

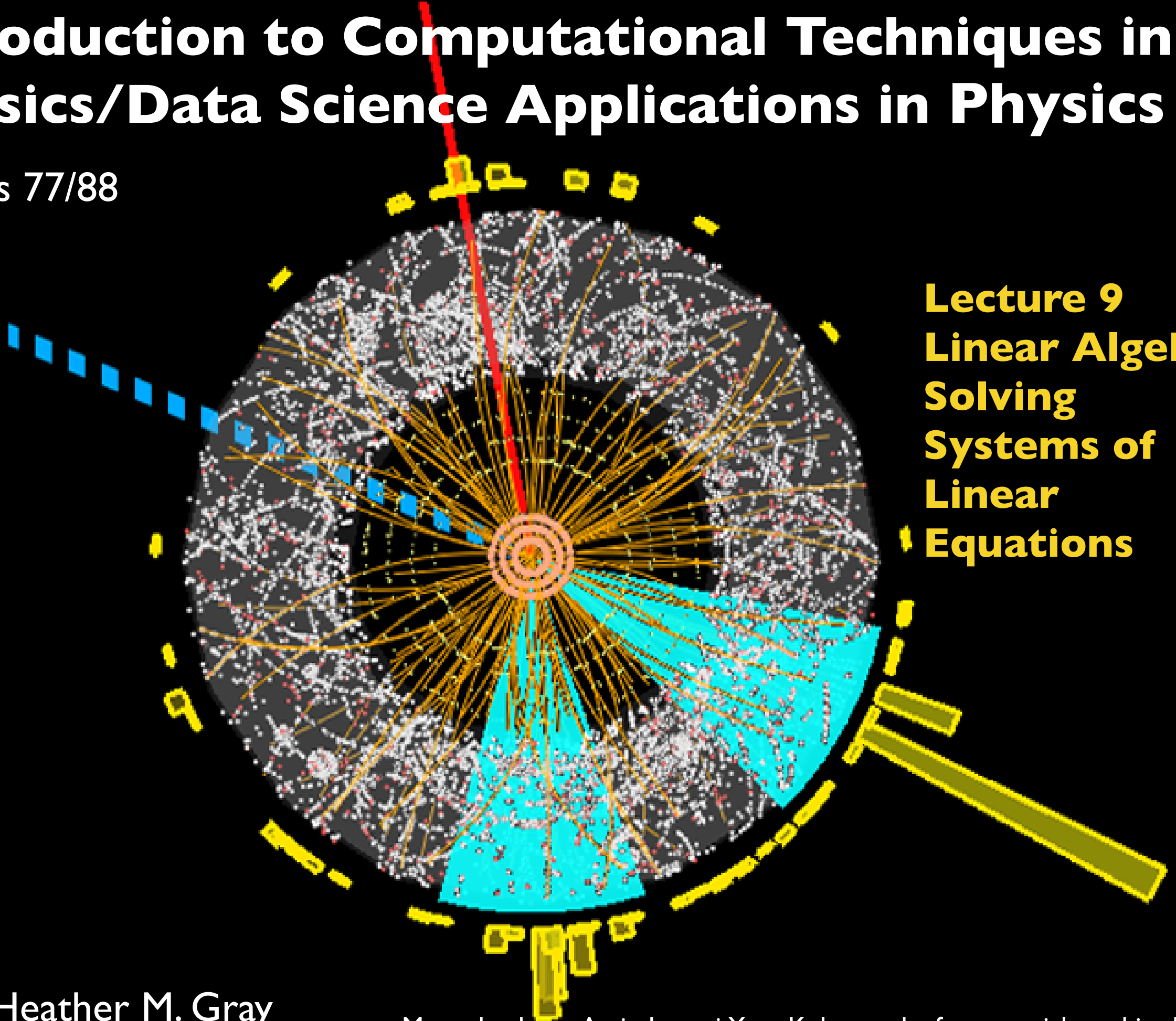
Introduction to Computational Techniques in Physics/Data Science Applications in Physics

