

233 Computational Techniques

Problem Sheet for Tutorial 5

Problem 1

Solve the following system of equations using Gauss-Jordan elimination. Identify basic variables. Express all solutions in terms of non-basic variables. Determine the space of solutions and verify the result.

$$\begin{array}{rcccccc} 2x_1 & + & x_2 & - & x_3 & + & 2x_4 & - & x_5 & = & -2 \\ 4x_1 & + & 2x_2 & & & + & 3x_4 & - & 2x_5 & = & 2 \\ x_1 & + & x_2 & + & x_3 & + & x_4 & + & x_5 & = & 3 \end{array}$$

Problem 2

(a) Find the *Cholesky factorization* of the matrix

$$\mathbf{A} = \begin{bmatrix} 1 & 1 & -1 \\ 1 & 5 & -5 \\ -1 & -5 & 6 \end{bmatrix}.$$

(b) Then solve $\mathbf{Ax} = \mathbf{b}$ for $\mathbf{b} = [1, -3, 6]^T$ by forward and backward substitution, using the triangular shape of the factorization matrices. (See the end of section 3.5.1 in the lecture notes.)

Problem 3

Let $\mathbf{a}_1 = [1, 2, 2]^T$, $\mathbf{a}_2 = [1, 0, 1]^T$, and

$$P = \{x_1\mathbf{a}_1 + x_2\mathbf{a}_2 : x_1, x_2 \in \mathbb{R}\}$$

be the plane in \mathbb{R}^3 spanned by the vectors \mathbf{a}_1 and \mathbf{a}_2 .

(a) Find the matrix $\mathbf{A} \in \mathbb{R}^{3 \times 2}$ of which P is the range space. Then determine the vector \mathbf{p} in P with the shortest ℓ_2 distance from the vector $\mathbf{b} = [3, -1, 1]^T$, using the normal equations with Cholesky factorization.

(b) What are the range and nullspace components \mathbf{b}_R and \mathbf{b}_N of the vector \mathbf{b} ? Check that they are orthogonal.

(c) What is the shortest distance of \mathbf{b} from P ?

(d) Alternative geometric approach: Determine \mathbf{b}_N and \mathbf{b}_R using orthogonal projection onto the nullspace of the relevant matrix \mathbf{A}^T . *Hint:* The cross product is a good tool to find this nullspace.

Solution

Problem 1

In tableau notation, a possible sequence of steps is the following (with the third equation as the first row; pivot elements underlined):

$$\begin{aligned} \left[\begin{array}{ccccc|c} \underline{1} & 1 & 1 & 1 & 1 & 3 \\ 2 & 1 & -1 & 2 & -1 & -2 \\ 4 & 2 & 0 & 3 & -2 & 2 \end{array} \right] &\longrightarrow \left[\begin{array}{ccccc|c} 1 & 1 & 1 & 1 & 1 & 3 \\ 0 & \underline{-1} & -3 & 0 & -3 & -8 \\ 0 & -2 & -4 & -1 & -6 & -10 \end{array} \right] \longrightarrow \\ \left[\begin{array}{ccccc|c} 1 & 0 & -2 & 1 & -2 & 5 \\ 0 & 1 & 3 & 0 & 3 & 8 \\ 0 & 0 & \underline{2} & -1 & 0 & 6 \end{array} \right] &\longrightarrow \left[\begin{array}{ccccc|c} 1 & 0 & 0 & 0 & -2 & 1 \\ 0 & 1 & 0 & 3/2 & 3 & -1 \\ 0 & 0 & 1 & -1/2 & 0 & 3 \end{array} \right]. \end{aligned}$$

The first three columns of the final tableau contain the three unit vectors \mathbf{e}_1 , \mathbf{e}_2 and \mathbf{e}_3 . (A different choice of pivot elements would have produced them in other columns.) So x_1 , x_2 and x_3 are the basic variables, and we can read off

$$x_1 = 1 + 2x_5, \quad x_2 = -1 - \frac{3}{2}x_4 - 3x_5, \quad x_3 = 3 + \frac{1}{2}x_4,$$

where the non-basic variables x_4 and x_5 can take arbitrary values. Hence the set of solutions is

$$\{\mathbf{x} = [1, -1, 3, 0, 0]^T + x_4[0, -3/2, 1/2, 1, 0]^T + x_5[2, -3, 0, 0, 1]^T : x_4, x_5 \in \mathbb{R}\}.$$

(Note that this is *not* a vector space! For instance, the zero vector $\mathbf{x} = \mathbf{0}$ is not a solution. In general, for a given matrix \mathbf{A} and a given vector \mathbf{b} (with dimension equal to the row dimension of \mathbf{A}) the set $\{\mathbf{x} : \mathbf{Ax} = \mathbf{b}\}$ is a vector space if and only if $\mathbf{b} = \mathbf{0}$.)

