

Linear Equations

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

Examples of linear systems

Optimization Problems

* Quantum Mechanics (Hamiltonian systems) $\rightarrow \hat{H}\psi = E\psi$
 (Coupled) spring systems

* Classical Mechanics Systems (first-order) $\rightarrow \nabla^2 \psi = \rho$
 $\vec{F} = m\vec{a}$ \uparrow \uparrow p_m gravity
 p_e charge

Stochastic Processes $A^p \psi$

- Solution Methods
- (1) Gauss-Jordan Elimination
 - (2) LU decomposition
 - (3) Matrix Inversion
 - (4) SVD - Friday

(1) G.J. Elimination $A\vec{x} = \vec{b}$

$$\vec{A} = \begin{pmatrix} 10 & -7 & 1 \\ -3 & 2 & 6 \\ 5 & -1 & 5 \end{pmatrix} \quad \vec{b} = \begin{pmatrix} 8 \\ 4 \\ 6 \end{pmatrix}$$

(1) $(A|b)$ look at Augmented Matrix

(2) take linear combinations of rows to get upper-triangular form

(3) back-substituting to solve.

$$\left(\begin{array}{ccc|c} 10 & -7 & 1 & 8 \\ -3 & 2 & 6 & 4 \\ 5 & -1 & 5 & 6 \end{array} \right) \rightarrow \begin{array}{l} R_1 \\ R_2 \rightarrow (R_2 + \frac{3}{10}R_1)10 \\ R_3 \rightarrow (R_3 - \frac{5}{10}R_1)2 \end{array} \left(\begin{array}{ccc|c} 10 & -7 & 1 & 8 \\ 0 & -1 & 63 & 64 \\ 0 & 5 & 9 & 4 \end{array} \right)$$

* Maintain Precision:
Keep rows w/ large
coeffs ↓
log₁₀
X =

$$\begin{matrix} R_1 \\ R_2 \\ R_3 \rightarrow \frac{1}{384} (R_3 + 5R_2) \end{matrix} \left(\begin{array}{cc|c} 10 & -7 & 8 \\ 0 & -1 & 63 & 64 \\ 0 & 0 & 1 & 1 \end{array} \right)$$

$$R_2 \rightarrow -(R_2 - 63R_3) \left(\begin{array}{cc|c} 10 & -7 & 8 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 1 \end{array} \right)$$

$$R_1 \rightarrow \frac{1}{10} (R_1 + 7R_2 - R_3) \left(\begin{array}{cc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 1 \end{array} \right)$$

$$\hat{X} = \begin{pmatrix} 0 \\ -1 \\ 1 \end{pmatrix}$$

(2) L-U decomposition

$$A = L \cdot U$$

$$\left(\begin{array}{cc|c} 10 & -7 & 1 \\ -3 & 2 & 6 \\ 5 & -1 & 5 \end{array} \right) = \left(\begin{array}{ccc} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & l_{33} \end{array} \right) \left(\begin{array}{ccc} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{array} \right)$$

↳ diagonal components of $L = 1$
are a free choice (under-determined system)

$$= \left(\begin{array}{ccc} 1 & 0 & 0 \\ -\frac{3}{10} & 1 & 0 \\ \frac{5}{10} & l_{32} & 1 \end{array} \right) \left(\begin{array}{ccc} 10 & -7 & 1 \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{array} \right)$$

↓
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→ Numerical Recipes 2.3.1
has explicit sum

$$u_{22} = (2 - 21/10)$$

$$u_{23} = (6 + 3/10)$$

$$u_{33} = 2$$

$$\star \det(A) = \det(L \cdot U) = \det(L) \det(U) = \det(U) = \prod_i u_{ii}$$

$$\vec{y} \equiv U \cdot \vec{x}$$

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

$$= L \cdot U \vec{x} = \vec{b}$$

∴ $\vec{z} = \vec{b}$ → use forward sub. to find \vec{y}

$$= L \cdot U \vec{x} = \vec{b}$$

$$\underline{L \vec{y} = \vec{b}} \rightarrow \text{use forward sub. to find } \vec{y}$$

$$U \vec{x} = \vec{y} \rightarrow \text{Use back-sub. to find } \vec{x}$$

Inversion Similar to Gauss-Jordan Elimination

$$(1) (A|I)$$

(2) Perform row interchanges, linear combinations to get $A \rightarrow I, I \rightarrow A^{-1}$

Null spaces

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

• vectors with no eigenvalue ($= 0$)

• A not invertible

• Friday: SVD will help/handle

Eigenvalue finding

- Rayleigh Method ✓
- Iterative Methods
- If diagonalization is known ✓
- look @ polynomial roots ✓

$$A \vec{v}_{\lambda_1} = \lambda_1 \vec{v}_{\lambda_1}$$

Power Iteration

$$\vec{v} = c_1 \vec{v}_{\lambda_1} + c_2 \vec{v}_{\lambda_2} + \dots$$

$$\lim_{p \rightarrow \infty} A^p \vec{v} = c_1 \lambda_1^p \vec{v}_{\lambda_1} + c_2 \lambda_2^p \vec{v}_{\lambda_2} + \dots$$

Isolate largest λ

$$\lim_{p \rightarrow \infty} (A - \mu I)^{-p} \vec{v} = c_1 (\lambda_1 - \mu)^{-p} \vec{v}_{\lambda_1} + \dots$$

↑
converge linearly

$$\frac{1}{(\lambda_1 - \mu)^p} \text{ v.s. } \frac{1}{(\lambda_2 - \mu)^p}$$

isolates smallest λ - μ , eigenvalue closest to μ

Refinement: each time we multiply by $(A - \mu I)^{-1}$,

$$\rightarrow \underline{\mu} \rightarrow \frac{\vec{v}^T A \vec{v}}{|\vec{v}|^2} \rightarrow \lambda$$

Converge "Cubically"

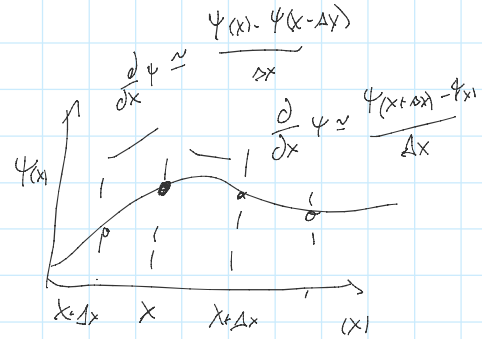
QM:

$$\begin{aligned} H\psi &= E\psi \\ \nabla^2\psi &= p \end{aligned}$$

$$\nabla^2 \psi \rightarrow \frac{\partial^2}{\partial x^2} \psi(x)$$

$$\frac{1}{\Delta x^2} \begin{pmatrix} & & & & \\ & 1 & -2 & 1 & \\ & & . & & \\ & & & 1 & \\ . & 1 & -2 & 1 & \\ o & o & 1 & -2 & 1 \\ o & o & o & 1 & -2 \\ & & & 1 & -2 \\ & & & & 1 \end{pmatrix} \cdot \begin{pmatrix} \psi(x-\Delta x) \\ \vdots \\ \psi(x) \\ \psi(x+\Delta x) \\ \vdots \end{pmatrix} = V(x)$$

↑



$$\frac{d^2}{dx^2} \psi = \frac{\psi(x+\Delta x) - 2\psi(x) + \psi(x-\Delta x)}{\Delta x^2}$$