Numerical Integration PHYS4840

Trapz, Simpsons, Romberg, errors

github.com/nialljmiller/PHYS4840_Numerical_integration_I

(This is currently the pinned repo on my page)

Dr Niall Miller

Office hours - 2 hours before class or by request

Numerical Integration

- Why do we need numerical integration methods?
- Why is there more than one?
- Why is my boss mad at me for using trapz?
- Cant AI do this?

"There is no known way to calculate the area under a curve exactly in all cases on a computer" - Mark Newman

Numerical Integration

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Computers will never integrate

- Computers may integrate into society one day
- However, as computers can only ever speak in a discretized way, they won't ever be able to compute pure integration - or anything in a truly continuous way
- Everything a computer does to interface with the real world is based on this

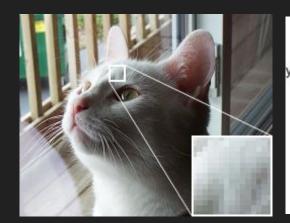


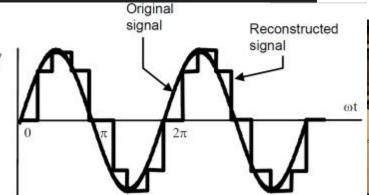
Computers will never integrate

- This is a fundamental limitation of computing

All the visuals we see rendered by a computer are quantized into pixels.

Any audio we hear from a computer must first go through a Digital to Analogue Converter.



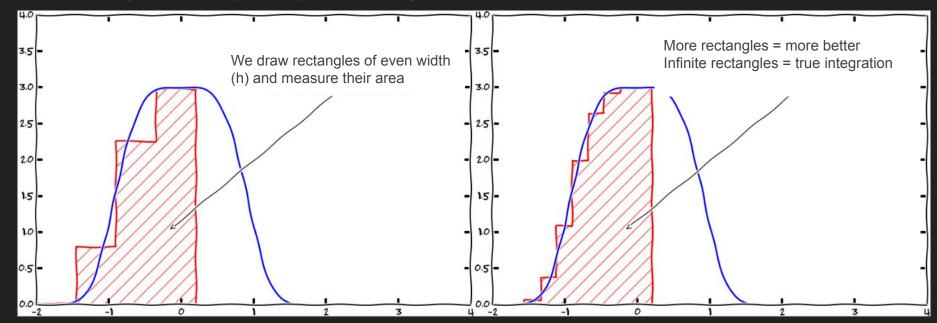




EVALUATING INTEGRALS

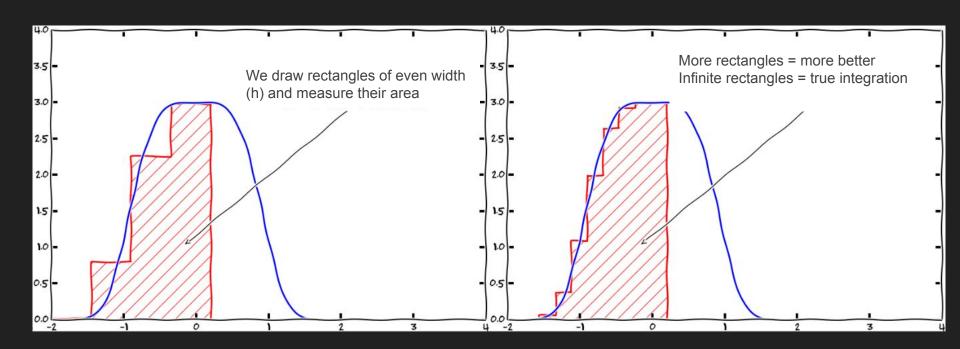
So we approximate integration

- Computers may never be able to comprehend the beauty of the analogue world
- But they are really really fast and good at discrete maths.



So we approximate integration

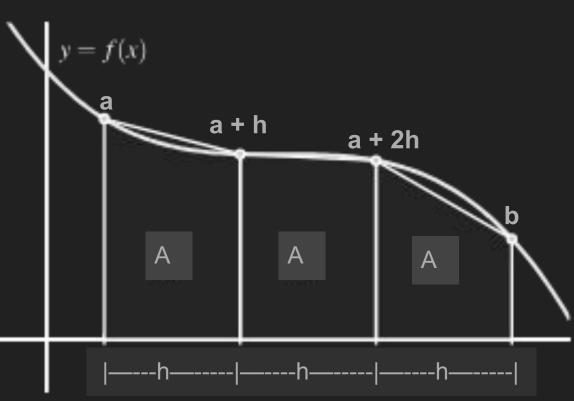
So what if we want to find the area under a continuously defined curve (f(x))?



