

Lecture 06 – Algorithms for Linear Algebra

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NERS/ENGR 570 - Methods and Practice of Scientific Computing (F22)



COLLEGE OF ENGINEERING

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Outline

- Recap of our progress so far...
- Linear Algebra Fundamentals
- Dense vs. Sparse Linear Systems
 - Where do they come from?
- Solution of Linear Systems
 - Direct methods and LU Decomposition

Learning Objectives: By the end of Today's Lecture you should be able to

- (*Knowledge*) interpretate meaning of some vector norms
- (*Value/Knowledge*) explain how to think about programming equations in linear algebra
- (*Knowledge*) program the basic kernels needed for linear algebra
- (*Knowledge*) explain the differences of dense and sparse matrix storage formats

Overview

- What types of problems are solved with Linear Algebra?
 - Linear systems of equations
 - Eigenvalue problems
 - Matrix factorizations
 - Overdetermined system of equations
 - Least-Squares (Data fitting)

	Embed	SPEC	DB	Games	ML	HPC
1 Finite State Mach.	Red	Red	Red	Yellow	Yellow	Light Blue
2 Combinational	Red	Light Blue	Green	Light Blue	Green	Light Blue
3 Graph Traversal	Red	Yellow	Yellow	Yellow	Red	Light Blue
4 Structured Grid	Red	Red	Light Blue	Yellow	Light Blue	Red
5 Dense Matrix	Red	Red	Yellow	Red	Red	Red
6 Sparse Matrix	Yellow	Yellow	Light Blue	Red	Red	Red
7 Spectral (FFT)	Yellow	Light Blue	Light Blue	Yellow	Yellow	Red
8 Dynamic Prog	Yellow	Light Blue	Red	Light Blue	Red	Light Blue
9 N-Body	Light Blue	Yellow	Light Blue	Yellow	Light Blue	Red
10 MapReduce	Light Blue	Green	Red	Light Blue	Red	Red
11 Backtrack/ B&B	Light Blue	Light Blue	Yellow	Light Blue	Red	Light Blue
12 Graphical Models	Light Blue	Light Blue	Yellow	Light Blue	Red	Light Blue
13 Unstructured Grid	Light Blue	Light Blue	Light Blue	Yellow	Yellow	Red



Linear Algebra Fundamentals

Basics of Linear Systems

$$\rightarrow a_{11}x_1 + a_{21}x_2 + \dots + a_{n1}x_n = b_1$$

$$a_{12}x_1 + a_{22}x_2 + \dots + a_{n2}x_n = b_2$$

$$\dots$$

$$\underline{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \quad \underline{b} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}$$

$m > n \rightarrow$ over-determined (Tall-Skinny)
 $m = n \rightarrow$ square, fully determined
 $m < n \rightarrow$ underdetermined (short-wide)

$$\underline{A} = \begin{bmatrix} a_{11} & a_{21} & \dots & a_{n1} \\ a_{12} & a_{22} & \dots & a_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \quad \underline{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

$a_{n,m}$

$$\underline{A} \underline{x} = \underline{b}$$

dot product
matrix vector multiplication

The Residual and Norms

$$\underline{A}\underline{x} = \underline{b} \quad b - Ax = 0$$

$$\tilde{x} = x + \varepsilon \quad b - A\tilde{x} \neq 0 = \text{residual } (\vec{r})$$

residual "dual" of the error

$$\|\vec{r}\|$$

$$\|\vec{r}\|_1$$

$$\|\vec{r}\|_2$$

$$\|\vec{r}\|_\infty$$



Norms of Vectors

$$\begin{aligned} \|r\|_1 &= |r_1| + |r_2| + \dots + |r_n| && \rightarrow \text{"total error"} \\ \|r\|_2 &= \sqrt{r_1^2 + r_2^2 + \dots + r_n^2} && \rightarrow \text{"average error"} \\ \|r\|_\infty &= \max(|r_1|, |r_2|, \dots, |r_n|) && \rightarrow \text{"max. local error"} \end{aligned}$$

