

Computational Physics (PHYS6350)

Lecture 8: Numerical Integration: Part 1

- Basic methods for numerical integration (rectangle, trapezoid, Simpson)
- Adaptive quadrature
- Improper integrals

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Course materials: <u>https://github.com/vlvovch/PHYS6350-ComputationalPhysics</u> **Online textbook:** <u>https://vovchenko.net/computational-physics/</u> Generic problem: evaluate

$$I=\int_a^b f(x)dx$$

We need numerical integration when

- Cannot/difficult integrate analytically
- Only know the integrand f(x) at certain points



References: Chapter 5 of Computational Physics by Mark Newman Chapter 4 of Numerical Recipes Third Edition by W.H. Press et al. Interpret the integral as the area under the curve and approximate by a rectangle evaluated at midpoint

$$\int_{a}^{b} f(x) \, dx \approx (b-a) \, f\left(\frac{a+b}{2}\right)$$



Error (from Euler-McLaurin formula):

$$\int_{a}^{b} f(x)dx - (b-a) f\left(\frac{a+b}{2}\right) \approx \frac{(b-a)^{3}}{24} f''(a)$$

The rule is exact for the integration of linear functions

Numerical integration: rectangular (midpoint) rule

Example:

$$I=\int_0^2 2x+3dx=10$$



Although the rectangle is a poor approximate of the line (which is a trapezoid here), the errors cancel out

Numerical integration: rectangular (midpoint) rule

Another example:

$$I = \int_{0}^{2} x^{4} - 2x + 2 = 6.4$$

$$I = 2.0$$

$$I = -10$$

$$I =$$



Rectangle rule gives $I_{rect} = 2$ which is way off

Extended (composite) rectangular rule

Split the integration interval into N sub-intervals and apply the rectangle rule separately to each one

$$\int_{a}^{b} f(x) \approx h \sum_{k=1}^{N} f(x_{k}), \qquad k = 1, \dots, N$$
$$x_{k} = a + \frac{2k - 1}{2}h.$$
$$h = (b - a)/N$$

Error estimate:

$$I - I_{\text{rect}} = (b - a)\frac{h^2}{24} f''(a) + \mathcal{O}(h^4)$$

 $\int_0^2 x^4 - 2x + 2$



Extended (composite) rectangular rule

$$I = \int_0^2 x^4 - 2x + 2 = 6.4$$



Numerical integration: trapezoidal rule

Approximate the integral by a trapezoid

$$\int_{a}^{b} f(x) dx \approx (b-a) \frac{f(a) + f(b)}{2}$$



Error:

$$\int_{a}^{b} f(x)dx - (b-a)\frac{f(a) + f(b)}{2} \approx -\frac{(b-a)^{3}}{12}f''(a)$$

The rule is exact for the integration of linear functions



Numerical integration: trapezoidal rule



Trapezoidal rule gives $I_{trap} = 16$, way off and in the opposite direction relative to rectangle rule

Extended trapezoidal rule

$$\int_{a}^{b} f(x) \approx h \sum_{k=0}^{N} \frac{f(x_{k}) + f(x_{k+1})}{2}, \quad i = 0, ..., N$$

$$I = 6.441650390625$$

$$I = 6.441650906$$

$$I = 6.44165090625$$

$$I = 6.441650906$$

$$I - I_{\text{trap}} = -(b - a)\frac{h^2}{12} f''(a) + \mathcal{O}(h^4)$$

Extended (composite) trapezoidal rule

$$I = \int_0^2 x^4 - 2x + 2 = 6.4$$



Numerical integration: Simpson's rule

Recall the error estimates for rectangular and trapezoidal rules

$$I - I_{\text{rect}} = (b - a)\frac{h^2}{24} f''(a) + \mathcal{O}(h^4) \qquad \qquad I - I_{\text{trap}} = -(b - a)\frac{h^2}{12} f''(a) + \mathcal{O}(h^4)$$

Combine them to eliminate the $O(h^2)$ error term:

$$I_S = \frac{2I_{\text{rect}} + I_{\text{trap}}}{3}$$



$$\int_{a}^{b} f(x) dx \approx \frac{(b-a)}{6} \left[f(a) + 4f\left(\frac{a+b}{2}\right) + f(b) \right]$$
 Simpson's rule

An equivalent way to obtain the rule: replace the integrand by the parabolic interpolation



Numerical integration: Simpson's rule

$$\int_{a}^{b} f(x) dx \approx \frac{(b-a)}{6} \left[f(a) + 4f\left(\frac{a+b}{2}\right) + f(b) \right]$$

The error for the Simpson's rule is

$$I - I_S = C h^4 + \mathcal{O}(h^6)$$

The method is exact for polynomials up to third order



Numerical integration: Simpson's rule

$$I = \int_0^2 x^4 - 2x + 2 = 6.4$$



Simpson's rule gives $\mathbf{I}_{trap}=6.66$ using three points, which is already not too bad!

