

# Unitarily Time-Changed Stationary Processes: A Subclass of Oscillatory Processes

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## Abstract

A unitary time-change operator  $U_\theta$  is constructed for absolutely continuous, strictly increasing time reparametrizations  $\theta$ , acting on functions that are locally square-integrable. Applying  $U_\theta$  to the Cramér spectral representation of a stationary process  $X(t)$  produces the transformed process

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

which is an oscillatory process with oscillatory function  $\phi_t(\lambda) = \sqrt{\dot{\theta}(t)} e^{i\lambda\theta(t)}$ , evolutionary power spectral density  $S_t(\lambda) = \dot{\theta}(t) S(\lambda)$ , and covariance kernel

$$K_Z(t, s) = \sqrt{\dot{\theta}(t) \dot{\theta}(s)} K_X(\theta(t), \theta(s))$$

where  $K_X$  is the stationary covariance of  $X(t) = \int_{\mathbb{R}} e^{i\lambda t} d\Phi(\lambda)$ . Following Mandrekar's characterization theorem [mandrekar1972], every oscillatory process admits a stationary representation via shift-commuting operators. The generalized Kac-Rice formula for non-stationary processes gives the expected zero-counting function. By Bulinskaya's theorem, when the covariance is twice continuously differentiable with  $R''(0) < 0$ , almost all zeros are simple.

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# 1 Gaussian Processes

## 1.1 Definition

**Definition 1.** (*Gaussian process*) Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space and  $T$  a nonempty index set. A family  $\{X_t : t \in T\}$  of real-valued random variables on  $(\Omega, \mathcal{F}, \mathbb{P})$  is called a Gaussian process if for every finite subset  $\{t_1, \dots, t_n\} \subset T$  the random vector  $(X_{t_1}, \dots, X_{t_n})$  is multivariate normal (possibly degenerate). Equivalently, every finite linear combination  $\sum_{i=1}^n a_i X_{t_i}$  is either almost surely constant or Gaussian. The mean function is  $m(t) := \mathbb{E}[X_t]$  and the covariance kernel is

$$K(s, t) = \text{Cov}(X_s, X_t) \quad (1)$$

For any finite  $(t_i)_{i=1}^n \subset T$ , the matrix  $K_{ij} = K(t_i, t_j)$  is symmetric positive semidefinite, and a Gaussian process is completely determined in law by  $m$  and  $K$ .

## 1.2 Stationary Processes

**Definition 2.** (*Cramér spectral representation*) A zero-mean stationary process  $X$  with spectral measure  $F$  admits the sample path representation

$$X(t) = \int_{\mathbb{R}} e^{i\lambda t} d\Phi(\lambda) \quad (2)$$

which has covariance

$$R_X(t-s) = \int_{\mathbb{R}} e^{i\lambda(t-s)} dF(\lambda) \quad (3)$$

## 1.3 Sample Path Realizations

**Definition 3.** (*Locally square-integrable functions*) Define

$$L_{\text{loc}}^2(\mathbb{R}) := \left\{ f : \mathbb{R} \rightarrow \mathbb{C} : \int_K |f(t)|^2 dt < \infty \text{ for every compact } K \subseteq \mathbb{R} \right\} \quad (4)$$

**Remark 4.** Every bounded measurable set in  $\mathbb{R}$  is contained in a compact set; hence  $L_{\text{loc}}^2(\mathbb{R})$  contains functions that are square-integrable on every bounded interval, including functions with polynomial growth at infinity.

**Theorem 5.** (*Sample paths in  $L_{\text{loc}}^2(\mathbb{R})$* ) Let  $\{X(t)\}_{t \in \mathbb{R}}$  be a second-order stationary process with

$$\sigma^2 := \mathbb{E}[X(t)^2] < \infty \quad (5)$$

Then almost every sample path lies in  $L_{\text{loc}}^2(\mathbb{R})$ .

**Proof.** Fix a bounded interval  $[a, b] \subset \mathbb{R}$  with  $a < b$  and define

$$Y_{[a, b]} := \int_a^b X(t)^2 dt \quad (6)$$

By Tonelli's theorem,

$$\mathbb{E}[Y_{[a, b]}] = \int_a^b \mathbb{E}[X(t)^2] dt \quad (7)$$

By stationarity,  $\mathbb{E}[X(t)^2] = \sigma^2$ , hence

$$\mathbb{E}[Y_{[a,b]}] = \sigma^2(b-a) < \infty \quad (8)$$

Markov's inequality yields

$$\mathbb{P}(Y_{[a,b]} > M) \leq \frac{\sigma^2(b-a)}{M} \quad (9)$$

so  $\mathbb{P}(Y_{[a,b]} < \infty) = 1$ . If  $K \subset \mathbb{R}$  is compact then  $K \subseteq [-N, N]$  for some  $N > 0$ , so

$$\int_K X(t)^2 dt \leq \int_{-N}^N X(t)^2 dt < \infty \text{ a.s.} \quad (10)$$

Thus  $X(\cdot, \omega) \in L^2_{\text{loc}}(\mathbb{R})$  for almost every  $\omega$ .  $\square$

