Math 381 - Fall 2020

Jay Newby

University of Alberta

Week 2

Last Week

- Course overview
- eClass
- Python
- 4 Jupyter
- 6 Programming
- 6 Hardware
- Numerical errors

This Week

- Representing numbers on a computer
- 2 Floating point overflow and underflow
- Roundoff error
- 4 Examples

Review from last time

Suppose we have the exact solution to a problem $x \neq 0$, and an approximate solution $\hat{x} \in \mathbb{R}$.

Absolute Error:

$$\mathcal{E}_{\text{abs}} = |\hat{x} - x|$$

Relative Error:

$$\mathcal{E}_{\mathrm{rel}} = \frac{|\hat{x} - x|}{|x|}$$

Representing integers

Indeed of leaving a migue symbol of every integer, we learn 10 by mbols

0,1,2,3,4,5,6,7,8,9

Polynomial representation of integers

Let $r_j, j=1,\ldots,N$ be a sequence of base β integers. An integer $n\in\mathbb{N}$ can be represented as

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$$n = \pm \sum_{i=0}^{N} r_{i} \beta^{j} \qquad \left\{ \Gamma_{0}, \Gamma_{1}, \Gamma_{2}, \cdots \right\}$$

How many symbols does a computer know?

Binary Example

Let $\beta = 2$ then

$$n=\pm\sum_{j=0}^N r_j 2^j$$

How about n = 1074?

$$r_{10-j} = \{10000110010\}$$

Representing Real numbers

How do we represent real numbers with pencil and paper?

Representing Real numbers

Any real number $x \in \mathbb{R}$ can be represented by three components: a sign \pm , an infinite sequence of base β digits d_j , $j=0,1,\ldots$ called the *mantissa*, and an exponent β^e for integer e. Thus, we can write

$$x = \pm \left(\sum_{j=0}^{\infty} \frac{d_j}{\beta^j}\right) \times \beta^e,$$

where

$$d_j \in \{0, 1, \dots \beta - 1\}.$$

Example for base 10
If $d_t \in \{0, 1, \dots 9\}$, (base 10) then

$$x = \pm \left(\frac{d_0}{1} + \frac{d_1}{10} + \frac{d_2}{10^2} + \dots + \frac{d_t}{10^t} + \dots\right) \times 10^e,$$

where e is an integer exponent.

Example for binary

If $d_t \in \{0,1\}$, (base 2) then $d_b > \delta$ by convention

$$x = \pm \left(1 + \frac{d_1}{2} + \frac{d_2}{2^2} + \dots + \frac{d_t}{2^t} + \dots\right) \times 2^e,$$

where e is an integer exponent.

Floating Point Number

A computer can store a finite number of bits ($\{0,1\}$ values) to represent a single floating point number

$$fl(x) = sign(x) \times (1.\tilde{d}_1 \tilde{d}_2 \tilde{d}_3 \cdots \tilde{d}_{t-1} \tilde{d}_t) \times 2^e,$$

where e is an integer exponent and $\tilde{d}_n \in \{0,1\}$.

Single and double precision

• Double Precision (Python default, AKA 'float64') 64bits total, with 1bit for the sign, t=52 for the matissa, and 11bits to store the exponent e. The exponent is bounded by $L \le e \le U$, where L=-1022 and U=1023.

• Single Precision (AKA 'float32') 32bits total, with 1bit for the sign, t=23 for the matissa, and 8bits to store the exponent e (where L=-126 and U=127).

Floating Point Number: roundoff error

For IEEE standard

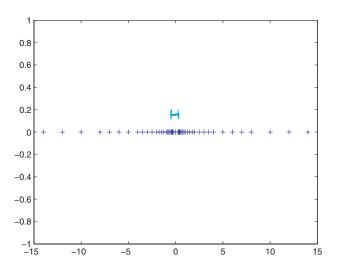
$$\mathcal{E}_{\text{rel}} = \frac{|fl(x) - x|}{|x|} \le 2^{-t-1}$$

$$\text{machine epsilon} : \mathcal{E} \quad \text{(book uses m)}$$

$$\text{floatly: } \mathcal{E}_{\text{N}} | 0^{-8}$$

$$\text{floatly: } \mathcal{E}_{\text{N}} | 0^{-16}$$

Distribution of floating point numbers on the real line



Smallest and largest possible (absolute) value of float32

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Special
$$b=0 \Rightarrow e=0-127=-127 \leftarrow Zero$$
 *Subnormal we want study

Numbers $b=255 \Rightarrow e=255-127=128 \leftarrow NaN \Rightarrow inf$

"not a number"

Recall $L \le e \le U$, L = -126, U = 127Let e = b - 127, where b is a non negative 8-bit integer (uint8)

Largest absolute value: $x_{\text{largest}}^{(32)} = (1.\tilde{d}_1\tilde{d}_2\tilde{d}_3\cdots\tilde{d}_{t-1}\tilde{d}_t) \times 2^{127} \nsim |D^{*}$

Smalest absolute value: $x_{\text{smallest}}^{(32)} = 2^{-126} \approx 10^{-3}$

Smallest and largest possible (absolute) value of float64

Recall $L \le e \le U$, L = -1022, U = 1023Let e = b - 1023, where b is a non negative 8-bit integer (uint8)

Largest absolute value: $x_{\mathrm{largest}}^{(64)} \approx 2 \cdot 2^{1023}$

Smalest absolute value: $x_{\text{smallest}}^{(64)} = 2^{-1022}$