



Introduction to Numerical Analysis for Engineers

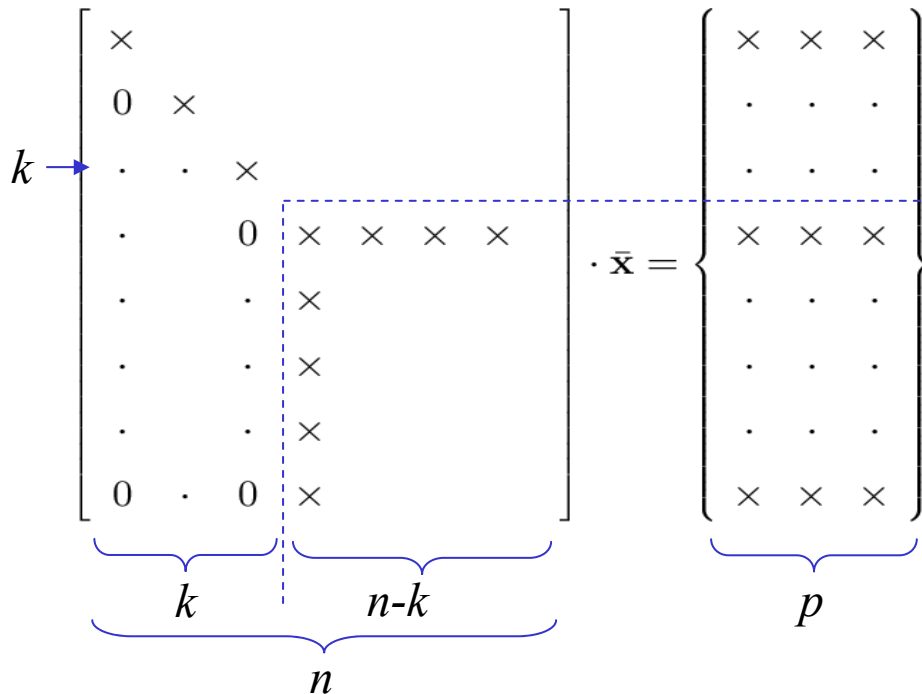
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Systems of Linear Equations Gaussian Elimination

Multiple Right-hand Sides

Reduction
Step k



Computation Count
Reduction Step k

$$(n - k)(n - k + p) \text{ Operations}$$

Total Computation Count

$$n \gg 1$$

Reduction

$$N_r = \sum_{k=1}^{n-1} (n - k)(n - k + p) \simeq \frac{1}{3}n^3 + \frac{1}{2}n^2(p - 1)$$

Back Substitution

$$N_b = \sum_{k=1}^{n-1} (n - k)p \simeq \frac{1}{2}n^2p$$

Reduction for each right-hand side inefficient.
However, RHS may be result of iteration and unknown a priori
(e.g. Euler's method) \rightarrow LU Factorization

$$n \gg 1 \Rightarrow N_r \gg N_b$$



Systems of Linear Equations

LU Factorization

The coefficient Matrix $\bar{\bar{A}}$ is decomposed as

$$\bar{\bar{A}} = \bar{\bar{L}} \cdot \bar{\bar{U}}$$

where $\bar{\bar{L}}$ is a lower triangular matrix
and $\bar{\bar{U}}$ is an upper triangular matrix

$$\bar{\bar{L}} = [l_{ij}] = \begin{bmatrix} l_{11} & 0 & \cdot & \cdot & \cdot & \cdot & 0 \\ l_{21} & l_{22} & \cdot & \cdot & \cdot & \cdot & \cdot \\ \vdots & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & l_{kk} & \cdot & \cdot & \cdot \\ \vdots & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ l_{n1} & \cdot & \cdot & \cdot & \cdot & l_{n,n-1} & l_{nn} \end{bmatrix}$$

