

# Numerical Solutions to Differential Equations

## Lecture Notes #2 — Calculus and Math 541 Review

Peter Blomgren,  
(blomgren.peter@gmail.com)

Department of Mathematics and Statistics  
Dynamical Systems Group  
Computational Sciences Research Center  
San Diego State University  
San Diego, CA 92182-7720

<http://terminus.sdsu.edu/>

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# Outline

- 1 **Calculus Review**
  - Limits, Continuity, and Convergence
  - Differentiability, Rolle's, and the Mean Value Theorem
  - Extreme Value, Intermediate Value, and Taylor's Theorem
  
- 2 **Math 541 Review**
  - Interpolation, Differentiation, Extrapolation
  - Integration, Degree of Accuracy
  - Newton-Cotes Formulas, Composite Integration

## Current Lecture — Reviewing Math 541 and Calculus

The purpose of this lecture is to “warm up” by reviewing some forgotten(?) material from the past.

Note that **complete lecture notes** for Math 541 are available on-line at [http://terminus.sdsu.edu/SDSU/Math541\\_f2014/](http://terminus.sdsu.edu/SDSU/Math541_f2014/).

## Why Review Calculus???

It's a good warm-up for our brains!

When developing numerical schemes we will use theorems from calculus to guarantee that our algorithms make sense.

If the theory is sound, when our programs fail we look for bugs in the code!

## Background Material — A Crash Course in Calculus

### Key concepts from Calculus

- Limits
- Continuity
- Convergence
- Differentiability
- Rolle's Theorem
- Mean Value Theorem
- Extreme Value Theorem
- Intermediate Value Theorem
- **Taylor's Theorem**

## Limit / Continuity

## Definition (Limit)

A function  $f$  defined on a set  $X$  of real numbers  $X \subset \mathbb{R}$  has the limit  $L$  at  $x_0$ , written

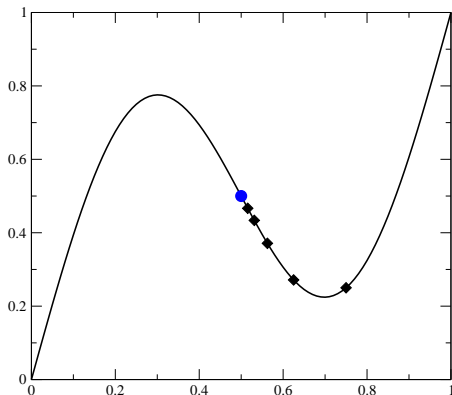
$$\lim_{x \rightarrow x_0} f(x) = L,$$

if given any real number  $\epsilon > 0$  ( $\forall \epsilon > 0$ ), there exists a real number  $\delta > 0$  ( $\exists \delta > 0$ ) such that  $|f(x) - L| < \epsilon$ , whenever  $x \in X$  and  $0 < |x - x_0| < \delta$ .

## Definition (Continuity (at a point))

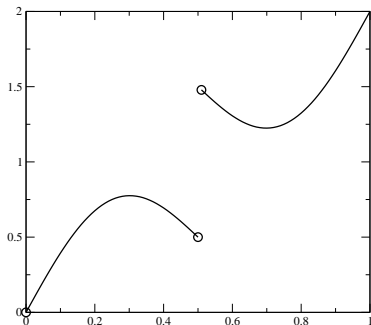
Let  $f$  be a function defined on a set  $X$  of real numbers, and  $x_0 \in X$ . Then  $f$  is continuous at  $x_0$  if

$$\lim_{x \rightarrow x_0} f(x) = f(x_0).$$

Example: Continuity at  $x_0$ 

Here we see how the limit  $x \rightarrow x_0$  (where  $x_0 = 0.5$ ) exists for the function  $f(x) = x + \frac{1}{2} \sin(2\pi x)$ .

## Examples: Jump Discontinuity



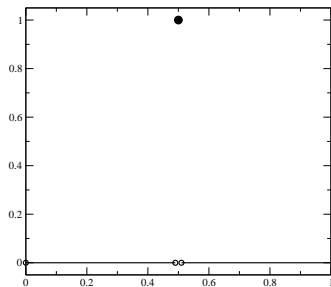
The function

$$f(x) = \begin{cases} x + \frac{1}{2} \sin(2\pi x) & x < 0.5 \\ x + \frac{1}{2} \sin(2\pi x) + 1 & x > 0.5 \end{cases}$$

has a jump discontinuity at  $x_0 = 0.5$ .



