Unitary Time Changes of Stationary Processes Yield Oscillatory Processes

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Abstract

A unitary time-change operator U_{θ} is constructed for absolutely continuous, strictly increasing time reparametrizations θ , acting on functions that are square-integrable over compact sets. Applying U_{θ} to the Cramér spectral representation of a stationary process yields an oscillatory process in the sense of Priestley with oscillatory function $\varphi_t(\lambda) = \sqrt{\dot{\theta}(t)} \ e^{i\lambda\theta(t)}$ and evolutionary spectrum $dF_t(\lambda) = \dot{\theta}(t)dF(\lambda)$. It is proved that sample paths of any non-degenerate second-order stationary process almost surely lie in $L^2_{\text{loc}}(\mathbb{R})$, making the operator applicable to typical realizations. A zero-localization measure $d\mu(t) = \delta(Z(t)) |\dot{Z}(t)| \ dt$ induces a Hilbert space $L^2(\mu)$ on the zero set of an oscillatory process Z, and the multiplication operator (Lf)(t) = tf(t) has simple pure point spectrum equal to the zero crossing set of Z. This produces a concrete operator scaffold consistent with a Hilbert–Pólya-type viewpoint.

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1 Function Spaces

1.1 σ -compact sets and locally square-integrable functions

Definition 1. [σ -compact sets] A subset $U \subseteq \mathbb{R}$ is σ -compact if

$$U = \bigcup_{n=1}^{\infty} K_n \tag{1}$$

with each K_n compact.

Definition 2. [Locally square-integrable functions] Define

$$L^2_{\mathrm{loc}}(\mathbb{R}) := \left\{ f \colon \mathbb{R} \to \mathbb{C} \colon \int_K |f(t)|^2 \ dt < \infty \ for \ every \ compact \ K \subseteq \mathbb{R} \right\} \tag{2}$$

Remark 3. Every bounded measurable set in \mathbb{R} is compact or contained in a compact set; hence $L^2_{loc}(\mathbb{R})$ contains functions that are square-integrable on every bounded interval, including functions with polynomial growth at infinity.

2 Gaussian Processes

A Gaussian process is a ...

2.1 Stationary processes

Definition 4. [Cramér representation] A zero-mean stationary process X with spectral measure F admits the sample path representation

$$X(t) = \int_{\mathbb{R}} e^{i\lambda t} \,\Phi\left(d\lambda\right) \tag{3}$$

which has covariance

$$R_X(t-s) = \int_{\mathbb{R}} e^{i\lambda(t-s)} dF(\lambda)$$
 (4)

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2.2 Oscillatory Processes

A particularly tractable class of non-stationary Gaussian processes is that of the oscillatory processes as defined by M.B. Priestley in 1965[1].

Definition 5. [Oscillatory process] link to Priestley 1965 Let F be a finite nonnegative Borel measure on \mathbb{R} . Let

$$A_t \in L^2(F) \forall t \in \mathbb{R} \tag{5}$$

be the gain function and

$$\varphi_t(\lambda) = A_t(\lambda) e^{i\lambda t} \tag{6}$$

be the corresponding oscillatory function then an oscillatory process is a stochastic process which can be represented as

$$Z(t) = \int_{\mathbb{R}} \varphi_t(\lambda) d\Phi(\lambda)$$

$$= \int_{\mathbb{R}} A_t(\lambda) e^{i\lambda t} d\Phi(\lambda)$$
(7)

where Φ is a complex orthogonal random measure with spectral measure F which satisfies the relation

$$d\mathbb{E}[\Phi(\lambda)\overline{\Phi(\mu)}] = \delta(\lambda - \mu) dF(\lambda)$$
(8)

and has the corresponding covariance kernel

$$R_{Z}(t,s) = \mathbb{E}[Z(t)\overline{Z(s)}]$$

$$= \int_{\mathbb{R}} A_{t}(\lambda) \overline{A_{s}(\lambda)} e^{i\lambda(t-s)} dF(\lambda)$$

$$= \int_{\mathbb{R}} \phi_{t}(\lambda) \overline{\phi_{s}(\lambda)} dF(\lambda)$$
(9)

Theorem 6. [Real-valuedness criterion for oscillatory processes] Let Z be an oscillatory process with oscillatory function

$$\varphi_t(\lambda) = A_t(\lambda) e^{i\lambda t} \tag{10}$$

and spectral measure F. Then Z is real-valued if and only if

$$A_t(-\lambda) = \overline{A_t(\lambda)} \tag{11}$$

for F-almost every $\lambda \in \mathbb{R}$, equivalently

$$\varphi_t\left(-\lambda\right) = \overline{\varphi_t(\lambda)} \tag{12}$$

for F-almost every $\lambda \in \mathbb{R}$.

Proof. Assume Z is real-valued, i.e.

$$Z(t) = \overline{Z(t)} \quad \forall t \in \mathbb{R}$$
 (13)

Writing its oscillatory representation,

$$Z(t) = \int_{\mathbb{R}} A_t(\lambda) e^{i\lambda t} d\Phi(\lambda)$$
 (14)

and taking the complex conjugate gives

$$\overline{Z(t)} = \int_{\mathbb{R}} \overline{A_t(\lambda)} e^{-i\lambda t} d\overline{\Phi(\lambda)}$$
(15)

For a real-valued process, the orthogonal random measure Φ must satisfy

$$d\overline{\Phi(\lambda)} = -d\Phi(\lambda) \tag{16}$$

which ensures that the spectral representation produces real values. Substituting this identity and using the substitution

$$\mu = -\lambda \tag{17}$$

it is shown that

$$\overline{Z(t)} = \int_{\mathbb{R}} \overline{A_t(-\mu)} e^{i\mu t} d\Phi(\mu)$$
(18)

