

LinearSystemsTimings

January 7, 2023

```
[ ]: import numpy as np
from BNumMet.LinearSystems import permute
import time
import matplotlib.pyplot as plt
import pandas as pd
from scipy.linalg import lu
```

```
[ ]: def lu1(Matrix): # LU decomposition with np subarrays
    if Matrix.shape[0] != Matrix.shape[1]:
        raise ValueError("Matrix must be square")

    n = Matrix.shape[0] # number of rows/columns
    A = Matrix.copy().astype(float) # Make a copy of A and convert to float
    P = np.eye(n, dtype=float) # Initialize P as the identity matrix

    # Loop over the columns of A
    for col in range(n):
        # Find the index of the row with the largest pivot element
        iMax = int(np.argmax(np.abs(A[col:, col])) + col)

        # Skip if the pivot is zero
        if A[iMax, col] != 0:
            # Swap the rows of A and P for those with the largest pivot element
            A = permute(A, iMax, col)
            P = permute(P, iMax, col)

            pivot = A[col, col]

            A[col + 1 :, col] = A[col + 1 :, col] / pivot
            A[col + 1 :, col + 1 :] = A[col + 1 :, col + 1 :] - (
                A[col + 1 :, :][:, [col]] @ A[[col], :][:, col + 1 :]
            ) # It is the same as the matrix multiplication (A[col + 1 :, col][:,
↪ np.newaxis] @ A[col, col + 1 :][np.newaxis, :]) or np.outer(A[col + 1 :,
↪ col], A[col, col + 1 :])

    L = np.tril(A, -1) + np.eye(
        n
```

```

    ) # Extract the lower triangular matrix from A and add the identity matrix
    ↪to make it a proper lower triangular matrix
    U = np.triu(A) # Extract the upper triangular matrix from A

    return P, L, U

def lu2(Matrix): # LU decomposition with np newaxis syntax
    if Matrix.shape[0] != Matrix.shape[1]:
        raise ValueError("Matrix must be square")

    n = Matrix.shape[0] # number of rows/columns
    A = Matrix.copy().astype(float) # Make a copy of A and convert to float
    P = np.eye(n, dtype=float) # Initialize P as the identity matrix

    # Loop over the columns of A
    for col in range(n):
        # Find the index of the row with the largest pivot element
        iMax = int(np.argmax(np.abs(A[col:, col])) + col)

        # Skip if the pivot is zero
        if A[iMax, col] != 0:
            # Swap the rows of A and P for those with the largest pivot element
            A = permute(A, iMax, col)
            P = permute(P, iMax, col)

            pivot = A[col, col]

            A[col + 1 :, col] = A[col + 1 :, col] / pivot
            A[col + 1 :, col + 1 :] = (
                A[col + 1 :, col][:, np.newaxis] @ A[col, col + 1 :][np.
    ↪newaxis, :]
            ) # It is the same as the matrix multiplication (A[col + 1 :, col][:
    ↪, np.newaxis] @ A[col, col + 1 :][np.newaxis, :]) or np.outer(A[col + 1 :,
    ↪col], A[col, col + 1 :])

            L = np.tril(A, -1) + np.eye(
                n
            ) # Extract the lower triangular matrix from A and add the identity matrix
    ↪to make it a proper lower triangular matrix
            U = np.triu(A) # Extract the upper triangular matrix from A

            return P, L, U

def lu3(Matrix): # LU decomposition with np outer syntax
    if Matrix.shape[0] != Matrix.shape[1]:

