No "final solutions" are provided for these types of questions, since the whole point of them is to encourage you to express *briefly* but *clearly* and *in your own words* what you understand. As explained in the directions, definitions taken from text books or the internet do not reflect a good understanding of these terms, nor do extremely long explanations. Equations do not express the meaning of these, nor do literal word translations of equations show that you know what they mean. Instead, we are looking for clear evidence that you understand what each term means. Possible definitions for each term is provided below:

- (a) Direct elimination: A method to solve a system of equations that uses row operations to perform forward elimination in order to solve the system using backwards substitution.
- (b) LU Factorization: Solving method that factors the initial matrix into lower and upper matrices in order to simplify the solving the system into forward and backwards substitutions.
- (c) Inverse matrix: A matrix such that any square matrix multiplied by its inverse gives the identity matrix.
- (d) Symmetric matrix: Any matrix that if you switch the rows and the columns you do not change the matrix.
- (e) Transpose: A matrix operation that flips a matrix along its diagonal, i.e. switches its rows and columns.
- (f) Permutation: Matrix operations that change the rows of an another matrix by multiplying it by a set of permutation matrices which are composed of the rows of the identity matrix.
- (g) Inner product: A matrix operation that gives the projection of one matrix onto another.
- (h) Singular matrix: A square matrix without an inverse.

Solve the following system of equations, showing all relevant steps and stating the method used to solve the system (e.g., LU factorization, direct numerical solution, matrix elimination, etc.,). Furthermore, if a system has no solution state it has no solution and briefly describe why it has no solution.

(a) The following system of equations can be solved using the inverse matrix approach, namely

$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{b} \tag{1a}$$

therefore we need to obtain  $A^{-1}$ . We can determine the inverse of matrix **A** using the augmented matrix method. Where the matrix **A** is augmented with the identity matrix **I**, and row operations are performed until the left hand side of the augmented matrix is the identity matrix. Starting with the augmented matrix

$$\begin{bmatrix}
1 & 3 & 1 & 1 & 0 & 0 \\
4 & 0 & 2 & 0 & 1 & 0 \\
1 & 3 & 4 & 0 & 0 & 1
\end{bmatrix}$$
(1b)

Then we can take  $4r_1 - r_2$  to zero out the  $a_{21}$  term giving

$$\begin{bmatrix}
1 & 3 & 1 & 1 & 0 & 0 \\
0 & 12 & 2 & 4 & -1 & 0 \\
1 & 3 & 4 & 0 & 0 & 1
\end{bmatrix}$$
(1c)

Next, the  $a_{31}$  is zeroed out by performing  $r_1 - r_3$  giving

$$\begin{bmatrix}
1 & 3 & 1 & 1 & 0 & 0 \\
0 & 12 & 2 & 4 & -1 & 0 \\
0 & 0 & -3 & 1 & 0 & -1
\end{bmatrix}$$
(1d)

Now we continue zeroing out entries of **A** moving up the matrix starting with  $a_{23}$  by performing  $-\frac{2}{3}r_3 - r_2$  giving

$$\begin{bmatrix}
1 & 3 & 1 & 1 & 0 & 0 \\
0 & -12 & 0 & -14/3 & 1 & 2/3 \\
0 & 0 & -3 & 1 & 0 & -1
\end{bmatrix}$$
(1e)

Moving up the third column we zero out  $a_{13}$  using  $-\frac{1}{3}r_3 - r_1$  giving

$$\begin{bmatrix} -1 & -3 & 0 & -1/3 & 0 & -1/3 \\ 0 & -12 & 0 & -14/3 & 1 & 2/3 \\ 0 & 0 & -3 & 1 & 0 & -1 \end{bmatrix}$$
 (1f)

Next we can zero  $a_{12}$  with  $\frac{1}{4}r_2 - r_1$  giving

$$\begin{bmatrix} 1 & 0 & 0 & 1/6 & 1/4 & -1/6 \\ 0 & -12 & 0 & -14/3 & 1 & 2/3 \\ 0 & 0 & -3 & 1 & 0 & -1 \end{bmatrix}$$
 (1g)

Lastly we can multiply each pivot by its reciprocal, i.e.,  $r_2 = -\frac{r_2}{12}$  and  $r_3 = -\frac{r_3}{3}$ , to finally get the inverse of matrix  $\mathbf{A}$ ,

$$\begin{bmatrix} 1 & 0 & 0 & 1/6 & 1/4 & -1/6 \\ 0 & 1 & 0 & 7/18 & -1/12 & -1/18 \\ 0 & 0 & 1 & -1/3 & 0 & 1/3 \end{bmatrix}$$
 (1h)

Thus

$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{b} = \begin{bmatrix} 1/6 & 1/4 & -1/6 \\ 7/18 & -1/12 & -1/18 \\ -1/3 & 0 & 1/3 \end{bmatrix} \begin{bmatrix} 5 \\ 6 \\ 3 \end{bmatrix} = \begin{bmatrix} 11/6 \\ 23/18 \\ -2/3 \end{bmatrix}$$
(1i)

(b) From the system of equations

$$\begin{bmatrix} 1 & 4 & 2 \\ 2 & 8 & 4 \\ 1 & 5 & 4 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$
 (2)

it is obvious that the second row is a linear combination of the first row, i.e.  $r_2 = 2r_1$ . Therefore the second equation in the system adds no new information, that we do not already have from the first row. Thus, we have a system with three unknowns, x, y, z, and only two equations, therefore the system does not have an unique solution.