Extended Simpson's rule

$$\int_{a}^{b} f(x) \approx \frac{h}{3} \left[f(x_{0}) + 4 \sum_{k=1}^{N/2} f(x_{2k-1}) + 2 \sum_{k=1}^{N/2-1} f(x_{2k}) + f(x_{N}) \right], \qquad i = 0, \dots, N \qquad \int_{0}^{2} x^{4} - 2x + 2$$

h = (b - a)/N

N must be even!

Error estimate:

$$I - I_S = C h^4 + \mathcal{O}(h^6)$$



Extended Simpson's rule

$$I = \int_0^2 x^4 - 2x + 2 = 6.4$$



Comparing the methods



We would like to control the error in our calculation

This can be achieved by doubling the number of subintervals and keeping track of the error estimate

Recall that in the rectangle/trapezoidal rule the error is proportional to h^2

$$I-I_{
m trap}pprox ch^2$$



At step k we have $h_k = h_{k-1}/2$ therefore $I - I_{trap}^k \approx ch_k^2$, $I - I_{trap}^{k-1} \approx 4ch_k^2$, and the error at step k is estimated as

$$\varepsilon_k \simeq (I_{\rm trap}^k - I_{\rm trap}^{k-1})/3$$

Adaptive trapezoidal rule

```
Trapezoidal rule for numerical integration with adaptive step
if trapezoidal rule adaptive(f, a, b, nst = 1, tol = 1.e-8, max iterations = 16):
  Iprev = 0.
  n = nst
  Iprev = trapezoidal rule(f, a, b, n)
  print("Iteration: {0:5}, I = {1:20.15f}".format(1, Iprev))
  for k in range(1, max iterations):
      n *= 2
      Inew = trapezoidal_rule(f, a, b, n)
      ek = (Inew - Iprev) / 3.
      print("Iteration: {0:5}, I = {1:20.15f}, error estimate = {2:10.15f}".format(k+1, Inew, ek))
      if (abs(ek) < tol):</pre>
          return Inew
                       Computing the integral of x^4 - 2x + 2 over the interval (0.0, 2.0) using adaptive trapezoidal rule
      Iprev = Inew
                       Iteration:
                                     Iteration:
                                     2, I =
                                               9.00000000000000, error estimate = -2.3333333333333333
  print("Failed to achi
                       Iteration:
                                     3, I =
                                               7.06250000000000, error estimate = -0.645833333333333
  return Inew
                       Iteration:
                                     4, I =
                                               6.566406250000000, error estimate = -0.165364583333333
                       Iteration:
                                    5, I =
                                               6.441650390625000, error estimate = -0.041585286458333
                       Iteration:
                                    6, I =
                                               6.410415649414062, error estimate = -0.010411580403646
                       Iteration:
                                    7, I =
                                               6.402604103088379, error estimate = -0.002603848775228
                                    8, I =
                       Iteration:
                                               6.400651037693024, error estimate = -0.000651021798452
                                    9, I =
                       Iteration:
                                               6.400162760168314, error estimate = -0.000162759174903
                       Iteration:
                                    10, I =
                                               6.400040690088645, error estimate = -0.000040690026556
                                    11, I =
                       Iteration:
                                               6.400010172525072, error estimate = -0.000010172521191
                                    12, I =
                       Iteration:
                                               6.400002543131352, error estimate = -0.000002543131240
                                    13, I =
                       Iteration:
                                               6.400000635782950, error estimate = -0.000000635782801
                                    14, I =
                                               6.400000158945742, error estimate = -0.000000158945736
                       Iteration:
                       Iteration:
                                    15, I =
                                               6.400000039736406, error estimate = -0.000000039736446
                                    16, I =
                       Iteration:
                                               6.40000009934106, error estimate = -0.000000009934100
```

```
For Simpson's rule \varepsilon_k \simeq (I_S^k - I_S^{k-1})/15 (understand why 15 and not 3?)
```

```
# Simpson's rule for numerical integration with adaptive step
def simpson rule adaptive(f, a, b, nst = 2, tol = 1.e-8, max iterations = 16):
   Iprev = 0.
   n = nst
   Iprev = simpson rule(f, a, b, n)
   print("Iteration: {0:5}, I = {1:20.15f}".format(1, Iprev))
   for k in range(1, max iterations):
       n *= 2
       Inew = simpson rule(f, a, b, n)
       ek = (Inew - Iprev) / 15.
       print("Iteration: {0:5}, I = {1:20.15f}, error estimate = {2:10.15f}".format(k+1, Inew, ek))
       if (abs(ek) < tol):</pre>
          return Inew
       Iprev = Inew
                     Computing the integral of x^4 - 2x + 2 over the interval (0.0, 2.0) using adaptive Simpson's rule
   print("Failed to ac Iteration:
                                  1, I =
                                            6.66666666666666
   return Inew
                     Iteration:
                                  2, I =
                                            Iteration:
                                  3, I =
                                            Iteration:
                                  4, I =
                                            6.400065104166666, error estimate = -0.000065104166667
                     Iteration:
                                  5, I =
                                            6.400004069010416, error estimate = -0.000004069010417
                     Iteration:
                                  6, I =
                                            6.400000254313150, error estimate = -0.000000254313151
                     Iteration:
                                  7, I =
                                            6.400000015894571, error estimate = -0.000000015894572
                     Iteration:
                                   8, I =
                                            6.40000000993410, error estimate = -0.00000000993411
```

