## On the randomized complexity of Banach space valued integration

by

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Dedicated to Albrecht Pietsch on the occasion of his 80th birthday

**Abstract.** We study the complexity of Banach space valued integration in the randomized setting. We are concerned with r times continuously differentiable functions on the *d*-dimensional unit cube Q, with values in a Banach space X, and investigate the relation of the optimal convergence rate to the geometry of X. It turns out that the *n*th minimal errors are bounded by  $cn^{-r/d-1+1/p}$  if and only if X is of equal norm type p.

1. Introduction. Integration of scalar valued functions is an intensively studied topic in the theory of information-based complexity (see [12], [10], [11]). Motivated by applications to parametric integration, recently the complexity of Banach space valued integration was considered in [2]. It was shown that the behaviour of the *n*th minimal errors  $e_n^{ran}$  of randomized integration in  $C^r(Q, X)$  is related to the geometry of the Banach space X in the following way: The infimum of the exponents of the rate is determined by the supremum of p such that X is of type p. In the present paper we further investigate this relation. We establish a connection between *n*th minimal errors and equal norm type p constants for n vectors. It follows that  $e_n^{ran}$  is bounded by  $cn^{-r/d-1+1/p}$  if and only if X is of equal norm type p.

**2. Preliminaries.** Let  $\mathbb{N} = \{1, 2, ...\}$  and  $\mathbb{N}_0 = \{0, 1, 2, ...\}$ . We introduce some notation and concepts from Banach space theory needed in what follows. For Banach spaces X and Y let  $B_X$  be the closed unit ball of X and

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 $\mathscr{L}(X,Y)$  the space of bounded linear operators from X to Y, endowed with the usual norm. If X = Y, we write  $\mathscr{L}(X)$ . The norm of X is denoted by  $\|\cdot\|$ , while other norms are distinguished by subscripts. We assume that all the Banach spaces considered are defined over the same scalar field  $\mathbb{K} = \mathbb{R}$ or  $\mathbb{K} = \mathbb{C}$ .

Let  $Q = [0, 1]^d$  and let  $C^r(Q, X)$  be the space of all r times continuously differentiable functions  $f : Q \to X$  equipped with the norm

$$\|f\|_{C^{r}(Q,X)} = \max_{0 \le |\alpha| \le r, t \in Q} \|D^{\alpha}f(t)\|,$$

where  $\alpha = (\alpha_1, \ldots, \alpha_d), |\alpha| = |\alpha_1| + \cdots + |\alpha_d|$  and  $D^{\alpha}$  denotes the respective partial derivative. For r = 0 we write  $C^0(Q, X) = C(Q, X)$ , which is the space of continuous X-valued functions on Q. If  $X = \mathbb{K}$ , we write  $C^r(Q)$ and C(Q).

Let  $1 \le p \le 2$ . A Banach space X is said to be of (Rademacher) type p if there is a constant c > 0 such that for all  $n \in \mathbb{N}$  and  $x_1, \ldots, x_n \in X$ ,

(1) 
$$\left(\mathbb{E}\left\|\sum_{i=1}^{n}\varepsilon_{i}x_{i}\right\|^{p}\right)^{1/p} \leq c\left(\sum_{k=1}^{n}\|x_{i}\|^{p}\right)^{1/p},$$

where  $(\varepsilon_i)_{i=1}^n$  is a sequence of independent Bernoulli random variables with  $\mathbb{P}\{\varepsilon_i = -1\} = \mathbb{P}\{\varepsilon_i = +1\} = 1/2$  on some probability space  $(\Omega, \Sigma, \mathbb{P})$  (we refer to [9, 7] for this notion and related facts). The smallest constant satisfying (1) is called the *type p constant* of X and is denoted by  $\tau_p(X)$ . If there is no such c > 0, we put  $\tau_p(X) = \infty$ . The space  $L_{p_1}(\mathcal{N}, \nu)$  with  $(\mathcal{N}, \nu)$  an arbitrary measure space and  $p_1 < \infty$  is of type p with  $p = \min(p_1, 2)$ .