(c) The simplest way to solve the following system

$$\begin{bmatrix} 1 & 3 & 1 \\ 4 & 1 & 2 \\ 0 & 0 & 4 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$
 (3a)

is to use to use the first row to cancel out the first entry of the second row, i.e.  $4r_1 - r_2$ , giving

$$\begin{bmatrix} 1 & 3 & 1 \\ 0 & 11 & 2 \\ 0 & 0 & 4 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$
 (3b)

and them backwards substituting. Thus,

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 15/44 \\ 3/22 \\ 1/4 \end{bmatrix} \tag{3c}$$

(d) The easiest way to solve this system of equations

$$\begin{bmatrix} 0 & 1 & 2 & 0 & 0 \\ 5 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{bmatrix}$$

$$(4a)$$

is to swap the rows so that all the pivots are non-zero giving

$$\begin{bmatrix} 5 & 0 & 2 & 0 & 0 \\ 0 & 1 & 2 & 0 & 0 \\ 0 & 0 & 1 & 0 & 3 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \\ 5 \\ 3 \\ 4 \end{bmatrix}$$

$$(4b)$$

Now the system of equations can be solved quite easily using backwards substitution. Thus,

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} 16/5 \\ 15 \\ -7 \\ -1 \\ 4 \end{bmatrix}$$
 (5)

The following system of equations

$$2x + 3y = 4 \tag{6a}$$

$$x + 2y = 5 \tag{6b}$$

can easily be solve by hand. First solve for x in terms of y in Eq. (6b),

$$x = 5 - 2y \tag{6c}$$

Next substitute x into Eq. (6a) to obtain a value for y, namely

$$2(-5 - 2y) + 3y = 4 \to y = 6 \tag{6d}$$

Thus,

$$x = -7 \tag{6e}$$

$$y = 6 \tag{6f}$$

Furthermore, the solution can be verified by solving for y in both equations and plotting both curves and observing their intersection point, as shown in Fig (1).

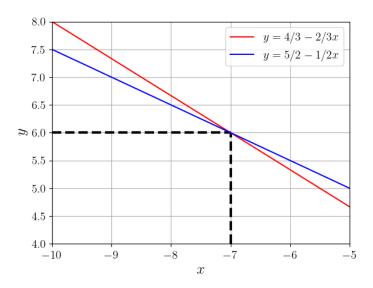


Figure 1: Verification plot for Exercise 3

Please see the Appendix for the full version of the verification code.

Prove the following matrix properties:

(a) Prove the following:

$$(AB)^T = B^T A^T (7a)$$

Start be defining two  $N \times N$  matrices **A** and **B**,

$$\mathbf{A} = A_{ij} = \begin{bmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n} \\ a_{2,1} & a_{2,2} & \cdots & a_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n,1} & a_{n_2} & \cdots & a_{n,n} \end{bmatrix}$$
 (7b)

$$\mathbf{B} = B_{ij} = \begin{bmatrix} b_{1,1} & b_{1,2} & \cdots & b_{1,n} \\ b_{2,1} & b_{2,2} & \cdots & b_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n,1} & b_{n,2} & \cdots & b_{n,n} \end{bmatrix}$$
(7c)

which gives the following for the LHS of Eq. (7a)

$$AB = \underbrace{\sum_{k=1}^{N} A_{ik} B_{kj}}_{\equiv C_{i,i}} = \begin{bmatrix} a_{1,1}b_{1,1} + a_{1,2}b_{2,1} & a_{1,1}b_{1,2} + a_{1,2}b_{2,2} & \cdots \\ a_{2,1}b_{1,1} + a_{2,2}b_{2,1} & a_{2,1}b_{1,2} + a_{2,2}b_{2,2} & \cdots \\ \vdots & \ddots & \vdots \end{bmatrix}$$
(7d)

Therefore,

$$C^{T} = \begin{bmatrix} a_{1,1}b_{1,1} + a_{1,2}b_{2,1} & a_{2,1}b_{1,1} + a_{2,2}b_{2,1} & \cdots \\ a_{1,1}b_{1,2} + a_{1,2}b_{2,2} & a_{2,1}b_{1,2} + a_{2,2}b_{2,2} & \cdots \\ \vdots & \ddots & \vdots \end{bmatrix}$$
(7e)