Slight improvement to the rectangles (this is foreshadowing)

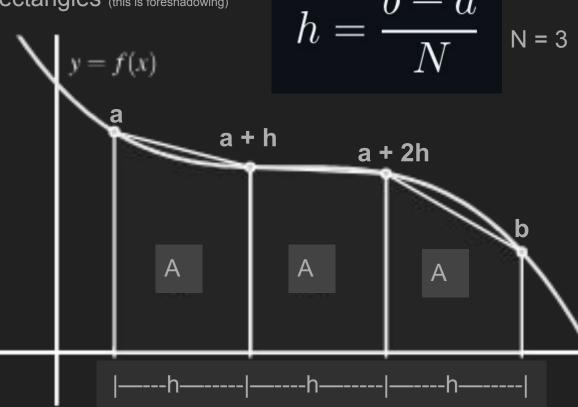
Now for some Maths

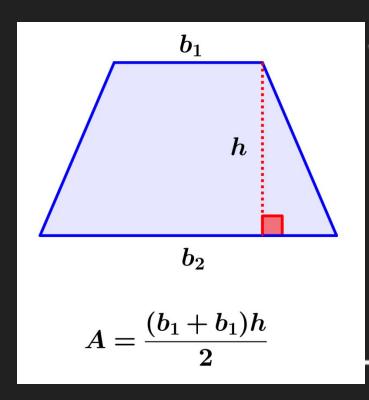
If we want to find the area between \mathbf{a} and \mathbf{b} We can split this into equal length trapezoids h = width of the trapezoidWe can calculate the area of each trapezoid
The area of $f(x) \sim \text{Sum of trapezoid areas}$

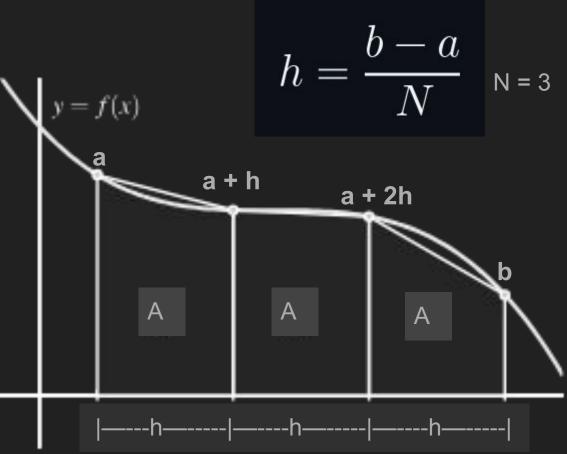


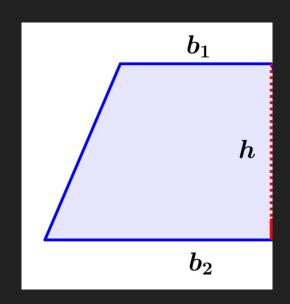
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Now for some Maths

