

## Math 218D Problem Session: Week 2

### Answer Key

#### 1. Reduced Row Echelon Form.

For each of the following augmented matrices:

$$\text{a) } \left( \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 4 & 0 \\ 3 & 0 & 0 & 1 \end{array} \right) \quad \text{b) } \left( \begin{array}{ccc|c} 1 & 0 & 3 & 2 \\ 0 & 2 & 0 & 3 \\ 0 & 0 & 1 & 2 \end{array} \right) \quad \text{c) } \left( \begin{array}{cc|c} 2 & 0 & -1 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \end{array} \right) \quad \text{d) } \left( \begin{array}{c|c} 1 & -1 \\ 0 & 1 \\ 0 & 0 \end{array} \right)$$

answer the following questions:

- (1) Is the matrix in REF?
- (2) Is the matrix in RREF?
- (3) If the matrix is not in RREF, use the Gauss–Jordan algorithm to convert it into RREF.
- (4) Circle the pivot positions of the augmented matrices and determine the number of solutions to the corresponding system of linear equations.

#### Solution.

- (1) No; Yes; Yes; Yes.
- (2) No; No; No; No.

(3) (a) Use Gaussian elimination to obtain  $\left( \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 4 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right)$ .

(b) Use Jordan substitution to obtain  $\left( \begin{array}{ccc|c} 1 & 0 & 0 & -4 \\ 0 & 1 & 0 & 3/2 \\ 0 & 0 & 1 & 2 \end{array} \right)$ .

(c) Use Jordan substitution to obtain  $\left( \begin{array}{cc|c} 1 & 0 & -1/2 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{array} \right)$ .

(d) Use Jordan substitution to obtain  $\left( \begin{array}{c|c} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{array} \right)$ .

- (4) (a) Pivots: 1, 1, 1. No solution.
- (b) Pivots: 1, 2, 1. Unique solution.
- (c) Pivots: 2, -1. Unique solution.
- (d) Pivots: 1, 1. No solution.

**2. Elementary matrices**

- a) Write down the  $3 \times 3$  elementary matrix which  $E_1$  which performs the row operation  $R_2 += 3R_1$ .
- b) Write down the  $3 \times 3$  elementary matrix  $E_2$  which performs the row operation  $R_1 \leftrightarrow R_3$ .
- c) What (non-elementary) matrix first performs the row operation  $R_2 += 3R_1$  and *then* does  $R_1 \leftrightarrow R_3$ ?

**Solution.**

a)  $E_1 = \begin{pmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$

b)  $E_2 = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}$

c) The matrix is  $\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 1 \\ 3 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}$ .

**3. True or false?**

If true, give an explanation; if false, give a counterexample.

- a) All elementary matrices are invertible.
- b) A  $3 \times 2$  matrix can have 3 pivots.
- c) If a  $3 \times 4$  augmented matrix has 3 pivots, then the corresponding system of equations has a unique solution.

**Solution.**

- a) Yes, since we can “un-do” the row operation.
- b) No, since the number of pivots can not be larger than the number of rows or columns.
- c) No, consider  $\left( \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 4 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right)$ .

**4. Solving  $Ax = b$  using  $A = LU$ .**

When you know the  $LU$  factorization of a matrix  $A$ , you can use it to solve the matrix equation  $Ax = b$ . In this problem we will go through this process in an example.

Solve the matrix equation

$$\begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 2 & 2 & 3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 1 \\ 3 \\ 2 \end{pmatrix},$$

using the  $A = LU$  decomposition

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

- Identify  $A$ ,  $b$ ,  $L$ , and  $U$ .
- Convert  $Ly = b$  into 3 linear equations, and solve for  $y = (y_1, y_2, y_3)$  using forward-substitution.
- Convert  $Ux = y$  into 3 linear equations, and solve for  $x$  using back-substitution.
- Check your answer, by multiplying  $A \cdot x$  and confirming that it equals  $b$ .

Why does this work? If  $Ly = b$  and  $y = Ux$ , then  $L(Ux) = b$ . Since  $A = LU$ , this means that  $Ax = b$ .

