

Basic Numerical Integration

• We want to find integration of functions of various forms of the equation known as the *Newton-Cotes* integration formulas ("*rules*").

• Newton-Cotes formula

Assume the value of f(x) defined on [a, b] is known at equally spaced points x_i (i = 0, 1, ..., n), where $x_0 = a$, and $x_n = b$. Then,

$$\int_a^b f(x) \, dx = \sum_{i=1}^n c_i f(x_i),$$

where $x_i = h \ i + x_0$, with h ("step size") = $(x_n - x_0)/n = (b - a)/n$.

The c_i 's are called *weights*. The x_i 's are called *nodes*. The precision of the approximation depends on n.

<u>Note</u>: The N-C rules use nodes **equally spaced**. But, they do not have to be –unequally spaced nodes are OK too.

Basic Numerical Integration

• Weights

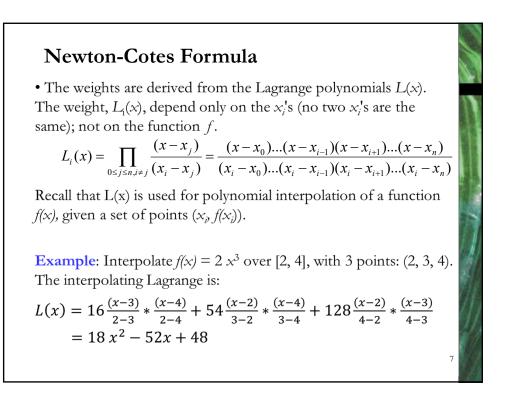
In the N-C formulae, they are derived from an approximation required to be equal for a polynomial of order lower or equal to the degree of the polynomials used to approximate the function. In other methods, weights and nodes can be derived jointly.

• Error analysis

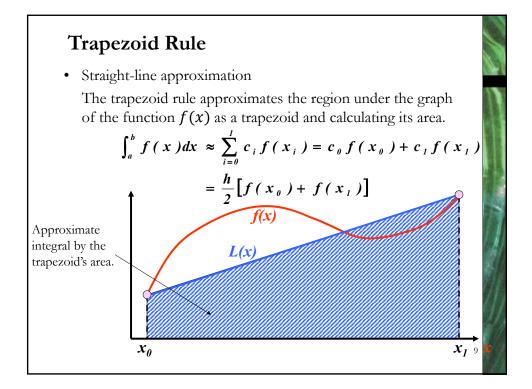
The error of the approximation is the difference between the value of the integral and the numerical result:

error =
$$\varepsilon = \int_a^b f(x) dx - \sum_{i=1}^n c_i f(x_i)$$

The errors are frequently approximated using Taylor series for f(x). The error analysis gives a strict upper bound on the error, if the derivatives of f are available.



Newton-Cotes Formula Different methods use different polynomials to get the c_i's. Newton-Cotes Closed Formulae – Use both end points Trapezoidal Rule : Linear Simpson's 1/3-Rule : Quadratic Simpson's 3/8-Rule : Cubic Boole's Rule : Fourth-order Newton-Cotes Open Formulae – Use only interior points midpoint rule



Trapezoid Rule – Derivation

• We use a Lagrange approximation (a polynomial) for f(x) over the interval $(x_n - x_0)$ (*Lagrange interpolation*), given by

$$f(x) \approx f(x_0)L_0(x) + f(x_1)L_1(x) + \dots + f(x_n)L_n(x)$$

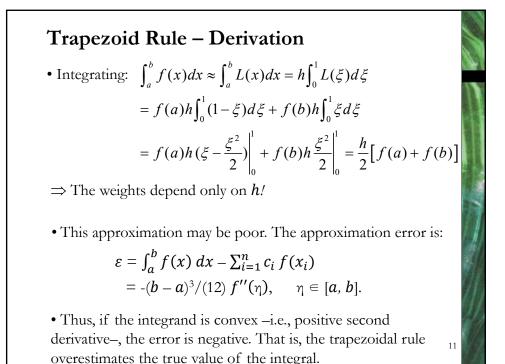
• For the case, n = 2, with the interval $(x_1 - x_0)$:

