

# CS 137 Week 3

Floating Numbers, Math Library, Polynomials and Root Finding

September 25th, 2017

# Floating Point Numbers

- How do we store decimal numbers in a computer?
- In scientific notation, we can represent numbers say by

$$-2.61202 \cdot 10^{30}$$

where  $-2.61202$  is called the precision and  $30$  is called the range.

- On a computer, we can do a similar thing to help store decimal numbers.

# Data Types

Type	Size	Precision	Exponent
float	4 bytes	7 digits	$\pm 38$
double	8 bytes	16 digits	$\pm 308$

Note: You will almost always use the type double

# Conversion Specifications

There are many different ways we can display these numbers using the `printf` command. They in general have the format `% $\pm$  m.pX` where

- $\pm$  is the right or left justification of the number depending on if the sign is positive or negative respectively
- `m` is the minimum field width, that is, how many spaces to leave for numbers
- `p` is the precision (this heavily depends on `X` as to what it means)
- `X` is a letter specifying the type (see next slide)

## Conversion Specifications Continued

Some of the possible values for X

- %d refers to a decimal number. The precision here will refer to the minimum number of digits to display. Default is 1.
- %e refers to a float in exponential form. The precision here will refer to the number of digits to display after the decimal point. Default is 6.
- %f refers to a float in “fixed decimal” format. The precision here is the same as above.
- %g refers to a float in one of the two aforementioned forms depending on the number's size. The precision here is the maximum number of **significant digits** (not the number of decimal points!) to display. This is the most versatile option useful if you don't know the size of the number.

## Example

```
#include <stdio.h>
int main(void) {
    double x = -2.61202e30;
    printf("%zu\n",
        sizeof(double));
    printf("%f\n", x);
    printf("%.2e\n", x);
    printf("%g\n", x);
    return 0;
}
```

Notice that on the %f line above we get some garbage at the end (it is tough for a computer to store floating numbers!).

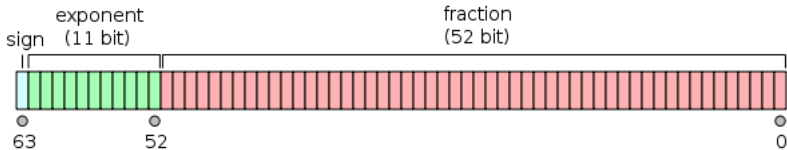
## Exercise

Write the code that displays the following numbers (Ensure you get the white space correct as well!)

1. 3.14150e+10
2.    0436 (two leading white spaces)
3. 436     (three white spaces at the end)
4. 2.00001

# IEEE 754 Floating Point Standard

- IEEE - Institute of Electrical and Electronics Engineers



- Number is

$$(-1)^{\text{sign}} \cdot \text{fraction} \cdot 2^{\text{exponent}}$$

(This is a bit of a lie but good enough for us - the details of this can get messy. See Wikipedia if you want more information)

(Picture courtesy of Wikipedia)



## A Fun Aside

- How do I convert 0.1 as a decimal number to a decimal number in binary?
- Binary fractions are sometimes called 2-adic numbers.
- Idea: Write 0.1 as below where each  $a_i$  is one of 0 or 1 for all integers  $i$ .

$$0.1 = \frac{a_1}{2} + \frac{a_2}{4} + \frac{a_3}{8} + \dots + \frac{a_k}{2^k} + \dots$$

- Multiplying by 2 yields

$$0.2 = a_1 + \frac{a_2}{2} + \frac{a_3}{4} + \dots + \frac{a_k}{2^{k-1}} + \dots (\text{Eqn1})$$

and so  $a_1 = 0$  since  $0.2 < 1$ .

- Repeating gives

$$0.4 = a_2 + \frac{a_3}{2} + \frac{a_4}{4} + \dots + \frac{a_k}{2^{k-2}} + \dots$$

and again  $a_2 = 0$ .

