

Unitarily Time-Changed Stationary Processes: A Subclass of 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 locally square-integrable. Applying U_θ 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 ω . \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 Φ 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, λ) . Define

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

Then there exists a complex orthogonal random measure Φ 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 $f \in L_{\text{loc}}^2(\mathbb{R})$:

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

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

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

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 Φ 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 Φ 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 θ 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. □

3.2 Stationary to Oscillatory

Theorem 15. (Unitary time change produces oscillatory process) Let X be zero-mean stationary with spectral representation (2). Let θ 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). □

Corollary 16. (EPSP 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:

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

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). □

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. □

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 λ 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:

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

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. \square

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:

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

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 θ 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_Θ 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 $z = \frac{1}{4} + \frac{it}{2}$ 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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