## Examples: "Spike" Discontinuity



The function

$$f(x) = \begin{cases} 1 & x = 0.5 \\ 0 & x \neq 0.5 \end{cases}$$

has a discontinuity at  $x_0 = 0.5$ .

The **limit exists**, but

$$\lim_{x \rightarrow 0.5} f(x) = 0 \neq 1$$

## Continuity / Convergence

## Definition (Continuity (in an interval))

The function  $f$  is continuous on the set  $X$  ( $f \in C(X)$ ) if it is continuous at each point  $x$  in  $X$ .

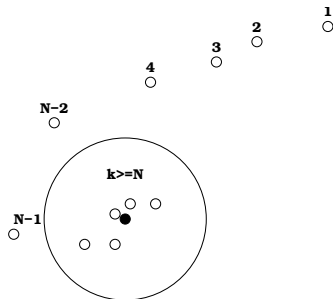
## Definition (Convergence of a sequence)

Let  $\underline{x} = \{x_n\}_{n=1}^{\infty}$  be an infinite sequence of real (or complex numbers). The sequence  $\underline{x}$  converges to  $x$  (has the limit  $x$ ) if  $\forall \epsilon > 0, \exists N(\epsilon) \in \mathbb{Z}^+ : |x_n - x| < \epsilon \forall n > N(\epsilon)$ . The notation

$$\lim_{n \rightarrow \infty} x_n = x$$

means that the sequence  $\{x_n\}_{n=1}^{\infty}$  converges to  $x$ .

## Illustration: Convergence of a Complex Sequence



A sequence in  $\underline{z} = \{z_k\}_{k=1}^{\infty}$  converges to  $z_0 \in \mathbb{C}$  (the black dot) if for any  $\epsilon$  (the radius of the circle), there is a value  $N$  (which depends on  $\epsilon$ ) so that the “tail” of the sequence  $\underline{z}_t = \{z_k\}_{k=N}^{\infty}$  is inside the circle.

# Differentiability

## Theorem

If  $f$  is a function defined on a set  $X$  of real numbers and  $x_0 \in X$ , the the following statements are **equivalent**:

- (a)  $f$  is continuous at  $x_0$
- (b) If  $\{x_n\}_{n=1}^{\infty}$  is any sequence in  $X$  converging to  $x_0$ , then  $\lim_{n \rightarrow \infty} f(x_n) = f(x_0)$ .

## Definition (Differentiability (at a point))

Let  $f$  be a function defined on an open interval containing  $x_0$  ( $a < x_0 < b$ ).  $f$  is differentiable at  $x_0$  if

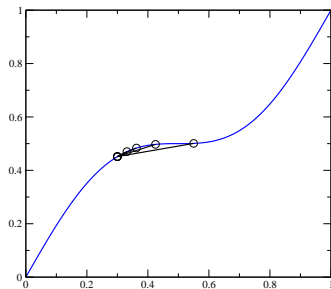
$$f'(x_0) = \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0} \text{ exists.}$$

If the limit exists,  $f'(x_0)$  is the derivative at  $x_0$ .

## Definition (Differentiability (in an interval))

If  $f'(x_0)$  exists  $\forall x_0 \in X$ , then  $f$  is differentiable on  $X$ .

## Illustration: Differentiability



Here we see that the limit

$$\lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0}$$

exists — and approaches the slope / derivative at  $x_0$ ,  $f'(x_0)$ .

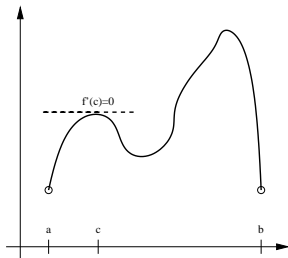
## Continuity / Rolle's Theorem

Theorem (Differentiability  $\Rightarrow$  Continuity)

If  $f$  is differentiable at  $x_0$ , then  $f$  is continuous at  $x_0$ .

Theorem (Rolle's Theorem [Wiki-Link](#))

Suppose  $f \in C[a, b]$  and that  $f$  is differentiable on  $(a, b)$ . If  $f(a) = f(b)$ , then  $\exists c \in (a, b): f'(c) = 0$ .