## Continuing

- From

$$0.4 = 0 + \frac{a_3}{2} + \frac{a_4}{4} + \dots + \frac{a_k}{2^{k-2}} + \dots$$

multiplying by 2 gives

$$0.8 = a_3 + \frac{a_4}{2} + \frac{a_5}{4} \dots + \frac{a_k}{2^{k-3}}$$

and again  $a_3 = 0$ . Doubling again gives

$$1.6 = a_4 + \frac{a_5}{2} + \frac{a_6}{4} \dots + \frac{a_k}{2^{k-4}}$$

and so  $a_4 = 1$ . Now, we subtract 1 from both sides and then repeat to see that... (see next slide)

## Continuing

$$1.6 - 1 = \frac{a_5}{2} + \frac{a_6}{4} \dots + \frac{a_k}{2^{k-4}}$$

$$0.6 = \frac{a_5}{2} + \frac{a_6}{4} \dots + \frac{a_k}{2^{k-4}}$$

$$1.2 = a_5 + \frac{a_6}{2} + \frac{a_7}{4} \dots + \frac{a_k}{2^{k-4}}$$

giving  $a_5 = 1$  as well. At this point, subtracting 1 from both sides gives

$$0.2 = \frac{a_6}{2} + \frac{a_7}{4} \dots + \frac{a_k}{2^{k-4}}$$

which is the same as (Eqn 1) from two slides ago and hence,

$$(0.1)_{10} = (0.00011)_2$$

## Short Hand

$$0.1 \cdot 2 = \mathbf{0.2}$$

$$0.2 \cdot 2 = \mathbf{0.4}$$

$$0.4 \cdot 2 = \mathbf{0.8}$$

$$0.8 \cdot 2 = \mathbf{1.6}$$

$$0.6 \cdot 2 = \mathbf{1.2}$$

$$0.2 \cdot 2 = \mathbf{0.4}$$

and so  $(0.1)_{10} = (0.000\overline{11})_2$

# Errors

- Notice that these floating point numbers only store rational numbers, that is, they cannot store real numbers (though there are CAS packages like Sage which try to).
- This for us is okay since the rationals can approximate real numbers as accurately as we need.
- When we discuss errors in approximation, we have two types of measures we commonly use, namely **absolute error** and **relative error**.

## Errors (Continued)

- Let  $r$  be the real number we're approximating and let  $p$  be the exact value.
- Absolute Error  $|p - r|$ . Eg.  $|3.14 - \pi| \approx 0.0015927\dots$
- Relative Error  $\frac{|p-r|}{r}$ . Eg.  $\frac{|3.14-\pi|}{\pi} = 0.000507$ .
- Note: Relative error can be large when  $r$  is small even if the absolute error is small.

## Errors (Continued)

Be wary of...

- Subtracting nearly equal numbers
- Dividing by very small numbers
- Multiplying by very large numbers
- Testing for equality

## An Example

```
#include <stdio.h>
int main(void) {
    double a = 7.0/12.0;
    double b = 1.0/3.0;
    double c = 1.0/4.0;
    if (b+c==a) printf("Everything is Awesome!");
    else printf("Not cool... %g",b+c-a);
}
```



## Watch out...

- Comparing  $x == y$  is often risky.
- To be safe, instead of using `if (x==y)` you can use `if (x-y < 0.0001 && y-x < 0.0001)` (or use absolute values - see next lecture!)
- We sometimes call  $\epsilon = 0.0001$  the **tolerance**.
- Note: Sometimes it is okay to compare floats to constants such as `if (x==0.0)` but you're best to exercise caution. Comparing to 0 is a surprisingly difficult problem.

## One Note

- What happens when you type `double a = 1/3`? Do you get 0.33333?
- In C, most operators are overloaded. When it sees `1/3`, C reads this as integer division and so returns the value of 0.
- There are a few ways to fix this, one of them is to make at least one of the value a double (or a float) by writing `double a = 1.0/3` (dividing a double by an integer or a double gives a double).
- Another way is by **typecasting**, that is, explicitly telling C to make a value something else.
- For example, `double a = ((double)1)/3` will work as expected.

