

Identification of Scale Parameter of the Matérn Model

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Outline

- Matérn (Whittle) family
 - Covariance
 - Semivariogram
 - Spectral density
 - Stochastic partial differential equation
- Generalized Gaussian random fields
 - Dobrushin, Surgailis
 - Rozanov

Outline 2.

- Orthogonality and equivalence of Gaussian measures
- Levy-Baxter type theorems for stochastic elliptic differential equations
- Orthogonality and equivalence of the Matérn fields
 - Zhang
 - Anderes
 - $d=4$?
- Maximum likelihood estimation of the scale parameter for nonstationary case ($d \leq 3$)
- Orthogonality of Gaussian measures ($d=4$)

Bentil Matem



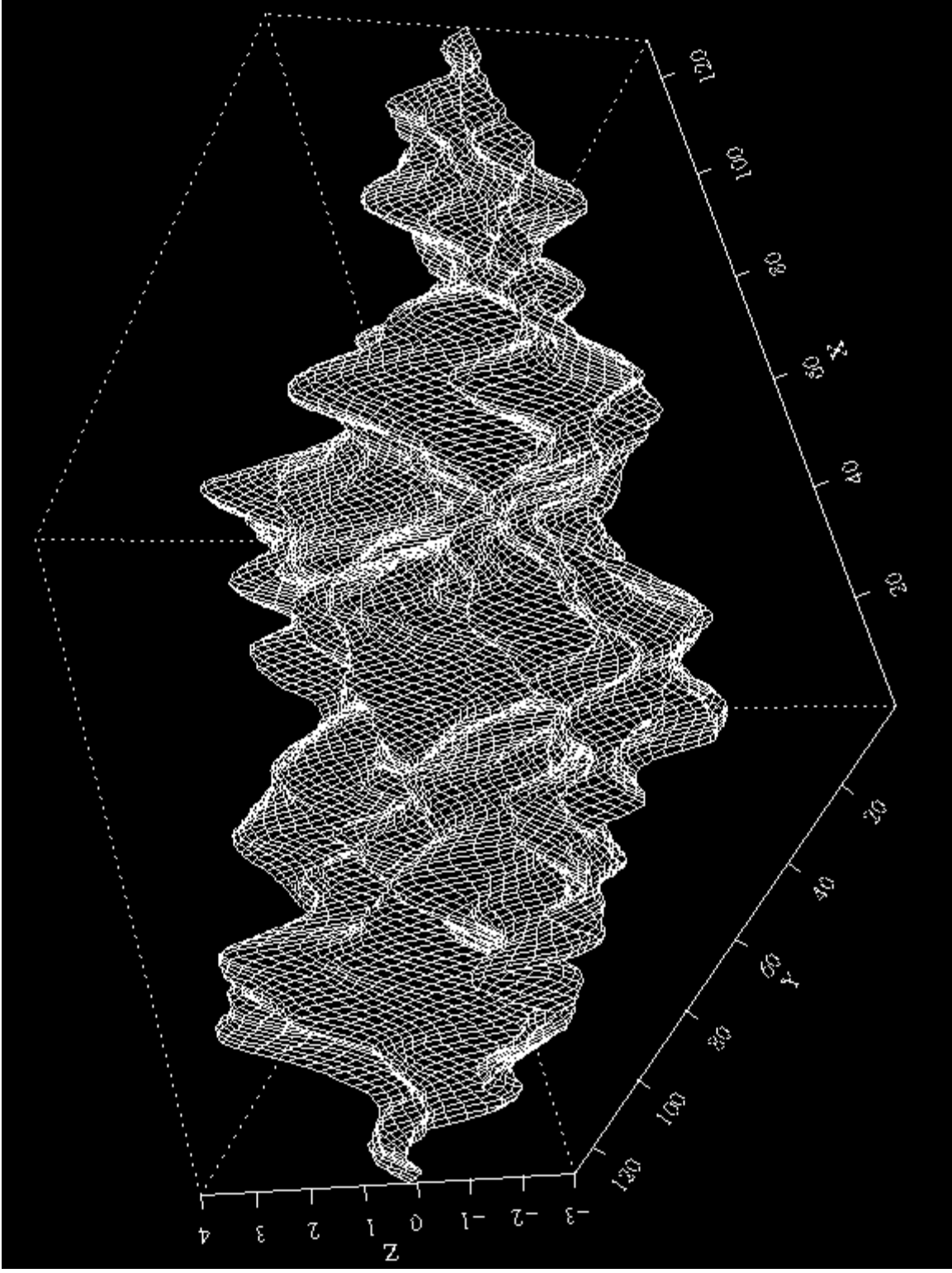
Matérn covariance model

- Matérn (1947, 1960) covariance function between locations $\mathbf{x}, \mathbf{y} \in \mathbf{R}^d$:

$$B(\mathbf{x}, \mathbf{y}) = \rho_\nu (\|\mathbf{x} - \mathbf{y}\|) = \frac{\sigma^2}{2^{\nu-1} \Gamma(\nu)} (\kappa \|\mathbf{x} - \mathbf{y}\|)^\nu K_\nu (\kappa \|\mathbf{x} - \mathbf{y}\|),$$

where

- K_ν is the modified Bessel function of the second kind and order $\nu > 0$,
- ν : smoothness parameter,
- κ : scale parameter,
- σ^2 : marginal variance.



<http://www2.math.umd.edu/~bnk//CLIP/clip.gauss.htm>

Special cases of the Matérn model

($\kappa=1$)

ν	$\rho_\nu(s)$	<i>name</i>
$\frac{1}{3}$	$\frac{2^{2/3}}{\Gamma(1/3)} s^{1/3} K_{1/3}(s)$	<i>Kármán (von Kármán)</i>
1	$sK_1(s)$	<i>Whittle</i>
$\frac{1}{2}$	$\exp(-s)$	<i>Exponential</i>
$\frac{3}{2}$	$(1+s)\exp(-s)$	<i>Second - order autoregressive</i>
$\frac{5}{2}$	$\left(1+s+\frac{1}{3}s^2\right)\exp(-s)$	<i>Third - order autoregressive</i>