Physics 77/88

Lecture 9 Linear Algebra: Solving Systems of Linear Equations

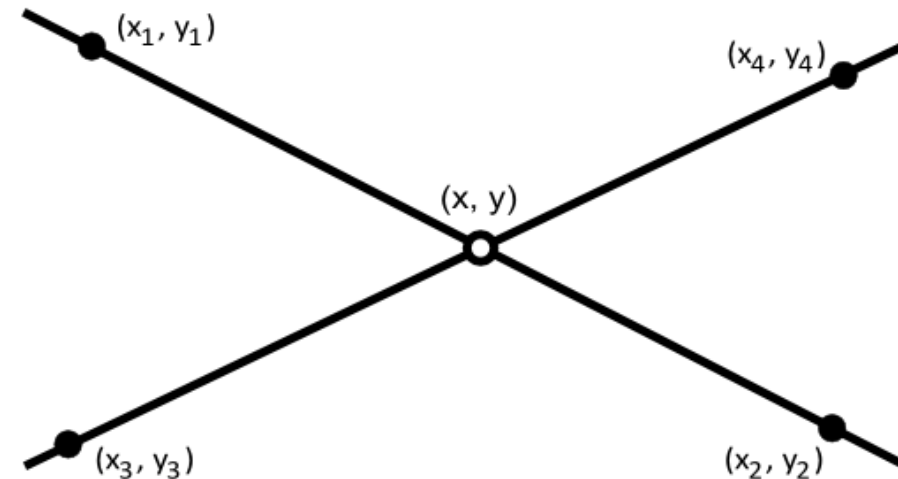
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Example: Finding Intersection of Two Lines

- Example 1:
 - $y = 2 - x$
 - $y = x - 1$
- Solution:



- $\Rightarrow x =$

- $y =$

- $(x, y) =$

$$\begin{bmatrix} \\ \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \\ \end{bmatrix}$$

Example: Finding Intersection of Two Lines

- Example 2 (generic):

- $a_{11}x + a_{12}y = c_1$

- $a_{21}x + a_{22}y = c_2$

- Solution:

-

$$-x =$$

$$x$$

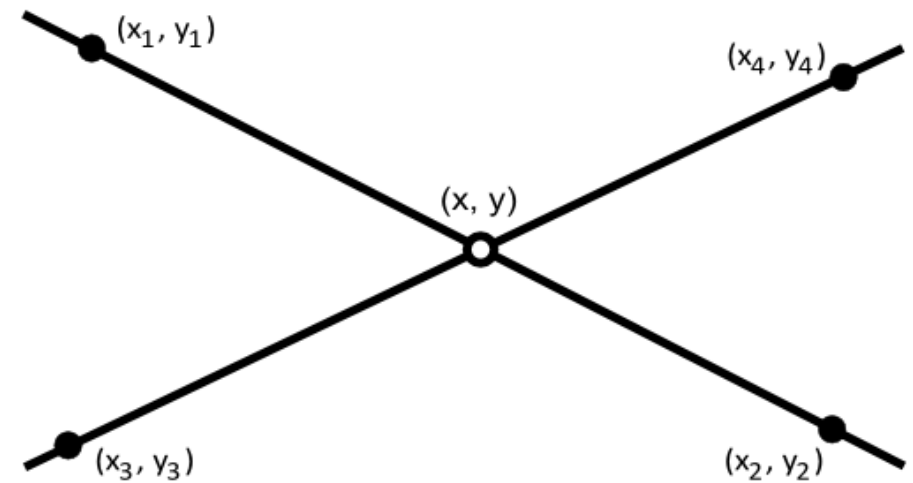
$$\Rightarrow x = \underline{\hspace{2cm}}$$

-

$$= ($$

$$)x$$

Substitute and
solve for y



Linear Algebraic Equations

$$\begin{aligned}
 & a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \cdots a_{1n}x_n &= c_1 \\
 & a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \cdots a_{2n}x_n &= c_2 \\
 \bullet \quad & \vdots &= \vdots \\
 & a_{n1}x_1 + a_{n2}x_2 + a_{n3}x_3 + \cdots a_{nn}x_n &= c_n
 \end{aligned}$$

$$\bullet \quad \begin{bmatrix} \vdots & \vdots & \vdots & \vdots \\ \dots & \dots & \dots & \dots \end{bmatrix} \begin{Bmatrix} \vdots \end{Bmatrix} = \begin{Bmatrix} \vdots \end{Bmatrix}$$