Since $Z(t) = \overline{Z(t)}$, comparison of the integrands (which are unique elements of $L^2(F)$) yields

$$A_t(\lambda) = \overline{A_t(-\lambda)}$$
 for F -a.e. λ (19)

Equivalently, because the oscillatory function (6) is given by

$$\varphi_t(\lambda) = A_t(\lambda) e^{i\lambda t} \tag{20}$$

we have

$$\varphi_t(-\lambda) = \overline{\varphi_t(\lambda)}$$
 for F -a.e. λ (21)

Conversely, if

$$A_t(-\lambda) = \overline{A_t(\lambda)} \tag{22}$$

for F-a.e. λ , then the same substitution shows that

$$\overline{Z(t)} = Z(t) \quad \forall t \in \mathbb{R} \tag{23}$$

so Z is real-valued. This completes the proof.

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Theorem 7. [Existence] Let F be an absolutely continuous spectral measure and the gain function

$$A_t(\lambda) \in L^2(F) \forall \mathbb{R} \ni t < \infty \tag{24}$$

be measurable in both time and frequency then the time-dependent spectral density is defined by

$$S_t(\lambda) = \int_{\mathbb{R}} |A_t(\lambda)|^2 dF(\lambda) < \infty$$

$$= \int_{\mathbb{R}} |A_t(\lambda)|^2 S(\lambda) d\lambda$$
(25)

and there exists a complex orthogonal random measure Φ with spectral measure F such that for each sample path $\varpi \in \Theta$ in the space of sample paths having given covariance constituting the ensemble denoted Θ

$$Z(t) = \int_{\mathbb{R}} A_t(\lambda) e^{i\lambda t} d\Phi(\lambda)$$
 (26)

is well-defined in $L^2(\Omega)$ and has covariance R_Z as in (9) above.

Proof. The proof proceeds by constructing the stochastic integral using the standard extension procedure. First, the integral is defined for simple functions of the form

$$g(\lambda) = \lim_{n \to \infty} \sum_{j=1}^{n} c_j \mathbf{1}_{E_j}(\lambda)$$
 (27)

where $\{E_j\}$ are disjoint Borel sets with $F(E_j) < \infty$ and $c_j \in \mathbb{C}$:

$$\int_{\mathbb{R}} g(\lambda) \, d\Phi(\lambda) = \lim_{n \to \infty} \sum_{j=1}^{n} c_j \, \Phi(E_j)$$
(28)

For simple functions such as this, the isometry property holds:

$$\mathbb{E}\left[\left|\int_{\mathbb{R}} g(\lambda) \, \Phi d(\lambda)\right|^{2}\right] = \mathbb{E}\left[\lim_{n \to \infty} \left|\sum_{j=1}^{n} c_{j} \Phi(E_{j})\right|^{2}\right]$$

$$= \lim_{n \to \infty} \sum_{j=1}^{n} \sum_{k=1}^{n} c_{j} \bar{c_{k}} \mathbb{E}\left[\Phi(E_{j}) \overline{\Phi(E_{k})}\right]$$

$$= \lim_{n \to \infty} \sum_{j=1}^{n} |c_{j}|^{2} F(E_{j})$$

$$= \int_{\mathbb{R}} |g(\lambda)|^{2} dF(\lambda)$$

$$(29)$$

Since simple functions are dense in $L^2(F)$, the integral is extended by continuity $\forall g \in L^2(F)$ since the oscillatory function (6) is defined by

$$\varphi_t(\lambda) = A_t(\lambda) e^{i\lambda t} \in L^2(F) \forall t \in \mathbb{R}$$
(30)

and $A_t \in$. Therefore

$$Z(t) = \int_{\mathbb{R}} \varphi_t(\lambda) \, d\Phi(\lambda) \tag{31}$$

is well-defined in $L^2(\Omega)$. The covariance is computed as:

$$R_{Z}(t,s) = \mathbb{E}[Z(t)\overline{Z(s)}]$$

$$= \mathbb{E}\left[\int_{\mathbb{R}} \varphi_{t}(\lambda) d\Phi(\lambda) \int_{\mathbb{R}} \overline{\varphi_{s}(\mu)} d\overline{\Phi(\mu)}\right]$$

$$= \int_{\mathbb{R}} \int_{\mathbb{R}} \varphi_{t}(\lambda) \overline{\varphi_{s}(\mu)} d\mathbb{E}\left[\Phi(\lambda)\overline{\Phi(\mu)}\right]$$

$$= \int_{\mathbb{R}} \varphi_{t}(\lambda) \overline{\varphi_{s}(\lambda)} dF(\lambda)$$

$$= \int_{\mathbb{R}} A_{t}(\lambda) \overline{A_{s}(\lambda)} e^{i\lambda(t-s)} dF(\lambda)$$
(32)

3 Unitarily Time-Changed Stationary Processes

3.1 Unitary time-change operator $U_{\theta} f$

Definition 8. [Unitary time-change] Let the time-scaling function $\theta: \mathbb{R} \to \mathbb{R}$ be absolutely continuous, strictly increasing, and bijective, with $\dot{\theta}(t) > 0$ almost everywhere and $\dot{\theta}(t) = 0$ only on sets of Lebesgue measure zero. The function θ maps σ -compact sets to σ -compact sets. Define, for f measurable,

$$(U_{\theta} f)(t) = \sqrt{\dot{\theta}(t)} f(\theta(t)) \tag{33}$$

Proposition 9. [Inversion of Unitary time-change] The inverse of the unitary time-change operator U in Equation (33) is given by

$$(U_{\theta}^{-1}g)(s) = \frac{g(\theta^{-1}(s))}{\sqrt{\dot{\theta}(\theta^{-1}(s))}}$$
(34)

which is well-defined almost everywhere on every σ -compact set.

