## SECTION 5: INTEGRATION

ESC 440 - Computational Methods for Engineers

## Introduction

## Integration

- Integration, or quadrature, has many engineering applications
$\square$ A few examples:
- Mean value

$$
\bar{y}=\frac{\int_{a}^{b} f(x) d x}{b-a}
$$

- Constitutive physical laws

$$
\begin{aligned}
& \Delta p=\int F(t) d t \\
& \Delta v=\frac{1}{C} \int i(t) d t \\
& \Delta x=\int v(t) d t
\end{aligned}
$$

- Total flux through a surface

$$
Q=\iint U(x, y) d x d y
$$

- Etc. ...


## Numerical Integration

$\square$ The numerical integration algorithms we'll look at can be divided into two broad categories:

- Algorithms for integration of data or functions
- No flexibility to choose the points, $f\left(x_{i}\right)$, used for calculation of the integral
- Points, $f\left(x_{i}\right)$, may or may not be uniformly-spaced
- Newton-Cotes formulas
- Algorithms for the integration of functions
- Exploit the ability to calculate $f(x)$ at any value of $x$
- Improved accuracy and efficiency
- Adaptive quadrature, Romberg integration, Gauss quadrature


## Newton-Cotes Formulas

This first category of numerical integration algorithms can be applied either to functions or to discrete data sets.

## Newton-Cotes Formulas

$\square$ Want to approximate the integral of a function or data set

$$
I=\int_{a}^{b} f(x) d x
$$

$\square$ Approximate $f(x)$ with something that is easy to integrate

- An $n^{\text {th }}$-order polynomial

$$
f(x) \approx f_{n}(x)=a_{0}+a_{1} x+\cdots+a_{n} x^{n}
$$

$\square$ Integral approximation:

$$
\hat{I}=\int_{a}^{b} f_{n}(x) d x \approx I
$$

$\square$ Unless otherwise noted, Newton-Cotes formulas assume evenlyspaced data points

## Closed Forms vs. Open Formulas

$\square$ Two different versions of the Newton-Cotes integral formulas:

- Closed forms
- Values of the function at the limits of integration, $f(a)$ and $f(b)$, are known
- Open forms
- $f(a)$ and $f(b)$ are unknown
$\square$ We'll focus on closed forms of the Newton-Cotes formulas



## Single-Segment vs. Composite

$\square$ Newton-Cotes formulas may be applied in two different ways:
$\square$ Single-segment

- Entire integration interval, $[a, b]$, approximated with a single polynomial

$\square$ Composite
- Integration interval divided into multiple segments
- Integral approximated for each segment - results summed



## 9 <br> Trapezoidal Rule

In the following sections, we'll look at three different Newton-Cotes integration formulas:

- Trapezoid rule
- Simpson's 1/3 rule
- Simpson's $3 / 8$ rule


## Trapezoidal Rule

$\square$ Approximate $f(x)$ as a first-order polynomial

$$
\begin{aligned}
& f(x) \approx f_{1}(x)=a_{0}+a_{1} x \\
& f_{1}(x)=f(a)+\frac{f(b)-f(a)}{b-a}(x-a)
\end{aligned}
$$

$\square$ Integral approximation:

$$
\hat{I}=\int_{a}^{b} f_{1}(x) d x=\int_{a}^{b}\left[f(a)+\frac{f(b)-f(a)}{b-a}(x-a)\right] d x \approx I
$$

$\square$ Trapezoidal rule formula:

$$
\hat{I}=(b-a) \frac{f(a)+f(b)}{2}
$$

## Trapezoidal Rule

$\square$ The trapezoidal rule formula

$$
\hat{I}=(b-a) \frac{f(a)+f(b)}{2}
$$

can be interpreted as

$$
\hat{I}=(\text { width }) \times(\text { avg.value })
$$

$\square$ All Newton-Cotes formulas can be expressed this way

- Only the approximation of the average value of $f(x)$ varies
- More accurate approx. of avg. value yields more accurate integral estimate

$\square$ Integral approximation is the area under the polynomial approximation of $f(x)$