Next we can look at the RHS of Eq. (7a) where,

$$A^{T} = \begin{bmatrix} a_{1,1} & a_{2,1} & \cdots & a_{n,1} \\ a_{1,2} & a_{2,2} & \cdots & a_{n,2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1,n} & a_{2,n} & \cdots & a_{n,n} \end{bmatrix}$$

$$(7f)$$

$$B^{T} = \begin{bmatrix} b_{1,1} & b_{2,1} & \cdots & b_{n,1} \\ b_{1,2} & b_{2,2} & \cdots & b_{n,2} \\ \vdots & \vdots & \ddots & \vdots \\ b_{1,n} & b_{2,n} & \cdots & b_{n,n} \end{bmatrix}$$
(7g)

Therefore,

$$B^{T}A^{T} = \begin{bmatrix} b_{1,1}a_{1,1} + b_{2,1}a_{1,2} & b_{1,1}a_{2,1} + b_{2,1}a_{2,2} & \cdots \\ b_{1,2}a_{1,1} + b_{2,2}a_{1,2} & b_{1,2}a_{2,1} + b_{2,2}a_{2,2} & \cdots \\ \vdots & \ddots & \vdots \end{bmatrix}$$

$$(7h)$$

which is equivalent to Eq. (7e).

(b) To prove the following

$$\left(A^{-1}\right)^T = \left(A^T\right)^{-1} \tag{8a}$$

start with

$$A^{-1}A = I \tag{8b}$$

where from part(a) we know that

$$(A^{-1}A)^T = A^T (A^{-1})^T = I^T$$
 (8c)

and since

$$I^T = I (8d)$$

then

$$A^{-1}A = A^{T} (A^{-1})^{T} = I (8e)$$

Thus,

$$\left(A^{-1}\right)^T = \left(A^T\right)^{-1} \tag{8f}$$

(c) Lets consider the following symmetric matrix

$$C = \begin{bmatrix} c_{1,1} & c_{1,2} & c_{1,3} & \cdots & c_{1,n} \\ c_{1,2} & c_{2,2} & c_{2,3} & \cdots & c_{2,n} \\ c_{1,3} & c_{2,3} & c_{3,3} & \cdots & c_{3,n} \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ c_{1,n} & c_{2,n} & c_{3,n} & \cdots & c_{n,n} \end{bmatrix}$$
(9a)

where for in index notation  $C_{ij} = C_{ji}$ . Therefore,

$$C^{T} = \begin{bmatrix} c_{1,1} & c_{1,2} & c_{1,3} & \cdots & c_{1,n} \\ c_{1,2} & c_{2,2} & c_{2,3} & \cdots & c_{2,n} \\ c_{1,3} & c_{2,3} & c_{3,3} & \cdots & c_{3,n} \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ c_{1,n} & c_{2,n} & c_{3,n} & \cdots & c_{n,n} \end{bmatrix} = C$$

$$(9b)$$

(a)  $A = \begin{bmatrix} 1 & 3 & 2 \\ 3 & 5 & 4 \\ 2 & 4 & 6 \end{bmatrix}$  (10a)

As covered in lecture we know that A can be factored into a product of lower and upper matrices, namely

$$A = LU = \begin{bmatrix} 1 & 0 & 0 \\ l_{2,1} & 1 & 0 \\ l_{3,1} & l_{3,2} & 1 \end{bmatrix} \begin{bmatrix} u_{1,2} & u_{1,2} & u_{1,2} \\ 0 & u_{2,2} & u_{2,2} \\ 0 & 0 & u_{3,2} \end{bmatrix}$$
(10b)

where the non-zero off diagonal entries of L are the multipliers used in forward elimination, and the entries of U are the values after all the row operations are performed. Therefore, the easiest way to perform LU decomposition is to set U = A and perform forward elimination. Starting with the first row, since  $r_2 - 3r_1$  zeros out  $a_{2,1}$ , we know that  $l_{2,1} = 3$ , and that the updated U matrix is

$$U = \begin{bmatrix} 1 & 3 & 2 \\ 0 & -4 & -2 \\ 2 & 4 & 6 \end{bmatrix} \tag{10c}$$

To cancel out the third row we need  $r_3 - 2r_1$ , therefore  $l_{31} = 2$  and the updated U is

$$U = \begin{bmatrix} 1 & 3 & 2 \\ 0 & -4 & -2 \\ 0 & -2 & 2 \end{bmatrix} \tag{10d}$$

Lastly,  $r_3 - \frac{1}{2}r_2$  would zero out the last entry, therefore  $l_{32} = 1/2$ , and

$$U = \begin{bmatrix} 1 & 3 & 2 \\ 0 & -4 & -2 \\ 0 & 0 & 3 \end{bmatrix} \tag{10e}$$

Thus

$$A = LU = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 2 & 1/2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 3 & 2 \\ 0 & -4 & -2 \\ 0 & 0 & 3 \end{bmatrix}$$
 (10f)

