Data Representations

- Computers store data as different variable types, e.g. integer, floating point, complex, etc.
- Different machines have different wordlengths,
 e.g. 4-byte ints on a 32-bit machine (Pentium),
 8-byte ints on a 64-bit machine (Alpha).
- This makes (binary) data non-portable.

Integers

- All data types represented by 0's and 1's.
- An integer value:

$$j = \sum_{i=1}^{N} s_i \times 2^{N-i}$$

- N = # of bits in word
- s_i = value of bit *i* in binary string *s*
- e.g. $0\ 0\ 0\ 0\ 1\ 1\ 0 = 2^2 + 2^1 = 6$ for 8-bit word.
- Use "two's complement" method for sign.

Integers, Cont'd

- Largest value that can be represented is $2^N 1$.
- For 32-bit word this is 4,294,967,295.
- Arithmetic with integers is exact, except:
 - When division results in remainder.
 - Result exceeds largest representable integer e.g. $2 \times 10^9 + 3 \times 10^9$ = overflow error
- Note multiplication by 2's can be achieved by left-shift, which is very fast (in C: "<<" operator).

Two's Complement

- Exploits finite size of data representations (cyclic groups) and properties of binary arithmetic.
- To get negative of binary number, invert all bits and add 1 to the result.
 - e.g. 1 = 0 0 0 0 0 0 0 1 in 8-bit

invert bits: 11111110

add 1: 00000001

result: $1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ = -1$

• In 8 bits, signed char ranges from -128 to +127.

Negative Powers of 2

 Binary notation can be extended to cover negative powers of 2, e.g. "110.101" is:

 $1 \times 2^{2} + 1 \times 2^{1} + 1 \times 2^{-1} + 1 \times 2^{-3} = 6.625$

- Can represent real numbers by specifying some location in the word as the "binary point" → fixed-point representation.
- In practice, use some bits for an exponent → floating-point representation.

Floats

 For most machines these days, real numbers are represented by floating-point format:

 $x = s \times M \times B^{e-E}$

- s = sign B = base (usually 2, sometimes 16)
- M = mantissa e = exponent

E = bias, usually 127.

• In past, manufacturers used different number of bits for each of *M* and $e \rightarrow$ non-portable code.

Floats, Cont'd

- Currently, most manufacturers adopt IEEE standard:
 - $-s = 1^{st}$ bit
 - Next 8 bits are e
 - Last 23 bits are M, expressed as a binary fraction, either 1.F, or, if e=0, 0.F, where F is in base 2.
- Largest single-precision float $f_{max} = 2^{127} \approx 10^{38}$.
- Smallest (and least precise!)

$$f_{\min} = 2^{-149} \approx 10^{-45}.$$

Round-off Error

- Not all values along real axis can be represented.
- There are 10^{38} integers between $f_{min} \& f_{max}$, but only $2^{32} \approx 10^9$ bit patterns.



- Values $< |10^{-45}|$ result in "underflow" error.
- If value cannot be represented, next nearest value is produced. Difference between desired and actual value is called "round-off error" (RE).

Round-off Error, Cont'd

- Smallest value e_m for which $1 + e_m > 1$ is called "machine accuracy", typically ~10⁻⁷ for 32 bits.
- Double precision greatly reduces $e_m (\sim 10^{-16})$.
- RE accumulates in a calculation:
 - Random walk: total error $N^{1/2} e_m$ after N operations.
 - But algorithms rarely random \rightarrow linear error N e_m.

Round-off Error, Cont'd

 Subtraction of two very nearly equal numbers can give rise to large RE.

e.g. Solution of quadratic equation...

$$x = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

...can go badly wrong whenever $ac \ll b^2$ (Cf. PS#2).

 RE cannot be avoided—it is a consequence of using a finite number of bits to represent real values.

Truncation Error

- In practice, most numerical algorithms approximate desired solution with a finite number of artithmetic operations.
 - e.g. evaluating integral by quadrature summing series using finite number of terms
- Difference between true solution and numerical approximation to solution is called "truncation error" (TE).

Truncation Error, Cont'd

- TE exists even on "perfect" machine with no RE.
- TE is under programmer's control; much effort goes into reducing it.
- Usually RE and TE do not interact.
- Sometimes TE can amplify RE until it swamps calculation. Solution is then called <u>unstable</u>.

e.g. Integer powers of Golden Mean (Cf. PS#2).