Problem 2

(a) The aim is to split \mathbf{A} in the following way:

$$\begin{bmatrix} 1 & 1 & -1 \\ 1 & 5 & -5 \\ -1 & -5 & 6 \end{bmatrix} = \mathbf{A} = \mathbf{LL}^T = \begin{bmatrix} \ell_{11} & & \\ \ell_{21} & \ell_{22} & \\ \ell_{31} & \ell_{32} & \ell_{33} \end{bmatrix} \begin{bmatrix} \ell_{11} & \ell_{21} & \ell_{31} \\ & \ell_{22} & \ell_{32} \\ & & \ell_{33} \end{bmatrix}.$$

With the usual sign convention $\ell_{ii} > 0$ for the diagonal elements of \mathbf{L} , one obtains

- from the first column of \mathbf{A} :

$$1 = \ell_{11}^2 \Rightarrow \ell_{11} = 1, \quad 1 = \ell_{21}\ell_{11} = \ell_{21}, \quad -1 = \ell_{31}\ell_{11} = \ell_{31},$$

- from the second column of \mathbf{A} , starting with a_{22} (as \mathbf{A} is symmetric, the equation for a_{21} is the same as the one for a_{12} , which we have just solved)

$$5 = \ell_{21}^2 + \ell_{22}^2 = 1 + \ell_{22}^2 \Rightarrow \ell_{22} = 2, \quad -5 = \ell_{31}\ell_{21} + \ell_{32}\ell_{22} = -1 + 2\ell_{32} \Rightarrow \ell_{32} = -2,$$

- and from the third column (only the last element can give anything new),

$$6 = \ell_{31}^2 + \ell_{32}^2 + \ell_{33}^2 = 1 + 4 + \ell_{33}^2 \Rightarrow \ell_{33} = 1 .$$

So

- the factorization was successful (which implies that \mathbf{A} is positive definite), and
- the Cholesky factor of \mathbf{A} is

$$\mathbf{L} = \begin{bmatrix} 1 & & \\ 1 & 2 & \\ -1 & -2 & 1 \end{bmatrix} . \quad (1)$$

(b) The idea is to solve $\mathbf{L}\mathbf{L}^T\mathbf{x} = \mathbf{A}\mathbf{x} = \mathbf{b}$ in two steps by defining $\mathbf{y} := \mathbf{L}^T\mathbf{x}$ and solving (1) $\mathbf{L}\mathbf{y} = \mathbf{b}$ by forward substitution and (2) $\mathbf{L}^T\mathbf{x} = \mathbf{y}$ by backward substitution. So:

$$\begin{aligned} \begin{bmatrix} 1 \\ -3 \\ 6 \end{bmatrix} = \mathbf{b} = \mathbf{L}\mathbf{y} &= \begin{bmatrix} 1 & & \\ 1 & 2 & \\ -1 & -2 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_1 + 2y_2 \\ -y_1 - 2y_2 + y_3 \end{bmatrix} \Rightarrow \mathbf{y} = \begin{bmatrix} 1 \\ -2 \\ 3 \end{bmatrix} \\ \begin{bmatrix} 1 \\ -2 \\ 3 \end{bmatrix} = \mathbf{y} = \mathbf{L}^T\mathbf{x} &= \begin{bmatrix} 1 & 1 & -1 \\ & 2 & -2 \\ & & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_1 + x_2 - x_3 \\ 2x_2 - 2x_3 \\ x_3 \end{bmatrix} \Rightarrow \mathbf{x} = \begin{bmatrix} 2 \\ 2 \\ 3 \end{bmatrix} . \end{aligned}$$

Problem 3

(a) The matrix is

$$\mathbf{A} = [\mathbf{a}_1, \mathbf{a}_2] = \begin{bmatrix} 1 & 1 \\ 2 & 0 \\ 2 & 1 \end{bmatrix} .$$

- We seek $\mathbf{p} = x_1\mathbf{a}_1 + x_2\mathbf{a}_2$ minimizing $\|\mathbf{b} - \mathbf{p}\|_2^2 = \|\mathbf{b} - \mathbf{A}\mathbf{x}\|_2^2$, where $\mathbf{x} = [x_1, x_2]^T$.
- Any such \mathbf{x} will be a solution of the normal equation, $\mathbf{A}^T\mathbf{A}\mathbf{x} = \mathbf{A}^T\mathbf{b}$. So we compute $\mathbf{A}^T\mathbf{A}$ and apply Cholesky factorization:

$$\mathbf{A}^T\mathbf{A} = \begin{bmatrix} 9 & 3 \\ 3 & 2 \end{bmatrix} = \underbrace{\begin{bmatrix} 3 & \\ 1 & 1 \end{bmatrix}}_{\mathbf{L}} \underbrace{\begin{bmatrix} 3 & 1 \\ & 1 \end{bmatrix}}_{\mathbf{L}^T} .$$

(As the factorization was successful, we know that the solution \mathbf{x} is unique.)

