

Linear Algebra Review

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Recall: Last class we wrapped up our discussion on Gaussian quadratures for a given interval $[a,b]$ and weight function $w(x)$. The main results for these quadratures are as follows:

- We construct or obtain the **3-term recursion** for the family of orthogonal polynomials $Q_n(x)$.
- The **nodes** $\{x_j\}_{j=0}^n$ for the $n+1$ point quadrature are the **unique $n+1$ zeroes of $Q_{n+1}(x)$** . There is a theorem that shows Q_{n+1} always has $n+1$ unique zeroes, and that all of them are in $[a,b]$.
- These zeroes can be obtained via **rootfinding (good first guess using interlacing + Newton)** or the **Golub-Welch algorithm**.
- The **weights w_j** can be obtained via the integrals of Lagrange polynomials or via the linear system that results from asking the rule be exact for polynomials of degree $\leq n$. However, the most efficient way to compute them is via formulas involving Q_n , Q_{n+1} and Q'_{n+1} (all which can be computed using the 3-term recursion and formulas obtained from it).
- **Weights are positive and small;** they add up to the integral of $w(x)$ from a to b . This makes Gaussian quadratures very stable.
- We can either use one interval and let n go to infinity OR do a composite quadrature for a given $n+1$ point rule. In the first case, order of accuracy goes faster to 0 than any power of $h = (b-a)/N$. In the second case, we get an $O(h^{2n+2})$ composite quadrature (the best for an $n+1$ pt quadrature).

We then introduced the concept of Adaptive Quadrature, what it is and how the algorithm to subdivide an interval adaptively works. We finished our discussion describing a recursive and non-recursive algorithm to perform Adaptive Quadrature.

Today, we start our section on numerical methods for linear algebra by going over a review of important linear algebra concepts.

LOOKING FORWARD

- ① Gauss Elimination - LU decomp
 - ② Gram Schmidt - QR decomp
 - ③ Eigenvalues, eigenvectors $\leq \begin{matrix} UDU^* \\ U\Sigma V^* \end{matrix}$
-

KEY CONCEPTS:

- Vector space / subspace - $(V, +, *)$
 ↳ Scalar field $\mathbb{F} \rightarrow \mathbb{R}/\mathbb{C}$.
- Linear combination $\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n$ $v_i \in V$
 $\alpha_i \in \mathbb{F}$
- Linear span $(\{v_1, \dots, v_n\})$
 ↳ Set of all possible (finite) linear comb.
- Linear dependence / Linear independence
 $\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n = 0$, α_j not all 0.

$$v_1 + v_2 + \dots + v_n = 0, \quad \text{if all } 0.$$

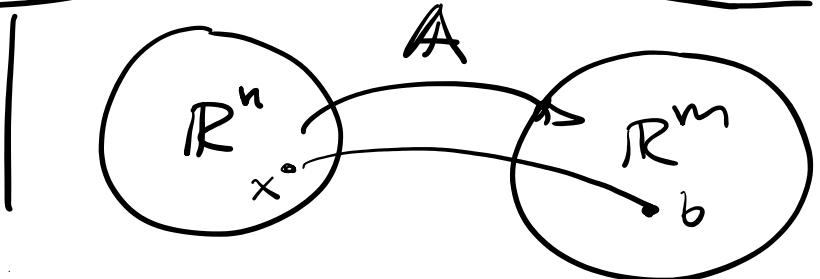
Can I write v_j as a LC of the rest?

• Basis & Dimension.

$B \subseteq V$ is a basis of V if $\text{span}(B) = V$ and B is LI set.

$$\rightarrow |B| = \dim V$$

$$\underset{m \times n}{A} \underset{n \times 1}{\vec{x}} = \underset{m \times 1}{\vec{b}}$$



1 $R(A)$, Range / Column space of A

$$R(A) = \{ \vec{b} \in \mathbb{R}^m \mid \text{There is } \vec{x} \text{ s.t. } A\vec{x} = \vec{b} \}$$

$$A\vec{x} = x_1 v_1 + x_2 v_2 + \dots + x_n v_n \quad A = [v_1 \ v_2 \ \dots \ v_n]$$

$$\dim(R(A)) = r(A) \text{ "rank of } A\text{".}$$

2 $N(A)$ / Nullspace / Kernel

$$N(A) = \{ \vec{x} \in \mathbb{R}^n \mid A\vec{x} = \vec{0} \}$$

$$A\vec{x} = x_1 v_1 + x_2 v_2 + \dots + x_n v_n = \vec{0}$$

Uniqueness: If $A\vec{x}_p = \vec{b}$, then

$$A(\vec{x}_p + \vec{w}) = \vec{b} \text{ for all } \vec{w} \in N(A).$$

$A(x_p + w) = b$ for all $w \in N(A)$.

$N(A) = \{\vec{0}\}$, then if $Ax=b$ has a sol, it is unique.