# Math Library (Highlights)

- `#include <math.h>`
- Lots of interesting functions including:
  - `double sin(double x)` and similarly for `cos`, `tan`, `asin`, `acos`, `atan` etc.
  - `double exp(double x)` and similarly for `log`, `log10`, `log2`, `sqrt`, `fabs`, `ceil`, `floor` etc. (note `log` is the natural logarithm and `fabs` is the absolute value)
  - `int abs(int x)` is the absolute value function
  - `double pow(double x, double y)` gives  $x^y$ , the power function.
  - Constants: `M_PI`, `M_PI_2`, `M_PI_4`, `M_E`, `M_LN2`, `M_SQRT2`
  - Other values: `INFINITY`, `NAN`, `MAXFLOAT`

# Polynomials

- A polynomial is an expression with at least one indeterminate and coefficients lying in some set.
- For example,  $3x^3 + 4x^2 + 9x + 2$ .
- In general:  $p(x) = a_0 + a_1x + \dots + a_nx^n$
- We will primarily use ints for the coefficients. (maybe doubles later)
- Question: Brainstorm some different ways we can represent polynomials in memory. Discuss the pros and cons of each.

# Our Representation

- We will represent it as an array of  $n + 1$  coefficients where  $n$  is the degree.
- For our example  $3x^3 + 4x^2 + 9x + 2$ , we have  
`double p[] = {2.0, 9.0, 4.0, 3.0};`
- How do we evaluate a polynomial? That is, how can we implement:

```
double eval(double p[], int n, double x);
```

## Traditional Method

- Compute  $x, x^2, x^3, \dots, x^n$  for  $n - 1$  multiplications.
- Multiply each by  $a_1, a_2, \dots, a_n$  for another  $n$  multiplications.
- Add all the results  $a_0 + a_1x + \dots + a_nx^n$  for a final  $n$  multiplications.
- This gives a total of  $2n - 1$  multiplications and  $n$  additions.
- A note: Multiplication is an expensive operation compared to addition. Is there a way to reduce the number of multiplication operations?

# Horner's Method

- Named after William George Horner (1786-1837) but known long before him (dating back as early as pre turn of millennium Chinese mathematicians).
- Idea:

$$2 + 9x + 4x^2 + 3x^3 = 2 + x(9 + x(4 + 3x))$$

- Start inside out. Total operations are  $n$  multiplications and  $n$  additions.

## Horner's Method

```
#include <stdio.h>
#include <assert.h>
double horner(double p[], int n, double x){
    assert(n > 0);
    double y = p[n-1];
    for(int i=n-2; i >= 0; i--){
        y = y*x + p[i];
    }
    return y;
}
```

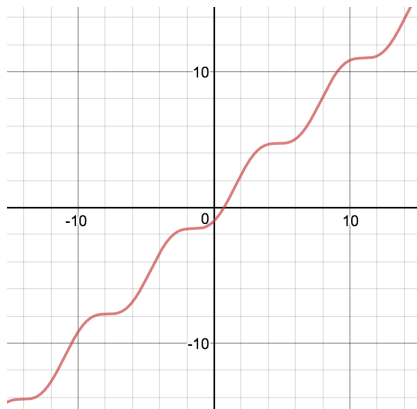


## Horner's Method (Continued)

```
int main(void) {  
    double p[] = {2,9,4,3};  
    int len = sizeof(p)/sizeof(p[0]);  
    printf("2 = %g\n", horner(p, len, 0));  
    printf("18 = %g\n", horner(p, len, 1));  
    printf("60 = %g\n", horner(p, len, 2));  
    printf("-6 = %g\n", horner(p, len, -1));  
    return 0;  
}
```

# Root Finding

- Given a function  $f(x)$ , how can we determine a root?
- Example:  $f(x) = x - \cos(x)$ . Courtesy: Desmos.



# Idea

- Notice that  $f(-10) < 0 < f(10)$  so a root must be in the interval of  $[-10, 10]$  (why!?)
- Look at the midpoint of the interval (namely 0) and evaluate  $f(0)$ .
- If  $f(0) > 0$ , look for a root in the interval  $[-10, 0]$ . Otherwise, look for a root in  $[0, 10]$ .
- Repeat until a root is found.

