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 **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

$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{b} = \begin{bmatrix} 1/6 & 1/4 & -1/6 \\ 7/18 & -1/12 & -1/18 \\ -1/3 & 0 & 1/3 \end{bmatrix} \begin{bmatrix} 5 \\ 6 \\ 3 \end{bmatrix} = \begin{bmatrix} 23/18 \\ 44/24 \\ -2/3 \end{bmatrix}$$
(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).

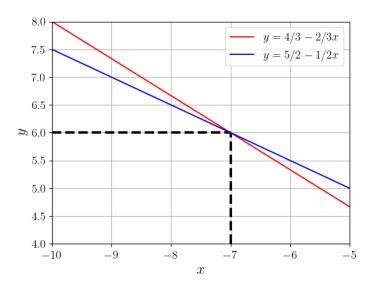


Figure 1: Verification plot for Exercise 3

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

$$I^T = I \tag{8d}$$

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

$$A = LDU = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 2 & 1/2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & -4 & 0 \\ 0 & 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & 3 & 2 \\ 0 & 1 & -2 \\ 0 & 0 & 1 \end{bmatrix}$$
(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 $l_{21} = 2$ 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 $l_{31} = 3$ and

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

Lastly, we need $r_3 - 2r_2$ giving $l_{32} = 2$ and

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

Thus,

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

and

$$B = LDU = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$
(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, $r_3 - \frac{1}{3}r_3$, therefore $l_{3,2} = \frac{1}{3}$ and

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

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

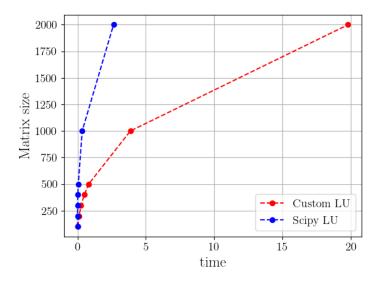


Figure 2: Performance study of user defined LU factorization tool

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.

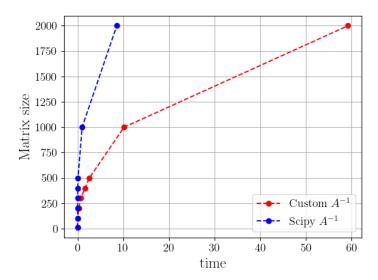


Figure 3: Performance study of user defined A^{-1} tool

9 Appendix

9.1 Exercise 3 verification code

```
1 #!/usr/bin/env python3
 Purpose:
   Verification of exercise 3
6 Author:
   Emilio Torres
 # Preamble
# Python packages
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
19 # - - - - -
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
```

19

```
L[J,K]
                    = U[J,K]/U[K,K]
94
                   = U[J,K:]-L[J,K]*U[K,K:]
           U[J,K:]
95
           U[J,K]
96
97
     return (L, U)
98
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
                 = random.rand(5,5)
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
     #-----
118
     # Time study
119
     #-----#
120
          = [100, 200, 300, 400, 500, 1000, 2000]
121
     times = zeros(len(N))
     times2 = zeros(len(N))
123
     for k, i in enumerate(N):
124
                    = random.rand(i,i)
                                          # random matrix
        Α
125
        tic
                     = time.time()
                                           # start time
126
        (Lower, Upper) = LU_factorization(A)
                                           # LU
127
128
        toc
                    = time.time()
                                           # end time
        times[k]
                    = toc-tic
                                           # time elapsed
129
                     = time.time()
        tic
130
                    = la.lu(A)
131
        (p, l, u)
132
        toc
                    = time.time()
                 = toc-tic
     #-----
134
     # Plotting the solutions
135
     #------
136
137
     plot_setting()
     plt.plot(times, N, 'ro--', lw=1.5, label='Custom LU')
138
     plt.plot(times2, N, 'bo--', lw=1.5, label='Scipy LU')
139
140
     # Plot settings
                                                             #
141
     #-----
142
     plt.ylabel('Matrix size')
143
     plt.xlabel('time')
144
     plt.grid(True)
145
     plt.legend(loc=0)
146
     plt.savefig(media_path + 'exercise-6.png')
147
148
     plt.show()
```

```
149 plt.close()
150
151 print('**** Successful run ****')
152 sys.exit(0)
```

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
 # Preamble
 # Python packages
16 import os
17 import sys
18 from subprocess import call
19 import time
20 from numpy import copy, identity, random, zeros, array
 import scipy.linalg as la
 import matplotlib.pyplot as plt
 # User defined functions
 #----#
 # Pretty print matrix
28 # - - - - -
 def print_matrix(
29
                       # input matrix
       mat,
30
       var_str):
31
32
    """ Pretty printing a matrix """
33
34
35
    # Looping over columns and rows
36
                     # initialize string
37
38
    for I in range(0, mat.shape[0]):
       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)
43
45 # Plotting settings
47 def plot_setting():
```