Adaptive quadratures: Romberg method

Recall that we obtained error estimate for trapezoidal method at step k

$$arepsilon_k \simeq (I_{ ext{trap}}^k - I_{ ext{trap}}^{k-1})/3$$

On the other hand, by definition, $\varepsilon_k = I - I_{\text{trap}}^k$

Therefore, we can improve our estimate of the integral as

$$I = R_{k,1} = I_{trap}^{k} + \frac{I_{trap}^{k} - I_{trap}^{k-1}}{3} + \mathcal{O}(h^{4})$$

Romberg method: continue this procedure iteratively

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

until the desired accuracy is reached

Romberg method

```
def romberg(
   f,
    a,
    b,
    accuracy=1e-8,
    max order=10
):
    R = np.zeros((max order, max order))
    h = (b - a) / 2.
    R[0, 0] = h * (f(a) + f(b)) # The initial trapezoidal rule
   for n in range(1, max order):
        trapezoid = 0.0
       for j in range(2**(n-1)):
            trapezoid += f(a + (2*j+1)*h)
        R[n, 0] = 0.5 * R[n-1, 0] + h * trapezoid # The trapezoidal rule
        1 = 1
        # The Romberg iterations
       for m in range(1, n+1):
            1 *= 4
            R[n, m] = (1 * R[n, m-1] - R[n-1, m-1]) / (1-1)
        print("Iteration: {0:5}, I = {1:20.15f}, error estimate = {2:10.15f}".format(n, R[n, m], abs(R[n, m] - R[n-1, m-1])))
        if abs(R[n, m] - R[n-1, m-1]) < accuracy:
            return R[n, m]
        h /= 2.
    print("Romberg method did not converge to required accuracy")
    return R[-1, -1]
```

Nothing wrong with integrating discontinuous functions



$$f(x) = 3x^{2} + x + 3 \qquad \text{for } x < 1,$$

$$f(x) = 2x^{3} - 3x^{2} + x + 3. \qquad \text{for } x > 1$$

$$f = \int_0^2 f(x) = 9.5$$





$$f(x) = 3x^2 + x + 3$$
 for $x < 1$,

 $f(x) = 2x^3 - 3x^2 + x + 3.$ for x > 1,

Adaptive rectangle rule
print("Rectangle rule:")
rectangle_rule_adaptive(fdist,0,2,1,1.e-8)

Adaptive trapezoidal rule
print("Trapezoidal rule:")
trapezoidal_rule_adaptive(fdist,0,2,1,1.e-8)

```
# Adaptive Simpson rule
print("Simpson's rule:")
simpson_rule_adaptive(fdist,0,2,2,1.e-8)
```

Romberg method
print("Romberg method:")
romberg(fdist,0,2,1e-8,18)

