Numerical Linear Algebra: An Introduction - Notes

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0.2 Error, Stability and Conditioning

Definition 0.2.1: Floating point number

B-adic, normalized floating point number of precision m is either $x = 0$ or:

$$x = B^e \sum_{k=-m}^{-1} x_k B^{-k},\ x_{-1} \neq 0,\ x_k \in \{0, 1, \ldots, B-1\}$$

Where:

- $B \geq 2$ is the base of the number system.
- $e_{min} \leq e \leq e_{max}$ is the exponent.
- $\sum_{k=-m}^{-1} x_k B^{-k}$ is the mantissa.

Many programming languages use the IEEE 754 standard for floating point arithmetic. In this standard, for a double precision number, the base is $B = 2$, the mantissa has $m = 52$ bits and the exponent has 11 bits.

Two additional numbers are added to the set of floating point numbers: $\pm\infty$ and NaN (Not a Number) which is used to represent undefined or unrepresentable values.

Definition 0.2.2: Machine epsilon

he machine epsilon eps is the smallest positive number which satisfies:

$$|x - rd(x)| \leq eps|x|$$

Where $rd(x)$ is the floating point representation of x.

Usually this rounding function is taken to be the nearest machine number. A B-system with precision m the associated the machine epsilon is given by:

Theorem 0.2.1: Floating point machine epsilon

or a floating point number system with base B and precision m, the machine epsilon is given by: The machine epsilon is given by $eps = B^{1-m}$, i.e we have:

$$|x - rd(x)| \leq B^{1-m}|x|$$

Theorem 0.2.2: Floating point ope

et $\star$ be one of the operations $+, -, \times, /$ and let $\circledast$ be the equivalent floating point operation, then $\forall x, y$ in the floating point system, there exists an ϵ such that:

$$x \star y = (x \circledast y)(1 + \epsilon).$$

Definition 0.2.3: Compatibility of matrix norms

iven the norms $\|\cdot\|_{(n)}$ and $\|\cdot\|_{(m)}$ on $\mathbb{R}^n$ and $\mathbb{R}^m$ respectively, we say that a matrix norm $\|\cdot\|_\star$ is **compatible** with these norms if:

$$\|Ax\|_{(m)} \leq \|A\|_\star \|x\|_{(n)}, \quad \forall x \in \mathbb{R}^n.$$

Matrix-vector multiplication arises naturally when solving linear systems of equations. Given a matrix $\mathbf{A} \in \mathbb{R}^{m \times n}$, $x, b \in \mathbb{R}^n$. Let us consider the situation

$$\begin{aligned} b &= \mathbf{A}x, \\ \delta b &= \mathbf{A}\delta x \end{aligned}$$

If the matix $\mathbf{A} : \mathbb{R}^n \to \mathbb{R}^n$ is non-singular, then $\|x\|_\star := \|\mathbf{A}x\|$ is a norm on $\mathbb{R}^n$. From the fact that all norms in $\mathbb{R}^n$ are equivalent follows that there exits two constants C_1 and C_2 such that:

$$C_1\|x\| \leq \|\mathbf{A}x\| \leq C_2\|x\| \implies \frac{\|\delta b\|}{\|b\|} \leq \frac{C_2}{C_1}\frac{\|\delta x\|}{\|x\|}.$$