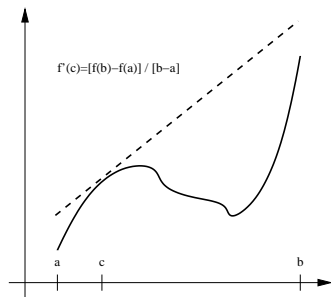


## Mean Value Theorem

Theorem (Mean Value Theorem [Wiki-Link](#))

If  $f \in C[a, b]$  and  $f$  is differentiable on  $(a, b)$ , then  $\exists c \in (a, b)$ :

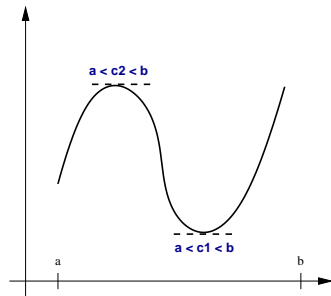
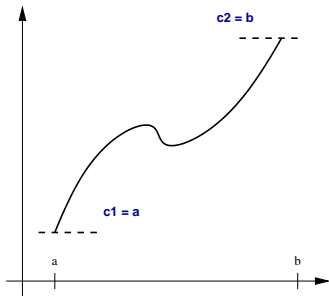
$$f'(c) = \frac{f(b) - f(a)}{b - a}.$$



## Extreme Value Theorem

Theorem (Extreme Value Theorem [Wiki-Link](#))

If  $f \in C[a, b]$  then  $\exists c_1, c_2 \in [a, b]: f(c_1) \leq f(x) \leq f(c_2)$   
 $\forall x \in [a, b]$ . If  $f$  is differentiable on  $(a, b)$  then the numbers  $c_1, c_2$  occur either at the endpoints of  $[a, b]$  or where  $f'(x) = 0$ .





## Intermediate Value Theorem

### Theorem (Intermediate Value Theorem [Wiki-Link](#))

*if  $f \in C[a, b]$  and  $K$  is any number between  $f(a)$  and  $f(b)$ , then there exists a number  $c$  in  $(a, b)$  for which  $f(c) = K$ .*

## Taylor's Theorem

Theorem (Taylor's Theorem [Wiki-Link](#))

Suppose  $f \in C^n[a, b]$ ,  $f^{(n+1)} \exists$  on  $[a, b]$ , and  $x_0 \in [a, b]$ . Then  $\forall x \in (a, b)$ ,  $\exists \xi(x) \in (x_0, x)$  with  $f(x) = P_n(x) + R_n(x)$  where

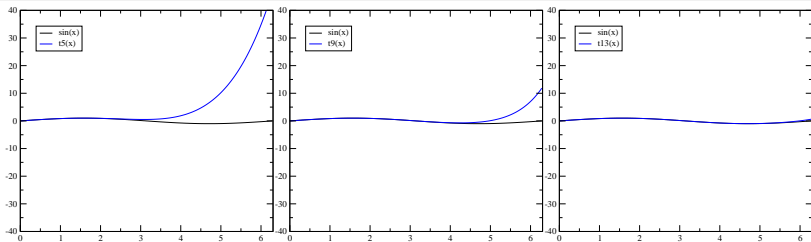
$$P_n(x) = \sum_{k=0}^n \frac{f^{(k)}(x_0)}{k!} (x-x_0)^k, \quad R_n(x) = \frac{f^{(n+1)}(\xi(x))}{(n+1)!} (x-x_0)^{(n+1)}.$$

$P_n(x)$  is called the **Taylor polynomial of degree  $n$** , and  $R_n(x)$  is the **remainder term** (truncation error).

This theorem is **extremely important** for numerical analysis; Taylor expansion is a fundamental step in the derivation of many of the algorithms we see in this class (and in Math 542 & 693ab).

## Illustration: Taylor's Theorem

$$f(x) = \sin(x)$$

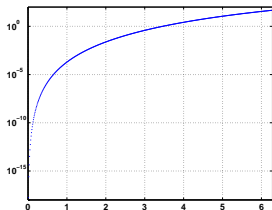
 $P_5(x)$  $P_9(x)$  $P_{13}(x)$ 

$$P_{13}(x) = \underbrace{x - \frac{1}{3!}x^3 + \frac{1}{5!}x^5}_{P_5(x)} - \frac{1}{7!}x^7 + \frac{1}{9!}x^9 - \frac{1}{11!}x^{11} + \frac{1}{13!}x^{13}$$
$$\underbrace{\hspace{15em}}_{P_9(x)}$$