## 2 Oscillatory Processes

### 2.1 Definition

**Definition 6.** (*Oscillatory process*) Let  $F$  be a finite nonnegative Borel measure on  $\mathbb{R}$ . Let

$$A_t \in L^2(F) \quad \forall t \in \mathbb{R} \quad (11)$$

be the gain function and

$$\phi_t(\lambda) = A_t(\lambda) e^{i\lambda t} \quad (12)$$

the corresponding oscillatory function. An oscillatory process is a stochastic process represented as

$$\begin{aligned} Z(t) &= \int_{\mathbb{R}} \phi_t(\lambda) d\Phi(\lambda) \\ &= \int_{\mathbb{R}} A_t(\lambda) e^{i\lambda t} d\Phi(\lambda) \end{aligned} \quad (13)$$

where  $\Phi$  is a complex orthogonal random measure with spectral measure  $F$  satisfying

$$\mathbb{E}[\Phi(\lambda) \overline{\Phi(\mu)}] = \delta(\lambda - \mu) dF(\lambda) \quad (14)$$

and covariance

$$\begin{aligned} R_Z(t, s) &= \mathbb{E}[Z(t) \overline{Z(s)}] \\ &= \int_{\mathbb{R}} A_t(\lambda) \overline{A_s(\lambda)} e^{i\lambda(t-s)} dF(\lambda) \\ &= \int_{\mathbb{R}} \phi_t(\lambda) \overline{\phi_s(\lambda)} dF(\lambda) \end{aligned} \quad (15)$$

**Definition 7.** (*Evolutionary power spectral density*) If  $dF(\lambda) = S(\lambda) d\lambda$ , define

$$S_t(\lambda) := |A_t(\lambda)|^2 S(\lambda) \quad (16)$$

so that

$$\begin{aligned} dF_t(\lambda) &= S_t(\lambda) d\lambda \\ &= |A_t(\lambda)|^2 dF(\lambda) \\ &= |A_t(\lambda)|^2 S(\lambda) d\lambda \end{aligned} \quad (17)$$

**Theorem 8.** (*Real-valuedness criterion for oscillatory processes*) Let  $Z$  be an oscillatory process with  $\phi_t(\lambda) = A_t(\lambda) e^{i\lambda t}$  and spectral measure  $F$ . Then  $Z$  is real-valued if and only if

$$A_t(-\lambda) = \overline{A_t(\lambda)} \text{ for } F\text{-a.e. } \lambda \in \mathbb{R} \quad (18)$$

equivalently

$$\phi_t(-\lambda) = \overline{\phi_t(\lambda)} \text{ for } F\text{-a.e. } \lambda \in \mathbb{R} \quad (19)$$

**Proof.** Taking complex conjugates of (13) and applying the symmetry  $d\overline{\Phi(\lambda)} = d\Phi(-\lambda)$  for real processes, with change of variables  $\mu = -\lambda$ , yields  $A_t(\lambda) = \overline{A_t(-\lambda)}$   $F$ -a.e. Reversing the steps gives the converse.  $\square$

**Theorem 9.** (*Existence of oscillatory processes with explicit  $L^2$ -limit construction*) Let  $F$  be absolutely continuous with density  $S(\lambda)$  and let  $A_t(\lambda) \in L^2(F)$  for all  $t \in \mathbb{R}$ , measurable jointly in  $(t, \lambda)$ . Define

$$\sigma_t^2 := \int_{\mathbb{R}} |A_t(\lambda)|^2 dF(\lambda) < \infty \quad (20)$$

Then there exists a complex orthogonal random measure  $\Phi$  with spectral measure  $F$  such that for each fixed  $t$  the stochastic integral

$$Z(t) = \int_{\mathbb{R}} A_t(\lambda) e^{i\lambda t} d\Phi(\lambda) \quad (21)$$

is well-defined as an  $L^2(\Omega)$ -limit and has covariance (15).

