## Answers/Solutions of Exercise 4 (Version: August 18, 2014)

**Remark:** Please note that bases for vector spaces are not unique. In the following, if a question asks for a basis, the answer given is only one of the possible answers.

- 1. In order to answer (iv), we obtain the reduced row-echelon form of each of the matrices. (To answer (i)-(iii), we only need a row-echelon form.)
  - (a) Performing Gauss-Jordan elimination, we obtain the reduced row-echelon form of A:

$$\begin{pmatrix} 1 & 0 & 0 & 1 & -\frac{2}{7} \\ 0 & 1 & 0 & 1 & \frac{4}{7} \\ 0 & 0 & 1 & -1 & \frac{13}{7} \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}.$$

(i)  $\{(1,0,0,1,-\frac{2}{7}), (0,1,0,1,\frac{4}{7})), (0,0,1,-1,\frac{13}{7})\}$  is a basis for the row space.

 $\{(1,2,-1,1)^{\mathrm{T}}, (4,1,3,-1)^{\mathrm{T}}, (0,0,0,1)^{\mathrm{T}}\}\$  is a basis for the column space.

- (ii)  $\{(1,0,0,1,-\frac{2}{7}), (0,1,0,1,\frac{4}{7})), (0,0,1,-1,\frac{13}{7}), (0,0,0,1,0), (0,0,0,0,1)\}$  is a basis for  $\mathbb{R}^5$ .
- (iii)  $\begin{pmatrix} 1 & 2 & -1 & 1 \\ 4 & 1 & 3 & -1 \\ 0 & 0 & 0 & 1 \end{pmatrix}$  Gaussian  $\begin{pmatrix} 1 & 2 & -1 & 1 \\ 0 & -7 & 7 & -6 \\ 0 & 0 & 0 & 1 \end{pmatrix}$ . So  $\{(1, 2, -1, 1)^{\mathrm{T}}, (4, 1, 3, -1)^{\mathrm{T}}, (0, 0, 0, 1)^{\mathrm{T}}, (0, 0, 1, 0)^{\mathrm{T}}\}$  is a basis for  $\mathbb{R}^4$ .
- (iv)  $\{(-1,-1,1,1,0)^{\mathrm{T}}, (\frac{2}{7},-\frac{4}{7},-\frac{13}{7},0,1)^{\mathrm{T}}\}$  is a basis for the nullspace.
- (v)  $rank(\mathbf{A}) = 3$  and  $rank(\mathbf{A}) = 2$ . Hence  $rank(\mathbf{A}) + rank(\mathbf{A}) = 3 + 2 = 5 =$ the number of column in  $\mathbf{A}$ .
- (vi) No.
- (b) Performing Gauss-Jordan elimination, we obtain the reduced row-echelon form of  $\boldsymbol{B}$ :

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}.$$

- (i)  $\{1,0,0\}$ , (0,1,0), (0,0,1)} is a basis for the row space.  $\{(1,0,-1,2,3)^{\mathrm{T}}, (2,1,3,1,1)^{\mathrm{T}}, (0,1,6,0,-1)^{\mathrm{T}}\}$  is a basis for the column space.
- (ii)  $\{1,0,0\}, (0,1,0), (0,0,1)\}$  is already a basis for  $\mathbb{R}^3$ .

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(iii) 
$$\begin{pmatrix} 1 & 0 & -1 & 2 & 3 \\ 2 & 1 & 3 & 1 & 1 \\ 0 & 1 & 6 & 0 & -1 \end{pmatrix}$$
 Gaussian  $\begin{pmatrix} 1 & 0 & -1 & 2 & 3 \\ 0 & 1 & 5 & -3 & -5 \\ 0 & 0 & 1 & 3 & 4 \end{pmatrix}$ .  
So  $\{(1,0,-1,2,3)^{\mathrm{T}}, (2,1,3,1,1)^{\mathrm{T}}, (0,1,6,0,-1)^{\mathrm{T}}, (0,0,0,1,0)^{\mathrm{T}}, (0,0,0,0,1)^{\mathrm{T}}\}$  is a basis for  $\mathbb{R}^5$ .