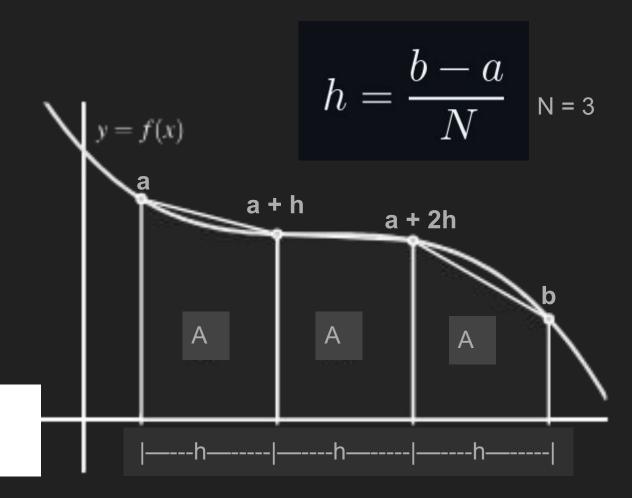


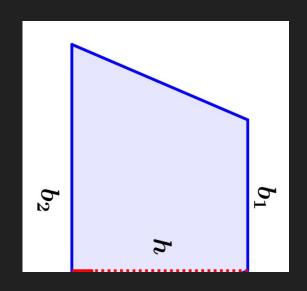




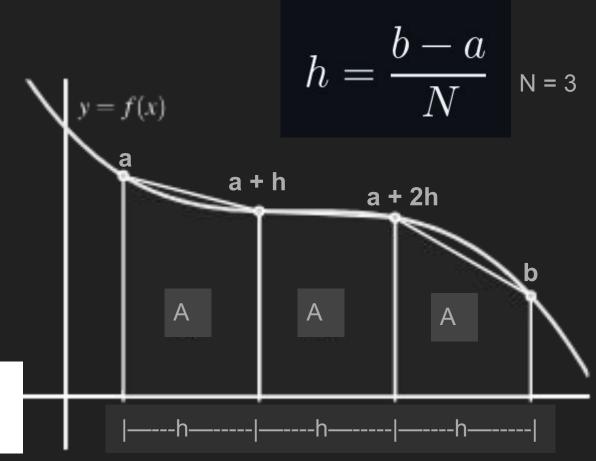


$$A=\frac{(b_1+b_1)h}{2}$$





$$A=\frac{(b_1+b_1)h}{2}$$



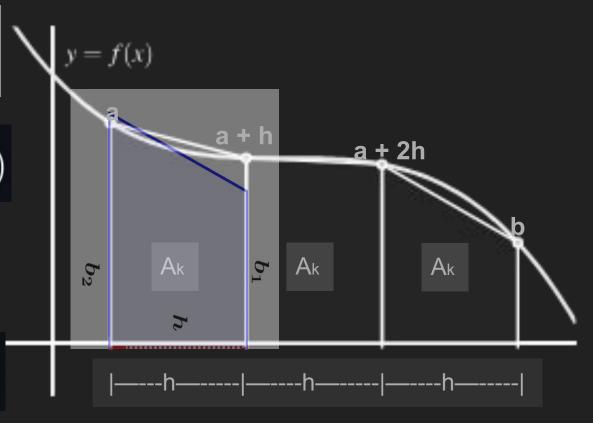
 $Area = \frac{1}{2}$ height [length of two parallel lines added together]

$$A=\frac{(b_1+b_1)h}{2}$$

b1 =
$$f(a+h(k-1))$$

b2 =
$$f(a+hk)$$

$$A_k = \frac{1}{2} h \left[f(a+h(k-1)) + f(a+hk) \right]$$



Summing over all subintervals from k = 1 to N:

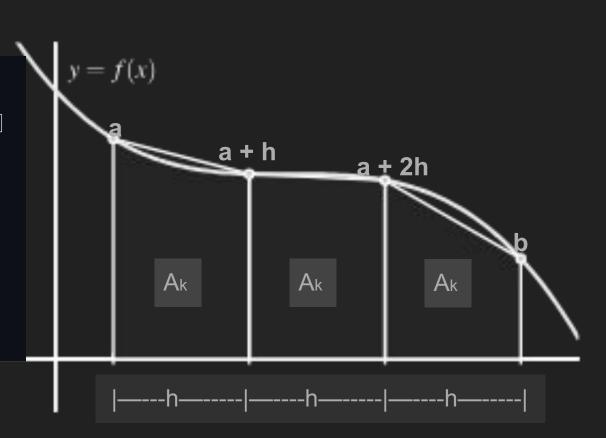
$$A \approx \frac{1}{2} h \sum_{k=1}^{N} \left[f(a + h(k-1)) + f(a + hk) \right]$$

Rewriting in a more compact form:

$$A \approx \frac{h}{2} \left[f(a) + f(b) + 2 \sum_{k=1}^{N-1} f(a+hk) \right]$$

...or as Mark prefers it:

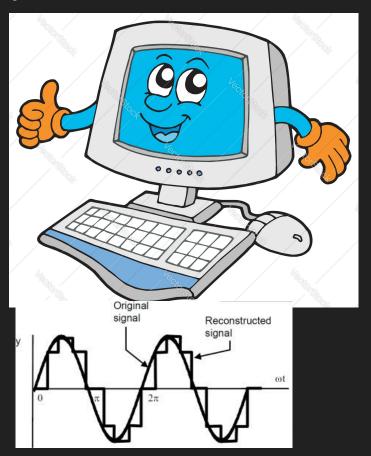
$$A \approx h \left[\frac{1}{2} f(a) + \frac{1}{2} f(b) + \sum_{k=1}^{N-1} f(a+hk) \right]$$



So if we integrate over data a computer has measured...