**Solution.**

$$\text{a) } A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 2 & 2 & 3 \end{pmatrix}, b = \begin{pmatrix} 1 \\ 3 \\ 2 \end{pmatrix}, L = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 2 & 0 & 1 \end{pmatrix}, U = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix}.$$

b)

$$\begin{aligned} y_1 &= 1 \\ y_1 + y_2 &= 3 \\ 2y_1 + y_3 &= 2, \end{aligned}$$

and substitution gives  $(y_1, y_2, y_3) = (1, 2, 0)$ .

c)

$$\begin{aligned} x_1 + x_2 + x_3 &= 1 \\ x_2 + 2x_3 &= 2 \\ x_3 &= 0, \end{aligned}$$

and substitution gives  $(x_1, x_2, x_3) = (-1, 2, 0)$ .

$$\text{d) Check } \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 2 & 2 & 3 \end{pmatrix} \begin{pmatrix} -1 \\ 2 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 3 \\ 2 \end{pmatrix}.$$

**5. Finding  $A = LU$  and  $A^{-1}$  using elementary matrices**

Consider the matrix

$$A = \begin{pmatrix} 1 & -1 & 2 \\ 2 & -1 & 4 \\ 1 & 4 & 6 \end{pmatrix}.$$

- a) Explain how to reduce  $A$  to a matrix  $U$  in REF (not RREF) using three row replacements.
- b) Let  $E_1, E_2, E_3$  be the elementary matrices for these row operations, in order. Fill in the blank with a product involving the  $E_i$ :

$$U = \underline{\hspace{2cm}} A.$$

- c) Fill in the blank with a product involving the  $E_i^{-1}$ :

$$A = \underline{\hspace{2cm}} U$$

- d) Evaluate that product to produce a lower-unitriangular matrix  $L$  such that  $A = LU$ .
- e) Explain how to reduce  $U$  to the  $3 \times 3$  identity matrix using three more elementary matrices  $E_4, E_5, E_6$  (scaling, followed by row replacements).
- f) Fill in the blank with a product involving the  $E_i$ :

$$A^{-1} = \underline{\hspace{2cm}}.$$

- g) Compute  $A^{-1}$  by row reducing  $(A \mid I_n)$ . This is exactly the same as evaluating the product above!

**Solution.**

- a) We perform Gaussian elimination as follows:

$$\begin{aligned} A = \begin{pmatrix} 1 & -1 & 2 \\ 2 & -1 & 4 \\ 1 & 4 & 6 \end{pmatrix} & \xrightarrow{\substack{R_2 \leftarrow 2R_1 \\ \text{~~~~~}}} \begin{pmatrix} 1 & -1 & 2 \\ 0 & 1 & 0 \\ 1 & 4 & 6 \end{pmatrix} \\ & \xrightarrow{\substack{R_3 \leftarrow R_1 \\ \text{~~~~~}}} \begin{pmatrix} 1 & -1 & 2 \\ 0 & 1 & 0 \\ 0 & 5 & 4 \end{pmatrix} \\ & \xrightarrow{\substack{R_3 \leftarrow 5R_2 \\ \text{~~~~~}}} \begin{pmatrix} 1 & -1 & 2 \\ 0 & 1 & 0 \\ 0 & 0 & 4 \end{pmatrix} = U. \end{aligned}$$

- b) The elementary matrices are

$$E_1 = \begin{pmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad E_2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1 & 0 & 1 \end{pmatrix}, \quad E_3 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -5 & 1 \end{pmatrix}.$$

Therefore,

$$U = E_3 E_2 E_1 A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -5 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} A.$$

$$\text{c) } A = E_1^{-1} E_2^{-1} E_3^{-1} U = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 5 & 1 \end{pmatrix} U.$$

d) Multiplying elementary matrices together is easy: it's just doing row operations!

$$\begin{aligned} L &= \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 5 & 1 \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 5 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 1 & 5 & 1 \end{pmatrix}. \end{aligned}$$

e) We perform Jordan substitution as follows:

$$\begin{aligned} U &= \begin{pmatrix} 1 & -1 & 2 \\ 0 & 1 & 0 \\ 0 & 0 & 4 \end{pmatrix} \begin{array}{l} R_3 \div = 4 \\ \hline \end{array} \begin{pmatrix} 1 & -1 & 2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \\ &\begin{array}{l} R_1 - = 2R_3 \\ \hline \end{array} \begin{pmatrix} 1 & -1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \\ &\begin{array}{l} R_1 + = R_2 \\ \hline \end{array} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}. \end{aligned}$$

The corresponding elementary matrices are

$$E_4 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1/4 \end{pmatrix}, \quad E_5 = \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad E_6 = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