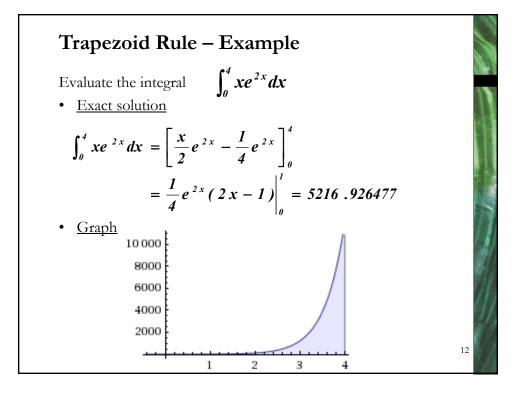
$$L(x) = \frac{x - x_1}{x_0 - x_1} f(x_0) + \frac{x - x_0}{x_1 - x_0} f(x_1)$$

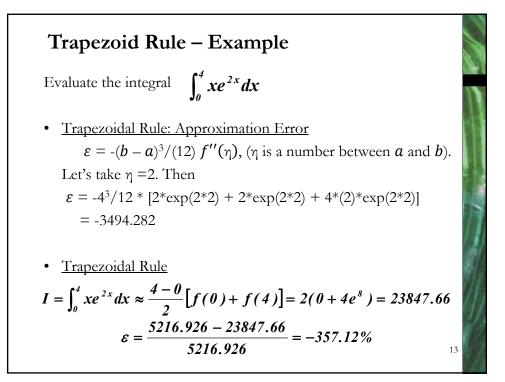
$$let \quad \mathbf{a} = \mathbf{x}_0, \ \mathbf{b} = \mathbf{x}_1, \ \xi = \frac{x - a}{b - a}, \ \mathbf{d}\xi = \frac{\mathbf{d}\mathbf{x}}{\mathbf{h}}; \ h = b - a$$

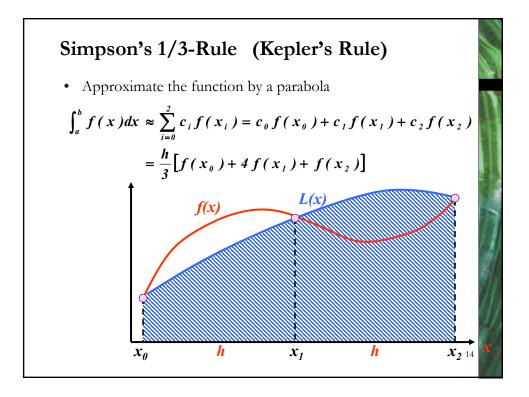
$$\begin{cases} x = a \quad \Rightarrow \xi = 0\\ x = b \quad \Rightarrow \xi = 1 \end{cases} \Rightarrow L(\xi) = (1 - \xi)f(a) + (\xi)f(b)$$

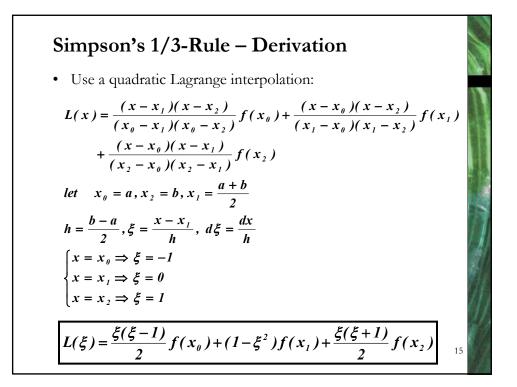
• Then, we integrate the Lagrange polynomial to obtain the
trapezoid rule

