

# Linear Algebra Practice Midterm 1

## Spring 2019

1. Let

$$A = \begin{pmatrix} -2 & 3 & -3 & -1 \\ 4 & 1 & 13 & -5 \end{pmatrix}$$

and consider the homogeneous system  $A\mathbf{x} = \mathbf{0}$ , where  $\mathbf{x} \in \mathbb{R}^4$  and  $\mathbf{0} \in \mathbb{R}^2$ .

(a) Compute  $\text{rref}(A \mid \mathbf{0})$ .

**Solution:**

$$\text{rref}(A \mid \mathbf{0}) = \left( \begin{array}{cccc|c} 1 & 0 & 3 & -1 & 0 \\ 0 & 1 & 1 & -1 & 0 \end{array} \right)$$

(b) Identify the pivot columns  $\mathbf{b}_j$  in  $B = \text{rref}(A \mid \mathbf{0})$ .

**Solution:** Let  $B = \text{rref}(A \mid \mathbf{0})$ . Then the pivot columns of  $B$  are

$$\mathbf{b}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad \text{and} \quad \mathbf{b}_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

- (c) Write the remaining, non-pivot, columns of  $B$  as linear combinations of the pivot columns you identified in part (b).

Solution: The non-pivot columns (ignoring the last column of zeros) are  $\mathbf{b}_3$  and  $\mathbf{b}_4$ , and

$$\begin{aligned}\mathbf{b}_3 &= \begin{pmatrix} 3 \\ 1 \end{pmatrix} \\ &= 3 \begin{pmatrix} 1 \\ 0 \end{pmatrix} + 1 \begin{pmatrix} 0 \\ 1 \end{pmatrix} \\ &= 3\mathbf{b}_1 + 1\mathbf{b}_2\end{aligned}$$

and

$$\begin{aligned}\mathbf{b}_4 &= \begin{pmatrix} -1 \\ -1 \end{pmatrix} \\ &= (-1) \begin{pmatrix} 1 \\ 0 \end{pmatrix} + (-1) \begin{pmatrix} 0 \\ 1 \end{pmatrix} \\ &= (-1)\mathbf{b}_1 + (-1)\mathbf{b}_2\end{aligned}$$

- (d) Returning to the *original* augmented matrix  $(A|\mathbf{0})$ , show directly that the same dependency relations hold among the columns of  $A$  (e.g. if you found that  $\mathbf{b}_4 = 5\mathbf{b}_1 + 2\mathbf{b}_3$ , then you would demonstrate, starting with the RHS, that  $\mathbf{a}_4 = 5\mathbf{a}_1 + 2\mathbf{a}_3$ ).

Solution: For convenience, let's rewrite  $A$  and its columns here, so we could see what we're doing:

$$A = \begin{pmatrix} -2 & 3 & -3 & -1 \\ 4 & 1 & 13 & -5 \end{pmatrix},$$

so

$$\mathbf{a}_1 = \begin{pmatrix} -2 \\ 4 \end{pmatrix}, \quad \mathbf{a}_2 = \begin{pmatrix} 3 \\ 1 \end{pmatrix}, \quad \mathbf{a}_3 = \begin{pmatrix} -3 \\ 13 \end{pmatrix}, \quad \mathbf{a}_4 = \begin{pmatrix} -1 \\ -5 \end{pmatrix}$$

We start with the RHS of the proposed equation which we're trying to prove, the easier side, and perform the computation, hoping that we get what we're supposed to:

$$3\mathbf{a}_1 + 1\mathbf{a}_2 = 3 \begin{pmatrix} -2 \\ 4 \end{pmatrix} + 1 \begin{pmatrix} 3 \\ 1 \end{pmatrix} = \begin{pmatrix} -6+3 \\ 12+1 \end{pmatrix} = \begin{pmatrix} -3 \\ 13 \end{pmatrix} = \mathbf{a}_3 \quad \checkmark$$

and similarly

$$(-1)\mathbf{a}_1 + (-1)\mathbf{a}_2 = (-1) \begin{pmatrix} -2 \\ 4 \end{pmatrix} + (-1) \begin{pmatrix} 3 \\ 1 \end{pmatrix} = \begin{pmatrix} 2-3 \\ -4-1 \end{pmatrix} = \begin{pmatrix} -1 \\ -5 \end{pmatrix} = \mathbf{a}_4 \quad \checkmark$$

- (e) Parametrize the free variables, those corresponding to the non-pivot columns of  $B$ , and use the form of  $B$  to solve for the dependent variables in terms of the independent.