Using these matrices,

$$E_6 E_5 E_4 U = I_3.$$

$$\text{f) } A^{-1} = E_6 E_5 E_4 E_3 E_2 E_1.$$

g) First we do Gaussian elimination:

$$\begin{aligned}
 (A | I_3) &= \left( \begin{array}{ccc|ccc} 1 & -1 & 2 & 1 & 0 & 0 \\ 2 & -1 & 4 & 0 & 1 & 0 \\ 1 & 4 & 6 & 0 & 0 & 1 \end{array} \right) \xrightarrow{R_2 \leftarrow 2R_1} \left( \begin{array}{ccc|ccc} 1 & -1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 0 & -2 & 1 & 0 \\ 1 & 4 & 6 & 0 & 0 & 1 \end{array} \right) \\
 &\xrightarrow{R_3 \leftarrow R_1} \left( \begin{array}{ccc|ccc} 1 & -1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 0 & -2 & 1 & 0 \\ 0 & 5 & 4 & -1 & 0 & 1 \end{array} \right) \\
 &\xrightarrow{R_3 \leftarrow 5R_1} \left( \begin{array}{ccc|ccc} 1 & -1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 0 & -2 & 1 & 0 \\ 0 & 0 & 4 & 9 & -5 & 1 \end{array} \right) \\
 &= (U | E_3 E_2 E_1)
 \end{aligned}$$

We are halfway done—the right half of this matrix is now

$$E_3 E_2 E_1 = \begin{pmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 9 & -5 & 1 \end{pmatrix}.$$

**An aside:** one convenient fact about  $L = E_1^{-1} E_2^{-1} E_3^{-1}$  is the way in which its entries precisely correspond to the row operations performed. It is harder to interpret the entries of  $E_3 E_2 E_1$ . For example, why does 9 appear in  $E_3 E_2 E_1$ ? It is because

$$\text{final } R_3 = 9(\text{original } R_1) - 5(\text{original } R_2) + (\text{original } R_3).$$

Continuing onwards:

$$\begin{aligned}
 \left( \begin{array}{ccc|ccc} 1 & -1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 0 & -2 & 1 & 0 \\ 0 & 0 & 4 & 9 & -5 & 1 \end{array} \right) &\xrightarrow{R_3 \div 4} \left( \begin{array}{ccc|ccc} 1 & -1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 0 & -2 & 1 & 0 \\ 0 & 0 & 1 & 9/4 & -5/4 & 1/4 \end{array} \right) \\
 &\xrightarrow{R_1 \leftarrow 2R_3} \left( \begin{array}{ccc|ccc} 1 & -1 & 0 & -7/2 & 5/2 & -1/2 \\ 0 & 1 & 0 & -2 & 1 & 0 \\ 0 & 0 & 1 & 9/4 & -5/4 & 1/4 \end{array} \right) \\
 &\xrightarrow{R_1 \leftarrow R_2} \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & -11/2 & 7/2 & -1/2 \\ 0 & 1 & 0 & -2 & 1 & 0 \\ 0 & 0 & 1 & 9/4 & -5/4 & 1/4 \end{array} \right) \\
 &= (I_3 | A^{-1})
 \end{aligned}$$

We conclude that

$$A^{-1} = \begin{pmatrix} -11/2 & 7/2 & -1/2 \\ -2 & 1 & 0 \\ 9/4 & -5/4 & 1/4 \end{pmatrix}.$$

**6. Finding  $PA = LU$** 

a) Find the  $PA = LU$  decomposition of  $A = \begin{pmatrix} 1 & 1 & 1 \\ 2 & 2 & 5 \\ 1 & 2 & 3 \end{pmatrix}$  using the 3-column method.

b) Find the  $A = LU$  decomposition of  $A = \begin{pmatrix} 1 & 1 & 1 \\ -10 & -20 & -30 \\ 5 & 15 & 10 \end{pmatrix}$  using the 2-column method.

c) Find the  $PA = LU$  decomposition of  $A = \begin{pmatrix} 1 & 1 & 1 \\ -10 & -20 & -30 \\ 5 & 15 & 10 \end{pmatrix}$ , this time using *maximal partial pivoting*.

