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1 IEEE Stuff

Algorithm 1 EFT of the sum of two floating point numbers.

```
\begin{aligned} & \text{function } [x,y] = \texttt{TwoSum}(a,b) \\ & x = a \oplus b \\ & z = x \ominus a \\ & y = (a \ominus (x \ominus z)) \oplus (b \ominus z) \\ & \text{end function} \end{aligned}
```

Algorithm 2 Splitting of a floating point number into two parts.

```
function [x,y] = \mathtt{Split}(a)

z = a \otimes (2^r + 1)

x = z \ominus (z \ominus a)

y = a \ominus x

end function
```

Algorithm 3 EFT of the product of two floating point numbers.

```
\begin{aligned} & \textbf{function} \; [x,y] = \texttt{TwoProd}(a,b) \\ & x = a \otimes b \\ & [a_h,a_l] = \texttt{Split}(a) \\ & [b_h,b_l] = \texttt{Split}(b) \\ & y = a_l \otimes b_l \ominus (((x\ominus a_h \otimes b_h) \ominus a_l \otimes b_h) \ominus a_h \otimes b_l) \\ & \textbf{end function} \end{aligned}
```

Algorithm 4 EFT of the sum of two floating point numbers with a FMA.

```
\begin{aligned} & \text{function } [x,y] = \texttt{TwoProdFMA}(a,b) \\ & x = a \otimes b \\ & y = \texttt{FMA}(a,b,-x) \\ & \text{end function} \end{aligned}
```

Algorithm 5 de Casteljau algorithm for polynomial evaluation.

```
function \operatorname{res} = \operatorname{deCasteljau}(p,s) r = 1 \ominus s \operatorname{for} \ j = 0, \dots, n \ \operatorname{do} b_j^{(n)} = p_j end for \operatorname{for} \ k = n-1, \dots, 0 \ \operatorname{do} for j = 0, \dots, k \ \operatorname{do} b_j^{(k)} = r \otimes b_j^{(k+1)} \oplus s \otimes b_{j+1}^{(k+1)} end for end for end function
```

2 de Casteljau's Method

Consider de Casteljau's method to evaluate a degree n polynomial in Bernstein-Bézier form with control points p_i :

$$b_j^{(0)} = p_j$$

$$b_j^{(k)} = (1 - s)b_j^{(k-1)} + sb_{j+1}^{(k-1)}$$

$$b(s) = b_0^{(n)}.$$

2.1 Condition Number

For a polynomial p(x) in the power basis, we have ([LGL06]):

cond
$$(p(x)) = \frac{\tilde{p}(|x|)}{|p(x)|} = \frac{\sum_{j} |a_{j}| |x|^{j}}{|p(x)|}.$$

In particular, this means that if $x \ge 0$ and each $a_j \ge 0$ we must necessarily have cond (p(x)) = 1. To see an example in use, consider $p(x) = (x-1)^n$ and input values of the form $x = 1 + \delta$ (with $|\delta| \ll 1$). Since $a_j = \binom{n}{j} (-1)^{n-j}$ we have $\widetilde{p}(x) = (x+1)^n$ hence

$$\operatorname{cond}\left(p\left(1+\delta\right)\right) = \frac{(2+\delta)^n}{\left|\delta\right|^n} = \left|1 + \frac{2}{\delta}\right|^n.$$

As $\delta \to 0$, this value approaches ∞ (as expected).

For a polynomial p(s) in Bernstein form, we have ([JLCS10]):

$$\operatorname{cond}\left(p(s)\right) = \frac{\widetilde{p}\left(s\right)}{\left|p(s)\right|} = \frac{\sum_{j}\left|p_{j}\right|\left|b_{j,n}(s)\right|}{\left|p(s)\right|}.$$

The Bernstein form is suited for $s \in [0,1]$, which means $b_{j,n}(s) \ge 0$ typically. If $s \in [0,1]$ and each $p_j \ge 0$ we must necessarily have cond (p(s)) = 1. To see an example in use, consider

$$p(s) = (1 - 2s)^n = [(1 - s) - s]^n = \sum_{j} \binom{n}{j} (1 - s)^{n-j} (-s)^j = \sum_{j} (-1)^j b_{j,n}(s)$$

and input values of the form $x = \frac{1}{2} + \delta$ (with $|\delta| \ll \frac{1}{2}$). Since $p_j = (-1)^j$ we have $\widetilde{p}(s) = [(1-s) + s]^n = 1$

$$\operatorname{cond}\left(p\left(\frac{1}{2} + \delta\right)\right) = \frac{1}{|2\delta|^n}.$$

As $\delta \to 0$, this value approaches ∞ (as expected).

2.2 Selection of Test Cases

From [DP15] (end of Section 3):

We can observe that, in this case, the algorithm with a good behavior everywhere is the de Casteljau algorithm

In the same paper (when referring to [Bez13] at the beginning of Section 2):

assuming that all control points are positive. This assumption avoided ill-conditioned polynomials. In this section, we shall show that this is a natural assumption in Computer Aided Geometric Design (from now on, C.A.G.D.) and that it permits to assure high relative precision for the evaluation through a large family of representations in C.A.G.D.

From the same author, in [MP05] (towards the end of Section 5, at the bottom of page 109):

Let us observe that in this case, the de Casteljau algorithm presents better stability properties for the evaluation near the roots. In fact, the de Casteljau algorithm has good behaviour even when using simple precision, although the running error bound is not so accurate in points close to the roots.

2.3 K-Fold Error Filtering

After implementing for K = 2, 3, ..., 12 and instrumenting all relevant floating point operations, the K-fold Horner requires

$$(5 \cdot 2^K - 8)n + ((K+8)2^K - 12K - 6) = \mathcal{O}((n+K)2^K)$$

flops to evaluate a degree n polynomial (this only applies when $n \ge K - 1$). As a comparison, the non-compensated form of Horner requires 2n flops. Of these, $(2^{K-1} - 1)n - 2^{K-1}(K - 3) - 2$ are FMA (fused-multiply-add) instructions.

After implementing for K = 2, 3, 4, 5 and instrumenting all relevant floating point operations, the K-fold de Casteljau requires

$$(15K^2 - 34K + 26)T_n + K + 5 = \mathcal{O}(n^2K^2)$$

flops to evaluate a degree n polynomial. (Here T_n is the nth triangular number.) As a comparison, the non-compensated form of de Casteljau requires $3T_n + 1$ flops. Of these, $(3K - 4)T_n$ are FMA instructions. On hardware that doesn't support FMA, every FMA will be exchanged for $10 \ominus$'s and $6 \otimes$'s so the count will increase by $(10 + 6 - 1)(3K - 4)T_n$.

3 Bogus Section for Refs

Here they are, for now

- Compensated Horner (K = 2) ([LGL06])
- Compensated de Casteljau ([JLCS10])
- Newton with compensated Horner ([Gra08])
- *K*-fold Sum ([ORO05])
- K-fold Horner ([GLL09])

References

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- [ORO05] Takeshi Ogita, Siegfried M. Rump, and Shin'ichi Oishi. Accurate sum and dot product. SIAM Journal on Scientific Computing, 26(6):1955–1988, jan 2005.