

Value distribution of general Dirichlet series. VIII

A. Laurinčikas

Communicated by V. V. Kirichenko

ABSTRACT. A joint limit theorem on the complex plane for a new class of general Dirichlet series is proved.

1. Introduction

Let $s = \sigma + it$ be a complex variable, $\{a_m : m \in \mathbb{N}\}$ be a sequence of complex numbers, and let $\{\lambda_m : m \in \mathbb{N}\}$ be an increasing sequence of positive numbers, $\lim_{m \rightarrow \infty} \lambda_m = +\infty$. The series of the form

$$f(s) = \sum_{m=1}^{\infty} a_m e^{-\lambda_m s} \quad (1)$$

is called a general Dirichlet series. If $\lambda_m = \log m$, we obtain the ordinary Dirichlet series

$$\sum_{m=1}^{\infty} \frac{a_m}{m^s}.$$

It is well known that the region of convergence as well as of absolute convergence of Dirichlet series is a half-plane.

The first probabilistic results for Dirichlet series were obtained by H. Bohr and B. Jessen. In [2] and [3] they proved theorems for the Riemann zeta-function which are similar to modern limit theorems in the sense of weak convergence of probability measures. The investigations

2000 Mathematics Subject Classification: 11M41.

Key words and phrases: Compact topological group, general Dirichlet series, Haar measure, limit theorem, probability measure, random element, weak convergence.

of H.Bohr and B.Jessen were developed and generalized by A. Wintner, V.Borchsenius, A.Selberg, P.D.T.A. Elliott, A.Ghosh, K.Matsumoto, B.Bagchi, E.M.Nikishin, E.Stankus, J.Steuding, W.Schwarz, the author and others. The results of such a kind can be found in [7], [8], [14] and [20].

Limit theorems in the sense of weak convergence of probability measures in various spaces for general Dirichlet series were obtained [4]-[6], [10]-[14] and [18], [19]. Limit theorems on the complex plane for general Dirichlet series were proved in [12]-[14]. Denote by $meas\{A\}$ the Lebesgue measure of a measurable set $A \in \mathbb{R}$, and let, for $T > 0$,

$$\nu_T(\dots) = \frac{1}{T} \text{meas}\{t \in [0, T] : \dots\},$$

where in place of dots a condition satisfied by t is to be written. Moreover, let $\mathcal{B}(S)$ be the class of Borel sets of the space S .

Denote by γ the unit circle $\{s \in \mathbb{C} : |s| = 1\}$ on the complex plane \mathbb{C} , and define

$$\Omega = \prod_{m=1}^{\infty} \gamma_m,$$

where $\gamma_m = \gamma$ for each $m \in \mathbb{N}$. Then the infinite-dimensional torus Ω in view of the Tikhonov theorem is a compact topological Abelian group, therefore the probability Haar measure m_H on $(\Omega, \mathcal{B}(\Omega))$ can be defined. This gives a probability space $(\Omega, \mathcal{B}(\Omega), m_H)$. Denote by $\omega(m)$ the projection of $\omega \in \Omega$ to the coordinate space γ_m , $m \in \mathbb{N}$.

Suppose that the series (1) converges absolutely for $\sigma > \sigma_a$. Then the function $f(s)$ is analytic in the half-plane $\{s \in \mathbb{C} : \sigma > \sigma_a\}$. Moreover, we require that the function $f(s)$ should be meromorphically continuable to the half-plane $\{s \in \mathbb{C} : \sigma > \sigma_1\}$, $\sigma_1 < \sigma_a$, all poles being included in a compact set, and that, for $\sigma > \sigma_1$, the estimates

$$f(\sigma + it) \ll |t|^\alpha, \quad \alpha > 0, \quad |t| \geq t_0 > 0, \quad (2)$$

and

$$\int_{-T}^T |f(\sigma + it)|^2 dt \ll T, \quad T \rightarrow \infty, \quad (3)$$

should be satisfied. Suppose that the exponents λ_m satisfy the inequality

$$\lambda_m \geq (\log m)^\delta \quad (4)$$

with some positive $\delta > 0$. Then in [12] it was proved that under the last conditions, for $\sigma > \sigma_1$,

$$f(\sigma, \omega) = \sum_{m=1}^{\infty} a_m \omega(m) e^{-\lambda_m \sigma},$$

is a complex-valued random variable defined on the probability space $(\Omega, \mathcal{B}(\Omega), m_H)$. If the system $\{\lambda_m\}$ is linearly independent over the field of rational numbers, then it was obtained in [12] that, for $\sigma > \sigma_1$, the probability measure

$$\nu_T(f(\sigma + it) \in A), \quad A \in \mathcal{B}(\mathbb{C}), \quad (6)$$

converges weakly to the distribution of the random variable $f(\sigma, \omega)$ as $T \rightarrow \infty$.