## Trapezoidal Rule - Error

$\square$ The error of the trapezoidal rule estimate is

$$
E_{t}=\hat{I}-I=\frac{1}{12} f^{\prime \prime}(\xi)(b-a)^{3}
$$

where $\xi$ is some unknown value of $x$ on $[a, b]$
$\square$ Since $\xi$ is unknown, approximate the error as

$$
E_{a}=\frac{1}{12} \bar{f}^{\prime \prime}(b-a)^{3}
$$

where $\bar{f}^{\prime \prime}$ is the mean curvature of $f(x)$ on $[a, b]$

## Trapezoidal Rule - Error

$\square$ The error of the trapezoidal rule estimate is

$$
E_{t}=\frac{1}{12} f^{\prime \prime}(\xi)(b-a)^{3}
$$

$\square$ If the curvature of $f(x)$ is zero on $[a, b]$

$$
f^{\prime \prime}(x)=0, \text { for } a \leq x \leq b
$$

$\square$ Then the trapezoidal rule approximation is exact

$$
E_{t}=0
$$

$\square$ First-order polynomial is an exact representation of a linear $f(x)$

## Trapezoidal Rule - Example

$\square$ Trapezoidal rule may provide an accurate integral estimate

- Over regions with low curvature
- Where $f(x)$ is reasonably approximated as linear


$\square$ Or, large errors may result
- Over regions with large curvature
- Where a linear approximation is unacceptable


## Composite Trapezoidal Rule

$\square$ Accuracy can be improved by dividing the interval $[a, b]$ into $n$ segments

- $n+1$ evenly-spaced sample points of $f(x)$ : $x_{0} \ldots x_{n}$
$\square$ Segment width:

$$
h=\frac{b-a}{n}
$$

$\square$ Now approximating $f(x)$ as piece-wise linear


## Composite Trapezoidal Rule

$\square$ Divide the integral into $n$ segments

$$
I=\int_{x_{0}}^{x_{1}} f(x) d x+\int_{x_{1}}^{x_{2}} f(x) d x+\cdots+\int_{x_{n-1}}^{x_{n}} f(x) d x
$$

$\square$ Approximate each term using the trapezoidal rule

$$
\hat{I}=h \frac{f\left(x_{0}\right)+f\left(x_{1}\right)}{2}+h \frac{f\left(x_{1}\right)+f\left(x_{2}\right)}{2}+\cdots+h \frac{f\left(x_{n-1}\right)+f\left(x_{n}\right)}{2}
$$

$\square$ Using summation notation

$$
\hat{I}=\frac{h}{2}\left[f\left(x_{0}\right)+2 \sum_{i=1}^{n-1} f\left(x_{i}\right)+f\left(x_{n}\right)\right]
$$

$\square$ Or, in (width) $\times$ (avg.value) form

$$
\hat{I}=(b-a) \frac{\left[f\left(x_{0}\right)+2 \sum_{i=1}^{n-1} f\left(x_{i}\right)+f\left(x_{n}\right)\right]}{2 n}
$$

## Composite Trapezoidal Rule - Error

$\square$ Total error is the sum of the individual errors

$$
E_{t}=\sum_{i=1}^{n} E_{t, i}=\frac{1}{12} h^{3} \sum_{i=1}^{n} f^{\prime \prime}\left(\xi_{i}\right)=\frac{(b-a)^{3}}{12 n^{3}} \sum_{i=1}^{n} f^{\prime \prime}\left(\xi_{i}\right)
$$

$\square$ Again, approximate using $\bar{f}^{\prime \prime}$, the mean curvature

$$
E_{a}=\frac{(b-a)^{3}}{12 n^{3}} \sum_{i=1}^{n} \bar{f}^{\prime \prime}
$$

where

$$
\sum_{i=1}^{n} \bar{f}^{\prime \prime}=n \bar{f}^{\prime \prime}
$$

so

$$
E_{a}=\frac{(b-a)^{3}}{12 n^{2}} \bar{f}^{\prime \prime}
$$

## Composite Trapezoidal Rule - Example

$\square$ Accuracy improves as the number of segments increases

- Average curvature over each segment decreases
- $f(x)$ better approximated as linear over smaller regions

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## Trapezoidal Rule - Unequally-Spaced Data

$\square$ Trapezoidal rule can be easily modified to accommodate unequally-spaced data points
$\square$ Account for the width of each of the $n$ individual segments explicitly

$$
\hat{I}=h_{1} \frac{f\left(x_{0}\right)+f\left(x_{1}\right)}{2}+h_{2} \frac{f\left(x_{1}\right)+f\left(x_{2}\right)}{2}+\cdots+h_{n} \frac{f\left(x_{n-1}\right)+f\left(x_{n}\right)}{2}
$$

