

Spline approximation of random processes with singularity

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Suppose a random process $X(t)$, $t \in [0, 1]$, with finite second moment is observed in a finite number of points (**sampling designs**). At any unsampled point t , we approximate the value of the process. The approximation performance on the entire interval is measured by mean errors. In this talk we deal with two problems:

- ▶ Investigating accuracy of such interpolator in mean norms
- ▶ Constructing a sequence of sampling designs with asymptotically optimal properties

Basic notation

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We set $\|\xi\| = (E\xi^2)^{1/2}$ for all $\xi \in L^2(\Omega)$ and consider the approximation based on the normed linear space $\mathcal{C}^m[0, 1]$ of random processes having continuous q.m. (quadratic mean) derivatives up to order $m \geq 0$.

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We define the **integrated mean norm** for any $X \in \mathcal{C}^m[0, 1]$ by setting

$$\|X\|_p = \left(\int_0^1 \|X(t)\|^p dt \right)^{1/p}, \quad 1 \leq p < \infty,$$

and the **uniform mean norm** $\|X\|_\infty = \max_{[0,1]} \|X(t)\|$.

Hölder's conditions and local stationarity

We define the classes of processes used throughout the paper. Let $X \in \mathcal{C}^m[a, b]$. We say that

- i) $X \in \mathcal{C}^{m,\beta}([a, b], C)$ if $Y(t) = X^{(m)}(t)$ is *Hölder continuous*, i.e., if there exist $0 < \beta \leq 1$ and a positive constant C such that, for all $t, t + s \in [a, b]$,

$$\|Y(t + s) - Y(t)\| \leq C|s|^\beta, \quad (1)$$

- ii) $X \in \mathcal{V}^{m,\beta}([a, b], V(\cdot))$ if $Y(t) = X^{(m)}(t)$ is *locally Hölder*, i.e., if there exist $0 < \beta \leq 1$ and a positive continuous function $V(\cdot)$ such that, for all $t, t + s, \in [a, b], s > 0$,

$$\|Y(t + s) - Y(t)\| \leq V(t)^{1/2} |s|^\beta, \quad (2)$$

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- iii) $X \in \mathcal{B}^{m,\beta}([a, b], c(\cdot))$ if $Y(t) = X^{(m)}(t)$ is *locally stationary* (see, Berman (1974)), i.e., if there exist $0 < \beta \leq 1$ and a positive continuous function $c(t)$ such that

$$\lim_{s \rightarrow 0} \frac{\|Y(t + s) - Y(t)\|}{|s|^\beta} = c(t)^{1/2} \text{ uniformly in } t \in [a, b]. \quad (3)$$

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We say that $X \in \mathcal{BV}^{m,\beta}((0, 1], c(\cdot), V(\cdot))$ if $X \in \mathcal{C}^m[a, b]$ and its m -th q.m. derivative satisfies (2) and (3) for any $[a, b] \subset (0, 1]$.

Composite Splines

For any $f \in C^l[0, 1]$, $l \geq 0$, the piecewise Hermite polynomial $H_k(t) := H_k(f, T_n)(t)$, of degree $k = 2l + 1$, $l \geq 0$, is the unique solution of the interpolation problem

$H_k^{(j)}(t_i) = f^{(j)}(t_i)$, where $i = 0, \dots, n$, $j = 0, \dots, l$. Define $H_{q,k}(X, T_n)$, $q \leq k$, to be a *composite Hermite spline*

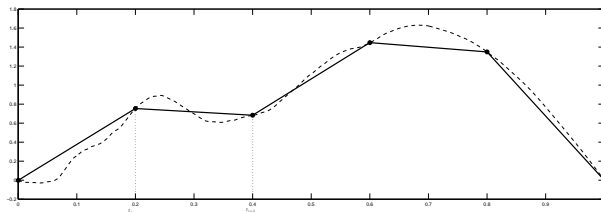
$$H_{q,k}(X, T_n) := \begin{cases} H_q(X, T_n)(t), & t \in [0, t_1] \\ H_k(X, T_n)(t), & t \in [t_1, 1] \end{cases}.$$