- (iv)  $\emptyset$  is the basis for the nullspace.
- (v)  $rank(\mathbf{B}) = 3$  and  $rank(\mathbf{B}) = 0$ . Hence  $rank(\mathbf{B}) + rullity(\mathbf{B}) = 3 + 0 = 3 = the number of column in <math>\mathbf{B}$ .
- (vi) Yes.
- (c) Performing Gauss-Jordan elimination, we obtain the reduced row-echelon form of  $\boldsymbol{C}$ :

- (i)  $\{(1, \frac{1}{2}, 0, \frac{5}{6}, \frac{1}{3}), (0, 0, 1, -\frac{1}{6}, \frac{1}{3})\}$  is a basis for the row space.  $\{(2, 4, 2, 6)^{\mathrm{T}}, (4, 2, -2, 6)^{\mathrm{T}}\}$  is a basis for the column space.
- (ii)  $\{(1,\frac{1}{2},0,\frac{5}{6},\frac{1}{3}), (0,0,1,-\frac{1}{6},\frac{1}{3}), (0,1,0,0,0), (0,0,0,1,0), (0,0,0,0,1)\}$  is a basis for  $\mathbb{R}^5$ .

(iii) 
$$\begin{pmatrix} 2 & 4 & 2 & 6 \\ 4 & 2 & -2 & 6 \end{pmatrix} \xrightarrow{\text{Gaussian}} \begin{pmatrix} 2 & 4 & 2 & 6 \\ 0 & -6 & -6 & -6 \end{pmatrix}$$
.

So  $\{(2,4,2,6)^{\mathrm{T}}, (4,2,-2,6)^{\mathrm{T}}, (0,0,1,0)^{\mathrm{T}}, (0,0,0,1)^{\mathrm{T}}\}\$  is a basis for  $\mathbb{R}^4$ .

- (iv)  $\{(-\frac{1}{2},1,0,0,0)^{\mathrm{T}}, (-\frac{5}{6},0,\frac{1}{6},1,0)^{\mathrm{T}}, (-\frac{1}{3},0,-\frac{1}{3},0,1)^{\mathrm{T}}\}$  is the basis for the nullspace.
- (v)  $\operatorname{rank}(\mathbf{C}) = 2$  and  $\operatorname{nullity}(\mathbf{C}) = 3$ . Hence  $\operatorname{rank}(\mathbf{C}) + \operatorname{nullity}(\mathbf{C}) = 2 + 3 = 5 =$ the number of column in  $\mathbf{C}$ .
- (vi) No.
- (d) Performing Gauss-Jordan elimination, we obtain the reduced row-echelon form of D:

$$\begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}.$$

- (i)  $\{(1,0,1,0), (0,1,1,0), (0,0,0,1)\}$  is a basis for the row space.  $\{(1,-1,2)^{\mathrm{T}}, (4,4,0)^{\mathrm{T}}, (8,0,1)^{\mathrm{T}}\}$  is a basis for the column space.
- (ii)  $\{(1,0,1,0), (0,1,1,0), (0,0,0,1), (0,0,1,0)\}$  is a basis for  $\mathbb{R}^4$ .

(iii) 
$$\{(1,-1,2)^{\mathrm{T}}, (4,4,0)^{\mathrm{T}}, (8,0,1)^{\mathrm{T}}\}\$$
 is already a basis for  $\mathbb{R}^3$ .

- (iv)  $\{(-1,-1,1,0)^{\mathrm{T}}\}$  is a basis for the nullspace.
- (v)  $rank(\mathbf{D}) = 3$  and  $rank(\mathbf{D}) = 1$ . Hence  $rank(\mathbf{D}) + rullity(\mathbf{D}) = 3 + 1 = 4 = the$  number of column in  $\mathbf{D}$ .
- (vi) Yes.

2. (a) 
$$\begin{pmatrix} 1 & -2 & 0 & 0 & 3 \\ 2 & -5 & -3 & -2 & 6 \\ 0 & 5 & 15 & 10 & 0 \\ 2 & 1 & 15 & 8 & 6 \end{pmatrix}$$
Gaussian 
$$\begin{pmatrix} 1 & -2 & 0 & 0 & 3 \\ 0 & -1 & -3 & -2 & 0 \\ 0 & 0 & 0 & -2 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$
. 
$$\{(1, -2, 0, 0, 3), (0, -1, -3, -2, 0), (0, 0, 0, -2, 0)\}$$
is a basis for  $W$ .