- N = len(x)

```
def empirical_trapezoidal_rule(y_values, x_values, N):
    a, b = x_values[0], x_values[-1]
    h = x_values[1] - x_values[0]
    integral = (1/2) * (y_values[0] + y_values[-1]) * h # First and last terms
    for k in range(1, N):
       yk = np.interp(xk, x_values, y_values) # Interpolate y at x_k manually in loop
        integral += vk * h
    return integral
x_{data} = np.array([0, 0.3, 0.6, 1]) # Given x data
y_data = np.array([1.0, 0.85, 0.55, 0.2]) # Corresponding y data
```



You can't learn to code from listening to me speak...

github.com/nialljmiller/PHYS4840_Numerical_integration_I



Download and complete "trapz.py"

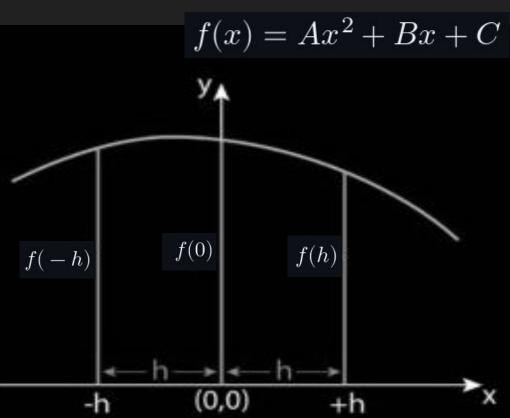
How do we compute how wrong we are?

How wrong is the trapz method with N = 2, 10, 100?