**Solution.**

$$\text{a) } P = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} \quad L = \begin{pmatrix} 1 & 0 & 0 \\ 1/2 & 1 & 0 \\ 1/2 & 0 & 1 \end{pmatrix} \quad U = \begin{pmatrix} 2 & 2 & 5 \\ 0 & 1 & 1/2 \\ 0 & 0 & -3/2 \end{pmatrix}$$

$$\text{b) } L = \begin{pmatrix} 1 & 0 & 0 \\ -10 & 1 & 0 \\ 5 & -1 & 1 \end{pmatrix} \quad U = \begin{pmatrix} 1 & 1 & 1 \\ 0 & -10 & -20 \\ 0 & 0 & -15 \end{pmatrix}$$

$$\text{c) } P = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} \quad L = \begin{pmatrix} 1 & 0 & 0 \\ -1/2 & 1 & 0 \\ -1/10 & -1/5 & 1 \end{pmatrix} \quad U = \begin{pmatrix} -10 & -20 & -30 \\ 0 & 5 & -5 \\ 0 & 0 & -3 \end{pmatrix}$$



**7. Using  $PA = LU$** 

In this problem, we'll solve the matrix equation

$$\begin{pmatrix} 1 & 1 & 1 \\ 2 & 2 & 5 \\ 1 & 2 & 3 \end{pmatrix} x = \begin{pmatrix} 1 \\ 5 \\ 2 \end{pmatrix}$$

using the  $PA = LU$  decomposition of problem 6a).

- Identify  $A$ ,  $b$ ,  $P$ ,  $L$ , and  $U$ .
- Compute  $Pb$ . What does  $P$  do to  $b$ ?
- Convert  $Ly = Pb$  into 3 linear equations, and solve for  $y = (y_1, y_2, y_3)$  using forward-substitution.
- Convert  $Ux = y$  into 3 linear equations, and solve for  $x$  using back-substitution.
- Check your answer, by multiplying  $A \cdot x$  and confirming that it equals  $b$ .

**Solution.**

- a) In this case,

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 2 & 2 & 5 \\ 1 & 2 & 3 \end{pmatrix} \quad b = \begin{pmatrix} 1 \\ 5 \\ 2 \end{pmatrix}$$

$$P = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} \quad L = \begin{pmatrix} 1 & 0 & 0 \\ 1/2 & 1 & 0 \\ 1/2 & 0 & 1 \end{pmatrix} \quad U = \begin{pmatrix} 2 & 2 & 5 \\ 0 & 1 & 1/2 \\ 0 & 0 & -3/2 \end{pmatrix}$$

- b) The permutation matrix  $P$  swaps the first and second rows, and then swaps the second and third rows of  $b$ :

$$Pb = \begin{pmatrix} 5 \\ 2 \\ 1 \end{pmatrix}.$$

c) 
$$\begin{array}{rcl} y_1 & = & 5 \\ \frac{1}{2}y_1 + y_2 & = & 2 \\ \frac{1}{2}y_1 + y_3 & = & 1 \end{array} \rightsquigarrow \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = \begin{pmatrix} 5 \\ -1/2 \\ -3/2 \end{pmatrix}.$$

d) 
$$\begin{array}{rcl} 2x_1 + 2x_2 + 5x_3 & = & 5 \\ x_2 + \frac{1}{2}x_3 & = & -\frac{1}{2} \\ -\frac{3}{2}x_3 & = & -\frac{3}{2} \end{array} \rightsquigarrow \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}.$$

e) 
$$\begin{pmatrix} 1 & 1 & 1 \\ 2 & 2 & 5 \\ 1 & 2 & 3 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 5 \\ 2 \end{pmatrix}.$$

**8. Maximal Partial Pivoting**

Consider the linear system

$$\begin{aligned}x_2 &= 1 \\x_1 + x_2 &= 2.\end{aligned}$$

Clearly the solution is  $x_1 = 1$  and  $x_2 = 1$ . Let's modify the system just a little bit:

$$\begin{aligned}10^{-17}x_1 + x_2 &= 1 \\x_1 + x_2 &= 2.\end{aligned}$$

Presumably the solution  $(x_1, x_2)$  will be very close to  $(1, 1)$ .

a) Perform Gauss–Jordan elimination on the augmented matrix

$$\left( \begin{array}{cc|c} 10^{-17} & 1 & 1 \\ 1 & 1 & 2 \end{array} \right)$$

to solve the modified system. You should obtain

$$x_1 = \frac{1}{1 - 10^{-17}} \quad x_2 = 1 + \frac{1}{1 - 10^{17}},$$

which are indeed very close to 1.

