Numerical integration is done by fitting a polyomial and finding a weighted-sum. Numerical differentiation is done by brute-forcing the limit.

**numerical-integration:~#** Newton-Cotes integration uses rational weights for evenly spaced data while Gaussian quadrature uses Legendre roots weights for unevenly spaced data. To pick a method, accuracy∝noise.

…

s = 0.5\*f(a) + 0.5\*f(b)

for k in range(1,N):

s += f(a+k\*h)

return h\*s

# First-order Newton-Cotes integration is called trapezoidal rule. Observe that it has 0.5 weights at boundaries a and b and unity weights at the middle.

def gaussxwab(N,a,b):

x,w = gaussxw(N)

return 0.5\*(b-a)\*x+0.5\*(b+a),0.5\*(b-a)\*w

# Gaussain quadrature weights can be directly called from a python library by assigning weights at domain [-1,1] and rescaling.

! Pitfall: Mind the order parity of Newton-Cotes method. Plugging in odd slices for, say, Simpson’s rule will yield an oscillating error.

**approximation-error:~#** A closed form expression for the approximation error for each iteration of an integration scheme allows us to precisely “budget” computational power-to-error ratio. Extensive exploitation of this leads to Romberg integration: an add-on to Newton-cotes integration for more accuracy.

…

s = 0.5\*f(a) + 0.5\*f(b)

for k in range(1,N):

s += f(a+k\*h)

error = abs((1/3)\*(((1/2)\*I+(h\*t))-I))

I = ((1/2)\*I+(h\*t))

# Information about error at each iteration coded within the loop can be used to halt the loop as needed.

T[i][m] = T[i][m-1] +

((1)/((4\*\*m)-1))\*(T[i][m-1]-T[i-1][m-1])

# Data about the 2D romberg integration can be stored in an array and calling the latest version as needed.

! Pitfall: Make sure to use the latest version (max. i and m) for Romberg integration. Error oscillates periodically throughout the Romberg cycle.

**numerical-derivative:~#** A numerical method to derivatives is a brute-force approach to the the limit definition. The central difference method offers the best accuracy compared to the one-sided forward and backward difference methods.

def d\_f(x):

return ((f(x+(h/2))-f(x-(h/2)))/h)

# Numerical derivatives are straightforward to code by approaching h -> 0 through brute force than analytical limit evaluation.

! Pitfall: Especially with derivatives, watch out for noisy data. We can smoothen them via Fourier transforms or interpolate a polynomial.