

NATIONAL UNIVERSITY OF SINGAPORE
Department of Mathematics

Module: MA1508E Linear Algebra for Engineering
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Tutorial: 9

1. For each of the following linear system $\mathbf{Ax} = \mathbf{b}$,

- (i) Show that the system is inconsistent;
- (ii) Find a least squares solution \mathbf{x}' to the system. Is there a unique least squares solution or infinitely many?
- (iii) Compute the least squares error, defined as $\|\mathbf{b} - \mathbf{Ax}'\|$. If there are infinitely many least squares solution and $\mathbf{x}'_1, \mathbf{x}'_2$ are any two of them, would the least squares error $\|\mathbf{b} - \mathbf{Ax}'_1\|$ and $\|\mathbf{b} - \mathbf{Ax}'_2\|$ be the same?

(a) $\mathbf{A} = \begin{pmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{pmatrix} \quad \mathbf{b} = \begin{pmatrix} 4 \\ 1 \\ 3 \end{pmatrix}.$

(b) $\mathbf{A} = \begin{pmatrix} 3 & 2 & -1 \\ 1 & -4 & 3 \\ 1 & 10 & -7 \end{pmatrix} \quad \mathbf{b} = \begin{pmatrix} 2 \\ -2 \\ 1 \end{pmatrix}.$

(a) (i)

$$\left(\begin{array}{cc|c} 1 & -1 & 4 \\ 3 & 2 & 1 \\ -2 & 4 & 3 \end{array} \right) \longrightarrow \left(\begin{array}{cc|c} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{array} \right).$$

So the system is inconsistent.

(ii) We solve $\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$:

$$\left(\begin{array}{cc|c} 14 & -3 & 1 \\ -3 & 21 & 10 \end{array} \right) \longrightarrow \left(\begin{array}{cc|c} 1 & 0 & \frac{17}{95} \\ 0 & 1 & \frac{143}{285} \end{array} \right).$$

So the least squares solution is $\mathbf{x}' = (x_1, x_2) = (\frac{17}{95}, \frac{143}{285})$, and it is unique.

(iii) The least squares error is

$$\|\mathbf{b} - \mathbf{Ax}'\| = \left\| \begin{pmatrix} 4 \\ 1 \\ 3 \end{pmatrix} - \begin{pmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{pmatrix} \begin{pmatrix} \frac{17}{95} \\ \frac{143}{285} \end{pmatrix} \right\| = \left\| \begin{pmatrix} \frac{1232}{285} \\ -\frac{154}{285} \\ \frac{77}{57} \end{pmatrix} \right\| \approx 4.5611.$$

(b) (i)

$$\left(\begin{array}{ccc|c} 3 & 2 & -1 & 2 \\ 1 & -4 & 3 & -2 \\ 1 & 10 & -7 & 1 \end{array} \right) \longrightarrow \left(\begin{array}{ccc|c} 1 & 0 & \frac{1}{7} & 0 \\ 0 & 1 & -\frac{5}{7} & 0 \\ 0 & 0 & 0 & 1 \end{array} \right).$$

So the system is inconsistent.

(ii) We solve $\mathbf{A}^T \mathbf{A} \mathbf{x} = \mathbf{A}^T \mathbf{b}$:

$$\left(\begin{array}{ccc|c} 11 & 12 & -7 & 5 \\ 12 & 120 & -84 & 22 \\ -7 & -84 & 59 & -15 \end{array} \right) \longrightarrow \left(\begin{array}{ccc|c} 1 & 0 & \frac{1}{7} & \frac{2}{7} \\ 0 & 1 & -\frac{5}{7} & \frac{13}{84} \\ 0 & 0 & 0 & 0 \end{array} \right).$$

So there are infinitely many least squares solutions. One of them is $\mathbf{x}' = (x_1, x_2, x_3) = (\frac{2}{7}, \frac{13}{84}, 0)$.