Furthermore, given  $n \in \mathbb{N}$ , let  $\sigma_{p,n}(X)$  be the smallest c > 0 for which (1) holds for any  $x_1, \ldots, x_n \in X$  with  $||x_1|| = \cdots = ||x_n||$ . The contraction principle for Rademacher series (see [7, Th. 4.4]) implies that  $\sigma_{p,n}(X)$  is the smallest constant c > 0 such that for  $x_1, \ldots, x_n \in X$ ,

(2) 
$$\left(\mathbb{E}\left\|\sum_{i=1}^{n}\varepsilon_{i}x_{i}\right\|^{p}\right)^{1/p} \leq cn^{1/p}\max_{1\leq i\leq n}\|x_{i}\|.$$

We say that X is of equal norm type p if there is a constant c > 0 such that  $\sigma_{p,n}(X) \leq c$  for all  $n \in \mathbb{N}$ . Clearly,  $\sigma_{p,n}(X) \leq \tau_p(X)$  and type p implies equal norm type p.

Let us comment a little more on the relation of the different notions of type which are used here and in the literature. The concept of equal norm type p was first introduced and used by R. C. James in the case p = 2 in [6]. There it is shown that X is of equal norm type 2 if and only if X is of type 2. This result is attributed to G. Pisier. Later, it even turned out in [1] that

the sequence  $\sigma_{2,n}(X)$  and the corresponding sequence  $\tau_{2,n}(X)$  of type 2 constants computed with *n* vectors are uniformly equivalent. In contrast, for 1 , L. Tzafriri [13] constructed Tsirelson spaces without type*p*but with equal norm type*p*. Finally, V. Mascioni introduced and studied the notion of weak type*p*for <math>1 in [8] and showed that, again in contrast to the situation for <math>p = 2, a Banach space X is of weak type *p* if and only if it is of equal norm type *p*.

Throughout the paper  $c, c_1, c_2, \ldots$  are constants, which depend only on the problem parameters r, d, but depend neither on the algorithm parameters n, l etc. nor on the input f. The same symbol may denote different constants, even in a sequence of relations.

For  $r, k \in \mathbb{N}$  we let  $P_k^{r,X} \in \mathscr{L}(C(Q,X))$  be X-valued composite tensor product Lagrange interpolation of degree r with respect to the partition of  $[0,1]^d$  into  $k^d$  subcubes of sidelength  $k^{-1}$  with disjoint interiors (see [2]). Given  $r \in \mathbb{N}_0$  and  $d \in \mathbb{N}$ , there are constants  $c_1, c_2 > 0$  such that for all Banach spaces X and all  $k \in \mathbb{N}$ ,

(3) 
$$\sup_{f \in B_{C^r(Q,X)}} \|f - P_k^{r,X}f\|_{C(Q,X)} \le c_2 k^{-r}$$

(see [2]).

**3. Banach space valued integration.** Let X be a Banach space,  $r \in \mathbb{N}_0$ , and let the integration operator  $S^X : C(Q, X) \to X$  be given by

$$S^X f = \int_Q f(t) \, dt.$$

We will work in the setting of information-based complexity theory (see [12, 10, 11]). Below,  $e_n^{\text{det}}(S^X, B_{C^r(Q,X)})$  and  $e_n^{\text{ran}}(S^X, B_{C^r(Q,X)})$  denote the *n*th minimal error of  $S^X$  on  $B_{C^r(Q,X)}$  in the deterministic, respectively randomized, setting, that is, the minimal possible error among all deterministic, respectively randomized, algorithms approximating  $S^X$  on  $B_{C^r(Q,X)}$  that use at most *n* values of the input function *f*. The precise notions are recalled in the appendix. The following was shown in [2].

THEOREM 1. Let  $r \in \mathbb{N}_0$  and  $1 \leq p \leq 2$ . Then there are constants  $c_{1-4} > 0$  such that for all Banach spaces X and  $n \in \mathbb{N}$ , the deterministic *n*th minimal error satisfies

$$c_1 n^{-r/d} \le e_n^{\det}(S^X, B_{C^r(Q,X)}) \le c_2 n^{-r/d}.$$

Moreover, if X is of type p and  $p_X$  is the supremum of all  $p_1$  such that X is of type  $p_1$ , then the randomized nth minimal error fulfills

$$c_3 n^{-r/d-1+1/p_X} \le e_n^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}) \le c_4 \tau_p(X) n^{-r/d-1+1/p}.$$

As a consequence, we obtain

COROLLARY 1. Let  $r \in \mathbb{N}_0$  and  $1 \leq p \leq 2$ . Then the following are equivalent:

(i) X is of type  $p_1$  for all  $p_1 < p$ .