Solution: The pivot columns are 1 and 2, so that means the free variables are  $x_3$  and  $x_4$  in the system  $A\mathbf{x} = \mathbf{0}$ . Let  $x_3 = s$  and  $x_4 = t$ . We interpret

$$B = \text{rref}(A | \mathbf{0}) = \left( \begin{array}{cccc|c} 1 & 0 & 3 & -1 & 0 \\ 0 & 1 & 1 & -1 & 0 \end{array} \right)$$

as

$$\begin{aligned} x_1 + 3x_3 - x_4 &= 0 \\ x_2 + x_3 - x_4 &= 0 \end{aligned}$$

Solving for the dependent variables in terms of the independent gives

$$\begin{array}{lcl} x_1 = -3x_3 + x_4 & \text{and} & x_2 = -x_3 + x_4 \\ = -3s + t & & = -s + t \end{array}$$

- (f) Use the previous part to write *all possible* solutions, namely the null space  $N(A)$  of  $A$ , as a span of column vectors.

Solution: We simply plug in what we got in part (e): the general solution vector  $\mathbf{x} \in \mathbb{R}^4$  of  $A\mathbf{x} = \mathbf{0}$  is

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} -3s + t \\ -s + t \\ s \\ t \end{pmatrix} = s \begin{pmatrix} -3 \\ -1 \\ 1 \\ 0 \end{pmatrix} + t \begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \end{pmatrix}$$

That is,

$$N(A) = \text{span}(\mathbf{u}, \mathbf{v}), \quad \text{where } \mathbf{u} = \begin{pmatrix} -3 \\ -1 \\ 1 \\ 0 \end{pmatrix}, \quad \mathbf{v} = \begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \end{pmatrix}$$

- (g) Use the previous step to find two different solutions of  $A\mathbf{x} = \mathbf{0}$ . Plug them in to demonstrate that they are, in fact, solutions.

Solution: Well, the two vectors  $\mathbf{u}$  and  $\mathbf{v}$  in part (f) above are two solutions, but so as to get maybe some others, try e.g. letting  $s = -2$ ,  $t = 1$  in the expression for the solution vector  $\mathbf{x}$ , and give it a different name, say

$$\mathbf{w} = (-2)\mathbf{u} + 1\mathbf{v} = (-2) \begin{pmatrix} -3 \\ -1 \\ 1 \\ 0 \end{pmatrix} + 1 \begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 7 \\ 3 \\ -2 \\ 1 \end{pmatrix}$$

Let's try these out and make sure they're solution vectors:

$$\begin{aligned} A\mathbf{u} &= \begin{pmatrix} -2 & 3 & -3 & -1 \\ 4 & 1 & 13 & -5 \end{pmatrix} \begin{pmatrix} -3 \\ -1 \\ 1 \\ 0 \end{pmatrix} \\ &= -3 \begin{pmatrix} -2 \\ 4 \end{pmatrix} - \begin{pmatrix} 3 \\ 1 \end{pmatrix} + \begin{pmatrix} -3 \\ 13 \end{pmatrix} + 0 \begin{pmatrix} -1 \\ -5 \end{pmatrix} \\ &= \begin{pmatrix} 6 - 3 - 3 + 0 \\ -12 - 1 + 13 + 0 \end{pmatrix} \\ &= \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \checkmark \end{aligned}$$

and similarly

$$\begin{aligned} A\mathbf{v} &= \begin{pmatrix} -2 & 3 & -3 & -1 \\ 4 & 1 & 13 & -5 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \end{pmatrix} \\ &= 1 \begin{pmatrix} -2 \\ 4 \end{pmatrix} + 1 \begin{pmatrix} 3 \\ 1 \end{pmatrix} + 0 \begin{pmatrix} -3 \\ 13 \end{pmatrix} + 1 \begin{pmatrix} -1 \\ -5 \end{pmatrix} \\ &= \begin{pmatrix} -2 + 3 + 0 - 1 \\ 4 + 1 + 0 - 5 \end{pmatrix} \\ &= \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \checkmark \end{aligned}$$