```

```

        raise ValueError("Matrix must be square")

    n = Matrix.shape[0] # number of rows/columns
    A = Matrix.copy().astype(float) # Make a copy of A and convert to float
    P = np.eye(n, dtype=float) # Initialize P as the identity matrix

    # Loop over the columns of A
    for col in range(n):
        # Find the index of the row with the largest pivot element
        iMax = int(np.argmax(np.abs(A[col:, col])) + col)

        # Skip if the pivot is zero
        if A[iMax, col] != 0:
            # Swap the rows of A and P for those with the largest pivot element
            A = permute(A, iMax, col)
            P = permute(P, iMax, col)

            pivot = A[col, col]

            A[col + 1 :, col] = A[col + 1 :, col] / pivot
            A[col + 1 :, col + 1 :] = np.outer(
                A[col + 1 :, col], A[col, col + 1 :]
            ) # It is the same as the matrix multiplition (A[col + 1 :, col][:
↪, np.newaxis] @ A[col, col + 1 :][np.newaxis, :]) or np.outer(A[col + 1 :,
↪col], A[col, col + 1 :])

            L = np.tril(A, -1) + np.eye(
                n
            ) # Extract the lower triangular matrix from A and add the identity matrix
↪to make it a proper lower triangular matrix
            U = np.triu(A) # Extract the upper triangular matrix from A

    return P, L, U

def lu4(Matrix): # LU decomposition with for loops
    if Matrix.shape[0] != Matrix.shape[1]:
        raise ValueError("Matrix must be square")

    n = Matrix.shape[0] # number of rows/columns
    A = Matrix.copy().astype(float) # Make a copy of A and convert to float
    P = np.eye(n).astype(float) # Initialize P as the identity matrix

    # Loop over the columns of A
    for col in range(n):
        # Find the index of the row with the largest pivot element
        iMax = int(np.argmax(np.abs(A[col:, col])) + col)

```

```

    # Skip if the pivot is zero
    if A[iMax, col] != 0:
        # Swap the rows of A and P for those with the largest pivot element
        A = permute(A, iMax, col)
        P = permute(P, iMax, col)

    # Gaussian elimination
    for row in range(col + 1, n):
        A[row, col] = (
            A[row, col] / A[col, col]
        ) # Calculate the multiplier and store in A for later use
        A[row, col + 1 :] = (
            A[row, col + 1 :] - A[row, col] * A[col, col + 1 :]
        ) # Update the remaining elements in the row using the
↪multiplier

    L = np.tril(A, -1) + np.eye(
        n
    ) # Extract the lower triangular matrix from A and add the identity matrix
↪to make it a proper lower triangular matrix
    U = np.triu(A) # Extract the upper triangular matrix from A

    return P, L, U

```

```

[ ]: functions = [
    (lu1, "Submatrices"),
    (lu2, "NewAxis"),
    (lu3, "Outer"),
    (lu4, "ForLoops"),
    (lu, "Scipy"),
]
sizes = [2**i for i in range(2, 14)]
results = []
for size in sizes:
    print(f"{size:_^50}")
    A = np.random.rand(size, size)
    for func, name in functions:

        start = time.time()
        func(A)
        end = time.time()
        results.append((name, size, end - start))
    print(f"\t {name} - {end - start:.3f}")

```

4

Submatrices - 0.001
NewAxis - 0.000

```

Outer - 0.000
ForLoops - 0.000
Scipy - 0.001
-----8-----
Submatrices - 0.000
NewAxis - 0.000
Outer - 0.000
ForLoops - 0.000
Scipy - 0.001
-----16-----
Submatrices - 0.000
NewAxis - 0.001
Outer - 0.000
ForLoops - 0.001
Scipy - 0.001
-----32-----
Submatrices - 0.001
NewAxis - 0.002
Outer - 0.001
ForLoops - 0.004
Scipy - 0.000
-----64-----
Submatrices - 0.003
NewAxis - 0.003
Outer - 0.003
ForLoops - 0.011
Scipy - 0.000
-----128-----
Submatrices - 0.009
NewAxis - 0.006
Outer - 0.005
ForLoops - 0.028
Scipy - 0.162
-----256-----
Submatrices - 0.035
NewAxis - 0.027
Outer - 0.016
ForLoops - 0.133
Scipy - 0.393
-----512-----
Submatrices - 0.224
NewAxis - 0.177
Outer - 0.129
ForLoops - 0.467
Scipy - 0.268
-----1024-----
Submatrices - 2.199
NewAxis - 1.244

```

```

Outer - 1.051
ForLoops - 1.961
Scipy - 0.029
-----2048-----
Submatrices - 16.728
NewAxis - 9.820
Outer - 8.374
ForLoops - 9.510
Scipy - 0.151
-----4096-----
Submatrices - 135.769
NewAxis - 75.582
Outer - 63.191
ForLoops - 51.393
Scipy - 0.811
-----8192-----
Submatrices - 1023.253
NewAxis - 603.968
Outer - 505.603
ForLoops - 302.093
Scipy - 3.854