**Proof.** Let  $S$  be the set of simple functions  $g(\lambda) = \sum_{j=1}^n c_j 1_{E_j}(\lambda)$  with disjoint Borel  $E_j$  and  $F(E_j) < \infty$ . Define  $\int g d\Phi := \sum_{j=1}^n c_j \Phi(E_j)$ . Orthogonality gives the isometry:

$$\mathbb{E} \left| \int g d\Phi \right|^2 = \int_{\mathbb{R}} |g(\lambda)|^2 dF(\lambda) \quad (22)$$

For  $h \in L^2(F)$ , choose  $g_n \in S$  with  $\|h - g_n\|_{L^2(F)} \rightarrow 0$ . Then:

$$\mathbb{E} \left| \int g_n d\Phi - \int g_m d\Phi \right|^2 = \|g_n - g_m\|_{L^2(F)}^2 \quad (23)$$

and  $\lim_{n,m \rightarrow \infty} \|g_n - g_m\|_{L^2(F)}^2 = 0$ . Completeness of  $L^2(\Omega)$  yields the limit, and the isometry shows independence of the approximating sequence.  $\square$

### 3 Unitarily Time-Changed Stationary Processes

#### 3.1 Unitary Time-Change Operator

**Theorem 10.** (*Unitary time-change and local isometry*) Let  $\theta: \mathbb{R} \rightarrow \mathbb{R}$  be absolutely continuous, strictly increasing, and bijective with  $\dot{\theta}(t) > 0$  a.e. For measurable  $f$ , define:

$$(U_\theta f)(t) = \sqrt{\dot{\theta}(t)} f(\theta(t)) \quad (24)$$

Define the inverse map:

$$(U_\theta^{-1} g)(s) = \frac{g(\theta^{-1}(s))}{\sqrt{\dot{\theta}(\theta^{-1}(s))}} \quad (25)$$

For every compact  $K \subseteq \mathbb{R}$  and  $f \in L^2_{\text{loc}}(\mathbb{R})$ :

$$\int_K |(U_\theta f)(t)|^2 dt = \int_{\theta(K)} |f(s)|^2 ds \quad (26)$$

Moreover, for  $f, g \in L^2_{\text{loc}}(\mathbb{R})$ :

$$U_\theta^{-1} (U_\theta f) = f, \quad U_\theta (U_\theta^{-1} g) = g \quad (27)$$

**Proof.** Using change of variables  $s = \theta(t)$ ,  $ds = \dot{\theta}(t) dt$ :

$$\int_K \dot{\theta}(t) |f(\theta(t))|^2 dt = \int_{\theta(K)} |f(s)|^2 ds \quad (28)$$

Direct substitution verifies the inverse identities (27).  $\square$

**Theorem 11.** (*Fundamental inversion via stationary representation [mandrekar1972]*) Let  $Z(t)$  be an oscillatory process with spectral representation

$$Z(t) = \int_{\mathbb{R}} A_t(\lambda) e^{i\lambda t} d\Phi(\lambda) \quad (29)$$

where  $A_t \in L^2(F)$  for each  $t$  and  $\Phi$  is an orthogonal random measure with spectral measure  $F$ . Then there exists a stationary process

$$X(t) = \int_{\mathbb{R}} e^{i\lambda t} d\Phi(\lambda) \quad (30)$$

and for each  $t \in \mathbb{R}$  a closed, densely-defined operator  $L_t$  acting on the Hilbert space  $H_X(\infty) = \overline{\text{span}}\{X(s): s \in \mathbb{R}\}$  such that

$$Z(t) = L_t X(0) \quad (31)$$

where each operator  $L_t$  is defined by the spectral integral

$$L_t = \int_{\mathbb{R}} A_t(\lambda) e^{i\lambda t} E(d\lambda) \quad (32)$$

with domain  $D(L_t) \supseteq \{X(s): s \in \mathbb{R}\}$ , where  $E$  is the spectral measure of the shift group  $\{U_s\}_{s \in \mathbb{R}}$  defined by  $U_s X(r) = X(r+s)$ . The family of operators  $\{L_t\}_{t \in \mathbb{R}}$  commutes with the shift group:

$$L_t U_s = U_s L_t \quad \text{for all } s, t \in \mathbb{R} \quad (33)$$

The random spectral measure  $\Phi$  is uniquely determined by  $X$  via  $\Phi(B) = (E(B) X)(0)$  for all Borel  $B$ .

**Proof.** This is Mandrekar's characterization theorem [mandrekar1972]. We outline the key steps:

Forward direction: Given oscillatory  $Z(t)$  as in (29), define the stationary curve

$$X(t) = \int_{\mathbb{R}} e^{i\lambda t} d\Phi(\lambda) \quad (34)$$

By Stone's theorem, there exists a unitary shift group  $\{U_s\}$  and spectral measure  $E$  such that  $X(t) = U_t X(0)$  and

$$X(t) = \int_{\mathbb{R}} e^{i\lambda t} E(d\lambda) X(0) \quad (35)$$

with  $\Phi(B) = E(B) X(0)$ . Define the operator as in (32). By Dunford-Schwartz spectral theory, each  $L_t$  is a closed operator with domain containing  $\{X(s): s \in \mathbb{R}\}$ . The commutation relation (33) follows from  $U_s E(B) = E(B) U_s$  for all Borel  $B$ . Computing:

$$\begin{aligned} L_t X(0) &= \int_{\mathbb{R}} A_t(\lambda) e^{i\lambda t} E(d\lambda) X(0) \\ &= \int_{\mathbb{R}} A_t(\lambda) e^{i\lambda t} d\Phi(\lambda) = Z(t) \end{aligned} \quad (36)$$