Then the solution is performed in two simple steps

1. $\bar{\bar{L}}\vec{y} = \vec{b}$ Forward substitution

2. $\bar{\bar{U}}\vec{x} = \vec{y}$ Back substitution

$$\bar{\bar{U}} = [u_{ij}] =$$

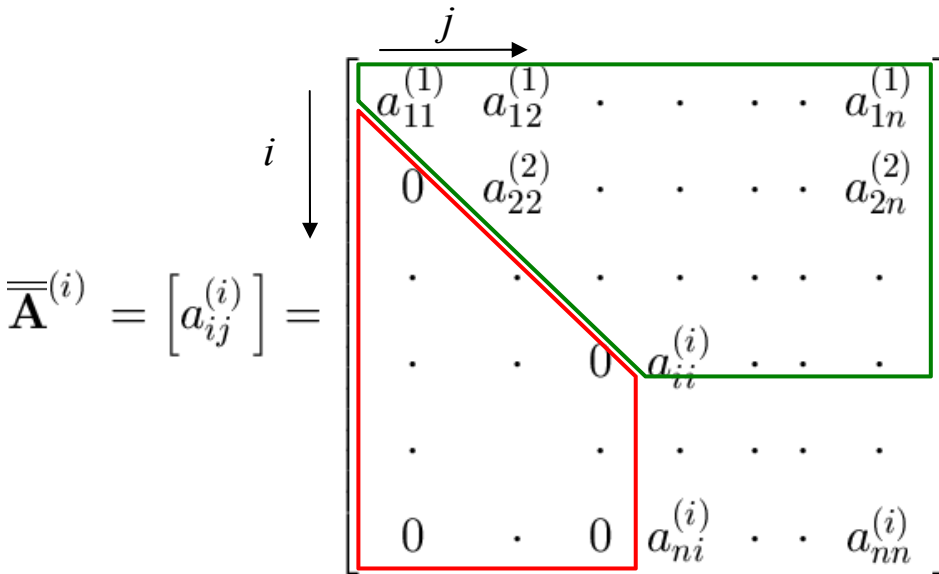
$$\begin{bmatrix} u_{11} & u_{12} & \cdot & \cdot & \cdot & \cdot & u_{1n} \\ 0 & u_{22} & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & u_{kk} & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & u_{n-1,n} \\ 0 & \cdot & \cdot & \cdot & \cdot & 0 & u_{nn} \end{bmatrix}$$

How to determine $\bar{\bar{L}}$ and $\bar{\bar{U}}$?



Systems of Linear Equations

LU Factorization



After reduction step $i-1$:

Above and on diagonal

$$i \leq j$$

Unchanged after step $i-1$

$$a_{ij}^{(n)} = \dots a_{ij}^{(i)}$$

Below diagonal

$$j < i$$

Become and remain 0 in step j

$$a_{ij}^{(n)} = \dots a_{ij}^{(j+1)} = 0$$

Change in reduction steps $1 - i-1$:

$$a_{ij}^{(k+1)} = a_{ij}^{(k)} - m_{ik} a_{kj}^{(k)}, \quad m_{ik} = a_{ik}^{(k)} / a_{kk}^{(k)}$$

Total change above diagonal

$$i \leq j : a_{ij}^{(i)} = a_{ij} - \sum_{k=1}^{i-1} m_{ik} a_{kj}^{(k)}$$

Total change below diagonal

$$i > j : a_{ij}^{(i)} = 0 = a_{ij} - \sum_{k=1}^j m_{ik} a_{kj}^{(k)}$$

Define

$$m_{ii} = 1, \quad i = 1, \dots, n$$

\Rightarrow

$$i \leq j : a_{ij} = \sum_{k=1}^i m_{ik} a_{kj}^{(k)}$$

$$i > j : a_{ij} = \sum_{k=1}^j m_{ik} a_{kj}^{(k)}$$

$$\Rightarrow a_{ij} = \sum_{k=1}^{\min(i,j)} m_{ik} a_{kj}^{(k)}$$



Systems of Linear Equations

LU Factorization

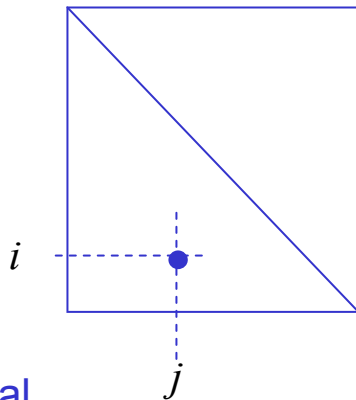
'Matrix product'