- In matrix format

-

- Solution

-

Linear Algebra Primer: Matrices

- Inverse

-

- Transpose

- \mathcal{I}

- Conjugate transpose

-

- Trace

-

Linear Algebra Primer: Matrices

- Symmetric

-

- Unitary

- Normal

-

Determinant

- If the $\det(A)$ of a matrix is $\neq 0$ then solutions exist and are unique (and $\det(A) \neq 0$)

- Our Generic Example 2:

- $a_{11}x + a_{12}y = c_1$

- $a_{21}x + a_{22}y = c_2$

- $$x = \frac{\frac{c_1}{a_{12}} - \frac{c_2}{a_{22}}}{\frac{a_{11}}{a_{12}} - \frac{a_{21}}{a_{22}}}, y = \frac{\frac{c_2}{a_{21}} - \frac{c_1}{a_{11}}}{\frac{a_{21}}{a_{12}} - \frac{a_{22}}{a_{22}}}$$

- If $\det(A) = 0$; x and y are

Determinant of a Matrix

- The determinant is a scalar value that can be used to determine if a matrix is invertible.

r that can be used

- $\det A = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$

- $= a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$

- $=$

Aside: Levi-Civita Symbol

- Define $\epsilon_{ij} = \begin{cases} 1 & \text{if } (i, j) = (1, 2) \\ -1 & \text{if } (i, j) = (2, 1) \\ 0 & \text{if } i = j \end{cases}$

- e.g. $\begin{pmatrix} \epsilon_{11} & \epsilon_{12} \\ \epsilon_{21} & \epsilon_{22} \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$

- For matrix $A \Rightarrow \det A = \sum$

Aside: Levi-Civita Symbol

- Extending to 3 dimensions

$$\epsilon_{ijk} = \begin{cases} 1 & \text{if } (i, j, k) = (1, 2, 3) \\ -1 & \text{if } (i, j, k) = (3, 2, 1) \\ 0 & \text{if } (i, j, k) \text{ is not a permutation of } (1, 2, 3) \end{cases}$$

- A famous application

$$(a \times b)_i = \epsilon_{ijk} a_j b_k = \sum_{j=1}^3 \sum_{k=1}^3 \epsilon_{ijk} a_j b_k$$

$$(a \times b)^i = \sum_{j=1}^3 \sum_{k=1}^3 \epsilon^{ijk} a_j b_k$$

Einstein
summation: sum
over repeated
indices is implied

Aside: Levi-Civita Symbol

- We can generalize to matrices of size
- $\epsilon_{i_1 i_2 \dots i_n} = \begin{cases} 1 & \text{if } p \text{ is the permutation of the indices from } 1 \text{ to } n \\ -1 & \text{if } p \text{ is an odd permutation of the indices from } 1 \text{ to } n \\ 0 & \text{if } p \text{ is not a permutation of the indices from } 1 \text{ to } n \end{cases}$
- the number of permutations of order n is $n!$ to get from $1, 2, \dots, n$ to i_1, i_2, \dots, i_n
- Then in n dimensions $A = [a_{ij}]$ and
- $\det(A) = \sum_{i_1, i_2, \dots, i_n} \epsilon_{i_1 i_2 \dots i_n} a_{1 i_1} a_{2 i_2} \dots a_{n i_n}$

Cramer's Rule

- a_{ij} is the i th element of the j th column with the i th row and j th column removed
- $(-1)^{i+j}$ is called the cofactor of element a_{ij}

$$\bullet \det(A) = \sum_{i=1}^n a_{ij} C_{ij}$$

Cramer's Rule

$$x_1 = \frac{\begin{vmatrix} a_{12} & a_{13} & \cdots & a_{1n} \\ a_{22} & a_{23} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n2} & a_{n3} & \cdots & a_{nn} \end{vmatrix}}{\begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix}}$$

$$x_j = \frac{\begin{vmatrix} a_{11} & \cdots & a_{1j-1} & a_{1j+1} & \cdots & a_{1n} \\ a_{21} & \cdots & a_{2j-1} & a_{2j+1} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nj-1} & a_{nj+1} & \cdots & a_{nn} \end{vmatrix}}{\begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix}}$$

