## ON THE SAUER-XU FORMULA FOR THE ERROR IN MULTIVARIATE POLYNOMIAL INTERPOLATION

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ABSTRACT. Use of a new notion of multivariate divided difference leads to a short proof of a formula by Sauer and Xu for the error in interpolation, by polynomials of total degree  $\leq n$  in d variables, at a 'correct' point set.

It is the purpose of this note to give a short proof of a remarkable formula for the error in polynomial interpolation given in [3].

In [3], Sauer and Xu consider interpolation from the space  $\Pi_n(\mathbf{R}^d)$  of d-variate polynomials of degree  $\leq n$  at a point set  $\mathcal{X}$  which is **correct** for it in the sense that, for an arbitrary q, there is exactly one  $p \in \Pi_n(\mathbf{R}^d)$ , denoted here by

$$P_n g$$

which agrees with g on  $\mathcal{X}$ . Such a correct point set can, as Sauer and Xu point out, be partitioned into subsets

$$x^{(i)} := \{x_r^{(i)} : r = 1, \dots, r_i^d := \dim \Pi_i - \dim \Pi_{i-1} = \binom{i-1+d}{i}\}, \quad i = 0, 1, \dots, n,$$

in such a way that, for each  $j \leq n$ , polynomial interpolation from  $\Pi_j$  at the points in

$$x^{(\leq j)} := x^{(0)} \cup \cdots \cup x^{(j)}$$

is uniquely possible. They denote the corresponding Lagrange polynomial in  $\Pi_j$  associated with the point  $x_r^{(j)}$  by

$$n^{[j]}$$
.

i.e.,  $p_r^{[j]}$  is the unique element of  $\Pi_j$  which satisfies

$$p_r^{[j]}(x_s^{(i)}) = \delta_{ji}\delta_{rs}, \quad s = 1, \dots, r_i^d; \ i = 0, \dots, j,$$

but is called a Newton polynomial in [3], probably because any  $p \in \Pi_j$  which vanishes on  $x^{(< j)}$  can be written in the form

(1) 
$$p = \sum_{r=1}^{r_j^d} p_r^{[j]} p(x_r^{(j)}).$$

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Let  $E_i g$  denote the error in the polynomial interpolant from  $\Pi_i$  to g, i.e.,

$$g = P_i g + E_i g$$
.

Sauer and Xu prove that (in roughly the same notation, except for a slight reordering and the use of a divided difference here)

(2) 
$$(E_j g)(x) = \sum_{r=1}^{r_j^d} p_r^{[j]}(x) \sum_{\mu \in \Lambda_r^{(j)}} c_{\mu}[x_{\mu}, x; \Delta x_{\mu}, x - x_r^{(j)}]g,$$

with the following definition of the various quantities appearing here:

$$\Lambda_r^{(j)} := \{ \mu \in \times_{i=0}^j \{1, \dots, r_i^d\} : \mu_j = r \};$$

$$c_{\mu} := \prod_{i=0}^{j-1} p_{\mu_i}^{[i]}(x_{\mu_{i+1}}^{(i+1)});$$

$$x_{\mu} := (x_{\mu_i}^{(i)} : i = 0, \dots, j), \qquad \Delta x_{\mu} := (x_{\mu_{i+1}}^{(i+1)} - x_{\mu_i}^{(i)} : i = 0, \dots, j-1);$$

and, finally,  $[t_0, \ldots, t_j; \xi_1, \ldots, \xi_j]$  is the *j*th divided difference introduced in [1], i.e.,

(3) 
$$[t_0, \dots, t_j; \xi_1, \dots, \xi_j]g := \int_{[t_0, \dots, t_j]} D_{\xi_1} \dots D_{\xi_j} g,$$

with

$$f \mapsto \int_{[t_0,\dots,t_j]} f := \int_0^1 \int_0^{s_1} \dots \int_0^{s_{j-1}} f(t_0 + s_1 \nabla t_1 + \dots + s_j \nabla t_j) \, \mathrm{d}s_j \dots \, \mathrm{d}s_1$$

the linear functional which was used to such advantage by Micchelli in his analysis of Kergin interpolation and the simplex spline, and which has been dubbed by him in [2] the **divided difference functional for R**<sup>d</sup>. The only facts about the *j*th divided difference (3) needed here are that it is symmetric in the *points*  $t_0, \ldots, t_j$ , and is symmetric and *linear* in the *directions*  $\xi_1, \ldots, \xi_j$  (which is obvious), and that, for an arbitrary point sequence T, points a, b, and arbitrary direction sequence  $\Xi$  (with  $\#T = \#\Xi$ ),

$$[T, a, b; \Xi, a - b] = [T, a; \Xi] - [T, b; \Xi],$$

which can be verified directly (see [1]).

