

# MIT 18.06 Exam 1 Solutions, Fall 2017

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### Problem 1:

You are given three vectors  $\vec{v}_1 = \begin{pmatrix} 1 \\ -2 \\ 3 \end{pmatrix}$ ,  $\vec{v}_2 = \begin{pmatrix} 1 \\ 0 \\ 5 \end{pmatrix}$ , and  $\vec{v}_3 = \begin{pmatrix} 0 \\ 2 \\ 4 \end{pmatrix}$ .

Your goal is to find a *linear combination of these three vectors* (that is, multiply them by some numbers  $x_1, x_2, x_3$  and add them) to give the vector  $\vec{b} = \begin{pmatrix} 2 \\ -2 \\ 12 \end{pmatrix}$ .

- (a) Write the equation in matrix form.
- (b) Solve it to find the correct linear combination  $(x_1, x_2, x_3)$  of  $\vec{v}_1$ ,  $\vec{v}_2$ , and  $\vec{v}_3$ .
- (c) Change *one number* in  $\vec{v}_3$  to make the problem have *no* solution for *most* vectors  $\vec{b}$ , but give a new vector  $\vec{b}'$  for which there *is* still a solution. This new  $\vec{b}'$  is in the \_\_\_\_\_ space of the matrix \_\_\_\_\_.

(There are multiple correct answers for your new  $\vec{v}_3$  and your new  $\vec{b}'$ .)

### Solution:

- (a) **Write the equation in matrix form.**

Recall that if  $A = [c_1 | c_2 | c_3]$  is a matrix with three columns  $c_1, c_2, c_3$  then

$$A \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = x_1 c_1 + x_2 c_2 + x_3 c_3.$$

So, finding numbers  $x_1, x_2, x_3$  such that  $x_1 \vec{v}_1 + x_2 \vec{v}_2 + x_3 \vec{v}_3 = \vec{b}$  is the same as solving the equation  $Ax = \vec{b}$  for the unknown vector  $x = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$ ,

where  $A = [\vec{v}_1 | \vec{v}_2 | \vec{v}_3] = \begin{pmatrix} 1 & 1 & 0 \\ -2 & 0 & 2 \\ 3 & 5 & 4 \end{pmatrix}$ .

- (b) **Solve it to find the correct linear combination**  $(x_1, x_2, x_3)$  **of**  $\vec{v}_1$ ,  $\vec{v}_2$ , **and**  $\vec{v}_3$ .

We can solve the matrix equation by performing elimination on the augmented matrix  $(A|b)$  to make it upper triangular:

$$\left( \begin{array}{ccc|c} 1 & 1 & 0 & 2 \\ -2 & 0 & 2 & -2 \\ 3 & 5 & 4 & 12 \end{array} \right) \rightsquigarrow \left( \begin{array}{ccc|c} 1 & 1 & 0 & 2 \\ 0 & 2 & 2 & 2 \\ 0 & 2 & 4 & 6 \end{array} \right) \rightsquigarrow \left( \begin{array}{ccc|c} 1 & 1 & 0 & 2 \\ 0 & 2 & 2 & 2 \\ 0 & 0 & 2 & 4 \end{array} \right).$$

Then backsubstitution yields  $2x_3 = 4 \implies x_3 = 2$ ,  $2x_2 + 2x_3 = 2 = 2x_2 + 4 \implies x_2 = -1$ ,  $x_1 + x_2 = 2 = x_1 - 1 \implies x_1 = 3$ , or

$$x = \begin{pmatrix} 3 \\ -1 \\ 2 \end{pmatrix}.$$

- (c) **Change one number in**  $\vec{v}_3$  **to make the problem have** *no* **solution for** *most* **vectors**  $\vec{b}$ , **but give a new vector**  $\vec{b}'$  **for which there is still a solution. This new**  $\vec{b}'$  **is in the** \_\_\_\_\_ **space of the matrix** \_\_\_\_\_.

To not have solutions for most right-hand sides, the matrix needs to be singular. We just need to change  $\vec{v}_3$  so that we get a 0 instead of a 2 for the last step of Gaussian elimination, to eliminate the third pivot, so we just

need to subtract 2 from the third component, and hence  $\vec{v}_3' = \begin{pmatrix} 0 \\ 2 \\ 2 \end{pmatrix}$ .

