CS319: Scientific Computing (with C++)

CS319 Lab 6: Numerical Integration 2

Week 8 (6+7 March, 2025)

Goal:

- Derive and Implement Simpson's Rule in 2D;
- Use a Jupyter Notebook to verify its convergence.
- Compare with a Monte-Carlo Method.

Deadline:

Submit your work for the assignment on Slide 10 by 17:00, Wednesday 12 March.

1. Review

- 1. Review your notes from Weeks 6 and 7 on Numerical Integration in one and two dimensions.
- Complete Lab 5: from that you should have working code for the Trapezium, Simpson's, and Boole's Rule for estimating a definite integral of a function of one variable. (If you've yet to complete Lab 5, focus on implementing Simpson's Rule).

In class we did a rough derivation of the Trapezium Rule in 2D, and implemented it as a function Trap2D().

Presently, we want to the two-dimensional Simpson's Rule. To have a single framework which works for both, we'll consider a slightly different presentation of the Trapezium Rule.

First, let's write the general formulation for a one-dimensional quadrature method:

- ▶ Define the (equally spaced) **quadrature points**: $\{x_0, x_1, \dots, x_N\}$. If we set h = (b-a)/N, then $x_i = a + ih$.
- ▶ Define the associated **quadrature points**: $y_0 = f(x_0), y_1 = f(x_1), \dots, y_N = f(x_N).$
- \blacktriangleright Suppose we have a set of quadrature weights w_0, w_1, \ldots, w_N .

▶ Then the general form of a quadrature rule is

$$\int_{a}^{b} f(x)dx \approx \sum_{i=0}^{N} w_{i}y_{i}.$$

▶ For the Trapezium Rule in one-dimension, we have

$$w_0 = \frac{h}{2}$$
, $w_1 = h$, $w_2 = h$,..., $w_{N-1} = h$, $w_N = \frac{h}{2}$.

The advantage of defining the method this way is that

- It is easy to apply this approach to other methods, such as Simpson's Rule or Boole's Rule.
- It is easy to generalise to higher dimensions.

Dealing with the last point first, we know that computing

$$\int_a^b \int_a^b f(x_1, x_2) dx_1 dx_2$$

can be thought of as integrating first in x_1 , and integrating then in x_2 .

In the same way, we can apply our quadrature rule in each direction...

That is

$$\int_{a}^{b} \int_{a}^{b} f(x_1, x_2) dx_1 dx_2 \approx \sum_{i=0}^{N} w_i \left(\sum_{j=0}^{N} w_j y_{i,j} \right)$$

where $y_{i,j} := f((x_1)_i, (x_2)_j)$. Expanding that, we get

$$\int_{a}^{b} \int_{a}^{b} f(x_{1}, x_{2}) dx_{1} dx_{2} \approx \sum_{i=0}^{N} \sum_{j=0}^{N} w_{i} w_{j} y_{i,j}$$
 (1)

This is implemented in the program Quad2D.cpp, which can be downloaded from

https://www.niallmadden.ie/2425-CS319/lab6/Quad2D.cpp Note that the quadrature weights are specified in main(), and the Quad2D() function uses them.

The code in Quad2D.cpp applies the 2D Trapezium Rule to estimating

$$\int_0^1 \int_0^1 e^{x_1 + x_2} dx_1 dx_2, \tag{2}$$

with N=16 intervals in each coordinate direction.

- Adapt that code in O1CheckConvergence.cpp from Week 6 (or Lab 5) so that the 2D Trapezium Rule is used to etimate the solution to (2) for various values of N, and the results are output as a NumPy array that can be copied into a Jupyter notebook.
- 2. Tip: make sure you allocate and de-allocate memory for x1, x2, and y within the for-loop that applies the method for various values of N.
- Adapt the Jupyter notebook at https://www.niallmadden.ie/ 2425-CS319/lab6/CS319-Lab6-Q1.ipynb to verify the convergence of the method.

- 4. Further modify the code so that it reports how much time a call to the 2D Trapezium function takes for a given N. (See code from Lab1-Q3.cpp from Lab 1 for an example of how to do this).
- 5. Have the C++ code output the times for each N as a numpy array. Again, copy that into the Jupyter Notebook. Determine K and r where the time taken, for a given N is

$$t(N) \approx KN^r$$
.

3. 2: Simpson's Rule in 2D

Based on our study of the Trapezium Rule, it is relatively easy to derive the 2D Simspon's Rule.

Recall first the 1D version:

$$\int_{a}^{b} f(x)dx \approx \sum_{i=0}^{N} w_{i}y_{i}.$$

where

$$w_0 = w_N = \frac{1}{3}h,$$

 $w_1 = w_3 = \dots = w_{N-1} = \frac{4}{3}h,$
 $w_2 = w_4 = \dots = w_{N-2} = \frac{2}{3}h,$

Now the method is exactly the same as in (1), except using these new weights.

4. Assignment

ASSIGNMENT

- (a) Write a C++ program that implements Simpson's Rule for estimating the solution to (2).
- (b) Your code should report the error, and time taken, for various values of N.
- (c) Use a Jupyter notebook to determine how the error, and time taken, depends on N.
- (d) Upload both your C++ code, and the Jupyter notebook (as PDF or HTML). Make sure all files contain, at the very least, your name and ID number. However, you should document your C++ code based on the verbal feedback from Lab 4.

5. Extra: Monte Carlo Methods

The approach given here is useful:

- ▶ We have a general framework for implementing any 1D method.
- We have a general framework for extending any method from 1D to 2D.
- ▶ In fact, this approach works in the "obvious" way if we extend to three or more dimensions...
- Moreover, it does get very slow. This is known as the "curse of dimensionality". For example, if we taken N=1,000 in one dimensional, we have to compute 1,000,000 values of f in two dimensions. In general, the time scales like N^d in d dimensions.

Monte Carlo methods are a useful way of estimating high dimensional integrals. They are fast, if not especially accurate...

5. Extra: Monte Carlo Methods

Monte Carlo Algorithm

- ▶ Choose some value of N and compute an $(N+1) \times (N+1)$ array, z, of values of f, as we did in previous methods.
- ▶ Choose some value M such that $M \ll N^2$. Usually, taking M = N is OK.
- ► Compute

$$Q_M := \frac{V}{M} \sum_{k=0}^{M} z_{i_k} j_k,$$

where i_k and j_k are randomly chosen integers in the range [0, N]. (That is, for each k, choose a random i_k and j_k).

Try implementing this method, and comparing with the results for our other methods, in terms of both accuracy and time. You don't have to submit your work for this part.