- (b)  $\dim(W) = 3$
- (c)  $\{(1, -2, 0, 0, 3), (0, -1, -3, -2, 0), (0, 0, 0, -2, 0), (0, 0, 1, 0, 0), (0, 0, 0, 0, 1)\}$  is a basis for  $\mathbb{R}^5$ .

So  $S' = \{(1,0,1,3), (2,-1,0,1)\}$  is a basis for V.

So  $S' = \{(1, 0, 1, 3, 4), (2, 1, -2, 1, 0), (0, 5, 2, 1, 1)\}$  is a basis for V.

## 4. Since

$$(a+b+3c+3d, b+2c+d, a+c+2d, -a-b-3c-3d, a+c+2d)$$
  
=  $a(1,0,1,-1,1) + b(1,1,0,-1,0) + c(3,2,1,-3,1) + d(3,1,2,-3,2),$ 

 $V = \text{span}\{(1,0,1,-1,1), (1,1,0,-1,0), (3,2,1,-3,1), (3,1,2,-3,2)\}.$  By

So  $\{(1,0,1,-1,1), (0,1,-1,0,-1)\}$  is a basis for V.

(b) (i) Note that  $\operatorname{span}(S)$  and  $\operatorname{span}(T)$  are the row spaces of  $\boldsymbol{B}$  and  $\boldsymbol{R}$  respectively. Since  $\boldsymbol{B}$  and  $\boldsymbol{R}$  are row equivalent,  $\operatorname{span}(S) = \operatorname{span}(T)$ . Also,  $\operatorname{dim}(\operatorname{span}(T)) = \operatorname{rank}(R) = 3$ . So by Theorem 3.6.7, S is a basis for  $\operatorname{span}(T)$ .

(You do not need to really do any computations to claim the result on the RHS. Why?)

So the transition matrix from S to T is  $\begin{pmatrix} 1 & 1 & -1 \\ 0 & 1 & 0 \\ -1 & 2 & 2 \end{pmatrix}$ .

6. 
$$\begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & a & a & a & a \\ 1 & a & a^2 & a & a^2 \\ 1 & a^3 & a & 2a - a^3 & a \end{pmatrix} \xrightarrow{\text{Gaussian}} \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 0 & a - 1 & a - 1 & a - 1 & a - 1 \\ 0 & 0 & a^2 - a & 0 & a^2 - a \\ 0 & 0 & 0 & 2a - 2a^3 & 0 \end{pmatrix}$$

- If a = 1, then  $\{(1, 1, 1, 1, 1)\}$  is a basis for V and  $\dim(V) = 1$ .
- If a = 0, then  $\{(1, 1, 1, 1, 1), (0, 1, 1, 1, 1)\}$  is a basis for V and  $\dim(V) = 2$ .
- If a = -1, then  $\{(1, 1, 1, 1, 1), (0, -2, -2, -2, -2), (0, 0, 2, 0, 2)\}$  is a basis for V and  $\dim(V) = 3$ .
- If  $a \notin \{1, 0, -1\}$ , then  $\{(1, 1, 1, 1, 1), (0, a 1, a 1, a 1, a 1), (0, 0, a^2 a, 0, a^2 a), (0, 0, 0, 2a 2a^3, 0)\}$  is a basis for V and  $\dim(V) = 4$ .

7. 
$$V + W = \text{span}\{(1, 1, 0, 0), (-1, 0, 1, 0), (-1, 2, 3, 0), (2, -1, 2, -1)\}$$

$$\begin{pmatrix} 1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 2 & 3 & 0 \\ 2 & -1 & 2 & -1 \end{pmatrix} \xrightarrow{\text{Gaussian}} \begin{pmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 5 & -1 \\ 0 & 0 & 0 & 0 \end{pmatrix} = \mathbf{R}$$

 $\{(1,1,0,0), (0,1,1,0), (0,0,5,-1)\}$  is a basis for V+W.

8. (a) We can choose 
$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$
.

(b) We can choose 
$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & -1 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$
.

(c) V is the solution space of the linear equation  $2x_1 - x_2 - x_3 + 0x_4 = 0$ .