Sum stops at diagonal

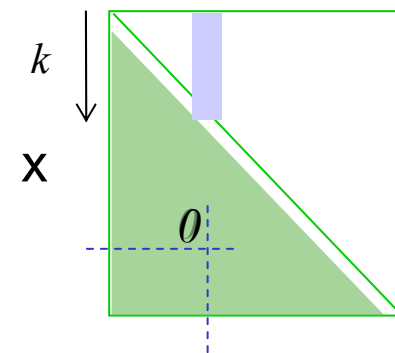
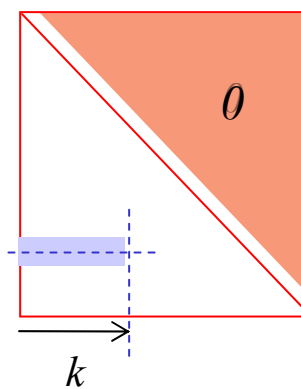
$$a_{ij} = \sum_{k=1}^{\min(i,j)} m_{ik} a_{kj}^{(k)}$$

Below diagonal

$$i > j :$$

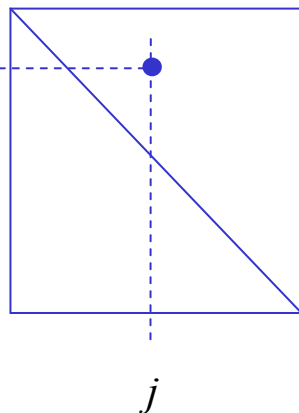


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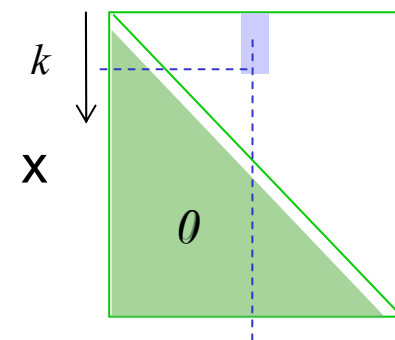
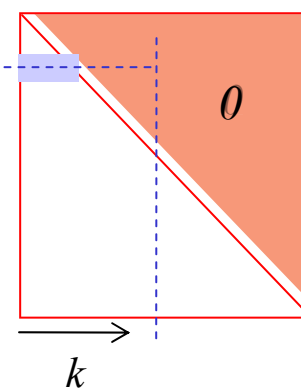


Above diagonal

$$i \leq j :$$



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Systems of Linear Equations

LU Factorization

GE Reduction directly yields LU factorization

$$\bar{\bar{\mathbf{A}}} = \bar{\bar{\mathbf{L}}} \cdot \bar{\bar{\mathbf{U}}}$$

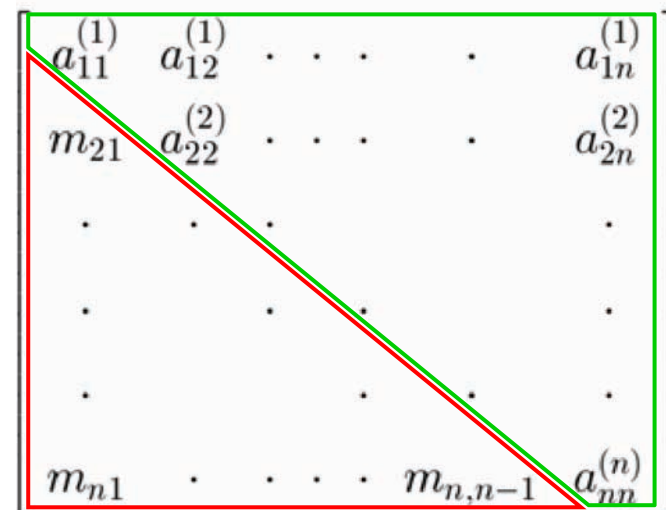
Lower triangular

$$\bar{\bar{\mathbf{L}}} = l_{ij} = \begin{cases} 0 & i < j \\ 1 & i = j \\ m_{ij} & i > j \end{cases}$$