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quasi Regular Sequences

We consider **quasi regular sequences** (qRS) of sampling designs $\{T_n = T_n(h)\}$ generated by a density function $h(\cdot)$ via

$$\int_0^{t_i} h(t) dt = \frac{i}{n}, \quad i = 1, \dots, n,$$

where $h(\cdot)$ is continuous for $t \in (0, 1]$ and if $h(\cdot)$ is unbounded in $t = 0$, then $h(t) \rightarrow +\infty$ as $t \rightarrow +0$. We denote this property of $\{T_n\}$ by: $\{T_n\}$ is qRS(h). Observe that if $h(\cdot)$ is positive and continuous on $[0, 1]$, then we obtain **regular sequences**.

Denote the distribution function by $H(t) = \int_0^t h(v) dv$, and the quantile function by $G(s) = H^{-1}(s)$, and $g(s) = G'(s)$, $t, s \in [0, 1]$, i.e., $t_j = G(j/n)$, $j = 0, \dots, n$.

Previous Results

- ▶ (Seleznev, Buslaev 1999)
Optimal approximation rate for linear methods for $X \in \mathcal{C}^{0,\beta}[0, 1]$ is $n^{-\beta}$
- ▶ (Seleznev, 2000)
Results on Hermite spline approximation when $X \in \mathcal{B}^{m,\beta}([0, 1], c(\cdot))$ and regular sequences of sampling designs are used

$$\|X - H_k(X, T_n)\| \sim n^{-(m+\beta)} \text{ as } n \rightarrow \infty, \quad m \leq k.$$

Processes of interest

Let $X(t)$, $t \in [0, 1]$, be a stochastic process which l – th q.m. derivative satisfies Hölder's condition on $[0, 1]$ with $0 < \alpha \leq 1$. Moreover the process is q.m. differentiable up to order m on the left-open interval $(0, 1]$. The m – th derivative is locally Hölder with $0 < \beta \leq 1$ on $[0, 1]$ and locally stationary on any $[a, b] \subset (0, 1]$ with β . We denote it by

$$X \in \mathcal{BV}^{m,\beta}((0, 1], c(\cdot), V(\cdot)) \cap \mathcal{C}^{l,\alpha}([0, 1], M).$$

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Examples:

- ▶ $X_1(t) = Y(t^{1/2})$, $t \in [0, 1]$, where $Y(t)$, $t \in [0, 1]$, is a fractional Brownian motion with Hurst parameter H ,
- ▶ $X_2(t) = t^{9/10}Y(t)$, $t \in [0, 1]$, where $Y(t)$, $t \in [0, 1]$, is a zero mean stationary process with $\text{Cov}(Y(t), Y(s)) = \exp(-(t-s)^2)$.

Problem formulation

We have a process which l – th derivative is α -Hölder on $[0, 1]$.
Can we get the approximation rate **better** than $n^{-(l+\alpha)}$?

Regularly varying function

A positive function $f(\cdot)$ is called **regularly varying** (on the right) at the origin with index ρ , if for any $\lambda > 0$,

$$\frac{f(\lambda x)}{f(x)} \rightarrow \lambda^\rho \quad \text{as } x \rightarrow 0+,$$

and denote this property by $f \in \mathcal{R}_\rho(O+)$.

Assumptions and conditions

Let us recall the notation: $H(t) = \int_0^t h(v)dv$, $G(s) = H^{-1}(s)$, and $g(s) = G'(s)$, $t, s \in [0, 1]$.

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We formulate the following assumptions about the local Hölder function and the sequence generating density:

- (A1) there exists a function $r \in \mathcal{R}_\rho(0+)$, $\rho > 0$, such that $r(s) \geq g(s)$, $s \in [0, a]$, for some $a > 0$,
- (A2) there exists a function $R \in \mathcal{R}_\rho(0+)$, $\rho > 0$, such that $R(H(t)) \geq V(t)^{1/2}$, $t \in [0, a]$, for some $a > 0$.