- 9. (a) Since  $(x_1, x_2, x_3, x_4)^{\mathrm{T}} = (t 2s, s + t, s, t)^{\mathrm{T}} = s(-2, 1, 1, 0)^{\mathrm{T}} + t(1, 1, 0, 1)^{\mathrm{T}}, \{(-2, 1, 1, 0)^{\mathrm{T}}, (1, 1, 0, 1)^{\mathrm{T}}\}$  is a basis for the nullspace of  $\boldsymbol{A}$ . The nullity of  $\boldsymbol{A}$  is 2.
  - (b) A general solution of  $\mathbf{A}\mathbf{x} = \mathbf{b}$  is  $x_1 = t 2s + 1$ ,  $x_2 = s + t$ ,  $x_3 = s 1$ ,  $x_4 = t$  where s, t are arbitrary parameters.
  - (c) The reduced row-echelon form of  $\mathbf{A}$  is  $\begin{pmatrix} 1 & 0 & 2 & -1 \\ 0 & 1 & -1 & -1 \\ 0 & 0 & 0 & 0 \end{pmatrix}$ .
  - (d)  $\{(1,0,2,-1), (0,1,-1,-1)\}$  is a basis for the row space of  $\boldsymbol{A}$ . The rank of  $\boldsymbol{A}$  is 2.
  - (e) No, we cannot find the column space of  $\boldsymbol{A}$  with the given information.
- 10. (a) Let R be the reduced row-echelon form of A. Since  $a_1, a_2, a_3$  are linearly independent, the first three columns of R are linearly independent. Thus

the first three columns of 
$$\mathbf{R}$$
 must be  $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$ . Together with the infor-

mation given for the fourth and fifth columns, 
$$\mathbf{R} = \begin{pmatrix} 1 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & -2 & 1 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$
.

- (b)  $\{(1,0,0,1,0), (0,1,0,-2,1), (0,0,1,1,1)\}$  is a basis for the row space of A; and  $\{a_1, a_2, a_3\}$  is a basis for the column space of A.
- 11. (a)  $\boldsymbol{x} = (2, -1, 3)$  is the solution to the linear system.

Thus 
$$\begin{pmatrix} 16\\13\\-4\\7 \end{pmatrix} = 2 \begin{pmatrix} 1\\2\\0\\1 \end{pmatrix} - \begin{pmatrix} 1\\3\\1\\1 \end{pmatrix} + 3 \begin{pmatrix} 5\\4\\-1\\2 \end{pmatrix}$$
.

(b)  $\boldsymbol{x} = (-3 + s + t, 13 - 3s - 2t, 1 - t, s, t)$ , where  $s, t \in \mathbb{R}$ , is a general solution for the linear system.

In particular, 
$$\begin{pmatrix} -1 \\ 9 \\ 4 \end{pmatrix} = -3 \begin{pmatrix} 1 \\ 1 \\ 3 \end{pmatrix} + 13 \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} + \begin{pmatrix} 2 \\ -1 \\ 0 \end{pmatrix} + 0 \begin{pmatrix} -1 \\ 2 \\ 0 \end{pmatrix} + 0 \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}.$$

- 12. (a) For example,  $\begin{pmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{pmatrix}$ .
  - (b) No. By Theorem 4.2.1, the dimensions of the row space and column space of a matrix must be the same.
  - (c) For example,  $\begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 3 & 6 & 9 \end{pmatrix}$ .
  - (d) For example,  $\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ .
- - (b) 4 2 (c) 3 0
- 14. (a) a = b = c = d = 0.
  - (b)  $ad bc \neq 0$ .
  - (c) ad bc = 0 but not all a, b, c, d are zero.
- 15. (a) If a = 1,  $rank(\mathbf{A}) = 1$ . If a = -2,  $rank(\mathbf{A}) = 2$ . If  $a \neq 1$  and  $a \neq -2$ ,  $rank(\mathbf{A}) = 3$ .
  - (b) If b = c = d = e = f = 0, rank $(\mathbf{B}) = 0$ . If either (i) b = c = 0 and not all d, e, f are zero or (ii) d = e = 0 and not all b, c, f are zero, rank $(\mathbf{B}) = 1$ .

If not all b, c are zero and not all d, e are zero, rank $(\mathbf{B}) = 2$ .

16. (a)  $\mathbf{X_1} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \xrightarrow{R_3 - R_1} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$ 

So rank $(X_1) = 2$  and nullity $(X_1) = 3 - 2 = 1$ .

