CS601: Software Development for Scientific Computing

Autumn 2023

Week1: Overview

Course Takeaways (intended)

- Non-CS majors:
 - 1. Write code (mostly in C/C++) and
 - 2. Develop software (not just write standalone code)
 - Numerical software
- CS-Majors:

In addition to the above two:

Learn to face mathematical equations and implement them with confidence

Why C++?

- C/C++/Fortran codes form the majority in scientific computing codes
- Catch a lot of errors early (e.g. at compile-time rather than at run-time)
- Has features for object-oriented software development
- Known to result in codes with better performance

What is this course about?

Software Development

+

Scientific Computing

Software Development

 Software development is the process of conceiving, specifying, designing, programming, documenting, testing, and bug fixing involved in creating and maintaining applications, frameworks, or other software components.

Software development is a process of writing and maintaining the source code, but in a broader sense, it includes all that is involved between the conception of the desired software through to the final manifestation of the software, ...

- Wikipedia on "Software Development"

Scientific Computing

- Also called computational science
 - Development of models to understand systems (biological, physical, chemical, engineering, humanities)

Collection of tools, techniques, and theories required to solve on a computer mathematical models of problems in science and engineering

This course NOT about...

- Software Engineering
 - Systematic study of Techniques, Methodology, and Tools to build correct software within time and price budget (topics covered in CS305)
 - People, Software life cycle and management etc.
- Scientific Computing
 - Rigorous exploration of numerical methods, their analysis, and theories
 - Programming models (topics covered in CS410)

Who this course is for?

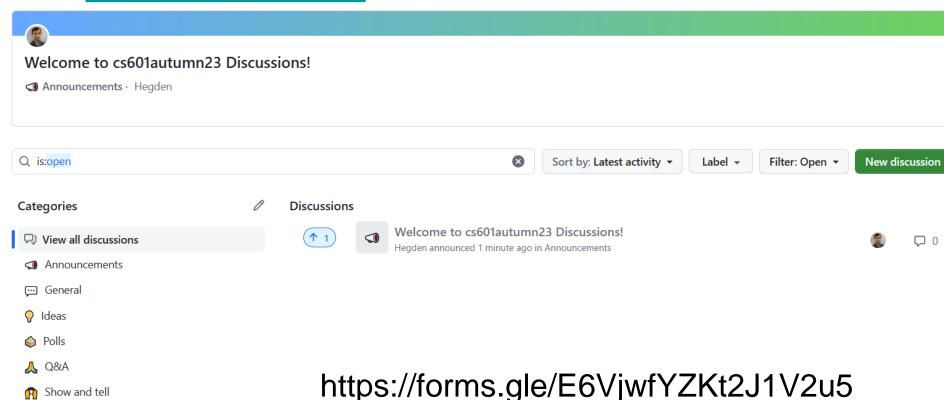
- You are interested in scientific computing
- You are interested in high-performance computing
- You want to build / add to a large software system

Course Webpage

https://hegden.github.io/cs601

GitHub Discussions

 https://github.com/IITDhCSE/cs601autumn 23/discussions



Let us try an exam question (from this course) on ChatGPT

Question:

Computing $\sqrt{x^2 + y^2}$ is a common problem. A common implementation strategy is as follows:

```
double ComputeHypotenuse(double x, double y) {
    return sqrt(x*x + y*y);
}
```

However, the above strategy is not robust. How would you implement a robust code?

 https://chat.openai.com/share/bcf4d871-21cf-4799-9275-1486345ce6dd

Takeaways:

- You still need to know the right questions to ask.
- Know if the answer provided makes sense.

Develop intuition

Let us dive into an example....

Example - Factorial

• $n! = n \times (n-1) \times (n-2) \times ... \times 3 \times 2 \times 1$ $(n-1)! = (n-1) \times (n-2) \times ... \times 3 \times 2 \times 1$ therefore,

Definition1: $n! = n \times (n-1)!$

is this definition complete?

plug 0 to n and the equation breaks.

Definition2:

$$n! = \begin{cases} n \times (n-1)! & \text{when } n>=1 \\ 1 & \text{when } n=0 \end{cases}$$

Exercise 1

 Does this code implement the definition of factorial correctly?

```
int fact(int n){
   if(n==0)
     return 1;

return n*fact(n-1);
}
```

Example - Factorial

Definition2:
$$n! = \begin{cases} n \times (n-1)! & \text{when } n>=1 \\ 1 & \text{when } n=0 \end{cases}$$

is this definition complete?

n! is not defined for negative n

Solution - Factorial

```
int fact(int n){
   if(n<0)
      return ERROR;
   if(n==0)
      return 1;

return n*fact(n-1);
}</pre>
```

Exercise 2

In how many flops does the code execute?
 assume 1 flop = 1 step executing any arithmetic operation

```
int fact(int n){
   if(n<0)
       return ERROR;
   if(n==0)
      return 1;

return n*fact(n-1);
}</pre>
```

Exercise 3

Does the code yield correct results for any n?

```
int fact(int n){
    if(n<0)
        return ERROR;
    if(n==0)
        return 1;

    return n*fact(n-1);
}</pre>
```

Who this course is for?