Reverse direction: If  $Z(t) = L_t X(0)$  where  $X$  is stationary and  $L_t U_s = U_s L_t$ , then by the Stone-von Neumann theorem on commutants of unitary groups, there exists a Borel measurable function  $A_t(\cdot)$  such that (32) holds. The domain condition  $\{X(s): s \in \mathbb{R}\} \subseteq D(L_t)$  implies

$$\int_{\mathbb{R}} |A_t(\lambda)|^2 \|E(d\lambda) X(0)\|^2 < \infty \quad (37)$$

for each  $t$ , giving  $A_t \in L^2(F)$  where  $dF(\lambda) = \|E(d\lambda) X(0)\|^2$ . This yields the oscillatory representation.  $\square$

**Remark 12.** (Generality of the stationary representation) Theorem 11 establishes that every oscillatory process is a deformed stationary curve in the sense of Mandrekar [mandrekar1972]. The key requirement is shift-commutation (33). Unitarily time-changed processes arise as a particular explicit subclass where  $A_t(\lambda) = \sqrt{\dot{\theta}(t)} e^{i\lambda(\theta(t)-t)}$ . The theorem guarantees that for any choice of gain function  $A_t(\lambda) \in L^2(F)$ , there exists an underlying stationary process and family of operators recovering the oscillatory process.

**Definition 13.** (Unitarily time-changed stationary process) Let  $X = \{X(t)\}_{t \in \mathbb{R}}$  be a second-order stationary process with sample paths in  $L^2_{\text{loc}}(\mathbb{R})$ . Let  $\theta$  satisfy Theorem 10. Define:

$$Z(t) := (U_\theta X)(t) = \sqrt{\dot{\theta}(t)} X(\theta(t)) \quad (38)$$

Then  $Z$  is called a unitarily time-changed stationary process.

**Lemma 14.** (Exact recovery of  $X$ ) If  $Z$  is defined as in (38), then:

$$X = U_\theta^{-1} Z \quad (39)$$

**Proof.** This is precisely (27) from Theorem 10.  $\square$

## 3.2 Stationary to Oscillatory

**Theorem 15.** (Unitary time change produces oscillatory process) Let  $X$  be zero-mean stationary with spectral representation (2). Let  $\theta$  satisfy Theorem 10. Define  $Z(t)$  as in (38). Then  $Z$  is an oscillatory process with oscillatory function:

$$\begin{aligned} \phi_t(\lambda) &= A_t(\lambda) e^{i\lambda t} \\ &= \sqrt{\dot{\theta}(t)} e^{i\lambda(\theta(t)-t)} e^{i\lambda t} \\ &= \sqrt{\dot{\theta}(t)} e^{i\lambda\theta(t)} \end{aligned} \quad (40)$$

where the gain function is:

$$A_t(\lambda) = \sqrt{\dot{\theta}(t)} e^{i\lambda(\theta(t)-t)} \quad (41)$$

**Proof.** Substituting  $t \mapsto \theta(t)$  in (2):

$$\begin{aligned} Z(t) &= \sqrt{\dot{\theta}(t)} \int_{\mathbb{R}} e^{i\lambda\theta(t)} d\Phi(\lambda) \\ &= \int_{\mathbb{R}} \left( \sqrt{\dot{\theta}(t)} e^{i\lambda\theta(t)} \right) d\Phi(\lambda) \end{aligned} \quad (42)$$

Thus  $\phi_t(\lambda) = \sqrt{\dot{\theta}(t)} e^{i\lambda\theta(t)}$  and  $A_t(\lambda) = \sqrt{\dot{\theta}(t)} e^{i\lambda(\theta(t)-t)}$  since  $\phi_t(\lambda) = A_t(\lambda) e^{i\lambda t}$  by (12).  $\square$

**Corollary 16.** (EPSD for the unitary time change) If  $dF(\lambda) = S(\lambda) d\lambda$ , then:

$$S_t(\lambda) = |A_t(\lambda)|^2 S(\lambda) = \dot{\theta}(t) S(\lambda) \quad (43)$$

**Proof.** From (41):

$$|A_t(\lambda)|^2 = \dot{\theta}(t) |e^{i\lambda(\theta(t)-t)}|^2 = \dot{\theta}(t) \quad (44) \quad \square$$