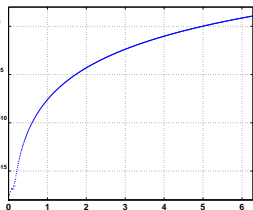
## Illustration: Taylor's Theorem

## Errors

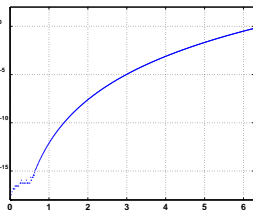
$f(x) = \sin(x)$



$E_5(x)$



$E_9(x)$



$E_{13}(x)$

$$P_{13}(x) = \underbrace{x - \frac{1}{3!}x^3 + \frac{1}{5!}x^5 - \frac{1}{7!}x^7 + \frac{1}{9!}x^9}_{P_5(x)} - \frac{1}{11!}x^{11} + \frac{1}{13!}x^{13}$$

$$\underbrace{\hspace{15em}}_{P_9(x)}$$

## Taylor Expansions — Matlab

- A **Taylor polynomial of degree  $n$**  requires all derivatives up to order  $n$ , and order  $n + 1$  for the **remainder**.
- Derivatives may be [more] complicated expression [than the original function].
- **Matlab** can compute derivatives for you:

Matlab: Symbolic Computations

Try this!!!

```
>> syms x
>> diff(sin(2*x))
>> diff(sin(2*x),3)
>> taylor(exp(x),5)
>> taylor(exp(x),5,1)
```

## Background Material

## Key concepts from Math 541

- Polynomial Interpolation and Approximation  
→ Approximation of Derivatives
- Numerical Integration  
→ Approximation of

$$\int_{t_i}^{t_{i+1}} f(t, y, y', \dots, y^{(n)}) dt$$

## Math 541 Review

### Key topics:

Lagrange Coefficients — Lagrange Polynomials

Newton's Divided Differences — An expression for the polynomial

Using polynomials to approximate  $f'(x)$

Richardson's Extrapolation

Numerical Integration — Return of the Lagrange Polynomials

## Interpolation: Lagrange Polynomials

Instead of working hard at *one point* (Taylor polynomials), we are going to construct a polynomial passing through the points  $(x_0, f(x_0)), (x_1, f(x_1)), (x_2, f(x_2)), \dots, (x_n, f(x_n))$ .

We define

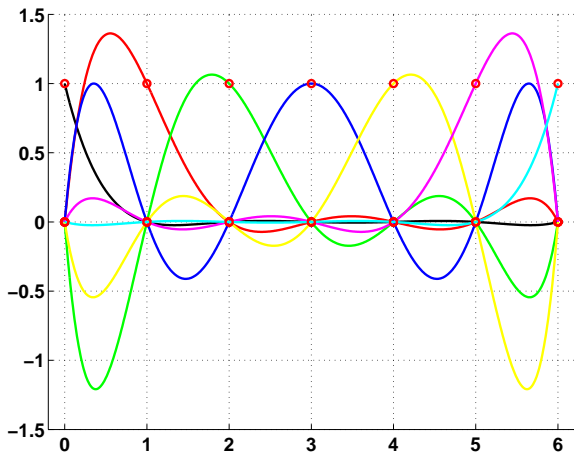
**Definition (the Lagrange coefficients,  $L_{n,k}(x)$ )**

$$L_{n,k}(x) = \prod_{i=0, i \neq k}^n \frac{x - x_i}{x_k - x_i},$$

$L_{n,k}(x)$  have the properties  $L_{n,i}(x_j) = \delta_{i,j}$ , and we use them as building blocks for the **Lagrange interpolating polynomial**:

$$P_n(x) = \sum_{k=0}^n f(x_k) L_{n,k}(x).$$



The Lagrange Coefficients,  $L_{n,k}(x)$ .E.g.  $L_{6,k}(x_i) = \delta_{i,k}$ ,  $\{x_i\}_{i=0}^6 = i$ :

## Newton's Divided Differences.

**Zeroth Divided Difference:**

$$f[x_i] = f(x_i).$$

**First Divided Difference:**

$$f[x_i, x_{i+1}] = \frac{f[x_{i+1}] - f[x_i]}{x_{i+1} - x_i}.$$

**Second Divided Difference:**

$$f[x_i, x_{i+1}, x_{i+2}] = \frac{f[x_{i+1}, x_{i+2}] - f[x_i, x_{i+1}]}{x_{i+2} - x_i}.$$

***k*th Divided Difference:**

$$f[x_i, x_{i+1}, \dots, x_{i+k}] = \frac{f[x_{i+1}, x_{i+2}, \dots, x_{i+k}] - f[x_i, x_{i+1}, \dots, x_{i+k-1}]}{x_{i+k} - x_i}.$$

## Newton's Divided Differences: Compute the Polynomial.

We can write the interpolating polynomial with the help of the divided differences:

$$P_n(x) = f[x_0] + \sum_{k=1}^n \left[ f[x_0, \dots, x_k] \prod_{m=0}^{k-1} (x - x_m) \right].$$

where  $f[x_0, \dots, x_k]$  are the diagonal entries from the divided difference table:

