Chapter 2: Number System and Errors

2.1 Floating Point Arithmetic

What can go wrong:

- modeling errors
- discretization and truncation errors
- human error
- roundoff and data errors

Computers represent numbers using *floating point number* systems, $\mathbf{F}(\beta, k, m, M)$, characterized by

- β : the base
- k: the number of digits in the base β expansion
- m: the minimum exponent
- *M*: the maximum exponent

$$\mathbf{F}(\beta, k, m, M) := \{ \pm (0.b_1 b_2 ... b_k)_{\beta} \times \beta^e \text{ with } m \le e \le M \}$$

Terminology:

- $b_1b_2...b_k$ is called the **mantissa**
- e is called the exponent
- If $b_1 \neq 0$, or else $b_1 = b_2 = ... = b_k = 0$, $\mathbf{F}(\beta, k, m, M)$ is said to be **normalized**

2.1 Floating Point Arithmetic

Example:

$$\begin{aligned} \textbf{F}(10,1,0,1) &= & \{\pm (0.b_1)_{10} \times 10^e \text{ with } 0 \leq e \leq 1\} \\ &= & \{0,\pm 0.1,\pm 0.2,...,\pm 0.9,\pm 1,\pm 2,...,\pm 9\} \end{aligned}$$

Properties:

- The number of elements of (normalized) $\mathbf{F}(\beta, k, m, M)$ is $1 + 2(\beta 1)\beta^{k-1}(M m + 1)$.
- The largest positive number of (normalized) $\mathbf{F}(\beta, k, m, M)$ is $(0.\beta 1\beta 1...\beta 1)_{\beta} \times \beta^{M} = (1 \beta^{-k})\beta^{M}$.
- The smallest positive number of (normalized) $\mathbf{F}(\beta, k, m, M)$ is $(0.10...0)_{\beta} \times \beta^m = \beta^{m-1}$.

A number that has a magnitude outside the above computer range is called an **underflow** or an **overflow**.

2.1 Floating Point Arithmetic

Most computers use the binary system ($\beta=2$). The two binary digits 0 and 1 are usually called **bits**, and the fixed-length group of binary bits is called a **computer word**.

Example: the floating-point number system of a 32-bit word length microcomputer. The internal representation of a word is as following:

- the leftmost bit is used for the sign of the number (0 \rightarrow + and 1 \rightarrow -)
- the next seven bits represent the exponent, with the first bit used for its sign
- the final 24 bits represent the normalized mantissa

2.2 Roundoff Errors

 $y = \pm (0.b_1b_2...b_kb_{k+1}...)_{\beta} \times \beta^e$ with $b_1 \neq 0$ and $m \leq e \leq M$. Denote by $fl(y) \in \mathbf{F}(\beta, k, m, M)$ the floating point equivalent of y. There are two natural ways to define fl(y):

- chopping the number, i.e. $fl_{chop}(y) = \pm (0.b_1b_2...b_k)_{\beta} \times \beta^e$;
- rounding the number, i.e.

$$\mathit{fl}_{\textit{round}}(y) = \left\{ \begin{array}{ll} \pm (0.b_1 b_2 ... b_k)_{\beta} \times \beta^e & \text{if } b_{k+1} < \beta/2 \\ \pm [(0.b_1 b_2 ... b_k)_{\beta} + \beta^{-k}] \times \beta^e & \text{if } b_{k+1} \ge \beta/2. \end{array} \right.$$

Definition. The error introduced by converting a real number y to its floating point equivalent fl(y) is called *roundoff error*.

Absolute roundoff error:

$$|fl_{chop}(y)-y| \leq \beta^{e-k}, \quad |fl_{round}(y)-y| \leq \frac{\beta^{e-k}}{2}.$$

Relative roundoff error:

$$\frac{|\mathit{fl}_{chop}(y)-y|}{|y|} \leq \beta^{1-k}, \quad \frac{|\mathit{fl}_{round}(y)-y|}{|y|} \leq \frac{\beta^{1-k}}{2}.$$



2.2 Roundoff Errors

Definition. The *machine precision* is given by

$$u = \begin{cases} \beta^{1-k}, & \text{chopping} \\ \frac{1}{2}\beta^{1-k}, & \text{rounding}, \end{cases}$$

where β is the base and k is the number of digits in the implemented floating point number system.