(iii) The least squares error is

$$\|\mathbf{b} - \mathbf{A}\mathbf{x}'\| = \left\| \begin{pmatrix} 2 \\ -2 \\ 1 \end{pmatrix} - \begin{pmatrix} 3 & 2 & -1 \\ 1 & -4 & 3 \\ 1 & 10 & -7 \end{pmatrix} \begin{pmatrix} \frac{2}{7} \\ \frac{13}{84} \\ 0 \end{pmatrix} \right\| = \left\| \begin{pmatrix} \frac{5}{6} \\ -\frac{5}{6} \\ -\frac{5}{6} \end{pmatrix} \right\| \approx 2.0412.$$

The least squares errors for different least squares solution would be the same since $\mathbf{A}\mathbf{x}'_1$ and $\mathbf{A}\mathbf{x}'_2$ are equal.

2. For each of the following, compute the orthogonal projection of \mathbf{u} onto the subspace spanned by $\mathbf{v}_1, \dots, \mathbf{v}_k$.

(a) $\mathbf{u} = (1, -6, 1)$, $\mathbf{v}_1 = (-1, 2, 1)$, $\mathbf{v}_2 = (2, 2, 4)$.

(b) $\mathbf{u} = (6, 12, 3, 6)$, $\mathbf{v}_1 = (1, 1, 0, 0)$, $\mathbf{v}_2 = (1, 0, 1, 0)$, $\mathbf{v}_3 = (3, 1, 1, 1)$.

(a) An orthogonal basis for $\text{span}\{\mathbf{v}_1, \mathbf{v}_2\}$ is $\{\mathbf{w}_1, \mathbf{w}_2\}$ where

$$\begin{aligned} \mathbf{w}_1 &= \mathbf{v}_1 \\ \mathbf{w}_2 &= \mathbf{v}_2 - \frac{\mathbf{v}_2 \cdot \mathbf{w}_1}{\mathbf{w}_1 \cdot \mathbf{w}_1} \mathbf{w}_1 \\ &= (2, 2, 4) - \frac{6}{6}(-1, 2, 1) = (3, 0, 3). \end{aligned}$$

So the orthogonal projection of \mathbf{u} onto $\text{span}\{\mathbf{v}_1, \mathbf{v}_2\} = \text{span}\{\mathbf{w}_1, \mathbf{w}_2\}$ is

$$\frac{\mathbf{u} \cdot \mathbf{w}_1}{\mathbf{w}_1 \cdot \mathbf{w}_1} \mathbf{w}_1 + \frac{\mathbf{u} \cdot \mathbf{w}_2}{\mathbf{w}_2 \cdot \mathbf{w}_2} \mathbf{w}_2 = -2\mathbf{w}_1 + \frac{6}{18} \mathbf{w}_2 = -2(-1, 2, 1) + \frac{1}{3}(3, 0, 3) = (3, -4, -1).$$

(b) Applying Gram-Schmidt Process, we obtain an orthogonal basis $\{\mathbf{w}_1, \mathbf{w}_2, \mathbf{w}_3\}$ for $\text{span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ where

$$\mathbf{w}_1 = (1, 1, 0, 0), \quad \mathbf{w}_2 = \left(\frac{1}{2}, -\frac{1}{2}, 1, 0\right), \quad \mathbf{w}_3 = \left(\frac{1}{3}, -\frac{1}{3}, -\frac{1}{3}, 1\right).$$

So the orthogonal projection of \mathbf{u} onto $\text{span}\{\mathbf{w}_1, \mathbf{w}_2, \mathbf{w}_3\}$ is

$$\begin{aligned} & \frac{\mathbf{u} \cdot \mathbf{w}_1}{\mathbf{w}_1 \cdot \mathbf{w}_1} \mathbf{w}_1 + \frac{\mathbf{u} \cdot \mathbf{w}_2}{\mathbf{w}_2 \cdot \mathbf{w}_2} \mathbf{w}_2 + \frac{\mathbf{u} \cdot \mathbf{w}_3}{\mathbf{w}_3 \cdot \mathbf{w}_3} \mathbf{w}_3 \\ &= 9(1, 1, 0, 0) + 0\left(\frac{1}{2}, -\frac{1}{2}, 1, 0\right) + \frac{9}{4}\left(\frac{1}{3}, -\frac{1}{3}, -\frac{1}{3}, 1\right) = \left(\frac{39}{4}, \frac{33}{4}, -\frac{3}{4}, \frac{9}{4}\right). \end{aligned}$$