(ii) For each 
$$p_1 < p$$
 there is a constant  $c > 0$  such that for all  $n \in \mathbb{N}$ ,

$$e_n^{\mathrm{ran}}(S^X, B_{C^r(Q,X)}) \le cn^{-r/d - 1 + 1/p_1}$$

The main result of the present paper is the following

THEOREM 2. Let  $1 \leq p \leq 2$  and  $r \in \mathbb{N}_0$ . Then there are constants  $c_1, c_2 > 0$  such that for all Banach spaces X and all  $n \in \mathbb{N}$ ,

(4) 
$$c_1 n^{r/d+1-1/p} e_n^{\operatorname{ran}}(S^X, B_{C^r(Q,X)})$$
  
 $\leq \sigma_{p,n}(X) \leq c_2 \max_{1 \leq k \leq n} k^{r/d+1-1/p} e_k^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}).$ 

This allows us to sharpen Corollary 1:

COROLLARY 2. Let  $r \in \mathbb{N}_0$  and  $1 \leq p \leq 2$ . Then the following are equivalent:

- (i) X is of equal norm type p.
- (ii) There is a constant c > 0 such that for all  $n \in \mathbb{N}$ ,

$$e_n^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}) \le cn^{-r/d-1+1/p}.$$

Recall from the preliminaries that the conditions in the corollary are also equivalent to

(iii) X is of type 2 if p = 2 and of weak type p if 1 .

For the proof of Theorem 2 we need a number of auxiliary results. The following lemma is a slight modification of Prop. 9.11 of [7], with essentially the same proof, which we include for the sake of completeness.

LEMMA 1. Let  $1 \leq p \leq 2$ . Then there is a constant c > 0 such that for each Banach space X, each  $n \in \mathbb{N}$  and each sequence of independent, essentially bounded, mean zero X-valued random variables  $(\eta_i)_{i=1}^n$  on some probability space  $(\Omega, \Sigma, \mathbb{P})$ ,

$$\left(\mathbb{E}\left\|\sum_{i=1}^{n}\eta_{i}\right\|^{p}\right)^{1/p} \leq c\sigma_{p,n}(X)n^{1/p}\max_{1\leq i\leq n}\|\eta_{i}\|_{L_{\infty}(\Omega,\mathbb{P},X)}.$$

*Proof.* Let  $(\varepsilon_i)_{i=1}^n$  be independent, symmetric Bernoulli random variables on some probability space  $(\Omega', \Sigma', \mathbb{P}')$  different from  $(\Omega, \Sigma, \mathbb{P})$ . Considering  $(\eta_i)_{i=1}^n$  and  $(\varepsilon_i)_{i=1}^n$  as random variables on the product probability space, we denote the expectation with respect to  $\mathbb{P}'$  by  $\mathbb{E}'$  (and the expectation with respect to  $\mathbb{P}$ , as before, by  $\mathbb{E}$ ). Using Lemma 6.3 of [7] and (2),

we get

$$\begin{split} \left(\mathbb{E} \left\| \sum_{i=1}^{n} \eta_{i} \right\|^{p} \right)^{1/p} &\leq 2 \left(\mathbb{E} \mathbb{E}' \left\| \sum_{i=1}^{n} \varepsilon_{i} \eta_{i} \right\|^{p} \right)^{1/p} \\ &\leq 2 \sigma_{p,n}(X) n^{1/p} \left(\mathbb{E} \max_{1 \leq i \leq n} \|\eta_{i}\|^{p} \right)^{1/p} \\ &\leq 2 \sigma_{p,n}(X) n^{1/p} \max_{1 \leq i \leq n} \|\eta_{i}\|_{L_{\infty}(\Omega, \mathbb{P}, X)}. \end{split}$$

Next we introduce an algorithm for the approximation of  $S^X f$ . Let  $n \in \mathbb{N}$ and let  $\xi_i : \Omega \to Q$  (i = 1, ..., n) be independent random variables on some probability space  $(\Omega, \Sigma, \mathbb{P})$ , uniformly distributed on Q. For  $f \in C(Q, X)$ define