Proof. Since $\dot{\theta}(t) = 0$ only on sets of measure zero, and θ^{-1} maps sets of measure zero to sets of measure zero because of the fact that absolutely continuous bijective functions preserve measure-zero sets, the denominator $\sqrt{\dot{\theta}(\theta^{-1}(s))}$ is positive almost everywhere. The expression is therefore well-defined almost everywhere on every σ -compact set, which suffices for defining an element of $L^2_{\rm loc}(\mathbb{R})$.

Theorem 10. [Local unitarity on σ -compact sets] For every σ -compact set $C \subseteq \mathbb{R}$ and $f \in L^2_{loc}(\mathbb{R})$,

$$\int_{C} |(U_{\theta} f)(t)|^{2} dt = \int_{\theta(C)} |f(s)|^{2} ds$$
(35)

Moreover, U_{θ}^{-1} is the inverse of U_{θ} on $L_{loc}^2(\mathbb{R})$.

Proof. Let $f \in L^2_{loc}(\mathbb{R})$ and let C be any σ -compact set. The local L^2 -norm of $U_{\theta} f$ over C is:

$$\int_{C} |(U_{\theta} f)(t)|^{2} dt = \int_{C} \left| \sqrt{\dot{\theta}(t)} f(\theta(t)) \right|^{2} dt
= \int_{C} \dot{\theta}(t) |f(\theta(t))|^{2} dt$$
(36)

Since θ is absolutely continuous and strictly increasing, applying the change of variables $s = \theta(t)$ gives

$$ds = \dot{\theta}(t) dt \tag{37}$$

almost everywhere. Since θ maps σ -compact sets to σ -compact sets, as t ranges over C, $s = \theta(t)$ ranges over $\theta(C)$, which is σ -compact. Therefore:

$$\int_{C} \dot{\theta}(t) |f(\theta(t))|^{2} dt = \int_{\theta(C)} |f(s)|^{2} ds$$
(38)

To verify that U_{θ}^{-1} is indeed the inverse, it is seen that:

$$(U_{\theta}^{-1}U_{\theta}f)(s) = \left(U_{\theta}^{-1}\sqrt{\dot{\theta}(s)}f(\theta(s))\right)(s)$$

$$= \frac{\sqrt{\dot{\theta}(\theta^{-1}(s))}}{\sqrt{\dot{\theta}(\theta^{-1}(s))}}f(\theta(\theta^{-1}(s))) \qquad \forall f \in L^{2}_{loc}(\mathbb{R})$$

$$= f(s)$$
(39)

since

$$\theta(\theta^{-1}(s)) = s \tag{40}$$

and similarly, its also plain to see that:

$$(U_{\theta}U_{\theta}^{-1}g)(t) = \sqrt{\dot{\theta}(t)} (U_{\theta}^{-1}g)(\theta(t))$$

$$= \frac{\sqrt{\dot{\theta}(t)}}{\sqrt{\dot{\theta}(\theta^{-1}(\theta(t)))}} g(\theta^{-1}(\theta(t))) \qquad \forall g \in L_{loc}^{2}(\mathbb{R})$$

$$= \frac{\sqrt{\dot{\theta}(t)}}{\sqrt{\dot{\theta}(t)}} g(t)$$

$$= g(t)$$

$$(41)$$

since

$$\theta^{-1}(\theta(t)) = t \tag{42}$$

Therefore

$$(U_{\theta} U_{\theta}^{-1} f)(t) = (U_{\theta}^{-1} U_{\theta} f)(t)$$

$$= f(t)$$
(43)

on
$$L^2_{\mathrm{loc}}(\mathbb{R})$$
.

3.1.1 Inverse Filter for Unitary Time Transformations

Theorem 11. [Inverse Filter for Unitary Time Transformations] Let $\theta: \mathbb{R} \to \mathbb{R}$ be absolutely continuous, strictly increasing, and bijective with $\theta'(t) > 0$ almost everywhere. Let Y(u) be a stationary process with unit variance, and define

$$Z(t) = \sqrt{\dot{\theta}(t)} \ Y(\theta(t)) \tag{44}$$

as the oscillatory process obtained by the unitary time transformation. Then:

1. The forward filter kernel is

$$h(t, u) = \sqrt{\dot{\theta}(t)} \, \delta \left(u - \theta(t) \right) \tag{45}$$

2. The inverse filter kernel is

$$g(t,s) = \frac{\delta(s - \theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))}}$$

$$\tag{46}$$

3. The composition $(g \circ h)$ recovers the identity:

$$Y(t) = \int_{\mathbb{R}} g(t,s) Z(s) ds$$

$$= \frac{Z(\theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))}}$$
(47)

Proof. Parts 1–3 are established in sequence.

Let

$$v = \theta(s) \text{ and } s = \theta^{-1}(v)$$
 (48)

so that

$$ds = \frac{1}{\dot{\theta}(\theta^{-1}(v))} dv \tag{49}$$

and

$$\delta(\theta^{-1}(v) - \theta^{-1}(t)) = \dot{\theta}(\theta^{-1}(t)) \,\,\delta(v - t) \tag{50}$$

then substitute the inverse filter in Equation (45)

$$g(t,s) = \frac{\delta(s - \theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))}}$$
(51)

and unitarily time-changed stationary process operator representation (44)

$$Z(t) = (U_{\theta}Y)(t)$$

$$= \sqrt{\dot{\theta}(t)} Y(\theta(t))$$
(52)

to verify that each of the equivalent expressions

$$Y(t) = \int_{\mathbb{R}} g(t,s) Z(s) ds$$

$$= \int_{\mathbb{R}} \frac{\delta(s - \theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))}} Z(s) ds$$