Lastly,

$$\begin{aligned} A\mathbf{w} &= \begin{pmatrix} -2 & 3 & -3 & -1 \\ 4 & 1 & 13 & -5 \end{pmatrix} \begin{pmatrix} 7 \\ 3 \\ -2 \\ 1 \end{pmatrix} \\ &= 7 \begin{pmatrix} -2 \\ 4 \end{pmatrix} + 3 \begin{pmatrix} 3 \\ 1 \end{pmatrix} - 2 \begin{pmatrix} -3 \\ 13 \end{pmatrix} + 1 \begin{pmatrix} -1 \\ -5 \end{pmatrix} \end{aligned}$$

$$= \begin{pmatrix} -14 + 9 + 6 - 1 \\ 28 + 3 - 26 - 5 \end{pmatrix}$$

$$= \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \checkmark$$

2. Let  $\mathbf{u} = \begin{pmatrix} 1 \\ 2 \\ -3 \end{pmatrix}$ ,  $\mathbf{v} = \begin{pmatrix} 0 \\ -4 \\ -3 \end{pmatrix}$ ,  $\mathbf{w} = \begin{pmatrix} 7 \\ 2 \\ 0 \end{pmatrix} \in \mathbb{R}^3$ . Are these three vectors linearly independent? If not, write  $\mathbf{w}$  as a linear combination of  $\mathbf{u}$  and  $\mathbf{v}$ .

Solution: The definition of linear independence, we recall, says "If  $a\mathbf{u} + b\mathbf{v} + c\mathbf{w} = \mathbf{0}$ , then  $a = b = c = 0$ ." Let's see whether this is the case for our  $\mathbf{u}$ ,  $\mathbf{v}$  and  $\mathbf{w}$ :

Suppose there are scalars  $a, b, c \in \mathbb{R}$  satisfying  $a\mathbf{u} + b\mathbf{v} + c\mathbf{w} = \mathbf{0}$ . Because in the case of vectors in  $\mathbb{R}^n$  this amounts to a matrix product (*by the definition matrix product!*), we can *rephrase* the original problem in terms of a system, then use row-reduction:

$$\begin{aligned} \mathbf{0} &= a\mathbf{u} + b\mathbf{v} + c\mathbf{w} \\ &= a \begin{pmatrix} 1 \\ 2 \\ -3 \end{pmatrix} + b \begin{pmatrix} 0 \\ -4 \\ -3 \end{pmatrix} + c \begin{pmatrix} 7 \\ 2 \\ 0 \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 & 7 \\ 2 & -4 & 2 \\ -3 & -3 & 0 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} \\ &= A\mathbf{x} \end{aligned}$$

where

$$A = \begin{pmatrix} 1 & 0 & 7 \\ 2 & -4 & 2 \\ -3 & -3 & 0 \end{pmatrix}, \quad \mathbf{x} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$

Let's see whether  $\mathbf{x} = \mathbf{0}$  or not, then, using the matrix method of row-reduction on  $(A \mid \mathbf{0})$ : since

$$\text{rref}(A \mid \mathbf{0}) = \left( \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right)$$

we conclude that  $a = b = c = 0$ , or  $\mathbf{x} = \mathbf{0}$ . Thus,

**$\mathbf{u}$ ,  $\mathbf{v}$  and  $\mathbf{w}$  are linearly independent!**

3. **True or False:** If a matrix  $A$  has trivial null space,  $N(A) = \{\mathbf{0}\}$ , then as a function  $A$  is one-to-one.

Solution: True. To see this, suppose there were two vectors  $\mathbf{x}_1, \mathbf{x}_2 \in \mathbb{R}^n$  satisfying  $A\mathbf{x}_1 = A\mathbf{x}_2$  (i.e.  $\mathbf{x}_1$  and  $\mathbf{x}_2$  are sent to the same  $\mathbf{y}$ -value). Then, subtracting  $A\mathbf{x}_2$  from both sides and factoring  $A$  out gives

$$\mathbf{0} = A\mathbf{x}_1 - A\mathbf{x}_2 = A(\mathbf{x}_1 - \mathbf{x}_2)$$

But this means  $\mathbf{x}_1 - \mathbf{x}_2 \in N(A) = \{\mathbf{0}\}$ , and therefore

$$\mathbf{x}_1 - \mathbf{x}_2 = \mathbf{0}$$

from which it follows that in fact  $\mathbf{x}_1 = \mathbf{x}_2$ . That is, there can only be one  $\mathbf{x} \in \mathbb{R}^n$  to each  $\mathbf{y} = A\mathbf{x} \in \mathbb{R}^m$ .  $A$  is one-to-one.

