## Lecture 16

niceguy

October 17, 2022

## 1 Improving Euler's Method

Recall Euler's method can be thought as approximating an integral

$$\int_{t_n}^{t_{n+1}} f(t, y) dt$$

We can improve Euler's method by finding a better approximation for f on the interval  $t \in [t_n, t_{n+1}]$ . Our original approximation is that f is approximately  $f(t_n, y(t_n))$ . A better approximate is to take the average value of f, i.e.

$$f(t, u(t)) \approx \frac{f(t_n, y(t_n)) + f(t_{n+1}, y(t_{n+11}))}{2}$$

However, we do not have access to  $y(t_{n+1})$ . Instead, we will use the approximated value of  $y(t_{n+1})$  using Euler's method, i.e.

$$f(t, y(t)) \approx \frac{f(t_n, y(t_n)) + f(t_{n+1}, y_n + (t_{n+1} - t_n f(t_n, y_n)))}{2}$$

We then have

$$y_{n+1} = y_n + \frac{f(n, y_n) + f(t_{n+1}, y_n + hf(t_n, y_n))}{2} \times h$$

where  $h = t_{n+1} - t_n$ .

However, there is a downside to this, as more calculations have to be performed for each successive approximation.

The Error comparisons are as follows

Method	Local Truncation Error	Global Truncation Error	Function Evaluations
Euler	$h^2$	h	1
Improved Euler	$h^3$	$h^2$	2

However, we can still improve on this. We first start with the slope  $f_1 = f(t_n, y(t_n))$ . Extending this to  $t_n + \frac{h}{2}$ , we get the slope  $f_2$ . Now put  $f_2$  at  $t_n$ , and extend it to get a second approximation for the slope at  $t_n + \frac{h}{2}$ , which is  $f_3$ . Finally,  $f_3$  is placed at  $t_n$  which is then extended to approximate the slope at  $t_{n_1}$ , which is  $f_4$ . We take a weighted average with  $f_1, f_2, f_3, f_4$  being given weights of 1, 2, 2, 1 respectively.

$$y_{n+1} = y_n + \frac{s_{n1} + 2s_{n2} + 2s_{n3} + s_{n4}}{6} \times h$$

where

$$s_{n1} = f(t_n, y_n)$$

$$s_{n2} = f(t_n + \frac{h}{2}, y_n + \frac{1}{2}hs_{n1})$$

$$s_{n3} = f(t_n + \frac{h}{2}, y_n + \frac{1}{2}hs_{n2})$$

$$s_{n4} = f(t_n + h, y_n + hs_{n3})$$

There are some shortcomings to using a constant step size. There might be certain bounds where a smaller step size (more oscillations) is desired, while there are other bounds where a larger step size is desired (approximately constant). Assuming we want the maximum truncation error to be  $\epsilon > 0$ . If the error is greater than that, we decrease the step size, and vice versa. We estimate the local truncation error by

$$e_{n+1} \approx |y_{n+1} - z_{n+1}|$$

where z is a better approximation than y.