$$= \int_{\mathbb{R}} \frac{\delta(s - \theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))}} (U_{\theta}Y)(s) ds$$

$$= \int_{\mathbb{R}} \frac{\delta(s - \theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))}} \sqrt{\dot{\theta}(s)} Y(\theta(s)) ds$$

$$= \int_{\mathbb{R}} \frac{\delta(\theta^{-1}(v) - \theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))}} \frac{\sqrt{\dot{\theta}(\theta^{-1}(v))}}{\dot{\theta}(\theta^{-1}(v))} Y(v) dv$$

$$= \int_{\mathbb{R}} \frac{\delta(\theta^{-1}(v) - \theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))} \sqrt{\dot{\theta}(\theta^{-1}(v))}} Y(v) dv$$

$$= \int_{\mathbb{R}} \frac{\dot{\theta}(\theta^{-1}(v) - \theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))} \sqrt{\dot{\theta}(\theta^{-1}(v))}} Y(v) dv$$

$$= \int_{\mathbb{R}} \frac{\dot{\theta}(\theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))} \sqrt{\dot{\theta}(\theta^{-1}(v))}} Y(v) dv$$

$$= \frac{\dot{\theta}(\theta^{-1}(t))}{\sqrt{\dot{\theta}(\theta^{-1}(t))} \sqrt{\dot{\theta}(\theta^{-1}(t))}} Y(t)$$

$$= \frac{\dot{\theta}(\theta^{-1}(t))}{\dot{\theta}(\theta^{-1}(t))} Y(t)$$

3.2 Transformation of Stationary \rightarrow Oscillatory Processes via U_{θ}

Theorem 12. [Unitary time change yields oscillatory process] Let X be zero-mean stationary as in Definition 4. For scaling function θ as in Definition 8, define

$$Z(t) = (U_{\theta} X)(t)$$

$$= \sqrt{\dot{\theta}(t)} X(\theta(t))$$
(54)

Then Z is a realization of an oscillatory process with oscillatory function

$$\varphi_t(\lambda) = \sqrt{\dot{\theta}(t)} \ e^{i\lambda\theta(t)} \tag{55}$$

 $gain\ function$

$$A_t(\lambda) = \sqrt{\dot{\theta}(t)} e^{i\lambda(\theta(t) - t)}$$
(56)

and covariance

$$R_{Z}(t,s) = \mathbb{E}[Z(t)\overline{Z(s)}]$$

$$= \mathbb{E}\left[\sqrt{\dot{\theta}(t)} X(\theta(t)) \sqrt{\dot{\theta}(t)} X(\theta(t))\right]$$

$$= \sqrt{\dot{\theta}(t) \dot{\theta}(s)} \mathbb{E}[X(\theta(t)) \overline{X(\theta(s))}]$$

$$= \sqrt{\dot{\theta}(t) \dot{\theta}(s)} R_{X} (\theta(t) - \theta(s))$$

$$= \sqrt{\dot{\theta}(t) \dot{\theta}(s)} \int_{\mathbb{R}} e^{i\lambda(\theta(t) - \theta(s))} dF(\lambda)$$
(57)

Proof. Applying the unitary time change operator to the spectral representation of X(t):

$$Z(t) = (U_{\theta} X)(t)$$

$$= \sqrt{\dot{\theta}(t)} X(\theta(t))$$

$$= \sqrt{\dot{\theta}(t)} \int_{\mathbb{R}} e^{i\lambda\theta(t)} d\Phi(\lambda)$$

$$= \int_{\mathbb{R}} \sqrt{\dot{\theta}(t)} e^{i\lambda\theta(t)} d\Phi(\lambda)$$

$$= \int_{\mathbb{R}} \varphi_t(\lambda) d\Phi(\lambda)$$
(58)

where

$$\varphi_t(\lambda) = \sqrt{\dot{\theta}(t)} e^{i\lambda\theta(t)} \tag{59}$$

To verify this constitutes an oscillatory representation according to Definition 5, $\varphi_t(\lambda)$ has

the form $A_t(\lambda) e^{i\lambda t}$:

$$\varphi_{t}(\lambda) = \sqrt{\dot{\theta}(t)} e^{i\lambda\theta(t)}
= \sqrt{\dot{\theta}(t)} e^{i\lambda(\theta(t)-t)} e^{i\lambda t}
= A_{t}(\lambda) e^{i\lambda t}$$
(60)

where

$$A_t(\lambda) = \sqrt{\dot{\theta}(t)} e^{i\lambda(\theta(t) - t)}$$
(61)

Since $\dot{\theta}(t) \geq 0$ almost everywhere and $\dot{\theta}(t) = 0$ only on sets of measure zero, $A_t(\lambda)$ is well defined almost everywhere. Moreover, $A_t \in L^2(F)$ for each t since:

$$\int_{\mathbb{R}} |A_{t}(\lambda)|^{2} dF(\lambda) = \int_{\mathbb{R}} \left| \sqrt{\dot{\theta}(t)} e^{i\lambda(\theta(t)-t)} \right|^{2} dF(\lambda)
= \int_{\mathbb{R}} \dot{\theta}(t) |e^{i\lambda(\theta(t)-t)}|^{2} dF(\lambda)
= \dot{\theta}(t) \int_{\mathbb{R}} dF(\lambda)
= \dot{\theta}(t) F(\mathbb{R}) < \infty$$
(62)

where $|e^{i\alpha}| = 1$ for all real α is used. The covariance (57) is computed by substituting the spectral representation and applying Fubuni's theorem to interchange the order of operations.

