

FF505
Computational Science

Lecture 2
Linear Algebra with Matlab
Linear Systems

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Outline

1. An Example: Electrical Networks
2. Matrices and Vectors in MatLab
3. Solving Linear Systems

Resume

- MATLAB, numerical computing vs symbolic computing
- MATLAB Desktop
- Script files
- 1D and 2D arrays
- Plot
- Interacting with matlab

- matrices and vectors
- solving linear systems
- determinants
- linear transformation
- eigenvalues and eigenvectors
- diagonalization?
- spectral theorem?

- projectile trajectory?
- car market assignment?

Outline

1. An Example: Electrical Networks
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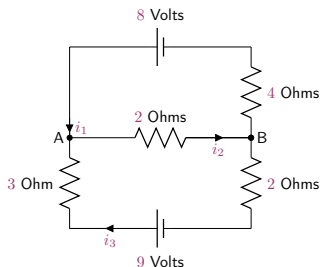
We want to determine the amount of current present in each branch.

Kirchoff's Laws

- At every node, the sum of the incoming currents equals the sum of the outgoing currents
- Around every closed loop, the algebraic sum of the voltage gains must equal the algebraic sum of the voltage drops.

Voltage drops E (by Ohm's law)

$$E = iR$$



$$i_1 - i_2 + i_3 = 0$$

$$-i_1 + i_2 - i_3 = 0$$

$$4i_1 + 2i_2 = 8$$

$$2i_2 + 5i_3 = 9$$

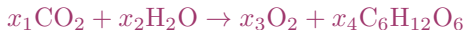
node A

node B

top loop

bottom loop

Chemical Equations



To balance the equation, we must choose x_1, x_2, x_3, x_4 so that the numbers of carbon, hydrogen, and oxygen atoms are the same on each side of the equation.

$$x_1 = 6x_4$$

carbon atoms

$$2x_1 + x_2 = 2x_3 + 6x_4$$

oxygen

$$2x_2 = 12x_4$$

hydrogen

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Matrix Multiplication

$$i_1 - i_2 + i_3 = 0$$

$$-i_1 + i_2 - i_3 = 0$$

$$4i_1 + 2i_2 = 8$$

$$2i_2 + 5i_3 = 9$$

node A

node B

top loop

bottom loop

$$\begin{bmatrix} 1 & -1 & 1 \\ -1 & 1 & -1 \\ 4 & 2 & 0 \\ 0 & 2 & 5 \end{bmatrix} \begin{bmatrix} i_1 \\ i_2 \\ i_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 8 \\ 9 \end{bmatrix}$$

$$Ax = \mathbf{b}$$

Matrix Multiplication

$$\begin{aligned}x_1 &= 6x_4 \\2x_1 + x_2 &= 2x_3 + 6x_4 \\2x_2 &= 12x_4\end{aligned}$$

carbon atoms

oxygen

hydrogen

$$\begin{bmatrix} 1 & 0 & 0 & -6 \\ 2 & 1 & 2 & 6 \\ 0 & 2 & 0 & 12 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$A\mathbf{x} = \mathbf{0}$$

Creating Matrices

```
eye(4) % identity matrix
zeros(4) % matrix of zero elements
ones(4) % matrix of one elements
```

```
A=rand(8)
triu(A) % upper triangular matrix
tril(A)
diag(A) % diagonal
```

```
>> [ eye(2), ones(2,3); zeros(2),
      [1:3;3:-1:1] ]
```

```
ans =
```

```
1 0 1 1 1
0 1 1 1 1
0 0 1 2 3
0 0 3 2 1
```

Can you create this matrix in one line of code?

```
-5    0    0    0    0    0    0    1    1    1    1
  0   -4    0    0    0    0    0    0    1    1    1
  0    0   -3    0    0    0    0    0    0    1    1
  0    0    0   -2    0    0    0    0    0    0    1
  0    0    0    0   -1    0    0    0    0    0    0
  0    0    0    0    0    0    0    0    0    0    0
  0    0    0    0    0    0    1    0    0    0    0
  1    0    0    0    0    0    0    2    0    0    0
  1    1    0    0    0    0    0    0    3    0    0
  1    1    1    0    0    0    0    0    0    4    0
  1    1    1    1    0    0    0    0    0    0    5
```

Matrix-Matrix Multiplication

In the product of two matrices $A * B$,
the number of columns in A must equal the number of rows in B .