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Moreover we formulate the following conditions on the behavior of the introduced functions in a neighborhood of zero

- (C1) the assumption (A1) holds with $r(\cdot)$ such that $V(t)r(H(t))^{2(m+\beta)} \rightarrow 0$ as $t \rightarrow 0$,
- (C2) for a given $1 \leq p < \infty$, the assumptions (A1) and (A2) hold with $r(\cdot)$ and $R(\cdot)$ such that, for some $b > 0$, $R(H(t))r(H(t))^{m+\beta} \in L_p[0, b]$.

Optimal rate recovery

In this section we use the convention, that if $p = \infty$, then $1/p = 0$.

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Theorem

Let $X \in \mathcal{BV}^{m,\beta}((0, 1], c(\cdot), V(\cdot)) \cap \mathcal{C}^{l,\alpha}([0, 1], M)$, $l + \alpha \leq m + \beta$, be interpolated by a composite Hermite spline $H_{q,k}(t, T_n)$, $l \leq q \leq 2l + 1$, $m \leq k \leq 2m + 1$, where T_n is a $qRS(h)$. Let the condition (C1) be satisfied if $p = \infty$, or alternatively, for $1 \leq p < \infty$, let the condition (C2) hold. If additionally,

$$r(s) = o(s^{(m+\beta)/(\alpha+l+1/p)-1}) \text{ as } s \rightarrow 0, \quad (4)$$

then

$$\lim_{n \rightarrow \infty} n^{m+\beta} \|X - H_{q,k}(t, T_n)\|_p = b_{k,p}^{m,\beta} \|c^{1/2} h^{-(m+\beta)}\|_p.$$

For a composite spline $H_{q,k}$, we define **asymptotic optimality** of the sequence of sampling designs T_n^* by

$$\lim_{n \rightarrow \infty} \|X - H_{q,k}(X, T_n^*)\|_p / \inf_{T \in D_n} \|X - H_{q,k}(X, T)\|_p = 1,$$

where $D_n := \{T_n : 0 = t_0 < t_1 < \dots < t_n = 1\}$ denotes the set of all $(n+1)$ -point designs.

Asymptotically optimal density

Let $\gamma := 1/(m + \beta + 1/p)$, $1 \leq p < \infty$. Define

$$h^*(t) := c(t)^{\gamma/2} / \int_0^1 c(s)^{\gamma/2} ds, \quad t \in [0, 1],$$

and denote by $g^*(s)$, $s \in [0, 1]$ the corresponding quantile density function.

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Proposition

Let $X \in \mathcal{BV}^{m,\beta}((0, 1], c(\cdot), V(\cdot)) \cap \mathcal{C}^{l,\alpha}([0, 1], M)$, $l + \alpha \leq m + \beta$, such that $V(t) = O(c(t))$ as $t \rightarrow 0$, and $c, V \in \mathcal{R}_\rho(0+)$, be interpolated by a composite Hermite spline $H_{q,k}(t, T_n)$, $l \leq q \leq 2l + 1$, $m \leq k \leq 2m + 1$, where T_n is a qRS(h). If the condition

$$g^*(s) = o(s^{(m+\beta)/(\alpha+l+1/p)-1}) \text{ as } s \rightarrow 0, \quad (5)$$

is satisfied, then $h^*(\cdot)$ is asymptotically optimal, and

$$\lim_{n \rightarrow \infty} n^{m+\beta} \|X - H_{q,k}(X, T_n(h^*))\|_p = b_{k,p}^{m,\beta} \|c^{1/2}\|_\gamma.$$

Remark

Theorem and Proposition above extend to the class

$$\mathcal{B}^{m,1}((0,1], c(\cdot)) \cap \mathcal{C}^{l,\alpha}([0,1], M), \quad l + \alpha \leq m + 1,$$

with the assumptions and the conditions formulated in terms of local stationarity function $c(\cdot)$ instead of function $V(\cdot)$.