$x_0$	$f[x_0]$			
$x_1$	$f[x_1]$	$f[x_0, x_1]$		
$x_2$	$f[x_2]$	$f[x_1, x_2]$	$f[x_0, x_1, x_2]$	
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\ddots$

$$\frac{f[x_1, x_2] - f[x_0, x_1]}{x_2 - x_0}.$$

## Numerical Differentiation — Using Polynomials

Suppose  $\{x_0, x_1, \dots, x_n\}$  are distinct points in an interval  $\mathcal{I}$ , and  $f \in C^{n+1}(\mathcal{I})$ , we can write

$$f(x) = \sum_{k=0}^n f(x_k)L_k(x) + \frac{\prod_{k=0}^n (x - x_k)}{(n+1)!} f^{(n+1)}(\xi).$$

Formal differentiation gives:

$$\begin{aligned} f'(x) &= \sum_{k=0}^n f(x_k)L'_k(x) + \frac{d}{dx} \left[ \frac{\prod_{k=0}^n (x - x_k)}{(n+1)!} \right] f^{(n+1)}(\xi) \\ &\quad + \frac{\prod_{k=0}^n (x - x_k)}{(n+1)!} \frac{d}{dx} \left[ f^{(n+1)}(\xi) \right]. \end{aligned}$$

Since we will be evaluating  $f'(x_j)$  the last term gives no contribution.

The  $(n + 1)$  point formula for approximating  $f'(x_j)$

$$f'(x_j) = \sum_{k=0}^n f(x_k) L'_k(x_j) + \frac{f^{(n+1)}(\xi)}{(n+1)!} \left[ \prod_{\substack{k=0 \\ k \neq j}}^n (x_j - x_k) \right]$$

The formula is most useful when the node points are equally spaced, *i.e.*

$$x_k = x_0 + kh.$$

## Example: 3-point Formulas, I/III

Building blocks:

$$L_{2,0}(x) = \frac{(x - x_1)(x - x_2)}{(x_0 - x_1)(x_0 - x_2)}, \quad L'_{2,0}(x) = \frac{2x - x_1 - x_2}{(x_0 - x_1)(x_0 - x_2)}$$

$$L_{2,1}(x) = \frac{(x - x_0)(x - x_2)}{(x_1 - x_0)(x_1 - x_2)}, \quad L'_{2,1}(x) = \frac{2x - x_0 - x_2}{(x_1 - x_0)(x_1 - x_2)}$$

$$L_{2,2}(x) = \frac{(x - x_0)(x - x_1)}{(x_2 - x_0)(x_2 - x_1)}, \quad L'_{2,2}(x) = \frac{2x - x_0 - x_1}{(x_2 - x_0)(x_2 - x_1)}.$$

Formulas:

$$f'(x_j) = f(x_0) \left[ \frac{2x_j - x_1 - x_2}{(x_0 - x_1)(x_0 - x_2)} \right] + f(x_1) \left[ \frac{2x_j - x_0 - x_2}{(x_1 - x_0)(x_1 - x_2)} \right] \\ + f(x_2) \left[ \frac{2x_j - x_0 - x_1}{(x_2 - x_0)(x_2 - x_1)} \right] + \frac{f^{(3)}(\xi_j)}{6} \prod_{\substack{k=0 \\ k \neq j}}^2 (x_j - x_k).$$

## Example: 3-point Formulas, II/III

$$\begin{cases} f'(x_0) = \frac{1}{2h} [-3f(x_0) + 4f(x_1) - f(x_2)] + \frac{h^2}{3} f^{(3)}(\xi_0) \\ f'(x_1) = \frac{1}{2h} [-f(x_0) + f(x_2)] - \frac{h^2}{6} f^{(3)}(\xi_1) \\ f'(x_2) = \frac{1}{2h} [f(x_0) - 4f(x_1) + 3f(x_2)] + \frac{h^2}{3} f^{(3)}(\xi_2) \end{cases}$$

Use  $x_k = x_0 + kh$ :

$$\begin{cases} f'(x_0) = \frac{1}{2h} [-3f(x_0) + 4f(x_0 + h) - f(x_0 + 2h)] + \frac{h^2}{3} f^{(3)}(\xi_0) \\ f'(x_0 + h) = \frac{1}{2h} [-f(x_0) + f(x_0 + 2h)] - \frac{h^2}{6} f^{(3)}(\xi_1) \\ f'(x_0 + 2h) = \frac{1}{2h} [f(x_0) - 4f(x_0 + h) + 3f(x_0 + 2h)] + \frac{h^2}{3} f^{(3)}(\xi_2) \end{cases}$$

