

# Week 2

Numerical Integration

# Finite differences

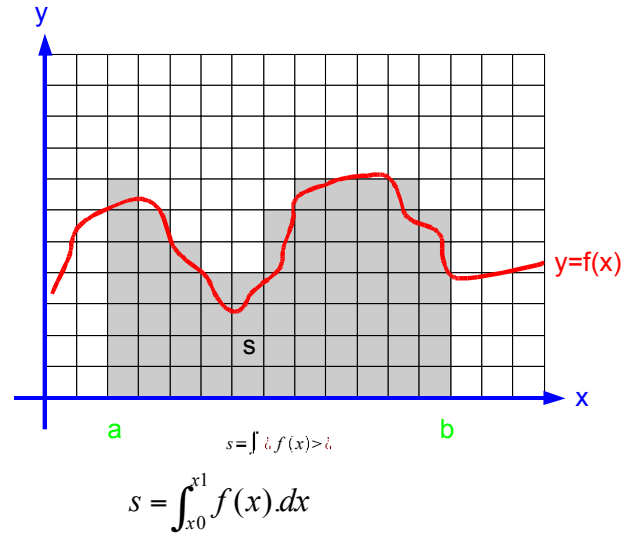
- Taylor expansion:
- $f(x_0+h) = f(x_0) + f'(x_0) h + f''(x_0) h^2/2! + \dots$
- $f(x_0+h) - f(x_0) = f'(x_0) h + f''(x_0) h^2/2! + \dots$
- $[f(x_0+h) - f(x_0)]/h = f'(x_0) + f''(x_0) h/2! + \dots$
  
- Approximation good to  $O(h)$

# Finite differences

- Taylor expansion: MIDPOINT
- $f(x_0 + h/2) = f(x_0) + f'(x_0) h/2 + f''(x_0) (h/2)^2/2! + \dots$
- $f(x_0 - h/2) = f(x_0) + f'(x_0) (-h/2) + f''(x_0) (-h/2)^2/2! + \dots$
- $f(x_0 + h/2) - f(x_0 - h/2) = f'(x_0)h + f'''(x_0) (h/2)^3 2/3! + \dots$
- $[f(x_0 + h/2) - f(x_0 - h/2)]/h = f'(x_0) + f'''(x_0) (h/2)^2 /3!$
- Approximation good to  $O(h^2)$

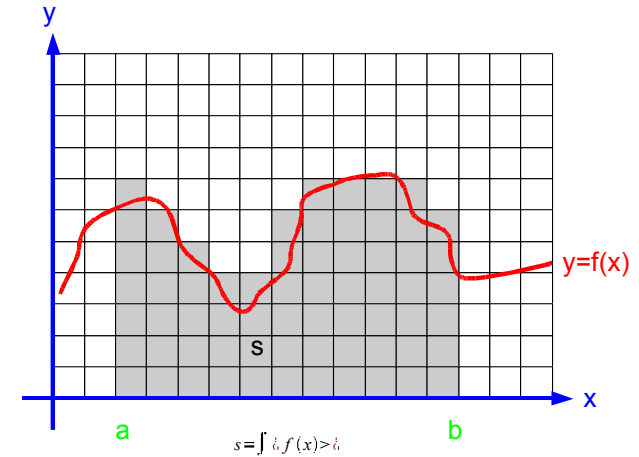
# Numerical Integration

- Finding the area under a curve
- An alternative to analytical solutions (i.e. doing the maths)
  - When a formula can't be symbolically integrated
  - When it is computationally cheaper to evaluate numerically than analytically
  - When a formula isn't available – only numerical data



# Numerical Integration

- Finding the area under a curve
- An alternative to analytical solutions (i.e. doing the maths)
- Simplest: N panels each width h
- Approximate as rectangle



$$s = \int_a^b f(x) \cdot dx = \sum_{n=0}^{N-1} \int_{x_n}^{x_{n+h}} f(x) \cdot dx$$

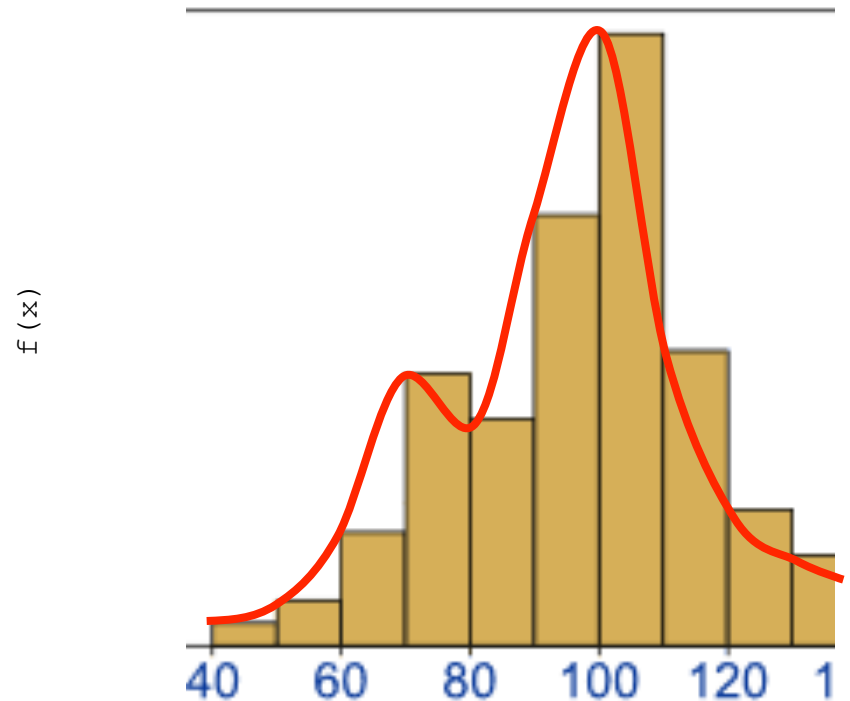
$$\approx \sum_{n=0}^{N-1} [f(x_n) + f'(x_n)h + f''(x_n)h^2 / 2 + \dots]h$$

$$\approx \sum_{n=0}^{N-1} f(x_n)h + O(h^2)$$

# Rectangles

$$\approx \sum_{n=0}^{N-1} f(x_n)h + O(h^2)$$

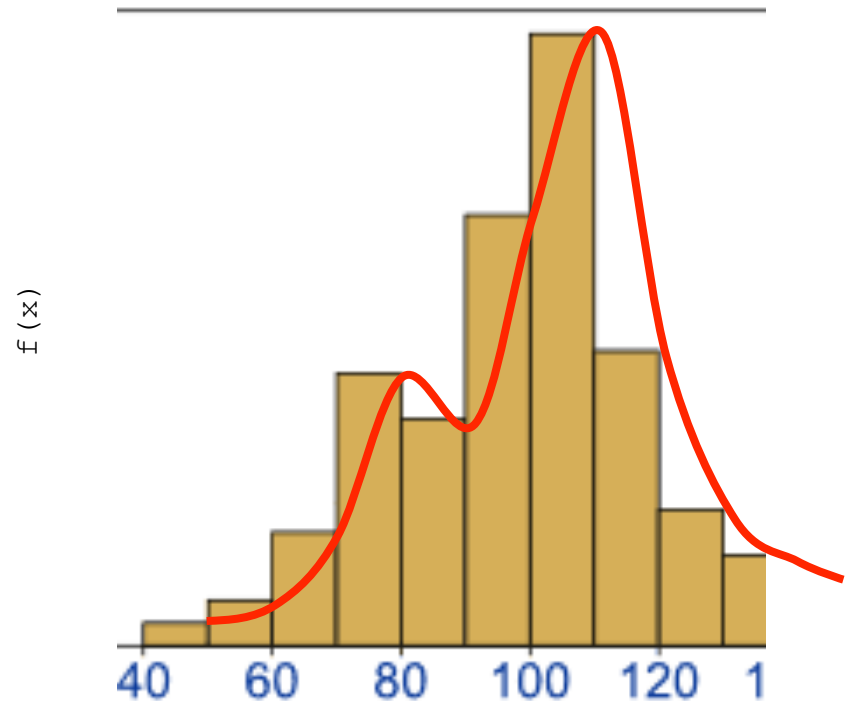
- Divide the function into a series of rectangular panels
- This is the simplest way. Height of the rectangle set by function value at start (left point)