# Bisection Method

- For which types of functions is this method guaranteed to work?
- What cases should we worry about?
- Can we run forever?
- What is our stopping condition?

# Bisection Method

- For which types of functions is this method guaranteed to work?
- What cases should we worry about?
- Can we run forever?
- What is our stopping condition?
- Two stopping conditions possible
  - Stop when  $|f(m)| < \epsilon$  for some fixed  $\epsilon > 0$  where  $m$  is the midpoint of the interval.
  - Stop when  $|m_{n-1} - m_n| < \epsilon$  (where  $m_n$  is the  $n$ th midpoint).
- Should include a safety escape, namely some fixed number of iterations.

# Algorithm Pseudocode

- Given some  $a$  and  $b$  with  $f(a) > 0$  and  $f(b) < 0$ , set  $m = (a + b)/2$ .
- If  $f(m) < 0$ , set  $b = m$ .
- Otherwise, set  $a = m$
- Loop until either  $|f(m)| < \epsilon$ ,  $|m_{n-1} - m_n| < \epsilon$ , or the number of iterations has been met.

# Bisection.h

## Bisection.h

```
#ifndef BISECTION_H
#define BISECTION_H
/*
Pre: None
Post: Returns the value of  $x - \cos(x)$ 
*/
double f(double x);
/*
Pre:  $\epsilon > 0$  is a tolerance,  $\text{iterations} > 0$ ,
 $f(x)$  has only one root in  $[a, b]$ ,  $f(a)f(b) < 0$ 
Post: Returns an approximate root of  $f(x)$  using
bisection method. Stops when either number of
iterations is exceeded or  $|f(m)| < \epsilon$ 
*/
double bisect(double a, double b,
              double epsilon, int iterations);
#endif
```



## Bisection.c

```
#include <assert.h>
#include <math.h>
#include "bisection.h"
double f(double x){return x - cos(x);}
double bisect(double a, double b,
    double epsilon, int iterations){
    double m=a, fm;
    assert(epsilon > 0.0 && f(a)*f(b) < 0);
    for(int i=0; i<iterations; i++){
        m = (a+b)/2.0;
        fm = f(m); //Why is this a good idea?
        if (fabs(fm) < epsilon) return m;
        if (fm*f(b) > 0) b=m;
        else a=m;
    }
    return m;
}
```

## Main.c

```
#include <stdio.h>
#include "bisection.h"
int main(void) {
    printf("%g\n", bisect(-10,10,0.0001,50));
    return 0;
}
```

## Calculating the Number of Iterations

- We used the break condition  $|f(m)| < \epsilon$  in our code.
- One advantage to using the condition  $|m_n - m_{n-1}| < \epsilon$  is that we can compute the number of iterations fairly easily.
- After each iteration, the length of the interval is cut in half, so, we seek to find a value for  $n$  such that

$$\epsilon > \frac{b - a}{2^n}$$

rearranging gives

$$2^n > \frac{b - a}{\epsilon}$$

and so after logarithms

$$n \log 2 > \log(b - a) - \log(\epsilon)$$

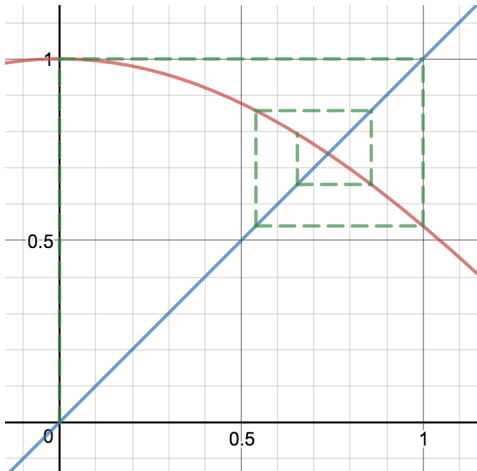
with  $b = 10$ ,  $a = -10$ ,  $\epsilon = 0.0001$ , we get  $n > 17.60964$ .