## Example: 3-point Formulas, III/III

$$\left\{ \begin{array}{l} f'(x_0) = \frac{1}{2h} [-3f(x_0) + 4f(x_0 + h) - f(x_0 + 2h)] + \frac{h^2}{3} f^{(3)}(\xi_0) \\ f'(x_0) = \frac{1}{2h} [-f(x_0 - h) + f(x_0 + h)] - \frac{h^2}{6} f^{(3)}(\xi_1) \\ f'(x_0) = \frac{1}{2h} [f(x_0 - 2h) - 4f(x_0 - h) + 3f(x_0)] + \frac{h^2}{3} f^{(3)}(\xi_2) \end{array} \right.$$

After the substitution  $x_0 + h \rightarrow x_0$  in the second equation, and  $x_0 + 2h \rightarrow x_0$  in the third equation.

**Note#1:** The third equation can be obtained from the first one by setting  $h \rightarrow -h$ .

**Note#2:** The error is smallest in the second equation.

**Note#3:** The second equation is a two-sided approximation, the first and third one-sided approximations.



## Richardson's Extrapolation

**What it is:** A general method for generating high-accuracy results using low-order formulas.

**Applicable when:** The approximation technique has an error term of predictable form, e.g.

$$M - N_j(h) = \sum_{k=j}^{\infty} E_k h^k,$$

where  $M$  is the unknown value we are trying to approximate, and  $N_j(h)$  the approximation (which has an error  $\mathcal{O}(h^j)$ .)

**Procedure:** Use two approximations of the same order, but with *different*  $h$ ; e.g.  $N_j(h)$  and  $N_j(h/2)$ . Combine the two approximations in such a way that the error terms of order  $h^j$  cancel.

## Building High Accuracy Approximations

1 of 2

Consider two first order approximations to  $M$ :

$$M - N_1(h) = \sum_{k=1}^{\infty} E_k h^k,$$

and

$$M - N_1(h/2) = \sum_{k=1}^{\infty} E_k \frac{h^k}{2^k}.$$

If we let  $N_2(h) = 2N_1(h/2) - N_1(h)$ , then

$$M - N_2(h) = \underbrace{2E_1 \frac{h}{2} - E_1 h}_0 + \sum_{k=2}^n E_k^{(2)} h^k,$$

where

$$E_k^{(2)} = E_k \left( \frac{1}{2^{k-1}} - 1 \right).$$

Hence,  $N_2(h)$  is now a **second order approximation** to  $M$ .

## Building High Accuracy Approximations

2 of 2

We can play the game again, and combine  $N_2(h)$  with  $N_2(h/2)$  to get a third-order accurate approximation, etc.

$$N_3(h) = \frac{4N_2(h/2) - N_2(h)}{3} = N_2(h/2) + \frac{N_2(h/2) - N_2(h)}{3}$$

$$N_4(h) = N_3(h/2) + \frac{N_3(h/2) - N_3(h)}{7}$$

$$N_5(h) = N_4(h/2) + \frac{N_4(h/2) - N_4(h)}{2^4 - 1}$$

In general, combining two  $j$ th order approximations to get a  $(j + 1)$ st order approximation:

$$N_{j+1}(h) = N_j(h/2) + \frac{N_j(h/2) - N_j(h)}{2^j - 1}$$

## Building Integration Schemes with Lagrange Polynomials

Given the nodes  $\{x_0, x_1, \dots, x_n\}$  we use the Lagrange interpolating polynomial

$$P_n(x) = \sum_{i=0}^n f_i L_i(x), \quad \text{with error} \quad E_n(x) = \frac{f^{(n+1)}(\xi(x))}{(n+1)!} \prod_{i=0}^n (x-x_i)$$

to obtain

$$\int_a^b f(x) dx = \int_a^b P_n(x) dx + \int_a^b E_n(x) dx.$$

## Identifying the Coefficients

$$\int_a^b P_n(x) dx = \int_a^b \sum_{i=0}^n f_i L_i(x) dx = \sum_{i=0}^n f_i \underbrace{\int_a^b L_i(x) dx}_{a_i} = \sum_{i=0}^n f_i a_i.$$

Hence we write

$$\int_a^b f(x) dx \approx \sum_{i=0}^n a_i f_i$$

with error given by

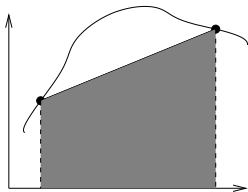
$$E(f) = \int_a^b E_n(x) dx = \int_a^b \frac{f^{(n+1)}(\xi(x))}{(n+1)!} \prod_{i=0}^n (x - x_i) dx.$$