(Equivalently, this makes  $\vec{v}_3' = \vec{v}_2 - \vec{v}_1$ , so the column space becomes two-dimensional.) Now we have a new matrix  $A' = [\vec{v}_1 | \vec{v}_2 | \vec{v}_3']$ , and to have a solution to  $A'\vec{x} = \vec{b}'$  we just need  $\vec{b}'$  in the **column space** of  $A'$ . We can just pick any  $\vec{x}$  we want and let  $\vec{b}' = A'\vec{x}$ , or equivalently we can pick  $\vec{b}'$  to be any linear combination of the columns of  $A'$ . For example, we can pick  $\vec{b}'$  to be *one of the columns* of  $A'$ , or the sum of two columns, or even  $\vec{b}' = \vec{0}$ .

### Problem 2:

Suppose  $A$  is some  $3 \times 3$  matrix. We will transform this into a *new*  $3 \times 3$  matrix  $B$  by doing operations on the rows or columns of  $A$  as follows. For each part, (i) **explain how to express  $B$  as  $B=AE$  or  $B=EA$  (say which!) for some matrix  $E$  (write down  $E$ !)**. Also, (ii) say **whether  $E$  is invertible** (that is, whether the transformation is reversible). (You don't need to compute  $E^{-1}$ , just say whether the inverse exists!)

- (a) **Swap the first and second rows of  $A$ .**

Remember that left-multiplications do row operations and that right-multiplications do column operations

- (i) Swapping the first and second rows is an elementary row operation, given by left-multiplication by the matrix

$$E = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

In this case we therefore have  $B = EA$  with  $E$  as above. (ii)  $E$  is invertible (in fact  $E^2 = I$ ), since you can undo a row swap by swapping again.

- (b) **Keep the first row the same, *then* add the second row to the third row, *then* replace the second row with the sum of the first and third rows.**

- (i) We are again performing row operations, so we'll have  $B = EA$ . To find  $E$ , we can simply apply the operations to the identity matrix  $I$ . Keeping the first row the same doesn't change  $I$ . Adding the second row to the third row yields the matrix:

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

Replacing the second row with the sum of the first and third rows gives the final answer:

$$E = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix}$$

- (ii)  $E$  is not invertible; its columns are linearly dependent. In fact, the last two columns are equal. This means that the vector  $\begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix}$  is in the

nullspace of  $E$ . But the nullspace of an invertible matrix must include only the zero vector. Alternatively, we could just do elimination:

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

to see that we only have two pivots, hence  $E$  is singular.

(c) **Subtract the first *column* from the second and third columns.**

(i) We are now operating on columns, so we'll have  $\boxed{B = AE}$ . To compute  $E$ , as usual we can just apply the operation in question to the identity matrix. Subtracting the first column from the second and third columns gives the matrix:

$$\boxed{E = \begin{pmatrix} 1 & -1 & -1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}}$$

(ii)  $\boxed{E \text{ is invertible}}$ , because the corresponding column operation is invertible: just add the first column back to the other two! In fact, from this we can see that the inverse of  $E$  is

$$E^{-1} = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

### Problem 3:

Suppose you have a  $3 \times 3$  matrix  $A$  satisfying  $A = B^{-1}UL$  where

$$B = \begin{pmatrix} 1 & 2 & 1 \\ 3 & -1 & 1 \\ -2 & 0 & -1 \end{pmatrix}, \quad U = \begin{pmatrix} 1 & 1 & 2 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{pmatrix}, \quad L = \begin{pmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 1 & -2 & 1 \end{pmatrix}.$$

- (a) **The *second* column  $c$  of the matrix  $A^{-1}$  satisfies  $Ac = b$  for what right-hand side  $b$ ?**

Recall that if  $M$  is any  $3 \times 3$  matrix and if  $e_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$ , then  $Me_2$  is the second column of  $M$ . So  $c = A^{-1}e_2$ . We want to know what vector  $Ac$  is. Using our formula for  $c$ , we get

$$Ac = A(A^{-1}e_2) = (AA^{-1})e_2 = Ie_2 = e_2.$$

So  $\boxed{b = e_2}$ .

- (b) **The *second* column  $c$  of the matrix  $A^{-1}$  also satisfies  $ULc = d$  for what right-hand side  $d$ ?**

We're given that  $A = B^{-1}UL$ , and from part (1) we have that  $Ac = e_2$ . Putting these together, we have  $B^{-1}ULc = e_2$ . Multiplying both sides by  $B$  on the left, we then get  $ULc = Be_2$ . But  $Be_2$  is just the second column of  $B$ , so we get:

$$ULc = Be_2 = \begin{pmatrix} 2 \\ -1 \\ 0 \end{pmatrix}$$

So  $\boxed{d = \begin{pmatrix} 2 \\ -1 \\ 0 \end{pmatrix}}$ .

