

Tutorial 9 (Linear algebra I) -1-

Solutions

Exercise 1

$$A = \begin{pmatrix} 2 & 1 & 1 \\ 1 & 1 & 0 \\ 2 & -1 & 2 \end{pmatrix}, B = \begin{pmatrix} 0 & 1 & 2 & 3 \\ 1 & 3 & 0 & 2 \\ 2 & 1 & -5 & -1 \end{pmatrix}.$$

i) $AB = \begin{pmatrix} 3 & 6 & -1 & 7 \\ 1 & 4 & 2 & 5 \\ 3 & 1 & -6 & 2 \end{pmatrix}$

BA is not defined, since a $m \times n$ matrix can just be multiplied with a $n \times l$ matrix. But $B \in \mathbb{R}^{3 \times 4}$ and $A \in \mathbb{R}^{3 \times 3}$.

ii) B is not invertible because it is not a square matrix.
To check if A is invertible we consider the rref of $(A | I_3)$

$$(A | I_3) \xrightarrow{\text{RREF}} \left(\begin{array}{ccc|ccc} 2 & 1 & 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 0 \\ 2 & -1 & 2 & 0 & 0 & 1 \end{array} \right) \sim \left(\begin{array}{ccc|ccc} 1 & 0 & 1 & 1 & -1 & 0 \\ 1 & 1 & 0 & 0 & 1 & 0 \\ 0 & -3 & 2 & 0 & 0 & 1 \end{array} \right)$$

$$\sim \left(\begin{array}{ccc|ccc} 1 & 0 & 1 & 1 & -1 & 0 \\ 0 & 1 & -1 & -1 & 2 & 0 \\ 0 & -3 & 2 & 0 & 0 & 1 \end{array} \right) \sim \left(\begin{array}{ccc|ccc} 1 & 0 & 1 & 1 & -1 & 0 \\ 0 & 1 & -1 & -1 & 2 & 0 \\ 0 & 0 & -1 & -3 & 4 & 1 \end{array} \right) \sim \left(\begin{array}{ccc|ccc} 1 & 0 & 1 & 1 & -1 & 0 \\ 0 & 1 & -1 & -1 & 2 & 0 \\ 0 & 0 & 1 & 3 & -4 & -1 \end{array} \right)$$

$$\sim \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & -2 & 3 & 1 \\ 0 & 1 & 0 & 2 & -2 & 1 \\ 0 & 0 & 1 & 3 & -4 & -1 \end{array} \right) = (I_3 | \bar{A}^{-1}).$$

iii) With $C = \bar{A}^{-1}B$ we have $AC = \overset{I}{\underset{\sim}{\bar{A}\bar{A}'}} B = B$.

$$\left(\begin{array}{ccc} -2 & 3 & 1 \\ 2 & -2 & -1 \\ 3 & -4 & -1 \end{array} \right) \left(\begin{array}{cccc} 0 & 1 & 2 & 3 \\ 1 & 3 & 0 & 2 \\ 2 & 1 & -5 & -1 \end{array} \right) = \left(\begin{array}{cccc} 5 & 8 & -9 & -1 \\ -4 & -5 & 9 & 3 \\ -6 & -10 & 11 & 2 \end{array} \right).$$

Exercise 2

i) In general: If $u_1, u_2 \in \mathbb{R}^3$ are two linearly independent vectors, then $U = \text{span}\{u_1, u_2\}$ is a subspace with $\dim U = 2$.

Example: $u_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, u_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$

$$U = \text{span}\{u_1, u_2\} = \left\{ \begin{pmatrix} t_1 \\ t_2 \\ 0 \end{pmatrix} \in \mathbb{R}^3 \mid t_1, t_2 \in \mathbb{R} \right\}.$$

ii) In general: If (u_1, u_2) is a basis of U , then one can choose $u_3 = u_1 + u_2$.

$$\begin{aligned} \text{If } u \in U, \text{ then } u &= \lambda_1 u_1 + \lambda_2 u_2 \in \text{span}\{u_1, u_2\} \\ &= (\lambda_1 - \lambda_2)u_1 + \lambda_2 u_3 \in \text{span}\{u_1, u_3\} \\ &= (\lambda_2 - \lambda_1)u_2 + \lambda_1 u_3 \in \text{span}\{u_2, u_3\}. \end{aligned}$$

In our example: $u_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, u_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, u_3 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$.

iii) In general: If $B = (u_1, u_2)$ is a basis of U and $A = S_B = \begin{pmatrix} u_1 & u_2 \end{pmatrix}$, then $U^\perp = \left\{ x \in \mathbb{R}^3 \mid x \cdot u_1 = x \cdot u_2 = 0 \right\}$

$$\begin{aligned} &= \left\{ x \in \mathbb{R}^3 \mid \underbrace{\begin{pmatrix} T \\ S_B \end{pmatrix} x}_{= 0} = 0 \right\} \\ &= \ker S_B^T. \end{aligned}$$

\curvearrowleft This will always have a solution of the form $t \cdot w$ for $t \in \mathbb{R}$ and $w \in \mathbb{R}^3$.