Mean Square Error

$$\|r\|_p = (|r_1|^p + |r_2|^p + \dots + |r_n|^p)^{1/p}$$

$$\|r\|_2 = \sqrt{r^T \cdot r}$$

Going from Chalkboard to Terminal

$$\|r\|_2 = \left(\sum_{i=1}^n r_i^2 \right)^{1/2} \leftarrow$$

norm2(v(:), n) {

double sum = 0.0;

for(i=0; i++; i < n) {

sum += v(i)*v(i)

}
return sum;

}

1: sum variable

2: pow(v(i), 2)
or
multiplication

Types of Linear Algebra Operations

BLAS

~~scalar-scalar: 0 loops~~

scalar-vector: 1 loop } Level 1

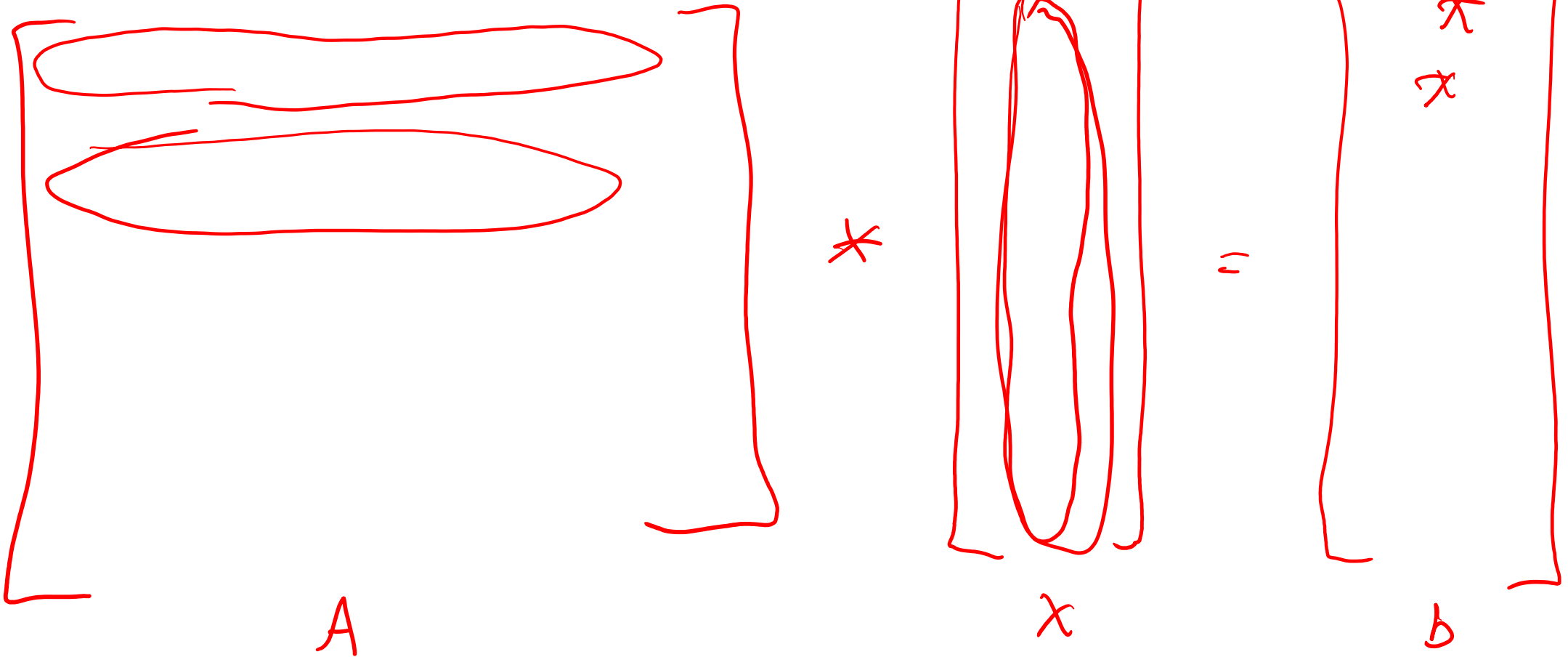
vector-vector: 1 loop }

Matrix-vector: 2 loops } Level 2