```
48
      """ Useful plotting settings """
49
      #-----
50
     # Plotting settings
51
      #------
52
      plt.rc('text', usetex=True)
      plt.rc('font', family='serif')
54
      SMALL_SIZE = 14
     MEDIUM_SIZE = 18
56
     BIGGER_SIZE = 20
57
                     size=SMALL_SIZE)
     plt.rc('font',
58
     plt.rc('axes',
                      titlesize=SMALL_SIZE)
59
     plt.rc('axes',
                      labelsize=MEDIUM_SIZE)
60
     plt.rc('xtick', labelsize=SMALL_SIZE)
plt.rc('ytick', labelsize=SMALL_SIZE)
plt.rc('legend', fontsize=SMALL_SIZE)
plt.rc('figure', titlesize=BIGGER_SIZE)
61
62
63
64
65
   LU factorization
67
  def LU_factorization(
68
                           # inpuit matrix
         mat):
69
70
     """ Calculating the LU factorization of the input vector """
71
      #-----
72
     # Preallocating matrices
73
      #-----#
74
        = mat.shape[0]
                          # matrix size
75
        = copy(mat)
                          # make sure you copy matrix (No!!! U = mat )
76
     L
        = identity(M) # Initialize L with the identity matrix
77
      #-----#
78
      # Cheking it is error matrix
79
80
     if not mat.shape[0] == mat.shape[1]:
81
         print('Input matrix must be square')
82
         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]
= U[J,K:]-L[J,K]*U[K,K:]
            L[J,K]
90
            U[J,K:]
91
             U[J,K]
92
93
     return L, U
94
95
  # Diagonal inverse tool
  def diagonal_inverse(
98
      mat):
99
100
     """ Subroutine to calculate the inverse of a diagonal matrix """
101
102
```

```
# Domain variables
                                                             #
103
     #-----
104
        = mat.shape[0]
     out = idnetity(M)
106
     #-----#
107
     # Diagonal inverse
108
     #-----
     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
  def upper_inverse(
117
       mat):
118
119
     """ Subroutine to calculate the inverse of an upper diagonal matrix """
120
121
     # Domain variables
123
     M = mat.shape[0]
124
     out = identity(M)
125
     #----#
     # Upper inverse
127
128
     for j in range (M-1, -1, -1):
        for i in range(j-1, -1, -1):
130
                    = mat[i,j]/mat[j,j]
131
           out[i,:] = out[i,:]-c*out[j,:]
132
        out[j,:] = out[j,:]/mat[j,j]
133
134
     return out
135
136
   Upper matrix inverse
137
  def lower_inverse(
138
       \mathtt{mat}):
139
140
     """ Subroutine to calculate the inverse of a lower diagonal matrix """
141
     # Domain variables
143
144
       = mat.shape[0]
145
     out = identity(M)
146
     #-----#
147
     # Lower inverse
148
149
     for j in range(0, M-1):
150
        for i in range(j,M-1):
151
152
                    = mat[i+1,j]
           out[i+1,:] = out[i+1,:]-c*out[j,:]
153
154
     return out
156 #
   Inverse tool
```

```
159
 def mat_inverse(
       \mathtt{mat}):
160
    """ Subroutine to determine a matrix inverse """
161
    #----#
162
    # Calculating the inverse using LU factors
163
    #-----
164
         = mat.shape[0]
165
    [L,U] = LU_factorization(mat)
166
    #-----#
167
    # Calculating the inverse
168
    #-----
         = lower_inverse(L)
    I.inv
170
         = upper_inverse(U)
171
         = Uinv@Linv
172
173
    return inv
174
 #-----#
175
176
 #-----#
177
 if __name__ == '__main__':
178
    #-----#
179
    # Main preamble
180
    #----#
181
    call(['clear'])
182
    sep
             = os.sep
183
             = os.getcwd()
184
                                       %(sep, sep, sep)
    media_path = pwd + '%c..%cmedia%c'
185
186
    # Testing LU
187
188
                  = 5
189
                   = random.rand(N,N)
190
    Ainv
                   = mat_inverse(A)
191
192
                   = identity(N)
    print_matrix(A, 'A')
193
    print_matrix(Ainv, 'A^{-1}')
194
    print_matrix(I-A@Ainv, 'I-AxA^{-1}')
195
    #------
196
197
    # Time study
    #-----#
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):
202
       print(i)
                   = random.rand(i,i)
                                      # random matrix
       Α
204
       tic
                   = time.time()
                                       # start time
205
                   = mat_inverse(A)
                                       # inv(A)
       Ainv
206
       toc
                   = time.time()
                                      # end time
                                      # time elapsed
       times[k]
                  = toc-tic
208
       tic
                   = time.time()
209
                   = la.inv(A)
       ainv
210
                   = time.time()
211
       toc
       times2[k]
                   = toc-tic
212
```

```
#-----#
213
                                                                         #
      # Plotting the solutions
214
215
      plot_setting()
216
      plt.plot(times, N, 'ro--', lw=1.5, label='Custom A^{-1}')
217
      plt.plot(times2, N, 'bo--', lw=1.5, label='Scipy $A^{-1}$')
218
219
      # Plot settings
220
221
      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
```