

# Approximation vs. small deviation of $S_{\alpha}S$ Lévy process

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13.09.2005 / Small Deviation probabilities and Related  
Topics

# Outline

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# Classical Widths

Let

- ▶  $E$  be a Banach space,
- ▶  $B \subseteq E$  precompact,
- ▶  $Q_N : E \rightarrow E/N$  quotient mapping  
( $N \subseteq E$  closed linear subspace)

define *linear widths*

$$a_n(B, E) := \inf \left\{ \sup_{x \in B} \|x - u(x)\| : u : E \rightarrow E \text{ linear, } \text{rk}(u) \leq n \right\}$$

and *Kolmogorov widths*

$$d_n(B, E) := \inf \left\{ \sup_{x \in B} \|Q_N(x)\|_{E/N} : N \subseteq E, \dim N \leq n \right\} .$$

Let now

- ▶  $E$  be a separable Banach space,
- ▶  $X$  r.v. on  $E$  with  $\mathbb{E}\|X\|^q < \infty$  ( $q \in (0, \infty)$ ),
- ▶  $Q_N : E \rightarrow E/N$  quotient mapping

define *linear widths*

$$a_n(X, E, q) := \inf \left\{ \left( \mathbb{E} \|X - u(X)\|^q \right)^{1/q} : \text{rk}(u) \leq n \right\}$$

and *Kolmogorov widths*

$$d_n(X, E, q) := \inf \left\{ \left( \|Q_N(X)\|_{E/N}^q \right)^{1/q} : N \subseteq E, \dim N \leq n \right\}.$$

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# Quantization numbers

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For  $C \subseteq E$ ,  $x \in E$  set

$$d(x, C) := \inf_{c \in C} \|x - c\| .$$

The  $n$ -th quantization number is

$$e_n(X, E, q) := \inf \left\{ \mathbb{E} \left( d(X, E)^q \right)^{1/q} : C \subseteq E, \#C \leq 2^n \right\} .$$

Also denote

$$\varphi(X, E, \varepsilon) := -\log \mathbb{P}(\|X\| < \varepsilon) .$$

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# Example: Fractional Brownian Motion

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Let  $B^\gamma$  denote the fractional Brownian Motion on  $[0, 1]$ .

## Theorem (Maïorov, Wasilkowski)

If  $p < \infty$ , then

$$a_n(B^\gamma, L_p, q) \asymp d_n(B^\gamma, L_p, q) \asymp n^{-\frac{\gamma}{2}}.$$

Furthermore,

$$d_n(B^\gamma, C, q) \asymp n^{-\frac{\gamma}{2}}, \quad a_n(B^\gamma, C, q) \asymp n^{-\frac{\gamma}{2}} (\log n)^{\frac{1}{2}}.$$

- ▶ Polynomial decay,  $p, q$  irrelevant for rate.
- ▶ Kolmogorov approximation does not help much.

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Denote now  $B_d^\gamma$  the  $d$ -dimensional fractional Brownian sheet defined on  $[0, 1]^d$ .

## Theorem (Kühn, Linde)

For  $p < \infty$ ,

$$a_n(B_d^\gamma, L_p, q) \asymp d_n(B_d^\gamma, L_p, q) \asymp n^{-\frac{\gamma}{2}} (\log n)^{\frac{\gamma}{2}(d-1)}$$

while

$$a_n(B_d^\gamma, C, q) \asymp n^{-\frac{\gamma}{2}} (\log n)^{\frac{\gamma}{2}(d-1) + \frac{1}{2}} .$$

## Theorem (Talagrand)

$$d_n(B_2^\gamma, C, q) \asymp a_n(B_2^\gamma, C, q) .$$

Many further results by Wasilkowski, Ritter, Buslaev/Seleznev, Majorov, Gensun, ...

# Linear vs. Kolmogorov

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## Theorem (Pisier)

For  $p \in (1, \infty)$  there is  $C_p$  such that for Gaussian  $X$ ,

$$a_n(X, L_p, q) \leq C_p \cdot d_n(X, L_p, q) .$$

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## Theorem

There is  $K > 0$  such that for any  $E$ , Gaussian  $X$

$$a_n(X, E, q) \leq K \cdot d_n(X, E, q) \cdot (1 + \log n) .$$

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**Conclusion:** Using Kolmogorov-type approximations **never** improves the rate for Gaussian processes.

**Question:** Is  $(1 + \log n)$  improvable to  $\sqrt{1 + \log n}$ ?



# Quantization and Small Deviation

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## Theorem (Dereich et al)

For any centered Gaussian  $X$  and  $q > 0$ , we have

$$e_n(X, E, q) \asymp n^{-q} \quad \text{iff} \quad \varphi(\varepsilon, X, E) \asymp \varepsilon^{-\frac{1}{q}}.$$

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Key Lemma for lower bounds: Define the pseudo-inverse

$$b_n(X, E) = \varphi^{-1}(X, E)(n^{-1}).$$

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## Lemma (Dereich et al)

If  $X$  is a r.v. with the Anderson property

$$\mathbb{P}(\|X - y\| < \varepsilon) \leq \mathbb{P}(\|X\| < \varepsilon), \quad y \in E, \varepsilon > 0$$

then

$$b_{n+2}(X, E) \leq 2^{\frac{1}{q}} \cdot e_n(X, E, q).$$

# Relations to Widths

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## Theorem (Kuelbs, Li, Linde, C)

*For  $X$  Gaussian,*

$$d_n(X, E, q) \asymp n^{-\varrho} \quad \text{iff} \quad \varphi(\varepsilon, X, E) \asymp \varepsilon^{-1/\varrho} .$$

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## Corollary

*For  $X$  Gaussian,*

$$d_n(X, E, q) \asymp n^{-\varrho} \quad \text{iff} \quad e_n(X, E, q) \asymp n^{-\varrho} .$$

- ▶  $a_n, d_n, e_n, b_n$  all 'walk hand in hand'.