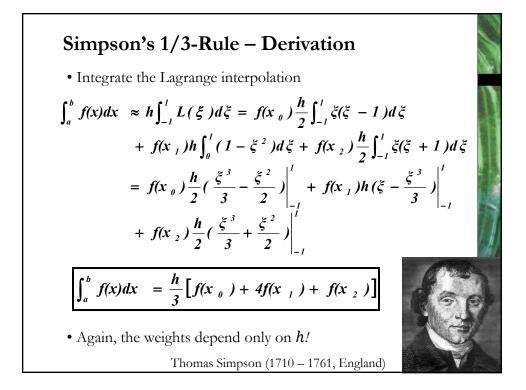


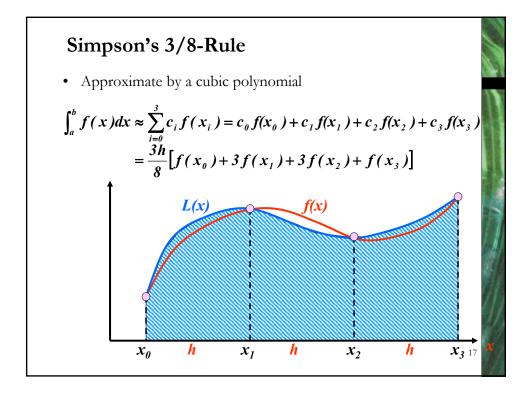


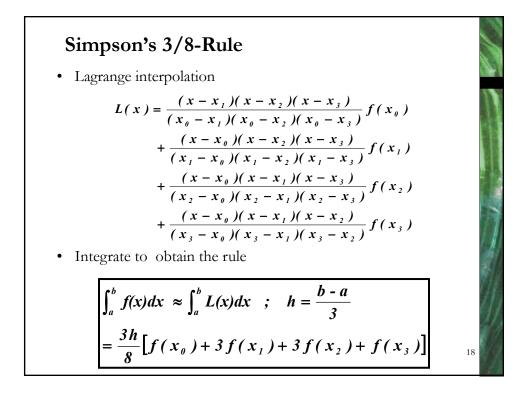


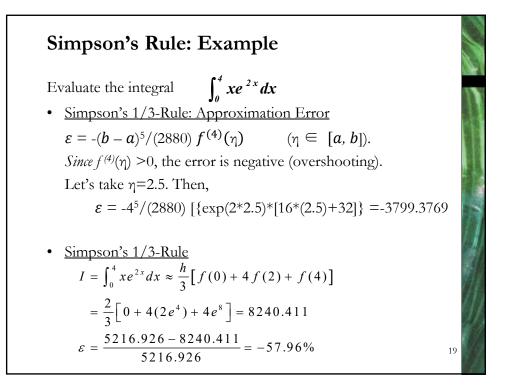


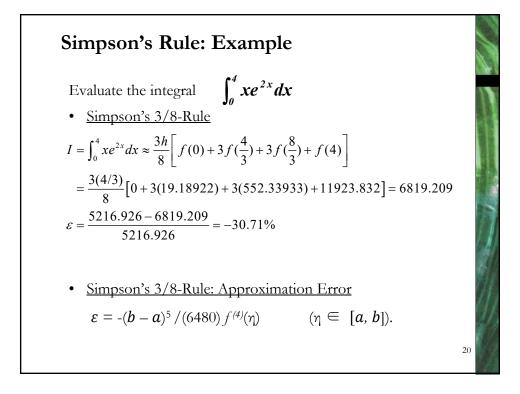


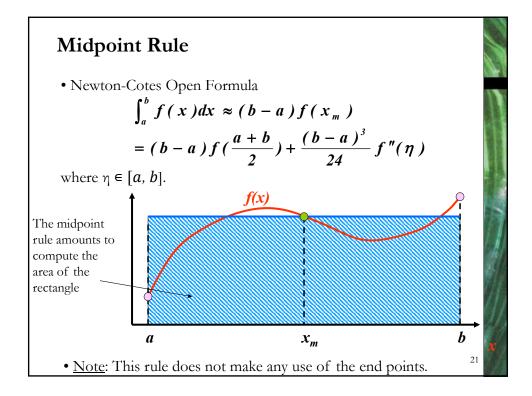


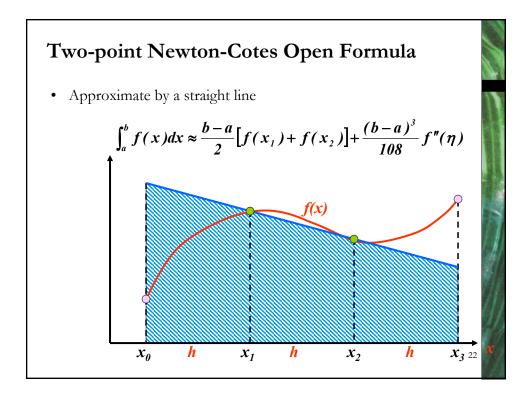


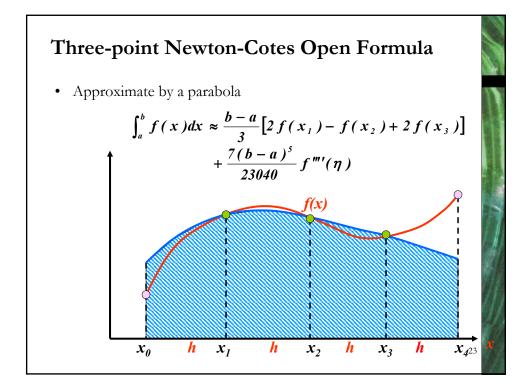


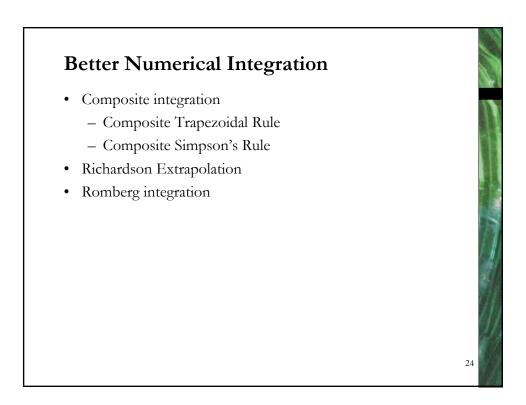


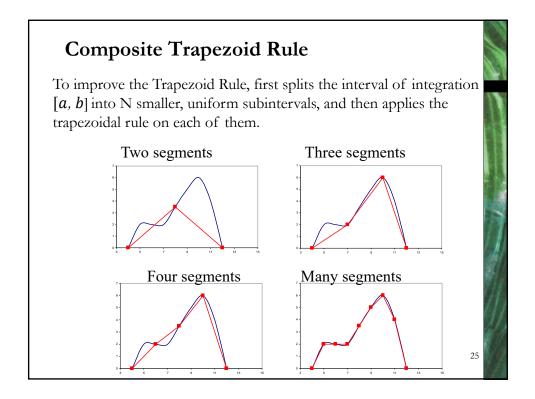


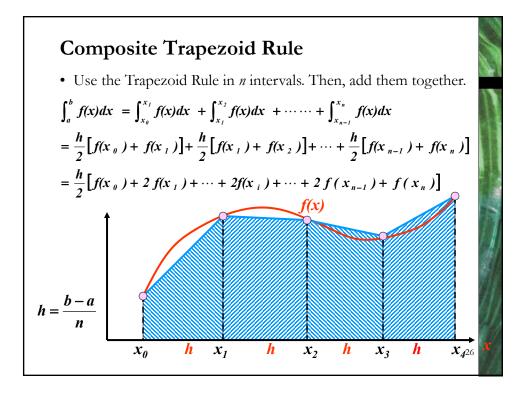


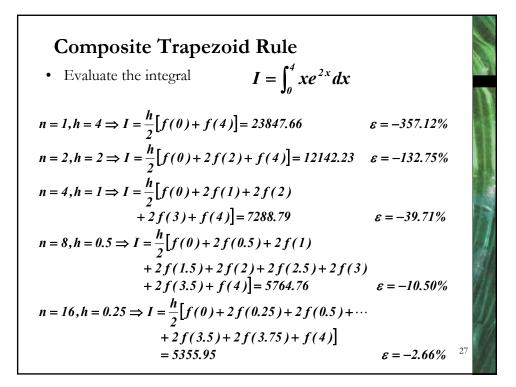


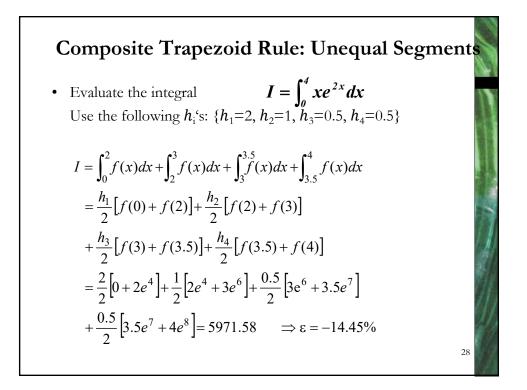


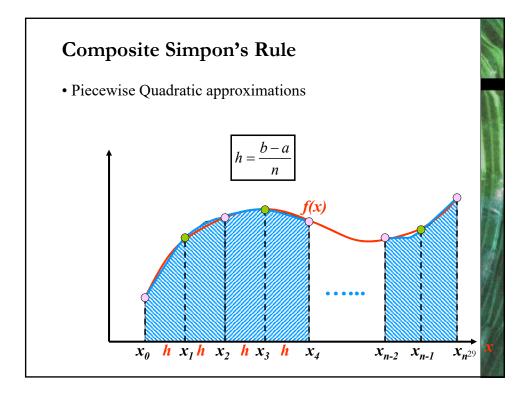


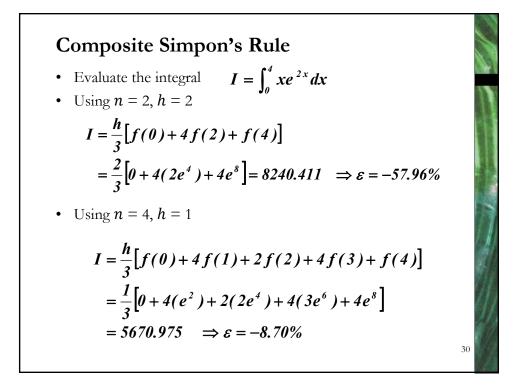


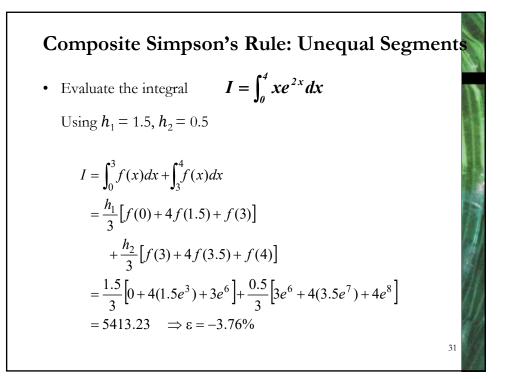












Gaussian Quadratures

• Newton-Cotes Formulae

- Nodes $(x_i$'s): Use evenly-spaced functional values

- Weights (c_i 's): Derived from an approximation required to be equal for a polynomial of order lower or equal to the degree of the polynomials used to approximate the function. Given nodes, best!