$\square$ Useful for measured data, where uneven spacing is not uncommon

## Trapezoidal Rule in Python - trapezoid( )

$\square$ The integrate module from the SciPy package includes several integration functions, including trapezoid rule

- Import it first:

> from scipy import integrate

## I = integrate.trapezoid(y, x)

- y: vector of dependent variable data
- x: vector of independent variable data
- I: trapezoidal rule approximation to the integral of $y$ with respect to $x$ (a scalar)
$\square$ Data need not be equally-spaced
- Segment widths calculated from $x$ values


## Cumulative Integral - cumulative_trapezoid()

$$
\begin{gathered}
\text { I = integrate.cumulative_trapezoid(y, x, } \\
\text { initial=0) }
\end{gathered}
$$

- y: n-vector of dependent variable data
$\square \mathrm{x}$ : n-vector of independent variable data
- initial: optional initial value inserted as the first value in I - if not given, I is an ( $\mathrm{n}-1$ )-vector
- I: trapezoidal rule approximation to the cumulative integral of y with respect to x (an n -vector)
$\square$ Result is a vector - equivalent to:

$$
I(x)=\int_{x_{1}}^{x} y(\tilde{x}) d \tilde{x}
$$

## trapezoid() and cumulative_trapezoid()



```
# trapz test.py
import numpy as np
from matplotlib import pyplot as plt
from scipy import integrate
x = np.linspace(0,1,2000)
f = lambda x: 1 / ((x-0.3)**2 + .01) + 1 / ((x-0.9)**2 + 0.04) + 14
y=f(x)
I = integrate.trapezoid(y, x)
Ic = integrate.cumulative_trapezoid(y, x)
plt.figure(1); plt.clf()
plt.subplot(211)
plt.plot(x,y,'-b',linewidth=2)
plt.ylabel('f(x)')
plt.title('Integrating with\ntrapezoid() and cumulative_trapezoid()',
    fontweight='bold')
plt.subplot(212)
plt.plot(x[1:],Ic,'-b', linewidth=2)
plt.xlabel('x'); plt.ylabel('I(x)')
plt.text(0.65,15,f'I = {I:1.3f}',
    fontsize=12,fontname='Tahoma')
```


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Simpson's 1/3 Rule

## Simpson's 1/3 Rule

$\square$ Approximate $f(x)$ with a second-order polynomial

$$
f(x) \approx f_{2}(x)
$$

where $f_{2}(x)$ can be expressed as a Lagrange polynomial:

$$
f_{2}(x)=\frac{\left(x-x_{1}\right)\left(x-x_{2}\right)}{\left(x_{0}-x_{1}\right)\left(x_{0}-x_{2}\right)} f\left(x_{0}\right)+\frac{\left(x-x_{0}\right)\left(x-x_{2}\right)}{\left(x_{1}-x_{0}\right)\left(x_{1}-x_{2}\right)} f\left(x_{1}\right)+\frac{\left(x-x_{0}\right)\left(x-x_{1}\right)}{\left(x_{2}-x_{0}\right)\left(x_{2}-x_{1}\right)} f\left(x_{2}\right)
$$

$\square$ Approximate the integral of $f(x)$ as the integral of the quadratic approximation

$$
I \approx \hat{I}=\int_{a}^{b} f_{2}(x) d x
$$

## Simpson's 1/3 Rule

$\square$ Now fitting a parabola to $f(x)$
$\square$ Three points required: $x_{0}, x_{1}$, and $x_{2}$
$\square$ Integration interval, [ $a, b$ ] divided into two segments
$\square$ Points must be evenly spaced

## Simpson's 1/3 Rule

$\square$ Evaluating the integral of the quadratic approximation, $f_{2}(x)$, yields Simpson's 1/3 rule:

$$
\hat{I}=\frac{h}{3}\left[f\left(x_{0}\right)+4 f\left(x_{1}\right)+f\left(x_{2}\right)\right]
$$

$\square$ Or, in $\hat{I}=($ width $) \times($ avg.value $)$ form:

$$
\hat{I}=(b-a) \frac{f\left(x_{0}\right)+4 f\left(x_{1}\right)+f\left(x_{2}\right)}{6}
$$

## Simpson's 1/3 Rule - Error

$\square$ The error associated with Simpson's $1 / 3$ rule is

$$
E_{t}=\frac{1}{90} h^{5} f^{(4)}(\xi)=\frac{(b-a)^{5}}{2880} f^{(4)}(\xi)
$$

$\square$ Error is proportional to the fourth derivative of $f(x)$

- For third- and lower-order polynomials, $f^{(4)}=0$
$\square$ The Simpson's $1 / 3$ rule integral estimate is exact for cubic and lower-order polynomials
$\square$ An interesting result, given that $f(x)$ is approximated with only a quadratic