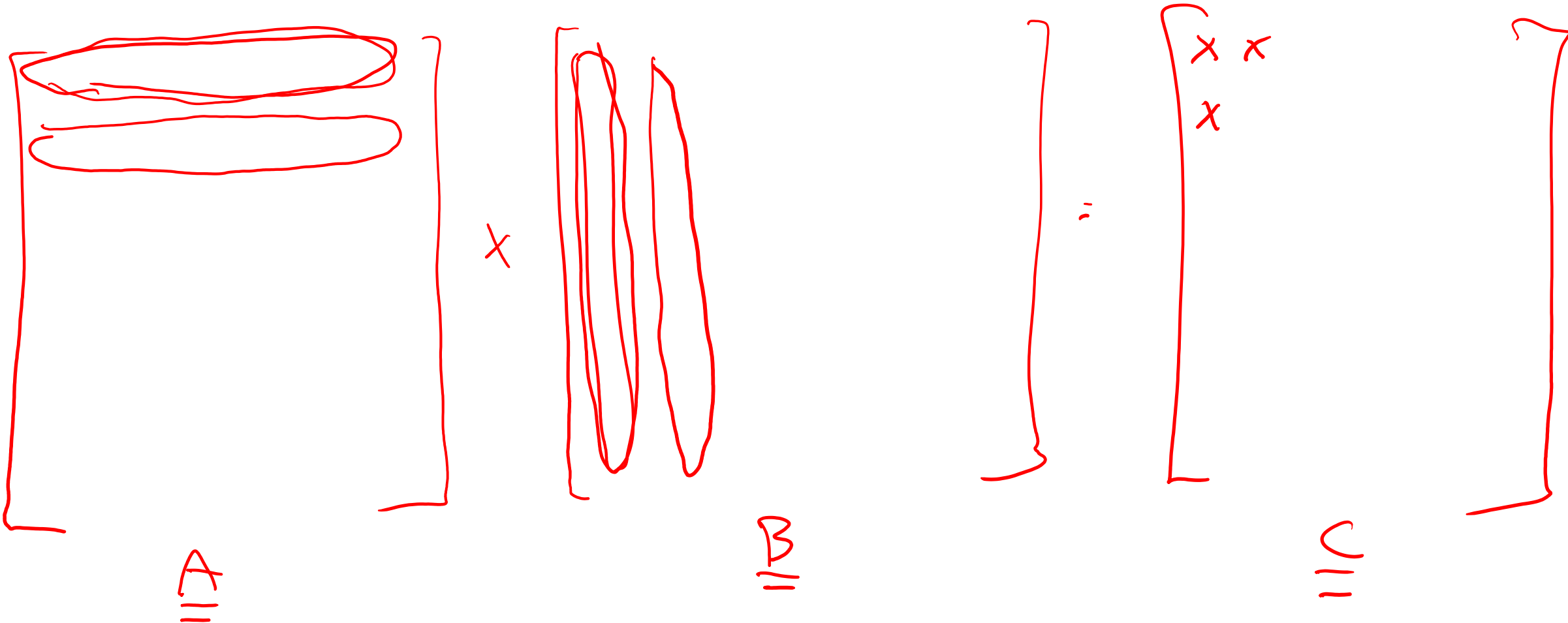
scalar-Matrix: 2 loops }

Matrix-Matrix: 3-loops } level 3

Matrix-Vector Multiplication



Matrix-Matrix Multiplication



Complexity and Cost of Solving Linear Systems

$O(n^3)$ for direct methods $\rightarrow \underline{n^{2.376}}$

Historically, what's a big N

1950's: $N=20$ - 6 hours

1965: $N=200$

1980: $N=2,000$

1995: $N=20,000$

2010: $N=200,000$
(4 GB)

10^4

MFLOPS

GFLOPS

TFLOPS

PFLOPS

EFLOPS

10^6 10^{12}

$$\frac{12}{4} = 3$$

$O(n^3) \rightarrow \# \text{ of operations}$
"classical view"