So rank $(\mathbf{X_2}) = 3$  and nullity $(\mathbf{X_2}) = 5 - 3 = 2$ .

(b) 
$$X_n \xrightarrow{R_{2n+1} - R_1} R_{2n} - R_2 \xrightarrow{R_{n+2} - R_n} \begin{pmatrix} 1 & 0 & \cdot & \cdot & \cdot & \cdot & 0 & 1 \\ 0 & 1 & 0 & \cdot & \cdot & \cdot & 0 & 1 & 0 \\ & \cdot & \cdot & & \cdot & & \cdot & & \\ & & \cdot & & \cdot & & \cdot & & \\ 0 & \cdot & \cdot & 0 & 1 & 0 & \cdot & \cdot & 0 \\ 0 & \cdot & \cdot & 0 & 0 & 0 & \cdot & \cdot & 0 \\ & & \cdot & & & \cdot & & \\ & & \cdot & & & & \cdot & & \\ 0 & \cdot & \cdot & 0 & 0 & 0 & \cdot & \cdot & 0 \end{pmatrix}$$

So rank $(\boldsymbol{X_n}) = n+1$  and nullity $(\boldsymbol{X_n}) = (2n+1) - (n+1) = n$ .

17. When the rank is 0, the solution set is the entire  $\mathbb{R}^3$ .

When the rank is 1, the solution set is a plane in  $\mathbb{R}^3$  that passes through the origin.

When the rank is 2, the solution set is a line in  $\mathbb{R}^3$  that passes through the origin.

When the rank is 3, the solution set is  $\{0\}$ .

18. Let  $\mathbf{A} = (\mathbf{a_1} \ \mathbf{a_2} \ \cdots \ \mathbf{a_n})$  and  $\mathbf{B}$  be  $m \times n$  matrices where  $\mathbf{a_i}$  is the *i*th column of  $\mathbf{A}$ . Suppose  $\mathbf{A}$  and  $\mathbf{B}$  are row equivalent, i.e. there exists elementary matrices  $\mathbf{E_1}, \mathbf{E_2}, \ldots, \mathbf{E_k}$  such that

$$B = E_k \cdots E_2 E_1 A$$
.

Define  $P = E_k \cdots E_2 E_1 A$ . Then  $B = PA = (Pa_1 \ Pa_2 \cdots \ Pa_n)$  where  $Pa_i$  is the *i*th column of B. By Theorem 2.4.7, P is invertible.

Let  $S_1 = \{a_{i_1}, a_{i_2}, \ldots, a_{i_r}\}$  be a set of columns of A. Note that  $S_2 = \{Pa_{i_1}, Pa_{i_2}, \ldots, Pa_{i_r}\}$  is the set of corresponding columns of B.

- (a) Since P is invertible, by Question 3.30,  $S_1$  is linearly independent if and only if  $S_2$  is linearly independent.
- (b) Suppose  $S_1$  is a basis for the column space of  $\mathbf{A}$ . We want to show that  $S_2$  is a basis for the column space of  $\mathbf{B}$ :

- (i) By (a),  $S_2$  is linearly independent.
- (ii) It is obvious that  $\operatorname{span}(S_2) \subseteq \operatorname{the column space of } \boldsymbol{B}$ .

Take any  $\mathbf{u} \in \text{the column space of } \mathbf{B}, \text{ i.e. for some } c_1, c_2, \dots, c_n \in \mathbb{R},$ 

$$\boldsymbol{u} = c_1 \boldsymbol{P} \boldsymbol{a}_1 + c_2 \boldsymbol{P} \boldsymbol{a}_2 + \dots + c_n \boldsymbol{P} \boldsymbol{a}_n.$$

Since span( $S_1$ ) = the column space of  $\mathbf{A}$ ,

$$a_1, a_2, \dots, a_n \in \text{span}(S_1) = \text{span}\{a_{i_1}, a_{i_2}, \dots, a_{i_r}\}$$

and hence

$$Pa_1, Pa_2, \dots, Pa_n \in \text{span}\{Pa_{i_1}, Pa_{i_2}, \dots, Pa_{i_r}\} = \text{span}(S_2).$$

By Theorem 3.2.9.2,  $\mathbf{u} \in \text{span}(S_2)$ . So the column space of  $\mathbf{B} \subseteq \text{span}(S_2)$ .