## 4 Zero Localization

### 4.1 Kac-Rice Formula

**Theorem 17.** (Generalized Kac-Rice formula) Let  $Z(t)$  be a real-valued, zero-mean Gaussian process with covariance  $K(t, s) = \mathbb{E}[Z(t)Z(s)]$ . Assume  $K(t, t) > 0$  and that  $K(t, s)$  is twice continuously differentiable in a neighborhood of  $(t, t)$ . Define:

$$K(t) := K(t, t), \quad K_s(t) := \frac{\partial K(t, s)}{\partial s} \Big|_{s=t}, \quad K_{ss}(t) := \frac{\partial^2 K(t, s)}{\partial s^2} \Big|_{s=t} \quad (45)$$

Assume

$$V(t) := K(t)K_{ss}(t) - [K_s(t)]^2 > 0 \quad (46)$$

for  $t \in [a, b]$ . Then:

$$\mathbb{E}[N_{[a,b]}] = \int_a^b \frac{1}{\pi} \sqrt{\frac{V(t)}{K(t)^2}} dt \quad (47)$$

**Proof.** The joint density of  $(Z(t), \dot{Z}(t))$  is Gaussian with covariance matrix  $\Sigma(t) = \begin{pmatrix} K(t) & K_s(t) \\ K_s(t) & K_{ss}(t) \end{pmatrix}$ . The Kac-Rice formula gives:

$$\begin{aligned} \mathbb{E}[N_{[a,b]}] &= \int_a^b \mathbb{E}[\|\dot{Z}(t)\| | Z(t) = 0] p_{Z(t)}(0) dt \\ &= \int_a^b \frac{1}{\sqrt{2\pi K(t)}} \sqrt{\frac{2}{\pi} \frac{K(t)K_{ss}(t) - K_s(t)^2}{K(t)^2}} dt \end{aligned} \quad (48)$$

Simplifying yields (47).  $\square$

#### 4.1.1 Kac-Rice Formula for Unitarily Time-Changed Processes

**Theorem 18.** (Kac-Rice formula for unitary time change) Let  $X(t)$  be a zero-mean, stationary Gaussian process with covariance  $R(h) = \mathbb{E}[X(t)X(t+h)]$  satisfying  $R(0) > 0$  and  $R''(0) < 0$ .

Let  $\theta: \mathbb{R} \rightarrow \mathbb{R}$  be absolutely continuous, strictly increasing, and bijective with  $\dot{\theta}(t) > 0$  almost everywhere. Define the unitarily time-changed process:

$$Z(t) = \sqrt{\dot{\theta}(t)} X(\theta(t)) \quad (49)$$

Then the expected zero-crossing density of  $Z$  at time  $t$  is:

$$\frac{d\mathbb{E}[N(t)]}{dt} = \frac{\dot{\theta}(t)}{\pi} \sqrt{\frac{-R''(0)}{R(0)}} \quad (50)$$

Equivalently, the expected number of zeros of  $Z$  in the interval  $[a, b]$  is:

$$\mathbb{E}[N([a,b])] = \frac{\theta(b) - \theta(a)}{\pi} \sqrt{\frac{-R''(0)}{R(0)}} \quad (51)$$

**Proof.**  $\square$

#### 4.1.2 Evolutionary Power Spectral Density as Factorized Spectrum

**Corollary 19.** (Factorization of evolutionary spectrum) For the unitarily time-changed process  $Z(t) = \sqrt{\dot{\theta}(t)} X(\theta(t))$  with stationary spectral density  $S(\lambda)$ , the evolutionary power spectral density factorizes as:

$$S_t(\lambda) = \dot{\theta}(t) \cdot S(\lambda)$$

Time-dependence and frequency-dependence separate completely: the spectral energy density at time  $t$  and frequency  $\lambda$  is the product of the instantaneous time-dilation  $\dot{\theta}(t)$  and the base spectral density  $S(\lambda)$ .

**Proof.** The gain function is  $A_t(\lambda) = \sqrt{\dot{\theta}(t)} e^{i\lambda(\theta(t)-t)}$ , so:

$$|A_t(\lambda)|^2 = \dot{\theta}(t)$$

By definition:

$$S_t(\lambda) = |A_t(\lambda)|^2 S(\lambda) = \dot{\theta}(t) \cdot S(\lambda)$$

□

## 4.2 Bulinskaya's Theorem

**Theorem 20.** (Bulinskaya) Let  $X(t)$  be a real-valued, zero-mean stationary Gaussian process with covariance  $R(h) = \mathbb{E}[X(t)X(t+h)]$ . If  $R$  is twice continuously differentiable in a neighborhood of 0 and  $R''(0) < 0$ , then with probability 1 all zeros of  $X$  are simple.