Condition (4) is rather strong, it limits a class of Dirichlet series for which a limit theorem is true. Suppose that, for $\sigma > \sigma_1$,

$$\sum_{m=1}^{\infty} |a_m|^2 e^{-2\lambda_m \sigma} \log^2 m < \infty. \quad (7)$$

Then in [14] the following statement has been obtained.

Theorem A. *Suppose that the system $\{\lambda_m\}$ is linearly independent over the field of rational numbers, and conditions (2), (3) and (7) are satisfied. Then the probability measure (6) converges weakly to the distribution of the random element $f(\sigma, \omega)$ as $T \rightarrow \infty$.*

In [13] a joint limit theorem on the complex plane for general Dirichlet series was proved. Let, for $\sigma > \sigma_{aj}$,

$$f_j(s) = \sum_{m=1}^{\infty} a_{mj} e^{-\lambda_{mj} s},$$

where $\{a_{mj}\}$ and $\{\lambda_{mj}\}$ are a sequence of complex numbers and an increasing sequence of positive numbers, $\lim_{m \rightarrow \infty} \lambda_{mj} = +\infty$, respectively, $j = 1, \dots, r$, $r > 1$. Suppose that the function $f_j(s)$ is meromorphically continuable to the region $\{s \in \mathbb{C} : \sigma > \sigma_{1j}\}$, $\sigma_{1j} < \sigma_{aj}$, $j = 1, \dots, r$, all poles being included in a compact set, and, for $\sigma > \sigma_{1j}$, the estimates

$$f_j(\sigma + it) \ll |t|^{\alpha_j}, \quad \alpha_j > 0, \quad |t| \geq t_0 > 0, \quad (8)$$

and

$$\int_{-T}^T |f_j(\sigma + it)|^2 dt \ll T, \quad T \rightarrow \infty, \quad (9)$$

$j = 1, \dots, r$, are satisfied. Moreover, we assume that $\lambda_{mj} = \lambda_m$, $j = 1, \dots, r$, and

$$\lambda_m \geq c(\log m)^\delta, \quad c > 0, \quad \delta > 0. \quad (10)$$

Let $\mathbb{C}^r = \underbrace{\mathbb{C} \times \dots \times \mathbb{C}}_r$. On the probability space $(\Omega, \mathcal{B}(\Omega), m_H)$ define, for $\sigma_1 > \sigma_{11}, \dots, \sigma_r > \sigma_{1r}$, an \mathbb{C}^r -valued random element $F = F(\sigma_1, \dots, \sigma_r; \omega)$ by

$$F = F(\sigma_1, \dots, \sigma_r, \omega) = (f_1(\sigma_1, \omega), \dots, f_n(\sigma_r, \omega)),$$

where

$$f_j(\sigma_j, \omega) = \sum_{m=1}^{\infty} a_{mj} \omega(m) e^{-\lambda_m \sigma_j}, \quad \omega \in \Omega.$$

Theorem B [13]. *Suppose that the system $\{\lambda_m\}$ is linearly independent over the field of rational numbers, and that conditions (8)-(10) are satisfied. Then the probability*

$$P_T(A) \stackrel{\text{def}}{=} \nu_T((f_1(\sigma_1 + it), \dots, f_r(\sigma_r + it)) \in A), \quad A \in \mathcal{B}(\mathbb{C}^r),$$

for $\sigma_1 > \sigma_{1j}, \dots, \sigma_r > \sigma_{1r}$, converges weakly to the distribution of the random element $F(\sigma_1, \dots, \sigma_r; \omega)$ as $T \rightarrow \infty$.

