

 1 Exact solution of system

$$\begin{pmatrix} 2 & 4 & 4 \\ 1 & 5 & 6 \\ 1 & 3 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 6 \\ 4 \\ 8 \end{pmatrix}$$

Solution: You will find $x = [1, 3, -2]^T$. \square

 2 Justify the column version of Back-substitution algorithm.

Solution: The system $Ax = b$ can be written in column form as follows:

$$x_1 a_{:,1} + x_2 a_{:,2} + \cdots + x_n a_{:,n} = b$$

In first step we compute $x_n = b_n / a_{n,n}$. Now move last term in left-hand side of above system to the right:

$$x_1 a_{:,1} + x_2 a_{:,2} + \cdots + x_{n-1} a_{:,n-1} = b - x_n a_{:,n} \equiv b^{(1)}$$

This is a new system of n equations that has $(n - 1)$ unknowns and the right-hand-side $b^{(1)}$. The last equation of this system is of the form $0 = 0$ and can therefore be ignored. Thus, we end up with a system of

size $(n - 1) \times (n - 1)$ that is still upper triangular and we can repeat the above argument recursively. \square

 3 Exact operation count for GE.

Solution:

$$\begin{aligned}
 T &= \sum_{k=1}^{n-1} \sum_{i=k+1}^n (2(n - k) + 3) \\
 &= \sum_{k=1}^{n-1} (2(n - k) + 3)(n - k) \\
 &= 2 \sum_{k=1}^{n-1} (n - k)^2 + 3 \sum_{k=1}^{n-1} (n - k) \\
 &= 2 \sum_{j=1}^{n-1} j^2 + 3 \sum_{j=1}^{n-1} j
 \end{aligned}$$

In the last step we made a change of variables $j = n - k$. Now we know that $\sum_{k=1}^n k^2 = n(n + 1)(2n + 1)/6$ and $\sum_{k=1}^n k = n(n + 1)/2$ and so

$$\begin{aligned}
 T &= 2 \frac{(n - 1)(n)(2n - 1)}{6} + 3 \times \frac{n(n - 1)}{2} \\
 &= \dots \\
 &= n(n - 1) \left(\frac{2n}{3} + \frac{7}{6} \right) \tag{1}
 \end{aligned}$$

Note in passing the remarkable fact that the above final expression is

always an integer (it has to be) no matter what (integer) value n takes.

□

4 Practical use: Show how to use the LU factorization to solve linear systems with the same matrix A and different b 's.

Solution: If we have the LU factorization $A = LU$ available then we can solve the linear system $Ax = b$ by writing it as

$$L(\underbrace{Ux}_y) = b$$

So we solve for y : $Ly = b$ then once y is computed we solve for x : $Ux = y$. This involves two triangular solves at the cost of n^2 each instead of the $O(n^3)$ cost of redoing everything with Gaussian elimination. □


5 LU factorization of the matrix $A = \begin{pmatrix} 2 & 4 & 4 \\ 1 & 5 & 6 \\ 1 & 3 & 1 \end{pmatrix}$?

Solution: You will find

$$L = \begin{pmatrix} 1 & 0 & 0 \\ 1/2 & 1 & 0 \\ 1/2 & 1/3 & 1 \end{pmatrix} \quad U = \begin{pmatrix} 2 & 4 & 4 \\ 0 & 3 & 4 \\ 0 & 0 & -7/3 \end{pmatrix} \quad \square$$

 6 Determinant of A ?

Solution: It is the determinant of U which is -12 .

 7 True or false: “Computing the LU factorization of matrix A involves more arithmetic operations than solving a linear system $Ax = b$ by Gaussian elimination”.

Solution: The number of arithmetic operations is identical. (The LU factorization involves additional memory moves to store the factors - but these are no floating point operations)

 8 Operation count for Gauss-Jordan. Order of the cost? How does it compare with Gaussian Elimination?

Solution: From the notes:

$$\begin{aligned} T &= \sum_{k=1}^{n-1} \sum_{i=1}^{n-1} [2(n-k) + 3] = \sum_{k=1}^{n-1} (n-1)[2(n-k) + 3] \\ &= (n-1) \sum_{j=1}^{n-1} [2j + 3] \\ &= (n-1) [n(n-1) + 3(n-1)] \\ &= (n-1)^2(n+3) = (n-1)^3 + 4(n-1)^2 \end{aligned}$$

The bottom line is that the cost is $\approx n^3$ which is 50% more expensive than GE. This additional cost is not worth it in spite of the simplicity

of the algorithm. For this Gauss-Jordan is seldom used in practice.