# Rectangles

$$\approx \sum_{n=1}^N f(x_n)h + O(h^2)$$

- Divide the function into a series of rectangular panels
- This is the simplest way. Height of the rectangle set by function value at right hand point



# Rectangles – MIDPOINT

$$s = \int_{x_0}^{x_1} f(x) \cdot dx$$

$$= \sum_{n=1}^{N-1} \int_{x_{n-h/2}}^{x_{n+h/2}} f(x) \cdot dx$$

$$= \sum_{n=1}^{N-1} \int_{x_{n-h/2}}^{x_n} f(x) \cdot dx + \sum_{n=1}^{N-1} \int_{x_n}^{x_{n+h/2}} f(x) \cdot dx$$

$$\approx \sum_{n=1}^{N-1} [f(x_n) + f'(x_n)(-h/2) + f''(x_n)(-h/2)^2 / 2 + \dots] h / 2 +$$

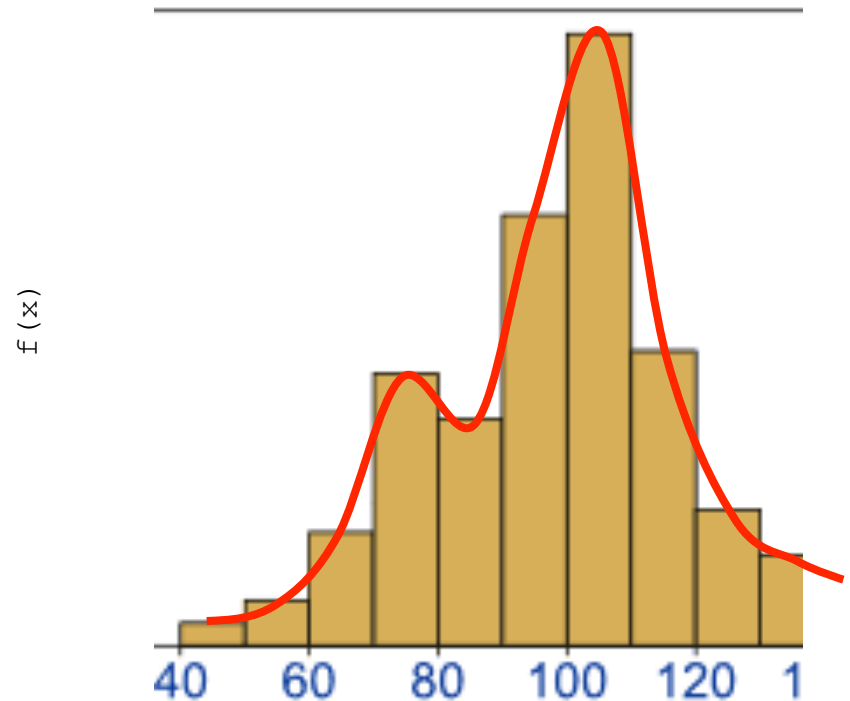
$$+ [f(x_n) + f'(x_n)(h/2) + f''(x_n)(h/2)^2 / 2 + \dots] h / 2$$

$$= \sum_{n=1}^{N-1} [f(x_n) + f''(x_n)(h/2)^2 / 2 + \dots] h \approx \sum_{n=1}^{N-1} f(x_n) h + O(h^3)$$



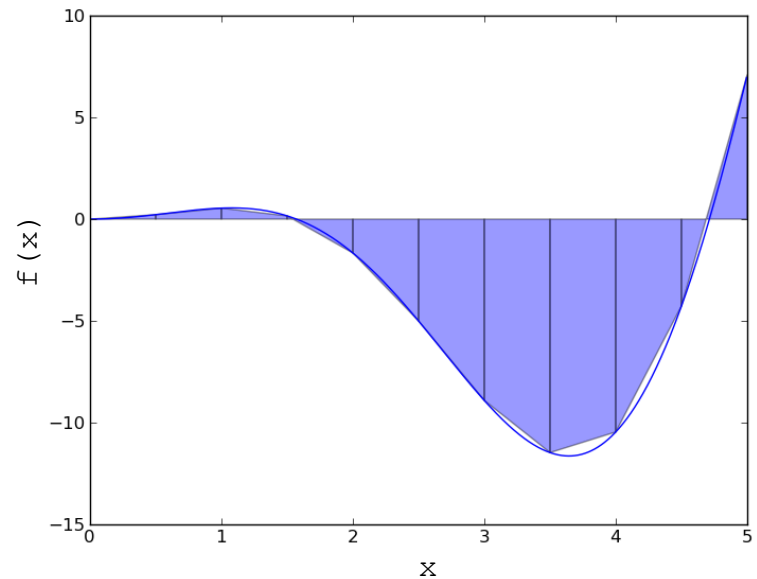
# Rectangles

- Divide the function into a series of rectangular panels
- This is the simplest way
- And with midpoint its same number of calculations but better accuracy!



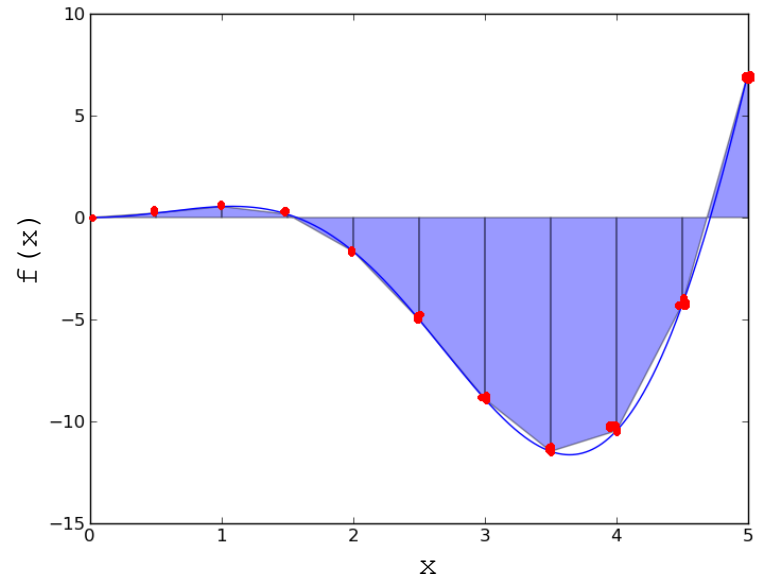
# Trapezium Rule

- Instead of rectangles, use trapeziums
- I.e. use first order derivative information



# Computational Cost

- 2 function evaluations per panel
- But edges are shared
  - 1 per panel + 1
- More accurate than rectangles for no extra function evaluations