 $(63) \square$

Corollary 13. [Evolutionary spectrum of unitarily time-changed stationary process][1] Link to The evolutionary spectrum, also called the time-varying spectral density, is

$$dF_t(\lambda) = |A_t(\lambda)|^2 dF(\lambda)$$

= $\dot{\theta}(t)dF(\lambda)$ (64)

Proof. By definition of the evolutionary spectrum and using the gain function from Theorem 12:

$$dF_{t}(\lambda) = |A_{t}(\lambda)|^{2} dF(\lambda)$$

$$= \left| \sqrt{\dot{\theta}(t)} e^{i\lambda(\theta(t)-t)} \right|^{2} dF(\lambda)$$

$$= \dot{\theta}(t) |e^{i\lambda(\theta(t)-t)}|^{2} dF(\lambda)$$

$$= \dot{\theta}(t) dF(\lambda)$$
(65)

since

$$|e^{i\alpha}| = 1 \forall a \in \mathbb{R} \tag{66} \quad \Box$$

3.3 Covariance operator conjugation

Proposition 14. [Operator conjugation] Let

$$(T_K f)(t) := \int_{\mathbb{R}} K(|t - s|) f(s) \, ds \tag{67}$$

with stationary kernel

$$K(h) = \int_{\mathbb{R}} e^{i\lambda h} dF(\lambda) \tag{68}$$

Define the transformed kernel

$$K_{\theta}(s,t) := \sqrt{\dot{\theta}(t)\,\dot{\theta}(s)} \ K(|\theta(t) - \theta(s)|) \tag{69}$$

then the corresponding integral covariance operator is conjugated $\forall f \in L^2_{loc}(\mathbb{R})$ by

$$(T_{K_{\theta}}f)(t) = \int_{\mathbb{R}} K_{\theta}(s,t)f(s) ds$$
$$= (U_{\theta} T_{K} U_{\theta}^{-1} f)(t)$$

$$(70)$$

Proof. For any $g \in L^2_{loc}(\mathbb{R})$, compute:

$$((U_{\theta}T_{K}U_{\theta}^{-1})g)(t) = (U_{\theta}(T_{K}U_{\theta}^{-1}g))(t)$$

$$= \sqrt{\dot{\theta}(t)} (T_{K}U_{\theta}^{-1}g)(\theta(t))$$

$$= \sqrt{\dot{\theta}(t)} \int_{\mathbb{R}} K(|\theta(t) - \theta(s)|) (U_{\theta}^{-1}g)(\theta(s))\dot{\theta}(s)ds$$

$$= \sqrt{\dot{\theta}(t)} \int_{\mathbb{R}} K(|\theta(t) - \theta(s)|) \frac{g(s)}{\sqrt{\dot{\theta}(s)}} \dot{\theta}(s) ds$$

$$= \sqrt{\dot{\theta}(t)} \int_{\mathbb{R}} K(|\theta(t) - \theta(s)|) g(s) \sqrt{\dot{\theta}(s)} ds$$

$$= \int_{\mathbb{R}} \sqrt{\dot{\theta}(t)\dot{\theta}(s)} K(|\theta(t) - \theta(s)|) g(s) ds$$

$$= \int_{\mathbb{R}} K_{\theta}(t, s) g(s) ds$$

$$= (T_{K_{\theta}}g)(t)$$

$$(71) \Box$$

4 The Ensemble of Sample Path Realizations

Question: is this called local integrability? state this more eloquently

Theorem 15. [Sample paths in $L^2_{loc}(\mathbb{R})$] Let $\{X(t)\}_{t\in\mathbb{R}}$ be a second-order stationary process with

$$\sigma^2 := \mathbb{E}[X(t)^2] < \infty \tag{72}$$

then, almost surely, every sample path $t \mapsto X(\omega, t) \in L^2_{loc}(\mathbb{R})$.

Proof. Fix any bounded interval [a, b] and consider the random variable

$$Y_{[a,b]} := \int_{a}^{b} X(t)^{2} dt \tag{73}$$

By stationarity and Fubini's theorem:

$$\mathbb{E}[Y_{[a,b]}] = \mathbb{E}\left[\int_{a}^{b} X(t)^{2} dt\right] = \int_{a}^{b} \mathbb{E}[X(t)^{2}] dt$$

$$= \int_{a}^{b} \sigma^{2} dt$$

$$= \sigma^{2} (b-a) < \infty$$

$$(74)$$

By Markov's inequality, for any M > 0:

$$P(Y_{[a,b]} > M) \le \frac{\mathbb{E}[Y_{[a,b]}]}{M} = \frac{\sigma^2(b-a)}{M}$$
 (75)

Taking $M \to \infty$, the conclusion is

$$P\left(Y_{[a,b]} < \infty\right) = 1\tag{76}$$

i.e., almost surely the sample path is square-integrable on [a,b]. Since $\mathbb R$ is the countable union of bounded intervals:

$$\mathbb{R} = \bigcup_{n=1}^{\infty} \left[-n, n \right] \tag{77}$$

by countable subadditivity of probability:

$$P\left(\bigcap_{n=1}^{\infty} \left\{ \int_{-n}^{n} X(t)^2 dt < \infty \right\} \right) = 1 \tag{78}$$

Now let K be any compact set. Then K is bounded, so

$$K \subseteq [-N, N] \tag{79}$$

for some N. Therefore:

$$\int_{K} X(t)^{2} dt \le \int_{-N}^{N} X(t)^{2} dt < \infty$$
 (80)

almost surely. This holds for every compact set K, so almost surely every sample path lies in $L^2_{loc}(\mathbb{R})$.