- Slight improvement to the trapezoids
- Now for some Maths

$$f(-h) = Ah^2 - Bh + C$$
$$f(0) = C$$

$$f(h) = Ah^2 + Bh + C$$



$$f(-h) = Ah^2 - Bh + C$$

$$f(0) = C$$

$$f(h) = Ah^2 + Bh + C$$

$$A = \frac{1}{2h^2}(f(h) - 2f(0) + f(-h))$$

$$B = \frac{1}{2h}(f(h) - f(-h))$$

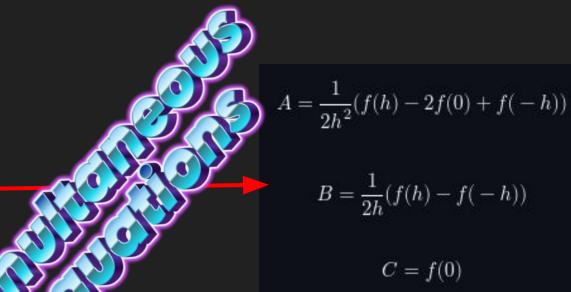
$$C = f(0)$$



$$f(-h) = Ah^2 - Bh + C$$

$$f(0) = C$$

$$f(h) = Ah^2 + Bh + C$$

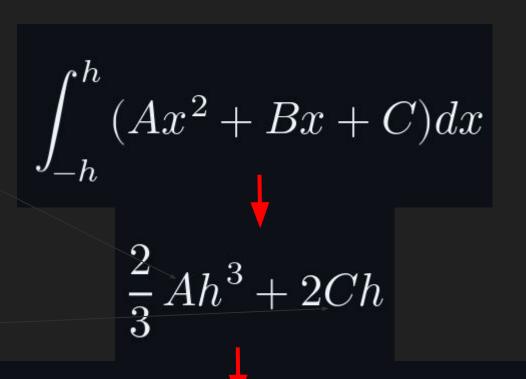




$$A = \frac{1}{2h^2}(f(h) - 2f(0) + f(-h))$$

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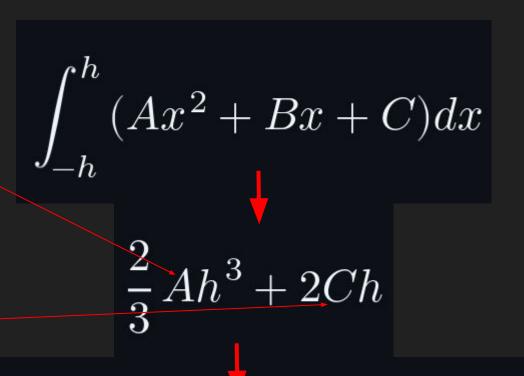
$$\frac{1}{3}h[f(-h)+4f(0)+f(h)]$$



$$A = \frac{1}{2h^2}(f(h) - 2f(0) + f(-h))$$

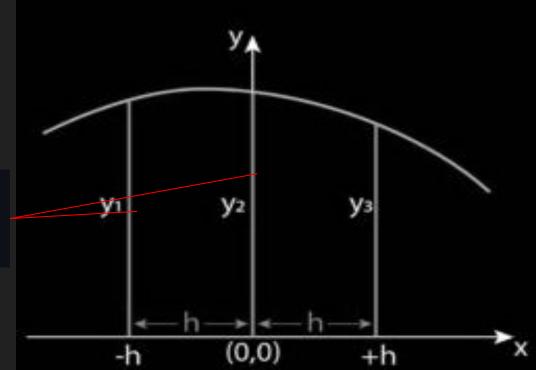
$$B = \frac{1}{2h}(f(h) - f(-h))$$

$$C = f(0)$$



$$\frac{1}{3}h[f(-h)+4f(0)+f(h)]$$

$$\frac{1}{3}h[f(-h) + 4f(0) + f(h)]$$



$$\frac{1}{3} h[f(-h) + 4f(0) + f(h)]$$

$$-h \qquad (0,0) \qquad +h$$

$$\int_{a}^{b} f(x), dx \approx \frac{h}{3} \left[f(a) + f(b) + 4 \sum_{\text{odd } k}^{1...N-1} f(a+kh) + 2 \sum_{\text{over } k}^{2...N-2} f(a+kh) \right]$$

Same as before...

github.com/nialljmiller/PHYS4840_Numerical_integration_I



Download and complete "simpson.py"

How wrong is the simpson method with N = 2, 10, 100?

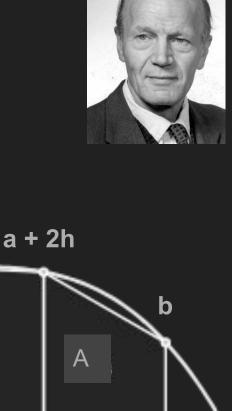
How does this compare to trapz?

Romberg Integration (1955)

- Improvement to the trapezoids
- We do some fancy maths to extrapolate
- This generally outperforms the others

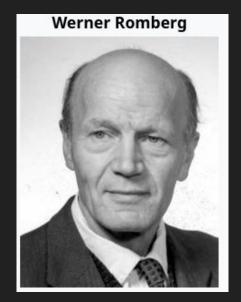
The Romberg Integration method builds upon the Trapezoidal Rule by applying Richardson Extrapolation, which systematically removes error terms to produce a more accurate result.

Instead of computing a single approximation, Romberg Integration refines the integral estimation step by step using a sequence of trapezoidal approximations.

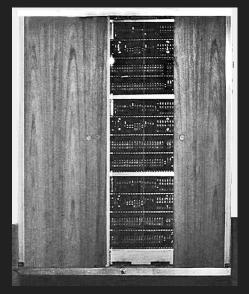


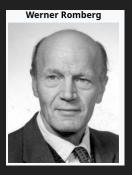
a + h

Werner Romberg was a German "half-jew" in the 1940s he fled to uppsala sweden, after WW2 he set up in trondheim Norway and went on to further numerical analysis and was instrumental in the installment of the first computer at trondheim.

