Computational Methods
CMSC/AMSC/MAPL 460

Linear Systems, LU Decomposition,

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Matrix norms

- Can be defined using corresponding vector norms
  - Two norm
  - One norm
  - Infinity norm
- Two norm is hard to define … need to find maximum singular value
  - related to idea that matrix acting on unit sphere converts it in to an ellipsoid
- Frobenius norm is defined just using matrix elements

$$\|A\|_2 = \max_{\|x\|_2=1} \|Ax\|_2$$

$$\|A\|_1 = \max_{\|x\|_1=1} \|Ax\|_1$$

$$= \max_{j=1,\ldots,n} \sum_{i=1}^{m} |a_{ij}|$$

$$\|A\|_\infty = \max_{\|x\|_1=1} \|Ax\|_\infty$$

$$= \max_{i=1,\ldots,n} \sum_{j=1}^{m} |a_{ij}|$$

$$\|A\|_F = \left(\sum_{i=1}^{m} \sum_{j=1}^{n} |a_{i,j}|^2 \right)^{1/2}$$

Condition Number of a Matrix

A measure of how close a matrix is to singular

$$\text{cond}(A) = \kappa(A) = \|A\| \cdot \|A^{-1}\|$$

$$= \frac{\text{maximum stretch}}{\text{maximum shrink}} = \frac{\max_i |\lambda_i|}{\min_i |\lambda_i|}$$

- $\text{cond}(I) = 1$
- $\text{cond}(\text{singular matrix}) = \infty$

Solving Linear Systems

- One idea compute inverse
- Not usually a good idea
  - (unless inverse is computable easily and accurately using some matrix property)
- Leads to increased errors, and is more expensive usually

$$Ax = b$$

$$7x = 21$$

$$x = \frac{21}{7} = 3$$

$$x = 7^{-1} \times 21$$

$$= .142857 \times 21 = 2.99997$$

Representing linear systems as matrix-vector equations

$$\begin{aligned} 10x_1 - 7x_2 &= 7 \\ -3x_1 + 2x_2 + 6x_3 &= 4 \\ 5x_1 - x_2 + 5x_3 &= 6 \end{aligned}$$

- Represent it as a matrix-vector equation (linear system)
- We will apply the familiar elimination technique, and then see its matrix equivalent

$$\begin{pmatrix} 10 & -7 & 0 \\ -3 & 2 & 6 \\ 5 & -1 & 5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 7 \\ 4 \\ 6 \end{pmatrix}$$

Cramer's Rule

- Familiar algorithm often taught in school years
- Get determinant of the matrix
  - Determinant defined by recursion
  - Complexity of determinant determination
  - Factorial and polynomial complexity
- Get determinant of the matrix with column corresponding to variable replaced with the right hand side

$$x_1 = \frac{\begin{vmatrix} b_1 & a_{12} \\ b_2 & a_{22} \end{vmatrix}}{\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix}}, \quad x_2 = \frac{\begin{vmatrix} a_{11} & b_1 \\ a_{21} & b_2 \end{vmatrix}}{\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix}}$$

Cramer's Rule (contd.)

- In 3D

$$(x,y,z) = \left(\frac{\begin{vmatrix} d_1 & b_1 & c_1 \\ d_2 & b_2 & c_2 \\ d_3 & b_3 & c_3 \end{vmatrix}}{D}, \frac{\begin{vmatrix} a_1 & d_1 & c_1 \\ a_2 & d_2 & c_2 \\ a_3 & d_3 & c_3 \end{vmatrix}}{D}, \frac{\begin{vmatrix} a_1 & b_1 & d_1 \\ a_2 & b_2 & d_2 \\ a_3 & b_3 & d_3 \end{vmatrix}}{D} \right).$$

- In higher dimensions it generalizes similarly
- Home reading:
  - Why does Cramer's rule work?
  - (From "Cramer's rule article *Eric W. Weisstein* math encyclpedia … available online)

Gaussian Elimination

- Zero elements of first column below 1st row
  - multiplying 1st row by 0.3 and add to 2nd row
  - multiplying 1st row by -0.5 and add to 3rd row
  - Results in
- Zero elements of first column below 2nd row
  - Swap rows
  - Multiply 2nd row by 0.04 and add to 3rd

$$\begin{pmatrix} 10 & -7 & 0 \\ -3 & 2 & 6 \\ 5 & -1 & 5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 7 \\ 4 \\ 6 \end{pmatrix}$$

$$\begin{pmatrix} 10 & -7 & 0 \\ 0 & -0.1 & 6 \\ 0 & 2.5 & 5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 7 \\ 6.1 \\ 2.5 \end{pmatrix}$$

$$\begin{pmatrix} 10 & -7 & 0 \\ 0 & 2.5 & 5 \\ 0 & -0.1 & 6 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 7 \\ 2.5 \\ 6.1 \end{pmatrix}$$

$$\begin{pmatrix} 10 & -7 & 0 \\ 0 & 2.5 & 5 \\ 0 & 0 & 6.2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 7 \\ 2.5 \\ 6.2 \end{pmatrix}$$