- Problem: Can explode for large *n* (Runge's phenomenon)

• Q: Can we use more efficient weights and nodes? Yes!

• Gaussian Quadratures

- Gaussian quadrature rules set the nodes and the weights in such a way that the approximation is exact when f(.) is a low order polynomial. Best choice for both, nodes and weights! 32

Gaussian Quadratures

• Gaussian quadrature computes an approximation to the integral:

$$\int_a^b f(x)dx = \sum_{i=1}^n c_i f(x_i),$$

 c_i 's are weights, x_i 's are the *quadrature nodes*, also called *cusps*. These values are not predetermined, but unknowns to be determined in some "optimal" fashion.

<u>Optimal Goal</u>: Get an exact answer if *f* is a $(2n - 1)^{th}$ -order polynomial. With n = 2, we get an exact answer *f* is a 3^{th} -order polynomial. (With n = 5, we get an exact answer *f* is a 9^{th} -order polynomial).

<u>Note</u>: A Gauss quadrature rule with 3 points yields exact value of an integral for a polynomial of degree $2 \times 3 - 1 = 5$. Simpson's 1/3 rule also uses 3 points, but the order of accuracy is 3.

Gaussian Quadratures – Features Gaussian Quadratures Features Select functional values at non-uniformly distributed points. The values are not predetermined, but unknowns determined by Legendre polynomials and integrating over a Lagrange interpolation. Several Gauss quadrature rules; we cover the Gauss-Legendre rules, which integrate from [-1, 1]. A change of variables is needed: t = b-a/2 x + a+b/h ⇒ the interval of integration is [-1,1]. Gauss-Legendre formulae for nodes and weights can be easily found online up to order n=100. With n nodes, delivers exact answer if f is (2n -1)th-order polynomial. Gauss-Legendre quadrature rule is not typically used for integrable functions with endpoint singularities.