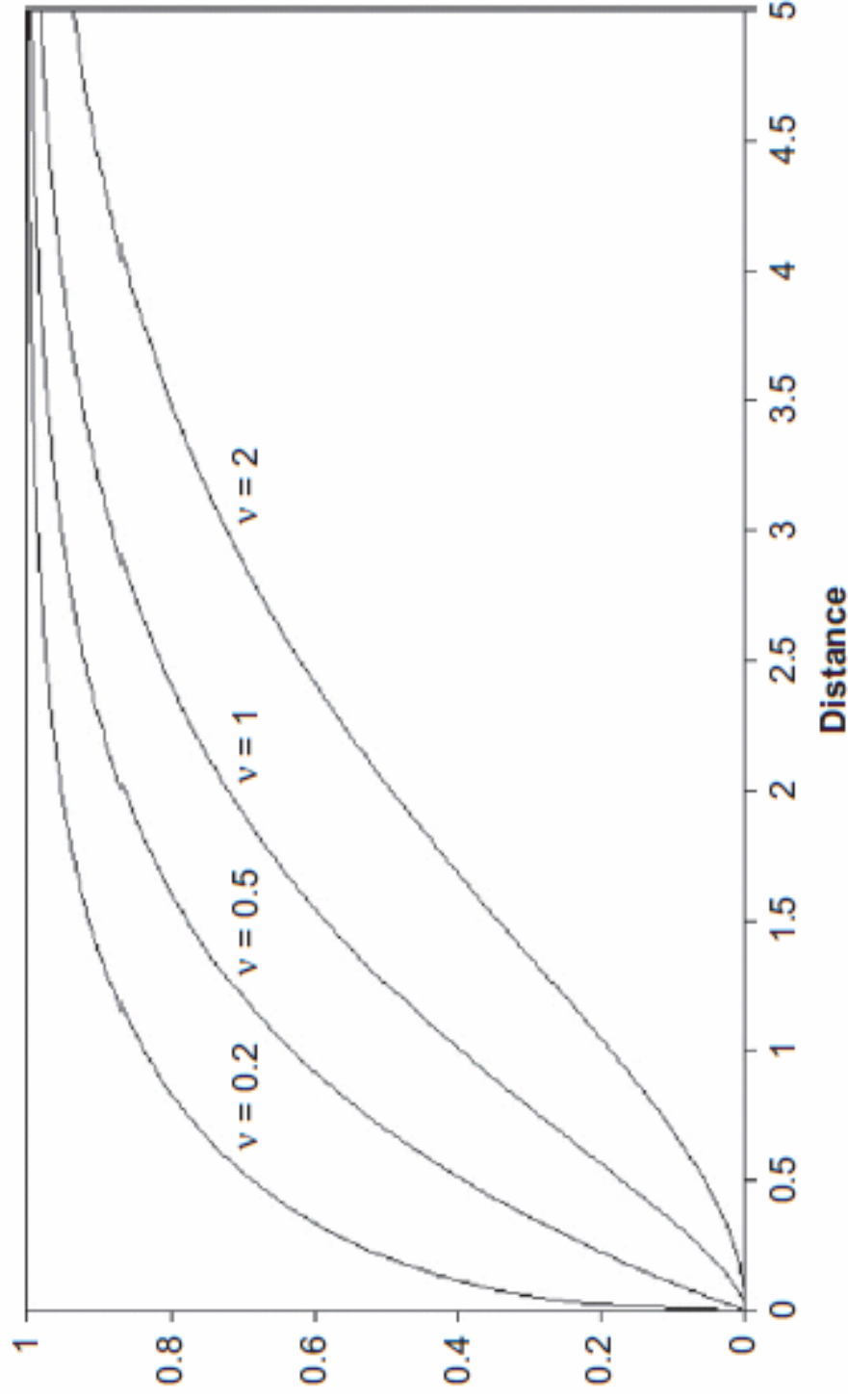
Uses of the Matérn model

- Environmental applications (temperature data)
- Hydrological spatial processes
- Geostatistics (gravitational fields, geodetic networks, satellite sensor images, soil science, topography)
- Machine learning

Semivariogram

- In geostatistics:

$$\gamma_\nu(s) = \sigma^2 - \rho_\nu(s) = \sigma^2 \left(1 - \frac{(\kappa s)^\nu K_\nu(\kappa s)}{2^{\nu-1} \Gamma(\nu)} \right),$$



Spectral density

$$f_\nu(\mathbf{u}) = \frac{\sigma^2 \kappa^{2\nu}}{\pi^{d/2} (\kappa^2 + \|\mathbf{u}\|^2)^{\nu+d/2}},$$

$$B(\mathbf{x}, \mathbf{y}) = \int_{\mathbf{R}^d} \exp(i(\mathbf{x} - \mathbf{y})^T \mathbf{u}) f_\nu(\mathbf{u}) d\mathbf{u}$$

Linear fractional SPDE

- Whittle (1954, 1963):
 - ξ is a zero-mean Gaussian random field and
 - ξ is a solution of the equation:

$$(\kappa^2 - \Delta)^{(v+d/2)/2} \xi = \eta \text{ in } \mathbf{R}^d \quad (\eta : \text{Gaussian white noise})$$

\Rightarrow

$$\text{cov}(\xi(\mathbf{x}), \xi(\mathbf{y})) = \rho_v(\|\mathbf{x} - \mathbf{y}\|)$$

Dobrushin, Surgailis (1979)

$T \subseteq \mathbf{R}^d$ open domain, X is a Gaussian generalized random field on T if

$\{X(\varphi), \varphi \in C_0^\infty(T)\}$ is a system of Gaussian random variables,

X is linear and $X(\varphi)$ in probability as $\varphi \rightarrow 0$.

$$A\varphi = \sum_{|\alpha|, |\beta| \leq p} (-1)^\alpha D^\alpha (a_{\alpha, \beta}(\mathbf{t}) D^\beta \varphi), \varphi \in C_0^\infty(T),$$

is a symmetric, uniformly strongly elliptic operator \Rightarrow

$\exists X$ Gaussian g.r.f. with mean zero

$$\text{and } R(\varphi, \psi) = E(X(\varphi)X(\psi)) = \int_T \varphi(\mathbf{u}) \left(\overline{A} \right)^{-1} \psi(\mathbf{u}) d\mathbf{u}, \text{ where}$$

\overline{A} is the Friedrichs' extension of A .