## Example #1: Trapezoidal Rule

Let  $a = x_0 < x_1 = b$ , and use the linear interpolating polynomial

$$P_1(x) = f_0 \left[ \frac{x - x_1}{x_0 - x_1} \right] + f_1 \left[ \frac{x - x_0}{x_1 - x_0} \right], \quad \text{Then...}$$

$$\int_a^b f(x) dx = h \left[ \frac{f(x_0) + f(x_1)}{2} \right] - \frac{h^3}{12} f''(\xi), \quad h = b - a.$$



## Example #2: Simpson's Rule (with optimal error bound)

$$\int_{x_0}^{x_2} f(x) dx = h \left[ \frac{f(x_0) + 4f(x_1) + f(x_2)}{3} \right] - \frac{h^5}{90} f^{(4)}(\xi).$$

Taylor expand  $f(x)$  about  $x_1$ :

$$f(x) = f(x_1) + f'(x_1)(x - x_1) + \frac{f''(x_1)}{2}(x - x_1)^2 + \frac{f'''(x_1)}{6}(x - x_1)^3 + \frac{f^{(4)}(\xi(x))}{24}(x - x_1)^4$$

Integrating the error term gives

$$\int_a^b \frac{f^{(4)}(\xi(x))}{24} (x - x_1)^4 dx = \frac{f^{(4)}(\xi_1)}{60} h^5.$$

Using the approximation  $f''(x_1) = \frac{1}{h^2} [f(x_0) - 2f(x_1) + f(x_2)] - \frac{h^2}{12} f^{(4)}(\xi)$

$$\begin{aligned} \int_{x_0}^{x_2} f(x) dx &= 2hf(x_1) + \frac{h^3}{3} \left[ \frac{1}{h^2} [f(x_0) - 2f(x_1) + f(x_2)] - \frac{h^2}{12} f^{(4)}(\xi_2) \right] + \frac{f^{(4)}(\xi_1)}{60} h^5 \\ &= h \left[ \frac{f(x_0) + 4f(x_1) + f(x_2)}{3} \right] - \frac{h^5}{90} f^{(4)}(\xi). \end{aligned}$$

## Degree of Accuracy (Precision) of an Integration Scheme

## Definition (Degree of Accuracy)

The **Degree of Accuracy**, or **precision**, of a quadrature formula is the largest positive integer  $n$  such that the formula is exact for  $x^k$   $\forall k = 0, 1, \dots, n$ .

With this definition:

Scheme	Degree of Accuracy
Trapezoidal	1
Simpson's	3

Trapezoidal and Simpson's are examples of a class of methods known as **Newton-Cotes formulas**.



## Newton-Cotes Formulas — Two Types

Two types of Newton-Cotes Formulas:

**Closed** The  $(n + 1)$  point closed NCF uses nodes  $x_i = x_0 + ih$ ,  $i = 0, 1, \dots, n$ , where  $x_0 = a$ ,  $x_n = b$  and  $h = (b - a)/n$ . It is called closed since the endpoints are included as nodes.

**Open** The  $(n + 1)$  point open NCF uses nodes  $x_i = x_0 + ih$ ,  $i = 0, 1, \dots, n$  where  $h = (b - a)/(n + 2)$  and  $x_0 = a + h$ ,  $x_n = b - h$ . (We label  $x_{-1} = a$ ,  $x_{n+1} = b$ .)

## Closed Newton-Cotes Formulas

The approximation is

$$\int_a^b f(x) dx \approx \sum_{i=0}^n a_i f(x_i),$$

where

$$a_i = \int_{x_0}^{x_n} L_{n,i}(x) dx = \int_{x_0}^{x_n} \prod_{\substack{j=0 \\ j \neq i}}^n \frac{(x - x_j)}{(x_i - x_j)} dx.$$

## Closed Newton-Cotes Formulas — Error

## Theorem (Newton-Cotes Formulas, Error Term)

Suppose that  $\sum_{i=0}^n a_i f(x_i)$  denotes the  $(n+1)$  point closed Newton-Cotes formula with  $x_0 = a$ ,  $x_n = b$ , and  $h = (b-a)/n$ . Then there exists  $\xi \in (a, b)$  for which

$$\int_a^b f(x) dx = \sum_{i=0}^n a_i f(x_i) + \frac{h^{n+3} f^{(n+2)}(\xi)}{(n+2)!} \int_0^n t^2(t-1)\cdots(t-n) dt,$$

if  $n$  is even and  $f \in C^{n+2}[a, b]$ , and

$$\int_a^b f(x) dx = \sum_{i=0}^n a_i f(x_i) + \frac{h^{n+2} f^{(n+1)}(\xi)}{(n+1)!} \int_0^n t(t-1)\cdots(t-n) dt,$$

if  $n$  is odd and  $f \in C^{n+1}[a, b]$ .