Practical Details: Cramer's Rule

- operations for the
- operations for
- for matrix
- error propagation
- for small matrices ($n < 20$)

Gaussian Elimination

- Divide a_{ii} by the a_{ii}
- Subtract a_{ij} from all the a_{ij}
- Move to the $a_{i+1,i+1}$ and continue

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ a_{31} & a_{32} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \\ \vdots \\ c_n \end{bmatrix}$$

Gaussian Elimination

- Divide each row by the leading element

$$\begin{bmatrix} \frac{a_{11}}{a_{11}} & \frac{a_{12}}{a_{11}} & \dots & \frac{a_{1n}}{a_{11}} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{a_{31}}{a_{31}} & \frac{a_{32}}{a_{31}} & \dots & \frac{a_{3n}}{a_{31}} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{a_{n1}}{a_{n1}} & \frac{a_{n2}}{a_{n1}} & \dots & \frac{a_{nn}}{a_{n1}} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_3 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} \frac{c_1}{a_{11}} \\ \vdots \\ \frac{c_3}{a_{31}} \\ \vdots \\ \frac{c_n}{a_{n1}} \end{bmatrix}$$

Gaussian Elimination

- Subtract row 1 from all the other rows

$$\begin{bmatrix} 1 & \frac{a_{12}}{a_{11}} & \frac{a_{13}}{a_{11}} & \dots & \frac{a_{1n}}{a_{11}} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & \frac{a_{32}}{a_{31}} & \frac{a_{33}}{a_{31}} & \dots & \frac{a_{3n}}{a_{31}} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & \frac{a_{n2}}{a_{n1}} & \frac{a_{n3}}{a_{n1}} & \dots & \frac{a_{nn}}{a_{n1}} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} \frac{c_1}{a_{11}} \\ \vdots \\ \frac{c_3}{a_{31}} \\ \vdots \\ \frac{c_n}{a_{n1}} \end{bmatrix}$$

Gaussian Elimination

$$\begin{bmatrix} 1 & & & \dots & \\ 0 & 1 & & \dots & \\ 0 & 0 & 1 & \dots & \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} \\ \\ \\ \vdots \\ \end{bmatrix}$$

- Back substitution

- $x_n =$

- $x_i =$

Gaussian Elimination: Practical Issues

- Division by
 - May occur in
- - Prone to

Gaussian Elimination: Example

- Let's look at the following system of equations
- We're going to use five significant figures with chopping

$$\begin{bmatrix} 10 & -7 & 0 \\ -3 & 2.099 & 6 \\ 5 & -1 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 7 \\ 3.901 \\ 6 \end{bmatrix}$$

- After forward elimination

$$\begin{bmatrix} 1 & & & \\ 0 & 1 & & \\ 0 & 0 & & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \\ \\ \end{bmatrix}$$

Gaussian Elimination: Example

- Next, we apply back substitution

$$\begin{bmatrix} 1 & -.7 & 0 \\ 0 & 1 & -588.23524 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0.7 \\ -588.3323 \\ 0.9992235 \end{bmatrix}$$

- $x_3 =$

- $x_2 =$

- $x_1 =$

Gaussian Elimination: Example

- Compare the calculated values with the exact solution

- $[X]_{\text{exact}} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \\ \\ \end{bmatrix}$

- $[X]_{\text{calculated}} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -0.35 \\ -1.5 \\ 0.99993 \end{bmatrix}$

Now let's see how this looks in python

Gaussian Elimination: Improvements

- We can increase the number of
 - Decreases the error
 - Does avoid division
- Gaussian elimination with
 - division by zero
 - rounding off error

Partial Pivoting

- Gaussian elimination with partial pivoting applies to normal Gaussian elimination
- At the beginning of the k th step of forward elimination, find the maximum of $|a_{kj}|$ for $j = k, k+1, \dots, n$
- If the maximum of the values is $|a_{kp}|$ in the k th row, where $k \leq p \leq n$
 - Switch row k with row p
- Gaussian elimination with partial pivoting ensures that forward elimination is performed with the pivot element having the largest magnitude