The aim of this note is to change condition (10) in Theorem B by a weaker one and to consider a general case of different exponents λ_{mj} . Therefore, for the proof we will apply a method different from that of [13]. Suppose that, for $\sigma_j > \sigma_{1j}$,

$$\sum_{m=1}^{\infty} |a_{mj}|^2 e^{-2\lambda_{mj} \sigma_j} \log^2 m < \infty, \quad j = 1, \dots, r. \quad (11)$$

Moreover, define $\Omega^r = \Omega_1 \times \dots \times \Omega_r$ where $\Omega_j = \Omega$ for $j = 1, \dots, r$. Then Ω^r is also a compact topological Abelian group. Denote by m_{H^r} the probability Haar measure on $(\Omega^r, \mathcal{B}(\Omega^r))$.

In the next section it will be proved that, under condition (11), for $\sigma_1 > \sigma_{11}, \dots, \sigma_r > \sigma_{1r}$,

$$F(\sigma_1, \dots, \sigma_r; \underline{\omega}) = (f_1(\sigma_1, \omega_1), \dots, f_r(\sigma_r, \omega_r)),$$

where

$$f_j(\sigma_j, \omega_j) = \sum_{m=1}^{\infty} a_{mj} \omega_j(m) e^{-\lambda_{mj} \sigma_j}, \quad \omega_j \in \Omega_j, \quad j = 1, \dots, r, \quad \underline{\omega} = (\omega_1, \dots, \omega_r),$$

is a \mathbb{C}^n -valued random element defined on the probability space $(\Omega^r, \mathcal{B}(\Omega^r), m_{H^r})$.

Theorem 1. *Suppose that the set $\bigcup_{j=1}^r \bigcup_{m=1}^{\infty} \{\lambda_{mj}\}$ is linearly independent over the field of rational numbers, and that conditions (8), (9) and (11)*

are satisfied. Then the probability measure P_T converges weakly to the distribution of the random element $F(\sigma_1, \dots, \sigma_r; \underline{\omega})$ as $T \rightarrow \infty$.

Note that joint limit theorems can be used to derive the joint universality for considered functions, see, for example, [16] and [17].

2. The random element $F(\sigma_1, \dots, \sigma_r; \underline{\omega})$

In this section we will prove that, under condition (11), $F(\sigma_1, \dots, \sigma_r; \underline{\omega})$ is a \mathbb{C}^r -valued random element. For the proof, we will apply a Rademacher's theorem on series of pairwise orthogonal random variables. Denote by $\mathbb{E}\xi$ the expectation of the random element ξ .

Lemma 2.[20] *Suppose that $\{X_n\}$ is a sequence of orthogonal random variables such that*

$$\sum_{m=1}^{\infty} \mathbb{E}|X_m|^2 \log^2 m < \infty.$$

Then the series

$$\sum_{m=1}^{\infty} X_m$$

converges almost surely.

Theorem 3. *Suppose that condition (11) holds. Then $F(\sigma_1, \dots, \sigma_r; \underline{\omega})$, for $\sigma_1 > \sigma_{11}, \dots, \sigma_r > \sigma_{1r}$, is a \mathbb{C}^r -valued random element defined on the probability space $(\Omega, \mathcal{B}(\Omega), m_{Hr})$.*

Proof. Clearly, it suffices to prove that, for each $j = 1, \dots, r$,

$$f_j(\sigma_j, \omega) = \sum_{m=1}^{\infty} a_{mj} \omega(m) e^{-\lambda_{mj} \sigma_j}, \quad \omega \in \Omega,$$

for $\sigma_j > \sigma_{1j}$, is a complex-valued random variable on the probability space $(\Omega, \mathcal{B}(\Omega), m_H)$.

We fix $j \in \{1, \dots, r\}$. Let $\sigma > \sigma_{1j}$ be fixed, and

$$\xi_{mj} = \xi_{mj}(\omega) = a_{mj} \omega(m) e^{-\lambda_{mj} \sigma}.$$

Then $\{\xi_{mj}\}$ is a sequence of pairwise orthogonal complex-valued random variables defined on the probability space $(\Omega, \mathcal{B}(\Omega), m_H)$. Really, denoting by \bar{z} the complex conjugate of $z \in \mathbb{C}$, we find

$$\begin{aligned} \mathbb{E}(\xi_{mj}, \bar{\xi}_{kj}) &= \int_{\Omega} \xi_{mj}(\omega) \bar{\xi}_{kj}(\omega) dm_H = a_{mj} \bar{a}_{kj} e^{-(\lambda_{mj} + \lambda_{kj})\sigma} \int_{\Omega} \omega(m) \overline{\omega(k)} dm_H \\ &= \begin{cases} 0 & \text{if } m \neq k, \\ |a_{mj}|^2 e^{-2\lambda_{mj}\sigma} & \text{if } m = k. \end{cases} \end{aligned}$$