## Another Method - Fixed Point Iteration

- Given a function  $g(x)$ , we seek to find a value  $x_0$  such that  $g(x_0) = x_0$ .
- We call such a point a fixed point.
- These are of significant importance in dynamical systems.
- In our example, looking for a root of  $f(x) = x - \cos(x)$  is the same problem as finding a fixed point of  $g(x) = \cos(x)$ .
- Note: Not all functions have fixed points (but we can transfer between root solving problems and fixed point problems).
- There is another more visual way to interpret this...

# Cobwebbing

Also known as **Cobwebbing**. (Courtesy Desmos)



## A Note

$$x_0 = 0$$

$$g(x_0) = 1$$

$$g(g(x_0)) = g(1) = 0.540$$

$$g(g(g(x_0))) = g(g(1)) = g(0.540) = 0.858$$

$$g(g(g(g(x_0)))) = g(g(g(1))) = g(g(0.540)) = g(0.858) = 0.654$$

- It turns out by the Banach Contraction Mapping Theorem (or the Banach Fixed Point Theorem) that if the slope of the tangent line at a fixed point has magnitude less than 1, this cobwebbing process will eventually converge to a suitable starting point.

# Pseudocode

- Start with some point  $x_0$ .
- Compute  $x_1 = g(x_0)$ .
- If  $|x_1 - x_0| < \epsilon$ , stop.
- Otherwise go back to the beginning with  $x_0 = x_1$ .

Fixed.h



## Fixed.h

```
#ifndef FIXED_H
#define FIXED_H
/* Pre: None
   Post: Returns the value of  $\cos(x)$  */
double g(double x);
/*
Pre:  $\epsilon > 0$  is a tolerance,  $\text{iterations} > 0$ ,
 $x_0$  is sufficiently close to a stable fixed point
Post: Returns an approximate fixed point of  $g(x)$ 
      using cobwebbing. Stops when either number of
      iterations is exceeded or  $|g(x_i) - x_i| < \epsilon$ 
      where  $x_i$  is the value of  $x_0$  after  $i$  iterations.
*/
double fixed(double x0, double epsilon,
             int iterations);
#endif
```

## Fixed.c

```
#include <assert.h>
#include <math.h>
#include "fixed.h"
double g(double x){return cos(x);}
double fixed(double x0,
    double epsilon, int iterations){
    double x1;
    assert(epsilon > 0.0);
    for(int i=0; i<iterations; i++){
        x1 = g(x0);
        if (fabs(x1-x0) < epsilon) return x1;
        x0 = x1;
    }
    return x0;
}
```

## Main.c

```
#include <stdio.h>
#include "fixed.h"
int main(void) {
    printf("%g\n", fixed(0,0.0001,50));
    return 0;
}
```

## Improving the previous two codes

- Notice in each of the two previous examples, we hard coded a definition of a function.
- Ideally, the code would also have as a parameter the function itself.
- C lets us do this using function pointers.
- Syntax: Pass a parameter `double (*f)(double)` a pointer to a function that consumes a double and returns a double.
- Note: The brackets around `(*f)` are important to not confuse this with a function that returns a pointer.

Bisection2.h

## Bisection2.h

```
#ifndef BISECTION2_H
#define BISECTION2_H
double bisect2(double a, double b,
               double epsilon, int iterations,
               double (*f)(double));
#endif
```

## Bisection2.c

```
#include <assert.h>
#include <math.h>
#include "bisection2.h"
double bisect2(double a, double b,
               double epsilon, int iterations,
               double (*f)(double)){
    double m=a, fm;
    assert(epsilon > 0.0 && f(a)*f(b) < 0);
    for(int i=0; i<iterations; i++){
        m = (a+b)/2.0;
        fm = f(m);
        if (fabs(fm) < epsilon) return m;
        if (fm*f(b) > 0) b=m;
        else a=m;
    }
    return m;
}
```

## Main.c

```
#include <stdio.h>
#include <math.h>
#include "bisection2.h"

double g(double x){return x - cos(x);}
double h(double x){return x*x*x-x+1;}

int main(void) {
    printf("%g\n", bisect2(-10,10,0.0001,50,g));
    printf("%g\n", bisect2(-10,10,0.0001,50,h));
    return 0;
}
```