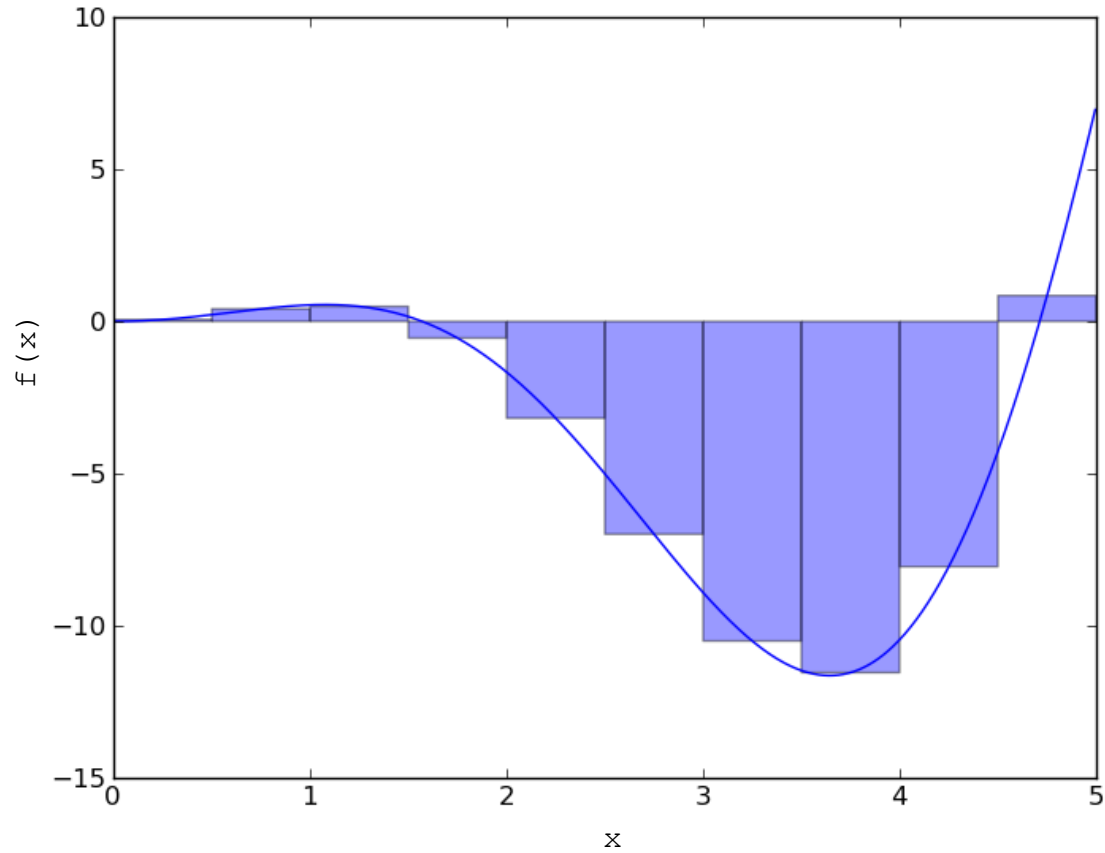
# What's going on?

- We are fitting analytical expressions to each panel of our function
- $N^{\text{th}}$  order Lagrange polynomial expansions
- We then analytically integrate these small chunks

Rule	Expression
Rectangle	$y = k_0$
Trapezium	$y = k_0 + k_1x$

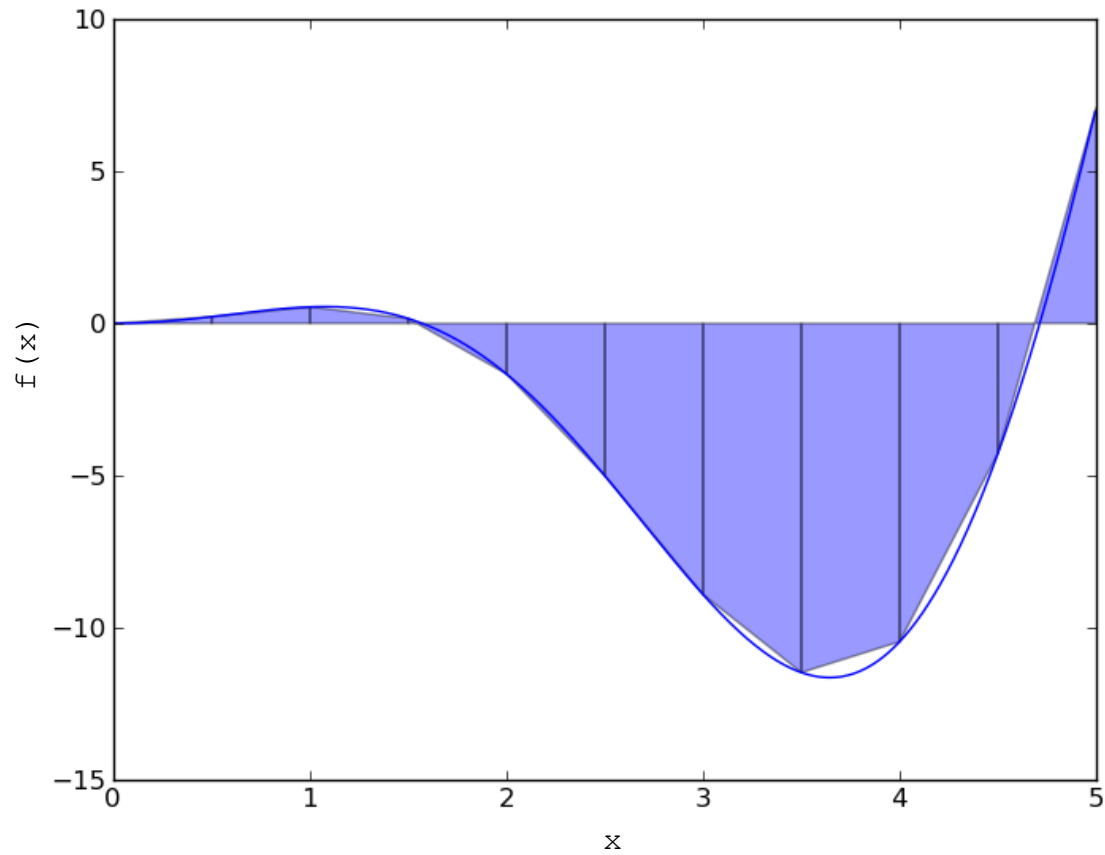
# Midpoint rule (rectangle)

Panel area :  $y = k_0 + k_1x + k_2x^2 + k_3x^3$



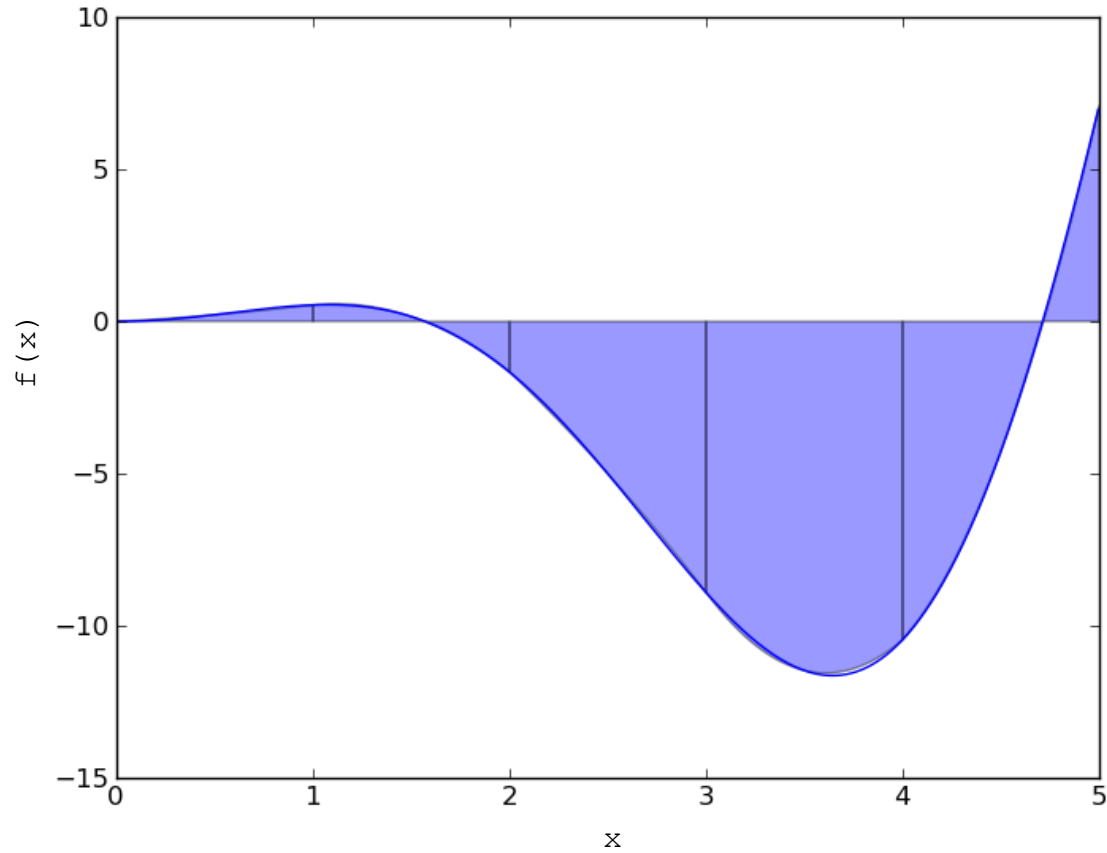
# Trapezium Rule (GCSE!)