In our example: $S_B = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix}, S_B^T = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$

$$S_B^T x = 0 \rightsquigarrow \begin{pmatrix} 1 & 0 & 0 & | & 0 \\ 0 & 1 & 0 & | & 0 \end{pmatrix} \rightsquigarrow \text{Solutions are } x = t \cdot \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \stackrel{=} w$$

$$\Rightarrow U^\perp = \text{span}\left\{\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}\right\}. \text{ Basis: } \left(\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}\right).$$

iv) In general: If (u_1, u_2) is a basis of U and
 (w) a basis of U^\perp , then
 (u_1, u_2, w) is a basis of \mathbb{R}^3 .

In our example: $u_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$, $u_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$, $w = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$.

v) In general: • If (u_1, u_2) is a basis of U , then we can
 define $G: \mathbb{R}^3 \rightarrow \mathbb{R}^3$
 $x \mapsto \begin{pmatrix} 1 & u_1 & 0 \\ 0 & u_2 & 0 \\ 0 & 0 & 0 \end{pmatrix}$.

Clearly $\text{im } G = \text{span}\{u_1, u_2\} = U$.

• Let (u_1, u_2) be a basis of U and $w \in \mathbb{R}^3$ such
 that (u_1, u_2, w) is a basis of \mathbb{R}^3 (for example the
 one from iv).

Then we would like to find a linear map

$H: \mathbb{R}^3 \rightarrow \mathbb{R}^3$ with $[H]_B = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$, because

this would mean that for $x = \lambda_1 \cdot u_1 + \lambda_2 \cdot u_2 + \lambda_3 w$

we have $H(x) = \lambda_3 w$.

By definition/proposition in the Lecture (see Tutorial 8)

we have $[H]_B = S_B^{-1} [H]_{S_B} S_B$ where

$S_B = \begin{pmatrix} 1 & u_1 & w \\ 0 & u_2 & 0 \\ 0 & 0 & 1 \end{pmatrix}$. Therefore we need to

calculate $[H] = \underbrace{S_B}_{\begin{pmatrix} 1 & u_1 & w \\ 0 & u_2 & 0 \\ 0 & 0 & 1 \end{pmatrix}} \underbrace{[H]_B}_{\begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}} S_B^{-1} = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & w \end{pmatrix} S_B^{-1}$.

This will give a linear map $H: \mathbb{R}^3 \rightarrow \mathbb{R}^3$ with $\text{Ker } H = U$.

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In our example: • $G: \mathbb{R}^3 \rightarrow \mathbb{R}^3$
 $x \mapsto \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} x$

$$\text{im } G = U.$$

• $S_B = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ i.e. $S_B^{-1} = S_B = I_3$.

$$[H] = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, I_3 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

The map $H: \mathbb{R}^3 \rightarrow \mathbb{R}^3$ satisfies $\text{Ker } H = U$.
 $x \mapsto \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$

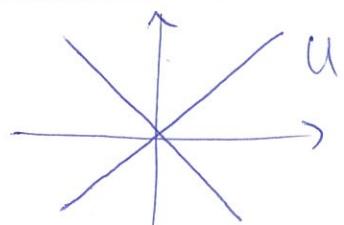
Exercise 3

Recall the definition of subspace:

$U \subset \mathbb{R}^2$ is a subspace if

- 1) $0 \in U$
- 2) $u, v \in U \Rightarrow u+v \in U$
- 3) $\lambda \in \mathbb{R}, u \in U \Rightarrow \lambda \cdot u \in U$.

Conditions			Example
1)	2)	3)	
X	X		$U = \left\{ \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \in \mathbb{R}^2 \mid x_1, x_2 \in \mathbb{Z} \right\}$ We have $0 \in U$ and $u+v \in U$ if $u, v \in U$, but for example $(1) \in U$ but $\frac{1}{2} \cdot (1) \notin U$.
X	X		$U = \{3\}$ does not contain 0 , but 2) and 3) are satisfied.
X	X		$U = \text{span}\{(1)\} \cup \text{span}\{(-1)\}$. We have $0 \in U$ and $\lambda u \in U$ for $u \in U$, but $(1) + (-1) = (0) \notin U$.



Exercise 4

i) To determine a basis of $\text{im}(F)$ we consider the rref of $[F]$

$$[F] = \begin{pmatrix} 1 & 1 & 1 \\ 1 & -1 & -5 \\ 0 & 0 & -2 \end{pmatrix} \xrightarrow{\frac{1}{2}\cdot\frac{1}{2}} \begin{pmatrix} 1 & 1 & 1 \\ 0 & -2 & -6 \\ 0 & -1 & -3 \end{pmatrix} \xrightarrow[-1]{\begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 3 \\ 0 & 0 & 0 \end{pmatrix}} \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 3 \\ 0 & 0 & 0 \end{pmatrix}$$

↑
pivot positions.