5 Zero Localization

The construction

stationary
$$X \xrightarrow{U_{\theta}}$$
 oscillatory $Z \xrightarrow{\mu = \delta(Z)|\dot{Z}| dt} L^{2}(\mu) \xrightarrow{L:tf(t)} (L, \sigma(L))$ (81)

produces a self-adjoint operator whose eigenvalues equal the zero set of the realization sample path realization Z(t) from the ensemble of possible sample path functions having the given covariance structure—and whose spectrum equals the closure of the zero set, determined by the choice of time-change $\theta(t)$, spectral measure $F(\lambda)$, and complex orthogonal random measure $\Phi(\lambda)$ which uniquely corresponds to a given sample path from the ensemble.

5.1 Zero localization measure

Definition 16. [Zero localization measure] Let Z be real-valued with $Z \in C^1(\mathbb{R})$ having only simple zeros

$$Z(t_0) = 0 \Rightarrow \dot{Z}(t_0) \neq 0 \tag{82}$$

Define, for Borel $B \subset \mathbb{R}$,

$$\mu(B) = \int_{\mathbb{R}} \mathbf{1}_{B}(t) \, \delta(Z(t)) \, |\dot{Z}(t)| \, dt \tag{83}$$

Theorem 17. [Atomicity on the zero set] For every $\phi \in C_c^{\infty}(\mathbb{R})$,

$$\int_{\mathbb{R}} \phi(t) \, \delta(Z(t)) \, |\dot{Z}(t)| \, dt = \sum_{t_0: Z(t_0) = 0} \phi(t_0) \tag{84}$$

hence

$$\mu(t) = \sum_{t_0: Z(t_0) = 0} \delta_{t_0}(t) \tag{85}$$

Proof. Since all zeros of Z are simple and $Z \in C^1(\mathbb{R})$, by the inverse function theorem each zero t_0 is isolated. Near each zero t_0 , Z is locally monotonic, so the one-dimensional change of variables formula for the Dirac delta can be applied. Specifically, near t_0 where $Z(t_0) = 0$ and $\dot{Z}(t_0) \neq 0$, locally

$$Z(t) = (t - t_0) \dot{Z}(t_0) + O((t - t_0)^2)$$
(86)

holds. The distributional identity for the Dirac delta under smooth changes of variables gives:

$$\delta(Z(t)) = \sum_{t_0: Z(t_0) = 0} \frac{\delta(t - t_0)}{|\dot{Z}(t_0)|}$$
(87)

Therefore:

$$\int_{\mathbb{R}} \phi(t) \, \delta(Z(t)) \, |\dot{Z}(t)| \, dt = \int_{-\infty}^{\infty} \phi(t) \, |\dot{Z}(t)| \sum_{t_0: Z(t_0) = 0} \frac{\delta(t - t_0)}{|\dot{Z}(t_0)|} \, dt$$

$$= \sum_{t_0: Z(t_0) = 0} \int_{\mathbb{R}} \phi(t) \frac{|\dot{Z}(t)| \, \delta(t - t_0)}{|\dot{Z}(t_0)|} \, dt$$

$$= \sum_{t_0: Z(t_0) = 0} \frac{|\dot{Z}(t_0)|}{|\dot{Z}(t_0)|} \, \phi(t_0)$$

$$= \sum_{t_0: Z(t_0) = 0} \phi(t_0)$$
(88)

This shows that μ is the discrete measure

$$\mu(t) = \sum_{t_0: Z(t_0) = 0} \delta_{t_0}(t) \tag{89}$$

assigning unit mass to each zero.

5.2 Hilbert space on zeros and multiplication operator

Definition 18. [Hilbert space on the zero set] Let $\mathcal{H} = L^2(\mu)$ with inner product

$$\langle f, g \rangle = \int_{-\infty}^{\infty} f(t) \overline{g(t)} \, \mathrm{d}\mu (t)$$
 (90)

Proposition 19. [Atomic structure] Let

$$\mu = \sum_{t_0: Z(t_0) = 0} \delta_{t_0} \tag{91}$$

then

$$\mathcal{H} \cong \left\{ f : \{ t_0 : Z(t_0) = 0 \} \to \mathbb{C} : \sum_{t_0 : Z(t_0) = 0} |f(t_0)|^2 < \infty \right\} \cong \ell^2$$
(92)

with orthonormal basis $\{e_{t_0}\}_{t_0:Z(t_0)=0}$ where

$$e_{t_0}(t_1) = \delta_{t_0}(t_1) \tag{93}$$

Proof. By the atomic form of μ , for any $f \in L^2(\mu)$:

$$||f||_{\mathcal{H}}^2 = \int |f(t)|^2 d\mu(t)$$
 (94)

$$= \int |f(t)|^2 \sum_{t_0: Z(t_0)=0} \delta_{t_0}(t)$$
 (95)

$$= \sum_{t_0: Z(t_0)=0} |f(t_0)|^2 \tag{96}$$

This shows the isomorphism with ℓ^2 where the functions e_{t_0} defined by

$$e_{t_0}(t_1) = \delta_{t_0}(t_1) \tag{97}$$

satisfy the relations

$$\langle e_{t_0}, e_{t_1} \rangle = \int e_{t_0}(t) \overline{e_{t_1}(t)} d\mu(t)$$

$$= \sum_{t: Z(t) = 0} \delta_{t_0}(t) \delta_{t_1}(t)$$

$$= \delta_{t_0}(t_1)$$

$$= \delta t_1(t_0)$$

$$(98)$$

thus forming an orthonormal set. Thus, any $f(t) \in \mathcal{H}$ can be written as

$$f(t) = \sum_{t_0: Z(t_0) = 0} f(t_0) e_{t_0}(t)$$
(99)

proving they form a basis.