(5) 
$$A_{n,\omega}^{0,X} f = \frac{1}{n} \sum_{i=1}^{n} f(\xi_i(\omega))$$

and, if  $r \ge 1$ , put  $k = \lceil n^{1/d} \rceil$  and

(6) 
$$A_{n,\omega}^{r,X}f = S^X(P_k^{r,X}f) + A_{n,\omega}^{0,X}(f - P_k^{r,X}f).$$

These are the Banach space valued versions of the standard Monte Carlo method (r = 0) and the Monte Carlo method with separation of the main part  $(r \ge 1)$ . The following extends the second part of Proposition 1 of [2].

PROPOSITION 1. Let  $r \in \mathbb{N}_0$  and  $1 \leq p \leq 2$ . Then there is a constant c > 0 such that for all Banach spaces  $X, n \in \mathbb{N}$ , and  $f \in C^r(Q, X)$ ,

(7) 
$$(\mathbb{E} \| S^X f - A_{n,\omega}^{r,X} f \|^p)^{1/p} \le c \sigma_{p,n}(X) n^{-r/d - 1 + 1/p} \| f \|_{C^r(Q,X)}.$$

*Proof.* Let us first consider the case r = 0. Let  $f \in C(Q, X)$  and put

$$\eta_i(\omega) = \int_Q f(t) \, dt - f(\xi_i(\omega)).$$

Clearly,  $\mathbb{E}\eta_i(\omega) = 0$ ,

$$S^{X}f - A_{n,\omega}^{0,X}f = \frac{1}{n}\sum_{i=1}^{n}\eta_{i}(\omega) \text{ and } \|\eta_{i}(\omega)\| \leq 2\|f\|_{C(Q,X)}.$$

An application of Lemma 1 gives (7). If  $r \ge 1$ , we have

$$S^{X}f - A_{n,\omega}^{r,X}f = S^{X}(f - P_{k}^{r,X}f) - A_{n,\omega}^{0,X}(f - P_{k}^{r,X}f)$$

and the result follows from (3) and the case r = 0.

LEMMA 2. Let  $1 \le p \le 2$ . Then there are constants c > 0 and  $0 < \gamma < 1$ such that for each Banach space X, each  $n \in \mathbb{N}$ , and  $(x_i)_{i=1}^n \subset X$  there is a subset  $I \subseteq \{1, \ldots, n\}$  with  $|I| \ge \gamma n$  and

$$\mathbb{E} \left\| \sum_{i \in I} \varepsilon_i x_i \right\| \le c n^{1/p} \|(x_i)\|_{\ell_{\infty}^n(X)} \max_{1 \le k \le n} k^{r/d+1-1/p} e_k^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}).$$

$$\begin{split} \textit{Proof. Since for all } n \in \mathbb{N}, \\ \max_{1 \leq k \leq n} k^{r/d+1-1/p} e_k^{\mathrm{ran}}(S^X, B_{C^r(Q,X)}) \geq e_1^{\mathrm{ran}}(S^\mathbb{K}, B_{C^r(Q,\mathbb{K})}) > 0, \end{split}$$

the statement is trivial for  $n < 8^d$ . Therefore we can assume  $n \ge 8^d$ . Clearly, we can also assume  $||(x_i)||_{\ell_{\infty}^n(X)} > 0$ . Let  $m \in \mathbb{N}$  be such that

(8) 
$$m^d \le n < (m+1)^d,$$

hence

$$(9) m \ge 8$$

Let  $\psi$  be an infinitely differentiable function on  $\mathbb{R}^d$  such that  $\psi(t) > 0$  for  $t \in (0,1)^d$  and  $\operatorname{supp} \psi \subset [0,1]^d$ . Let  $(Q_i)_{i=1}^{m^d}$  be the partition of Q into closed cubes of side length  $m^{-1}$  with disjoint interiors, let  $t_i$  be the point in  $Q_i$  with minimal coordinates and define  $\psi_i \in C(Q)$  by

$$\psi_i(t) = \psi(m(t-t_i)) \quad (i = 1, \dots, m^d).$$

It is easily verified that there is a constant  $c_0 > 0$  such that for all  $(\alpha_i)_{i=1}^{m^d}$ in  $[-1,1]^{m^d}$ ,  $m^d$ 

$$\sum_{i=1}^{m^{r}} \alpha_{i} x_{i} \psi_{i} \Big\|_{C^{r}(Q,X)} \le c_{0} m^{r} \|(x_{i})\|_{\ell_{\infty}^{n}(X)}.$$