The short proof of (2) about to be given here is by induction (as is the proof in [3]). For j = 0, (2) is the special case  $T = () = \Xi$ , a = x,  $b = x_1^{(0)}$  of (4). Assuming (2) to hold for j = n, we observe that

$$(E_n g)(x) = G_n(x, x),$$

with

$$G_n(x,y) := \sum_{r=1}^{r_n^d} p_r^{[n]}(x) \sum_{\mu \in \Lambda_r^{(n)}} c_\mu [x_\mu, y; \Delta x_\mu, x - x_r^{(n)}] g.$$

Now note that, for arbitrary y,  $G_n(\cdot, y)$  is a polynomial of degree  $\leq n+1$  (since the  $p_r^{[n]}$  are in  $\Pi_n$  while  $x \mapsto [T; \Xi, x]g$  is a linear (scalar-valued) function), and  $G_n(\cdot, y)$  vanishes at every  $x_r^{(n)}$ , hence, as in (1),  $G_n(\cdot, y)$  is writeable as

$$G_n(\cdot, y) = \sum_{s=1}^{r_{n+1}^d} p_s^{[n+1]} G_n(x_s^{(n+1)}, y).$$

On the other hand, since  $g = P_n g + E_n g$ , the function

$$P_n g + \sum_{s=1}^{r_{n+1}^d} p_s^{[n+1]} (E_n g)(x_s^{(n+1)})$$

agrees with g at  $x^{(\leq n+1)}$  and is in  $\Pi_{n+1}$ , hence must equal  $P_{n+1}g$ . Therefore,

$$E_{n+1}g = E_n g - \sum_s p_s^{[n+1]} (E_n g)(x_s^{(n+1)})$$

$$= \sum_s p_s^{[n+1]} (G_n(x_s^{(n+1)}, \cdot) - G_n(x_s^{(n+1)}, x_s^{(n+1)})),$$

and this implies (2) for j = n + 1 since

$$G_n(x_s^{(n+1)}, \cdot) - G_n(x_s^{(n+1)}, x_s^{(n+1)}) = \sum_{r=1}^{r_n^d} p_r^{[n]}(x_s^{(n+1)}) \sum_{\mu \in \Lambda_r^{(n)}} c_\mu d_{\mu,r} g$$

with

$$d_{\mu,r} := [x_{\mu}, \cdot; \Delta x_{\mu}, x_s^{(n+1)} - x_r^{(n)}] - [x_{\mu}, x_s^{(n+1)}; \Delta x_{\mu}, x_s^{(n+1)} - x_r^{(n)}]$$
$$= [x_{\mu}, x_s^{(n+1)}, \cdot; \Delta x_{\mu}, x_s^{(n+1)} - x_r^{(n)}, \cdot - x_s^{(n+1)}],$$

by (4).

It follows that

$$P_n g = \sum_{j=0}^{n} \sum_{r=1}^{r_j^d} p_r^{[j]} (E_{j-1}g)(x_r^{(j)}),$$

with

$$E_{-1}g := g.$$

Since  $(p_r^{[j]}: r = 1, \ldots, r_j^d; j = 0, 1, \ldots)$  is linearly independent, it follows that, for any j,  $[x;]E_j$  is a linear combination of the linear functionals [x;] and  $[x_r^{(i)};]$ ,  $r = 1, \ldots, r_i^d; i = 0, \ldots, j$ . However, its constituents, i.e., the jth divided differences

$$[x_{\mu}, x; \Delta x_{\mu}, x - x_{\mu_j}^{(j)}]$$

by themselves, are not necessarily such linear combinations, as the simple example

$$[0, \mathbf{i}_1, \mathbf{i}_1 + \mathbf{i}_2; \mathbf{i}_1, \mathbf{i}_2],$$

with  $\mathbf{i}_j$  the jth unit vector, readily shows.

## References

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