- Using the factorization of \mathbf{A} , we can write this equation as $\mathbf{L}\mathbf{L}^T\mathbf{x} = \mathbf{A}^T\mathbf{b}$ and hence solve it in two steps, (1) $\mathbf{L}\mathbf{y} = \mathbf{A}^T\mathbf{b}$, and (2) $\mathbf{L}^T\mathbf{x} = \mathbf{y}$:

$$\begin{aligned} \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \mathbf{A}^T\mathbf{b} = \mathbf{L}\mathbf{y} &= \begin{bmatrix} 3 & \\ 1 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} 3y_1 \\ y_1 + y_2 \end{bmatrix} \Rightarrow \mathbf{y} = \begin{bmatrix} 1 \\ 3 \end{bmatrix} , \\ \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \mathbf{y} = \mathbf{L}^T\mathbf{x} &= \begin{bmatrix} 3 & 1 \\ & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3x_1 + x_2 \\ x_2 \end{bmatrix} \Rightarrow \mathbf{x} = \begin{bmatrix} -2/3 \\ 3 \end{bmatrix} . \end{aligned}$$

(Alternatively, $\mathbf{A}^T \mathbf{A} \mathbf{x} = \mathbf{A}^T \mathbf{b}$ can also be solved by Gauss elimination.)

- For the vector \mathbf{p} we obtain

$$\mathbf{p} = -(2/3) \cdot [1, 2, 2]^T + 3 \cdot [1, 0, 1]^T = [7/3, -4/3, 5/3]^T .$$

(b)

$\mathbf{b}_R = \mathbf{A} \mathbf{x} = \mathbf{p}$, since this is how the minimizing solution \mathbf{x} of the least squares problem is constructed. The component of \mathbf{b} in the nullspace of \mathbf{A}^T is $\mathbf{b}_N = \mathbf{b} - \mathbf{b}_R = [2/3, 1/3, -2/3]^T$. The dot product of \mathbf{b}_R and \mathbf{b}_N is $\mathbf{b}_R^T \mathbf{b}_N = 14/9 - 4/9 - 10/9 = 0$, so they are indeed orthogonal.

(c) The minimal distance is $\|\mathbf{b}_N\|_2 = \sqrt{4/9 + 1/9 + 4/9} = 1$.

(d) By definition, $\mathbf{b}_R \in \text{range}(\mathbf{A}) = P$ and $\mathbf{b}_N \in \text{null}(\mathbf{A}^T)$. As $\text{range}(\mathbf{A}) \perp \text{null}(\mathbf{A}^T)$ and $\text{null}(\mathbf{A}^T)$ must be one-dimensional, it must be spanned by the cross-product of \mathbf{a}_1 and \mathbf{a}_2 ,

$$\text{null}(\mathbf{A}^T) = \mathbb{R} \mathbf{a}_1 \times \mathbf{a}_2 = \mathbb{R} \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} \times \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \mathbb{R} \underbrace{\begin{bmatrix} 2 \\ 1 \\ -2 \end{bmatrix}}_{\mathbf{s}},$$

and $\mathbf{b}_N = t\mathbf{s}$ for some $t \in \mathbb{R}$. From $\mathbf{b} = \mathbf{b}_R + \mathbf{b}_N = \mathbf{b}_R + t\mathbf{s}$, we find the value of t by taking the dot product with \mathbf{s} , since as an element of $\text{null}(\mathbf{A}^T)$, \mathbf{s} is orthogonal to \mathbf{b}_R :

$$\mathbf{s}^T \mathbf{b} = t \|\mathbf{s}\|_2^2 \implies t = \frac{\mathbf{s}^T \mathbf{b}}{\|\mathbf{s}\|_2^2} = \frac{1}{3} .$$

So $\mathbf{b}_N = [2/3, 1/3, -2/3]$, and $\mathbf{b}_R = \mathbf{b} - \mathbf{b}_N = [7/3, -4/3, 5/3]$, as above.