4. **True or False:** If  $A \in M_{3,5}(\mathbb{R})$ , then as a function  $A$  cannot be one-to-one.

Solution: True. Actually,  $A$  *must* have nontrivial null space  $N(A)$ , because it can have max 3 pivot columns, therefore *at least* 2 free variables. Parametrizing them as in number 1 above would give *two linearly independent* solution vectors of  $A\mathbf{x} = \mathbf{0}$ , nevermind all their possible linear combinations, which would also be solutions. Thus, there are in fact infinitely many vectors sent to the same  $\mathbf{y}$ -value of  $\mathbf{0}$ , and this merely by dimension-count, because  $3 < 5$ .

5. **True or False:** The matrix  $A$  in problem #1 above is onto.

Solution: True. First,  $A$  can be thought of as a function  $\mathbb{R}^4 \rightarrow \mathbb{R}^2$ , and the range  $R(A)$  is a subspace of its target space  $\mathbb{R}^2$ , which moreover, *by the definition of matrix multiplication*, is characterized by

$$R(A) = \text{span}(\mathbf{a}_1, \mathbf{a}_2, \mathbf{a}_3, \mathbf{a}_4)$$

Since two of the columns are pivots, namely  $\mathbf{a}_1 = \begin{pmatrix} -2 \\ 4 \end{pmatrix}$ ,  $\mathbf{a}_2 = \begin{pmatrix} 3 \\ 1 \end{pmatrix}$ , and those are always linearly independent (since we have seen that the dependency relations among the columns of  $B = \text{rref}(A)$  are the same as those among the columns of  $A$ , and this is true of  $\mathbf{b}_1$  and  $\mathbf{b}_2$ ), this reduces the span to

$$R(A) = \text{span}(\mathbf{a}_1, \mathbf{a}_2)$$

(since  $\mathbf{a}_3, \mathbf{a}_4 \in \text{span}(\mathbf{a}_1, \mathbf{a}_2)$  already, as we showed in # 1(d)). The matrix  $C \in M_2(\mathbb{R})$  whose columns are  $\mathbf{a}_1$  and  $\mathbf{a}_2$ , as a function  $\mathbb{R}^2 \rightarrow \mathbb{R}^2$ , is one-to-one (because its columns are linearly independent) and onto (since if  $\mathbf{y} \in \mathbb{R}^2$ , then

$$\text{rref}(\mathbf{a}_1 \ \mathbf{a}_2 \mid \mathbf{y}) = \left( \begin{array}{cc|c} 1 & 0 & h \\ 0 & 1 & k \end{array} \right)$$

for some  $h, k \in \mathbb{R}$ , namely the components of the solution  $\mathbf{x}$  in  $C\mathbf{x} = \mathbf{y}$ , whence  $R(C) = \mathbb{R}^2$  and  $A$  is onto).  $C$  is therefore invertible. Alternatively, we can compute the inverse directly,

$$C^{-1} = \begin{pmatrix} -2 & 3 \\ 4 & 1 \end{pmatrix}^{-1} = \frac{1}{(-2) \cdot 1 - 4 \cdot 3} \begin{pmatrix} 1 & -3 \\ -4 & -2 \end{pmatrix} = \begin{pmatrix} -1/16 & 3/16 \\ 1/4 & 1/8 \end{pmatrix}$$

Either way, we can see that  $C$  is onto, and since  $C$  has range the same as  $A$ ,  $A$  is onto.

6. **True or False:** A square matrix  $A$  is invertible (i.e. both one-to-one and onto as a function) if and only if the columns of  $A$  are linearly independent.

