

COMPSCI 590N

Lecture 4: Computing Special Functions

Roy J. Adams

College of Information and Computer Sciences
University of Massachusetts Amherst

Outline

1 Preliminaries

2 Computing Special Functions

Miscellaneous Python Stuff

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- Mixing tabs and spaces when indenting code will cause an error.

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- In numerical computing we must also manipulate numbers, which can introduce further errors.
- **Numerical algorithms** approximate mathematical functions under the restriction of finite precision and finite time.

Long Division

Divide:	$\begin{array}{r} 2 \\ 3 \overline{)75} \end{array}$ <p>3 goes into 7 2 times... with some extra!</p>
Multiply:	$\begin{array}{r} 2 \\ 3 \overline{)75} \\ \underline{6} \end{array}$ <p>$2 \times 3 = 6$</p>
Subtract:	$\begin{array}{r} 2 \\ 3 \overline{)75} \\ \underline{-6} \\ 1 \end{array}$
Bring Down:	$\begin{array}{r} 2 \\ 3 \overline{)75} \\ \underline{-6} \\ 15 \end{array}$
Repeat:	$\begin{array}{r} 25 \\ 3 \overline{)75} \\ \underline{-6} \\ 15 \\ \underline{-15} \\ 0 \end{array}$ <p>$15 \div 3 = 5$ $5 \times 3 = 15$</p>

Types of Numerical Algorithms

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A direct method is guaranteed to terminate in a finite number of steps and would return the exact answer if we had infinite precision.

Iterative

An iterative method is never guaranteed to terminate, but instead builds a sequence of increasingly accurate approximations and terminates when a desired accuracy is achieved.

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DEMO

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■ Examples:

- 1 If an algorithm allows representability errors to accumulate through addition, then we would call this algorithm unstable.
- 2 Most iterative algorithms start with an initial guess and refine it. If the accuracy of a method is highly sensitive to the initial guess, then we would call this method unstable.

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- While you will likely never need to implement these algorithms (most are implemented in hardware!), it is important to know how they are implemented as it can have a large impact on the speed of your programs.
- In the following slides we will talk about some general techniques for approximating these functions and then look at a few examples.

Bit-by-bit Methods

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1	3	5	7
9	11	13	15
17	19	21	23
25	27	29	31

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2	3	6	7
10	11	14	15
18	19	22	23
26	27	30	31

Bit-by-bit Methods

4	5	6	7
12	13	14	15
20	21	22	23
28	29	30	31

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- Can often be implemented using only simple operations: add, shift (multiplying/dividing by two), comparison.
- Bit-by-bit methods exist for many standard math functions.
- In particular, trigonometric functions (sine, cosine, etc.) are often calculated using a bit-by-bit algorithm called the CORDIC (**CO**ordinate **R**otation **DI**gital **C**omputer) method.

Bit-by-bit Methods

Bit-by-bit method for calculating $f(x) = \sqrt{x}$:

```
Input: x, n_bits    Output: y = sqrt(x)
base = 2**(n_bits-1)
y = 0
for i = 1,...,n_bits:
    if (y + base)**2 <= x:
        y += base
    base = base / 2
return y
```


Iterative Methods

Assume we want to evaluate the function $f(x)$. One strategy is to pick a number a that is close to x such that we know $f(a)$. Using $f(a)$ as our initial guess we can refine it using a number of methods:

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For a given value a any infinitely differentiable function can be rewritten as:

$$f(x) = f(a) + \frac{f'(a)}{1!}(x - a) + \frac{f''(a)}{2!}(x - a)^2 + \dots$$

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- Values for $f(a)$ can be precomputed and stored in a table.

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Newton's Method

Newton's Method is used to find the solution to problems of the form $g(x) = 0$. Newton's method repeatedly applies the following update:

$$x_{n+1} = x_n - \frac{g(x_n)}{g'(x_n)}$$

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- Can be unstable depending on the initialization.

Newton's Method

For example: Let $f(x) = \sqrt{x}$. Note that if we find $1/\sqrt{x}$, we can find \sqrt{x} as $\sqrt{x} = x/\sqrt{x}$. So, let

$$y = \frac{1}{\sqrt{x}}$$
$$0 = y^{-2} - x$$

If we apply Newton's method to solve for y and apply some algebraic simplifications we get updates of the form:

$$y_{n+1} = \frac{y_n}{2} (3 - xy_n^2)$$