## Composite Simpson's 1/3 Rule

$\square$ Accuracy can be improved by dividing the interval $[a, b]$ into $n$ segments
$\square$ Each application of Simpson's $1 / 3$ rule requires three points, and two segments

- Total number of segments must be even
- Total number of points must be odd
$\square f(x)$ approximated as a quadratic over each pair of adjacent segments



## Composite Simpson's 1/3 Rule

$\square$ Divide $[a, b]$ into $n$ segments, and the integral into $n / 2$ segments

$$
I=\int_{x_{0}}^{x_{2}} f(x) d x+\int_{x_{2}}^{x_{4}} f(x) d x+\cdots+\int_{x_{n-2}}^{x_{n}} f(x) d x
$$

$\square$ Approximate each term using Simpson's 1/3 rule

$$
\hat{I}=\frac{h}{3}\left[f\left(x_{0}\right)+4 f\left(x_{1}\right)+f\left(x_{2}\right)\right]+\frac{h}{3}\left[f\left(x_{2}\right)+4 f\left(x_{3}\right)+f\left(x_{4}\right)\right]+\cdots+\frac{h}{3}\left[f\left(x_{n-2}\right)+4 f\left(x_{n-1}\right)+f\left(x_{n}\right)\right]
$$

$\square$ Using summation notation

$$
\hat{I}=\frac{h}{3}\left[f\left(x_{0}\right)+4 \sum_{i=1,3,5 \ldots}^{n-1} f\left(x_{i}\right)+2 \sum_{j=2,4,6 \ldots}^{n-2} f\left(x_{j}\right)+f\left(x_{n}\right)\right]
$$

$\square$ Or, in (width) $\times$ (avg.value) form

$$
\hat{I}=(b-a) \frac{\left[f\left(x_{0}\right)+4 \sum_{i=1,3,5 \ldots . .}^{n-1} f\left(x_{i}\right)+2 \sum_{j=2,4,6 \ldots . .}^{n-2} f\left(x_{j}\right)+f\left(x_{n}\right)\right]}{3 n}
$$

## Composite Simpson's 1/3 Rule - Example




## 31 <br> Simpson's 3/8 Rule

## Simpson's 3/8 Rule

$\square$ Approximate $f(x)$ with a third-order polynomial

$$
f(x) \approx f_{3}(x)
$$

where $f_{3}(x)$ can, again, be expressed as a Lagrange polynomial
$\square$ Approximate the integral of $f(x)$ as the integral of the cubic approximation

$$
I \approx \hat{I}=\int_{a}^{b} f_{3}(x) d x
$$

## Simpson's 3/8 Rule

$\square$ Now fitting a cubic to $f(x)$
$\square$ Four points required: $x_{0}, x_{1}$, $x_{2}$, and $x_{3}$
$\square$ Integration interval, [ $a, b$ ] divided into three segments
$\square$ Points must be evenly spaced

## Simpson's 3/8 Rule

$\square$ Evaluating the integral of the cubic approximation, $f_{3}(x)$, yields Simpson's 3/8 rule:

$$
\hat{I}=\frac{3 h}{8}\left[f\left(x_{0}\right)+3 f\left(x_{1}\right)+3 f\left(x_{2}\right)+f\left(x_{3}\right)\right]
$$

$\square$ Or, in $\hat{I}=($ width $) \times($ avg.value $)$ form:

$$
\hat{I}=(b-a) \frac{f\left(x_{0}\right)+3 f\left(x_{1}\right)+3 f\left(x_{2}\right)+f\left(x_{3}\right)}{8}
$$

## Simpson's 3/8 Rule - Error

$\square$ The error associated with Simpson's 3/8 rule is

$$
E_{t}=\frac{3}{80} h^{5} f^{(4)}(\xi)=\frac{(b-a)^{5}}{6480} f^{(4)}(\xi)
$$

$\square$ Error is proportional to the fourth derivative of $f(x)$
$\square$ Third-order accuracy

- Same as Simpson's 1/3 rule
$\square$ For nonzero $f^{(4)}$, error is slightly lower than Simpson's 1/3 rule