Solution: True. Linear independence in  $\mathbb{R}^n$  can be rephrased in terms of a matrix equation (by the definition of matrix multiplication), so the statement “ $A$  is linearly independent” is, by definition, equivalent to the statement “If  $\sum_{i=1}^n c_i \mathbf{a}_i = \mathbf{0}$ , then all  $c_i = 0$ ,” but by the definition of matrix multiplication this statement is equivalent to “If  $A\mathbf{x} = \mathbf{0}$ , where  $\mathbf{x}$  is the column vector with components the  $c_i$ , then  $\mathbf{x} = \mathbf{0}$ .” Thus, to say all  $c_i = 0$  and to say the column vector  $\mathbf{x} = \mathbf{0}$  are equivalent statements.

What does this show? It shows that “The columns of  $A$  are linearly independent” iff “The null space of  $A$  is trivial,” i.e.  $N(A) = \{\mathbf{0}\}$ , and this is equivalent to “ $A$  is *one-to-one*” by #3 above. What about *onto*? This is where the *square* part of  $A$  comes in. We know that  $A$  is  $n \times n$ , so if it has  $n$  linearly independent columns, the system  $A\mathbf{x} = \mathbf{y}$  will have a unique solution for each  $\mathbf{y} \in \mathbb{R}^n$ , as  $\text{rref}(A | \mathbf{y})$  will be  $(I | \mathbf{h})$ , which means  $\mathbf{x} = \mathbf{h}$  is a solution. Thus,  $R(A) = \mathbb{R}^n$  and  $A$  is onto.

**Remark 0.1** Observe that the *uniqueness* of  $\mathbf{x}$  actually is the one-to-one part. That part of the problem could be attacked in this row-reduction way, but the connection to linear independence above still needs to be made via the null space  $N(A)$ , as in number 3. ■

7. Let  $A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \in M_2(\mathbb{R})$  and  $B = \begin{pmatrix} -1 & 2 & -3 \\ 4 & -5 & 6 \end{pmatrix} \in M_{2,3}(\mathbb{R})$ . Compute  $A^{-1}B$ .

Solution:  $A^{-1} = \frac{1}{1 \cdot 4 - 3 \cdot 2} \begin{pmatrix} 4 & -2 \\ -3 & 1 \end{pmatrix} = -\frac{1}{2} \begin{pmatrix} 4 & -2 \\ -3 & 1 \end{pmatrix}$ , which I'll leave unscaled for now, to make the computation easier. Then, using the rows of  $A^{-1}$  (prior to scaling by  $-1/2$ ) and the columns of  $B$ ,

$$\begin{aligned}
 A^{-1}B &= -\frac{1}{2} \begin{pmatrix} 4 & -2 \\ -3 & 1 \end{pmatrix} \begin{pmatrix} -1 & 2 & -3 \\ 4 & -5 & 6 \end{pmatrix} \\
 &= -\frac{1}{2} \begin{pmatrix} \vec{A}_1^{-1} \cdot \mathbf{b}_1 & \vec{A}_1^{-1} \cdot \mathbf{b}_2 & \vec{A}_1^{-1} \cdot \mathbf{b}_3 \\ \vec{A}_2^{-1} \cdot \mathbf{b}_1 & \vec{A}_2^{-1} \cdot \mathbf{b}_2 & \vec{A}_2^{-1} \cdot \mathbf{b}_3 \end{pmatrix} \\
 &= -\frac{1}{2} \begin{pmatrix} \langle 4, -2 \rangle \cdot \begin{pmatrix} -1 \\ 4 \end{pmatrix} & \langle 4, -2 \rangle \cdot \begin{pmatrix} 2 \\ -5 \end{pmatrix} & \langle 4, -2 \rangle \cdot \begin{pmatrix} -3 \\ 6 \end{pmatrix} \\ \langle -3, 1 \rangle \cdot \begin{pmatrix} -1 \\ 4 \end{pmatrix} & \langle -3, 1 \rangle \cdot \begin{pmatrix} 2 \\ -5 \end{pmatrix} & \langle -3, 1 \rangle \cdot \begin{pmatrix} -3 \\ 6 \end{pmatrix} \end{pmatrix} \\
 &= -\frac{1}{2} \begin{pmatrix} -12 & 18 & -24 \\ 7 & -11 & 15 \end{pmatrix} \\
 &= \boxed{\begin{pmatrix} 6 & -9 & 12 \\ -\frac{7}{2} & \frac{11}{2} & -\frac{15}{2} \end{pmatrix}}
 \end{aligned}$$