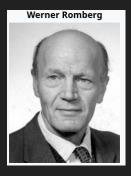




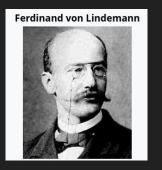


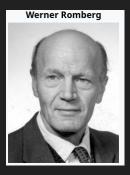




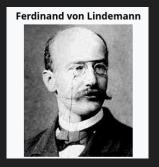


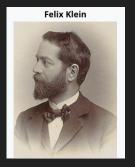






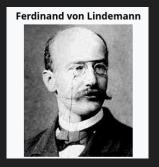


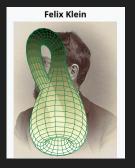


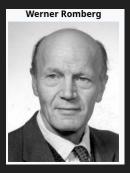




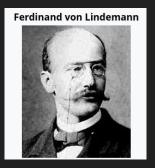


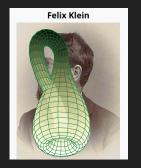


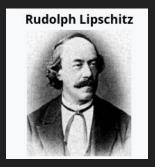




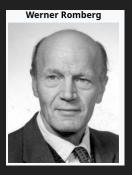




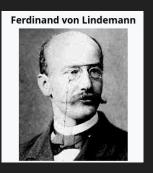


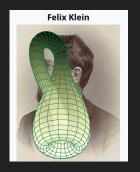


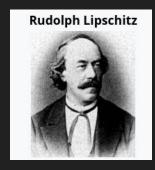




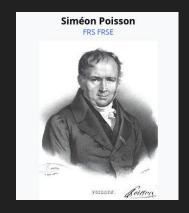


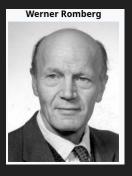




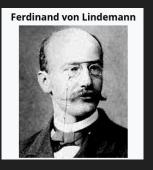




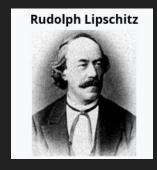




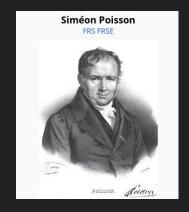


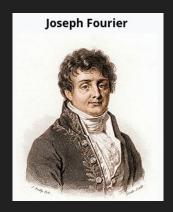






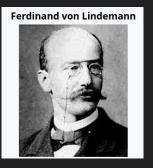




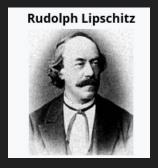




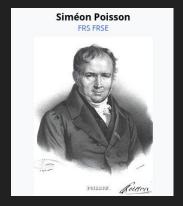




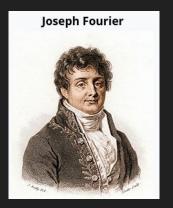










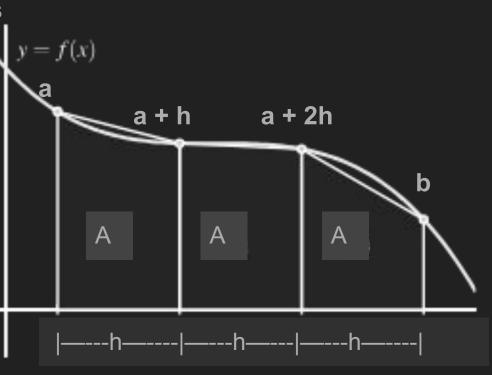


Romberg Integration

- Slight improvement to the trapezoids
- We do some fancy maths to extrapolate
- This generally outperforms the others

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Instead of computing a single approximation, Romberg Integration refines the integral estimation step by step using a sequence of trapezoidal approximations.



Romberg Integration

- Slight improvement to the trapezoids
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Instead of computing a single approximation, Romberg Integration refines the integral estimation step by step using a sequence of trapezoidal approximations.

See github for derivation

$$R_{m,n} = \frac{4^n R_{m,n-1} - R_{m-1,n-1}}{4^n - 1}$$

Romberg Integration Table

The Romberg method fills in a triangular table as follows:

(m)	($R_{m,0}$)	($R_{m,1}$)	($R_{m,2}$)	($R_{m,3}$)	
0	(T_0)				
1	(T_1)	($R_{1,1}$)			-
2	(T_2)	($R_{2,1}$)	($R_{2,2}$)		
3	(T ₃)	($R_{3,1}$)	($R_{3,2}$)	(R _{3,3})	

Nested functions detour



"functionA" uses arg 'a' and passes args 'b' and 'c' to "functionB".