Gaussian Quadratures - Nodes and Weights

Example: For n = 2, we choose (c_1, c_2, x_1, x_2) such that the method yields "exact integral" for $f(x) = x^0, x^1, x^2, x^3$.

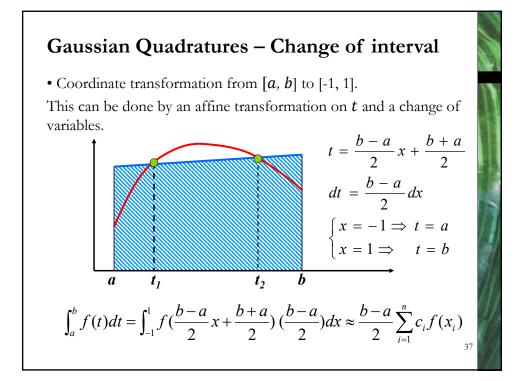
$$\begin{cases} f = 1 \implies \int_{-1}^{1} 1 dx = 2 = c_1 + c_2 \\ f = x \implies \int_{-1}^{1} x dx = \theta = c_1 x_1 + c_2 x_2 \\ f = x^2 \implies \int_{-1}^{1} x^2 dx = \frac{2}{3} = c_1 x_1^2 + c_2 x_2^2 \\ f = x^3 \implies \int_{-1}^{1} x^3 dx = \theta = c_1 x_1^3 + c_2 x_2^3 \end{cases}$$

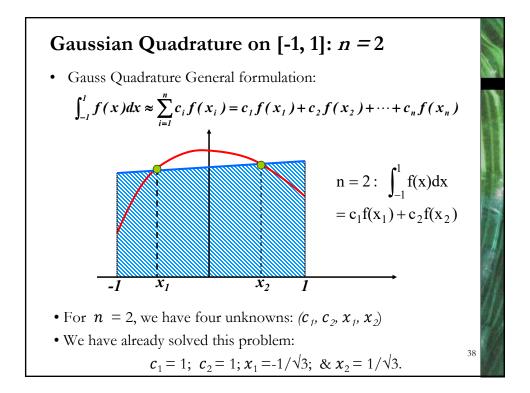
We solve this 4x4 system of equations to get (c_1, c_2, x_1, x_2) .

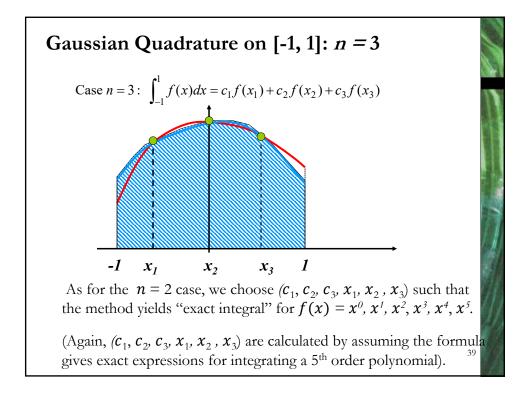
• By construction we get right answer for

$$f(x) = 1$$
 $(j = 0), f(x) = x$ $(j = 1), ..., f(x) = x^{j}$ $(j = 2 n - 1),$
 \Rightarrow enough to get the right answer for any polynomial of order $2n-1$.