Now let's see if a computer can do the same. Load up the Sage cell on the course homepage. Create the augmented matrix as follows:

```
from sympy import *
# 1e-17 is 10^(-17)
A = Matrix([[1e-17, 1.0, 1.0], [1.0, 1.0, 2.0]])
pprint(A)
```

b) Now let's perform Gauss–Jordan elimination:

```
# This does R2 -= 10^(17) R1
# (force sympy to use the smaller pivot)
A.row_op(1, lambda v, j: v - 1e17*A[0,j])
pprint(A)
# Now do Jordan substitution
pprint(A.rref(pivots=False))
```

c) Verify that the last matrix has the form

$$\left( \begin{array}{cc|c} 1 & 0 & (?) \\ 0 & 1 & (?) \end{array} \right).$$

What does the computer think  $x_1$  and  $x_2$  are? What went wrong?

d) Python uses 64-bit floating point numbers. This means that they have about 16 decimal digits of precision. Try evaluating  $1+1e17$ . What did you get?

The problem was that you produced enormous numbers by dividing by the tiny number  $10^{-17}$ . When you're doing math on a computer, *you never want to divide by tiny numbers*.

e) Now try performing Gauss–Jordan elimination again, after selecting the maximal pivot in the first column:

```
A = Matrix([[1e-17, 1.0, 1.0], [1.0, 1.0, 2.0]])
A.row_swap(0,1)
pprint(A.rref(pivots=False))
```

Did that work? What does the computer think  $x_1$  and  $x_2$  are now?

### Solution.

a) First we perform Gaussian elimination, which only takes one row operation:

$$A = \left( \begin{array}{cc|c} 10^{-17} & 1 & 1 \\ & 1 & 2 \end{array} \right) \xrightarrow{R_2 - 10^{17}R_1} \left( \begin{array}{cc|c} 10^{-17} & 1 & 1 \\ & 0 & 1 - 10^{17} \end{array} \right).$$

Now we do back-substitution on the corresponding system of equations:

$$\begin{aligned} 10^{-17}x_1 + x_2 &= 1 \\ (1 - 10^{17})x_2 &= 2 - 10^{17}. \end{aligned}$$

The result is

$$\begin{aligned} x_2 &= \frac{2 - 10^{17}}{1 - 10^{17}} = 1 + \frac{1}{1 - 10^{17}} \\ x_1 &= -\frac{10^{17}}{1 - 10^{17}} = \frac{1}{1 - 10^{-17}}. \end{aligned}$$

c) Sympy outputs

```
[1  0  0 ]
[      ]
[0  1  1.0]
```

d) The computer thinks  $1 - 10^{17} = -10^{17} = 2 - 10^{17}$ , so  $x_2 = \frac{-10^{17}}{-10^{17}} = 1$ , and by back-substitution  $x_1 = 0$ .

e) 1e+17

f) Sympy outputs

```
[1  0  1.0]
[      ]
[0  1  1.0]
```

That works: now  $x_1 = 1$ ,  $x_2 = 1$ .

**9.  $PA = LU$  on a computer**

The purpose of this problem is to convince you that computing a  $PA = LU$  decomposition really is faster for solving  $Ax = b$  for many values of  $b$ . Load up the Sage cell on the course homepage.

a) Let's create a  $10 \times 10$  invertible matrix and a column vector of length 10:

```
from sympy import *
from time import time
A = (eye(10) + ones(10)) * 1.0
b = ones(10, 1) * 1.0
```

Here  $\text{eye}(n)$  is the  $n \times n$  identity matrix, and  $\text{ones}(m, n=m)$  is the  $m \times n$  matrix of 1's. We multiply by 1.0 in order to force Sympy to do floating-point (as opposed to symbolic) arithmetic. Now

$$A = \begin{pmatrix} 2 & 1 & 1 & \cdots & 1 & 1 \\ 1 & 2 & 1 & \cdots & 1 & 1 \\ 1 & 1 & 2 & \cdots & 1 & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 1 & 1 & 1 & \cdots & 2 & 1 \\ 1 & 1 & 1 & \cdots & 1 & 2 \end{pmatrix} \quad \text{and} \quad b = \begin{pmatrix} 1 \\ 1 \\ 1 \\ \vdots \\ 1 \\ 1 \end{pmatrix}.$$

b) Let's solve  $Ax = b$  using a  $PA = LU$  decomposition.

```
start = time()
L, U, _ = A.LUdecomposition()
for _ in range(1000):
    U.upper_triangular_solve(
        L.lower_triangular_solve(b))
end = time()
print(end - start)
```

This solves  $Ax = (1, 1, \dots, 1)$  1000 times, using the  $PA = LU$  decomposition.

c) Now let's solve  $Ax = b$  without using  $PA = LU$ .

```
start = time()
for _ in range(1000):
    A.solve(b)
end = time()
print(end - start)
```

What do you notice?

**Solution.**

b) When I ran it, it took 4.0 seconds.

c) When I ran it, it took 15.8 seconds.