## Definition

A r.v.  $\xi$  is  $S_{\alpha}S$  iff

$$\hat{\xi}(\lambda) := \mathbb{E}e^{i\lambda\xi} = \exp\{|\lambda|^{\alpha}\sigma^{\alpha}\}$$

for some  $\sigma \geq 0$ .

An  $E$ -valued r.v.  $X$  is  $S_{\alpha}S$  iff for any  $a \in E'$ ,  $a(X)$  is  $S_{\alpha}S$ .

**Note:**  $S_2S$  = centered Gaussian.

## Features

- ▶ **Stability:** Sum of two independent  $S_{\alpha}S$  processes is  $S_{\alpha}S$  again.
- ▶ Different smoothness properties than Gaussian counterparts.
- ▶ Different integrability:  $\mathbb{E}\|X\|_F^q < \infty$  only iff  $q < \alpha$  for  $\alpha < 2 \Rightarrow$  No second-order theory.
- ▶ Conditional laws more delicate to handle.
- ▶ Representation as mixture of Gaussian processes.
- ▶ Connection with type/cotype theory of Banach spaces.

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## Theorem

For  $X \in \mathcal{S}$  in  $E$ ,

$$d_n(X, E, q) \preceq n^{-\varrho} \quad \Rightarrow \quad e_n(X, E, q) \preceq n^{-\varrho} .$$

- ▶ Valid in slightly weaker form for arbitrary process  $X$ .
- ▶ Semi-constructive: Proof shows how to construct 'good' quantizers using only information about 'good subspaces' for  $d_n$ .
- ▶ Optimal, as will be shown below.

# Ordering of the quantities

$X$   $S\alpha S$  variable mixture of centered Gaussian  
 $\Rightarrow$  has Anderson property

$$\mathbb{P}(\|X - y\| < \varepsilon) \leq \mathbb{P}(\|X\| < \varepsilon) \quad \varepsilon > 0, y \in E,$$

Hence, the Lemma of Dereich et al is applicable.

## Corollary

For  $X$   $S\alpha S$  and  $q < \alpha$  we have

$$b_{n+2}(X, F) \preceq e_n(X, F, q) \text{ "}\preceq\text{" } d_n(X, F, q) \leq a_n(X, F, q).$$

Consider a spectral measure  $\nu_X$  of  $X$ , i.e., a Borel measure on  $F$  satisfying

$$\mathbb{E}e^{i\langle \lambda, X \rangle} = \exp\left\{-\frac{1}{2} \int_F |\langle \lambda, \mathbf{x} \rangle|^\alpha d\nu_X(\mathbf{x})\right\}, \quad \lambda \in F^*.$$

Let  $S_X$  be a r.v. distributed after  $\nu_X/\nu_X(F)$ . Then  $\|S_X\|_F$  is  $\alpha$ -integrable, and considerations of type/cotype lead to

## Theorem

*If  $p > \alpha$ , then*

$$d_n(X, L_p, q) \asymp n^{-\varrho} \quad \text{iff} \quad d_n(S_X, L_p, \alpha) \asymp n^{-\varrho}.$$

*Same result for  $a_n$  as well.*

# $S_{\alpha}S$ Lévy motion

The  $S_{\alpha}S$  Lévy motion is a  $S_{\alpha}S$  process  $X = (X_t)_{t \in [0,1]}$  with

- ▶ independent increments,  $X_0 = 0$ ,
- ▶  $(X_{ct}) \stackrel{d}{=} c^{1/\alpha}(X_t)$  for  $c > 0$ ,
- ▶  $X$  has càdlàg trajectories.

## Theorem (Folklore)

$$b_n(X, L_p) \asymp n^{-1/\alpha} .$$

## Theorem (Dereich)

$$e_n(X, L_p, q) \asymp n^{-1/\alpha} .$$



## Theorem

$$a_n(X, L_p, q) \asymp d_n(X, L_p, q) \asymp \begin{cases} n^{-1/\alpha}, & p < \alpha, \\ n^{-1/p}, & \alpha < p \leq 2, \\ n^{-1/2}, & \alpha > 2. \end{cases}$$

- ▶ Same asymptotics for  $b_n$ ,  $e_n$  and for  $a_n$ ,  $d_n$ .
- ▶ *Different* rates for approximation and quantization/small deviations when  $p > \alpha$ .
- ▶ Optimal rates for approximation in the case  $p \leq 2$  achievable by naive sampling, for  $p > \alpha$  needs results of Gluskin about random projections.

## Questions

- ▶ Conditions when

$$a_n(X, E, q) \preceq d_n(X, E, q)$$

does hold?

- ▶ Examples for  $a_n \not\asymp d_n$ ?
- ▶ Is  $b_n \asymp \varphi_n$  a general principle for Lévy processes?
- ▶ Generalizations/Applications for SDE after  $S_\alpha S$  motion?