- Anybody who wishes to develop "computational thinking"
 - A skill necessary for everyone, not just computer programmers
 - An approach to problem solving, <u>designing</u> <u>systems</u>, and understanding human behavior that draws on concepts fundamental to computer science.

Computational Thinking - Examples

- How difficult is the problem to solve? And what is the best way to solve?
- · Modularizing something in anticipation of multiple users
- Prefetching and caching in anticipation of future use
- Thinking recursively
- Reformulating a seemingly difficult problem into one which we know how to solve by <u>reduction</u>, <u>embedding</u>, <u>transformation</u>, <u>simulation</u>
 - Are approximate solutions accepted?
 - False positives and False negatives allowed? etc.
- Using <u>abstraction</u> and <u>decomposition</u> in tackling large problem

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Computational Thinking – 2 As

Abstractions

- Our "mental" tools
- Includes: <u>choosing right abstractions</u>, operating at multiple <u>layers</u> of abstractions, and defining <u>relationships</u> among layers

Automation

- Our "metal" tools that <u>amplify</u> the power of "mental" tools
- Is mechanizing our abstractions, layers, and relationships
 - Need precise and exact notations / models for the "computer" below ("computer" can be human or machine)

Computing - 2 As Combined

- Computing is the automation of our abstractions
- Provides us the ability to <u>scale</u>
 - Make infeasible problems feasible
 - E.g. SHA-1 not safe anymore
 - Improve the answer's precision
 - E.g. capture the image of a black-hole

Summary: choose the right abstraction and computer

Computational Thinking – applied to the factorial exercise

- Need to be precise (formulating)
 - recall: n! = 1 for n=0, not defined for negative n
- Choosing right abstractions
 - recall: use of recursion, correct data type
- Ability to define the complexity
 - recall: flop calculation
- What else?

Computational Thinking – applied to the factorial exercise

- Need to be precise (formulating)
 - recall: n! = 1 for n=0, not defined for negative n
- Choosing right abstractions
 - recall: use of recursion, correct data type
- Ability to define the complexity
 - recall: flop calculation
- Choose the right "computer" for mechanizing the abstractions chosen

General Approach to Solving a Computational Problem

- 1. Problem statement: more precise this is, the easier it is to design and implement
- 2. Solution Algorithm: exactly how is the problem going to be solved
- 3. Implementation: breaking the algorithm into manageable pieces and putting it all together to solve the problem using a language of choice.
- **4. Verification:** checking that the implementation solves the original problem.
 - For numerical software this is often most difficult step, because you don't know the correct answer.

Scientific Software - Characteristics

- The answer is not a typical yes/no, red/blue/green
- The answer varies continuously. Think of computing the value of pi = 3.141592...
- Uses approximations. Think of discretization
- Employs efficient kernels
 - Kernels are core operations that are executed very frequently
- Should be able to adapt to change.
 - Writing everything from scratch is not an option
- Deals with large-scale problems
 - Lot of input/output data or both
 - Computationally hard

Toward Scientific Software

- Necessary Skills:
 - Understanding the mathematical problem
 - Understanding numerics
 - Designing algorithms and data structures
 - Selecting language and using libraries and tools
 - Verify the correctness of the results
 - Quick learning of new programming languages

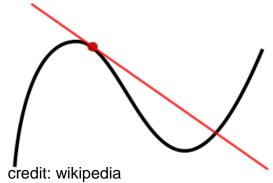
Exercise

Compute root(s) of:

$$x = \cos x$$
; $x \in \mathbb{R}$

roots, also called zeros, is the value of the argument/input to the function when the function output vanishes i.e. becomes zero

- let y = f(x) $f(x) = \cos(x) - x$
- At $x = x_n$, the value of y is $f(x_n)$. The coordinates of the point are $(x_n, f(x_n)) = known$ point.
- From calculus: <u>derivative</u> of a function of single variable at a chosen input value, when it exists, is the <u>slope of</u> <u>the tangent</u> to the graph at that input value.
 - $f'(x_n)$ is the slope of the line that is tangent to f(x) at x_n



From high-school math: point-slope formula for equation of a line

$$y - y_1 = m(x - x_1),$$

given the slope m and any known point (x_1, y_1)

- Substituting with:
 - $(x_n, f(x_n)) = \text{known point}$
 - $f'(x_n)$ = slope