Furthermore, we can obtain the A = LDU factorization, by dividing U by a diagonal matrix D that contains the pivots, namely

$$A = LDU = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 2 & 1/2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & -4 & 0 \\ 0 & 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & 3 & 2 \\ 0 & 1 & 1/2 \\ 0 & 0 & 1 \end{bmatrix}$$
(10g)

(b)

$$B = \begin{bmatrix} 1 & 0 & 1 \\ 2 & 2 & 2 \\ 3 & 4 & 5 \end{bmatrix} \tag{11a}$$

Using the same methodology as part(a), we use  $r_2 - 2r_1$  which gives  $l_{2,1} = 2$  and

$$U = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \\ 3 & 4 & 5 \end{bmatrix} \tag{11b}$$

Next, we need  $r_3 - 3r_1$ , therefore  $l_{3,1} = 3$  and

$$U = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \\ 0 & 4 & 2 \end{bmatrix} \tag{11c}$$

Lastly, we need  $r_3 - 2r_2$  giving  $l_{3,2} = 2$  and

$$U = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \tag{11d}$$

Thus,

$$B = LU = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$
 (11e)

and

$$B = LDU = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$
(11f)

(c)

$$\begin{bmatrix} 2 & 1 & 3 \\ 3 & 2 & 3 \\ 3 & 1 & 0 \end{bmatrix}$$
 (12a)

First, since there is a zero pivot in the third row we know were going to need a permutation matrix. A good chose for this problem would be to make the following row switches,

$$r_1 \to r_3$$
 (12b)

$$r_2 \to r_1$$
 (12c)

$$r_1 \to r_3$$
 (12d)

Therefore we get

$$PA = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 & 3 \\ 3 & 2 & 3 \\ 3 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 3 & 2 & 3 \\ 3 & 1 & 0 \\ 2 & 1 & 3 \end{bmatrix}$$
(12e)

Now we can apply the same methodology as above, starting with  $r_2 - r_1$ , therefore  $l_{2,1} = 1$  and

$$U = \begin{bmatrix} 3 & 2 & 3 \\ 0 & -1 & -3 \\ 2 & 1 & 3 \end{bmatrix} \tag{12f}$$

Next,  $r_3 - \frac{2}{3}r_1$ , therefore  $l_{3,1} = \frac{2}{3}$  and

$$U = \begin{bmatrix} 3 & 2 & 3 \\ 0 & -1 & -3 \\ 0 & -1/3 & 1 \end{bmatrix}$$
 (12g)

Lastly,  $r_3 - \frac{1}{3}r_2$ , therefore  $l_{3,2} = \frac{1}{3}$  and

$$U = \begin{bmatrix} 3 & 2 & 3 \\ 0 & -1 & -3 \\ 0 & 0 & 2 \end{bmatrix} \tag{12h}$$

Thus,

$$PA = LU = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 2/3 & 1/3 & 1 \end{bmatrix} \begin{bmatrix} 3 & 2 & 3 \\ 0 & -1 & -3 \\ 0 & 0 & 2 \end{bmatrix}$$
(12i)

and

$$PA = LDU = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 2/3 & 1/3 & 1 \end{bmatrix} \begin{bmatrix} 3 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & 2/3 & 1 \\ 0 & 1 & 3 \\ 0 & 0 & 1 \end{bmatrix}$$
(12j)

The following algorithm was used to perform the LU factorization

```
Algorithm 1: LU Factorization

Result: Factoring matrix A into L and U

Input: A
Output: L, U

1: L = I(3)
2: U = A

3: for k in range(0, M-1) do

4: | for j in range(k+1, M) do

5: | L[J,K] = U[J,K]/U[K,K]

6: | U[J,K:] = U[J,K:]-L[J,K]*U[K,K:]

7: | U[J,K:] = 0.0

8: | end

9: end
```

Secondly, we can plot the number of points versus computational time, as seen in Fig .(2), where we see how important it is to use built in Python packages.

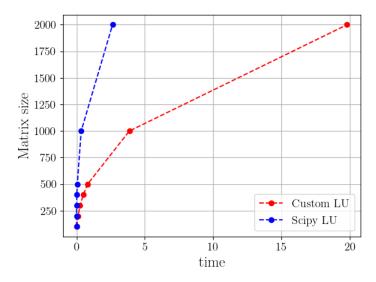


Figure 2: Performance study of user defined LU factorization tool

For an example of the complete code please refer to the Appendix.

The following inverses were calculated using the augmented matrix approach.