Definition 20. [Multiplication operator] Define the linear operator

$$L: \mathcal{D}(L) \subset \mathcal{H} \to \mathcal{H} \tag{100}$$

by

$$(Lf)(t) = tf(t) \tag{101}$$

on the support of μ with domain

$$\mathcal{D}(L) := \left\{ f \in \mathcal{H}: \int |t f(t)|^2 d\mu(t) < \infty \right\}$$
(102)

Theorem 21. [Self-adjointness and spectrum] L is self-adjoint on \mathcal{H} and has pure point, simple spectrum

$$\sigma(L) = \overline{\{t \in \mathbb{R}: Z(t) = 0\}} \tag{103}$$

with eigenvalues $\lambda = t_0$ for each zero t_0 and corresponding eigenvectors e_{t_0} .

Proof. First, self-adjointness is verified. For $f, g \in \mathcal{D}(L)$:

$$\langle Lf, g \rangle = \int (Lf)(t)\overline{g(t)}d\mu(t)$$

$$= \int tf(t)\overline{g(t)}d\mu(t)$$

$$= \int f(t)\overline{t}\overline{g(t)}d\mu(t)$$

$$= \int f(t)\overline{(Lg)(t)}d\mu(t)$$

$$= \langle f, Lg \rangle$$
(104)

Thus L is symmetric and acts as

$$(Lf)(t_0) = t_0 f(t_0) \tag{105}$$

for each t_0 in the atomic representation where

$$Z(t_0) = 0 (106)$$

This is unitarily equivalent to the diagonal operator on ℓ^2 with diagonal entries

$$\{t_0: Z(t_0) = 0\} \tag{107}$$

Such diagonal operators are self-adjoint. For the spectrum calculation:

$$L e_{t_0} = t_0 e_{t_0} \forall \{t_0: Z(t_0) = 0\}$$
(108)

holds, so each t_0 is an eigenvalue of L with eigenvector e_{t_0} and since $\{e_{t_0}\}$ forms an orthonormal basis, L has pure point spectrum. The spectrum of a diagonal operator equals the closure of the set of diagonal entries, hence

$$\sigma(L) = \overline{\{t_0: Z(t_0) = 0\}} \tag{109}$$

The eigenvalues are simple.

5.3 Regularity and Simplicity of Sample Path Zero Crossings

TODO: insert the fundamental theorem on the non-tangency of zero crossings so that it doesnt have to be assumed but is in fact a fundamental theorem of non-degenerate Gaussian processes

Definition 22. [Regularity and simplicity] Assume $Z \in C^1(\mathbb{R})$ and every zero is simple:

$$Z(t_0) = 0 \Rightarrow \dot{Z}(t_0) \neq 0 \tag{110}$$

Lemma 23. [Local finiteness and delta decomposition] Under Definition 22, zeros are locally finite and

$$\delta(Z(t)) = \sum_{t_0: Z(t_0) = 0} \frac{\delta(t - t_0)}{|\dot{Z}(t_0)|}$$
(111)

whence

$$\mu = \sum_{t_0: Z(t_0) = 0} \delta_{t_0} \tag{112}$$

Proof. Since $Z \in C^1(\mathbb{R})$ and $\dot{Z}(t_0) \neq 0$ at each zero t_0 , the inverse function theorem implies that Z is locally invertible near each zero. Specifically, there exists a neighborhood U_{t_0} of t_0 such that $Z|_{U_{t_0}}$ is strictly monotonic and invertible.

This implies zeros are isolated: if $Z(t_0) = 0$ and $\dot{Z}(t_0) \neq 0$, then there exists $\epsilon > 0$ such that $Z(t) \neq 0$ for $0 < |t - t_0| < \epsilon$. Therefore zeros are locally finite (finitely many in any bounded interval).

For the distributional identity, the one-dimensional change of variables formula for the Dirac delta is considered. If $g: I \to \mathbb{R}$ is C^1 on interval I with $\dot{g}(x) \neq 0$ for all $x \in I$, then

$$\delta(g(x)) = \sum_{x_0: g(x_0) = 0} \frac{\delta(x - x_0)}{|\dot{g}(x_0)|}$$
(113)

Applying this locally around each zero t_0 of Z, and since zeros are isolated, the local results can be patched together to obtain the global identity:

$$\delta(Z(t)) = \sum_{t_0: Z(t_0) = 0} \frac{\delta(t - t_0)}{|\dot{Z}(t_0)|}$$
(114)

Consequently:

$$d\mu(t) = \delta(Z(t))|\dot{Z}(t)| dt$$

$$= \sum_{t_0: Z(t_0)=0} \frac{|\dot{Z}(t)|}{|\dot{Z}(t_0)|} \delta(t - t_0) dt$$

$$= \sum_{t_0: Z(t_0)=0} \delta_{t_0}(dt)$$
(115)

where the last equality uses the fact that

$$\frac{|\dot{Z}(t_0)|}{|\dot{Z}(t_0)|} = 1\tag{116}$$

when evaluating at $t = t_0$.

5.4 The Kac-Rice Formula For The Expected Zero Counting Function

Theorem 24. (Kac-Rice Formula for Zero Crossings) Let Z(t) be a centered Gaussian process on [a,b] with covariance $K(s,t) = \mathbb{E}[Z(s)|Z(t)]$ then the expected number of zeros in [a,b] is

$$\mathbb{E}[N_{[a,b]}] = \int_{a}^{b} \sqrt{\frac{2}{\pi}} \frac{\sqrt{K(t,t) K_{\dot{Z}}(t,t) - K_{Z,\dot{Z}}(t,t)^{2}}}{K(t,t)} dt$$
(117)

where

$$K(t,t) = \mathbb{E}[Z(t)^2] \tag{118}$$

$$K_{\dot{Z}}(t,t) = -\partial_s^2 \partial_t K(s,t)|_{s=t}$$
(119)

and

$$K_{Z,\dot{Z}}(t,t) = \partial_s K(s,t)|_{s=t} \tag{120}$$

Proof.