Set

$$f_i = c_0^{-1} m^{-r} \|(x_i)\|_{\ell_{\infty}^n(X)}^{-1} x_i \psi_i;$$

it follows that

$$\sum_{i=1}^{m^{d}} \alpha_{i} f_{i} \in B_{C^{r}(Q,X)} \quad \text{for all } (\alpha_{i})_{i=1}^{m^{d}} \in [-1,1]^{m^{d}}.$$

Moreover, with  $\sigma = \int_{Q} \psi(t) dt$  we have

$$\begin{split} \left\| \sum_{i=1}^{m^d} \alpha_i S^X f_i \right\| &= c_0^{-1} m^{-r} \| (x_i) \|_{\ell_{\infty}^m(X)}^{-1} \left\| \sum_{i=1}^{m^d} \alpha_i x_i \int_Q \psi_i(t) \, dt \right\| \\ &= c_0^{-1} \sigma m^{-r-d} \| (x_i) \|_{\ell_{\infty}^m(X)}^{-1} \left\| \sum_{i=1}^{m^d} \alpha_i x_i \right\|. \end{split}$$

Next we use Lemmas 5 and 6 of [3] with K = X (although stated for  $K = \mathbb{R}$ , Lemma 6 is easily seen to hold for K = X as well) to obtain, for all  $l \in \mathbb{N}$ with  $l < m^d/4$ ,

$$e_l^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}) \ge \frac{1}{4} \min_{I \subseteq \{1, \dots, m^d\}, |I| \ge m^d - 4l} \mathbb{E} \left\| \sum_{i \in I} \varepsilon_i S^X f_i \right\|$$
$$\ge cm^{-r-d} \| (x_i) \|_{\ell_{\infty}^n(X)}^{-1} \mathbb{E} \left\| \sum_{i \in I} \varepsilon_i x_i \right\|.$$

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We put 
$$l = \lfloor m^d/8 \rfloor$$
. Then  
(10)  $m^d/16 < l \le m^d/8.$ 

Indeed, by (9) the left-hand inequality clearly holds for  $m^d < 16$ , while for  $m^d \ge 16$  we get  $\lfloor m^d/8 \rfloor > m^d/8 - 1 \ge m^d/16$ . We conclude that there is an  $I \subseteq \{1, \ldots, m^d\}$  with  $|I| \ge m^d - 4l \ge m^d/2$  and

$$\mathbb{E} \left\| \sum_{i \in I} \varepsilon_i x_i \right\| \le cm^{r+d} \|(x_i)\|_{\ell_{\infty}^n(X)} e_l^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}) \le cm^{r+d} l^{-r/d+1/p-1} \|(x_i)\|_{\ell_{\infty}^n(X)} \max_{1 \le k \le n} k^{r/d+1-1/p} e_k^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}) \le cn^{1/p} \|(x_i)\|_{\ell_{\infty}^n(X)} \max_{1 \le k \le n} k^{r/d+1-1/p} e_k^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}),$$

where we used (8) and (10). Finally, (8) and (9) give

$$|I| \ge m^d/2 \ge \frac{m^d}{2(m+1)^d} n \ge \frac{8^d}{2 \cdot 9^d} n.$$

Proof of Theorem 2. The left-hand inequality of (4) follows directly from Proposition 1, since the number of function values involved in  $A_{n,\omega}^{r,X}$  is bounded by  $ck^d + n \leq cn$ ; see also (16).