```

```

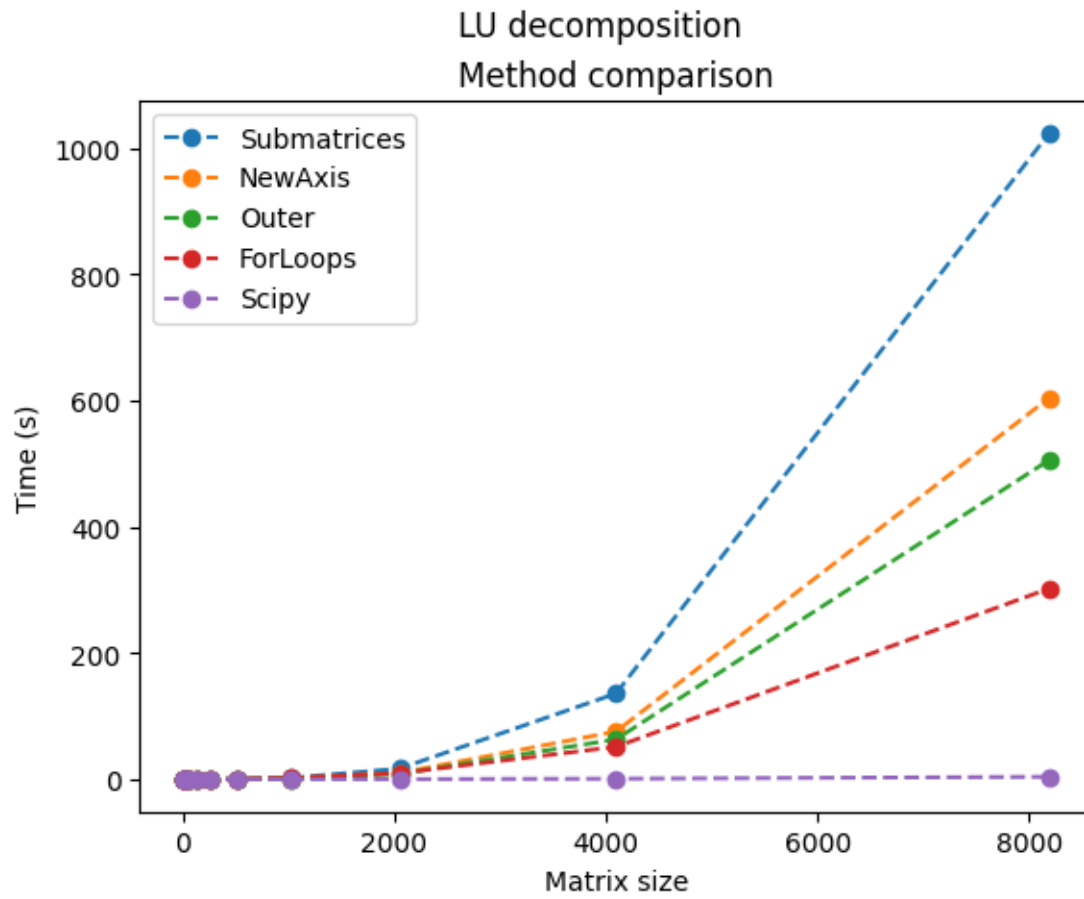
[ ]: import matplotlib.pyplot as plt

for fun in functions:

    plt.plot(
        [i[1] for i in results if i[0] == fun[1]],
        [i[2] for i in results if i[0] == fun[1]],
        "o--",
        label=fun[1],
    )
    plt.loglog(
        [i[1] for i in results if i[0] == fun[1]],
        [i[2] for i in results if i[0] == fun[1]],
        "o--",
        label=fun[1],
    )

plt.legend()
plt.xlabel("Matrix size")
plt.ylabel("Time (s)")
plt.suptitle("LU decomposition")
plt.title("Method comparison")
plt.show()

```



SCIPY : https://netlib.org/lapack/explore-html/dd/d9a/group___double_g_ecomputational_ga0019443faea0827

Uses internal LAPACK routines to solve a system of linear equations with a square coefficient matrix. <https://epubs.siam.org/doi/10.1137/S0895479896297744> Sivan Toledo, 1996