3. A series of experiments were performed to investigate the relationship between two physical quantities x and y . The results of the experiments are shown in the table below.

x	0	1	2	3
y	3	2	4	4

- (a) Find a least squares solution $\mathbf{x} = (\hat{a}, \hat{b})$ if it is believed that x and y are related linearly, that is, $y = ax + b$.
- (b) Find a least squares solution $\mathbf{x} = (\hat{a}, \hat{b}, \hat{c})$ if it is believed that x and y are related by the quadratic polynomial $y = ax^2 + bx + c$.
- (c) Which model (linear or quadratic) would produce a smaller least squares error?

- (a) We find a least squares solution to

$$\begin{pmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} 3 \\ 2 \\ 4 \\ 4 \end{pmatrix}.$$

Solving $\mathbf{A}^T \mathbf{A} \mathbf{x} = \mathbf{A}^T \mathbf{b}$,

$$\left(\begin{array}{cc|c} 14 & 6 & 22 \\ 6 & 4 & 13 \end{array} \right) \longrightarrow \left(\begin{array}{cc|c} 1 & 0 & \frac{1}{2} \\ 0 & 1 & \frac{5}{2} \end{array} \right)$$

So a least squares solution is $(\hat{a}, \hat{b}) = (\frac{1}{2}, \frac{5}{2})$.

- (b) We find a least squares solution to

$$\begin{pmatrix} 0 & 0 & 1 \\ 1 & 1 & 1 \\ 4 & 2 & 1 \\ 9 & 3 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} 3 \\ 2 \\ 4 \\ 4 \end{pmatrix}.$$

Solving $\mathbf{A}^T \mathbf{A} \mathbf{x} = \mathbf{A}^T \mathbf{b}$,

$$\left(\begin{array}{ccc|c} 98 & 36 & 14 & 54 \\ 36 & 14 & 6 & 22 \\ 14 & 6 & 4 & 13 \end{array} \right) \longrightarrow \left(\begin{array}{ccc|c} 1 & 0 & 0 & \frac{1}{4} \\ 0 & 1 & 0 & -\frac{1}{4} \\ 0 & 0 & 1 & \frac{11}{4} \end{array} \right).$$

So a least squares solution is $(\hat{a}, \hat{b}, \hat{c}) = (\frac{1}{4}, -\frac{1}{4}, \frac{11}{4})$.

- (c) For the linear model, the least squares error is

$$\left\| \begin{pmatrix} 3 \\ 2 \\ 4 \\ 4 \end{pmatrix} - \begin{pmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} \frac{1}{2} \\ \frac{5}{2} \end{pmatrix} \right\| = \left\| \begin{pmatrix} \frac{1}{2} \\ -1 \\ \frac{1}{2} \\ 0 \end{pmatrix} \right\| \approx 1.2247.$$

For the quadratic model, the least squares error is

$$\left\| \begin{pmatrix} 3 \\ 2 \\ 4 \\ 4 \end{pmatrix} - \begin{pmatrix} 0 & 0 & 1 \\ 1 & 1 & 1 \\ 4 & 2 & 1 \\ 9 & 3 & 1 \end{pmatrix} \begin{pmatrix} \frac{1}{4} \\ -\frac{1}{4} \\ \frac{11}{4} \end{pmatrix} \right\| = \left\| \begin{pmatrix} \frac{1}{4} \\ -\frac{3}{4} \\ \frac{3}{4} \\ -\frac{1}{4} \end{pmatrix} \right\| \approx 1.118.$$

So the quadratic model has a smaller least squares error.

4. Prove that if \mathbf{A} has linearly independent column vectors, and if \mathbf{b} is orthogonal to the column space of \mathbf{A} , then the least squares solution of $\mathbf{Ax} = \mathbf{b}$ is $\mathbf{x} = \mathbf{0}$.

Let \mathbf{A} be a $m \times n$ matrix. Since \mathbf{A} has linearly independent columns, we have $\text{rank}(\mathbf{A}) = n$. By dimension theorem for matrices, $\text{nullity}(\mathbf{A}) = 0$. From Tutorial 8, we know that the nullspace of \mathbf{A} is equal to the nullspace of $\mathbf{A}^T \mathbf{A}$, so we know that $\text{nullity}(\mathbf{A}^T \mathbf{A}) = 0$, in other words, $\mathbf{A}^T \mathbf{A}$ (which has n columns) has rank n and thus is invertible.