"functionB" does not care about the argument names.

"functionA" must have been given 'b' and 'c' to be able to give them to "functionB"

```
def functionA(a, b, c):
    value = a
    value = value + functionB(b, c) + functionC(y = b, x = a)
    return value

#function B does not care what the arguments are called, but it does care about the order.

def functionB(x, y):
    value = x * y
    return value
```

Nested functions detour



```
def functionA(a, b, c):
    value = a
    value = value + functionB(b, c) + functionC(y = b, x = a)
    return value
#function B does not care what the arguments are called, but it does care about the order.
def functionB(x, y):
    value = x * y
    return value
#function C does care -- this is the difference between args and kwargs
def functionC(x = 'a', y = 'b'):
    return x + y
# Oops, we forgot to pass any arguments :(
result = functionA()
print(result)
```

Nested functions detour

print(result)



```
def functionA(a, b, c):
    value = a
    value = value +
    return value
                        github.com/niallimiller/PHYS4840 Numerical integration I
#function B does not
                                                                          care about the order.
def functionB(x, y):
    value = x * y
    return value
#function C does care
def functionC(x = 'a
    return x + y
                                nested_functions.py
# Oops, we forgot to pass any arguments :(
result = functionA()
```

In-class Exercise:

github.com/nialljmiller/PHYS4840_Numerical_integration_I

Download and complete "<u>integral_test.py</u>" (This will take longer than previous, consult the github/textbook)

Run the timing test in a loop and save the error, time, N

Make a plot showing (see plot_example.py):

- the accuracy of each method vs N
- the accuracy of each method vs compute time

Numerical integration II

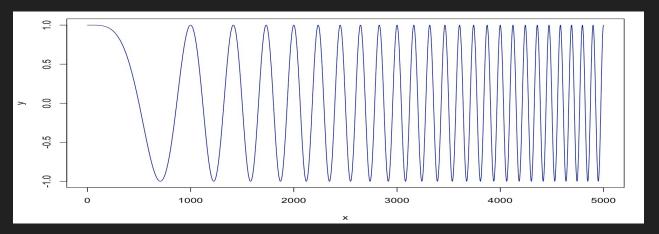
Dr Niall Miller

Gauss Legendre Quadrature

- Named as such because it predates computers
- Motivates by 19th century folk wanting to compute integrals for low N
 Simpsons rule for N = 1000 by hand is not fun
- Legendre polynomials are used to compute compute quaderature nodes and weights
- These 'nodes' are not determined as static. (i.e. not uniformly spaced)
- This allows this method to achieve high accuracy with fewer points as compared to the uniformly spaced method

An explanation of this method

- A limitation of the previously discussed methods is that they assume the 'complexity' of our function/distribution is consistent across the integrated region
- Does a consistent width size make sense for this?



- Instead of using equally spaced points (like in Trapezoidal or Simpson's Rule), we choose points that make the approximation as accurate as possible for the least effort.
- The best points to use turn out to be the roots of Legendre polynomials.

Gauss-Legendre

As usual, we approximate an integral with a sum. This time a weighted sum of the form:

$$\left|\int_{-1}^1 f(x)\,dxpprox \sum_{i=1}^n w_i f(x_i)
ight|$$



So all this method is:

- Consult legendre polynomials to find our weights and roots
- Calculate the value of our function at each of our Legendre 'roots'
- Multiply the value by the weight
- Sum everything together
- Done

Legendre polynomials?

These are polynomials $P_n(x)$ that appear in many areas of mathematics and physics.

$$(1-x^2)P_n''(x)-2xP_n'(x)+n(n+1)P_n(x)=0.$$

$$(1-x^2)P_n''(x)-2xP_n'(x)+n(n+1)P_n(x)=0. \ \int_{-1}^1 P_m(x)P_n(x)\,dx=0 \quad ext{if } n
eq m.$$

- The key point we need to know is that we use the root values This means we take the value of x where $P_n(x) = 0$
- The Legendre polynomials will give use the **positions** and **weights** for our integral

 $P_n(x)$

Legendre polynomials?