The product AB has the same number of rows as A and the same number of columns as B . For example

```
>> A=randi(10,3,2) % returns a 3-by-2 matrix containing pseudorandom integer values
drawn from the discrete uniform distribution on 1:10
A =
     6    10
    10     4
     5     8

>> C=randi(10,2,3)*100
C =
    1000    900    400
     200    700    200

>> A*C % matrix multiplication
ans =
    8000   12400   4400
   10800   11800   4800
    6600   10100   3600
```

Exercise: create a small example to show that in general, $AB \neq BA$.

Matrix Operations

```
%% matrix operations  
A * C % matrix multiplication  
B = [5 6; 7 8; 9 10] * 100 % same dims as A  
A .* B % element-wise multiplication  
% A .* C or A * B gives error - wrong dimensions  
A .^ 2  
1./B  
log(B) % functions like this operate element-wise on vecs or matrices  
exp(B) % overflow  
abs(B)  
v = [-3:3] % = [-3 -2 -1 0 1 2 3]  
-v % -1*v  
  
v + ones(1,length(v))  
% v + 1 % same  
  
A' % (conjugate) transpose
```

Array Operations

- **Addition/Subtraction:** trivial
- **Multiplication:**
 - of an array by a scalar is easily defined and easily carried out.
 - of two arrays is not so straightforward:
MATLAB uses two definitions of multiplication:
 - array multiplication (also called element-by-element multiplication)
 - matrix multiplication
- **Division and exponentiation** MATLAB has two forms on arrays.
 - element-by-element operations
 - matrix operations

↪ Remark:

the operation division by a matrix is not defined. In MatLab it is defined but it has other meanings.

Element-by-Element Operations

Symbol	Operation	Form	Examples
+	Scalar-array addition	$A + b$	$[6, 3] + 2 = [8, 5]$
-	Scalar-array subtraction	$A - b$	$[8, 3] - 5 = [3, -2]$
+	Array addition	$A + B$	$[6, 5] + [4, 8] = [10, 13]$
-	Array subtraction	$A - B$	$[6, 5] - [4, 8] = [2, -3]$
.*	Array multiplication	$A.*B$	$[3, 5].*[4, 8] = [12, 40]$
./	Array right division	$A./B$	$[2, 5]./[4, 8] = [2/4, 5/8]$
.\	Array left division	$A.\B$	$[2, 5].\[4, 8] = [2\4, 5\8]$
.^	Array exponentiation	$A.^B$	$[3, 5].^2 = [3^2, 5^2]$ $2.^[3, 5] = [2^3, 2^5]$ $[3, 5].^[2, 4] = [3^2, 5^4]$

Backslash or Matrix Left Division

$A \setminus B$ is roughly like $\text{INV}(A) * B$ except that it is computed in a different way:
 $X = A \setminus B$ is the solution to the equation $A * X = B$ computed by Gaussian elimination.

Slash or right matrix division:

A / B is the matrix division of B into A , which is roughly the same as $A * \text{INV}(B)$, except it is computed in a different way. More precisely, $A / B = (B' \setminus A')$ '.

Dot and Cross Products

`dot(A,B)` **inner** or **scalar product**: computes the projection of a vector on the other. eg. `dot(Fr,r)` computes component of force \mathbf{F} along direction \mathbf{r}
//Inner product, generalization of dot product

```
v=1:10  
u=11:20  
u*v' % inner or scalar product  
ui=u+i  
ui'  
v*ui' % inner product of C^n  
norm(v,2)  
sqrt(v*v')
```

`cross(A,B)` cross product: eg: moment $\mathbf{M} = \mathbf{r} \times \mathbf{F}$

Exercise: Projectile trajectory

\mathbf{p} position vector

$$\mathbf{p}_t = \mathbf{p}_0 + \mathbf{u}_t s_m t + \frac{g t^2}{2}$$

s_m muzzle velocity (speed at which the projectile left the weapon)

\mathbf{u}_t is the direction the weapon was fired

$$g = -9.81 \text{ms}^{-1}$$

Predict the landing spot

$$t_i = \frac{-u_x s_m \pm \sqrt{u_y^2 s_m^2 - 2g_y(p_{y0} - p_{yt})}}{g_y} \quad \mathbf{p}_E = \begin{bmatrix} p_{x0} + u_x s_m t_i \\ p_{y0} \\ p_{z0} + u_z s_m t_i \end{bmatrix}$$

Plot the trajectory in 2D.