To prove the right-hand inequality of (4), let  $n \in \mathbb{N}$  and  $x_1, \ldots, x_n \in X$ . We construct by induction a partition of  $K = \{1, \ldots, n\}$  into a sequence of disjoint subsets  $(I_l)_{l=1}^{l^*}$  such that for  $1 \leq l \leq l^*$ ,

(11) 
$$|I_l| \ge \gamma \left| K \setminus \bigcup_{j < l} I_j \right|$$

and

(12) 
$$\mathbb{E} \left\| \sum_{i \in I_l} \varepsilon_i x_i \right\|$$
$$\leq c \left| K \setminus \bigcup_{j < l} I_j \right|^{1/p} \| (x_i) \|_{\ell_{\infty}^n(X)} \max_{1 \leq k \leq n} k^{r/d + 1 - 1/p} e_k^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}),$$

where c and  $\gamma$  are the constants from Lemma 2. For l = 1 the existence of an  $I_1$  satisfying (11)–(12) follows directly from Lemma 2. Now assume that we already have a sequence of disjoint subsets  $(I_l)_{l=1}^m$  of K satisfying (11)–(12). If

$$J := K \setminus \bigcup_{j \le m} I_j \neq \emptyset,$$

we apply Lemma 2 to  $(x_i)_{i \in J}$  to find  $I_{m+1} \subseteq J$  with

$$(13) |I_{m+1}| \ge \gamma |J|$$

and

(14) 
$$\mathbb{E} \left\| \sum_{i \in I_{m+1}} \varepsilon_i x_i \right\|$$
  
  $\leq c |J|^{1/p} \| (x_i)_{i \in J} \|_{\ell_{\infty}(J,X)} \max_{1 \leq k \leq |J|} k^{r/d+1-1/p} e_k^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}).$ 

Observe that for l = m + 1, (13) is just (11) and (14) implies (12). Furthermore, (11) implies

$$\left| K \setminus \bigcup_{j \le l} I_j \right| \le (1 - \gamma) \left| K \setminus \bigcup_{j \le l - 1} I_j \right|$$

and therefore

(15) 
$$\left| K \setminus \bigcup_{j \le l} I_j \right| \le (1 - \gamma)^l n$$

It follows that the process stops with  $K = \bigcup_{j \leq l} I_j$  for a certain  $l = l^* \in \mathbb{N}$ . This completes the construction.

Using the equivalence of moments (Theorem 4.7 of [7]), we find from (12) and (15) that

$$\left(\mathbb{E}\left\|\sum_{i=1}^{n}\varepsilon_{i}x_{i}\right\|^{p}\right)^{1/p} \leq c\mathbb{E}\left\|\sum_{i=1}^{n}\varepsilon_{i}x_{i}\right\| \leq c\sum_{l=1}^{l^{*}}\mathbb{E}\left\|\sum_{i\in I_{l}}\varepsilon_{i}x_{i}\right\|$$
$$\leq cn^{1/p}\|(x_{i})\|_{\ell_{\infty}^{n}(X)}\max_{1\leq k\leq n}k^{r/d+1-1/p}e_{k}^{\mathrm{ran}}(S^{X},B_{C^{r}(Q,X)})\sum_{l=1}^{l^{*}}(1-\gamma)^{(l-1)/p}.$$

This gives the upper bound of (4).  $\blacksquare$ 

Let us mention that results analogous to Theorem 2 and Corollary 2 above also hold for Banach space valued indefinite integration (see [2] for the definition) and for the solution of initial value problems for Banach space valued ordinary differential equations [5]. Indeed, an inspection of the respective proofs together with Lemma 1 of the present paper shows that Proposition 2 of [2] also holds with  $\tau_p(X)$  replaced by  $\sigma_{p,n}(X)$ , and similarly Proposition 3.4 of [5]. Moreover, in both papers the lower bounds on  $e_n^{\text{ran}}$ are obtained by reduction to (definite) integration and thus the right-hand side inequality of (4) carries over directly.

4. Appendix. In this appendix we recall some basic notions of information-based complexity—the framework we used above. We refer to [10, 12] for more on this subject and to [3, 4] for the particular notation applied here. First we introduce the class of deterministic adaptive