Upper triangular

$$\bar{\bar{\mathbf{U}}} = u_{ij} = \begin{cases} a_{ij}^{(i)} & i \leq j \\ 0 & i > j \end{cases}$$

Compact storage



Lower diagonal implied

$$m_{ii} = 1, \quad i = 1, \dots, n$$



Systems of Linear Equations

Pivoting in LU Factorization

Before reduction, step k

$$\begin{bmatrix}
 a_{11}^{(1)} & a_{12}^{(1)} & \cdot & \cdot & \cdot & \cdot & a_{1n}^{(1)} \\
 m_{21} & a_{22}^{(2)} & \cdot & \cdot & \cdot & \cdot & a_{2n}^{(2)} \\
 \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\
 \cdot & \cdot & m_{k,k-1} & a_{kk}^{(k)} & \cdot & \cdot & \cdot \\
 \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\
 m_{n1} & \cdot & m_{n,k-1} & a_{nk}^{(k)} & \cdot & \dots & a_{nn}^{(n)}
 \end{bmatrix}$$

Pivoting if

$$|a_{ik}^{(k)}| \gg |a_{kk}^{(k)}|, \quad i > k$$

Interchange rows i and k

$$p_k = i$$

else

$$p_k = k$$

Pivot element vector

$$p_i, \quad i = 1, \dots, n$$

Forward substitution, step k

$$\bar{\bar{L}}\vec{y} = \vec{b}$$

Interchange rows i and k

$$\begin{bmatrix}
 b_1 \\
 \cdot \\
 \cdot \\
 b_k \\
 \cdot \\
 b_i \\
 \cdot \\
 b_n
 \end{bmatrix}$$

$$p_k = i \Rightarrow \begin{cases} b_i^{(k)} = b_k \\ b_k = b_i \\ b_i = b_i^{(k)} \end{cases}$$



Linear Systems of Equations Error Analysis

Function of one variable

$$y = f(x)$$

Condition number

$$\left| \frac{f(\bar{x}) - f(x)}{f(x)} \right| = K \left| \frac{\bar{x} - x}{x} \right|, \quad \bar{x} = x + \delta x$$

$$\left| \frac{\delta y}{y} \right| = K \left| \frac{\delta x}{x} \right|$$

The condition number K is a measure of the **amplification** of the **relative error** by the function $f(x)$

Linear systems

How is the relative error of $\bar{\mathbf{x}}$ dependent on errors in $\bar{\mathbf{b}}$?

$$\bar{\mathbf{A}}\bar{\mathbf{x}} = \bar{\mathbf{b}}$$

Example

$$\bar{\mathbf{A}} = \begin{bmatrix} 1.0 & 1.0 \\ 1.0 & 1.0001 \end{bmatrix}, \quad \det(\bar{\mathbf{A}}) = 0.0001$$

$$\bar{\mathbf{b}} = \begin{Bmatrix} 2 \\ 2 \end{Bmatrix} \Rightarrow \bar{\mathbf{x}} = \begin{Bmatrix} 2 \\ 0 \end{Bmatrix}$$

$$\bar{\mathbf{b}} = \begin{Bmatrix} 2 \\ 2.0001 \end{Bmatrix} \Rightarrow \bar{\mathbf{x}} = \begin{Bmatrix} 1 \\ 1 \end{Bmatrix}$$

Small changes in $\bar{\mathbf{b}}$ give large changes in $\bar{\mathbf{x}}$
The system is **ill-Conditioned**



Linear Systems of Equations Error Analysis

Vector and Matrix Norm

$$\|\bar{\mathbf{x}}\|_{\infty} = \max_i |x_i|$$

$$\|\bar{\mathbf{A}}\|_{\infty} = \max_i \sum_{j=1}^n |a_{ij}|$$

Properties

$$\bar{\mathbf{A}} \neq \bar{\mathbf{0}} \Rightarrow \|\bar{\mathbf{A}}\| > 0$$

$$\|\alpha \bar{\mathbf{A}}\| = |\alpha| \|\bar{\mathbf{A}}\|$$

$$\|\bar{\mathbf{A}} + \bar{\mathbf{B}}\| \leq \|\bar{\mathbf{A}}\| + \|\bar{\mathbf{B}}\|$$

$$\|\bar{\mathbf{A}}\bar{\mathbf{B}}\| \leq \|\bar{\mathbf{A}}\| \|\bar{\mathbf{B}}\|$$

$$\|\bar{\mathbf{A}}\bar{\mathbf{x}}\| \leq \|\bar{\mathbf{A}}\| \|\bar{\mathbf{x}}\|$$

