

Numerical Analysis: homework 02

Due on Tuesday, March 14, 2017

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1 Introduction

To solve such an equation:

$$Ax = b \quad (1)$$

where A is $n \times n$ given nonsingular matrix and b is a given vector, while x is a vector to solve. Although Gaussian Elimination is an easy method for this problem, everytime when we adjust b , we have to re-calculate it again, which will be time-consuming. As a result, LU Decomposition provides a more flexible method to solve this problem.

1.1 LU Decomposition

Given a $n \times n$ matrix A , we can split it by:

$$A = L \times U \quad (2)$$

where L is lower-triangular matrix and U is upper-triangular matrix. And we can express L and U by this format:

$$L = \begin{bmatrix} 1 & 0 & 0 \\ u_{21} & 1 & 0 \\ u_{31} & 0 & 1 \end{bmatrix} \quad (3)$$

$$U = \begin{bmatrix} l_{11} & l_{12} & l_{13} \\ 0 & l_{22} & l_{23} \\ 0 & 0 & l_{33} \end{bmatrix} \quad (4)$$

So

$$A = \begin{bmatrix} 1 & 0 & 0 \\ u_{21} & 1 & 0 \\ u_{31} & 0 & 1 \end{bmatrix} \times \begin{bmatrix} l_{11} & l_{12} & l_{13} \\ 0 & l_{22} & l_{23} \\ 0 & 0 & l_{33} \end{bmatrix}$$

After we find out L and U , we can express problem by

$$LUx = b$$

Let $Y = Ux$, then

$$LY = b$$

Y can be easily calculated, this is called **Forward Substitution**. After getting Y , the equation will be $Ux = Y$. x can be also easily calculated, this is called **Backward substitution**.

1.2 Error Calculation

Since the accuracy of our algorithm is quite important, I use square error to evaluate the performance:

$$Error = \sqrt{\sum_{i=1}^n (b'[i] - b[i])^2} \quad (5)$$

where b is the given vector and b' is the multiplication of A and x' which is calculated by our algorithm.

2 C++ Implementation

2.1 LU Decomposition

```

MAT &luFact(MAT &m1){
    for(int i=0; i<m1.m; i++){
        for(int j=i+1; j < m1.m; j++){
            m1[j][i] /= m1[i][i];
5          }
        for(int j=i+1; j < m1.m; j++){
            for(int k=i+1; k < m1.m; k++){
                m1[j][k] -= m1[j][i] * m1[i][k];
10          }
        }
    }
    return m1;
}

```

2.2 Forward Substitution

```

VEC fwdSubs(MAT &m1, VEC b){
    VEC Y(b);
    for(int i=0; i < m1.m; i++){
        for(int j=0; j < i; j++){
5          Y[i] -= m1[i][j] * Y[j];
        }
    }
    return Y;
}

```

2.3 Backward Substitution

```

VEC bckSubs(MAT &m1, VEC b){
    VEC X(b);
    for(int i=m1.m-1; i >= 0; i--){
        for(int j=m1.m-1; j > i; j--){
5          X[i] -= m1[i][j] * X[j];
        }
        X[i] /= m1[i][i];
    }
    return X;
10 }

```

3 Complexity

3.1 LU Decomposition

In the outer double for-loop, we do $\frac{n(n+1)}{2}$ times of one additional for-loop, so this is a $O(n^3)$ problem.

3.2 Forward/Backward substitution

The double for-loop do $\frac{n(n+1)}{2}$ times of calculation, so this is a $O(n^2)$ problem.

4 Discussion

4.1 Performance Evaluation

In this project, I use 5 numbers to indicate the accuracy and efficiency of our method.

1. **Error**(square error of the calculated result and answer)
2. **Runtime**(total execution time of program)
3. **LU**(total runtime consumed by LU Decomposition)
4. **FWD**(total runtime consumed by Forward Substitution)
5. **BCK**(total runtime consumed by Backward Substitution)

Table 1 shows detailed result of m3 to m10 execution result.

Table 1: Execution result of all matrix

	m3.dat	m4.dat	m5.dat	m6.dat	m7.dat	m8.dat	m9.dat	m10.dat
N	3	10	100	200	400	800	1600	3200
Error	0	1.11E-14	1.65E-10	2.52E-09	3.62E-08	6.30E-07	1.09E-05	0.000152939
Runtime(s)	0.005	0.004	0.007	0.026	0.162	1.137	9.343	67.129
LU(s)	0	2.00E-06	0.00187	0.015341	0.130141	1.02409	8.895	65.4746
FWD(s)	1.00E-06	1.00E-06	4.10E-05	0.000121	0.000467	0.002353	0.007206	0.028792
BCK(s)	0	1.00E-06	2.90E-05	0.00012	0.000479	0.001957	0.007655	0.03202