Panel area :  $y = k_0 + k_1x + k_2x^2 + k_3x^3$



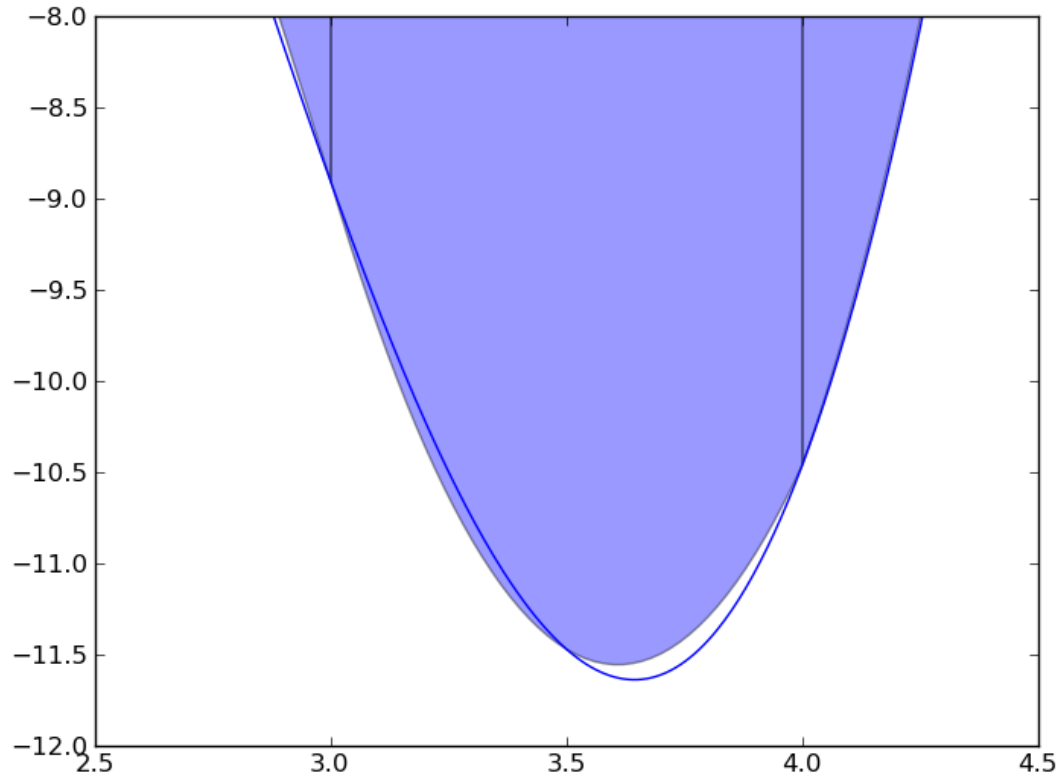
# Simpson's Rule

Panel area :  $y = k_0 + k_1x + k_2x^2 + k_3x^3$



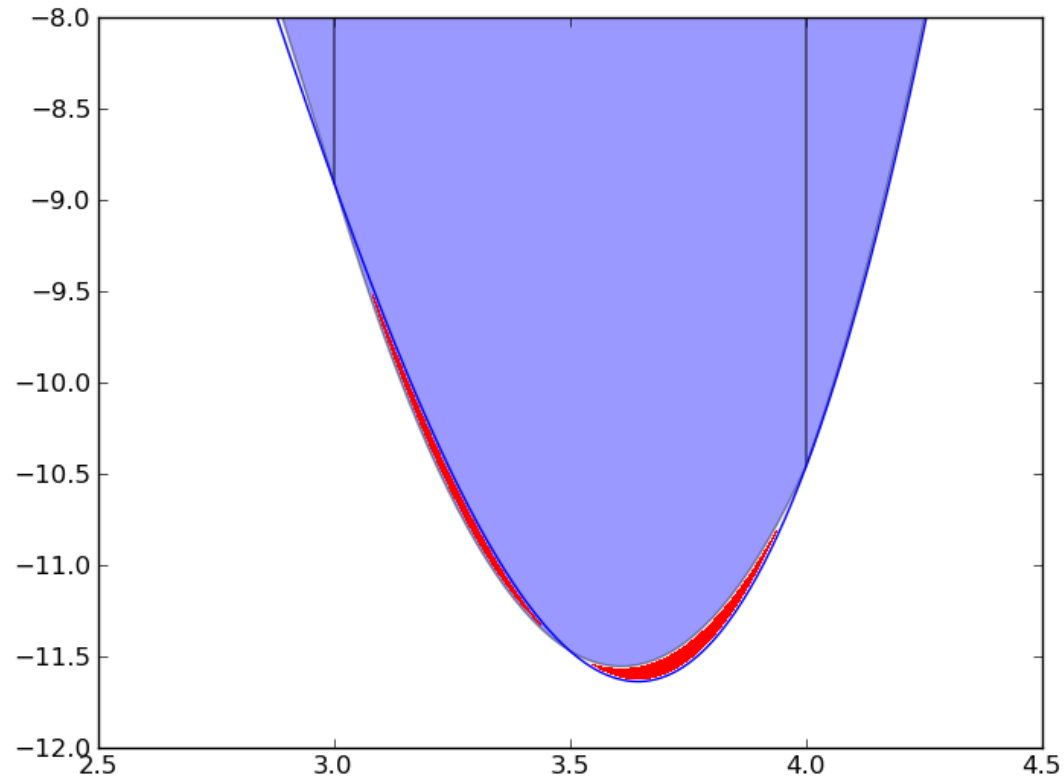
# Simpson's Rule

Panel area :  $y = k_0 + k_1x + k_2x^2 + k_3x^3$





# Simpson's Rule



# Simpson's Rule - formula

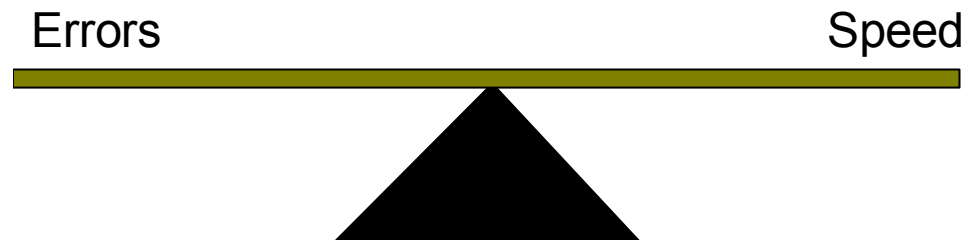
- Use quadratic information – second derivative
- Panel  $a \leq x \leq b$
- $m = (a + b) / 2$

*m for middle!*

$$\int_b^a f(x) = \frac{b-a}{6} (f(a) + 4.f(m) + f(b))$$

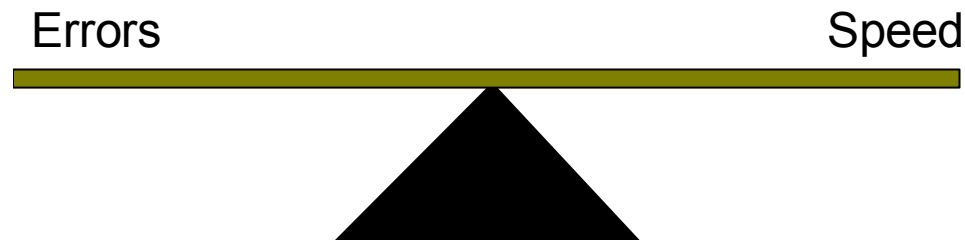
# It's all about the balance

- In real world use, computing the function calls costs time – complicated functions!
- You need some desired level of accuracy
- The choice of algorithm makes more difference than the panel size



# It's all about the balance

- In real world use, computing the function calls costs time – complicated functions!
- You need some desired level off accuracy
- The choice of algorithm makes more difference than the panel size



- How accurate do you need your answer?