The exact zero counting function is

$$N_{[a,b]} = \int_{a}^{b} \delta(Z(t)) |\dot{Z}(t)| \ dt \tag{121}$$

so

$$\mathbb{E}[N_{[a,b]}] = \int_{a}^{b} \mathbb{E}[\delta(Z(t))|\dot{Z}(t)|] dt$$

$$= \int_{a}^{b} \int_{-\infty}^{\infty} |v| \ p_{Z,\dot{Z}}(0,v) \ dv \ dt$$
(122)

The vector $(Z(t), \dot{Z}(t))$ is bivariate Gaussian with covariance matrix

$$\Sigma = \begin{pmatrix} K(t,t) & K_{Z,\dot{Z}}(t,t) \\ K_{Z,\dot{Z}}(t,t) & K_{\dot{Z}}(t,t) \end{pmatrix}$$

$$\tag{123}$$

whose determinant is given by

$$\det \Sigma = K(t,t) K_{\dot{Z}}(t,t) - K_{Z,\dot{Z}}(t,t)^2$$
(124)

the inverse of which satisfies

$$\Sigma_{22}^{-1} = \frac{K(t,t)}{\det \Sigma} \tag{125}$$

yielding

$$p_{Z,\dot{Z}}(0,v) = \frac{1}{\sqrt{2\pi K(t,t)}} \cdot \frac{e^{-\frac{K(t,t)v^2}{2\det\Sigma}}}{\sqrt{2\pi \det\Sigma/K(t,t)}}$$
(126)

which factorizes as $p_Z(0) \cdot p_{\dot{Z}|Z}(v|0)$ where

$$p_Z(0) = \frac{1}{\sqrt{2\pi K(t,t)}} \tag{127}$$

and

$$\dot{Z}|Z = 0 \sim \mathcal{N}(0, \det \Sigma / K(t, t)) \tag{128}$$

For zero-mean Gaussian $Y \sim \mathcal{N}(0, \sigma^2)$, direct integration gives

$$\mathbb{E}[|Y|] = 2\int_0^\infty \frac{y}{\sqrt{2\pi\sigma^2}} e^{-y^2/(2\sigma^2)} dy$$

$$= \frac{2\sigma}{\sqrt{2\pi}} \int_0^\infty e^{-u} du$$

$$= \sqrt{\frac{2}{\pi}} \sigma$$
(129)

so that combining results yields

$$\int_{-\infty}^{\infty} |v| \ p_{Z,\dot{Z}}(0,v) \ dv = \frac{\sqrt{\frac{2}{\pi}} \sqrt{\frac{\det \Sigma}{K(t,t)}}}{\sqrt{2\pi K(t,t)}}$$

$$= \sqrt{\frac{2}{\pi}} \frac{\sqrt{\det \Sigma}}{K(t,t)}$$

$$= \sqrt{\frac{2}{\pi}} \frac{\sqrt{\det \Sigma}}{K(t,t)}$$
(130)

Theorem 25. [Expected Zero-Counting Function] Let $\theta \in \mathcal{F}$ and let

$$K(t,s) = \operatorname{cov}(Z(t), Z(s)) \tag{131}$$

be twice differentiable at s=0 and t=0 then expected number of zeros of the process Z(t) in [a,b] is

$$\mathbb{E}[N_{[a,b]}] = \sqrt{-K(0)} \left(\theta(b) - \theta(a)\right) \tag{132}$$

Proof. The covariance function of the time-changed process is

$$K_{\theta}(s,t) = \operatorname{cov}(Z(t), Z(s))$$

$$= \sqrt{\dot{\theta}(s)\dot{\theta}(t)}K(|\theta(t) - \theta(s)|)$$
(133)

For the zero-crossing analysis, consider the normalized process. By the Kac-Rice formula:

$$\mathbb{E}[N_{[a,b]}] = \int_{a}^{b} \sqrt{-\lim_{s \to t} \frac{\partial^{2}}{\partial s \, \partial t} \, K_{\theta}(s,t)} \, dt \tag{134}$$

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Computing the mixed partial derivative:

$$\frac{\partial}{\partial t} K_{\theta}(s,t) = \frac{1}{2} \frac{\dot{\theta}(t)}{\sqrt{\theta(t)}} \sqrt{\dot{\theta}(s)} K(|\theta(t) - \theta(s)|) + \sqrt{\theta(s)} \sqrt{\theta(t)} \dot{K}(|\theta(t) - \theta(s)|) \operatorname{sgn}(\theta(t) - \theta(s)) \dot{\theta}(t)$$
(135)

Taking the limit as $s \to t$ and using the fact that $\dot{K}(0) = 0$ for stationary processes:

$$\lim_{s \to t} \frac{\partial^2}{\partial s \, \partial t} K_{\theta}(s, t) = \lim_{\substack{s \to t \\ = \dot{\theta}(t)^2 \ddot{K}(0)}} \dot{\theta}(t) \, \ddot{K}(0)$$

$$(136)$$

Substituting into the Kac-Rice formula we have

$$\mathbb{E}[N_{[a,b]}] = \int_{a}^{b} \sqrt{-\dot{\theta}(t)^{2} \, \ddot{K}(0)} \, dt$$

$$= \sqrt{-\ddot{K}(0)} \int_{a}^{b} \dot{\theta}(t) \, dt$$

$$= \sqrt{-\ddot{K}(0)} \, (\theta(b) - \theta(a))$$
(137)

since $\dot{\theta}(t) \ge 0$ almost everywhere.

Bibliography

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