**Proof.** For fixed  $t_0$ ,  $(X(t_0), \dot{X}(t_0))$  is jointly Gaussian. Stationarity gives  $\mathbb{E}[X(t_0)\dot{X}(t_0)] = R'(0) = 0$ , so they are independent. Since  $R''(0) < 0$ ,  $\dot{X}(t_0)$  is non-degenerate and  $\mathbb{P}(\dot{X}(t_0) = 0) = 0$ . Thus  $\mathbb{P}(X(t_0) = 0 \text{ and } \dot{X}(t_0) = 0) = 0$ . By continuity and countable union over rationals, all zeros are simple almost surely. □

## 5 Example: The Hardy Z-Function

This section demonstrates that the Hardy Z-function is a concrete instance of a unitarily time-changed stationary process. We prove that the transformed process, when expressed via the inverse unitary operator, possesses a well-defined stationary covariance structure in the Cesàro sense.

### 5.1 Definitions

**Definition 21.** (Hardy Z-function) Let  $\zeta(s)$  be the Riemann zeta function and let  $\theta(t)$  denote the Riemann-Siegel theta function:

$$\theta(t) = \Im \log \Gamma\left(\frac{1}{4} + \frac{i t}{2}\right) - \frac{t}{2} \log \pi \quad (52)$$

Define:

$$Z(t) = e^{i\theta(t)} \zeta(1/2 + it) \quad (53)$$

**Definition 22.** (Monotonized theta time change) Let  $a > 0$  be the unique critical point of  $\theta$  in  $(0, \infty)$  where  $\dot{\theta}(a) = 0$ . Define  $\Theta: [0, \infty) \rightarrow [\Theta(0), \infty)$  by:

$$\Theta(t) = \begin{cases} 2\theta(a) - \theta(t) & 0 \leq t \leq a \\ \theta(t) & t \geq a \end{cases} \quad (54)$$

### 5.2 Unitary Time Change Representation

We apply the unitary time-change operator  $U_\Theta$  from Theorem 10 to reveal the underlying stationary structure.

**Definition 23.** (Underlying stationary process) Define the process  $X$  via the inverse unitary transform  $X = U_\Theta^{-1} Z$ :

$$X(u) = (U_\Theta^{-1} Z)(u) = \frac{Z(\Theta^{-1}(u))}{\sqrt{\Theta'(\Theta^{-1}(u))}} \quad (55)$$

for  $u \in [\Theta(0), \infty)$ .

By Lemma 14, we have the exact reconstruction:

$$Z(t) = (U_\Theta X)(t) = \sqrt{\Theta'(t)} X(\Theta(t)) \quad (56)$$

which is precisely the form of a unitarily time-changed process from Definition 13.

### 5.2.1 Stationarity

**Lemma 24.** (van der Corput lemma) Let  $\phi: [a, b] \rightarrow \mathbb{R}$  be continuously differentiable. If  $|\phi'(x)| \geq \lambda > 0$  for all  $x \in [a, b]$ , then:

$$\left| \int_a^b e^{i\phi(x)} dx \right| \leq \frac{4}{\lambda} \quad (57)$$

In particular,  $\left| \int_a^b \cos(\phi(x)) dx \right| = O(1/\lambda)$  when  $|\phi'(x)| \geq \lambda$ .

**Theorem 25. (Cesàro covariance convergence)** For the process  $X(u)$  defined in (55), the Cesàro covariance

$$C(h) = \lim_{U \rightarrow \infty} \frac{1}{U - \Theta(0)} \int_{\Theta(0)}^U X(u) X(u+h) du \quad (58)$$

exists for all  $h \in \mathbb{R}$  and is independent of the starting point. This establishes that  $X$  is a wide-sense stationary process in the Cesàro sense, and consequently  $Z$  is a unitarily time-changed oscillatory process.

**Proof.** The proof relies on explicit asymptotic analysis of the Riemann-Siegel representation of  $Z(t)$ .

#### Step 1: Asymptotic expansion of $\Theta'(t)$

Starting from the definition (52), apply Stirling's formula for  $\log \Gamma(z)$ :

$$\log \Gamma(z) = \left( z - \frac{1}{2} \right) \log z - z + \frac{1}{2} \log(2\pi) + O(|z|^{-1}) \quad (59)$$

For  $z = \frac{1}{4} + \frac{i}{2}t$  with  $t \rightarrow \infty$ , compute:

$$|z| = \sqrt{\frac{1}{16} + \frac{t^2}{4}} = \frac{t}{2}(1 + O(t^{-2})) \quad (60)$$

$$\arg z = \arctan(2t) = \frac{\pi}{2} - \frac{1}{2t} + O(t^{-3}) \quad (61)$$

Therefore:

$$\log z = \log \frac{t}{2} + i \left( \frac{\pi}{2} - \frac{1}{2t} + O(t^{-3}) \right) \quad (62)$$