# Error scaling

Method	Order	Panel area formula	Function evaluations	Error (order)
Rectangle (midpoint)	0	$(b - a)f(m)$	N	$2(b-a)^3$
Trapezium	1	$\frac{(b - a)}{2} [f(a) + f(b)]$	N+1	$(b-a)^3$
Simpson	2	$\frac{(b - a)}{6} [f(a) + 4f(m) + f(b)]$	2N+1	$(b-a)^5$

$$b - a \propto \frac{1}{N}$$

**So doubling the number of panels decreases the error:**

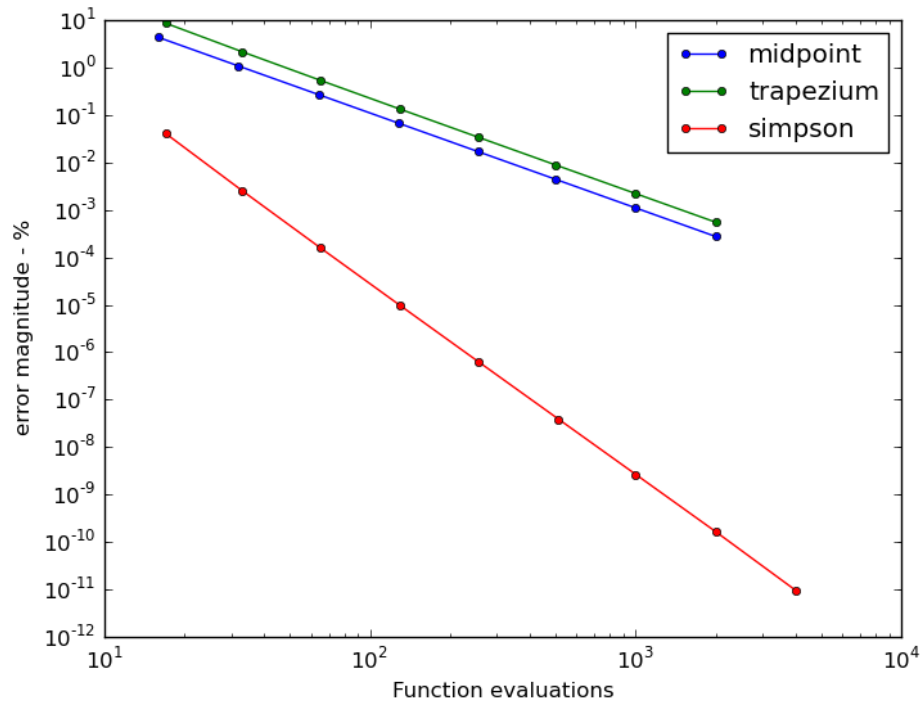
**Rectangle – 8x**

**Trapezium – 8x**

**Simpson – 32x**

# Accuracy vs. computational cost

$$\int_0^4 x^2 \sin(x)$$



**Simpson's rule is the clear winner – higher order methods are even better, but are rarely needed**

# Higher order methods

- Simpson's  $3/8$ 's rule
- Boole's rule
- Any higher order you want
  
- Generally, Simpson's rule is enough

Making some code...



# A Practical Algorithm

- Let's code an integrator with the **midpoint** rule
- Weekly assessment is to code an integrator with Simpson's rule
- Find the definite integral of  $f(x)$  between  $x_0$  and  $x_1$
- Let's use 5 panels

$$area = \int_{x_0}^{x_1} f(x).dx$$

# Start with the equation

- $S = h \cdot f_1 + h \cdot f_2 + h \cdot f_3 + h \cdot f_4 + h \cdot f_5$
- 5 function evaluations
- 5 multiplies

# Factorise

- $S = h^*(f_1 + f_2 + f_3 + f_4 + f_5)$
- 5 function evaluations
- 1 multiply
- Potentially less rounding errors

# Specify panel width?

- Your integration function needs to decide on a panel width.
- We could tell the code to use a specific width, *panel\_width*, but depending on the integration range we may not get an integer number of panels
- E.g. integrate  $0 \leq x \leq 1.2$  with a *panel\_width* of 0.5
  - *Blackboard example*
- We would have to add some more code to handle this ‘special case’ (e.g. use a different width final panel)
  - **It’s ‘special case’ code that makes most of the bugs!**

# Specify number of panels?

- Instead we could specify the number of panels to use,  $N_{panels}$
- The code then computes
  - $panel\_width = (x1-x0)/N_{panels}$
- Now we know that the panels always fit the integration range – no special case code needed.

# Specify number of panels?

```
from __future__ import division
```

```
import numpy
```

```
def f(x):
```

```
    return x**4
```

# Specify number of panels?

```
def integrate_rect(a,b,n_panels):  
    h=(b-a)/n_panels  
    func_sum=0.0  
    for ix in range(n_panels):  
        x=a+ix*h+h/2 # not x=x+h as cumulative  
        func_sum=func_sum+f(x)  
  
    return func_sum*h #at end so only do it once
```

# Specify number of panels?

```
a=0
```

```
b=2
```

```
num= integrate_rect(a,b,100)
```

```
#test the code using the analytic solution
```

```
ana=(b**5)/5-(a**5)/5
```

```
print num, ana, (num-ana)/ana
```



```
demo0.py - G:\teaching\2010-2011\CompPhys\02 numerical integratio...
File Edit Format Run Options Windows Help
demo0.py - G:\teaching\2010-2
from __future__ import division

import numpy

def f(x):
    return x**4

# Variables

# a - left (x-axis) of a panel
# b - right (x-axis) of a panel
# m - middle (x-axis) of a panel

def integrate_rect(x0, x1, n_panels):
    ''' Integrate the function f between x0 and x1 '''
    # Split the interval x0 <= x <= x1 into panels
    panel_width = (x1-x0) / n_panels

    # Sum of f(0) + f(1) + f(2) + ...
    func_sum = 0

    for ix in range(n_panels):
        # Find the left edge of this panel
        a = x0 + ix * panel_width
        # Find the midpoint
        m = a + panel_width / 2
        func_sum += f(m)

    return panel_width * func_sum

x0, x1 = 0, 2
print integrate_rect(x0, x1, 100)
# Analytical solution is x**5/5
print (x1**5/5) - (x0**5/5)
```

**Test code**

**Outputs are**

**6.399466676**

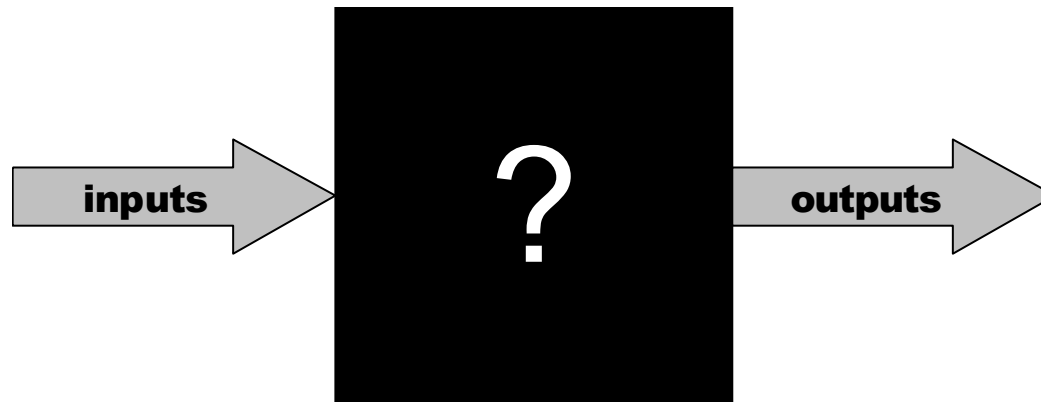
**and**

**6.4**

Black Box integrators

# Black Box

- ‘black box’ code is some third party module
- You know how to use it (*API Documentation*)
- Perhaps you don’ t know or care about the details of how it works
  - *Caveat Emptor*
  - *Brain rot!*