Since $\sigma > \sigma_{1j}$, hence we have in view of (11) that

$$\sum_{m=1}^{\infty} \mathbb{E}|\xi_{mj}|^2 \log^2 m < \infty.$$

This and Lemma 2 show that the series

$$\sum_{m=1}^{\infty} \xi_{mj} = \sum_{m=1}^{\infty} a_{mj} \omega(m) e^{-\lambda_{mj} \sigma} = f(\sigma, \omega) \quad (12)$$

converges almost surely with respect the Haar measure m_H . Then

$$\left(\sum_{m=1}^{\infty} a_{m1} \omega_1(m) e^{-\lambda_{m1} \sigma_1}, \dots, \sum_{m=1}^{\infty} a_{mr} \omega_r(m) e^{-\lambda_{mr} \sigma_r} \right)$$

converges almost surely in \mathbb{C}^r , and this proves the theorem. We note that $m_{Hr} = \underbrace{m_H \times \dots \times m_H}_r$.

3. Joint limit theorems for Dirichlet polynomials

We start with a joint limit theorem on the torus Ω^r . Define the probability measure

$$Q_{T,r}(A) = \nu_T(((e^{it\lambda_{m1}} : m \in \mathbb{N}), \dots, (e^{it\lambda_{mr}} : m \in \mathbb{N})) \in A).$$

Lemma 4. *The probability measure $Q_{T,r}$ converges weakly to the Haar measure m_{Hr} on $(\Omega^r, \mathcal{B}(\Omega^r))$ as $T \rightarrow \infty$.*

Proof. The dual group of Ω^r is

$$\bigoplus_{j=1}^r \bigoplus_{m=1}^{\infty} \mathbb{Z}_{m,j},$$

where $\mathbb{Z}_{m,j} = \mathbb{Z}$ for all $m \in \mathbb{N}$ and $j = 1, \dots, r$.

$$(\underline{k}_1, \dots, \underline{k}_r) = (k_{11}, k_{21}, \dots, k_{1r}, k_{2r}, \dots) \in \bigoplus_{j=1}^r \bigoplus_{m=1}^{\infty} \mathbb{Z}_{m,j},$$

where only a finite number of integers k_{mj} , $m \in \mathbb{N}$, $j = 1, \dots, r$, are distinct from zero, acts on Ω^r by

$$(\underline{x}_1, \dots, \underline{x}_r) \rightarrow (\underline{x}_1^{\underline{k}_1}, \dots, \underline{x}_r^{\underline{k}_r}) = \prod_{j=1}^r \prod_{m=1}^{\infty} x_{mj}^{k_{mj}}, \quad \underline{x}_j = (x_{1j}, x_{2j}, \dots), x_{mj} \in \gamma,$$

$m \in \mathbb{N}$, $j = 1, \dots, r$. Therefore, the Fourier transform $g_{T,r}(\underline{k}_1, \dots, \underline{k}_r)$ of the measure $Q_{T,r}$ is

$$\begin{aligned} g_{T,r}(\underline{k}_1, \dots, \underline{k}_r) &= \int_{\Omega^r} \prod_{j=1}^r \prod_{m=1}^{\infty} x_{mj}^{k_{mj}} dQ_{T,r} = \frac{1}{T} \int_0^T \prod_{j=1}^r \prod_{m=1}^{\infty} e^{itk_{mj}\lambda_{mj}} dt \\ &= \frac{1}{T} \int_0^T \exp\left\{it \sum_{j=1}^r \sum_{m=1}^{\infty} k_{mj}\lambda_{mj}\right\} dt. \end{aligned}$$

Since the set $\bigcup_{j=1}^r \bigcup_{m=1}^{\infty} \{\lambda_{mj}\}$ is linearly independent over the field of rational numbers, hence we find that

$$g_{T,r}(\underline{k}_1, \dots, \underline{k}_r) = \begin{cases} 1 & \text{if } (\underline{k}_1, \dots, \underline{k}_r) = (\mathcal{O}, \dots, \mathcal{O}), \\ \frac{\exp\left\{iT \sum_{j=1}^r \sum_{m=1}^{\infty} k_{mj}\lambda_{mj}\right\}^{-1}}{iT \sum_{j=1}^r \sum_{m=1}^{\infty} k_{mj}\lambda_{mj}} & \text{if } (\underline{k}_1, \dots, \underline{k}_r) \neq (\mathcal{O}, \dots, \mathcal{O}). \end{cases}$$