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algorithms of varying cardinality,  $\mathcal{A}^{\det}(C(Q, X), X)$ . It consists of tuples  $A = ((L_i)_{i=1}^{\infty}, (\varrho_i)_{i=0}^{\infty}, (\varphi_i)_{i=0}^{\infty})$ , with  $L_1 \in Q, \ \varrho_0 \in \{0, 1\}, \ \varphi_0 \in X$ ,

$$L_i: X^{i-1} \to Q \quad (i=2,3,\dots)$$

and

$$\varrho_i: X^i \to \{0, 1\}, \quad \varphi_i: X^i \to X \quad (i = 1, 2, \dots)$$

being arbitrary mappings. To each  $f \in C(Q, X)$ , we associate a sequence  $(t_i)_{i=1}^{\infty}$  with  $t_i \in Q$  as follows:

$$t_1 = L_1, \quad t_i = L_i(f(t_1), \dots, f(t_{i-1})) \quad (i \ge 2).$$

Define  $\operatorname{card}(A, f)$ , the cardinality of A at input f, to be 0 if  $\rho_0 = 1$ . If  $\rho_0 = 0$ , let  $\operatorname{card}(A, f)$  be the first integer  $n \ge 1$  with  $\rho_n(f(t_1), \ldots, f(t_n)) = 1$  if there is such an n, and  $\operatorname{card}(A, f) = \infty$  otherwise. For  $f \in C(Q, X)$  with  $\operatorname{card}(A, f) < \infty$  we define the output Af of algorithm A at input f as

$$Af = \begin{cases} \varphi_0 & \text{if } n = 0, \\ \varphi_n(f(t_1), \dots, f(t_n)) & \text{if } n \ge 1. \end{cases}$$

Let  $r \in \mathbb{N}_0$ . Given  $n \in \mathbb{N}_0$ , we let  $\mathcal{A}_n^{\det}(B_{C^r(Q,X)}, X)$  be the set of those  $A \in \mathcal{A}^{\det}(C(Q,X), X)$  for which

$$\max_{f \in B_{C^r(Q,X)}} \operatorname{card}(A, f) \le n.$$

The error of  $A \in \mathcal{A}_n^{\det}(B_{C^r(Q,X)}, X)$  as an approximation of  $S^X$  is defined as

$$e(S^X, A, B_{C^r(Q,X)}) = \sup_{f \in B_{C^r(Q,X)}} \|S^X f - Af\|.$$

The deterministic nth minimal error of  $S^X$  is defined for  $n \in \mathbb{N}_0$  as

$$e_n^{\det}(S^X, B_{C^r(Q,X)}) = \inf_{A \in \mathcal{A}_n^{\det}(B_{C^r(Q,X)})} e(S^X, A, B_{C^r(Q,X)}).$$

It follows that no deterministic algorithm that uses at most n function values can have an error smaller than  $e_n^{\text{det}}(S^X, B_{C^r(Q,X)})$ .

Next we introduce the class of randomized adaptive algorithms of varying cardinality,  $\mathcal{A}_n^{\operatorname{ran}}(B_{C^r(Q,X)}, X)$ , consisting of tuples  $A = ((\Omega, \Sigma, \mathbb{P}), (A_\omega)_{\omega \in \Omega})$ , where  $(\Omega, \Sigma, \mathbb{P})$  is a probability space,  $A_\omega \in \mathcal{A}^{\operatorname{det}}(C(Q,X), X)$  for all  $\omega \in \Omega$ , and for each  $f \in B_{C^r(Q,X)}$  the mapping  $\Omega \ni \omega \mapsto \operatorname{card}(A_\omega, f)$  is  $\Sigma$ -measurable and satisfies  $\mathbb{E} \operatorname{card}(A_\omega, f) \leq n$ . Moreover, the mapping  $\Omega \ni \omega \mapsto A_\omega f \in X$  is  $\Sigma$ -to-Borel measurable and essentially separably valued, i.e., there is a separable subspace  $X_0 \subseteq X$  such that  $A_\omega f \in X_0$  for  $\mathbb{P}$ -almost all  $\omega \in \Omega$ . The error of  $A \in \mathcal{A}_n^{\operatorname{ran}}(C(Q,X), X)$  in approximating  $S^X$  on

 $B_{C^r(Q,X)}$  is defined as

$$e(S^X, A, B_{C^r(Q,X)}) = \sup_{f \in B_{C^r(Q,X)}} \mathbb{E} \| S^X f - A_\omega f \|,$$

and the randomized nth minimal error of  $S^X$  as

$$e_n^{\mathrm{ran}}(S^X, B_{C^r(Q,X)}) = \inf_{A \in \mathcal{A}_n^{\mathrm{ran}}(B_{C^r(Q,X)})} e(S^X, A, B_{C^r(Q,X)}).$$

Consequently, no randomized algorithm that uses (on the average) at most n function values has an error smaller than  $e_n^{\operatorname{ran}}(S^X, B_{C^r(Q,X)}, X)$ .