## Composite Simpson's 3/8 Rule

$\square$ Accuracy can be improved by dividing the interval $[a, b]$ into $n$ segments
$\square$ Each application of Simpson's $3 / 8$ rule requires four points, and three segments

- Total number of segments must be divisible by three
- Can be used in conjunction with Simpson's $1 / 3$ rule to accommodate an odd number of segments
$\square f(x)$ approximated as a cubic over each group of three adjacent segments



## Composite Simpson's 3/8 Rule

$\square$ Divide $[a, b]$ into $n$ segments, and the integral into $n / 3$ segments

$$
I=\int_{x_{0}}^{x_{3}} f(x) d x+\int_{x_{3}}^{x_{6}} f(x) d x+\cdots+\int_{x_{n-3}}^{x_{n}} f(x) d x
$$

$\square$ Approximate each term using Simpson's 3/8 rule

$$
\begin{aligned}
& \hat{I} \\
& =\frac{3 h}{8}\left[f\left(x_{0}\right)+3 f\left(x_{1}\right)+3 f\left(x_{2}\right)+f\left(x_{3}\right)\right]+\frac{3 h}{8}\left[f\left(x_{3}\right)+3 f\left(x_{4}\right)+3 f\left(x_{5}\right)+f\left(x_{6}\right)\right]+\cdots \\
& +\frac{3 h}{8}\left[f\left(x_{n-3}\right)+3 f\left(x_{n-2}\right)+3 f\left(x_{n-1}\right)+f\left(x_{n}\right)\right]
\end{aligned}
$$

$\square$ Using summation notation

$$
\hat{I}=\frac{3 h}{8}\left[f\left(x_{0}\right)+3 \sum_{i=1,4,7 \ldots}^{n-2} f\left(x_{i}\right)+3 \sum_{j=2,5,8 \ldots . .}^{n-1} f\left(x_{j}\right)+2 \sum_{k=3,6,9 \ldots}^{n-3} f\left(x_{k}\right)+f\left(x_{n}\right)\right]
$$

$\square$ Or, in (width) $\times($ avg.value $)$ form

$$
\hat{I}=(b-a) \frac{3\left[f\left(x_{0}\right)+3 \sum_{i=1,4,7 \ldots}^{n-2} f\left(x_{i}\right)+3 \sum_{j=2,5,8 \ldots .}^{n-1} f\left(x_{j}\right)+2 \sum_{k=3,6,9 \ldots}^{n-3} f\left(x_{k}\right)+f\left(x_{n}\right)\right]}{8 n}
$$

## Composite Simpson's 3/8 Rule - Example

Accuracy improves as the number of segments increases

- $\bar{f}^{(4)}$ over each segment decreases
- $f(x)$ better approximated as a cubic over smaller regions

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## Higher-Order Formulas

## Higher-Order Formulas

$\square$ Typically, Simpson's $1 / 3$ rule, used in conjunction with Simpson's $3 / 8$ rule (for odd $n$ ), is sufficient
$\square$ Possible to use higher-order polynomials to approximate $f(x)$
$\square n$ segments and $n+1$ points needed for $n^{t h}$-order polynomial approximation
$\square$ Closed and open integration formulas exist
$\square$ Boole's rule will show up in a different form later when we cover adaptive quadrature

## Higher-Order Newton-Cotes Formulas - Closed

| $n$ | Name |  | Error <br> prop. <br> to |
| :---: | :--- | :--- | :--- |
| 1 | Trapezoidal <br> rule | $\hat{I}=\frac{h}{2} \frac{f\left(x_{0}\right)+f\left(x_{1}\right)}{2}$ | $f^{\prime \prime}(\xi)$ |
| 2 | Simpson's <br> $1 / 3$ | $\hat{I}=\frac{h}{3}\left[f\left(x_{0}\right)+4 f\left(x_{1}\right)+f\left(x_{2}\right)\right]$ | $f^{(4)}(\xi)$ |
| 3 | Simpson's <br> $3 / 8$ | $\hat{I}=\frac{3 h}{8}\left[f\left(x_{0}\right)+3 f\left(x_{1}\right)+3 f\left(x_{2}\right)+f\left(x_{3}\right)\right]$ | $f^{(4)}(\xi)$ |
| 4 | Boole's rule | $\hat{I}=\frac{2 h}{45}\left[7 f\left(x_{0}\right)+32 f\left(x_{1}\right)+12 f\left(x_{2}\right)+32 f\left(x_{3}\right)+7 f\left(x_{4}\right)\right]$ | $f^{(6)}(\xi)$ |
| 5 | - | $\hat{I}=\frac{5 h}{288}\left[19 f\left(x_{0}\right)+75 f\left(x_{1}\right)+50 f\left(x_{2}\right)+50 f\left(x_{3}\right)+75 f\left(x_{4}\right)+19 f\left(x_{5}\right)\right]$ | $f^{(6)}(\xi)$ |