- These are polynomials P_n(x) that appear in many areas of mathematics and physics.

$$P_0(x) = 1$$

$$P_1(x) = x$$

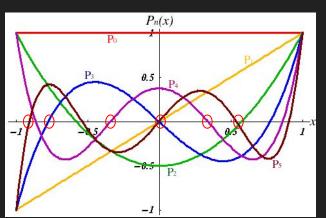
$$P_2(x) = \frac{1}{2}(3x^2 - 1)$$

$$P_3(x) = \frac{1}{2}(5x^3 - 3x)$$

$$P_4(x) = \frac{1}{8}(35x^4 - 30x^2 + 3)$$

$$P_5(x) = \frac{1}{8}(63x^5 - 70x^3 + 15x)$$

$$P_6(x) = \frac{1}{16}(231x^6 - 315x^4 + 105x^2 - 5)$$





- The key point we need to know is that we use the root values This means we take the value of x where $P_n(x) = 0$
- The Legendre polynomials will give use the positions and weights for our integral

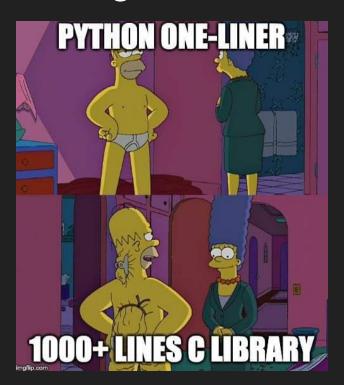
That sounds easy ...apart from Legendre

Numpy is amazing!



np.polynomial.legendre.leggauss(deg)

"Computes the sample points and weights for Gauss-Legendre quadrature."



That sounds easy ...apart from Legendre

- ♠ > NumPy reference > Routines and objects by topic > Polynomials
- > Legendre Series (numpy.polynomial.legendre) > numpy.polynomial.legendre.leggauss

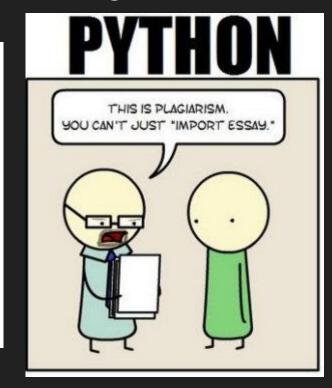
numpy.polynomial.legendre.leggauss

polynomial.legendre.leggauss(deg)

[source]

Gauss-Legendre quadrature.

Computes the sample points and weights for Gauss-Legendre quadrature. These sample points and weights will correctly integrate polynomials of degree 2*deg-1 or less over the interval [-1,1] with the weight function f(x)=1.



"Computes the sample points and weights for Gauss-Legendre quadrature."

That sounds easy ...apart from Legendre

np.polynomial.legendre.leggauss(deg)

$$P_0(x) = 1$$
 $P_1(x) = x$
 $P_1(x) = x$
 $P_2(x) = \frac{1}{2}(3x^2 - 1)$
 $P_3(x) = \frac{1}{2}(5x^3 - 3x)$
 $P_4(x) = \frac{1}{8}(35x^4 - 30x^2 + 3)$
 $P_5(x) = \frac{1}{8}(63x^5 - 70x^3 + 15x)$
 $P_6(x) = \frac{1}{16}(231x^6 - 315x^4 + 105x^2 - 5)$.

#how to do thing in python
import thing_doer as thingy

thing = thingy.do_thing()

Python thingys!

```
from numpy.polynomial.legendre import leggauss as legendre_thingy
     legendre_roots, weights = legendre_thingy(3)
    print(legendre_roots, weights)
njm@fedora [13:21:35] [~/PHYS4840_Numerical_integration_I] [main]
> % python python.py
Gauss-Legendre Quadrature Points (Roots of P_n(x)):
[-0.77459667 0. 0.77459667]
Weights for each point:
[0.55555556 0.88888889 0.55555556]
```

Lets code together!

github.com/nialljmiller/PHYS4840_Numerical_integration_I



gauss_legendre.py

In-class Exercise:

github.com/nialljmiller/PHYS4840_Numerical_integration_I

Download and complete "<u>integral_test.py</u>" (This will take longer than previous, consult the github/textbook)

Run the timing test in a loop and save the error, time, N

Make a plot showing (see plot_example.py):

- the accuracy of each method vs N
- the accuracy of each method vs compute time