

Tips for saving computations (cont'd)

Given $\mathbf{A}, \tilde{\mathbf{A}} \in \mathbb{R}^{m \times n}$, $\mathbf{B} \in \mathbb{R}^{n \times p}$, $\mathbf{x} \in \mathbb{R}^n$,

- $\alpha \mathbf{A}$: $nnz(\mathbf{A})$ ✗ $\neq 0$
- $\mathbf{A} + \tilde{\mathbf{A}}$: between 0 and $\min\{nnz(\mathbf{A}), nnz(\tilde{\mathbf{A}})\}$
 $\Rightarrow O(\min\{nnz(\mathbf{A}), nnz(\tilde{\mathbf{A}})\})$
- $\mathbf{A}\mathbf{x}$ with dense \mathbf{x} : $nnz(\mathbf{A})$ multiplications and a number of additions that is no more than $nnz(\mathbf{A})$, so between $nnz(\mathbf{A})$ and $2nnz(\mathbf{A})$ flops
 $\Rightarrow O(nnz(\mathbf{A}))$
 - For diagonal \mathbf{A} , only $nnz(\mathbf{A})$ multiplications are needed, no additions, so $nnz(\mathbf{A})$ flops
- \mathbf{AB} : At most $2 \min\{nnz(\mathbf{A})p, nnz(\mathbf{B})m\}$ flops
 $\Rightarrow O(\min\{nnz(\mathbf{A})p, nnz(\mathbf{B})m\})$

may not reach the upper bound depend on the distribution of nonzero elements

Reference: S. Boyd and L. Vandenberghe, *Introduction to Applied Linear Algebra – Vectors, Matrices, and Least Squares*, 2018. Available online at <https://web.stanford.edu/~boyd/vmls/vmls.pdf>

Matrix Computations

Chapter 2 Linear systems and LU decomposition

Section 2.1 Triangular Systems and LU Decomposition

Jie Lu
ShanghaiTech University

System of Linear Equations

Consider the system of linear equations (linear system)

$$\mathbf{Ax} = \mathbf{b}$$

- $\mathbf{A} \in \mathbb{R}^{n \times n}$ and $\mathbf{b} \in \mathbb{R}^n$ are given
- $\mathbf{x} \in \mathbb{R}^n$ is the solution to the system
- Extension to the complex case is simple

Solving the Linear System

Goal: Find the solution to $\mathbf{Ax} = \mathbf{b}$ in a numerically efficient way

- The problem is very easy if \mathbf{A} is nonsingular and \mathbf{A}^{-1} is known
 - How to compute \mathbf{A}^{-1} efficiently?
- Solving the linear system may be easier in some special cases, e.g., triangular \mathbf{A} , orthogonal \mathbf{A} , circulant \mathbf{A}

Lower Triangular Systems

Example: Consider the 3×3 lower triangular system

$$\begin{bmatrix} \ell_{11} & 0 & 0 \\ \ell_{21} & \ell_{22} & 0 \\ \ell_{31} & \ell_{32} & \ell_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

If $\ell_{11}, \ell_{22}, \ell_{33} \neq 0$, then

- The first equation gives $x_1 = b_1/\ell_{11}$
- The second equation gives $x_2 = (b_2 - \ell_{21}x_1)/\ell_{22}$. Then, substituting x_1 yields x_2
- The third equation gives $x_3 = (b_3 - \ell_{31}x_1 - \ell_{32}x_2)/\ell_{33}$. Then, substituting x_1, x_2 yields x_3

Question: What happens if some of $\ell_{11}, \ell_{22}, \ell_{33}$ is zero?

Forward Substitution

For a general lower triangular system $\mathbf{Lx} = \mathbf{b}$ with $\mathbf{L} \in \mathbb{R}^{n \times n}$,

$$x_i = \left(b_i - \sum_{j=1}^{i-1} \ell_{ij} x_j \right) / \ell_{ii}, \quad \text{for } i = 1, 2, \dots, n$$

The algorithm is called **Forward Substitution** for solving $\mathbf{Lx} = \mathbf{b}$

Forward substitution in MATLAB form:

```
function b = ForwardSubstitution(L,b)
n= length(b);
x= zeros(n,1);
b(1)= b(1)/L(1,1);
for i=2:n
    b(i)=(b(i)-L(i,1:i-1)*b(1:i-1))/L(i,i);
end
```