$B = (b_1, b_2)$ with $b_1 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ and $b_2 = \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}$ is a basis of $\text{im}(F)$.

ii) We need to find λ_1, λ_2 with $x = \lambda_1 b_1 + \lambda_2 b_2$:

$$\begin{pmatrix} 1 & 1 & 3 \\ 1 & -1 & t \\ 0 & 0 & t \end{pmatrix} \xrightarrow[\frac{1}{2}]{\begin{pmatrix} 1 & 1 & 3 \\ 0 & -2 & -2 \\ 0 & -1 & t-3 \end{pmatrix}} \xrightarrow[-1]{\begin{pmatrix} 1 & 1 & 3 \\ 0 & 1 & 1 \\ 0 & 0 & t-2 \end{pmatrix}} \sim \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & t-2 \end{pmatrix}.$$

We get a solution for $t=2$ given by $\lambda_1=2$ and $\lambda_2=1$.

$$\stackrel{t=2}{=} [x]_B = \begin{pmatrix} 2 \\ 1 \end{pmatrix}. \quad \left(\begin{pmatrix} 3 \\ 1 \\ 2 \end{pmatrix} = 2 \cdot \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} + 1 \cdot \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} \right).$$

iii) We use the Gram-Schmidt algorithm to get an ONB $E = (f_1, f_2)$:

$$f_1 = \frac{1}{\|b_1\|} b_1 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}.$$

$$w_2 = b_2 - (f_1 \cdot b_2) f_1 = \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} - \frac{1}{3} \cdot 0 \cdot \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}$$

(b_2 and b_1 are already orthogonal).

$$f_2 = \frac{1}{\|w_2\|} w_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}.$$

$E = \left(\frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} \right)$ is an ONB for $\text{im } F$.

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Exercise 5

i) For example $A = \begin{pmatrix} 1 & 1 \\ 2 & 3 \\ 1 & 1 \end{pmatrix}$ and $b = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$.

In this case $Ax=b$ has no solutions, since

$$(A|b) \xrightarrow{\text{Row Operations}} \left(\begin{array}{ccc|c} 1 & 1 & 1 \\ 2 & 3 & 3 \\ 1 & 1 & 3 \end{array} \right) \sim \left(\begin{array}{ccc|c} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 2 \end{array} \right).$$

ii) In general: We saw that we can minimize $\|Ax-b\|$ by choosing $AX = P_{\text{im } A}(b)$, i.e. the orthogonal projection of b onto $\text{im } A$ should be Ax . This can be done by solving $A^T A x = A^T b$. (see last Lecture).

In our example:

$$A^T A = \begin{pmatrix} 1 & 2 & 1 \\ 1 & 3 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 2 & 3 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 6 & 8 \\ 8 & 11 \end{pmatrix}$$

$$A^T b = \begin{pmatrix} 1 & 2 & 1 \\ 1 & 3 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 8 \\ 10 \end{pmatrix}.$$

Want to solve $\begin{pmatrix} 6 & 8 \\ 8 & 11 \end{pmatrix} x = \begin{pmatrix} 8 \\ 10 \end{pmatrix}$

$$\xrightarrow{\text{①}} \left(\begin{array}{cc|c} 6 & 8 & 8 \\ 8 & 11 & 10 \end{array} \right) \sim \xrightarrow{\text{③}} \left(\begin{array}{cc|c} 6 & 8 & 8 \\ 2 & 3 & 2 \end{array} \right) \sim \xrightarrow{\text{⑤}} \left(\begin{array}{cc|c} 0 & -1 & 2 \\ 2 & 3 & 2 \end{array} \right)$$

$$\sim \xrightarrow{-\frac{3}{2}\text{①}} \left(\begin{array}{cc|c} 1 & \frac{3}{2} & 1 \\ 0 & -1 & 2 \end{array} \right) \sim \left(\begin{array}{cc|c} 1 & 0 & 4 \\ 0 & 1 & -2 \end{array} \right)$$

$\Rightarrow x = \begin{pmatrix} 4 \\ -2 \end{pmatrix}$ minimizes $\|Ax-b\|$.

$$\left(Ax = \begin{pmatrix} 1 & 1 \\ 2 & 3 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 4 \\ -2 \end{pmatrix} = \begin{pmatrix} 2 \\ 2 \\ 2 \end{pmatrix} \sim \|Ax-b\| = \left\| \begin{pmatrix} 0 \\ -1 \\ 1 \end{pmatrix} \right\| = \sqrt{2} \right)$$