Exercise: Projectile trajectory

Given a firing point \mathbf{S} and s_m and a target point \mathbf{E} , we want to know the firing direction \mathbf{u} , $|\mathbf{u}| = 1$.

$$E_x = S_x + u_x s_m t_i + \frac{1}{2} g_x t_i^2$$

$$E_y = S_y + u_y s_m t_i + \frac{1}{2} g_y t_i^2$$

$$E_z = S_z + u_z s_m t_i + \frac{1}{2} g_z t_i^2$$

$$1 = u_x^2 + u_y^2 + u_z^2$$

four eq. in four unknowns, leads to:

$$|\mathbf{g}|^2 t_i^4 - 4(\mathbf{g} \cdot \mathbf{\Delta} + s_m^2) t_i^2 + 4|\mathbf{\Delta}|^2 = 0, \quad \mathbf{\Delta} = \mathbf{E} - \mathbf{S}$$

solve in t , and interpret the solution.

Reshaping

```
%% reshape and replication  
A = magic(3) % magic square  
A = [A [0;1;2]]  
reshape(A,[4 3]) % columnwise  
reshape(A,[2 6])  
v = [100;0;0]  
A+v  
A + repmat(v,[1 4])
```

Matrix Functions

Eigenvalues and eigenvectors:

```
A = ones(6)
trace(A)
A = A - tril(A)-triu(A,2)
eig(A)

diag(ones(3,1),-1)
[V,D]=eig(diag(1:4))

rank(A) % rank of A
orth(A) % orthonormal basis
```

Visualizing Eigenvalues

```
A=[5/4,0;0,3/4];
eigshow(A) %effect of operator A on unit
           vector
```

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Systems of Linear Equations

How many solutions have these linear systems? Find it out using the graphical approach.

$$6x - 10y = 2$$

$$3x - 4y = 5$$

% plot functions in implicit form
`ezplot`

$$3x - 4y = 5$$

$$6x - 8y = 10$$

$$3x - 4y = 5$$

$$6x - 8y = 3$$

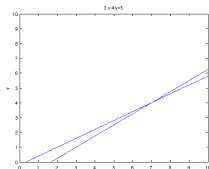
Systems of Linear Equations

$$6x - 10y = 2$$

$$3x - 4y = 5$$

% plot functions in implicit form
`ezplot('6*x-10*y=2',[0 10 0 10]),`
`hold,`
`ezplot('3*x-4*y=5',[0 10 0 10])`

has one single solution

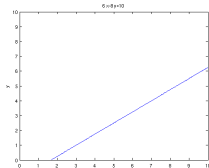


$$3x - 4y = 5$$

$$6x - 8y = 10$$

`ezplot('3*x-4*y=5',[0 10 0 10]),`
`hold,`
`ezplot('6*x-8*y=10',[0 10 0 10])`

has infinite solutions

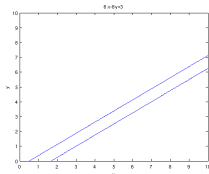


$$3x - 4y = 5$$

$$6x - 8y = 3$$

`ezplot('3*x-4*y=5',[0 10 0 10]),`
`hold,`
`ezplot('6*x-8*y=3',[0 10 0 10])`

has no solution



Matrix Form

The linear system:

$$2x_1 + 9x_2 = 5$$

$$3x_1 - 4x_2 = 7$$

can be expressed in vector-matrix form as:

$$\begin{bmatrix} 2 & 9 \\ 3 & -4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 5 \\ 7 \end{bmatrix}$$

In general, a set of m equations in n unknowns can be expressed in the form $A\mathbf{x} = \mathbf{b}$, where A is $m \times n$, \mathbf{x} is $n \times 1$ and \mathbf{b} is $m \times 1$.

The inverse of A is denoted A^{-1} and has property that

$$A^{-1}A = AA^{-1} = I$$

Hence the solution to our system is:

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

Inverse and Determinant

Compute the **inverse** and the **determinant** of this matrix in Matlab:

```
>> A=[3 -4; 6 -8];
```

Has the system solutions?