# scipy.integrate

```
Python Shell
File Edit Shell Debug Options Windows Help
>>> import scipy
>>> import scipy.integrate
>>> help (scipy.integrate)
Help on package scipy.integrate in scipy:

NAME
    scipy.integrate

FILE
    d:\lang\python25\lib\site-packages\scipy\integrate\__init__.py

DESCRIPTION
    Integration routines
    =====

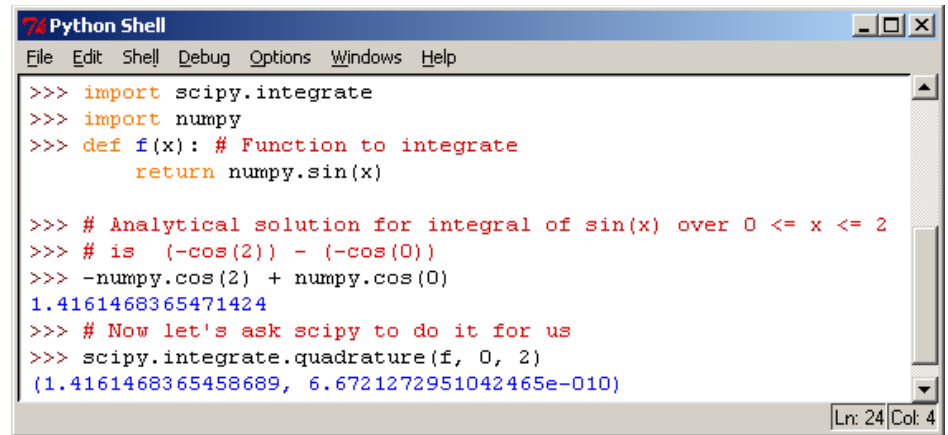
    Methods for Integrating Functions given function object.

    quad          -- General purpose integration.
    dblquad       -- General purpose double integration.
    tplquad       -- General purpose triple integration.
    fixed_quad    -- Integrate func(x) using Gaussian quadrature of order n.
    quadrature    -- Integrate with given tolerance using Gaussian quadrature.
    romberg       -- Integrate func using Romberg integration.
```

Ln: 689 | Col: 4

# Example

- Let's integrate  $\sin(x)$
- Simple function so we can also do this analytically

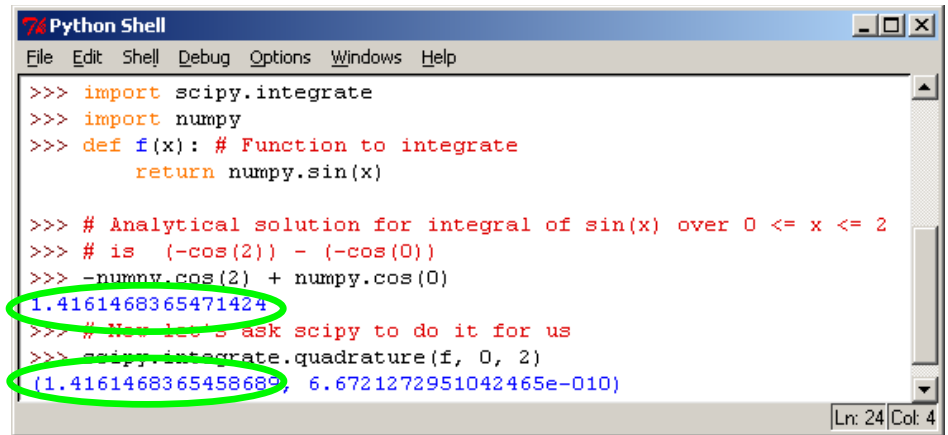


```
Python Shell
File Edit Shell Debug Options Windows Help
>>> import scipy.integrate
>>> import numpy
>>> def f(x): # Function to integrate
        return numpy.sin(x)

>>> # Analytical solution for integral of sin(x) over 0 <= x <= 2
>>> # is (-cos(2)) - (-cos(0))
>>> -numpy.cos(2) + numpy.cos(0)
1.4161468365471424
>>> # Now let's ask scipy to do it for us
>>> scipy.integrate.quadrature(f, 0, 2)
(1.4161468365458689, 6.6721272951042465e-010)
Ln: 24 Col: 4
```

# Example

- Let's integrate  $\sin(x)$
- Simple function so we can also do this analytically



```
Python Shell
File Edit Shell Debug Options Windows Help
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```

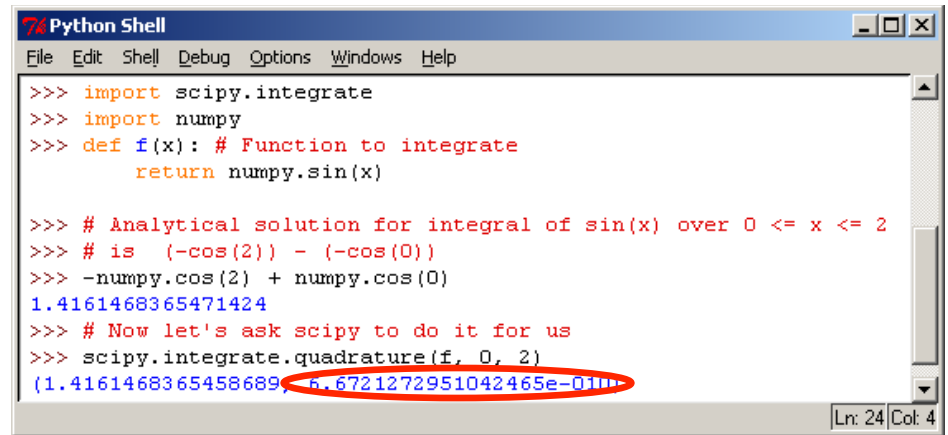
**SciPy is correct! – to ten decimal places anyway**

**It's always a good idea to compare third party *black box* code to a simple analytical case:**

- 1. This checks that their module isn't completely broken!**
- 2. It checks that you are using their module correctly**

# Example

- Let's integrate  $\sin(x)$
- Simple function so we can also do this analytically