Perturbed Right-hand Side

$$\bar{\mathbf{A}}\bar{\mathbf{x}} = \bar{\mathbf{b}}$$



$$\bar{\mathbf{A}}(\bar{\mathbf{x}} + \delta\bar{\mathbf{x}}) = \bar{\mathbf{b}} + \delta\bar{\mathbf{b}}$$

Subtract original equation

$$\bar{\mathbf{A}}\delta\bar{\mathbf{x}} = \delta\bar{\mathbf{b}}$$

$$\delta\bar{\mathbf{x}} = \bar{\mathbf{A}}^{-1}\delta\bar{\mathbf{b}}$$



$$\left. \begin{aligned} \|\delta\bar{\mathbf{x}}\| &\leq \|\bar{\mathbf{A}}^{-1}\| \|\delta\bar{\mathbf{b}}\| \\ \|\bar{\mathbf{b}}\| = \|\bar{\mathbf{A}}\bar{\mathbf{x}}\| &\leq \|\bar{\mathbf{A}}\| \|\bar{\mathbf{x}}\| \end{aligned} \right\} \Rightarrow$$

Relative Error Magnification

$$\frac{\|\delta\bar{\mathbf{x}}\|}{\|\bar{\mathbf{x}}\|} \leq \|\bar{\mathbf{A}}^{-1}\| \|\bar{\mathbf{A}}\| \frac{\|\delta\bar{\mathbf{b}}\|}{\|\bar{\mathbf{b}}\|}$$

Condition Number

$$K(\bar{\mathbf{A}}) = \|\bar{\mathbf{A}}^{-1}\| \|\bar{\mathbf{A}}\|$$



Linear Systems of Equations Error Analysis

Vector and Matrix Norm

$$\|\bar{\mathbf{x}}\|_{\infty} = \max_i |x_i|$$

$$\|\bar{\mathbf{A}}\|_{\infty} = \max_i \sum_{j=1}^n |a_{ij}|$$

Properties

$$\bar{\mathbf{A}} \neq \bar{\mathbf{0}} \Rightarrow \|\bar{\mathbf{A}}\| > 0$$

$$\|\alpha \bar{\mathbf{A}}\| = |\alpha| \|\bar{\mathbf{A}}\|$$

$$\|\bar{\mathbf{A}} + \bar{\mathbf{B}}\| \leq \|\bar{\mathbf{A}}\| + \|\bar{\mathbf{B}}\|$$

$$\|\bar{\mathbf{A}}\bar{\mathbf{B}}\| \leq \|\bar{\mathbf{A}}\| \|\bar{\mathbf{B}}\|$$

$$\|\bar{\mathbf{A}}\bar{\mathbf{x}}\| \leq \|\bar{\mathbf{A}}\| \|\bar{\mathbf{x}}\|$$

Perturbed Coefficient Matrix

$$(\bar{\mathbf{A}} + \delta\bar{\mathbf{A}})(\bar{\mathbf{x}} + \delta\bar{\mathbf{x}}) = \bar{\mathbf{b}}$$

Subtract unperturbed equation

$$\bar{\mathbf{A}}\delta\bar{\mathbf{x}} + \delta\bar{\mathbf{A}}(\bar{\mathbf{x}} + \delta\bar{\mathbf{x}}) = \bar{\mathbf{0}}$$

$$\delta\bar{\mathbf{x}} = -\bar{\mathbf{A}}^{-1} \delta\bar{\mathbf{A}}(\bar{\mathbf{x}} + \delta\bar{\mathbf{x}}) \simeq -\bar{\mathbf{A}}^{-1} \delta\bar{\mathbf{A}}\bar{\mathbf{x}}$$

$$\|\delta\bar{\mathbf{x}}\| \leq \|\bar{\mathbf{A}}^{-1}\| \|\delta\bar{\mathbf{A}}\| \|\bar{\mathbf{x}}\|$$

