 1.

Solution. This follows from $Q = AR^{-1}$ and the fact that R^{-1} is upper triangular. Therefore the kth column of Q is a linear combination of the first k columns of A.

Problem 5. Consider a square $(n + m) \times (n + m)$ -matrix

$$\left[\begin{array}{cc} A & B \\ C & D \end{array}\right]$$

with A of size $n \times n$ and D of size $m \times m$. We assume A is nonsingular. The matrix

$$S = D - CA^{-1}B$$

is called the *Schur complement* of A. Describe efficient algorithms for computing the Schur complement of each of the following types of matrices A.

- 1. A is diagonal.
- 2. A is lower triangular.
- 3. A is a general square matrix.

In each subproblem, give the different steps in the algorithm and their complexity. Include in the total flop count all terms that are order three (n^3, n^2m, nm^2, m^3) or higher. If you know different algorithms, choose the most efficient one.

Solution. In all three problems, we first compute $X = A^{-1}B$ and then compute S = D - CX. The complexity is $2m^2n$ for the product CX and m^2 for the subtraction. The complexity of computing $X = A^{-1}B$ depends on the properties of A.

- 1. If A is diagonal we compute the elements of $X = A^{-1}B$ as $X_{ij} = B_{ij}/A_{ii}$. This requires mn flops.
- 2. If A is lower triangular, we solve AX = B, column by column, via forward substitution. This requires mn^2 flops.
- 3. In the general case we compute an LU factorization of $A((2/3)n^3$ flops) and then solve AX = B column by column $(2mn^2$ flops).

Keeping only third order terms we obtain for the total complexity:

- 1. $2m^2n$.
- 2. $2m^2n + mn^2$.
- 3. $2m^2n + 2mn^2 + (2/3)n^3$.