Write the vectors in this question as column vectors. Let $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ be the columns of \mathbf{A} . Since \mathbf{b} is orthogonal to the column space of \mathbf{A} , we have $\mathbf{a}_i \cdot \mathbf{b} = 0$ for all $i = 1, 2, \dots, n$. Thus $\mathbf{A}^T \mathbf{b} = \mathbf{0}$.

So a least squares solution of $\mathbf{Ax} = \mathbf{b}$ is a solution of $\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$, which implies

$$(\mathbf{A}^T \mathbf{A})\mathbf{x} = \mathbf{A}^T \mathbf{b} \Rightarrow (\mathbf{A}^T \mathbf{A})\mathbf{x} = \mathbf{0} \Rightarrow \mathbf{x} = \mathbf{0}$$

since $(\mathbf{A}^T \mathbf{A})$ is invertible.

5. (**QR-factorisation**) Let $\mathbf{A} = \begin{pmatrix} 1 & -1 & -1 \\ 1 & 0 & 0 \\ 1 & -1 & 0 \\ 0 & 0 & -1 \end{pmatrix}$, $\mathbf{u}_1 = (1, 1, 1, 0)^T$, $\mathbf{u}_2 = (-1, 0, -1, 0)^T$, $\mathbf{u}_3 = (-1, 0, 0, -1)^T$.

- Use Gram-Schmidt Process to transform $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ into an orthonormal basis $\{\mathbf{w}_1, \mathbf{w}_2, \mathbf{w}_3\}$ for the column space of \mathbf{A} . (Do not change the order of $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$ when applying the Gram-Schmidt Process.)
- Write each of $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$ as a linear combination of $\mathbf{w}_1, \mathbf{w}_2, \mathbf{w}_3$.
- Hence or otherwise, write $\mathbf{A} = \mathbf{QR}$ where \mathbf{Q} is a 4×3 matrix with orthonormal columns and \mathbf{R} is a 3×3 upper triangular matrix with positive entries along its diagonal.

Remark: **QR**-factorisation is widely used in computer algorithms for various computations concerning matrices.

- (a) $\mathbf{w}_1 = \frac{1}{\sqrt{3}}(1, 1, 1, 0)^T$.

$$\mathbf{w}_2 = \frac{1}{\sqrt{6}}(-1, 2, -1, 0), \quad \mathbf{w}_3 = \frac{1}{\sqrt{6}}(-1, 0, 1, -2).$$

(b)

$$\begin{aligned}\mathbf{u}_1 &= \sqrt{3}\mathbf{w}_1 + 0\mathbf{w}_2 + 0\mathbf{w}_3 \\ \mathbf{u}_2 &= -\frac{2}{\sqrt{3}}\mathbf{w}_1 + \frac{2}{\sqrt{6}}\mathbf{w}_2 + 0\mathbf{w}_3 \\ \mathbf{u}_3 &= -\frac{1}{\sqrt{3}}\mathbf{w}_1 + \frac{1}{\sqrt{6}}\mathbf{w}_2 + \frac{\sqrt{6}}{2}\mathbf{w}_3\end{aligned}$$

(c) $\mathbf{A} = \mathbf{Q}\mathbf{R}$ where

$$\mathbf{Q} = (\mathbf{w}_1 \quad \mathbf{w}_2 \quad \mathbf{w}_3) = \begin{pmatrix} \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{6}} & -\frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & \frac{2}{\sqrt{6}} & 0 \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} \\ 0 & 0 & -\frac{2}{\sqrt{6}} \end{pmatrix} \quad \mathbf{R} = \begin{pmatrix} \sqrt{3} & -\frac{2}{\sqrt{3}} & -\frac{1}{\sqrt{3}} \\ 0 & \frac{2}{\sqrt{6}} & \frac{1}{\sqrt{6}} \\ 0 & 0 & \frac{\sqrt{6}}{2} \end{pmatrix}.$$