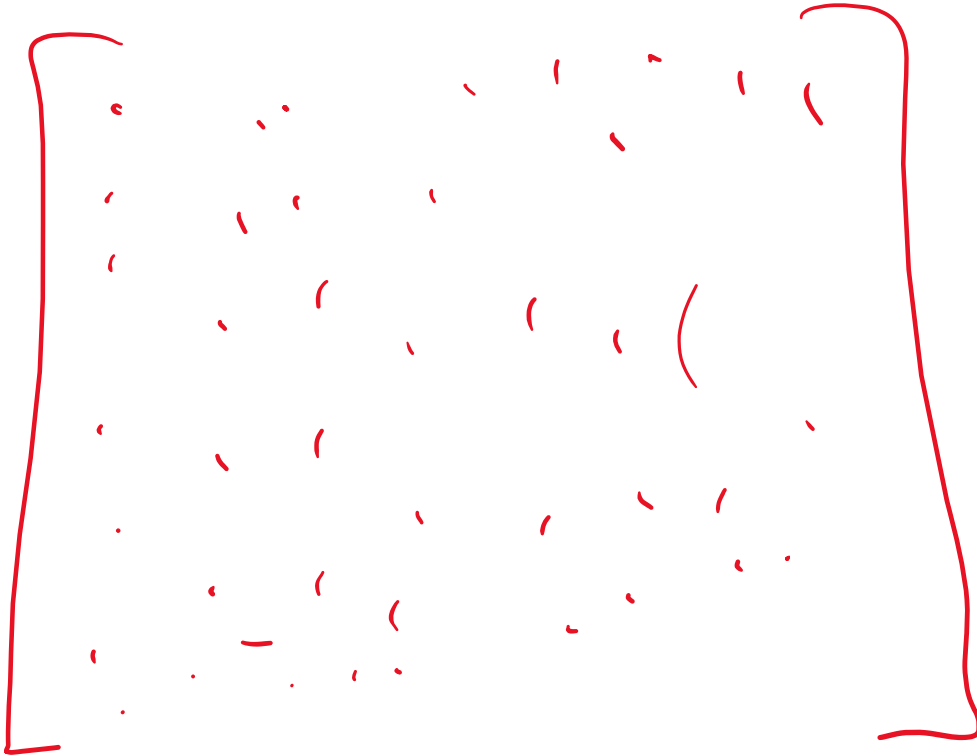
$\Rightarrow \# \text{ operations} + \underline{\underline{\text{data movement}}}$



Dense vs. Sparse Linear Algebra

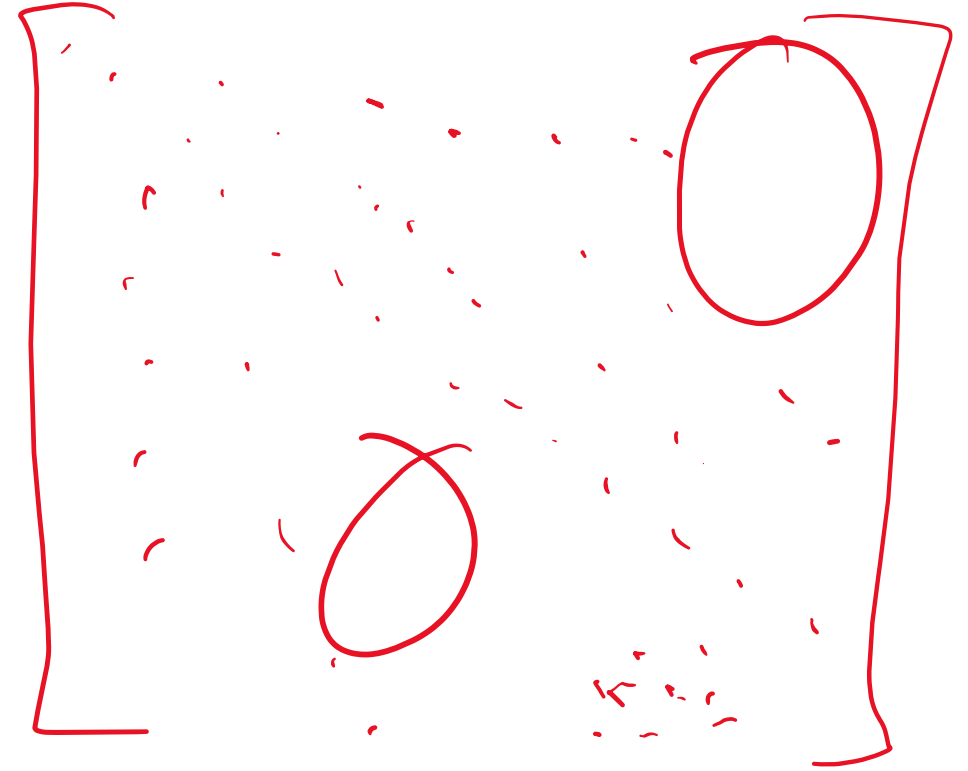
What are Dense and Sparse Linear Systems