The methods kind of work but not quite

$$I = \int_0^2 f(x) = 9.5$$

Iteration: 15, I = 9.499999988824127, error estimate = 0.000000011175870 Iteration: 16, I = 9.499999997206030, error estimate = 0.000000002793968 9.49999999720603

Iteration: 16, I = 9.499877935275437, error estimate = 0.000040684516232 Failed to achieve the desired accuracy after 16 iterations

9.499877935275437

Iteration: 16, I = 9.499959309895530, error estimate = 0.000002712673591 Failed to achieve the desired accuracy after 16 iterations

9.49995930989553

Iteration: 17, I = 9.499981410226075, error estimate = 0.000018589773871
Romberg method did not converge to required accuracy

9.499981410226075

Better strategy: split the integral into two and integrate separately

$$I = I_1 + I_2,$$
 $I_1 = \int_a^{x_{\text{distcont}}} f_1(x)$

Adaptive Simpson's rule
print("Simpson's rule:")
eps = 1.e-8
a = 0
b = xdistcont
I1 = simpson_rule_adaptive(fdist1,a,b,2,0.5*eps)
a = xdistcont
b = 2
I2 = simpson_rule_adaptive(fdist2,a,b,2,0.5*eps)
print("I1 =",I1)
print("I2 =",I2)
print("I =",I1 + I2)

Simpson's rule: Iteration: 1, I = 4.500000000000000 Iteration: 2, I = 4.50000000000000, error estimate = 0.00000000000000 Iteration: 1, I = 5.000000000000000 5.00000000000000, error estimate = 0.00000000000000 Iteration: 2, I = I1 = 4.5I2 = 5.0I = 9.5

$$I_2 = \int_{x_{\rm distcont}}^b f_2(x)$$

Romberg method
print("Romberg method:")
eps = 1.e-8
a = 0
b = xdistcont
I1 = romberg(fdist1,a,b,0.5*eps)
a = xdistcont
b = 2
I2 = romberg(fdist2,a,b,0.5*eps)
print("I1 =",I1)
print("I2 =",I2)
print("I =",I1 + I2)

Romberg method:

Iteration:	1,	I =	4.5000000000000000,	error	estimate	=	0.5000000000000000
Iteration:	2,	I =	4.5000000000000000,	error	estimate	=	0.000000000000000
Iteration:	1,	I =	5.0000000000000000,	error	estimate	=	1.000000000000000
Iteration:	2,	I =	5.0000000000000000,	error	estimate	=	0.000000000000000
I1 = 4.5							
I2 = 5.0							
I = 9.5							

Improper integrals

• Contain integrable singularities (typically at the endpoints)

$$\int_{0}^{1} \frac{1}{\sqrt{x}} dx = 2\sqrt{x} \Big|_{0}^{1} = 2$$

• (Semi-)infinite integration range

$$\int_0^\infty e^{-x} dx = 1 \qquad \qquad \int_{-\infty}^\infty e^{-x^2} dx = \sqrt{\pi}$$

Example: Momentum integration over thermal distributions (Fermi-Dirac/Bose-Einstein)

$$n = \frac{d}{2\pi^2} \int_0^\infty dk \; k^2 \; \left[\exp\left\{ \frac{\sqrt{m^2 + k^2} - \mu}{T} \right\} \pm 1 \right]^{-1}$$

Improper integrals: Singularities at endpoints

• Even though if the singularities at integration endpoints are integrable, the trapezoidal, Simpson, etc. methods will fail because they evaluate the integrand at the endpoints

$$\int_0^1 \frac{1}{\sqrt{x}} dx = 2\sqrt{x} \Big|_0^1 = 2$$

```
def fsing1(x):
    return 1./np.sqrt(x)
trapezoidal_rule(fsing1,0.,1.,10)
/tmp/ipykernel_31240/847063500.py:2: RuntimeWarning: divide by zero encountered in double_scalars
    return 1./np.sqrt(x)
```

• Solution: use method that does use the endpoints (e.g. rectangle rule)

Improper integrals: Singularities at endpoints

....

$$\int_{0}^{1} \frac{1}{\sqrt{x}} dx = 2\sqrt{x} \Big|_{0}^{1} = 2$$

def fsing1(x):
 return 1./np.sqrt(x)

print('Using rectangle rule to evaluate \int_0^1 1/\sqrt{x} dx')
nst = 1
rectangle_rule_adaptive(fsing1,0.,1.,1,1.e-3,20)