- (c) **Compute the second column  $c$  of the matrix  $A^{-1}$ . (Important: you don't *have* to compute the inverse of any matrix!)**

By (2), to get  $c$  we can just solve the system  $ULc = \begin{pmatrix} 2 \\ -1 \\ 0 \end{pmatrix}$  for  $c$ . First, we can solve for  $Lc$  by backsubstitution in the augmented matrix  $(U|d)$ :

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

so  $Lc = \begin{pmatrix} 3 \\ -1 \\ 0 \end{pmatrix}$ . We can now solve this lower-triangular system for  $c$  by forward-substitution:

$$\left( \begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 3 & 1 & 0 & -1 \\ 1 & -2 & 1 & 0 \end{array} \right) \rightarrow \left( \begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & 1 & 0 & -10 \\ 0 & -2 & 1 & -3 \end{array} \right) \rightarrow \left( \begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & 1 & 0 & -10 \\ 0 & 0 & 1 & -23 \end{array} \right)$$

So  $c = \begin{pmatrix} 3 \\ -10 \\ -23 \end{pmatrix}$ .

**Common mistake:** Many students think that the inverse of a matrix like  $U$  or  $L$  can be found just by flipping the signs of the off-diagonal entries. It is true that sometimes inverses have a simple form for matrices of special types, and it *is* true that you can just flip the signs to invert the lower-triangular matrix describing a *single* elimination step, but it is *not* true that you can invert a general  $U$  or  $L$  matrix this way (even if they have 1's on the diagonal).

### Problem 4 (20 points):

In class and homework, we showed that multiplying two arbitrary  $m \times m$  matrices, doing Gaussian elimination, or inverting an  $m \times m$  matrix requires  $\sim m^3$  arithmetic operations (that is, roughly proportional to  $m^3$  for large  $m$ ). We found that adding matrices, multiplying an  $m \times m$  matrix by a vector, or solving an  $m \times m$  upper/lower triangular system of equations requires  $\sim m^2$  operations.

Suppose that  $A$  is an  $m \times m$  matrix,  $x$  is an  $m$ -component *column* vector (an  $m \times 1$  matrix), and  $r$  is an  $m$ -component *row* vector (a  $1 \times m$  matrix).

- You could compute the same result  $xrAx$  by doing the multiplications in different orders, for example  $x(r(Ax))$  (multiplying terms from *right to left*) or  $((xr)A)x$  (multiplying from *left to right*). **Give the rough number of operations** (say whether proportional to  $\sim m$ ,  $\sim m^2$ ,  $\sim m^3$ , or  $\sim m^4$ ) **for these two different orders (right to left and left to right)**. Which one is the fastest for  $m = 1000$ ?

### Solution:

- Let's look at  $x(r(Ax))$  first.  $A$  is a matrix and  $x$  is a column vector, so computing  $Ax$  takes  $\sim m^2$  operations. Then  $r$  is a row vector and  $Ax$  is a column vector, so computing  $r(Ax)$  is a dot product, just  $\sim m$  operations. Finally  $r(Ax)$  is a  $1 \times 1$  matrix, so computing  $x(r(Ax))$  takes  $\sim m$  operations (multiplying each entry by a number). The largest power of  $m$  we saw was  $m^2$ , so the whole procedure takes  $\sim m^2$  operations.
- Now let's look at  $((xr)A)x$ . Note that  $x$  is a column vector and  $r$  is a row vector, so computing  $xr$  takes  $\sim m^2$  operations — that is, the result is an  $m \times m$  matrix, and each entry involves a different multiplication, hence exactly  $m^2$  multiplications are needed. Then  $xr$  is a  $m \times m$  matrix, and so is  $A$ , so computing  $(xr)A$  is a matrix–matrix product that takes  $\sim m^3$  operations (as given above). Finally result  $(xr)A$  is then another  $m \times m$  matrix, so computing the product  $((xr)A)x$  is a matrix–vector multiply that takes another  $\sim m^2$  operations (as given above). The largest power of  $m$  we saw was  $m^3$ , so the whole procedure takes  $\sim m^3$  operations.

When  $m = 1000$ , the first option (right to left) takes  $\sim 10^6$  operations, while the second option (left to right) takes  $\sim 10^9$  operations. So definitely right to left is better.

- **Remark:** More generally, if  $X$  is an  $m \times n$  matrix and  $Y$  is an  $n \times p$  matrix, then computing  $XY$  involves computing the  $mp$  entries of  $XY$ , and each entry involves computing a dot product of two length- $m$  vectors ( $\sim m$  operations, actually  $\approx 2m$ ), so computing  $XY$  takes  $\sim mnp$  operations (actually  $\approx 2mnp$ ). This gives the square matrix–matrix and square matrix–vector results given in the problem, and also tells us that computing  $xr$  takes  $\sim m^2$  operations, etc.