Overall flops

$$1 + \sum_{i=2}^n (1 + 2(i-1) + 1) = \sum_{i=1}^n (2i - 1) = n^2$$

- Complexity: n^2 flops
- You may overwrite b with the solution to save memory

Upper Triangular Systems

Example: Consider the 3×3 upper triangular system

$$\begin{bmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

If $u_{11}, u_{22}, u_{33} \neq 0$, then

- The third equation gives $x_3 = b_3/u_{33}$
- The second equation gives $x_2 = (b_2 - u_{23}x_3)/u_{22}$. Then, substituting x_3 yields x_2
- The first equation gives $x_1 = (b_1 - u_{12}x_2 - u_{13}x_3)/u_{11}$. Then, substituting x_3, x_2 yields x_1

Question: What happens if some of u_{11}, u_{22}, u_{33} is zero?

Backward Substitution

For a general upper triangular system $\mathbf{U}\mathbf{x} = \mathbf{b}$ with $\mathbf{U} \in \mathbb{R}^{n \times n}$,

$$x_i = \left(b_i - \sum_{j=i+1}^n u_{ij}x_j \right) / u_{ii}, \quad \text{for } i = n, n-1, \dots, 1$$

The algorithm is called **Backward Substitution** for solving $\mathbf{U}\mathbf{x} = \mathbf{b}$

Backward substitution in MATLAB form:

```
function x= BackwardSubstitution(U,b)
n= length(b);
x= zeros(n,1);
x(n)= b(n)/U(n,n);
for i= n-1:-1:1,
    x(i)= (b(i)- U(i,i+1:n)*x(i+1:n))/U(i,i);
end
```

$$1 + \sum_{i=1}^{n-1} (1 + 2(n-i) - 1 + 1) = n^2$$

- complexity: n^2 flops
- You may overwrite b with the solution to save memory

Column-Oriented Representation

Example: Consider the 3×3 lower triangular system

$$\begin{bmatrix} 2 & 0 & 0 \\ 1 & 5 & 0 \\ 7 & 9 & 8 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 6 \\ 2 \\ 5 \end{bmatrix}$$

From the first equation, we have $x_1 = 6/2 = 3$. Then the remaining two equations can be expressed as

$$\begin{bmatrix} 5 & 0 \\ 9 & 8 \end{bmatrix} \begin{bmatrix} x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ 5 \end{bmatrix} - x_1 \begin{bmatrix} 1 \\ 7 \end{bmatrix} = \begin{bmatrix} 2 \\ 5 \end{bmatrix} - 3 \begin{bmatrix} 1 \\ 7 \end{bmatrix} = \begin{bmatrix} -1 \\ -16 \end{bmatrix}$$

For a general $n \times n$ lower triangular system $\mathbf{L}\mathbf{x} = \mathbf{b}$, x_1 can be directly obtained. Then, form a $(n-1) \times (n-1)$ system according to

$$\mathbf{L}(2:n, 2:n)\mathbf{x}(2:n) = \mathbf{b}(2:n) - x_1 \cdot \mathbf{L}(2:n, 1)$$

Solving this new system for x_2 is simple

Repeated the process for the $(n-1) \times (n-1)$ system

Column-Oriented Representation (cont'd)

Column-Oriented Forward Substitution in MATLAB form:

```
for j=1:n-1
    b(j)=b(j)/L(j,j);
    % Compute the first element of the solution to the
    % latest system
    b(j+1:n)=b(j+1:n)-b(j)*L(j+1:n,j);
    % The right-hand side of the updated system
end
b(n)=b(n)/L(n,n);
% b has been overwritten by the solution
```

part of RHS of latest system solution x_j

subtraction multiplication

$$\sum_{j=1}^{n-1} (1 + (n-j) + (n-j))$$

$$+ 1 = n^2$$

Complexity: n^2 flops

Exercise: Derive Column-Oriented Backward Substitution for solving upper triangular systems

- See Section 3.1 of the textbook

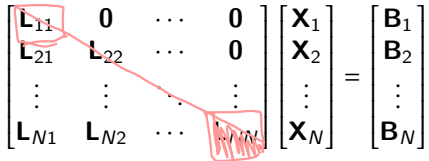
Multi-Right-Hand-side Problems

Compute the solution $\mathbf{X} \in \mathbb{R}^{n \times q}$ to

$$\mathbf{L}\mathbf{X} = \mathbf{B}$$

where $\mathbf{L} \in \mathbb{R}^{n \times n}$ is lower triangular and $\mathbf{B} \in \mathbb{R}^{n \times q}$