Using rectangle rule to evaluate \int_0^1 1/\sqrt{x} dx					
Iteration:	1, I =	1.414213562373095			
Iteration:	2, I =	1.577350269189626, error estimate = 0.054378902272177			
Iteration:	3, I =	1.698844079579673, error estimate = 0.040497936796682			
Iteration:	4, I =	1.786461001734842, error estimate = 0.029205640718390			
Iteration:	5, I =	1.848856684639738, error estimate = 0.020798560968299			
Iteration:	6, I =	1.893088359706383, error estimate = 0.014743891688882			
Iteration:	7, I =	1.924392755699513, error estimate = 0.010434798664376			
Iteration:	8, I =	1.946535279970520, error estimate = 0.007380841423669			
Iteration:	9, I =	1.962194152677056, error estimate = 0.005219624235512			
Iteration:	10, I =	1.973267083679453, error estimate = 0.003690977000799			
Iteration:	11, I =	1.981096937261288, error estimate = 0.002609951193945			
Iteration:	12, I =	1.986633507070365, error estimate = 0.001845523269692			
Iteration:	13, I =	1.990548459938304, error estimate = 0.001304984289313			
Iteration:	14, I =	1.993316751362098, error estimate = 0.000922763807931			

Improper integrals: (Semi-)infinite intervals

Solution: map to a finite interval [e.g. (0,1)] by a change of variables

• Semi-infinite:

• Infinite:

Then apply a standard method (e.g. rectangle rule to avoid endpoint singularities) to g(t)

NB: Other options for the change of variable are possible

Improper integrals: Semi-infinite intervals

```
\int_0^\infty e^{-x} dx = 1
```

dx

```
def fexp(x):
    return np.exp(-x)

def g(t, f, a = 0.):
    return f(a + t / (1. - t)) / (1. - t)**2

a = 0.
```

```
def frect(x):
    return g(x, fexp, a)
```

print('Using change of variable and the rectangle rule to evaluate \int_0^\infty \exp(-x) dx')
rectangle_rule_adaptive(frect,0.,1.,1,1.e-6,20)

Using change	of variable	and the rectangle rule to evaluate $int_0^{infty} \exp(-x)$
Iteration:	1, I =	1.471517764685769
Iteration:	2, I =	1.035213267452946, error estimate = -0.145434832410941
Iteration:	3, I =	0.984670579385046, error estimate = -0.016847562689300
Iteration:	4, I =	1.001784913275257, error estimate = 0.005704777963404
Iteration:	5, I =	1.000155714391028, error estimate = -0.000543066294743
Iteration:	6, I =	1.000040642390661, error estimate = -0.000038357333456
Iteration:	7, I =	1.000010172618432, error estimate = -0.000010156590743
Iteration:	8, I =	1.000002543136036, error estimate = -0.000002543160799
Iteration:	9, I =	1.000000635783161, error estimate = -0.000000635784292

Improper integrals: Infinite intervals

$$\int_{-\infty}^{\infty} e^{-x^2} dx = \sqrt{\pi} = 1.772454 \dots$$

```
def fexp2(x):
    return np.exp(-x**2)
def g2(t, f):
    return f(t / (1. - t**2)) * (1.+t**2) / (1. - t**2)**2
```

```
def frect2(x):
    return g2(x, fexp2)
```

```
print('Using change of variable and the rectangle rule to evaluate \int_{-\infty}^\infty \exp(-x^2) dx')
rectangle_rule_adaptive(frect2,-1.,1.,1,1.e-6,20)
```

```
print('Expected value: \sqrt{\pi} =', np.sqrt(np.pi))
```

Using change o	of variable	and the rectangle rule to evaluate \int_{-\infty}^\infty \exp(-x^2) dx				
Iteration:	1, I =	2.0000000000000				
Iteration:	2, I =	2.849690615244243, error estimate = 0.283230205081414				
Iteration:	3, I =	1.557994553948652, error estimate = -0.430565353765197				
Iteration:	4, I =	1.808005109208286, error estimate = 0.083336851753211				
Iteration:	5, I =	1.770118560572371, error estimate = -0.012628849545305				
Iteration:	6, I =	1.772492101507391, error estimate = 0.000791180311673				
Iteration:	7, I =	1.772453880915058, error estimate = -0.000012740197444				
Iteration:	8, I =	1.772453850905505, error estimate = -0.000000010003185				
Expected value: \sqrt{\pi} = 1.7724538509055159						