

$Ax = b$: $\left. \begin{array}{l} \textcircled{1} \underbrace{A = LU}_{O(\frac{2}{3}n^3)} \quad \textcircled{2} \underbrace{L(\overset{z}{U}x) = b}_{O(2n^2)} \end{array} \right\} \text{Direct Method}$

For n "Large" $O(\frac{2}{3}n^3)$ is PROHIBITIVE!

Iterative Method { Guess $X^{(0)}$, and generate $X^{(k)}$

For $k = 1, 2, 3, \dots$

$\lim_{k \rightarrow \infty} X^{(k)} = X$

$\cdot X^{(1)} \cdot X^{(2)} \cdot X^{(3)} \cdot X^{(4)} \cdot X^{(5)}$

Error Analysis

$$Ax = b$$

x : exact soln $x = A^{-1}b$

$\tilde{x}$: approximate $\tilde{x}$

e : $x - \tilde{x}$ error

r : $\underbrace{b - A\tilde{x}}_{\text{computable}}$ residual

$$(\lambda, x) = \left(\frac{1}{b}, e\right) \quad Ae = \frac{1}{b}e = r$$

Note

$$\textcircled{1} \quad Ae = r$$

$$\begin{aligned} \text{pf: } Ae &= A(x - \tilde{x}) \\ &= Ax - A\tilde{x} \\ &= b - A\tilde{x} = r \end{aligned}$$

$\textcircled{2} \quad e = 0 \text{ iff } r = 0$
which follows from (1) since
 A is invertible:

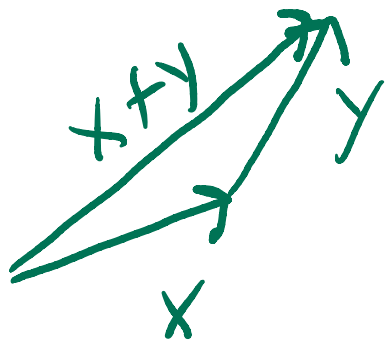
$$Ae = r, \quad e = A^{-1}r$$

Big Problem: Even if r is "small" e may be "large"

Main Tools Vector Norm and Matrix Norms

Defn: $x \in \mathbb{R}^n$, a vector norm $\|x\|$ satisfies

- ① $\|x\| \geq 0$ and $\|x\| = 0$ iff $x = 0$
- ② $\|\alpha x\| = |\alpha| \|x\|$ for $\alpha \in \mathbb{R}$
- ③ $\|x + y\| \leq \|x\| + \|y\|$ $y \in \mathbb{R}^n$



Triangle Inequality

Some Common Norms $x \in \mathbb{R}^n$

p-norm: $\|x\|_p = \left(\sum_{i=1}^n |x_i|^p \right)^{1/p} \quad 1 \leq p < \infty$

p=1: $x = [x_i]$ $\|x\|_p = \left(\sum_{i=1}^n |x_i|^p \right)^{1/p} = (|x|^p)^{1/p} = |x|$ Absolute value

p=1: 1-norm $\|x\|_1 = \sum_{i=1}^n |x_i|$

p=2: 2-norm $\|x\|_2 = \left(\sum_{i=1}^n |x_i|^2 \right)^{1/2} = \sqrt{x^T x}$
Euclidean Norm