Undersmoothing

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For any $q \leq k < m$ denote $c_k(t) := \|X^{(k+1)}(t)\|^2$, and consider the following modification of the assumption (A2),

(A2') there exists a function $R \in \mathcal{R}_\rho(0+)$, $\rho > 0$, such that $R(H(t)) \geq c_k(t)^{1/2}$, $t \in [0, a]$, for some $a > 0$.

Let us reformulate the conditions in terms of function $c_k(\cdot)$:

(C1') the assumption (A1) holds with $r(\cdot)$, such that $c_k(t)r(H(t))^{2(k+1)} \rightarrow 0$ as $t \rightarrow 0$,

(C2') for given $1 \leq p < \infty$, the assumptions (A1) and (A2'') hold with $r(\cdot)$ and $R(\cdot)$, such that, for some $b > 0$, $R(H(t))r(H(t))^{k+1} \in L_p[0, b]$.

Observe that in this case we do not require the local Hölder property, and the results are formulated for the local stationary class of processes.

Theorem

Let $X \in \mathcal{B}^{m,\beta}((0, 1], c(\cdot)) \cap \mathcal{C}^{l,\alpha}([0, 1], M)$, $l + \alpha \leq m + \beta$, be interpolated by a composite Hermite spline $H_{q,k}(t, T_n)$, $l \leq q \leq 2l + 1$, $q \leq k < m$, where T_n is a $qRS(h)$. Let the condition $(C1')$ be satisfied if $p = \infty$, or alternatively, for $1 \leq p < \infty$, let the condition $(C2')$ hold. If additionally,

$$r(s) = o(s^{(k+1)/(\alpha+l+1/p)-1}) \text{ as } s \rightarrow 0,$$

then

$$\lim_{n \rightarrow \infty} n^{k+1} \|X - H_{q,k}(t, T_n)\|_p = b_{k,p}^{k,1} \|c_k^{1/2} h^{-(k+1)}\|_p.$$

Asymptotically optimal density

We introduce the asymptotically optimal density, when $1 \leq p < \infty$. Let $\gamma_k := 1/(k+1+1/p)$. Define

$$h_k^*(t) := c_k(t)^{\gamma_k/2} / \int_0^1 c_k(s)^{\gamma_k/2} ds,$$

and denote by $g_k^*(s)$, $s \in [0, 1]$ the corresponding quantile density function.

Proposition

Let $X \in \mathcal{B}^{m,\beta}((0, 1], c(\cdot)) \cap \mathcal{C}^{l,\alpha}([0, 1], M)$, $l + \alpha \leq m + \beta$, such that $c_k \in \mathcal{R}_p(0+)$, be interpolated by a composite Hermite spline $H_{q,k}(t, T_n)$, $l \leq q \leq 2l + 1$, $q \leq k < m$, where T_n is a $qRS(h)$. If the condition

$$g_k^*(s) = o(s^{(k+1)/(\alpha+l+1/p)-1}) \text{ as } s \rightarrow 0,$$

is satisfied, then $h^(\cdot)$ is asymptotically optimal, and*

$$\lim_{n \rightarrow \infty} n^{k+1} \|X - H_{q,k}(X, T_n(h_k^*))\|_p = b_{k,p}^{k,1} \|c_k^{1/2}\|_{\gamma_k}.$$

Numerical Experiments and Examples

When choosing the knot distribution, we consider the densities of a form:

$$h_{\lambda}(t) = \frac{1}{\lambda} t^{\frac{1}{\lambda}-1}.$$

This leads to

$$t_i = \left(\frac{i}{n}\right)^{\lambda},$$

and therefore, such densities are called *power densities*.