Relative Error Magnification

$$\frac{\|\delta\bar{\mathbf{x}}\|}{\|\bar{\mathbf{x}}\|} \leq \|\bar{\mathbf{A}}^{-1}\| \|\bar{\mathbf{A}}\| \frac{\|\delta\bar{\mathbf{A}}\|}{\|\bar{\mathbf{A}}\|}$$

Condition Number

$$K(\bar{\mathbf{A}}) = \|\bar{\mathbf{A}}^{-1}\| \|\bar{\mathbf{A}}\|$$



III-Conditioned System

$$\begin{bmatrix} 1.0 & 1.0 \\ 1.0 & 1.0001 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\det(\overline{\mathbf{A}}) = 0.0001$$

$$a_{11} = \frac{1.0001}{0.0001} = 10,001$$

$$a_{12} = \frac{-1}{0.0001} = -10,000$$

$$a_{21} = \frac{-1}{0.0001} = -10,000$$

$$a_{22} = \frac{1.0}{0.0001} = 10,000$$

$$\left. \begin{array}{l} \|\overline{\mathbf{A}}\|_{\infty} = 2.0001 \\ \|\overline{\mathbf{A}}^{-1}\|_{\infty} = 20,001 \end{array} \right\} \Rightarrow K(\overline{\mathbf{A}}) \simeq \boxed{40,000}$$

III-conditioned system

```
n=4
a = [ [1.0 1.0]' [1.0 1.0001] ] tbt6.m
b= [1 2]

ai=inv(a);
a_nrm=max( abs(a(1,1)) + abs(a(1,2)) ,
           abs(a(2,1)) + abs(a(2,2)) )
ai_nrm=max( abs(ai(1,1)) + abs(ai(1,2)) ,
           abs(ai(2,1)) + abs(ai(2,2)) )
k=a_nrm*ai_nrm

r=ai * b

x=[0 0];
m21=a(2,1)/a(1,1);
a(2,1)=0;
a(2,2) = radd(a(2,2), -m21*a(1,2), n);
b(2)   = radd(b(2), -m21*b(1), n);

x(2)   = b(2)/a(2,2);
x(1)   = (radd(b(1), -a(1,2)*x(2), n))/a(1,1);
x'
```



Well-Conditioned System

$$\begin{bmatrix} 0.0001 & 1.0 \\ 1.0 & 1.0 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\det(\overline{\mathbf{A}}) = 0.9999$$

$$a_{11} = \frac{-1}{0.9999} = -1,0001$$

$$a_{12} = \frac{1}{0.9999} = 1.0001$$

$$a_{21} = \frac{1}{0.9999} = 1.0001$$

$$a_{22} = \frac{-0.0001}{0.9999} = -0.0001$$

$$\left. \begin{aligned} \|\overline{\mathbf{A}}\|_{\infty} &= 2.0 \\ \|\overline{\mathbf{A}}^{-1}\|_{\infty} &= 2.0002 \end{aligned} \right\} \Rightarrow K(\overline{\mathbf{A}}) \simeq \boxed{4}$$

Well-conditioned system

4-digit Arithmetic

```
n=4
a = [ [0.0001 1.0]' [1.0 1.0] ]  tbt7.m
b= [1 2]'

ai=inv(a);
a_nrm=max( abs(a(1,1)) + abs(a(1,2)) ,
           abs(a(2,1)) + abs(a(2,2)) )
ai_nrm=max( abs(ai(1,1)) + abs(ai(1,2)) ,
           abs(ai(2,1)) + abs(ai(2,2)) )
k=a_nrm*ai_nrm

r=ai * b

x=[0 0];
m21=a(2,1)/a(1,1);
a(2,1)=0;
a(2,2) = radd(a(2,2), -m21*a(1,2), n);
b(2)   = radd(b(2), -m21*b(1), n);

x(2)   = b(2)/a(2,2);
x(1)   = (radd(b(1), -a(1,2)*x(2), n))/a(1,1);
x'
```

Algorithmically ill-conditioned