We have shown that  $\operatorname{span}(S_2) = \operatorname{the column space of } \boldsymbol{B}$ .

By (i) and (ii),  $S_2$  is a basis for the column space of  $\mathbf{B}$ .

Similarly, follow the arguments above by replacing  $a_i$  by  $Pa_i$  and P by  $P^{-1}$ . We conclude that if  $S_2$  is a basis for the column space of B, then  $S_1$  is a basis for the column space of A.

19. (a) 
$$\begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 & 1 \end{pmatrix}$$
 Gauss-Jordan  $\begin{pmatrix} 1 & 0 & 0 & 0 & 1 & 0 & -1 \\ 0 & 1 & 1 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 1 & -1 & 1 & 1 \end{pmatrix}$  Elimination

(i) 
$$\begin{pmatrix} w \\ x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 1 \\ t \\ 1 - t \\ -1 \end{pmatrix}$$
 where  $t \in \mathbb{R}$ .

(ii) 
$$\begin{pmatrix} w \\ x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ t \\ -1 - t \\ 1 \end{pmatrix}$$
 where  $t \in \mathbb{R}$ .

(iii) 
$$\begin{pmatrix} w \\ x \\ y \\ z \end{pmatrix} = \begin{pmatrix} -1 \\ t \\ -t \\ 1 \end{pmatrix}$$
 where  $t \in \mathbb{R}$ .

If  $u_1$  is a solution of (i),  $u_2$  a solution of (ii) and  $u_3$  a solution of (iii), then  $(u_1 \ u_2 \ u_3)$  is a right inverse of B. The answer is certainly not unique.

For example, 
$$\begin{pmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ 1 & -1 & 0 \\ -1 & 1 & 1 \end{pmatrix}$$
 is a right inverse of  $\boldsymbol{B}$ .

- (b) For example,  $\boldsymbol{B} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$  has no right inverse.
- (c) Let  $\{e_1, e_2, \dots, e_m\}$  be the standard basis for  $\mathbb{R}^m$ .

 $\boldsymbol{B}$  has a right inverse

$$\Leftrightarrow$$
  $m{B}ig(m{u_1} \ m{u_2} \ \cdots \ m{u_m}ig) = ig(m{e_1} \ m{e_2} \ \cdots \ m{e_m}ig) ext{ for some } m{u_1}, m{u_2}, \dots, m{u_m} \in \mathbb{R}^n$ 

$$\Leftrightarrow$$
 All linear systems  $Bx = e_1, Bx = e_2, \ldots, Bx = e_m$  are consistent.

$$\Leftrightarrow$$
  $e_1, e_2, \dots, e_m \in$  the column space of  $B$ 

$$\Leftrightarrow$$
 the column space of  $\boldsymbol{B} = \mathbb{R}^m$ 

$$\Leftrightarrow$$
 dim(the column space of  $\mathbf{B}$ ) =  $m$ 

$$\Leftrightarrow \operatorname{rank}(\boldsymbol{B}) = m.$$

20. Let  $\mathbf{B} = (\mathbf{b_1} \cdots \mathbf{b_n})$  where  $b_j$  is the jth column of  $\mathbf{B}$ .

$$AB = 0 \Rightarrow (Ab_1 \cdots Ab_n) = 0 \Rightarrow Ab_i = 0 \text{ for all } j$$

i.e.  $b_1, \ldots, b_n$  are contained in the nullspace of A.

So the column space of  $B = \operatorname{span}\{b_1, \ldots, b_n\} \subseteq \operatorname{the nullspace}$  of A.

21. Let  $\mathbf{A} = \begin{pmatrix} \mathbf{a_1} \\ \vdots \\ \mathbf{a_n} \end{pmatrix}$  be a matrix where  $\mathbf{a_i}$  is the *i*th row of  $\mathbf{A}$ .