What about the system in the previous slide? What are its solutions?

A matrix is singular if $\det(A) = |A| = 0$

Inverse of a square matrix A is defined only if A is nonsingular.

If A is singular, the system has no solution

```
>> A=[3 -4; 6 -8];  
>> det(A)  
ans =  
    0  
>> inv(A)  
Warning: Matrix is singular to working precision.  
ans =  
    Inf Inf  
    Inf Inf
```

For a 2×2 matrix the matrix inverse is

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \quad A^{-1} = \frac{1}{|A|} \begin{bmatrix} d & -c \\ -b & a \end{bmatrix}^T = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

For a 3×3 matrix

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

the matrix inverse is

$$A^{-1} = \frac{1}{|A|} \begin{bmatrix} + \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} & - \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} & + \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix} \\ - \begin{vmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} & + \begin{vmatrix} a_{11} & a_{13} \\ a_{31} & a_{33} \end{vmatrix} & - \begin{vmatrix} a_{11} & a_{12} \\ a_{31} & a_{32} \end{vmatrix} \\ + \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix} & - \begin{vmatrix} a_{11} & a_{13} \\ a_{21} & a_{23} \end{vmatrix} & + \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} \end{bmatrix}^T$$

Calculating the inverse

$$A^{-1} = \frac{1}{|A|} \text{adj}(A)$$

$\text{adj}(A)$ is the adjugate matrix of A :

1. Calculate the (i, j) **minor** of A , denoted M_{ij} , as the determinant of the $(n-1) \times (n-1)$ matrix that results from deleting row i and column j of A .
2. Calculate the **cofactor** matrix of A , as the $n \times n$ matrix C whose (i, j) entry is the (i, j) cofactor of A

$$C_{ij} = (-1)^{i+j} M_{ij}$$

3. set $\text{adj}(A)_{ij} = C_{ji}$

Left Division Method

- $\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$ rarely applied in practice because calculation is likely to introduce numerical inaccuracy
- The inverse is calculated by LU decomposition, the matrix form of Gaussian elimination.

% left division method

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

$$\begin{aligned} \mathbf{A} &= \mathbf{LU} \\ \mathbf{PA} &= \mathbf{LU} \end{aligned}$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = \begin{bmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & l_{33} \end{bmatrix} \begin{bmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{bmatrix}$$

- for a matrix A , $n \times n$, $\det(A) \neq 0 \Leftrightarrow$ rank of A is n
 - for a system $Ax = b$ with m equations and n unknowns a solution exists iff $\text{rank}(A) = \text{rank}([Ab]) = r$
 - if $r = n \rightsquigarrow$ unique
 - if $r < n \rightsquigarrow$ infinite sol.
 - for a homogeneous system $Ax = 0$ it is always $\text{rank}(A) = \text{rank}([Ab])$ and there is a nonzero solution iff $\text{rank}(A) < n$
- $A \setminus b$ works for square and nonsquare matrices. If nonsquare ($m < n$) then the system is undetermined (infinite solutions). $A \setminus b$ returns one variable to zero
- $A \setminus b$ does not work when $\det(A) = 0$.

```
>> A=[2, -4,5;-4,-2,3;2,6,-8];
>> b=[-4;4;0];
>> rank(A)
ans =
     2
>> rank([A,b])
ans =
     2
>> x=A\b
Warning: Matrix is singular to working
precision.
x =
    NaN
    NaN
    NaN
```

However since

$$\text{rank}(A) = \text{rank}([Ab])$$

an infinite number of solutions exist (**undetermined system**).

$x = \text{pinv}(A)b$ solves with pseudoinverse and $\text{rref}([A,b])$ finds the reduced row echelon form

Overdetermined Systems

An overdetermined system is a set of equations that has more independent equations than unknowns ($m > n$).

For such a system the matrix inverse method will not work because the A matrix is not square.

However, some overdetermined systems have exact solutions, and they can be obtained with the left division method $\mathbf{x} = \mathbf{A} \setminus \mathbf{b}$

For other overdetermined systems, no exact solution exists. We need to check the ranks of A and $[Ab]$ to know whether the answer is the exact solution. If a solution does not exist, the left-division answer is the least squares solution.

Flowchart for Linear System Solver

