## ON A MODEL FOR COMPUTING ROUND-OFF ERROR OF A SUM

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Reproduction in whole or in part is permitted for any purpose of the United States Government. This document has been approved for public release and sale; its distribution is unlimited. Given real numbers  $a_1, a_2, \ldots, a_n$  we are interested in the classic problem of the error in computing  $S = \sum_{i=1}^{n} a_i$  when the sum is computed by  $S_0 = \sum_{i=1}^{n} a_i^*$  where  $a_i^*$  is the nearest integer to  $a_1$ . We shall first study this error as a function of a A shift, i.e., when all numbers  $a_i$  are each shifted A and then rounded;

(1) S - 
$$\mathbf{n}\Delta = \sum_{i=1}^{n} (\mathbf{a}_{i} - \mathbf{A})$$
  
(2)  $\tilde{\mathbf{S}}_{\Delta}$  -  $\mathbf{n}\Delta = \sum_{i=1}^{n} (\mathbf{a}_{i} - \mathbf{A})^{*}$ 

We will then let A become a random variable that can take on uniformly any value in the interval  $-\frac{1}{2} \leq A \leq +\frac{1}{2}$ . Different choices of A give rise to different rounding errors  $\tilde{S}_{\Delta}$  - s and the variance of the distribution of  $\tilde{S}_{\Delta}$  - s can be used to measure the variability of the <u>rounding error due to the random selection of</u> <u>the origin of the real numbers</u>  $a_i$  with respect to that of the computer.

The cumulative error from (1) and (2) is

(3) 
$$\tilde{s}_{\Delta} - s = \sum_{i=1}^{n} [(a_i - \Delta)^* - (a_i - A)]$$

Let  $f_i$  be the positive fractional part of  $a_i$  and let  $\alpha_i$  be the largest integer not exceeding  $a_i$ , i.e.,

(4) 
$$a_1 = \alpha_i + f_i$$

Denoting by  ${\bf r_i}$  the error of the i<sup>th</sup> term, we have

(5) ri = [(ai-A)\* - (a<sub>i</sub>-A)] = 
$$\begin{cases} 1 - (f_i - \Delta) & \text{if } -\frac{1}{2} \le A \le -\frac{1}{2} + f_i \\ -(f_i - \Delta) & \text{if } -\frac{1}{2} + f_i \le A \le +\frac{1}{2} \end{cases}$$

To prove the above, we note that  $f_i - \Delta = (a_i - \Delta) + \alpha_i$ . If  $-\frac{1}{2} \leq f_i - \Delta \leq +\frac{1}{2}$  then  $(a_i - \Delta)$  is rounded to  $a_i$ . Hence  $a_i - \Delta$  is rounded down if  $-\frac{1}{2} + f_i \leq A$  otherwise rounded up.

Denoting expected value by  $\ensuremath{\ensuremath{\mathsf{E}}}$  , we have by direct evaluation

(6) 
$$E(r_i) = \int_{-\frac{1}{2}}^{+\frac{1}{2}} r_i d\Delta = 0$$

Assume  $f_{i - j}$ , then

$$E(\mathbf{r}_{i}\mathbf{r}_{j}) = \int_{-\frac{1}{2}}^{-\frac{1}{2}+f_{i}} \mathbf{r}_{i}\mathbf{r}_{j}dA + \int_{-\frac{1}{2}+f_{i}}^{-\frac{1}{2}-f_{j}} \mathbf{r}_{i}\mathbf{r}_{j}dA + \int_{-\frac{1}{2}+f_{j}}^{+\frac{1}{2}} \mathbf{r}_{i}\mathbf{r}_{j}dA$$

$$= \int_{-\frac{1}{2}}^{+\frac{1}{2}} (f_{i}f_{j} - \Delta(f_{i} + f_{j}) + \Delta^{2}) d\Delta$$

$$+ \int_{-\frac{1}{2}}^{-\frac{1}{2}+f_{i}} [(1-f_{i}-f_{j}) + 2\Delta] d\Delta$$

$$+ \int_{-\frac{1}{2}+f_{j}}^{-\frac{1}{2}+f_{j}} [-f_{i} + A) d\Delta$$

Performing indicated integration yields:

(7) 
$$E(\mathbf{r}_{i}\mathbf{r}_{j}) = \frac{1}{2}[|\mathbf{f}_{j}-\mathbf{f}_{i}|^{2} - |\mathbf{f}_{j}-\mathbf{f}_{i}| + \frac{1}{6}]$$

which is one-half the 2<sup>nd</sup> order Bernoulli Polynomial in  $|f_j - f_i|$ . For  $f_j \leq f_i$  we also get (7). Note that the individual errors  $r_i$  and  $r_j$  are not independent of one another.