(a)  $A = \begin{bmatrix} 1 & 0 & 1 & 1 & 0 & 0 \\ 2 & 1 & 3 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{bmatrix}$  (13a)

First we can use  $2r_1 - r_2$  to cancel out the entries in the first column, giving

$$A = \begin{bmatrix} 1 & 0 & 1 & 1 & 0 & 0 \\ 0 & -1 & -1 & 2 & -1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{bmatrix}$$
 (13b)

Next we can zero out the entries in the third column using  $r_3 + r_2$  and  $r_3 - r_1$  giving

$$A = \begin{bmatrix} -1 & 0 & 0 & -1 & 0 & 1 \\ 0 & -1 & 0 & 2 & -1 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{bmatrix}$$
 (13c)

Thus,

$$A^{-1} = \begin{bmatrix} 1 & 0 & -1 \\ -2 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix}$$
 (13d)

(b)  $B = \begin{bmatrix} 2 & 1 & 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 & 1 & 0 \\ 1 & 1 & 2 & 0 & 0 & 1 \end{bmatrix}$  (14a)

First we can use  $r_2 - \frac{1}{2}r_1$  and  $r_3 - \frac{1}{2}r_1$  to cancel out the first in the first column, giving

$$B = \begin{bmatrix} 2 & 1 & 1 & 1 & 0 & 0 \\ 0 & 3/2 & 1/2 & -1/2 & 1 & 0 \\ 0 & 1/2 & 3/2 & -1/2 & 0 & 1 \end{bmatrix}$$
 (14b)

Next, we can use  $r_3 - \frac{1}{3}r_2$  to cancel out the entries of the second column beneath the pivot, giving

$$B = \begin{bmatrix} 2 & 1 & 1 & 1 & 0 & 0 \\ 0 & 3/2 & 1/2 & -1/2 & 1 & 0 \\ 0 & 0 & 4/3 & -1/3 & -1/3 & 1 \end{bmatrix}$$
 (14c)

Next, we can use  $r_2 - \frac{3}{8}r_3$  and  $r_1 - \frac{3}{4}r_3$  to cancel out the entries of the third column above the pivot, giving

$$B = \begin{bmatrix} 2 & 1 & 0 & 5/4 & 1/4 & -3/4 \\ 0 & 3/2 & 0 & -3/8 & 9/8 & -3/8 \\ 0 & 0 & 4/3 & -1/3 & -1/3 & 1 \end{bmatrix}$$
 (14d)

Lastly, we use  $r_1 - \frac{2}{3}r_2$  to zero out the entries above the pivot in the second column, giving

$$B = \begin{bmatrix} 2 & 0 & 0 & 3/2 & -1/2 & -1/2 \\ 0 & 3/2 & 0 & -3/8 & 9/8 & -3/8 \\ 0 & 0 & 4/3 & -1/3 & -1/3 & 1 \end{bmatrix}$$
 (14e)

Thus,

$$B^{-1} = \begin{bmatrix} 3/4 & -1/4 & -1/4 \\ -1/4 & 3/4 & -1/4 \\ -1/4 & -1/4 & 3/4 \end{bmatrix}$$
 (14f)

(c)

$$C = \begin{bmatrix} 2 & -1 & -1 & 1 & 0 & 0 \\ -1 & 2 & -1 & 0 & 1 & 0 \\ -1 & -1 & 2 & 0 & 0 & 1 \end{bmatrix}$$
 (15a)

Starting with  $r_2 + \frac{1}{2}r_1$  and  $r_3 + \frac{1}{2}r_1$  to cancel the entries beneath the pivot in the first column, giving

$$C = \begin{bmatrix} 2 & -1 & -1 & 1 & 0 & 0 \\ 0 & 3/2 & -3/2 & 1/2 & 1 & 0 \\ 0 & -3/2 & 3/2 & 1/2 & 0 & 1 \end{bmatrix}$$
(15b)

Next, we see that  $r_2 + r_3$  results in

$$C = \begin{bmatrix} 2 & -1 & -1 \\ 0 & 3/2 & -3/2 \\ 0 & 0 & 0 \end{bmatrix}$$
 (15c)

Thus, C is a singular matrix and has no inverse. Furthermore, there are many different ways to show a matrix is singular, therefore any solutions that rigorously proves that C is not invertible would be considered correct.

The inverse for each of the grids were calculated using the LU factorials of matrix, namely

$$A^{-1} = U^{-1}L^{-1} (16)$$

where the factorials were found using the subroutine discussed in Exercise 6. The following logic was used to find the inverse of matrix A:

#### **Algorithm 2:** $A^{-1}$ Subroutine

```
Result: Calculating A^{-1}
Input : A
Output: A^{-1}

1 [L, U] = \text{LU\_facrorization}(A)

2 Linv = lower\_inverse(L)
3 Uinv = upper\_inverse(L)
4 Ainv = Uinv \times Linv
```

Where the inverse subroutines use the following logic:

```
Algorithm 3: L^{-1} Subroutine
```

```
Result: Lower triangular inverse
Input: L
Output: L^{-1}

1 out = identity(M)
2 for j in range(0, M-1) do
3  | for i in range(j,M-1) do
4  | c = mat[i+1,j]
5  | out[i+1,:] = out[i+1,:]-c*out[j,:]
6 | end
7 end
```

#### Algorithm 4: $U^{-1}$ Subroutine

```
Result: Upper triangular inverse
Input: U
Output: U^{-1}

1 out = identity(M)
2 for j in range(M-1, -1, -1) do
3 | for i in range(j, 0, -1) do
4 | c = mat[i-1,j]/mat[j,j]
5 | out[i-1,:] = out[i-1,:]-c*out[j,:]
6 | end
7 | out[j,:] = out[j,:]/mat[j,j]
8 end
```

Using the above we get the following result shown in Fig. (3), where we can see once again the benefit in using the pre-compiled optimized code. Lastly, the full inverse code is provided in the Appendix.