The step size in the above formulas is: $\quad h=\frac{(b-a)}{n}$

## Open Integration Formulas

$\square$ Function values not know at the limits of integration

- $n$ segments
$\square(n-1)$ points
- $(n-2)^{n d}$-order polynomial approximation



## Higher-Order Newton-Cotes Formulas - Open

| Segments <br> $(n)$ | Points | Formula | Error prop. <br> to |
| :---: | :---: | :--- | :---: |
| 2 | 1 | $\hat{I}=(b-a) f\left(x_{1}\right)$ | $f^{\prime \prime}(\xi)$ |
| 3 | 2 | $\hat{I}=(b-a) \frac{f\left(x_{1}\right)+f\left(x_{2}\right)}{2}$ | $f^{(4)}(\xi)$ |
| 4 | 3 | $\hat{I}=(b-a) \frac{2 f\left(x_{1}\right)+f\left(x_{2}\right)+2 f\left(x_{3}\right)}{3}$ | $f^{(4)}(\xi)$ |
| 5 | 4 | $\hat{I}=(b-a) \frac{11 f\left(x_{1}\right)+f\left(x_{2}\right)+f\left(x_{3}\right)+11 f\left(x_{4}\right)}{24}$ | $f^{(6)}(\xi)$ |
| 6 | 5 | $\hat{I}=(b-a) \frac{11 f\left(x_{1}\right)-14 f\left(x_{2}\right)+26 f\left(x_{3}\right)-14 f\left(x_{4}\right)+11 f\left(x_{5}\right)}{20}$ | $f^{(6)}(\xi)$ |

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Integration of Functions

## Integration of Functions

$\square$ Newton-Cotes formulas can be used to integrate functions or discrete data points

- Evenly-spaced data points are assumed
$\square$ If $f(x)$ is known, spacing of $x$-values can be chosen to improve accuracy
$\square$ Spacing need not be uniform
$\square$ Can locate points specific distances from limits of integration or segment edges to improve accuracy
$\square$ Can use larger step size where acceptable, reduced step size where necessary
■ Effectively trade off accuracy and efficiency


## Methods for integrating functions

$\square$ Romberg integration

- Combine two trapezoidal rule estimates with different step sizes to yield a third, more accurate estimate
$\square$ Gauss quadrature
- Spacing of points within the integration segments chosen to improve accuracy of Newton-Cotes formulas
$\square$ Adaptive Quadrature
- Adaptively refine step size to achieve desired accuracy
- Smaller step size in some regions, larger in others
- Uses some of the techniques used by Romberg integration


# Adaptive Quadrature 

## Adaptive Quadrature

$\square$ Vary step size to achieve desired accuracy over each segment

- Smaller step size where $f(x)$ varies rapidly
- Larger step size where $f(x)$ varies gradually
$\square$ Integration method used is Simpson's 1/3 rule



## Adaptive Quadrature

- Apply Simpson's $1 / 3$ rule to approximate the integral at two different step sizes, $\hat{I}\left(h_{1}\right)$ and $\hat{I}\left(h_{2}\right)$, where $h_{2}=h_{1} / 2$


$$
\hat{I}\left(h_{1}\right)=\frac{h_{1}}{3}[f(a)+4 f(c)+f(b)]
$$


$\hat{I}\left(h_{2}\right)=\frac{h_{2}}{3}[f(a)+4 f(d)+2 f(c)+4 f(e)+f(b)]$

## Adaptive Quadrature

$\square$ Use $\hat{I}\left(h_{1}\right)$ and $\hat{I}\left(h_{2}\right)$ to approximate the error:

$$
\begin{equation*}
E_{a}=\hat{I}\left(h_{2}\right)-\hat{I}\left(h_{1}\right) \tag{1}
\end{equation*}
$$

$\square$ Two possible ways to proceed:

- If $E_{a} \leq a b s t o l$
- Using an approach similar to Romberg integration, combine $\hat{I}\left(h_{1}\right)$ and $\hat{I}\left(h_{2}\right)$ to yield a third, more accurate estimate of the integral
- If $E_{a}>a b s t o l$
- Divide [a,b] into two segments: $[a, c]$ and $[c, b]$
- Calculate $\hat{I}\left(h_{1}\right)$ and $\hat{I}\left(h_{2}\right)$ for each segment
- Single- and double-segment Simpson's 1/3 approximations
- Use (1) to approximate the error for each sub-interval


## Adaptive Quadrature $-E_{a} \leq$ abstol

$\square$ If $E_{a}$ as calculated by (1) is acceptable (i.e. $<a b s t o l$ ) we can use $\hat{I}\left(h_{1}\right)$ and $\hat{I}\left(h_{2}\right)$ to calculate a third, more accurate approximation

- This is the basic principal used in Romberg integration
- We'll now derive the formula used to combine $\hat{I}\left(h_{1}\right)$ and $\hat{I}\left(h_{2}\right)$
$\square$ Each estimate is the true integral plus some error

$$
\begin{equation*}
I=\hat{I}\left(h_{1}\right)-E\left(h_{1}\right)=\hat{I}\left(h_{2}\right)-E\left(h_{2}\right) \tag{2}
\end{equation*}
$$

$\square$ We've seen that Simpson's $1 / 3$ rule error can be approximated as

$$
\begin{equation*}
E_{a}(h)=\frac{(b-a) h^{4}}{180} \bar{f}^{(4)} \tag{3}
\end{equation*}
$$

where $\bar{f}^{(4)}$ is the average value of $f^{(4)}(x)$ over the integration interval

## Adaptive Quadrature $-E_{a} \leq a b s t o l$

$\square$ Equation (3) gives approximate error at each step size:

$$
\begin{align*}
& E_{a}\left(h_{1}\right)=\frac{(b-a) h_{1}^{4}}{180} \bar{f}^{(4)}  \tag{4}\\
& E_{a}\left(h_{2}\right)=\frac{(b-a) h_{2}^{4}}{180} \bar{f}^{(4)} \tag{5}
\end{align*}
$$

$\square$ Divide (4) by (5)

$$
\begin{equation*}
\frac{E_{a}\left(h_{1}\right)}{E_{a}\left(h_{2}\right)}=\frac{h_{1}^{4}}{h_{2}^{4}} \tag{6}
\end{equation*}
$$

$\square$ Solve for $E_{a}\left(h_{1}\right)$

$$
\begin{equation*}
E_{a}\left(h_{1}\right)=E_{a}\left(h_{2}\right) \frac{h_{1}^{4}}{h_{2}^{4}} \tag{7}
\end{equation*}
$$

## Adaptive Quadrature $-E_{a} \leq a b s t o l$

$\square$ Restate (2) as an approximation

$$
\begin{equation*}
\hat{I}\left(h_{1}\right)-E_{a}\left(h_{1}\right) \approx \hat{I}\left(h_{2}\right)-E_{a}\left(h_{2}\right) \tag{8}
\end{equation*}
$$

$\square$ Substitute (7) into (8)

$$
\begin{equation*}
\hat{I}\left(h_{1}\right)-E_{a}\left(h_{2}\right) \frac{h_{1}^{4}}{h_{2}^{4}} \approx \hat{I}\left(h_{2}\right)-E_{a}\left(h_{2}\right) \tag{9}
\end{equation*}
$$

$\square$ Solve (9) for the error of the more accurate approximation

$$
\begin{equation*}
E_{a}\left(h_{2}\right)=-\left[\frac{\hat{I}\left(h_{1}\right)-\hat{I}\left(h_{2}\right)}{1-\left(\frac{h_{1}}{h_{2}}\right)^{4}}\right] \tag{10}
\end{equation*}
$$

## Adaptive Quadrature $-E_{a} \leq a b s t o l$

$\square$ According to (2)

$$
\begin{equation*}
I=\hat{I}\left(h_{2}\right)-E\left(h_{2}\right) \tag{11}
\end{equation*}
$$

so

$$
\begin{equation*}
I \approx \hat{I}=\hat{I}\left(h_{2}\right)-E_{a}\left(h_{2}\right) \tag{12}
\end{equation*}
$$

$\square$ Substituting (10) into (12)

$$
\hat{I}=\hat{I}\left(h_{2}\right)+\frac{\hat{I}\left(h_{1}\right)-\hat{I}\left(h_{2}\right)}{1-\left(\frac{h_{1}}{h_{2}}\right)^{4}}
$$