Let  $\boldsymbol{u} = (u_1, \dots, u_n) \in \mathbb{R}^n$  such that  $\boldsymbol{u}^{\mathrm{T}}$  is a vector in the nullspace of  $\boldsymbol{A}$ . Then

$$m{A}m{u}^{ ext{T}} = m{0} \quad \Rightarrow \quad egin{pmatrix} m{a}_1m{u}^{ ext{T}} \ m{a}_nm{u}^{ ext{T}} \end{pmatrix} = egin{pmatrix} 0 \ dots \ 0 \end{pmatrix} \quad \Rightarrow \quad m{a}_im{u}^{ ext{T}} = 0 \ ext{for all } i.$$

Assume that  $\boldsymbol{u}$  is also contained in the row space of  $\boldsymbol{A}$ , i.e.  $\boldsymbol{u} = c_1 \boldsymbol{a_1} + \cdots + c_n \boldsymbol{a_n}$  for some  $c_1, \ldots, c_n \in \mathbb{R}$ . We have

$$\boldsymbol{u}\boldsymbol{u}^{\mathrm{T}} = c_1\boldsymbol{a_1}\boldsymbol{u}^{\mathrm{T}} + \dots + c_n\boldsymbol{a_n}\boldsymbol{u}^{\mathrm{T}} = 0.$$

On the other hand,  $\boldsymbol{u}\boldsymbol{u}^{\mathrm{T}}=u_1^2+\cdots+u_n^2$ . So  $u_1^2+\cdots+u_n^2=0$  which implies  $u_1=0,\ldots,u_n=0$ , i.e.  $\boldsymbol{u}$  is the zero vector.

22. (a) Since P is invertible, we can write  $P = E_n \cdots E_1$  where  $E_i$  are elementary matrices. So  $PA = E_n \cdots E_1 A$  and A are row-equivalent matrices. They have the same row space. Thus

$$rank(\mathbf{P}\mathbf{A}) = dim(the row space of \mathbf{P}\mathbf{A})$$
  
=  $dim(the row space of \mathbf{A}) = rank(\mathbf{A}).$ 

- (b) For example,  $\mathbf{P} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$  and  $\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ .
- (c) No. For example, let  $\mathbf{P} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$  and  $\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$ .
- 23. Let  $A = (a_1 \cdots a_n)$  and  $B = (b_1 \cdots b_n)$  where  $a_j$  is the jth column of A and  $b_j$  is the jth column of B. Let  $\{a'_1, \ldots, a'_r\}$  be a basis for the column space of A and let  $\{b'_1, \ldots, b'_s\}$  be a basis for the column space of B. Then

the column space of 
$$A + B = \operatorname{span}\{a_1 + b_1, \ldots, a_n + b_n\}$$
  
 $\subseteq \operatorname{span}\{a'_1, \ldots, a'_r, b'_1, \ldots, b'_s\}.$ 

So

$$rank(\mathbf{A} + \mathbf{B}) = dim(the column space of \mathbf{A} + \mathbf{B}) \le r + s = rank(\mathbf{A}) + rank(\mathbf{B}).$$

24. Since  $\mathbf{A}\mathbf{x} = \mathbf{b}$  is consistent for all  $\mathbf{b} \in \mathbb{R}^m$ , by Theorem 4.1.16, the column space of  $\mathbf{A}$  is  $\mathbb{R}^m$ , i.e.  $\operatorname{rank}(\mathbf{A}) = m$ . Hence

$$\operatorname{nullity}(\boldsymbol{A}^{\mathrm{T}}) = m - \operatorname{rank}(\boldsymbol{A}^{\mathrm{T}}) = m - \operatorname{rank}(\boldsymbol{A}) = 0.$$

It means that the linear system  $\boldsymbol{A}^{\scriptscriptstyle \mathrm{T}} \boldsymbol{y} = \boldsymbol{0}$  has only the trivial solution.

Alternative Solution: Let  $e_1, \ldots, e_m$  be the standard basis for  $\mathbb{R}^m$  and let  $u_1, \ldots, u_m$  be vectors in  $\mathbb{R}^n$  such that  $Au_i = e_i$  for each i. (In here, all the vectors are column vectors.) Suppose  $v = (v_1, \ldots, v_m)^T$  is a solution to the system  $A^Tv = 0$ . Then for  $i = 1, \ldots, m$ ,

$$v_i = \boldsymbol{e_i}^{\mathrm{T}} \boldsymbol{v} = (\boldsymbol{A} \boldsymbol{u_i})^{\mathrm{T}} \boldsymbol{v} = \boldsymbol{u_i}^{\mathrm{T}} \boldsymbol{A}^{\mathrm{T}} \boldsymbol{v} = \boldsymbol{u_i} \boldsymbol{0} = 0.$$

So v = 0. That is, the system  $A^{T}y = 0$  has only the trivial solution.