9 What is the matrix PA when

$$P = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \end{pmatrix} \quad A = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 9 & 0 & -1 & 2 \\ -3 & 4 & -5 & 6 \end{pmatrix} ?$$

Solution: Instead of multiplying you just permute the row: row 1 in new matrix is row 3 of old matrix, row 2 is row 1 of old matrix, etc.


$$PA = \begin{pmatrix} 9 & 0 & -1 & 2 \\ 1 & 2 & 3 & 4 \\ -3 & 4 & -5 & 6 \\ 5 & 6 & 7 & 8 \end{pmatrix}$$

10 In the previous example where

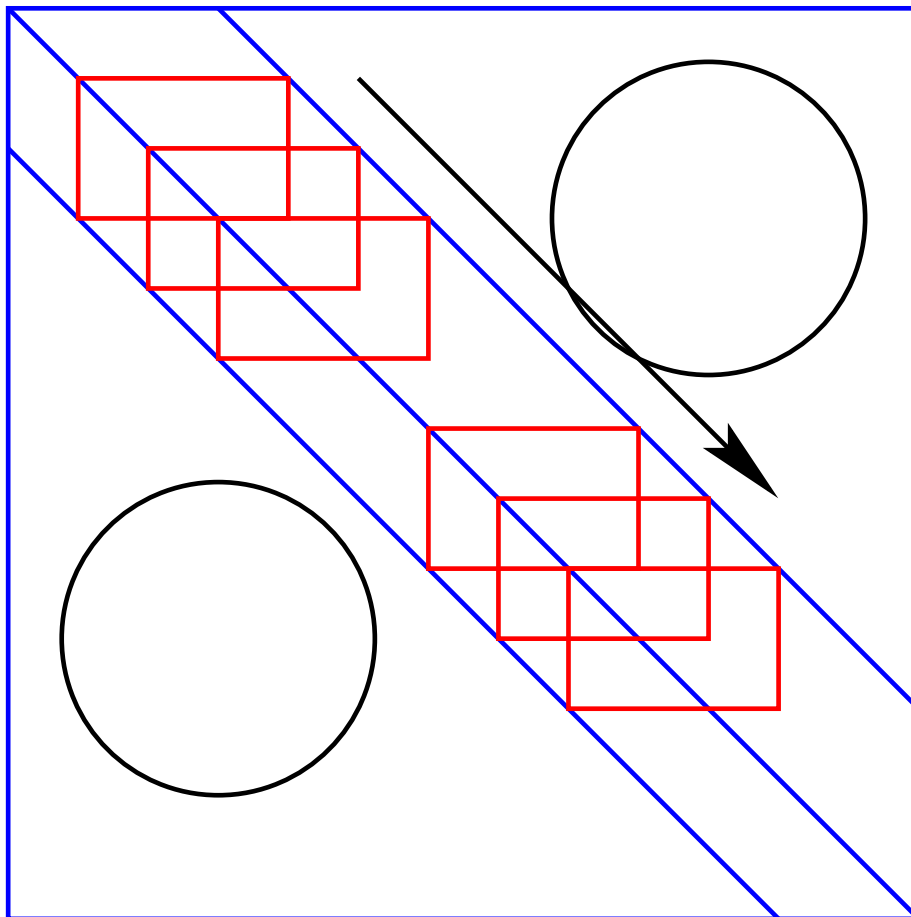
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>> A = [ 1 2 3 4; 5 6 7 8; 9 0 -1 2 ; -3 4 -5 6]
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Matlab gives $\det(A) = -896$. What is $\det(PA)$?

Solution: It changes sign so $\det(PA) = 896$. This is because the permutation $\pi = [3, 1, 4, 2]$ is made of 3 interchanges.

 11 Given a banded matrix with upper bandwidth q and lower bandwidth p , what is the operation count (leading term only) for solving the linear system $Ax = b$ with Gaussian elimination without pivoting? What happens when partial pivoting is used? Give the new operation count for the worst case scenario.

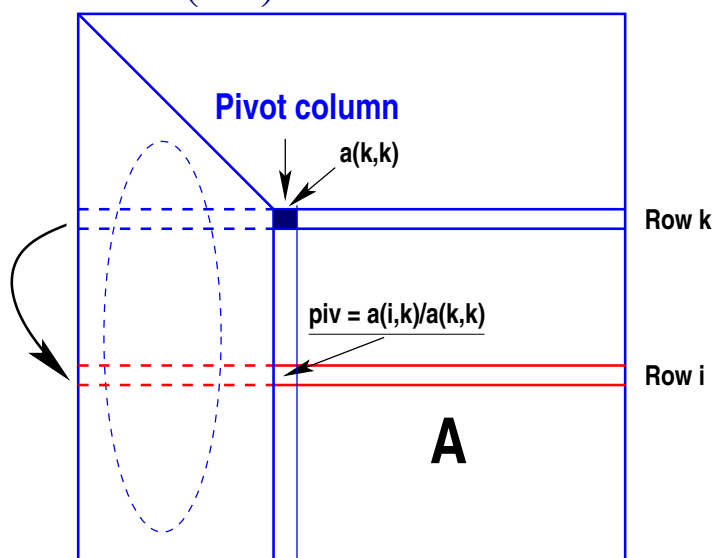
Solution: [Note: it is assumed that $p \ll n$ and $q \ll n$ but p and q are not related]. The important observation here is that Gaussian elimination without pivoting for this band matrix will operate on a rectangle: at step k only rows $k + 1$ to $k + p$ are affected and columns $k + 1$ to $k + q$ are affected.