$\square$ And, since $h_{1}=2 \cdot h_{2}$, the integral approximation is

$$
\begin{equation*}
\hat{I}=\hat{I}\left(h_{2}\right)+\frac{1}{15}\left[\hat{I}\left(h_{2}\right)-\hat{I}\left(h_{1}\right)\right] \tag{13}
\end{equation*}
$$

$\square$ Which can be shown to be equivalent to Boole's rule

## Adaptive Quadrature $-E_{a} \leq a b s t o l$

$\square$ Summarizing:

- Want to numerically integrate $f(x)$ over $[a, b]$
$\square$ Calculate $\hat{I}\left(h_{1}\right)$ and $\hat{I}\left(h_{2}\right)$
$\square$ Approximate the error as

$$
E_{a}=\hat{I}\left(h_{2}\right)-\hat{I}\left(h_{1}\right)
$$

- If $E_{a} \leq a b s t o l$, calculate the integral using Boole's rule

$$
\hat{I}=\hat{I}\left(h_{2}\right)+\frac{1}{15}\left[\hat{I}\left(h_{2}\right)-\hat{I}\left(h_{1}\right)\right]
$$

- Next, we'll look at what to do if $E_{a}>a b s t o l$


## Adaptive Quadrature $-E_{a}>$ abstol

$\square$ If $E_{a}>a b s t o l$, reduce the step size and try again
$\square$ Subdivide the integration interval, $[a, b]$, into two subintervals, $[a, c]$ and $[c, b]$

- For each subinterval:
- Calculate $\hat{I}\left(h_{1}\right)$ and $\hat{I}\left(h_{2}\right)$
- Calculate $E_{a}$
- If $E_{a} \leq a b s t o l$, calculate $\hat{I}$ for that sub-interval using Boole's rule
- If $E_{a}>a b s t o l$, further subdivide the sub-interval into two smaller sub-intervals
- Calculate $\hat{I}\left(h_{1}\right)$ and $\hat{I}\left(h_{2}\right)$, then $E_{a} \ldots$
- Eventually, total integral approximation is the sum of all individual sub-interval integral approximations


## Adaptive Quadrature - Recursive Algorithm

$\square$ Adaptive quadrature uses a recursive algorithm
$\square$ A function that calls itself
$\square$ Integration interval is continually subdivided until approximate error is acceptable
$\square \hat{I}$ returned by function is the sum of the individual $\hat{I}$ values

## Adaptive Quadrature - quadadapt()



Inputs: function handle, limits of integration, and tolerance

- On subsequent recursive calls, a and $b$ will be sub-interval limits
$\square$ Step sizes and $x$-values for two Simpson's $1 / 3$ rule estimates

Integral estimates at two different step sizes

Approximate error
Boole's rule estimate
Recursive function calls
Sum the sub-interval integral estimates

## Adaptive Quadrature - Examples



## 60 <br> Integrating Functions in Python

## Integrating Functions - integrate.quad()

$\square$ When we have an expression for the function to be integrated, we can use SciPy's integrate. quad() function:

## I, err = integrate.quad(f, $a, b)$

- f: the function to be integrated
- a: lower integration limit
- b: upper integration limit
- I: numerical approximation of the integral
- err: approximate absolute error
- Calculates $I=\int_{a}^{b} f(x) d x$


## Exercise - Integration in Python

$\square \quad$ The impulse response of a certain $2^{\text {nd }}$-order system is given by

$$
h(t)=5.2414 e^{-\alpha t} \sin \left(\omega_{d} t\right)
$$

where $\alpha=1.5$ and $\omega_{d}=4.7697 \mathrm{rad} / \mathrm{sec}$
$\square$ A system's step response is the integral of its impulse response. For this system, the step response is

$$
g(t)=1-e^{-\alpha t} \cos \left(\omega_{d} t\right)-0.3145 e^{-\alpha t} \sin \left(\omega_{d} t\right)
$$

$\square$ Plot $g(t)$ for $0 \leq t \leq 10 \mathrm{sec}$ using a small sampling interval (e.g. 1msec)
$\square$ For a variety of step sizes (e.g. 500, 200, 100, 10, 1msec)

- Calculate $\hat{g}(t)$ using cumulative_trapezoid () and superimpose on the plot of $g(t)$
- Calculate the steady-state value of the step response using trapezoid()
- Notice the effect of step size on the accuracy of the integral