X can be extended to the Sobolev space $H_0^{-p}(T)$

Dobrushin, Surgailis stationary case

$T = \mathbf{R}^d$ open domain, X is a stationary Gaussian g.r.f

with the spectral density $f(\mathbf{u}) = \frac{1}{(2\pi)^d} P(i\mathbf{u})$, where

$$P(i\mathbf{u}) = \sum_{|\alpha|, |\beta| \leq p} (-1)^\alpha a_{\alpha, \beta}(i\mathbf{u})^\alpha (i\mathbf{u})^\beta, P(i\mathbf{u}) = P(-i\mathbf{u}),$$

$$P(i\mathbf{u}) \geq c(1 + \|\mathbf{u}\|^2)^p, \mathbf{u} \in \mathbf{R}^d.$$

Corresponding operator :

$$A\varphi = \sum_{|\alpha|, |\beta| \leq p} (-1)^\alpha a_{\alpha, \beta} D^\alpha (D^\beta \varphi), \varphi \in C_0^\infty(\mathbf{R}^d).$$

SPDE (ROZANOV)

$$\text{I. } A\varphi = \sum_{|\alpha|, |\beta| \leq p} (-1)^\alpha D^\alpha (a_{\alpha, \beta}(\mathbf{t}) D^\beta \varphi) = \sum_{|\alpha| \leq 2p} b_\alpha(\mathbf{t}) D^\alpha \varphi, \varphi \in C_0^\infty(T),$$

is a positive elliptic operator.

δ is a Gaussian g.r.f. with mean zero and

$$E(\delta(\varphi)\delta(\psi)) = \int_T \varphi(\mathbf{u})(A\psi)(\mathbf{u})d\mathbf{u}.$$

$A\xi = \delta$ in T , if

ξ is a Gaussian g.r.f. with mean zero and $\xi(A\varphi) = \delta(\varphi)$, $\varphi \in C_0^\infty(T)$.

II. ξ is a Gaussian g.r.f. with mean zero and

$$E(\xi(\varphi)\xi(\psi)) = \int_T \varphi(\mathbf{u}) \left(\overline{A} \right)^{-1} \psi(\mathbf{u}) d\mathbf{u}, \varphi, \psi \in C_0^\infty(T) \Rightarrow$$

$A\xi = \delta$ in T .

Orthogonality and equivalence of Gaussian measures

- General results:
 - Feldman
 - Hajek
 - Golosov, Tempelman
- Isotropic fields:
 - Yadrenko
- Etc.

Theorem (Rozanov 1968., p. 46-49.)

Let (Ω, A) be a measurable space with two probability measures Q_1 and Q_2 . Let U be a Hilbert space and $\xi: U \times \Omega \rightarrow \mathbf{R}$ be a Gaussian linear cont. functional with zero mean with respect to both measures and the covariance functions fulfill

$$E_{Q_i}(\xi(u)\xi(v)) = \langle u, B_i v \rangle_U, \forall u, v \in U, i = 1, 2, \text{ where } B_i \text{ are linear}$$

continuous operators and there exists $\varepsilon > 0$ for which

$$\langle B_1 u, u \rangle_U \geq \varepsilon \|u\|_U^2 \quad \forall u \in U.$$

Then the measures Q_1 and Q_2 are equivalent if and only if the operator $(I - B_1^{-1} B_2)$ is a Hilbert - Schmidt operator with eigenvalues $(1 - \sigma_k^2) \neq 1$. In this case the

Radon - Nikodym derivative is given by

$$\frac{dQ_1}{dQ_2} = \lim_{n \rightarrow \infty} \frac{1}{\prod_{k=1}^n \sigma_k} \exp \left\{ -\frac{1}{2} \sum_{k=1}^n \frac{1 - \sigma_k^2}{\sigma_k^2} \xi^2(u_k) \right\}, \text{ where}$$

$\{u_k\}_{k=1}^{\infty}$ is a complete orthonormal system of eigenvectors of the operator $(I - B_1^{-1} B_2)$.

