

## PROBLEM SET #1

APC 523/MAE 507/AST 523 : Numerical Algorithms for Scientific Computing

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### 1 Error in (symmetric) rounding vs chopping

**Assertion:** When mapping a real number  $x$  to a nearby machine number in  $\mathbb{R}(p, q)$ , the upper bound in the relative error for symmetric rounding is:

$$\left| \frac{x - \text{rd}(x)}{x} \right| \leq 2^{-p}$$

**Proof:**

Consider the number  $x$  to be represented as:

$$x = \pm \left( \sum_{l=1}^{\infty} b_{-l} 2^{-l} \right) 2^e$$

If the number is to be rounded to  $p$  terms, two cases arise:

**CASE I.** The  $(p+1)^{\text{th}}$  is 0.

In this scenario the difference between the true value and the rounded value is given by:

$$x - \text{rd}(x) = \pm \left( \sum_{l=p+2}^{\infty} b_{-l} 2^{-l} \right) 2^e$$

The maximum relative error can then be computed as:

$$\begin{aligned} \max \left| \frac{x - \text{rd}(x)}{x} \right| &= \frac{\max |x - \text{rd}(x)|}{\min |x|} \\ &= \frac{2^{-p-1} 2^e}{2^{-1} 2^e} \\ &= 2^{-p} \end{aligned}$$

which is what we set to prove.

**CASE II.** The  $(p+1)^{\text{th}}$  is 1.

In this scenario the maximum difference between the true and the rounded value is obtained as:

$$\max |x - \text{rd}(x)| = (2^{-p} - 2^{-p-1}) 2^e$$

This is the case we all the leading terms from  $(p+2)$  are 1. Hence the maximum relative error can be computed as before:

$$\begin{aligned} \max \left| \frac{x - \text{rd}(x)}{x} \right| &= \frac{\max |x - \text{rd}(x)|}{\min |x|} \\ &= \frac{(2^{-p} - 2^{-p-1}) 2^e}{2^{-1} 2^e} \\ &= 2^{-p} \end{aligned}$$

which is what we set to prove

Both the cases show that the maximum symmetric rounding off error is  $2^{-p}$

## 2 Recurrence in reverse

(a) The reverse recurrence relation is given by:

$$y_{n-1} = \frac{e - y_n}{n}$$

Computing for a few terms down the chain we obtain:

$$\begin{aligned} y_{n-2} &= \frac{e - y_{n-1}}{n-1} \\ &= \frac{ne - e + y_n}{n(n-1)} \\ y_{n-3} &= \frac{e - y_{n-2}}{n-2} \\ &= \frac{n(n-1)e - ne + e - y_n}{n(n-1)(n-2)} \end{aligned}$$

One can denote the pattern as:

$$y_{n-p} = (-1)^p \frac{y_n}{n(n-1)(n-2)\dots(n-p+1)} + e \left[ \frac{1}{n-(p-1)} - \frac{1}{(n-(p-1))(n-(p-2))} + \frac{1}{(n-(p-1))(n-(p-2))(n-(p-3))} + \dots \right]$$

To obtain the value of  $y_k$  in terms of  $y_N$  we replace  $n-p$  with  $k$  and simplify:

$$\begin{aligned} y_k &= (-1)^{n-k} \frac{y_n}{n(n-1)(n-2)\dots(k+1)} + \text{exponent terms} \\ &= (-1)^{n-k} \frac{y_n k!}{n!} + \text{exponent terms} \end{aligned}$$

The condition number is given as:

$$\begin{aligned} (\text{cond } g_k)(y_k) &= \left| \frac{y_N g'(y_N)}{y_k} \right| \\ &= \left| \frac{y_N \frac{k!}{N!}}{y_k} \right| \end{aligned}$$

Since the  $k$  is less than  $N$ ,  $y_k$  is greater than  $y_N$ , the upper bound on the condition number,  $(\text{cond } g_k)(y_k)$ , obtained as:

$$(\text{cond } g_k)(y_k) \leq \frac{k!}{N!}$$

as  $\frac{y_N}{y_k} \leq 1$

(b) We know the condition number represents:

$$\begin{aligned} \varepsilon_y &= (\text{cond } g_k) \varepsilon_x \\ \frac{\Delta y_k}{y_k} &\leq \frac{k!}{N!} \leq \varepsilon \\ N! &\geq \frac{k!}{\varepsilon} \end{aligned}$$

Here, we have assumed  $\varepsilon_x = 1$  and  $\varepsilon$  is a predefined target error in  $y_k$ .

- (c) For `python3` the machine epsilon for float is  $1.0e^{-15}$  (Obtained using `numpy.finfo(float)`). Using this machine epsilon the value of  $N$  obtained is 31. [Check code]
- (d) The computed value of  $y_{20}$  is 0.123803830762570 and the value of  $y_{20}$  directly by integration is 0.123803830762570. [Check code]