Figure 1: Error vs N^3

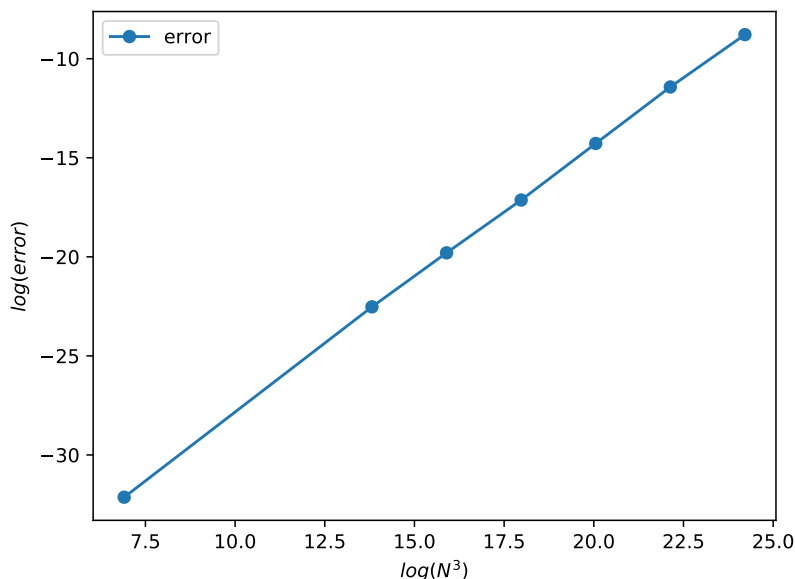


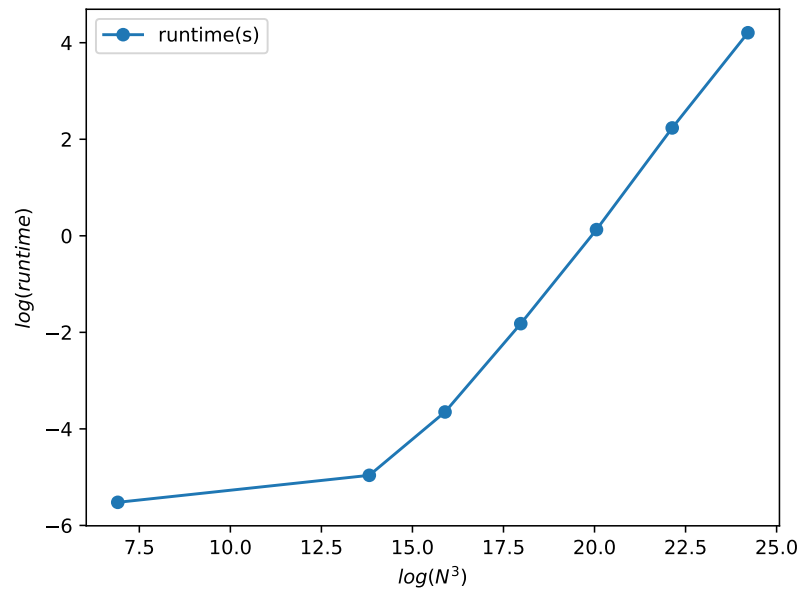
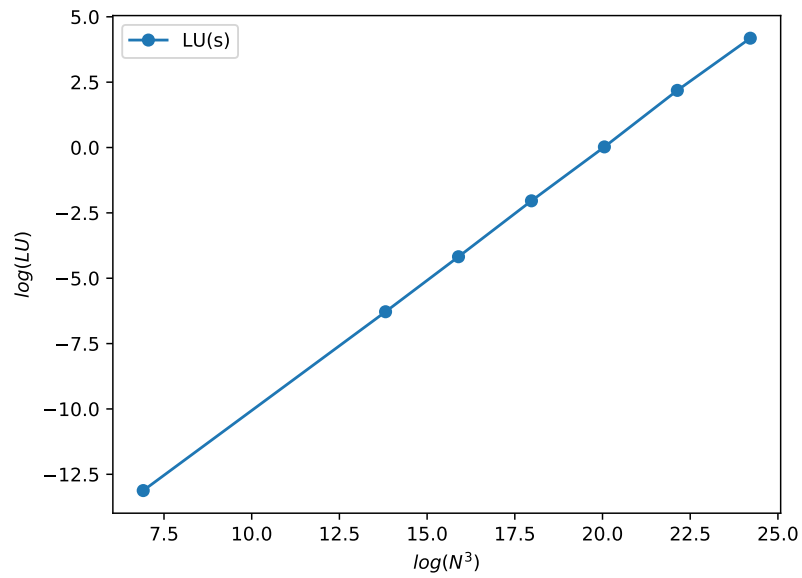
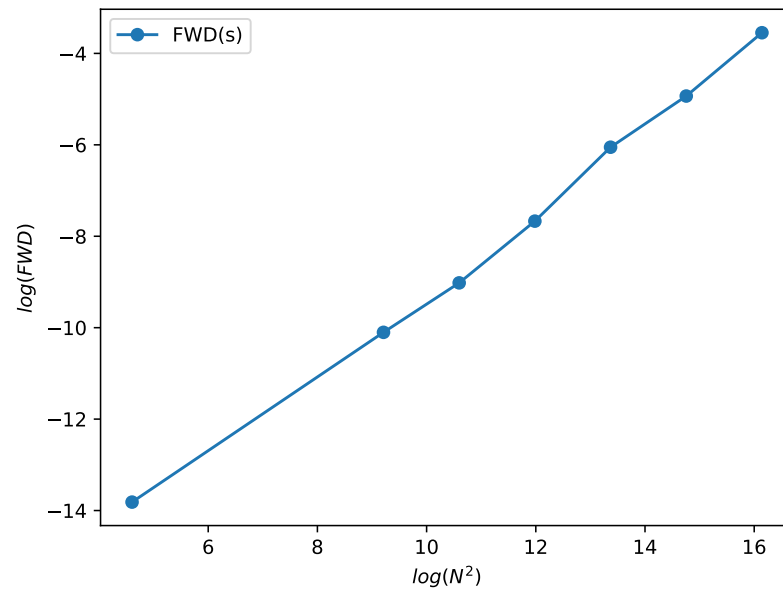
Figure 2: Runtime vs N^3 Figure 3: LU vs N^3 

Figure 4: FWD vs N^2 Figure 5: BCK vs N^2 