p $\rightarrow\infty$: ∞ -norm $\|x\|_\infty = \max_{1 \leq i \leq n} |x_i|$

Ex! $X = [1 \ 2 \ 3]^T$

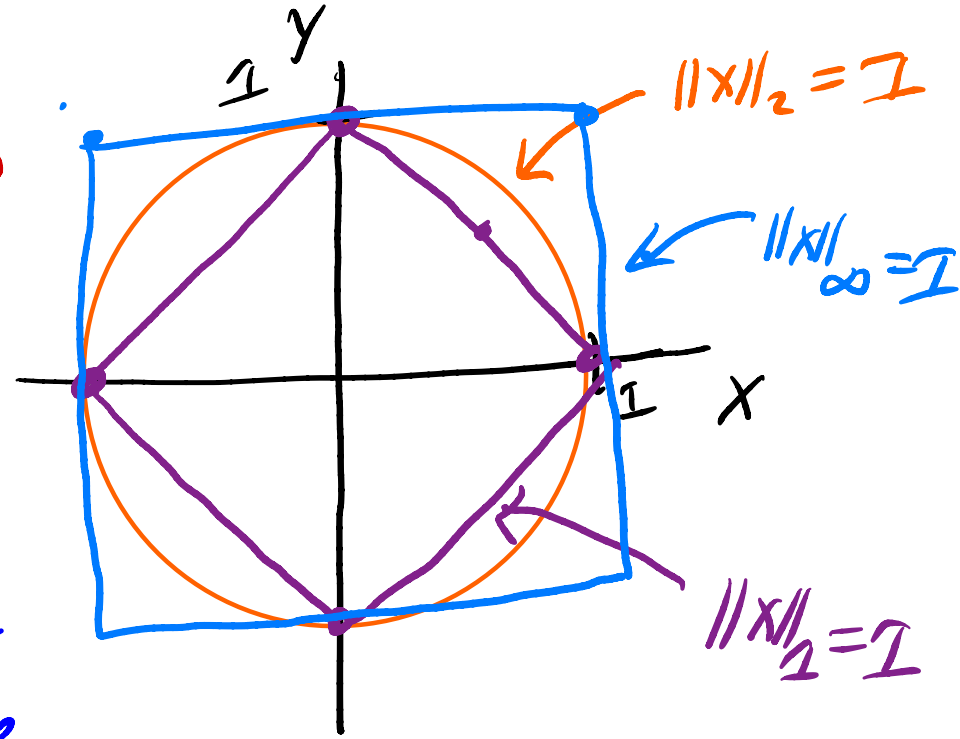
$$\|X\|_{\infty} = \max\{|1|, |2|, |3|\} = 3$$

$$\|X\|_1 = |1| + |2| + |3| = 6$$

$$\|X\|_2 = (|1|^2 + |2|^2 + |3|^2)^{1/2} = \sqrt{14}$$

Unit Ball: Set of all x such that $\|x\|=1$

$n=2$: Geometrically comparing norms.



Fact: All norms on $\mathbb{R}^n$ are EQUIVALENT in the sense that if a sequence $\{x^{(k)}\}_{k=0}^{\infty}$ converges to x in one norm, i.e. $\lim_{k \rightarrow \infty} \|x - x^{(k)}\| = 0$, then

$$\lim_{k \rightarrow \infty} |||x - x^{(k)}||| = 0. \quad (\text{Math 545})$$

Matrix Norms

Defn: Given $A, B \in \mathbb{R}^{n \times n}$, a MATRIX norm satisfies

- ① $\|A\| \geq 0$ and $\|A\| = 0$ iff $A = 0$
- ② $\|\alpha A\| = |\alpha| \|A\|$ for $\alpha \in \mathbb{R}$
- ③ $\|A + B\| \leq \|A\| + \|B\|$
- ④ $\|AB\| \leq \|A\| \|B\|$

Where can we get a matrix norm?

ans: A matrix norm INDUCED

by a vector norm.

Defn: Given a vector norm $\|\cdot\|$, the subordinate or INDUCED matrix norm is

$$\|A\| \stackrel{\text{defn}}{=} \max_{x \neq 0} \frac{\|Ax\|}{\|x\|} \quad \left. \begin{array}{l} \text{vector norm} \\ \text{vector norm} \end{array} \right\}$$

$$= \max_{\substack{\|x\|=1 \\ \text{Closed and Bounded Set}}} \|Ax\| \quad \text{Continuous}$$

And we also have

$$\textcircled{5} \quad \|Ax\| \leq \|A\| \|x\|$$