Equation of the tangent line to graph of f(x) at x_n :

$$y - f(x_n) = f'(x_n)(x - x_n)$$

- Interested in finding roots i.e. value of x at y=0 i.e. at point $(x_{np1}, 0)$.
- Substituting in the equation of the tangent line,

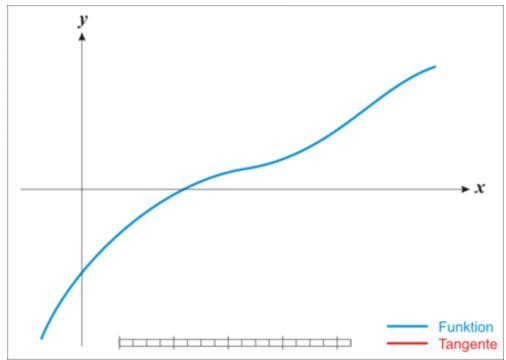
$$y - f(x_n) = f'(x_n)(x - x_n)$$

$$= -f(x_n) = f'(x_n)(x_{np1} - x_n)$$

$$= x_{np1} = x_n - f(x_n) / f'(x_n)$$

Visualizing

(SOURCE: https://en.wikipedia.org/wiki/Newton's_method):



The function f is shown in blue and the tangent line is in red. We see that x_{n+1} is a better approximation than x_n for the root x of the function f.

$$x_2 = x_1 - f(x_1) / f'(x_1)$$

 $x_3 = x_2 - f(x_2) / f'(x_2)$
 $x_4 = x_3 - f(x_3) / f'(x_3)$

Numerical Analysis

Talk to domain experts

- Choosing the initial value of x
- Does the method converge?
- What is an acceptable approximation?
- etc.

Designing Algorithms and Data Structures

Start with x₁

$$x_2 = x_1 - f(x_1) I f'(x_1)$$

 $x_3 = x_2 - f(x_2) I f'(x_2)$
 $x_4 = x_3 - f(x_3) I f'(x_3)$

. . .

- Repeat for up to maxIterations
- Check for x_{n+1} x_n to be "sufficiently small"
- Choose appropriate data types for x

Selecting libraries and tools

E.g. use the math library in C++ (cmath)

Verify the correctness of results

- Compare with 'gold' code / benchmark
- Compare with empirical data

Scientific Software - Examples

Entertainment

- Toy Story, Shrek rendered using data-center nodes

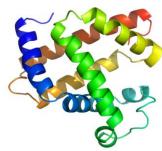
Economics

- ad-placement

Scientific Software - Examples

Biology

- Shotgun algorithm expedites sequencing of human genome



Credit: Wikipedia

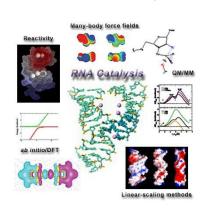
- Analyzing fMRI data



Credit: Wikipedia

Chemistry

- optimization and search algorithms to identify best chemicals for improving reaction conditions to improve yields



Scientific Software - Examples

Geology

- Modeling the Earth's surface to the core



Credit: Wikipedia

Astronomy

 kd-trees help analyze very large multidimensional data sets

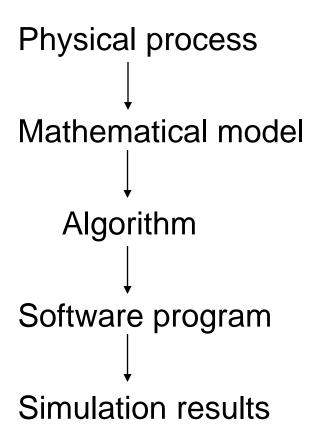


Credit: Kaggle.com

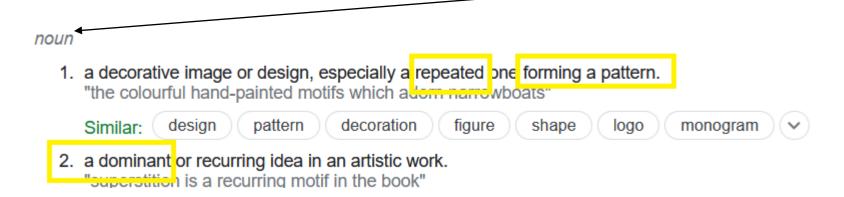
Engineering

 Boeing 777 tested via computer simulation (not via wind tunnel)

Recap: Toward Scientific Software



Scientific Software - Motifs



- 1. Finite State Machines
- 2. Combinatorial
- 3. Graph Traversal
- 4. Structured Grid
- 5. Dense Matrix
- 6. Sparse Matrix
- 7. <u>FFT</u>

- 8. Dynamic Programming
- 9. N-Body (/particle)
- 10. MapReduce
- 11. Backtrack / B&B
- 12. Graphical Models
- 13. Unstructured Grid