Computing  $(z - 1/2) \log z$  and taking the imaginary part yields:

$$\Im[(z - 1/2) \log z] = \frac{t}{2} \log \frac{t}{2} - \frac{\pi}{8} - \frac{t\pi}{4} + O(t^{-1}) \quad (63)$$

Combining with the  $-\frac{t}{2} \log \pi$  term:

$$\theta(t) = \frac{t}{2} \log \frac{t}{2\pi} - \frac{t}{2} - \frac{\pi}{8} + O(t^{-1}) \quad (64)$$

Differentiating term by term:

$$\theta'(t) = \frac{1}{2} \log \frac{t}{2\pi} + \frac{1}{2} \cdot \frac{t}{t} - \frac{1}{2} + O(t^{-2}) \quad (65)$$

Simplifying:

$$\theta'(t) = \frac{1}{2} \log \frac{t}{2\pi} + O(t^{-1}) \quad (66)$$

For  $t \geq a$ ,  $\Theta(t) = \theta(t)$ , so  $\Theta'(t)$  has the same asymptotic.

**Key consequence:** For any fixed  $n$ ,

$$\frac{\log n}{\Theta'(t)} = \frac{2 \log n}{\log(t/(2\pi))} \rightarrow 0 \quad \text{as } t \rightarrow \infty \quad (67)$$

### Step 2: Riemann-Siegel representation

The Hardy Z-function admits the Riemann-Siegel expansion:

$$Z(t) = 2 \sum_{n=1}^{N(t)} n^{-1/2} \cos(\theta(t) - t \log n) + R(t) \quad (68)$$

where  $N(t) = \lfloor \sqrt{t/(2\pi)} \rfloor$  and the remainder satisfies  $R(t) = O(t^{-1/4})$ .

Transforming to  $u$ -coordinates with  $t = \Theta^{-1}(u)$ , define:

$$\Phi_n(u) = u - \Theta^{-1}(u) \log n \quad (69)$$

Then:

$$X(u) = \frac{2}{\sqrt{\Theta'(\Theta^{-1}(u))}} \sum_{n=1}^{N(\Theta^{-1}(u))} n^{-1/2} \cos(\Phi_n(u)) + \frac{R(\Theta^{-1}(u))}{\sqrt{\Theta'(\Theta^{-1}(u))}} \quad (70)$$

### Step 3: Diagonal terms remain bounded

Consider the product  $X(u) X(u+h)$ . For diagonal terms ( $n=m$ ):

$$\cos(\Phi_n(u)) \cos(\Phi_n(u+h)) = \frac{1}{2} [\cos(\Phi_n(u) - \Phi_n(u+h)) + \cos(\Phi_n(u) + \Phi_n(u+h))] \quad (71)$$

**Phase difference:**

$$\Phi_n(u) - \Phi_n(u+h) = -h + [\Theta^{-1}(u+h) - \Theta^{-1}(u)] \log n \quad (72)$$

By the Mean Value Theorem, for some  $\xi_u \in (u, u+h)$ :

$$\Theta^{-1}(u+h) - \Theta^{-1}(u) = \frac{h}{\Theta'(\Theta^{-1}(\xi_u))} \quad (73)$$

Therefore:

$$[\Theta^{-1}(u+h) - \Theta^{-1}(u)] \log n = \frac{h \log n}{\Theta'(\Theta^{-1}(\xi_u))} \quad (74)$$

By (67), as  $u \rightarrow \infty$  (so  $\Theta^{-1}(\xi_u) \rightarrow \infty$ ):

$$\frac{h \log n}{\Theta'(\Theta^{-1}(\xi_u))} \rightarrow 0 \quad (75)$$

Hence:

$$\Phi_n(u) - \Phi_n(u+h) \rightarrow -h \quad (76)$$

The diagonal oscillatory term  $\cos(\Phi_n(u) - \Phi_n(u+h))$  remains bounded by 1.

**Phase sum:** The sum  $\Phi_n(u) + \Phi_n(u+h) = 2u + h - \Theta^{-1}(u) \log n - \Theta^{-1}(u+h) \log n$  has derivative:

$$\frac{d}{du} [\Phi_n(u) + \Phi_n(u+h)] = 2 - \frac{\log n}{\Theta'(\Theta^{-1}(u))} - \frac{\log n}{\Theta'(\Theta^{-1}(u+h))} \quad (77)$$

By (67), both reciprocal terms vanish as  $u \rightarrow \infty$ , so:

$$\frac{d}{du} [\Phi_n(u) + \Phi_n(u+h)] \rightarrow 2 \quad (78)$$

For sufficiently large  $u > U_0$ , we have  $\left| \frac{d}{du} [\Phi_n(u) + \Phi_n(u+h)] \right| \geq 1$ .