Note that when  $n$  is an even integer, the degree of precision is  $(n+1)$ . When  $n$  is odd, the degree of precision is only  $n$ .

## Closed Newton-Cotes Formulas — Examples

**n = 2: Simpson's Rule**

$$\frac{h}{3} \left[ f(x_0) + 4f(x_1) + f(x_2) \right] - \frac{h^5}{90} f^{(4)}(\xi)$$

**n = 3: Simpson's  $\frac{3}{8}$ -Rule**

$$\frac{3h}{8} \left[ f(x_0) + 3f(x_1) + 3f(x_2) + f(x_3) \right] - \frac{3h^5}{80} f^{(4)}(\xi)$$

**n = 4: Boole's Rule**

$$\frac{2h}{45} \left[ 7f(x_0) + 32f(x_1) + 12f(x_2) + 32f(x_3) + 7f(x_4) \right] - \frac{8h^7}{945} f^{(6)}(\xi)$$

## Composite Simpson's Rule, I/II

For an even integer  $n$ : Subdivide the interval  $[a, b]$  into  $n$  sub-intervals, and apply Simpson's rule on each consecutive pair of sub-intervals. With  $h = (b - a)/n$  and  $x_j = a + jh$ ,  $j = 0, 1, \dots, n$ , we have

$$\begin{aligned}\int_a^b f(x) dx &= \sum_{j=1}^{n/2} \int_{x_{2j-2}}^{x_{2j}} f(x) dx \\ &= \sum_{j=1}^{n/2} \left\{ \frac{h}{3} \left[ f(x_{2j-2}) + 4f(x_{2j-1}) + f(x_{2j}) \right] - \frac{h^5}{90} f^{(4)}(\xi_j) \right\},\end{aligned}$$

for some  $\xi_j \in [x_{2j-2}, x_{2j}]$ , if  $f \in C^4[a, b]$ .

Since all the interior "even"  $x_{2j}$  points appear twice in the sum, we can simplify the expression a bit...

## Composite Simpson's Rule, II/II

$$\int_a^b f(x)dx = \frac{h}{3} \left[ f(x_0) - f(x_n) + \sum_{j=1}^{n/2} [4f(x_{2j-1}) + 2f(x_{2j})] \right] - \frac{h^5}{90} \sum_{j=1}^{n/2} f^{(4)}(\xi_j).$$

## Theorem (Composite Simpson's Rule)

Let  $f \in C^4[a, b]$ ,  $n$  be even,  $h = (b - a)/n$ , and  $x_j = a + jh$ ,  $j = 0, 1, \dots, n$ . There exists  $\mu \in (a, b)$  for which the **Composite Simpson's Rule** for  $n$  sub-intervals can be written with its error term as

$$\int_a^b f(x)dx = \frac{h}{3} \left[ f(a) - f(b) + \sum_{j=1}^{n/2} [2f(x_{2j}) + 4f(x_{2j+1})] \right] - \frac{(b-a)}{180} h^4 f^{(4)}(\mu).$$

## Romberg Integration

Romberg Integration is the combination of the **Composite Trapezoidal Rule** (CTR)

$$\int_a^b f(x)dx = \frac{h}{2} \left[ f(a) + f(b) + 2 \sum_{j=1}^{n-1} f(x_j) \right] - \frac{(b-a)}{12} h^2 f''(\mu)$$

and **Richardson Extrapolation**.

It yields a method for generating high-accuracy integral approximations using several “measurements” using the relatively crude (inaccurate) Trapezoidal Rule.

## Romberg Integration — Implemented

```
% Romberg Integration for sin(x) over [0,pi]
a = 0; b = pi; % The Endpoints
R = zeros(7,7);
R(1,1) = (b - a)/2 * (sin(a) + sin(b));
for k = 2 : 7
    h = (b - a)/2^(k-1);
    R(k,1) = 1/2 * (R(k-1,1) + 2 * h * sum(sin(a + (2 * (1 : (2^(k-2))) - 1) * h)));
end
for j = 2 : 7
    for k = j : 7
        R(k,j) = R(k,j-1) + (R(k,j-1) - R(k-1,j-1))/(4^(j-1) - 1);
    end
end
disp(R)
```



## Next Time, and Beyond

Simulating ODEs using Euler's method, and improvements...