It amounts to solving q triangular systems, but we can do [Block Back Substitution](#). Partitioning $\mathbf{L}\mathbf{X} = \mathbf{B}$ into


$$\begin{bmatrix} \boxed{\mathbf{L}_{11}} & \mathbf{0} & \cdots & \mathbf{0} \\ \mathbf{L}_{21} & \mathbf{L}_{22} & \cdots & \mathbf{0} \\ \vdots & \vdots & \ddots & \vdots \\ \mathbf{L}_{N1} & \mathbf{L}_{N2} & \cdots & \boxed{\mathbf{L}_{NN}} \end{bmatrix} \begin{bmatrix} \mathbf{X}_1 \\ \mathbf{X}_2 \\ \vdots \\ \mathbf{X}_N \end{bmatrix} = \begin{bmatrix} \mathbf{B}_1 \\ \mathbf{B}_2 \\ \vdots \\ \mathbf{B}_N \end{bmatrix}$$

L_{ii} lower triangular

Multi-Right-Hand-side Problems (cont'd)

Solve the triangular system $\mathbf{L}_{11}\mathbf{X}_1 = \mathbf{B}_1$ for \mathbf{X}_1 . Then, remove \mathbf{X}_1 from block equations 2 through N :

$$\begin{bmatrix} \mathbf{L}_{22} & \mathbf{0} & \cdots & \mathbf{0} \\ \mathbf{L}_{32} & \mathbf{L}_{33} & \cdots & \mathbf{0} \\ \vdots & \vdots & \ddots & \vdots \\ \mathbf{L}_{N2} & \mathbf{L}_{N3} & \cdots & \mathbf{L}_{NN} \end{bmatrix} \begin{bmatrix} \mathbf{X}_2 \\ \mathbf{X}_3 \\ \vdots \\ \mathbf{X}_N \end{bmatrix} = \begin{bmatrix} \mathbf{B}_2 \\ \mathbf{B}_3 \\ \vdots \\ \mathbf{B}_N \end{bmatrix} - \begin{bmatrix} \mathbf{L}_{21} \\ \mathbf{L}_{31} \\ \vdots \\ \mathbf{L}_{N1} \end{bmatrix} \mathbf{X}_1$$

Repeat this process to the above system

```
pseudo code, not Matlab
```

```
for j=1:N
```

```
    Solve  $\mathbf{L}_{jj}\mathbf{X}_j = \mathbf{B}_j$ ;
```

```
    for i=j+1:N
```

```
         $\mathbf{B}_i = \mathbf{B}_i - \mathbf{L}_{ij}\mathbf{X}_j$ ;
```

```
    end
```

```
end
```

LU Decomposition

A “high-level” algebraic description of Gaussian Elimination

LU decomposition : Given $\mathbf{A} \in \mathbb{R}^{n \times n}$, find $\mathbf{L}, \mathbf{U} \in \mathbb{R}^{n \times n}$ such that

$$\mathbf{A} = \mathbf{LU}, \quad \text{where}$$

$\mathbf{L} \in \mathbb{R}^{n \times n}$ is unit lower triangular ($\ell_{ii} = 1$ for all i)

$\mathbf{U} \in \mathbb{R}^{n \times n}$ is upper triangular

Suppose \mathbf{A} has an LU decomposition. Then, solving $\mathbf{Ax} = \mathbf{b}$ can be recast as solving two triangular systems

1. solve $\mathbf{Lz} = \mathbf{b}$ for \mathbf{z}
2. solve $\mathbf{Ux} = \mathbf{z}$ for \mathbf{x}

Questions:

- Does LU decomposition always exist?
- How to find \mathbf{L} and \mathbf{U} ?

$$\begin{array}{c} \mathbf{Ax} = \mathbf{b} \\ \Downarrow \\ \underbrace{\mathbf{LU}}_{\mathbf{z}} \mathbf{x} = \mathbf{b} \\ \Downarrow \\ \mathbf{Ux} = \mathbf{z}, \quad \mathbf{Lz} = \mathbf{b} \end{array}$$

Gauss Transformations

A matrix description of the zeroing process in Gaussian elimination

Example: Suppose $x_1 \neq 0$ and $\tau = x_2/x_1$. Then,

$$\begin{bmatrix} 1 & 0 \\ -\tau & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ 0 \end{bmatrix}$$