```
Python Shell
File Edit Shell Debug Options Windows Help
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>>> def f(x): # Function to integrate
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>>> # Now let's ask scipy to do it for us
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(1.4161468365458689, 5.6721272951042465e-011)
```

SciPy gives us an error / accuracy value

But how does SciPy know this for any function in the absence of an analytical solution?

**Tolerance analysis**

# Tolerance driven approach

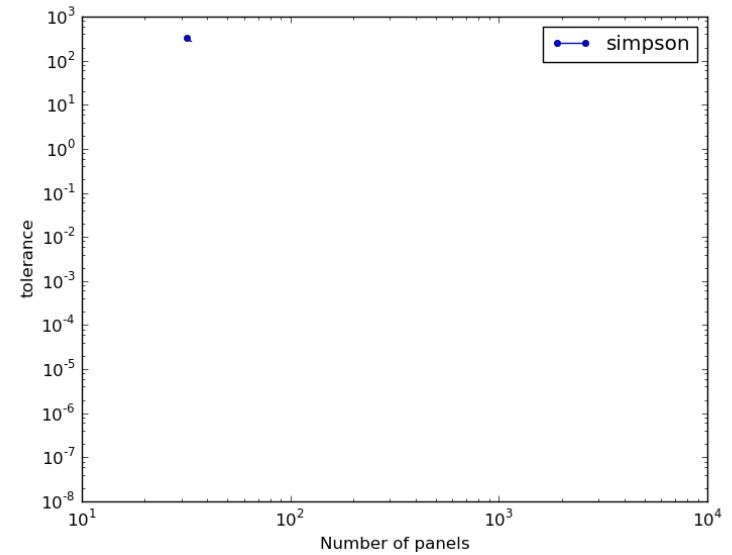
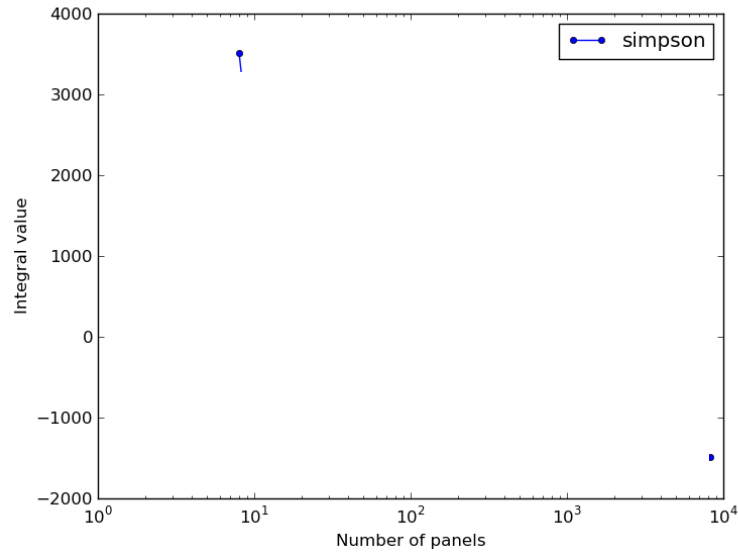
- Many third party integrators work to deliver a certain ‘tolerance’
- Tolerance – what is the change in the computed value if the number of panels is doubled?
- The tolerance asymptotically approaches zero as the error reduces with increasing step size
- This allows a known accuracy to be reached in the absence of an analytical solution (i.e. real problems!)



# Tolerance example

Evaluate for N panels

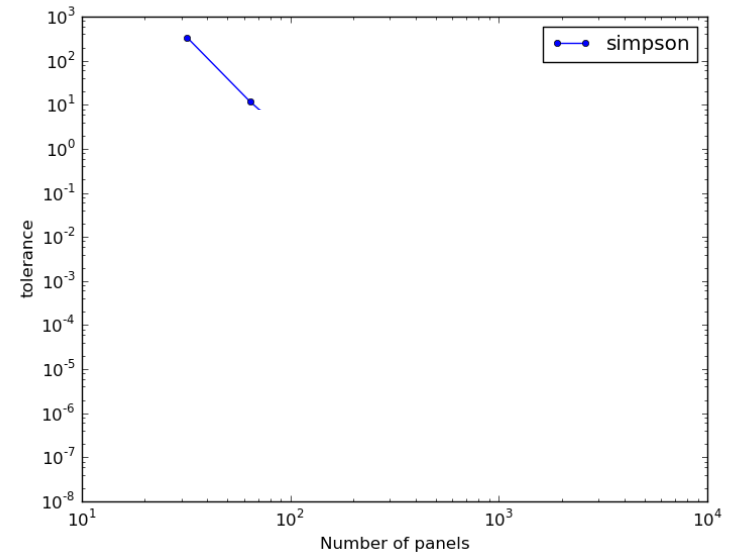
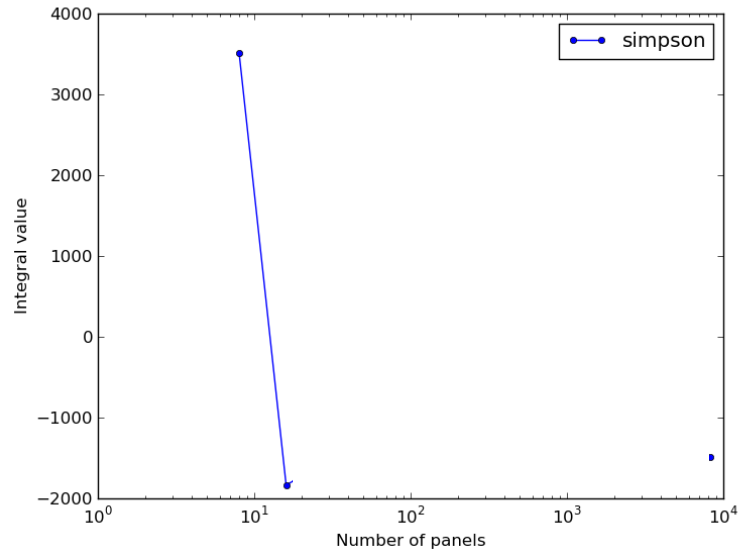
