CSC 165

floating-point

week 11, lecture 2

Danny Heap

heap@cs.toronto.edu

www.cdf.toronto.edu/~heap/165/F09

resources: chapter 7 of course notes

http://docs.python.org/tutorial/floatingpoint.html

http://en.wikipedia.org/wiki/IEEE\_754-2008

Due to peer pressure, we will celebrate Nov.

30th next Monday. — E4 due next Monday.

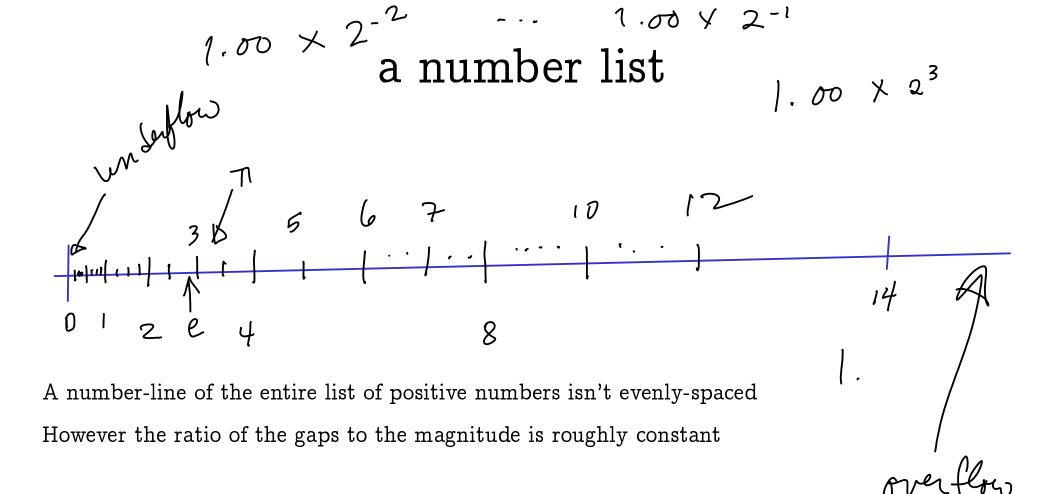
### recall: number representation

If you fix the cost of arithmetic operations, you fix the size of numbers Each number is given the same space (usually bits)

Result: floating numbers are represented in scientific notation using some base  $\beta$ , a fixed number of digits, t, a certain range of exponents  $e \in [e_{\min}, e_{\max}]$ , and some way to store the sign.

Suppose your base  $\beta=2$ , you allow a bit for the sign, you have room for t=3 digits, and your exponents are from [-2,3]. Normalize the representation of non-zero numbers so there is one non-zero digit to the left of the radix point

The smallest positive number you can represent in this system?  $1.00 \times 2^{-2} = \frac{1}{4}$ The largest positive number you can represent in this system?  $1.11 \times 2^3 = 14$ 



How many positive numbers are there in total?

## rounding

Since we're on a budget for digits and exponents, we can't represent infinitely many numbers. Suppose our base  $\beta=10$ , we have t=3 digits, and exponents [-3,-2,-1,0,1,2]. How should we represent  $e\approx 2.7182818284590451$  or  $\pi\approx 3.1415926535897931?$ 

truncate or round-to-nearest

This leads to three varieties of error (all with a sewage analogy):

- overflow: There is no way to represent numbers larger than  $9.99 \times 10^2$ , so 999.5 is a problem.
- underflow: There is no way to represent positive numbers smaller than  $1.00 \times 10^{-3}$ , so  $\frac{1}{1001}$  is a problem.
- rounding error: (throughflow?) There is no way to represent numbers strictly between adjacent numbers we can represent, so 1.001 is a problem.

## absolute, relative

In our base-ten number system of the previous page, using round-to-nearest, the absolute error representing e is |2.72 - 2.7182818284590451...|

We care about relative error. A millimeter error is a disaster in eye surgery but pretty acceptable in transcontinental air travel.

Compare the error to the quantity being sought. So, if  $x \neq 0$ , the relative error is  $\frac{|x-x'|}{|x|}$ 

 $\int e = \frac{0.1}{1.0}$ 

Compare the relative error when x = 1.0 and x' = 1.1? How about about when x = 100.0 and x' = 100.1?

# bounding round-to-nearest

If you had infinitely many digits and base  $\beta$ , you could exactly represent a number  $d_0.d_1d_2...d_{t-1}... \times \beta^e$ 



But, if you're limited to t digits, you have so round up or down:

$$\underbrace{d_0.d_1d_2\dots d_{t-1}}_{} imeseta^e ext{ or } d_0.d_1d_2\dots \left(d_{t-1}+1
ight) imeseta^e$$

What's the maximum difference between 
$$x$$
 and  $x'$ ? (take into account round-to-nearest)

Use the fact that  $1.0...0 \times \beta^e$  is the smallest possible denominator

$$1.00 - \times \beta^e = \beta^e$$

t-1 steps to right of radix

# big relative error!

By tweaking the numbers, you can make the error as bad as you like... billions of percent error

The internal (binary) representation of floats is obscured by the decimal display. Try shipwright.binFloat()