It now follows that

(9) E(S) = S

(10) 
$$E(S-S)^2 = E(\sum_{i=1}^n \sum_{j=1}^n r_i r_j) = \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n [|f_i - f_j|^2 - |f_i - f_j| + \frac{1}{6}]$$

The usual value of variance,  $E(S-S)^2 = n/12$ , will result if we further assume f<sub>i</sub> are independently drawn from uniform distributions on  $[0 \le f_i \le 1]$ .

<u>Theorem</u>: If the fractional parts of all  $a_i$  are equal to each other, then each term of (10) is maximum for  $0 \le f_i \le 1$  and

(11) Max 
$$E(S-S)^2 = \frac{1}{2} \sum_{i=1}^{n} \sum_{j=1}^{n} (\frac{1}{6}) = \frac{n^2}{12}$$

From (10) we have an interesting inequality, namely for all  $f_i$ 

(12) 
$$V(f) = \frac{1}{2} \sum_{i=1}^{n} \sum_{j=1}^{n} \{|f_i - f_j|^2 - |f_i - f_j| + \frac{1}{6}\} \ge 0$$

This function is not convex even for n=2, since  $f^{(1)} = (\frac{1}{2}, 0)$  and  $f^{(2)} = (-\frac{1}{2}, 0)$  yields  $V(f^0) = V(f^1) = \frac{1}{12} + \frac{1}{12} - \frac{1}{12} = \frac{1}{12}$  but

3

 $V(\frac{f^1 + f^2}{2}) = V(0) = \frac{3}{12}$ . There appears to be no obvious direct way to establish that  $V(f) \ge 0$  for all  $0 \le f_i \le 1$ . Our development shows V(f) to be a variance and this, of course, constitutes an indirect proof. We can replace (12) by a convex realization: Assume  $f_i \ge f_{i+1}$  for all i, then the problem of finding Min V(f) can be rewritten:

(13) Find Min 
$$[V(f)] = \sum_{i < j} (f_i - f_j)^2 + \frac{n^2}{12} - [(n-1)f_1 + (n-3)f_2 + (n-5)f_3 + ... + (n-2k+1)f_k + ... - (n-1)f_n]$$

subject to

(14) 
$$f_1 > f_2 \cdot \cdot \cdot > f_n$$
  
(15)  $0 \le f_i \le 1$ 

Formally (13), (14), (15), is a positive definite quadratic program. Fortunately, as we shall see this can be solved by classical calculus by ignoring inequalities (14) and (15).

<u>Theorem</u>: Equally spaced  $f_i = (n - i)/n$ , (i = 1,...,n) yields Min V(f) =  $\frac{1}{12}$  independent of n, i.e., the variance of the sum in this case is minimum and is the same as the variance of the individual terms forming the sum.

Proof: Setting partials = 0 in (13) yields:

$$(16) \begin{cases} 2(n-1)f_1 & -2f_2 & \cdots & -2f_n & = (n-1) \\ -2f_1 & +2(n-1)f_2 & -2f_n & = (n-3) \\ -2f_1 & -2f_2 & \cdots & 2(n-1)f_{n-1} - 2f_n & = -(n-3) \\ -2f_1 & -2f_2 & 2(n-1)f_n & = -(n-1) \end{cases}$$

Adding shows the equations to be dependent. Hence we may drop the last equation as redundant. Moreover, we can always translate the  $f_i$  so that the smallest  $f_i$ , namely  $f_n = 0$ 

Re-adding yields:

$$2f_1 + 2f_2 + \dots 2f_n + 0 = (n-1)$$
,  $f_n = 0$ .

Adding this last equation to each of the others gives

$$2nf_{i} = (n - 2i + 1) + (n - 1) = 2(n - i)$$

(17) 
$$f_{i} = (n - i)/n$$

Evidently the conditions  $0 \leq f_i \leq 1$  and  $f_i \geq f_{i+1}$  are (by good luck) also satisfied so that (17) yields the minimum, namely

(18) Min V(f) = 
$$\frac{n^2}{12} - \frac{1}{2} \sum_{i=1}^{n} (n-2_i+1)f_i = \frac{1}{12}$$