25. (a) Let  $\boldsymbol{u}$  be any vector in the nullspace of  $\boldsymbol{A}$ , i.e.  $\boldsymbol{A}\boldsymbol{u}=\boldsymbol{0}$ . Then  $\boldsymbol{A}^{\mathrm{T}}\boldsymbol{A}\boldsymbol{u}=\boldsymbol{A}^{\mathrm{T}}\boldsymbol{0}=\boldsymbol{0}$ . So  $\boldsymbol{u}$  is also a vector in the nullspace of  $\boldsymbol{A}^{\mathrm{T}}\boldsymbol{A}$ . We have shown that the nullspace of  $\boldsymbol{A}$  is a subspace of the nullspace of  $\boldsymbol{A}^{\mathrm{T}}\boldsymbol{A}$ .

Let  $\boldsymbol{v}$  be any vector in the nullspace of  $\boldsymbol{A}^{\mathrm{T}}\boldsymbol{A}$ , i.e.  $\boldsymbol{A}^{\mathrm{T}}\boldsymbol{A}\boldsymbol{v}=\boldsymbol{0}$ . Suppose  $\boldsymbol{A}\boldsymbol{v}=(b_1,b_2,\ldots,b_m)^{\mathrm{T}}$ . Then

$$(\mathbf{A}\mathbf{v})^{\mathrm{T}}(\mathbf{A}\mathbf{v}) = \mathbf{v}^{\mathrm{T}}\mathbf{A}^{\mathrm{T}}\mathbf{A}\mathbf{v} = \mathbf{v}^{\mathrm{T}}\mathbf{0} = \mathbf{0}$$

$$\Rightarrow b_{1}^{2} + b_{2}^{2} + \dots + b_{m}^{2} = 0$$

$$\Rightarrow b_{1} = b_{2} = \dots = b_{m} = 0.$$

That is, Av = 0. So v is also a vector in the null space of A.

We have shown that the nullspace of  $A^{T}A$  is a subspace of the nullspace of A.

Hence the nullspace of A is equal to the nullspace of  $A^{T}A$ .

(b) By (a),  $\operatorname{nullity}(\boldsymbol{A}) = \operatorname{nullity}(\boldsymbol{A}^{\mathsf{T}}\boldsymbol{A})$ .

Since  $\boldsymbol{A}$  is an  $m \times n$  matrix,  $\boldsymbol{A}^{\mathrm{T}} \boldsymbol{A}$  is an  $n \times n$  matrix. By the Dimension Theorem for Matrices (Theorem 4.3.4),

$$rank(\mathbf{A}) = n - nullity(\mathbf{A}) = n - nullity(\mathbf{A}^{\mathsf{T}}\mathbf{A}) = rank(\mathbf{A}^{\mathsf{T}}\mathbf{A}).$$

- (c) No. For example, let  $A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$ .
- (d) Yes. By (b) and Remark 4.2.5.3,  $\operatorname{rank}(\boldsymbol{A}) = \operatorname{rank}(\boldsymbol{A}^{\mathrm{T}}) = \operatorname{rank}((\boldsymbol{A}^{\mathrm{T}})^{\mathrm{T}}\boldsymbol{A}^{\mathrm{T}}) = \operatorname{rank}((\boldsymbol{A}\boldsymbol{A}^{\mathrm{T}}).$
- 26. (a) False. For example, let  $\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$  and  $\mathbf{B} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$ .
  - (b) True. By Theorem 4.1.7, the row space of  $\boldsymbol{A}$  and the trow space of  $\boldsymbol{B}$  are the same. Hence the column space of  $\boldsymbol{A}^{\mathrm{T}}$  and the column space of  $\boldsymbol{B}^{\mathrm{T}}$  are the same.
  - (c) False. For example, let  $\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$  and  $\mathbf{B} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$ .
  - (d) False. For example, let  $\mathbf{A} = \mathbf{B} = \mathbf{I}_2$ .
  - (e) False. For example, let  $\mathbf{A} = \mathbf{B} = \mathbf{0}_{2\times 2}$ .
  - (f) False. For example, let  $\mathbf{A} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$  and  $\mathbf{B} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$ .
  - (g) False. For example, let  $\mathbf{A} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$  and  $\mathbf{B} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$ .