Levy-Baxter type theorems for stochastic elliptic differential equations

In 1988 Goryainov proved that observing a stationary

Gaussian g.r.f with the spectral density $f(\mathbf{u}) = \frac{1}{P(i\mathbf{u})}$, where

P is an elliptic positive polynomial, in an arbitrary small neighborhood of a point t_0 , one can estimate with probability 1 the coefficients b_α , $|\alpha| > 2p - \frac{d}{2}$.

Corollary: if $d > 4$ we can estimate with probability 1 the coefficients κ (scale) and σ^2 (marginal variance) in the Matérn model (if the smoothing parameter is integer).

Arató N. M. (1989 - 1990) proved a similar result for the nonstationary case: if $P\xi = \delta$ in $T \subseteq \mathbf{R}^d$ than observing ξ in an arbitrary small neighborhood of a point $t_0 \in T$, one can estimate with probability 1 the coefficients

$$b_\alpha(t_0), |\alpha| > 2p - \frac{d}{2}.$$

We constructed explicitly test functions $\varphi_{m,n} \in C_0^\infty(S), S \subseteq T$ neighborhood of t_0 and statistics $\sum_m (\xi(\varphi_{m,n}))^2 \xrightarrow{n \rightarrow \infty} b_\alpha(t_0) a.s.$

Equivalence of the measures

- Inoue (1976) and Sokolova (1983) proved that the probability measures corresponding to different elliptic operators and Gaussian fields with covariance

$$\int_T \varphi(\mathbf{u}) \left((\overline{A_i})^{-1} \right) \psi(\mathbf{u}) d\mathbf{u}$$

are equivalent if $\deg(A_1 - A_2) < 2p - d/2$.

Estimation of the parameters of the

Matérn field

- Zhang (2004) proved that observing Matérn field in a bounded set there do not exist consistent estimators of the marginal variance and scale parameter ($d \leq 3$).
- Anderes (2010) constructed consistent estimators for the marginal variance and scale parameter ($d > 4$).
- $d=4$???

d (dimension)=4, ν (smoothing parameter) is integer

Statement :

Let (Ω, A) be a measurable space with two probability measures Q_1 and Q_2 . Let $T \subset \mathbf{R}^d$ be a bounded domain and $\xi : T \times \Omega \rightarrow \mathbf{R}$ be a Gaussian (Matérn) mean zero field with respect to both measures and the covariance functions fulfill

$$E_{Q_i}(\xi(\mathbf{x})\xi(\mathbf{y})) = \frac{\sigma^2}{2^{\nu-1}\Gamma(\nu)} (\kappa_i \|\mathbf{x} - \mathbf{y}\|)^\nu K_\nu(\kappa_i \|\mathbf{x} - \mathbf{y}\|), \forall \mathbf{x}, \mathbf{y} \in T, i = 1, 2,$$

Then the probability measures Q_1 and Q_2 are orthogonal if $d \geq 4$ and ν is integer.

Sketch of the proof

Dobrushin - Surgailis, Rozanov \Rightarrow

$\sigma_i^2 \kappa_i^{2\nu} (\kappa - \Delta)^{\nu+d/2} \xi = \delta$ in T , where

$$E_{Q_i}(\delta_i(\varphi)\delta_i(\psi)) = \int \varphi(\mathbf{x})\sigma_i^2 \kappa_i^{2\nu} (\kappa_i - \Delta)^{\nu+d/2} \psi(\mathbf{x})d\mathbf{x},$$

$\forall \varphi, \psi \in C_0^\infty(T)$.

Goryanov, Arató N.M. \Rightarrow

we can determine $\sigma_i^2 \kappa_i^{2\nu}$ with probability 1.

\Rightarrow we can assume that $\sigma_i^2 \kappa_i^{2\nu} = 1$.