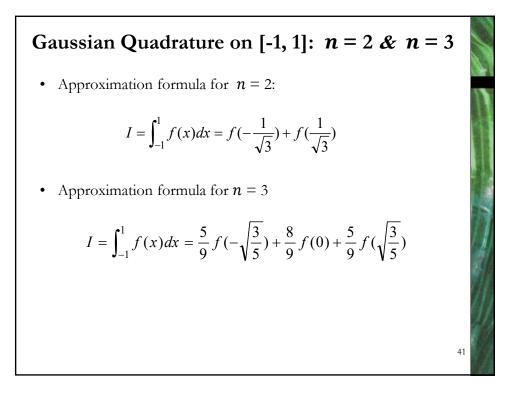
Gaussian Quadratures – Nodes and Weights Example (continuation): $n = 2 \Rightarrow$ Solve the 4x4 system: $\begin{cases}
f = 1 \Rightarrow \int_{-1}^{t} 1 dx = 2 = c_1 + c_2 \\
f = x \Rightarrow \int_{-1}^{t} x dx = \theta = c_1 x_1 + c_2 x_2 \\
f = x^2 \Rightarrow \int_{-1}^{t} x^2 dx = \frac{2}{3} = c_1 x_1^2 + c_2 x_2^2 \\
f = x^3 \Rightarrow \int_{-1}^{t} x^3 dx = \theta = c_1 x_1^3 + c_2 x_2^3
\end{cases} \Rightarrow \begin{cases}
c_1 = 1 \\
c_2 = 1 \\
x_1 = \frac{-1}{\sqrt{3}} \\
x_2 = \frac{1}{\sqrt{3}}
\end{cases}$ Note: This is <u>not</u> how it is done in practice: • x_i 's are chosen to be zeros of the degree-n Legendre polynomials $P_n(x)$ (not trivial to compute, but, they are tabulated). • Then, find the Lagrange polynomial that interpolates the integral f(x) at the selected x_i 's and integrate to get c_i 's.

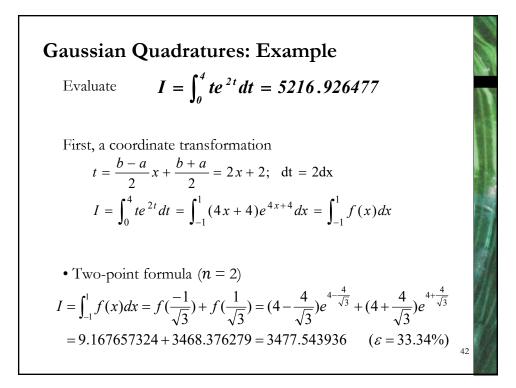


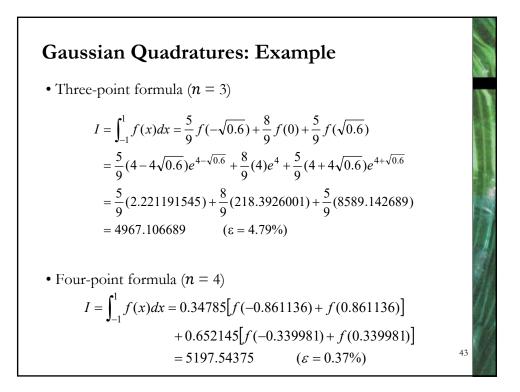


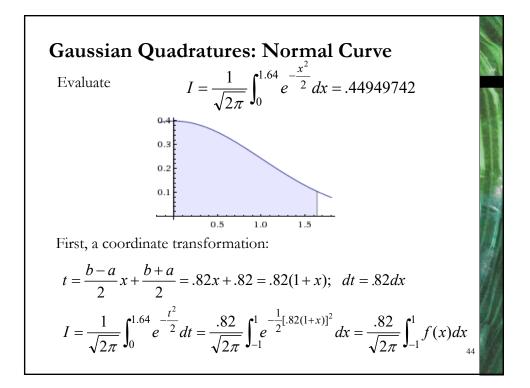


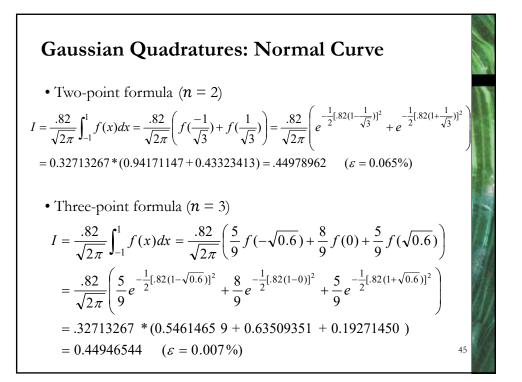
$$\begin{array}{l}
\textbf{Gaussian Quadrature on [-1, 1]: } n = 3 \\
f = 1 \Rightarrow \int_{-1}^{1} 1 dx = 2 = c_1 + c_2 + c_3 \\
f = x \Rightarrow \int_{-1}^{1} x dx = 0 = c_1 x_1 + c_2 x_2 + c_3 x_3 \\
f = x^2 \Rightarrow \int_{-1}^{1} x^2 dx = \frac{2}{3} = c_1 x_1^2 + c_2 x_2^2 + c_3 x_3^2 \\
f = x^3 \Rightarrow \int_{-1}^{1} x^3 dx = 0 = c_1 x_1^3 + c_2 x_2^3 + c_3 x_3^3 \\
f = x^4 \Rightarrow \int_{-1}^{1} x^4 dx = \frac{2}{5} = c_1 x_1^4 + c_2 x_2^4 + c_3 x_3^4 \\
f = x^5 \Rightarrow \int_{-1}^{1} x^5 dx = 0 = c_1 x_1^5 + c_2 x_2^5 + c_3 x_3^5 \\
\end{array}$$