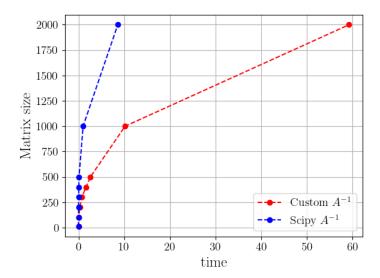


Figure 3: Performance study of user defined  $A^{-1}$  tool

16

# 9 Appendix

#### 9.1 Exercise 3 verification code

```
| #!/usr/bin/env python3
 Purpose:
   Verification of exercise 3
6 Author:
   Emilio Torres
 # Preamble
# Python packages
<sub>14</sub>| #-----#
15 import sys
16 import os
17 from subprocess import call
 from numpy import *
19 import matplotlib.pyplot as plt
20 | #-----
21 # User packages
 from ales_post.plot_settings import plot_setting
 #-----#
 # Main preamble
 #-----#
26
  __name__ == '__main__':
   #-----#
28
   # Main preamble
29
30
   call(['clear'])
31
   sep
          = os.sep
32
   pwd = os.getcwd()
33
   media_path = pwd + '%c..%cmedia%c'
                             %(sep, sep, sep)
34
35
   # Domain variables
36
    #-----#
37
   x = linspace(-10, -5, 100)
38
   y1 = 4./3. - 2./3*x
39
   y2 = 5./2.-x/2.
40
41
    # Plotting solution
42
    #-----#
43
    plot_setting()
44
   plt.plot([-10,-7], [6,6], 'k--', lw = 3.0)
45
    plt.plot([-7,-7], [4,6], k--1, [4,6]
46
   plt.plot(x,y1,'r', lw=1.5, label = 'y = 4/3 - 2/3 x')
47
   plt.plot(x,y2,'b', lw=1.5, label = 'y = 5/2 - 1/2 x')
48
49
```

```
# Plot settings
                                                                                   #
50
      #-----
51
      plt.legend(loc=0)
      plt.ylabel('$y$')
53
      plt.xlabel('$x$')
54
      plt.grid()
      plt.xlim([-10,-5])
56
57
      plt.ylim([4,8])
      plt.savefig(media_path + 'exercise-3.png')
58
      plt.close()
60
      print('**** Successful run ****')
61
      sys.exit(0)
62
```

