

Gaussian Processes Generated By Monotonically Modulated Stationary Gaussian Process Kernels

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Abstract

This paper investigates the properties of Gaussian processes generated by monotonically modulating the kernels of stationary Gaussian processes. A comprehensive analysis is presented of the relationship between original and modulated kernel eigenfunctions, demonstrating that the eigenfunctions of the modulated kernel are compositions of the original kernel's eigenfunctions with the modulating function, scaled by the square roots of modulating function's derivative. It is established that this transformation preserves both normalization and eigenvalues, providing an explicit isometry between the original and modulated kernel Hilbert spaces. Furthermore, the expected number of zeros of the process over $[0, T]$ is shown to be $\mathbb{E}[N([0, T])] = \sqrt{-K(0)}(f(T) - f(0))$, providing fundamental insights into how modulation by monotonic functions transform stationary Gaussian processes.

Definition 1

Let \mathcal{F} denote the class of functions $f: \mathbb{R} \rightarrow \mathbb{R}$ which are:

1. piecewise continuous with piecewise continuous first derivative,
2. strictly monotonically increasing

$$f(t) < f(s) \forall -\infty \leq t < s \leq \infty \quad (1)$$

3. and have a finite limiting derivative at infinity

$$\lim_{t \rightarrow \infty} \dot{f}(t) < \infty \quad (2)$$

Theorem 2

(Eigenfunctions) For any stationary kernel $K(t, s) = K(|t - s|)$, the eigenfunctions of the modulated kernel

$$K_f(s, t) = K(|f(t) - f(s)|) \quad (3)$$

take the form:

$$\phi_n(t) = \psi_n(f(t)) \sqrt{\dot{f}(t)} \quad (4)$$

where $f \in \mathcal{F}$ and ψ_n are the normalized eigenfunctions of the original unmodulated kernel $K(|t - s|)$.

Proof. The eigenfunction equation for the modulated kernel is:

$$\int_{-\infty}^{\infty} K(|f(t) - f(s)|) \phi_n(s) ds = \lambda_n \phi_n(t) \quad (5)$$

The variables can be changed by substituting $u = f(s)$, $v = f(t)$:

$$\int_{-\infty}^{\infty} K(|v - u|) \frac{\phi_n(f^{-1}(u))}{\dot{f}(f^{-1}(u))} du = \lambda_n \phi_n(f^{-1}(v)) \quad (6)$$

which is valid due to the strict monotonicity of f which assures its invertability. Let

$$\psi_n(u) = \frac{\phi_n(f^{-1}(u))}{\sqrt{\dot{f}(f^{-1}(u))}} \quad (7)$$

Then:

$$\int_{-\infty}^{\infty} K(|v - u|) \psi_n(u) du = \lambda_n \psi_n(v) \quad (8)$$

This is precisely the eigenfunction equation for the original kernel $K(|t - s|)$. Therefore, if ψ_n are the eigenfunctions of the original kernel, then

$$\phi_n(t) = \psi_n(f(t)) \sqrt{\dot{f}(t)} \quad (9)$$

are the eigenfunctions of the modulated kernel. □

Theorem 3

(Normalization) If ψ_n are normalized eigenfunctions of the original kernel, then $\phi_n(t) = \psi_n(f(t)) \sqrt{\dot{f}(t)}$ are automatically normalized eigenfunctions of the modulated kernel, requiring no additional normalization constants.

Proof. For normalized ψ_n :

$$\int_{-\infty}^{\infty} |\phi_n(t)|^2 dt = \int_{-\infty}^{\infty} |\psi_n(f(t))|^2 \dot{f}(t) dt \quad (10)$$

Under the change of variables $u = f(t)$:

$$\int_{-\infty}^{\infty} |\psi_n(u)|^2 du = 1 \quad (11)$$

Therefore the ϕ_n are already normalized without additional constants. \square

Corollary 4

(Eigenvalue Invariance) *The eigenvalues $\{\lambda_n\}$ of the modulated kernel K_f are identical to those of the original kernel K .*

Remark 5. This result demonstrates that monotonic modulation preserves the spectral structure of any stationary kernel through composition with the modulation function. The transformation operator

$$(T\phi)(t) = \sqrt{\dot{f}(t)} \phi(f(t)) \quad (12)$$

provides an explicit isometry between the original and modulated kernel Hilbert spaces, explaining why no additional normalization constants are needed.

Theorem 6

(Mean Zero-Counting Function) *Let $f \in \mathcal{F}$ and let $K(\cdot)$ be any positive-definite, stationary covariance function, twice differentiable at 0. Consider the centered Gaussian process with covariance*

$$K_f(s, t) = K(|f(t) - f(s)|) \quad (13)$$

Then the expected number of zeros in $[0, T]$ is

$$\mathbb{E}[N([0, T])] = \sqrt{-\ddot{K}(0)} (f(T) - f(0)) \quad (14)$$

Proof. By the Kac-Rice formula:

$$\mathbb{E}[N([0, T])] = \int_0^T \sqrt{-\lim_{s \rightarrow t} \frac{\partial^2}{\partial t \partial s} K_f(s, t)} dt \quad (15)$$

Computing the mixed partial derivative and taking the limit as $s \rightarrow t$:

$$\lim_{s \rightarrow t} \frac{\partial^2}{\partial t \partial s} K_f(s, t) = -\ddot{K}(0) \dot{f}(t)^2 \quad (16)$$

Therefore

$$\mathbb{E}[N([0, T])] = \sqrt{-\ddot{K}(0)} \int_0^T \dot{f}(t) dt = \sqrt{-\ddot{K}(0)} (f(T) - f(0)) \quad (17)$$

so that

$$\begin{aligned} \sqrt{-\ddot{K}(0)} (f(T) - f(0)) &= \sqrt{-\ddot{K}(0)} \int_0^T \dot{f}(t) dt \\ &= \int_0^T \sqrt{-\ddot{K}(0) \dot{f}(t)^2} dt \\ &= \int_0^T \sqrt{-\lim_{s \rightarrow t} \frac{\partial^2}{\partial t \partial s} K(|f(t) - f(s)|)} dt \end{aligned} \quad (18)$$

which is precisely the Kac-Rice formula for the expected zero-count. \square