Example

- Consider the same system of equations

- $10x_1 - 7x_2 = 7$

- $-3x_1 + 2.099x_2 + 3x_3 = 3.901$

- $5x_1 - x_2 + 5x_3 = 6$

- In matrix form

- $$\begin{bmatrix} & & \\ & & \\ & & \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \\ \\ \end{bmatrix}$$

- Let's solve them using Gaussian elimination with partial pivoting with

Partial Pivoting: Example

- Forward elimination: step 1
 - Examine the values of the
 - or
- The largest absolute value is 10, which means, following the rules of partial pivoting, we don't switch anything

$$\begin{bmatrix} 10 & 7 & 0 \\ -3 & 2.099 & 6 \\ 5 & -1 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 7 \\ 3.901 \\ 6 \end{bmatrix}$$

Forward elimination

$$\begin{bmatrix} & & \\ & & \\ & & \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \\ \\ \end{bmatrix}$$

Partial Pivoting: Example

- Forward elimination, step 2
 - Examine the values of the second column
 - or
 - The largest absolute value is 0.9 so switch row 2 and 3

$$\begin{bmatrix} 1 & -0.7 & 0 \\ 0 & -0.0034 & 2 \\ 0 & -0.9 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 7 \\ 2.000333 \\ -0.5 \end{bmatrix}$$

Row swap

$$\begin{bmatrix} 1 & -0.7 & 0 \\ 0 & -0.9 & -1 \\ 0 & -0.0034 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 7 \\ -0.5 \\ 2.000333 \end{bmatrix}$$

Partial Pivoting: Example

- Perform forward elimination

$$\begin{bmatrix} 1 & -0.7 & 0 \\ 0 & -0.9 & -1 \\ 0 & -0.0034 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 7 \\ -0.5 \\ 2.000333 \end{bmatrix}$$

$$\begin{bmatrix} & & \\ & & \\ & & \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \\ \\ \end{bmatrix}$$

Partial Pivoting Example

- Solve the equations through back substitution

$$\begin{bmatrix} 1 & -0.7 & 0 \\ 0 & 1 & 1.11111 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0.7 \\ 0.55555 \\ 0.999223 \end{bmatrix}$$

- $x_3 =$

- $x_2 =$

- $x_1 =$

Partial Pivoting: Example

- Compare the $[X]_{\text{exact}}$ and $[X]_{\text{calculated}}$ solution
- That they are different is a consequence of round-off error, but it illustrates the importance of partial pivoting

$$\bullet [X]_{\text{exact}} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}$$

$$\bullet [X]_{\text{calculated}} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \\ \\ \end{bmatrix}$$

Matrix Factorization

- Assume that our matrix **A** can be written as $\mathbf{A} = \mathbf{V}\mathbf{U}$ where **V** and **U** are triangular matrices

$$\bullet \begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \cdots & a_{nn} \end{bmatrix} =$$

$$\bullet \begin{bmatrix} & 0 & 0 & \cdots & 0 \\ & & 0 & \cdots & 0 \\ & & & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ & & & \cdots & \end{bmatrix} \begin{bmatrix} & & & \cdots \\ 0 & & & \cdots \\ & 0 & & \cdots \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & \end{bmatrix}$$

Matrix Factorization

- We generally look to solve
- Now we can decompose into and , so
- Then we solve for
- And then solve for

Matrix Factorization

- Method: Decompose $[A]$ to $[V]$ and $[U]$

$$[A] = [V] \cdot [U] = \begin{bmatrix} 1 & 0 & 0 \\ v_{21} & 1 & 0 \\ v_{31} & v_{32} & 1 \end{bmatrix} \begin{bmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{bmatrix}$$

- $[U]$ is the same as the $[U]$ at the end of the first step
- $[V]$ is obtained using the $[V]$ that were used in the first step

LU Decomposition: Example

- Let's start by finding $[U]$ matrix using the forward elimination procedure of Gaussian elimination