#### 9.2 Exercise 6 LU factorization

```
1 #!/usr/bin/env python3
 The purpose of this script is to build subroutines to perform the
   following:
     1. LU factorization
  **** Note:
       This is pseudo code to help with Assignment 1
10
11 Author:
   Emilio Torres
 13
#-----#
# Preamble
#-----#
18 # Python packages
20 import os
21 import sys
22 from subprocess import call
23 import time
24 from numpy import copy, identity, random, zeros, array
25 import scipy.linalg as la
26 import matplotlib.pyplot as plt
# User defined functions
 #-----#
30
# Pretty print matrix
        ------
32
def print_matrix(
                # input matrix
     mat,
34
     var_str):
35
36
   """ Pretty printing a matrix """
37
38
```

```
# Looping over columns and rows
39
40
     out = ''
# initialize string
41
     for I in range(0, mat.shape[0]):
42
        for J in range(0, mat.shape[1]):
43
            out += '%12.5f' %(mat[I,J])
44
        out += \cdot \setminus n
45
     print(var_str)
46
     print(out)
47
48
 # Plotting settings
49
 #-----
51 def plot_setting():
52
     """ Useful plotting settings """
53
     #-----#
54
     # Plotting settings
     #-----#
56
     plt.rc('text', usetex=True)
57
     plt.rc('font', family='serif')
58
     SMALL_SIZE = 14
59
     MEDIUM_SIZE = 18
60
     BIGGER_SIZE = 20
61
    plt.rc('font', size=SMALL_SIZE)
plt.rc('axes', titlesize=SMALL_SIZE)
plt.rc('axes', labelsize=MEDIUM_SIZE)
plt.rc('xtick', labelsize=SMALL_SIZE)
plt.rc('ytick', labelsize=SMALL_SIZE)
plt.rc('legend', fontsize=SMALL_SIZE)
plt.rc('figure', titlesize=BIGGER_SIZE)
62
63
64
65
66
67
68
69 # - -
70 # LU factorization
71
 def LU_factorization(
73
                           # input matrix
74
     """ Calculating the LU factorization of the input vector """
75
77
     # Preallocating matrices
78
        79
       = copy(mat)
                          # make sure you copy matrix (No!!! U = mat )
80
     L = identity(M) # Initialize L with the identity matrix
81
     #-----#
     # Cheking it is error matrix
83
     #-----#
84
     if not mat.shape[0] == mat.shape[1]:
85
         print('Input matrix must be square')
86
        87
88
        sys.exit(8)
89
     # Calculating both L and U
90
91
     for K in range(0, M-1):
92
         for J in range(K+1, M):
93
```

```
= U[J,K]/U[K,K]
          L[J,K]
94
                  = U[J,K:]-L[J,K]*U[K,K:]
          U[J,K:]
95
          U[J,K]
                   = 0.0
96
97
98
    return (L, U)
# Main
100
 |#-----#
101
 if __name__ == '__main__':
102
    #-----#
    # Main preamble
104
    #-----
105
    call(['clear'])
106
    sep = os.sep
107
    pwd
             = os.getcwd()
108
    media_path = pwd + '%c..%cmedia%c' %(sep, sep, sep)
109
110
    # Testing LU
111
112
                = array([[3, 2, 3], [3, 1, 0], [2, 1, 3]])
113
    (Lower, Upper) = LU_factorization(A)
114
    print_matrix(Lower, 'L')
115
    print_matrix(Upper, 'U')
116
    print_matrix(A - (Lower@Upper), 'A-LU')
117
    sys.exit(8)
118
    #-----
119
    # Time study
120
    #-----#
121
         = [100, 200, 300, 400, 500, 1000, 2000]
    times = zeros(len(N))
123
    times2 = zeros(len(N))
124
    for k, i in enumerate(N):
125
                   = random.rand(i,i)
                                        # random matrix
126
                                        # start time
                   = time.time()
127
       tic
128
       (Lower, Upper) = LU_factorization(A)
                                       # LU
                   = time.time()
                                        # end time
129
       toc
                   = toc-tic
       times[k]
                                        # time elapsed
130
       tic
                   = time.time()
131
132
       (p, l, u)
                  = la.lu(A)
       toc
                   = time.time()
       times2[k]
                = toc-tic
134
    #-----
135
    # Plotting the solutions
136
    #-----#
137
    plot_setting()
138
    plt.plot(times, N, 'ro--', lw=1.5, label='Custom LU')
139
    plt.plot(times2, N, 'bo--', lw=1.5, label='Scipy LU')
140
    #-----
141
    # Plot settings
142
    #------
143
    plt.ylabel('Matrix size')
144
    plt.xlabel('time')
145
    plt.grid(True)
146
    plt.legend(loc=0)
147
    plt.savefig(media_path + 'exercise-6.png')
148
```

#### 9.3 Exercise 8 Inverse

```
1 #!/usr/bin/env python3
 Purpose:
    The purpose pf this script is to build subroutines to perform the
    following:
       1. Matrix inverse
 Author:
    Emilio Torres
 ______"""
 # Python packages
             -----#
16 import os
 import sys
18 from subprocess import call
19 import time
20 from numpy import copy, identity, random, zeros, array
21 import scipy.linalg as la
 import matplotlib.pyplot as plt
23 #------
24 # User defined functions
 #-----
  Pretty print matrix
                 -----#
28
 def print_matrix(
29
                      # input matrix
30
      mat,
       var_str):
31
32
    """ Pretty printing a matrix """
33
34
    # Looping over columns and rows
35
36
    out = ''
37
                     # initialize string
    for I in range(0, mat.shape[0]):
38
       for J in range(0, mat.shape[1]):
39
         out += '%25.5f' %(mat[I,J])
40
       out += ' \ n
41
    print(var_str)
42
    print(out)
44
  Plotting settings
                                                  ----#
```

```
47 def plot_setting():
48
     """ Useful plotting settings """
49
     #-----
50
     # Plotting settings
51
     #-----#
     plt.rc('text', usetex=True)
     plt.rc('font', family='serif')
54
     SMALL_SIZE = 14
     MEDIUM_SIZE = 18
56
     BIGGER_SIZE = 20
57
     plt.rc('font',
                      size=SMALL_SIZE)
58
     plt.rc('axes',
                      titlesize=SMALL_SIZE)
59
     plt.rc('axes',
                       labelsize=MEDIUM_SIZE)
60
     plt.rc('xtick',
                      labelsize=SMALL_SIZE)
61
     plt.rc('xtick', labelsize=SMALL_SIZE)
plt.rc('ytick', labelsize=SMALL_SIZE)
plt.rc('legend', fontsize=SMALL_SIZE)
plt.rc('figure', titlesize=BIGGER_SIZE)
62
63
64
65
   LU factorization
66
  def LU_factorization(
68
        \mathtt{mat}):
                           # inpuit matrix
69
70
     """ Calculating the LU factorization of the input vector """
71
     #-----
72
     # Preallocating matrices
                                                                    #
73
     #-----#
74
                          # matrix size
        = mat.shape[0]
75
     IJ
        = copy(mat)
                         # make sure you copy matrix (No!!! U = mat )
76
         = identity(M)  # Initialize L with the identity matrix
77
     # Cheking it is error matrix
79
80
     if not mat.shape[0] == mat.shape[1]:
81
         print('Input matrix must be square')
82
         print('Input --> [%i, %i]'
                                         %(mat.shape[0], mat.shape[1]))
83
         sys.exit(8)
84
85
     # Calculating both L and U
86
     #-----#
87
     for K in range(0, M-1):
88
         for J in range(K+1, M):
89
                    = U[J,K]/U[K,K]
            L[J,K]
90
            U[J,K:] = U[J,K:]-L[J,K]*U[K,K:]

U[J,K] = 0.0
91
            U[J,K]
                      = 0.0
92
93
94
     return L, U
95
   Diagonal inverse tool
  def diagonal_inverse(
98
       mat):
99
100
     """ Subroutine to calculate the inverse of a diagonal matrix """
101
```