Dense



Mostly non-zero

Sparse



fill in usually $< 20\%$

How do dense linear systems arise?

photos, pictures,

[

]

x

]

$x^T x$

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

Integral Methods

- the solution at any point is a function of the solution at all points

Boundary Integral Equations

$$\nabla^2 u(x) = 0 \quad \text{domain}$$

$$u(x) = f(x) \quad \text{boundary}$$

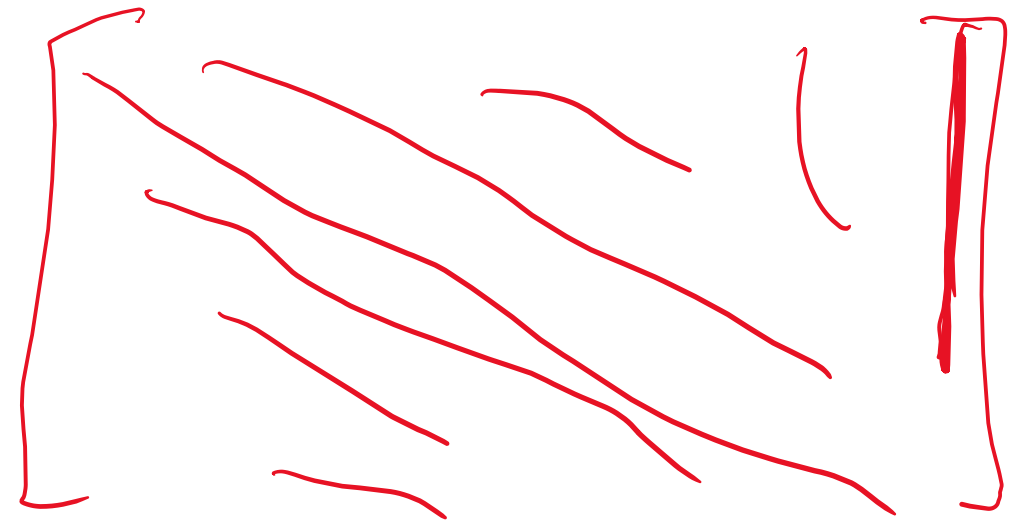
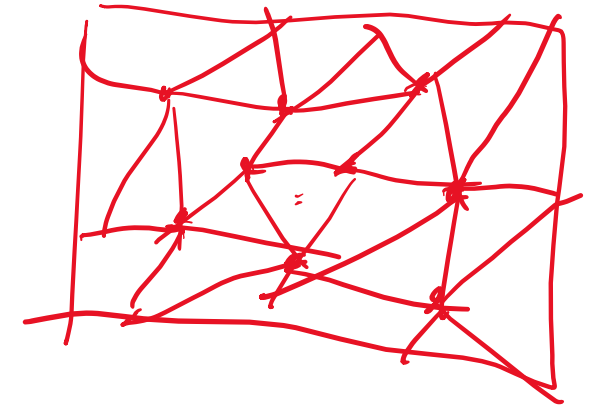
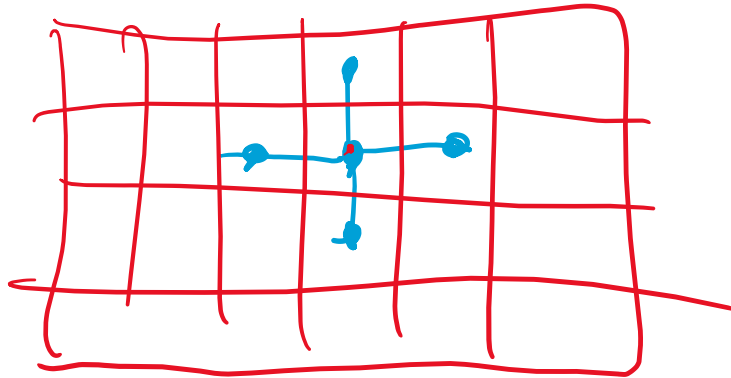
$$\rightarrow u(x) = \int_{\text{boundary}} G(x, s) f(s) ds$$

