

Gaussian Elimination

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October 27, 2011

1 Gaussian Elimination and LU Factorization

Gaussian Elimination is a way of solving systems of linear equations. When you apply the algorithm in a matrix form, it lets us write a matrix as the product of a unit lower-triangular matrix and an upper-triangular matrix.

Problem Statement: Solve $Ax = b$, where A is a N by M non-singular matrix.

Goal: Compute L , a unit lower-triangular matrix and U an upper-triangular matrix (with a P permutation matrix) s.t. $A = LU$. Then, we can solve $Ax = b$ by

$$\begin{aligned}Ax &= b \\LUx &= b \\L\hat{b} &= b \text{ where } \hat{b} = L^{-1}b \\&\Rightarrow Ux = \hat{b}\end{aligned}$$

Where since L is a lower triangular matrix, we can solve for \hat{b} using forward-substitution in $\mathcal{O}(n^2)$, and similarly, U is an upper triangular matrix so using backward-substitution we can solve for x in $\mathcal{O}(n^2)$

Doing the LU factorization costs $\mathcal{O}(\frac{2}{3}n^3) = \mathcal{O}(n^3)$.

2 How to do LU Factorization

1. Add multiple of first row of A and first entry of b to the second to last rows to produce zeros on the first column. i.e. Multiply A with L_1 to get $A^{(1)}$ ($A^{(i)}$ is our A after i th many steps).
2. Repeat, then we will get our upper triangular matrix $U = L_{n-1} \cdots L_1 A$, and our $L = L_1^{-1} \cdots L_{n-1}^{-1}$.
(Why?) because $A = L_1^{-1} L_1 A = L_1^{-1} U_1$ where $U_1 = L_1 A$ we get:

$$\begin{aligned}A &= (L_1^{-1} \cdots L_{n-1}^{-1})(L_{n-1} \cdots L_1 A) \\&= LU\end{aligned}$$

3 Step-by-Step example with 3 by 3 case

Let's walk through the entire thing with a 3 by 3 A :

3.1 LU Factorization

$$Ax = b \Rightarrow \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$

Where x is unknown.

1. Formally, our L_1 is $\begin{pmatrix} 1 & 0 & 0 \\ -a_{21}/a_{11} & 1 & 0 \\ -a_{31}/a_{11} & 0 & 1 \end{pmatrix}$, so

$$\begin{aligned} L_1 A &= \begin{pmatrix} 1 & 0 & 0 \\ -a_{21}/a_{11} & 1 & 0 \\ -a_{31}/a_{11} & 0 & 1 \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \\ &= \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ \frac{-a_{21}}{a_{11}}a_{11} + a_{21} & \frac{-a_{21}}{a_{11}}a_{12} + a_{22} & \frac{-a_{21}}{a_{11}}a_{13} + a_{23} \\ \frac{-a_{31}}{a_{11}}a_{11} + a_{31} & \frac{-a_{31}}{a_{11}}a_{12} + a_{32} & \frac{-a_{31}}{a_{11}}a_{13} + a_{33} \end{pmatrix} \\ &= \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22}^{(1)} & a_{23}^{(1)} \\ 0 & a_{32}^{(1)} & a_{33}^{(1)} \end{pmatrix} \end{aligned}$$

2. $L_2 L_1 A = L_2 U^{(1)} =$

$$\begin{aligned} &\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -a_{32}^{(1)}/a_{22}^{(1)} & 1 \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22}^{(1)} & a_{23}^{(1)} \\ 0 & a_{32}^{(1)} & a_{33}^{(1)} \end{pmatrix} \\ &= \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22}^{(1)} & a_{23}^{(1)} \\ 0 & \frac{-a_{32}^{(1)}}{a_{22}^{(1)}}a_{22}^{(1)} + a_{32}^{(1)} & \frac{-a_{32}^{(1)}}{a_{22}^{(1)}}a_{23}^{(1)} + a_{33}^{(1)} \end{pmatrix} \\ &= \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22}^{(1)} & a_{23}^{(1)} \\ 0 & 0 & a_{33}^{(2)} \end{pmatrix} = U \end{aligned}$$

Now What's L ? Note that for all of L_i 's we use, $L_i^{-1} = -L_i$ (check it).
So our $L = L_1^{-1} L_2^{-1} =$

$$\begin{pmatrix} 1 & 0 & 0 \\ a_{21}/a_{11} & 1 & 0 \\ a_{31}/a_{11} & a_{32}^{(1)}/a_{22}^{(1)} & 1 \end{pmatrix}$$

Or just all of the nonzero elements of $-L_i$'s.

(Remember that inverse/multiplication of a lower triangular matrix is also a lower triangular matrix)

Finally we have:

$$A = LU$$

$$= \begin{pmatrix} 1 & 0 & 0 \\ a_{21}/a_{11} & 1 & 0 \\ a_{31}/a_{11} & a_{32}^{(1)}/a_{22}^{(1)} & 1 \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22}^{(1)} & a_{23}^{(1)} \\ 0 & 0 & a_{33}^{(2)} \end{pmatrix}$$

3.2 Forward-substitution

Going back to our objective of solving $Ax = b$, now we do $\hat{b} = L^{-1}b$, or often written as $Ly = b$ ($y = \hat{b}$).