22

```
102
103
     # Domain variables
104
        = mat.shape[0]
105
106
     out = idnetity(M)
     #-----#
107
     # Diagonal inverse
108
109
     for i in range(0, M):
110
        out[i,i] = 1/mat[i,i]
111
112
     return out
113
114 #-----
  # Upper matrix inverse
115
116
117
  def upper_inverse(
        mat):
118
119
     """ Subroutine to calculate the inverse of an upper diagonal matrix """
120
     # Domain variables
122
123
124
     M = mat.shape[0]
     out = identity(M)
     #-----#
126
     # Upper inverse
127
     #-----#
     for j in range(M-1, -1, -1):
129
        for i in range(j-1, -1, -1):
130
                    = mat[i,j]/mat[j,j]
131
                   = out[i,:]-c*out[j,:]
132
        out[j,:] = out[j,:]/mat[j,j]
133
134
135
136
   Upper matrix inverse
137
  def lower_inverse(
138
        mat):
139
140
     """ Subroutine to calculate the inverse of a lower diagonal matrix """
141
     #----#
142
     # Domain variables
143
144
145
       = mat.shape[0]
     out = identity(M)
146
     #-----#
147
     # Lower inverse
148
149
     for j in range(0, M-1):
150
151
        for i in range(j,M-1):
                    = mat[i+1,j]
           out[i+1,:] = out[i+1,:]-c*out[j,:]
153
154
     return out
156
```

```
157 # Inverse tool
                                                          #
  #-----
158
  def mat_inverse(
159
       mat):
160
     """ Subroutine to determine a matrix inverse """
161
     #----#
     # Calculating the inverse using LU factors
163
164
     N = mat.shape[0]
165
         = LU_factorization(mat)
     [L,U]
     #-----#
167
     # Calculating the inverse
168
169
     Linv
          = lower_inverse(L)
170
    Uinv = upper_inverse(U)
171
     inv = Uinv@Linv
172
173
174
    return inv
176
  #-----#
  if __name__ == '__main__':
178
     #-----#
179
     # Main preamble
180
     #-----#
181
     call(['clear'])
182
             = os.sep
           = os.getcwd()
184
     media_path = pwd + '%c..%cmedia%c'
                                        %(sep, sep, sep)
185
186
     # Testing LU
187
188
                    = 5
     N
189
190
     Α
                    = random.rand(N,N)
191
                    = mat_inverse(A)
                    = identity(N)
192
     print_matrix(A, 'A')
193
     print_matrix(Ainv, 'A^{-1}')
194
     print_matrix(I-A@Ainv, 'I-AxA^{-1}')
195
196
     # Time study
197
     #----#
198
          = [10, 100, 200, 300, 400, 500, 1000, 2000]
199
     times = zeros(len(N))
200
     times2 = zeros(len(N))
201
     for k, i in enumerate(N):
       print(i)
203
                                        # random matrix
       Α
                    = random.rand(i,i)
204
       tic
                                         # start time
                    = time.time()
205
       Ainv
                    = mat_inverse(A)
                                        # inv(A)
                   = time.time()
        toc
                                        # end time
207
        times[k]
                    = toc-tic
                                         # time elapsed
208
                   = time.time()
       tic
209
                   = la.inv(A)
       ainv
210
                    = time.time()
211
        toc
```

```
times2[k] = toc-tic
212
213
214
       # Plotting the solutions
       #-----#
215
       plot_setting()
216
       plt.plot(times, N, 'ro--', lw=1.5, label='Custom A^{-1}') plt.plot(times2, N, 'bo--', lw=1.5, label='Scipy A^{-1}')
217
218
219
       # Plot settings
220
       plt.ylabel('Matrix size')
222
       plt.xlabel('time')
223
       plt.grid(True)
224
       plt.legend(loc=0)
225
       plt.savefig(media_path + 'exercise-8.png')
226
       plt.close()
227
228
       print('**** Successful run ****')
229
       sys.exit(0)
230
```