Computers perform calculations within their f.p.n.s.:

$$x@_{fl}y = fl(fl(x)@fl(y)),$$

where @ represents a binary arithmetic operators (e.g., +, -, x, /). Floating point arithmetic does not satisfy many of the properties of real arithmetic, such as (addition) associativity and distributivity.

Floating Point Calculations

Example: In 4 decimal digit rounding arithmetic:

$$(0.1329+1.543)+23.21=1.676+23.21=24.89$$
 but $0.1329+(1.543+23.21)=0.1329+24.75=24.88$ $(0.1351+23.21)\times 1.543=23.35\times 1.543=36.03$ but $0.1351\times 1.543+23.21\times 1.543=0.2085+35.81=36.02$

Accumulation of Roundoff Errors

$$x@_{fl}y - x@y = fl(fl(x)@fl(y)) - x@y$$

= $[fl(fl(x)@fl(y)) - fl(x)@fl(y)] + [fl(x)@fl(y) - x@y]$
= introduced error + propagated error

The introduced error is small; it is bounded by machine precision. Unfortunately the propagated error can be large.

Floating Point Number Systems: The IEEE Standard

A widely used internal representation of numbers in almost all new computers is the **IEEE Standard**.

The single-precision format

Floating-point number =
$$(-1)^{s} x (1.f)_{2} x (2^{c-127})_{10}$$

uses 32 bits:

- first bit is reserved for the sign bit s ($s = 0 \rightarrow +, s = 1 \rightarrow -$);
- next eigth bits are reserved for the (biased) exponent c;
- the remaining 23 bits are used for the fractional part *f* of the normalized mantissa.
- The double-precision format

Floating-point number =
$$(-1)^s x (1.f)_2 x (2^{c-1023})_{10}$$

uses 64 bits:

- first bit is reserved for the sign bit s ($s = 0 \rightarrow +, s = 1 \rightarrow -$);
- next 11 bits are reserved for the (biased) exponent c;
- the remaining 52 bits are used for the fractional part f of the normalized mantissa.

2.3 Truncation Error

Round-off errors arise in considering the floating point equivalent of numbers. In contrast, the **truncation error** terminates a process, usually related to considering only a finite number of terms of infinite series or sequences. An important tool is the Taylor series expansion of f(x) about a point x_0 :

$$f(x) = \sum_{k=0}^{n} \frac{f^{(k)}(x_0)}{k!} (x - x_0)^k + \frac{f^{(n+1)}(\xi)}{(n+1)!} (x - x_0)^{n+1},$$

for some $\xi = \xi(x)$ between x_0 and x.

Examples:

$$e^{x} = 1 + x + \frac{x^{2}}{2!} + \dots + \frac{x^{n}}{n!} + \frac{e^{\xi}}{(n+1)!} x^{n+1}$$

$$\sin x = x - \frac{x^{3}}{3!} + \frac{x^{5}}{5!} - \frac{x^{7}}{7!} + \dots + \frac{\sin^{(n+1)}(\xi)}{(n+1)!} x^{n+1}$$

2.4 Interval Arithmetic

Definition. Let \star be one of the symbols $\{+, -, \cdot, \div\}$. If A and B are intervals, we define arithmetic operations on intervals by

$$A \star B = \{x \star y | x \in A, y \in B\}$$

except that we do not define $A \div B$ if $0 \in B$.

If
$$A = [a_1, a_2]$$
 and $B = [b_1, b_2]$, then $A + B = [a_1 + b_1, a_2 + b_2]$ $A - B = [a_1 - b_2, a_2 - b_1]$ $A \cdot B = [\min{\{a_1b_1, a_1b_2, a_2b_1, a_2b_2\}}, \max{\{a_1b_1, a_1b_2, a_2b_1, a_2b_2\}}]$ $A \div B = [a_1, a_2] \cdot [1/b_2, 1/b_1]$ provided that $0 \not\in [b_1, b_2]$.