$N(A) \neq \{\vec{0}\}$, then if $Ax=b$ has a sol, it is not unique, and is

$$x_p + N(A)$$

$$\dim(N(A)) = v(A) \text{ "nullity of } A\text{"}$$

(0) Rank & Nullity thm: A $m \times n$,

$$r(A) + v(A) = n \text{ (# of cols)}$$

$Ax=b$, A $m \times n$ matrix.

A.

Case I: Underdetermined ($m < n$)



$$r(A) \leq s \rightarrow v(A) > 0 (> n-s)$$

Best case: $r(A)=m$, $v(A)=n-m$.

Case II: Overdetermined ($m > n$)

$$r(A) \leq n, v(A)=0 \text{ but } r(A) < m.$$



\rightarrow there are b for which $Ax=b$ has no solution.

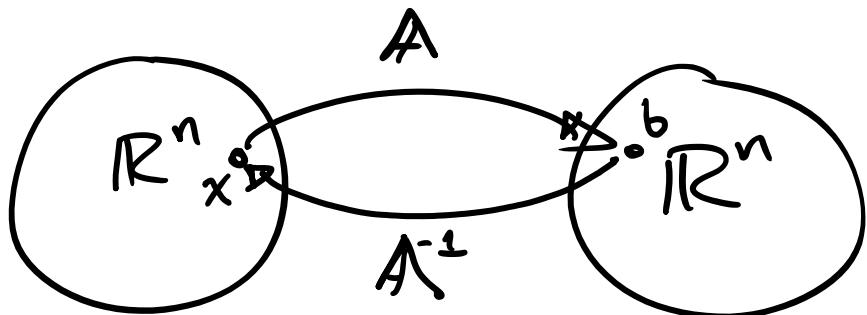
Case III: ($m=n$) (Determined)



Case III: ($m=n$) (Determined) □

↳ I can have $\nu(A)=0 \leftrightarrow r(A)=n$
Existence (for any b) and
Uniqueness?

$A \rightarrow$ invertible (non-singular)



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

Equivalent conditions:

- $Ax=b$ is uniquely solvable for every b .

• A is invertible

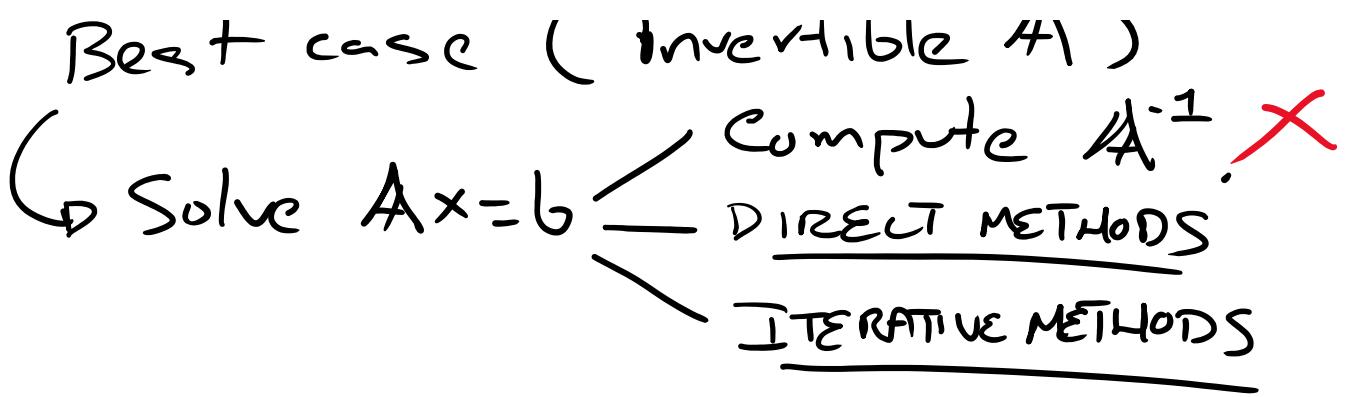
[• $r(A)=n$ (cols of A are LI)]

[• $\nu(A)=0$

• A has no zero eigenvalues

• $\det(A) \neq 0$.

Best case (invertible A)



DIRECT METHODS - Gauss Elim. (LU)

DO BEST WHEN

n is not too big

OR

I want to solve many $Ax = b_n$.

- Less dependent on cond #.

QR-based
UDU* / SVD.

Iterative Methods (CG, GMRES, Multigrid, Fixed pt method)

Producing seq of $x_n \xrightarrow{n \rightarrow \infty} x$.

Benefit: Cost per iter is low
(usually $A \cdot x$) Mat Vec

Cons: Cost $= (\# \text{Iters}) (\text{Cost per iter})$

A not invertible:

$$\|Ax - b\|^2$$

$$\rightarrow \min \|Ax - b\|^2$$

\rightarrow IF NOT
 $\rightarrow \underline{\min \text{ norm}}$