Looking at

$$\begin{pmatrix} 1 & 0 & 0 \\ a_{21}/a_{11} & 1 & 0 \\ a_{31}/a_{11} & a_{32}^{(1)}/a_{22}^{(1)} & 1 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$

The first two equations are $y_1 = b_1$ and $\frac{a_{21}}{a_{11}}y_1 + y_2 = b_2$. But since we know y_1 from the first line, we can substitute b_1 for y_1 and yeild $y_2 = b_2 - b_1 \frac{a_{21}}{a_{11}}$. We can keep on substituting earlier results to obtain all y_i and this is *forward-substitution*.

Using this, our \hat{b} is:

$$y_1 = b_1$$

$$b_2 = y_2 + y_1 \frac{a_{22}}{a_{11}}$$

$$\rightarrow y_2 = b_2 - b_1 \frac{a_{22}}{a_{11}} \text{ substitute } y_1$$

$$b_3 = y_3 + y_1 \frac{a_{31}}{a_{11}} + y_2 \frac{a_{32}^{(1)}}{a_{22}^{(1)}}$$

$$\rightarrow y_3 = b_3 - (y_1 \frac{a_{31}}{a_{11}} + y_2 \frac{a_{32}^{(1)}}{a_{22}^{(1)}})$$

Formally:

$$y_m = b_m - \sum_{i=1}^{m-1} l_{mi} y_i$$

This is long and details don't matter so just denote $\hat{b} = \begin{pmatrix} \hat{b}_1 \\ \hat{b}_2 \\ \hat{b}_3 \end{pmatrix}$

3.3 Backward-substitution

Similarly, now that we have our \hat{b} , we do $U\hat{b} = x$. Since U is an upper triangular, it's the same thing as forward substitution but we progress from x_n ($x_n = \hat{b}_1$) For us the x is:

$$\begin{aligned} x_3 &= \frac{\hat{b}_3}{u_{33}} \\ \hat{b}_2 &= u_{22}x_2 + u_{23}x_3 \\ \rightarrow x_2 &= \frac{\hat{b}_2 - u_{23}x_3}{u_{22}} \\ \rightarrow x_2 &= \frac{\hat{b}_2 - u_{23}\frac{\hat{b}_3}{u_{33}}}{u_{22}} \text{ substitute } x_3 \\ \hat{b}_1 &= u_{11}x_1 + u_{12}x_2 + u_{13}x_3 \\ \rightarrow x_1 &= \frac{\hat{b}_1 - (u_{12}x_2 + u_{13}x_3)}{u_{11}} \end{aligned}$$

Formally:

$$x_m = (\hat{b}_m - \sum_{i=m+1}^1 u_{mi}x_i)/u_{mm}$$

And finally we get our x !!

4 Pivoting

If $a_{11} = 1$, everything above falls apart. In such a case we need to permute the first row of A with another row of A with $\max_j |a_{j1}|$. This is the same as multiplying $Ax = b$ with a *permutation matrix*, P_1 , where all the diagonal values are 1 but a_{11}, a_{jj} and rest all 0 but a_{1j} and a_{j1}

$$P_1 = \begin{pmatrix} \mathbf{0} & \dots & \dots & \dots & \mathbf{1} = a_{1j} & \dots & 0 \\ 0 & 1 & 0 & \dots & \dots & \dots & 0 \\ 0 & 0 & 1 & \dots & \dots & \dots & 0 \\ \vdots & & & & & & \\ \mathbf{1}_{(a_{j1})} & 0 & 0 & \mathbf{0}_{(a_{jj})} & \dots & \dots & 0 \\ \vdots & & & & & & \\ 0 & \dots & \dots & \dots & \dots & \dots & 1 \end{pmatrix}$$

So in the end, we do the same as above and compute L , U , and $P = P_1P_2 \dots P_k$ where $k \leq n - 1$ and get $PA = LU$. Note that there is no mutliplication/division in doing PA .

5 Cost Analysis/Complexity

P is costless in terms of *flops* (in the context of number of multiplication and divisions). Components that contribute to the cost are:

1. **LU decomposition:** to make L_1 requires $n - 1$ divisions, and doing $L_1 A$ requires $(n-1)(n-1)$ multiplications. So total of $(n-1)^2 + n - 1 = n^2 - n$ or $n(n-1)$ operations for step 1 (n diagonal elements to multiply with elements that required $(n-1)$ flops to make).

So to do step 2 is $(n-1)(n-2)$, and so forth until $n-1$ steps are done, which is:

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

Recalling calc 2 arithmetic series and special sums, $\sum_{i=1}^n i = n(n+1)/2$ and $\sum_{i=1}^n i^2 = n(n+1)(2n+1)/6$, so the total sum is $\mathcal{O}n^3/3 + (n^2 + n)/2 = \mathcal{O}(2n^3/3)$

2. **forward/backward substitution:** The first(last for back) operation is 1 multiplication, the second equation requires 2 multiplication, and in the end it requires n multiplication. Disregarding the additions, this is just $\sum_{i=1}^n i = n(n+1)/2 = \mathcal{O}(n^2)$. This analysis holds for both, so total $\mathcal{O}(2n^2)$ here.

So in total, we have

$$\mathcal{O}(2n^3/3) + \mathcal{O}(2n^2) = \mathcal{O}(n^3)$$

for large n .