Example 1

Let $B_H = B_H(t)$, $t \in [0, 1]$ denote a fractional Brownian motion with Hurst parameter $0 \leq H \leq 1$, i.e., a zero mean Gaussian process, starting at zero, with covariance function:

$$K_{B_H}(s, t) = \frac{1}{2} \left(|t|^{2H} + |s|^{2H} - |t - s|^{2H} \right).$$

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Consider now a time change version of the process B_H ,

$$X(t) := B_H(\sqrt{t}), \quad t \in [0, 1].$$

Then

$$X \in \mathcal{BV}^{0,H}((0, 1], c(\cdot), V(\cdot)) \cap \mathcal{C}^{0,H/2}([0, 1], 1),$$

where $c(t) = V(t) = (4t)^{-H}$.

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where $c(t) = V(t) = (4t)^{-H}$.

Let $H = 0.8$. We measure the error of the **piecewise linear approximation** in mean uniform norm ($p = \infty$). The conditions of Theorem are satisfied if $\lambda > 2H/(H + 2/p)$. In our experiments we consider the following choice of λ :

$$\lambda_{1,\infty} = 1 \quad \lambda_{2,\infty} = 2.1$$

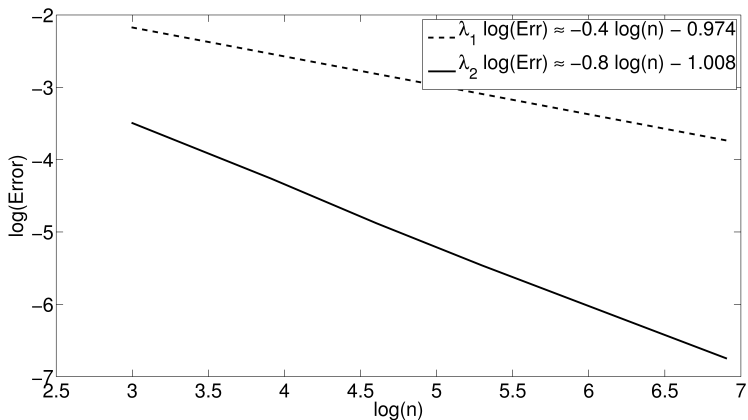


Figure: Comparison of the uniform mean errors for the uniform density $h_{\lambda_1, \infty}(\cdot)$ and $h_{\lambda_2, \infty}(\cdot)$ in the log-log scale.

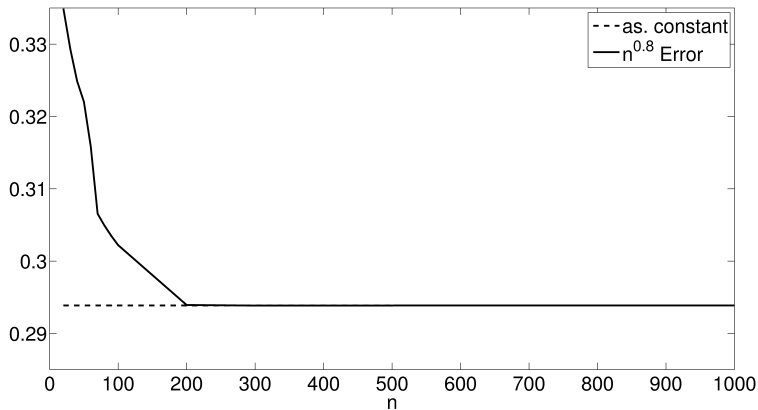


Figure: Convergence of the n^2 scaled uniform mean errors to the asymptotic constant for the generating density $h_{\lambda_2, \infty}(\cdot)$.

Example 2

Let $Y(t)$, $t \in [0, 1]$ be a zero mean Gaussian process with covariance kernel

$$K_Y(s, t) = \exp\{-(s - t)^2\}.$$

Such process has infinitely many q.m. derivatives, hence the rate of convergence of Hermite spline approximation is limited by the order of spline only.