$$\int_0^{64} x^2 \sin(x) dx$$



# Tolerance example

Evaluate for N panels

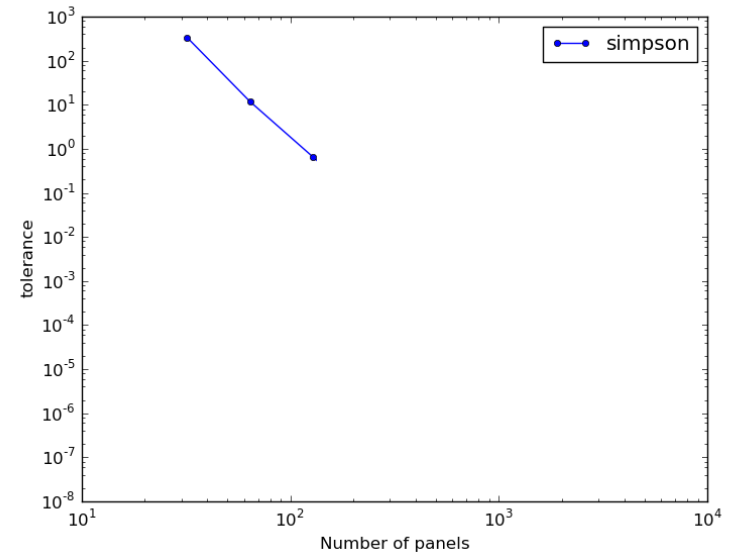
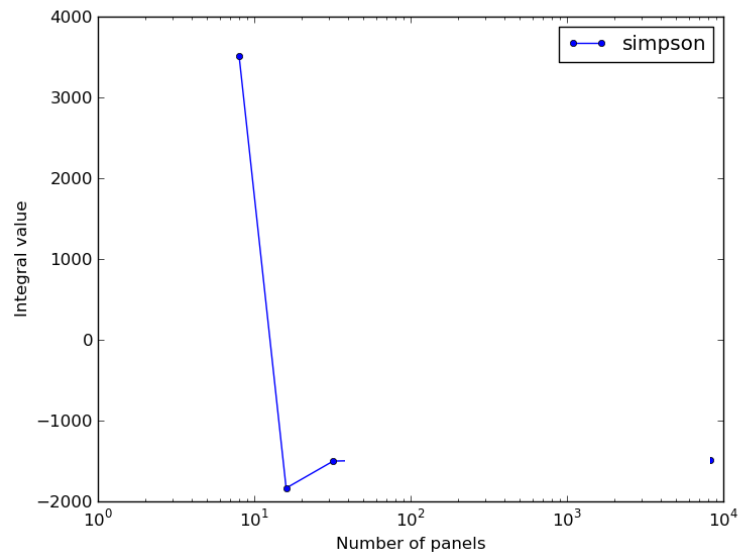
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# Tolerance example

Evaluate for N panels

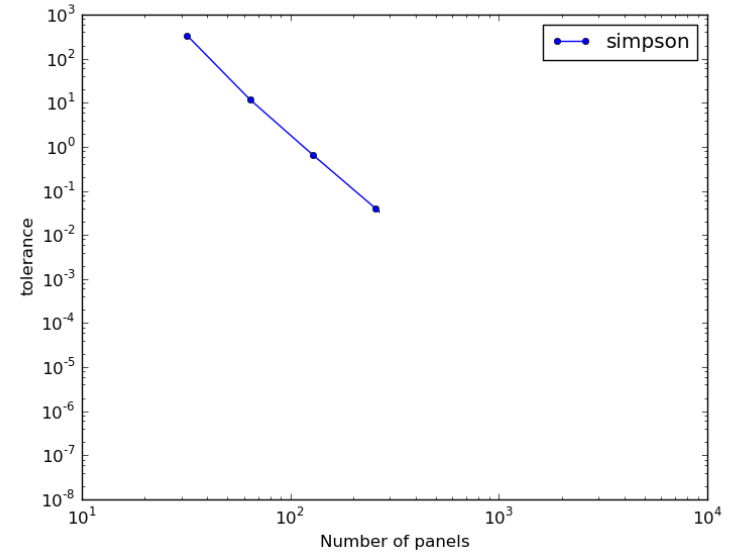
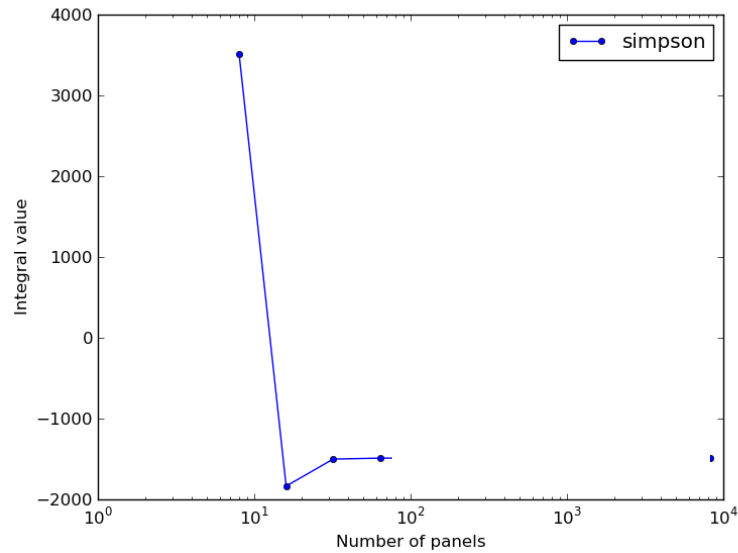
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# Tolerance example

Evaluate for N panels

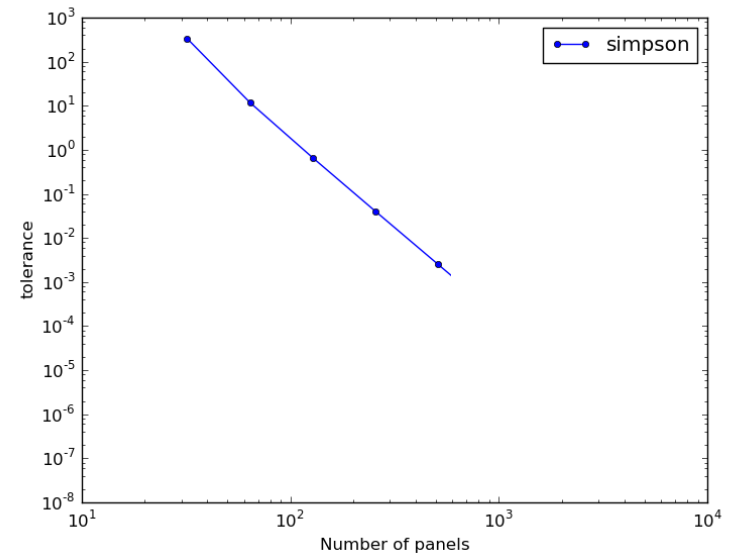
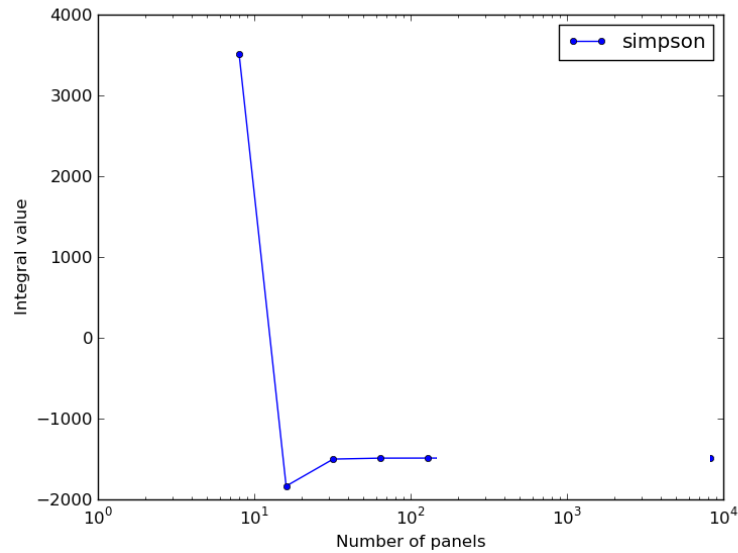
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# Tolerance example

Evaluate for N panels

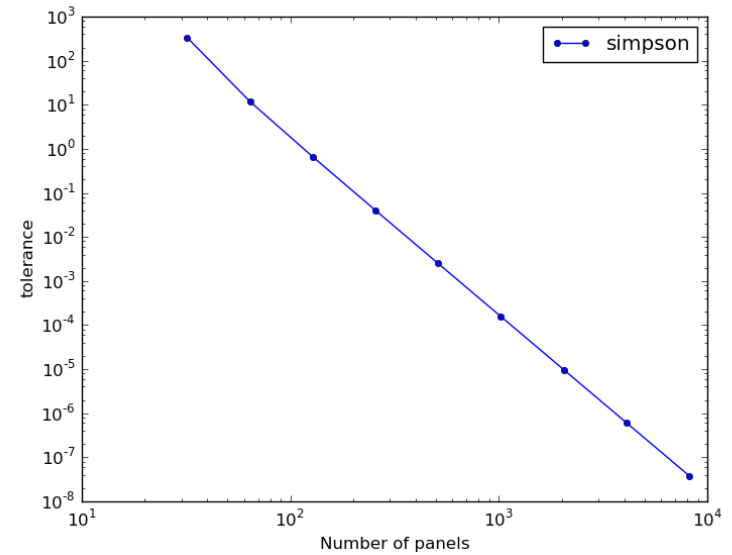
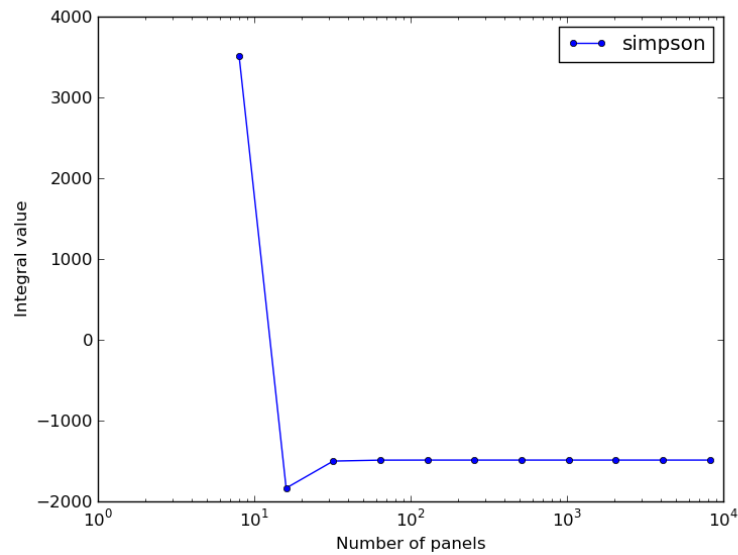
$$\int_0^{64} x^2 \sin(x) dx$$



# Tolerance example

Evaluate for N panels

$$\int_0^{64} x^2 \sin(x) dx$$



# Stretch exercises

- Derive Simpson's rule
  - On paper, fit a 2<sup>nd</sup> order polynomial to the left-, mid- and right- points of a panel;  $f(a)$ ,  $f(m)$ ;  $f(b)$
  - Integrate this polynomial fit
- Build a tolerance driven integrator:
- What happens if you make panel width too small?