- $$\begin{bmatrix} 25 & 5 & 1 \\ 64 & 8 & 1 \\ 144 & 12 & 1 \end{bmatrix}$$

- $$\text{Row2} - \left[\frac{\text{Row1}}{25} \right] \times = \begin{bmatrix} 0 & -4.8 & -1.56 \end{bmatrix}$$

- $$\text{Row3} - \left[\frac{\text{Row1}}{25} \right] \times = \begin{bmatrix} 25 & 5 & 1 \end{bmatrix}$$

Lower-Upper Decomposition: Example

- Finding the $[U]$ matrix using the forward elimination process of Gaussian elimination

- $$\begin{bmatrix} 25 & 5 & 1 \\ 0 & -4.8 & -1.56 \\ 0 & -16.8 & -4.76 \end{bmatrix}$$

- $$\text{Row3} - \left[\frac{\text{Row2}}{-4.8} \right] \times \begin{bmatrix} 25 & 5 & 1 \\ 0 & -4.8 & -1.56 \end{bmatrix}$$

- $$[U] = \begin{bmatrix} \\ \\ \end{bmatrix}$$

LU Decomposition: Example

- Finding the $[V]$ matrix using the multipliers used during the forward elimination process
- From the first step of forward elimination

$$\bullet \begin{bmatrix} 1 & 0 & 0 \\ v_{21} & 1 & 0 \\ v_{31} & v_{32} & 1 \end{bmatrix}$$

$$\bullet \Rightarrow v_{21} =$$

$$\bullet \Rightarrow v_{31} =$$

LU Decomposition: Example

- From the second step of forward elimination

- $$\begin{bmatrix} 25 & 5 & 1 \\ 0 & -4.8 & -1.56 \\ 0 & -16.8 & -4.76 \end{bmatrix}$$

- $\Rightarrow v_{32} =$

LU Decomposition: Example

- $[V] = \begin{bmatrix} 1 & 0 & 0 \\ 2.56 & 1 & 0 \\ 5.76 & 3.5 & 1 \end{bmatrix}$

- Cross-check: does $[V][U] = [A]$?

-

- $[V][U] = \begin{bmatrix} 1 & 0 & 0 \\ 2.56 & 1 & 0 \\ 5.76 & 3.5 & 1 \end{bmatrix} \begin{bmatrix} 25 & 5 & 1 \\ 0 & -4.8 & -1.56 \\ 0 & 0 & 0.7 \end{bmatrix} =$

LU Decomposition: Example

- Now let's use VU factorization to solve the following set of linear equations

$$\bullet \begin{bmatrix} 25 & 5 & 1 \\ 64 & 8 & 1 \\ 144 & 12 & 1 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} 106.8 \\ 177.2 \\ 279.2 \end{bmatrix}$$

- Using the procedure for finding the $[V]$ and $[U]$ matrices

$$\bullet [A] = [V][U] = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix} \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

LU Decomposition: Example

- Complete the forward substitution to solve for $[Z] : [L][Z] = [X]$

- $z_1 = 106.8$

$$\Rightarrow z_2 =$$

- $=$

$$=$$

$$z_3 =$$

- $=$

$$=$$

- $\Rightarrow [Z] = \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} \\ \\ \end{bmatrix}$

LU Decomposition: Example

- Now set $[U][X] = [Z]$

$$\bullet \begin{bmatrix} 25 & 5 & 1 \\ 0 & -.48 & -1.56 \\ 0 & 0 & 0.7 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} \\ \\ \end{bmatrix}$$

- Solve for $[X]$
- The three equations become
 -
 -
 -

LU Decomposition: Example

- From the 3rd equation

- $0.7a_3 = 0.735$

- $\Rightarrow a_3 =$

-

- $=$

- Substitute a_3 into the second equation

- $-4.8a_2 - 1.56a_3 = -96.21$

- $\Rightarrow a_2 =$

-

- $=$

- $=$

LU Decomposition: Example

- Substituting a_3 and a_2 using the first equation

- $25a_1 + 5a_2 + a_3 = 106.8$

$$\Rightarrow a_1 =$$

- $=$

$$=$$

- Hence the final solution is

- $\begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} \\ \\ \end{bmatrix}$