In this rectangle each entry will be modified at the cost of 2 operations (*, +). Total: $2pq$ for each step. So Gaussian elimination without pivoting for this band matrix costs *approximately* $2npq$ flops. Using band backward substitution to obtain the solution x costs $\approx 2nq$ flops. The total operation count (leading term only): $\approx 2npq + 2nq = 2nq(p+1)$. Note that when p is small the cost of substitution cannot be ignored.

For the Gaussian elimination with pivoting, the upper bandwidth of the resulting matrix will be $p + q$. In this case, the total operation count (leading term only) becomes: $\approx 2np(p + q)(p + 1)$. \square

Additional notes on the LU factorization The lecture notes mention an ‘algorithmic’ approach to understanding the LU factorization. Here again is the illustration of the k -th step of Gaussian Elimination (GE):



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For i=k+1:n Do
  piv = a(i,k)/a(k,k)
  row(i):=row(i) - piv*row(k)

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We will focus on how the i -th row is transformed throughout the algorithm. In words: the i th row undergoes $i - 1$ transformations (each indexed by k in the algorithm). After these $i - 1$ GE steps the row remains unchanged. Each of these $i - 1$ transformations - which corresponds to the steps $k = 1, 2, \dots, i - 1$, is as follows

$$\mathbf{a}_{i,:} = \mathbf{a}_{i,:} - \text{piv} * \mathbf{a}_{k,:}$$

We will need to make the following changes to the notation for better clarity. Once a row say $\mathbf{a}_{j,:}$ no longer changes [i.e., when it undergoes no further transformations] we will call it $\mathbf{u}_{j,:}$ - reflecting the fact that this will end up in the final \mathbf{U} matrix of the LU factorization. In addition we will change ‘piv’ in the above equation into l_{ik} which we recall is equal to $\mathbf{a}_{ik}/\mathbf{a}_{kk}$. Finally, we must also add a superscript to row i to index the transformation number k . With this, the above equation becomes

$$\mathbf{a}_{i,:}^{(k)} = \mathbf{a}_{i,:}^{(k-1)} - l_{ik} * \mathbf{u}_{k,:}$$

Notice how $\mathbf{a}_{k,:}$ has been changed to $\mathbf{u}_{k,:}$. Indeed the pivot row used for any elimination no longer changes. We will write the above relation for $k = 1, 2, \dots, i - 1$. After these $i - 1$ transformations $\mathbf{a}_{i,:}^{(k)}$ is no longer changed and becomes the row \mathbf{u}_k , the k th row of \mathbf{U} .

$$\mathbf{a}_{i,:}^{(1)} = \mathbf{a}_{i,:} - l_{i1} * \mathbf{u}_{1,:}$$

$$\mathbf{a}_{i,:}^{(2)} = \mathbf{a}_{i,:}^{(1)} - l_{i2} * \mathbf{u}_{2,:}$$

$$\mathbf{a}_{i,:}^{(3)} = \mathbf{a}_{i,:}^{(2)} - l_{i3} * \mathbf{u}_{3,:}$$

$$\dots = \dots - (\) * (\dots)$$

$$\mathbf{a}_{i,:}^{(i-2)} = \mathbf{a}_{i,:}^{(i-3)} - l_{i,i-2} * \mathbf{u}_{2,:}$$

$$\mathbf{a}_{i,:}^{(i-1)} = \mathbf{a}_{i,:}^{(i-2)} - l_{i,i-1} * \mathbf{u}_{2,:}$$

Notice that $\mathbf{a}_{i,:}^{(0)}$ is just $\mathbf{a}_{i,:}$. If you add all the equations on the left - things cancel out - and you will wind up with:

$$\mathbf{a}_{i,:}^{(i-1)} = \mathbf{a}_{i,:} - \sum_{k=1}^{i-1} l_{ik} \mathbf{u}_{k,:}$$

The row $\mathbf{a}_{i,:}^{(i-1)}$ is no longer modified.

Therefore, it should be change to $\mathbf{u}_{i,:}$. and so we get:

$$\mathbf{u}_{i,:} = \mathbf{a}_{i,:} - \sum_{k=1}^{i-1} l_{ik} \mathbf{u}_{k,:} \quad \text{or} \quad \mathbf{a}_{i,:} = \mathbf{u}_{i,:} + \sum_{k=1}^{i-1} l_{ik} \mathbf{u}_{k,:}$$

Next define the matrix \mathbf{L} whose entries l_{ij} 's are the same as above for $i > j$ (lower part), $l_{ii} = 1$ (diagonal), and $l_{ij} = 0$ for $j > i$ (upper part). The above equation can now be rewritten as

$$\mathbf{a}_{i,:} = \mathbf{u}_{i,:} + \sum_{k=1}^{i-1} l_{ik} \mathbf{u}_{k,:} = \sum_{k=1}^n l_{ik} \mathbf{u}_{k,:}$$

This translates **exactly** the equation $\mathbf{A} = \mathbf{LU}$ written in row-form. \square