Extension to \mathbb{R}^n : Let $\mathbf{x} \in \mathbb{R}^n$ s.t. $x_k \neq 0$ for some $1 \leq k \leq n$ and $\tau_i = \frac{x_i}{x_k}$ $\forall i = k+1, \dots, n$. Then, *kth column*

k x k identity matrix

$$\underbrace{\begin{bmatrix} 1 & & & & \\ & \ddots & & & \\ & & 1 & & \\ & & -\tau_{k+1} & 1 & \\ & & \vdots & & \ddots \\ & & -\tau_n & & & 1 \end{bmatrix}}_{\mathbf{M}_k} \begin{bmatrix} x_1 \\ \vdots \\ x_k \\ x_{k+1} \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} x_1 \\ \vdots \\ x_k \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

unit lower triangular

(k+1)-row:

$$\begin{aligned} & -\tau_{k+1} x_k + x_{k+1} \\ & = -\frac{x_{k+1}}{x_k} x_k + x_{k+1} \\ & = 0 \end{aligned}$$

Gauss Transformations (cont'd)

For $k = 1, \dots, n$,

$$\mathbf{M}_k \mathbf{x} = \begin{bmatrix} 1 & & & & & \\ & \ddots & & & & \\ & & 1 & & & \\ & & -\tau_{k+1} & 1 & & \\ & & \vdots & & \ddots & \\ & & -\tau_n & & & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_k \\ x_{k+1} \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} x_1 \\ \vdots \\ x_k \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad \tau_i = \frac{x_i}{x_k}$$

$$\mathbf{M}_k = \mathbf{I} - \boldsymbol{\tau} \mathbf{e}_k^T, \quad \boldsymbol{\tau} = [0, \dots, 0, \tau_{k+1}, \dots, \tau_n]^T$$

Handwritten derivation:

$$\boldsymbol{\tau} \mathbf{e}_k^T = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ \tau_{k+1} \\ \vdots \\ \tau_n \end{bmatrix} \begin{bmatrix} 0 & \dots & 0 & 1 & 0 & \dots & 0 \end{bmatrix} = \begin{bmatrix} 0 & \dots & 0 & 0 \\ \vdots & & \vdots & \vdots \\ 0 & & \tau_{k+1} & 0 \\ & & \vdots & \vdots \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \end{bmatrix}$$

kth column (with arrow pointing to the 1 in the matrix)

Multiplication by a Gauss Transformation

Let $\mathbf{M}_k = \mathbf{I} - \tau \mathbf{e}_k^T$ be a Gauss transformation and $\mathbf{C} \in \mathbb{R}^{n \times r}$

$$\mathbf{M}_k \mathbf{C} = (\mathbf{I} - \tau \mathbf{e}_k^T) \mathbf{C} = \mathbf{C} - \tau (\mathbf{e}_k^T \mathbf{C}) = \mathbf{C} - \tau \mathbf{C}(k, :) \quad (\text{outer product})$$

Here, τ does not necessarily depend on \mathbf{C} . Since $\tau(1:k) = \mathbf{0}$, only $\mathbf{C}(k+1:n, :)$ is affected

```
for i= k+1:n
```

```
    C(i,:) = C(i,:) - tau(i)*C(k,:)
```

```
end
```

the first k rows of C are unchanged under Gauss transformation

2r flops

Complexity: $2(n-k)r$ flops

compared to $O(n^2r)$ for $\mathbf{M}_k \mathbf{C}$

Exercise: Compute $(\mathbf{I} - \tau \mathbf{e}_1^T) \mathbf{C}$ with

$$\mathbf{C} = \begin{bmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \\ c_{31} & c_{32} & c_{33} \end{bmatrix} \quad \tau = \begin{bmatrix} 0 \\ \tau_2 \\ \tau_3 \end{bmatrix}$$

1st row of $\mathbf{M}_1 \mathbf{C}$

2nd row of $\mathbf{M}_1 \mathbf{C}$

3rd row of $\mathbf{M}_1 \mathbf{C}$

unchanged $[c_{11} \ c_{12} \ c_{13}]$

$[c_{21} \ c_{22} \ c_{23}] - \tau_2 [c_{11} \ c_{12} \ c_{13}]$

$[c_{31} \ c_{32} \ c_{33}] - \tau_3 [c_{11} \ c_{12} \ c_{13}]$