By van der Corput's lemma (Lemma 24):

$$\left| \int_{U_0}^U \cos(\Phi_n(u) + \Phi_n(u+h)) du \right| = O(1) \quad (79)$$

Therefore, the Cesàro contribution from the phase sum:

$$\frac{1}{U} \int_{\Theta(0)}^U \cos(\Phi_n(u) + \Phi_n(u+h)) du = O(U^{-1}) \rightarrow 0 \quad (80)$$

**Step 4: Off-diagonal terms vanish**

For  $n \neq m$ , the cross term has phase:

$$\Phi_n(u) + \Phi_m(u+h) = 2u + h - \Theta^{-1}(u) \log n - \Theta^{-1}(u+h) \log m \quad (81)$$

The derivative is:

$$\frac{d}{du} [\Phi_n(u) + \Phi_m(u+h)] = 2 - \frac{\log n}{\Theta'(\Theta^{-1}(u))} - \frac{\log m}{\Theta'(\Theta^{-1}(u+h))} \rightarrow 2 \quad (82)$$

Identically to Step 3, van der Corput's lemma applies and:

$$\frac{1}{U} \int_{\Theta(0)}^U \cos(\Phi_n(u) + \Phi_m(u+h)) du = O(U^{-1}) \rightarrow 0 \quad (83)$$

**Step 5: Remainder terms vanish**

The weight factor in the transformation is:

$$W(u, h) = \frac{1}{\sqrt{\Theta'(\Theta^{-1}(u)) \Theta'(\Theta^{-1}(u+h))}} \sim \frac{1}{\log(\Theta^{-1}(u))} \quad (84)$$

The sum  $\sum_{n=1}^{N(t)} n^{-1/2} \cos(\Phi_n(u))$  is bounded by  $O(\sqrt{N(t)}) = O(t^{1/4})$  where  $t = \Theta^{-1}(u)$ .

Cross terms involving the remainder  $R(\Theta^{-1}(u+h)) = O(t^{-1/4})$  give:

$$W(u, h) \cdot O(t^{1/4}) \cdot O(t^{-1/4}) = O((\log t)^{-1}) \quad (85)$$

Changing to  $t$ -coordinates with  $u = \Theta(t)$  and  $du = \Theta'(t) dt$ :

$$\frac{1}{U} \int_{t_0}^{t_1} \frac{\Theta'(t)}{\log t} dt \sim \frac{1}{U} \int_{t_0}^{t_1} \frac{\frac{1}{2} \log t}{\log t} dt = \frac{t_1 - t_0}{2U} \quad (86)$$

Since  $U = \Theta(t_1) - \Theta(t_0) \sim \frac{t_1}{2} \log t_1$  for large  $t_1$ :

$$\frac{t_1 - t_0}{2U} \sim \frac{t_1}{t_1 \log t_1} = (\log t_1)^{-1} \rightarrow 0 \quad (87)$$

Similarly,  $R(\Theta^{-1}(u)) R(\Theta^{-1}(u+h)) = O(t^{-1/2})$  gives Cesàro average  $O(t^{-1/2}) \rightarrow 0$ .

**Step 6: Independence of starting point**

For any bounded integrable function  $f$  and starting points  $u_0, \tilde{u}_0 \geq \Theta(0)$ :

$$\left| \frac{1}{U} \int_{u_0}^{u_0+U} f du - \frac{1}{U} \int_{\tilde{u}_0}^{\tilde{u}_0+U} f du \right| \leq \frac{2|\tilde{u}_0 - u_0| \sup |f|}{U} \rightarrow 0 \quad (88)$$

**Conclusion**

Combining Steps 3–6, the Cesàro covariance limit

$$C(h) = \lim_{U \rightarrow \infty} \frac{1}{U - \Theta(0)} \int_{\Theta(0)}^U X(u) X(u+h) du \quad (89)$$

exists and is independent of the starting point  $\Theta(0)$ . This establishes that  $X(u)$  is wide-sense stationary in the Cesàro sense.  $\square$

**Corollary 26.** *The Hardy Z-function is a unitarily time-changed stationary process  $Z = U_\Theta X$ , where  $X$  is the Cesàro-stationary process characterized by Theorem 25. Therefore,  $Z$  is an oscillatory process with evolutionary power spectral density*

$$S_t(\lambda) = \Theta'(t) S_X(\lambda) \quad (90)$$

where  $S_X(\lambda)$  is the spectral density of  $X$ .

**Remark 27.** The convergence of the Cesàro covariance rigorously establishes that the Hardy Z-function, when viewed through theta-time coordinates, admits a well-defined stationary structure. The explicit form of  $C(h)$  encodes the deep spectral properties of the Riemann zeta function and requires detailed harmonic analysis of the Riemann-Siegel coefficients.

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