Define for  $\varepsilon > 0$  the *information complexity* as

$$n_{\varepsilon}^{\mathrm{ran}}(S, B_{C^{r}(Q, X)}) = \min\{n \in \mathbb{N}_{0} : e_{n}^{\mathrm{ran}}(S, B_{C^{r}(Q, X)}) \leq \varepsilon\}$$

if there is such an n, and  $n_{\varepsilon}^{\operatorname{ran}}(S, B_{C^r(Q,X)}) = \infty$  if there is no such n. Thus, if  $n_{\varepsilon}^{\operatorname{ran}}(S, B_{C^r(Q,X)}) < \infty$ , it follows that any algorithm with error  $\leq \varepsilon$  needs at least  $n_{\varepsilon}^{\operatorname{ran}}(S, B_{C^r(Q,X)})$  function values, while  $n_{\varepsilon}^{\operatorname{ran}}(S, B_{C^r(Q,X)}) = \infty$ means that no algorithm at all has error  $\leq \varepsilon$ . The information complexity is essentially the inverse function of the *n*th minimal error. So determining the latter means determining the information complexity of the problem.

Let us also mention the subclasses consisting of quadrature formulas. Let  $n \geq 1$ . A mapping  $A : C(Q, X) \to X$  is called a *deterministic quadrature* formula with n nodes if there are  $t_i \in Q$  and  $a_i \in \mathbb{K}$   $(1 \leq i \leq n)$  such that

$$Af = \sum_{i=1}^{n} a_i f(t_i) \quad (f \in C(Q, X)).$$

In terms of the definition of  $\mathcal{A}^{\det}(C(Q,X),X)$  this means that the functions  $L_i$  and  $\varrho_i$  are constant,  $\varrho_0 = \varrho_1 = \cdots = \varrho_{n-1} = 0$ ,  $\varrho_n = 1$ , and  $\varphi_n$  has the form  $\varphi_n(x_1,\ldots,x_n) = \sum_{i=1}^n a_i x_i$ . Clearly,  $A \in \mathcal{A}_n^{\det}(B_{C^r(Q,X)},X)$ .

A tuple  $A = ((\Omega, \Sigma, \mathbb{P}), (A_{\omega})_{\omega \in \Omega})$  is called a randomized quadrature with n nodes if there exist random variables  $t_i : \Omega \to Q$  and  $a_i : \Omega \to \mathbb{K}$   $(1 \le i \le n)$  with

$$A_{\omega}f = \sum_{i=1}^{n} a_i(\omega)f(t_i(\omega)) \quad (f \in C(Q, X), \, \omega \in \Omega).$$

For each such A we have  $A \in \mathcal{A}_n^{\operatorname{ran}}(B_{C^r(Q,X)},X)$ . Finally we note that the algorithms  $A_{n,\omega}^{r,X}$  defined in (5) and (6) are quadratures. Indeed, for  $A_{n,\omega}^{0,X}$  given by (5) this is obvious. For  $r \geq 1$  we represent  $P_k^{r,X} \in \mathscr{L}(C(Q,X))$  as

$$P_k^{r,X}f = \sum_{j=1}^M f(u_j)\psi_j(t)$$

with  $M \leq ck^d$ ,  $u_j \in Q$ ,  $\psi_j \in C(Q)$   $(1 \leq i \leq M)$ , and obtain, setting  $b_j = \int_Q \psi_j(t) dt$ , (16)  $A_{n,\omega}^{r,X} f = S^X(P_k^{r,X} f) + A_{n,\omega}^{0,X}(f - P_k^{r,X} f)$   $= \sum_{j=1}^M b_j f(u_j) + \frac{1}{n} \sum_{i=1}^n (f(\xi_i(\omega)) - (P_k^{r,X} f)(\xi_i(\omega)))$  $= \sum_{i=1}^M b_j f(u_j) + \frac{1}{n} \sum_{i=1}^n f(\xi_i(\omega)) - \sum_{i=1}^M \left(\frac{1}{n} \sum_{i=1}^n \psi_j(\xi_i(\omega))\right) f(u_j).$ 

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