Solution

- Start from last equation which can be solved by division
- Next substitute in the previous line and continue
- This describes the way to do the algorithm by hand
- How to represent it using matrices?
- Also, how do we solve another system that has the same matrix?
  - Upper triangular matrix we end up with will be the same, but the sequence of operations on the r.h.s needs to be repeated

$$6.2x_3 = 6.2$$

$$2.5x_2 + (5)(1) = 2.5.$$

$$10x_1 + (-7)(-1) = 7$$

$$x = \begin{pmatrix} 0 \\ -1 \\ 1 \end{pmatrix}$$

Gaussian Elimination: LU Matrix decomposition

- It turns out that Gaussian elimination corresponds to a particular matrix decomposition …
  - Product of permutation, lower triangular and upper triangular matrices
- What is a permutation matrix?
  - It rearranges a system of equations and changes the order.
  - Multiplying by it swaps the order of rows in a matrix
  - Essentially a rearrangement of the identity
  - Nice property: transpose is its inverse: $PP^T=I$

$$P = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

$$Px = b$$

$$x = P^T b$$

LU Decomposition

- What is an upper triangular matrix?
  - Elements below diagonal are zero

$$U = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 0 & 5 & 6 & 7 \\ 0 & 0 & 8 & 9 \\ 0 & 0 & 0 & 10 \end{pmatrix}$$

- Lower triangular matrix
- Elements above diagonal are zero
- Unit lower triangular matrix
- Elements along diagonal are one
- Upper triangular part of Gauss Elimination is clear …
  - final matrix we end up with

$$L = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 3 & 5 & 1 & 0 \\ 4 & 6 & 7 & 1 \end{pmatrix}$$

- What about lower triangular and permutation?

$LU=PA$

$$L = \begin{pmatrix} 1 & 0 & 0 \\ 0.5 & 1 & 0 \\ -0.3 & -0.04 & 1 \end{pmatrix} \quad U = \begin{pmatrix} 10 & -7 & 0 \\ 0 & 2.5 & 5 \\ 0 & 0 & 6.2 \end{pmatrix} \quad P = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$$

- Identify the elements of L and P?
- L has the multipliers we used in the elimination steps
- P has a record of the row swaps we did to avoid dividing by small numbers
- In fact we can write each step of Gaussian elimination in matrix form

$$U = M_{n-1}P_{n-1}\cdots M_2P_2M_1P_1A$$
$$L_1L_2\cdots L_{n-1}U = P_{n-1}\cdots P_2P_1A$$

$$A = \begin{pmatrix} 10 & -7 & 0 \\ -3 & 2 & 6 \\ 5 & -1 & 5 \end{pmatrix} \qquad \begin{aligned} L &= L_1 L_2 \cdots L_{n-1} \\ P &= P_{n-1} \cdots P_2 P_1 \end{aligned}$$

the matrices defined during the elimination are

$$P_1 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \; M_1 = \begin{pmatrix} 1 & 0 & 0 \\ 0.3 & 1 & 0 \\ -0.5 & 0 & 1 \end{pmatrix},$$

$$P_2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}, \; M_2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0.04 & 1 \end{pmatrix},$$

$$A = \begin{pmatrix} 10 & -7 & 0 \\ -3 & 2 & 6 \\ 5 & -1 & 5 \end{pmatrix}$$

$$P_1 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \; M_1 = \begin{pmatrix} 1 & 0 & 0 \\ 0.3 & 1 & 0 \\ -0.5 & 0 & 1 \end{pmatrix},$$

$$P_2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}, \; M_2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0.04 & 1 \end{pmatrix},$$

$$L_1 = \begin{pmatrix} 1 & 0 & 0 \\ 0.5 & 1 & 0 \\ -0.3 & 0 & 1 \end{pmatrix}, \; L_2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -0.04 & 1 \end{pmatrix},$$

Solving a system with the LU decomposition

$$Ax=b$$

$$LU=PA$$

$$P^T LUx=b$$

$$L[Ux]=Pb$$

Solve $Ly=Pb$

Then $Ux=y$

Solving a triangular system

```
x = zeros(n,1);
for k = n:-1:1
    x(k) = b(k)/U(k,k);
    i = (1:k-1)';
    b(i) = b(i) - x(k)*U(i,k);
end
```

```
x = zeros(n,1);
for k = n:-1:1
    j = k+1:n;
    x(k) = (b(k) - U(k,j)*x(j))/U(k,k);
end
```