The g.r.f. δ_1 can be extend to Sobolev space $U = H_0^{\nu+d/2}$

with the equivalent norm $\|u\| = \int_T u(\mathbf{x})(\kappa_1 - \Delta)^{\nu+d/2} u(\mathbf{x})d\mathbf{x}$

$$\Rightarrow E_{Q_i}(\delta(u)\delta(v)) = \langle u, v \rangle_U$$

$$E_{Q_2}(\delta(u)\delta(v)) = \left\langle u, (\kappa_2 - \Delta)^{-(v+d/2)} (\kappa_1 - \Delta)^{v+d/2} v \right\rangle_U \Rightarrow$$

$$B_1 = I, B_2 = (\kappa_2 - \Delta)^{-(v+d/2)} (\kappa_1 - \Delta)^{v+d/2},$$

$$I - B_1^{-1} B_2 = I - (\kappa_2 - \Delta)^{-(v+d/2)} (\kappa_1 - \Delta)^{v+d/2}.$$

We can choose a complete orthogonal system $\{u_k\}_{k=1}^{\infty}$ in $H_0^{v+d/2}$:

$$-\Delta u_k = \lambda_k u_k \text{ in } T,$$

$$\alpha_0 k^{2/d} < \lambda_k < \alpha_1 k^{2/d}, \text{ for } k \geq k_0, (\alpha_0, \alpha_1, k_0 > 0).$$

$$u = \sum_k c_k u_k \in U:$$

$$(I - B_1^{-1} B_2)u = (I - (\kappa_2 - \Delta)^{-(v+d/2)} (\kappa_1 - \Delta)^{v+d/2})u =$$

$$\sum_k \left(1 - \left(\frac{\kappa_1 + \lambda_k}{\kappa_2 + \lambda_k} \right)^{v+d/2} \right) c_k u_k$$

eigenvalues $(1 - \sigma_k^2)$ of the $I - B_1^{-1} B_2$

are $1 - \left(\frac{\kappa_1 + \lambda_k}{\kappa_2 + \lambda_k} \right)^{\nu+d/2}$ and

$$\sum_k (1 - \sigma_k^2)^2 = \sum_k \left(1 - \left(\frac{\kappa_1 + \lambda_k}{\kappa_2 + \lambda_k} \right)^{\nu+d/2} \right)^2 = \infty \text{ if } d \geq 4.$$

Nonstationary case (and $v+d/2$ is integer)

Statement :

Let (Ω, A) be a measurable space with two probability measures Q_1 and Q_2 . Let $T \subset \mathbf{R}^d$ be a bounded domain and $\xi : T \times \Omega \rightarrow \mathbf{R}$ be a Dobrushin - Surgailis field with respect to both measures and $(\kappa_T - \Delta)^{v+d/2}$ differential operators.

Then the probability measures Q_1 and Q_2 are equivalent if and only if $d < 4$.

At the case $d < 4$

the Radon - Nikodym derivative is the following

$$\frac{dQ_2}{dQ_1} = \lim_{n \rightarrow \infty} \left\{ \left(\prod_{k=1}^n \frac{\kappa_2 + \lambda_k}{\kappa_1 + \lambda_k} \right)^{(v+d/2)/2} \exp \left(-\frac{1}{2} \sum_{k=1}^n \frac{(\kappa_2 + \lambda_k)^{v+d/2} - (\kappa_1 + \lambda_k)^{v+d/2}}{(\kappa_1 + \lambda_k)^{v+d/2}} \xi^2(u_k) \right) \right\},$$

where $\{u_k\}_{k=1}^{\infty}$ is a complete orthonormal system of the eigenfunctions of the operator - Δ .

Maximum likelihood estimation of the mean

If we observe in a sphere T the random field $m+\xi$ where m is an unknown constant and ξ is a Matérn field ($\alpha=(\nu+d/2)/2$ is integer) then the maximum likelihood estimator of the m has the following form.

$$\hat{m} = q \int_T \xi(\mathbf{x}) d\mathbf{x} + q_0 \int_{\partial T} \xi(\mathbf{x}) d\mathbf{x} + \sum_{i=1}^{\alpha-1} q_i \int_{\partial T} (\partial_{n,i} \xi)(\mathbf{x}) d\mathbf{x},$$

$\partial_{n,1} \xi, \dots, \partial_{n,\alpha-1} \xi$: generalized normal derivatives of ξ

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