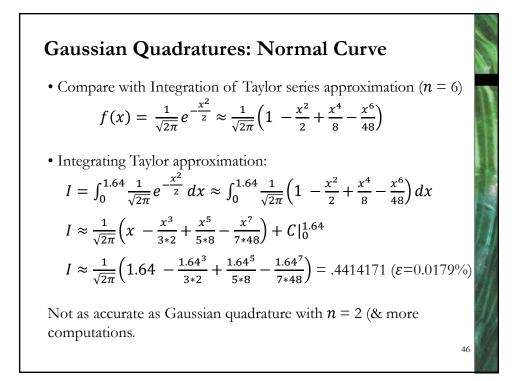












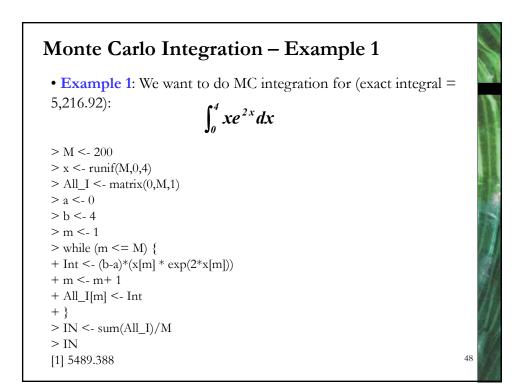
Monte Carlo Integration

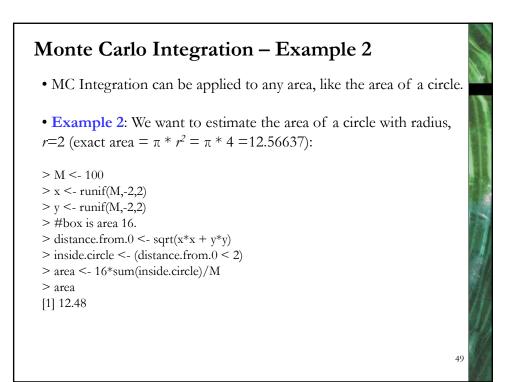
• In our motivation of integrals, we evaluated a one-dimensional integral by a sum of rectangles, using the end points of each interval to measure the height. Some of these rectangles overestimated the area, some underestimated the area.

• Let's focus on one of those rectangles, say with base [a, b]. We can also use as the height a *randomly* selected interior point, $x_1 \in [a, b]$ and estimate the integral, say $I(x_1)$. Of course, it may over- or underestimate the area.

• But, we can *randomly* select N interiors points and get N estimations of the area. Some points will under-estimate, some points will over-estimate, but, statistical intuition suggests that the average may work.

• In fact, as N increases, the average of the integral converges to the integral.





Monte Carlo Integration - Properties

- We formalize this idea with: $F_N = \frac{1}{N} \sum_{i=1}^N I(x_i)$
- This is our basic Monte Carlo (MC) estimator. Very simple.
- It can be shown it has good properties: unbiased, consistent (LLN applies), asymptotic normal (CLT applies).
- This results is very general and applies to many situations, for example, the trapezoid rule. Above, we selected two points to evaluate the integral (a and b). It produced a big over estimation.
- We can also randomly select two points between [a, b], say x_1 and calculate the integral, say $I(x_1)$. We repeat this evaluation of the integral at N randomly selected two points $\in [a, b]$: as N increases, the average of the integral converges to the integral.