How do sparse linear systems arise?

Discretize PDE

- Finite Difference
- Finite Element
- Finite volume

$$\nabla^2 u(x) = 0 \quad \text{domain}$$
$$u(x) = f(x) \quad \text{boundary}$$



Types of Matrices and Storage Formats

Matrices

symmetric

orthogonal

Triangular / upper or lower

Banded

invertible

positive definite

singular

unitary orthonormal

Formats

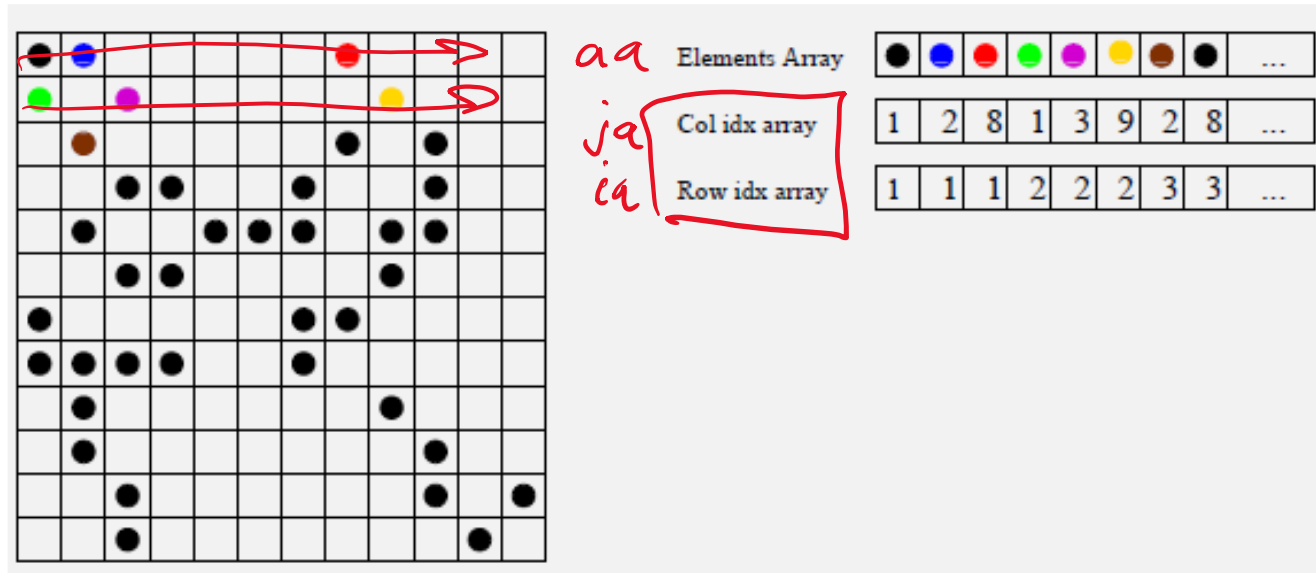
banded

symmetric

coordinate

Compressed sparse row

Example: Coordinate (COO) Format



} value
 } 2 integer
 arrays
 size nz
 nnz

```

do n=1, nnz
  i = ia(n)
  j = ja(n)
  y(i) = y(i) + aa(n) * x(j)
enddo
  
```



Solving Linear Systems



Overview of Solution Methods



LU Factorization



LU Factorization for Solving Linear Systems



LU: Forward Elimination



LU: Backward Substitution



Fast LU Factorizations