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$$X(t) := t^{0.9} Y(t), \quad t \in [0, 1].$$

Then

$$X \in \mathcal{B}^{m,1}((0, 1], c_m(\cdot)) \cap \mathcal{C}^{0,0.9}([0, 1], 1), \quad \text{for any } m \geq 0,$$

where $c_m = \|Y^{m+1}(t)\|^2$.

Consider now an approximation by the composite Hermite spline $H_{1,3}$. The conditions of Theorem are satisfied when $\lambda > 4/(0.9 + 1/p)$, with

$$c_3(t) = 1680t^{\frac{9}{5}} - \frac{44631}{1250}t^{-\frac{11}{5}} - \frac{16929}{125000}t^{-\frac{21}{5}} + \frac{8424}{5}t^{-\frac{1}{5}} + \frac{4322241}{10^8}t^{-\frac{31}{5}}$$

To evaluate mean integrated error ($p = 2$) we consider the following choice of the generating density parameter :

$$\lambda_{1,2} = 1 \quad \lambda_{2,2} = 3 \quad \lambda_{3,2} = 4 \quad \lambda_{4,2} = 5$$

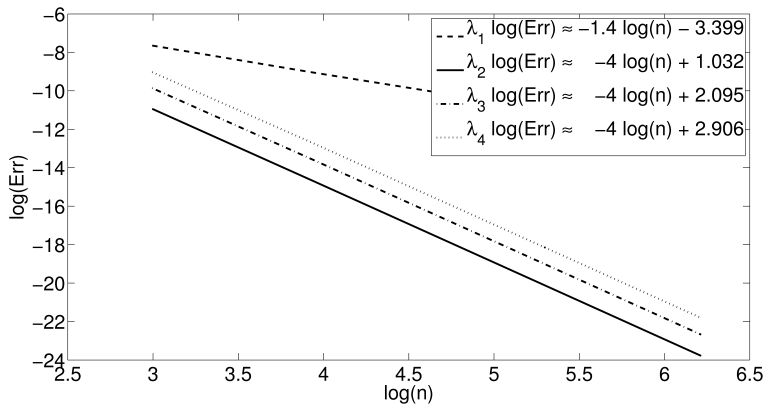


Figure: Comparison of the integrated mean errors, ($p = 2$), for the uniform density $h_{\lambda_{1,2}}(\cdot)$, $h_{\lambda_{2,2}}(\cdot)$, $h_{\lambda_{3,2}}(\cdot)$, and $h_{\lambda_{4,2}}(\cdot)$ in the log-log scale.

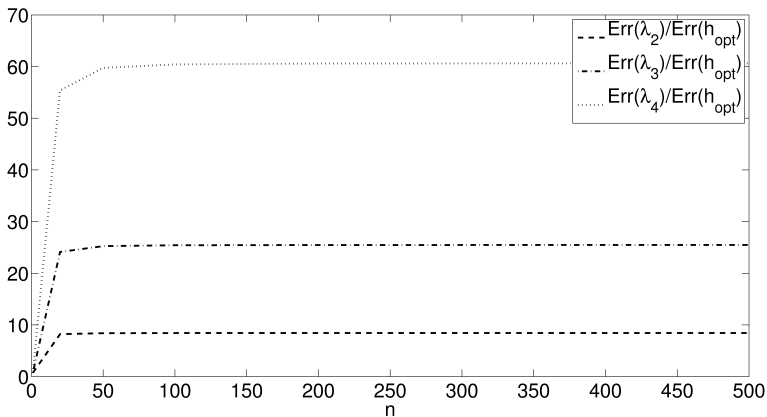


Figure: Ratio of the integrated mean errors, ($p = 2$), for the densities $h_{\lambda_{2,2}}(\cdot)$, $h_{\lambda_{3,2}}(\cdot)$, and $h_{\lambda_{4,2}}(\cdot)$ and the optimal density $h_3^*(\cdot)$.

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