Therefore,

$$\lim_{T \rightarrow \infty} g_{T,r}(\underline{k}_1, \dots, \underline{k}_r) = \begin{cases} 1 & \text{if } (\underline{k}_1, \dots, \underline{k}_r) = (\mathcal{O}, \dots, \mathcal{O}), \\ 0 & \text{if } (\underline{k}_1, \dots, \underline{k}_r) \neq (\mathcal{O}, \dots, \mathcal{O}). \end{cases}$$

This and continuity theorems for probability measures on compact groups [7] show that the probability measure $Q_{T,r}$ converges weakly to the Haar measure m_{Hr} as $T \rightarrow \infty$.

Let $\sigma_{2j} > \sigma_{aj} - \sigma_{1j}$, and, for $m, n \in \mathbb{N}$,

$$v_j(m, n) = \exp\{-e^{(\lambda_m - \lambda_n)\sigma_{2j}}\}, \quad j = 1, \dots, r.$$

Define, for $N_j \in \mathbb{N}$, $\sigma_j > \sigma_{1j}$ and $\widehat{\omega}_j \in \Omega$,

$$f_{N_j, j, n}(\sigma_j + it) = \sum_{m=1}^{N_j} a_{mj} v_j(m, n) e^{-\lambda_{mj}(\sigma_j + it)},$$

$$f_{N_j, j, n}(\sigma_j + it, \widehat{\omega}_j) = \sum_{m=1}^{N_j} a_{mj} \widehat{\omega}_j(m) v_j(m, n) e^{-\lambda_{mj}(\sigma_j + it)}, \quad j = 1, \dots, r,$$

and consider the weak convergence of the probability measures

$$P_{T, N_1, \dots, N_r, n}(A) = \nu_T((f_{N_1, 1, n}(\sigma_1 + it), \dots, f_{N_r, r, n}(\sigma_r + it))) \in A$$

and

$$\widehat{P}_{T, N_1, \dots, N_r, n}(A) = \nu_T((f_{N_1, 1, n}(\sigma_1 + it, \widehat{\omega}_1), \dots, f_{N_r, r, n}(\sigma_r + it, \widehat{\omega}_r))) \in A,$$

where $(\widehat{\omega}_1, \dots, \widehat{\omega}_r) \in \Omega^r$ and $A \in \mathcal{B}(\mathbb{C}^r)$.

Theorem 5. *The probability measures $P_{T, N_1, \dots, N_r, n}$ and $\widehat{P}_{T, N_1, \dots, N_r, n}$ both converge weakly to the same probability measure on $(\mathbb{C}^r, \mathcal{B}(\mathbb{C}^r))$ as $T \rightarrow \infty$.*

Proof. Let a function $h : \Omega^r \rightarrow \mathbb{C}^r$ be given by

$$h(\omega_1, \dots, \omega_r) = \left(\sum_{m=1}^{N_1} a_{m1} v(m, n) e^{-\lambda_{m1} \sigma_1} \omega_1^{-1}(m), \dots, \sum_{m=1}^{N_r} a_{mr} v(m, n) e^{-\lambda_{mr} \sigma_r} \omega_r^{-1}(m) \right),$$

$(\omega_1, \dots, \omega_r) \in \Omega^r$. Then, clearly,

$$\begin{aligned} & h\left((e^{it\lambda_{m1}} : m \in \mathbb{N}), \dots, (e^{it\lambda_{mr}} : m \in \mathbb{N}) \right) \\ &= (f_{N_1, 1, n}(\sigma_1 + it), \dots, f_{N_r, r, n}(\sigma_r + it)) \\ & \stackrel{\text{def}}{=} f_{N_1, \dots, N_r, n}(\sigma_1, \dots, \sigma_r; t), \end{aligned}$$

and the function h is continuous. Therefore, $P_{T, N_1, \dots, N_r, n} = Q_{T, r} h^{-1}$, and by Theorem 5.1 of [1] and Lemma 4 the probability measure $P_{T, N_1, \dots, N_r, n}$ converges weakly to $m_{Hr} h^{-1}$ as $T \rightarrow \infty$.

Now let $h_1 : \Omega^r \rightarrow \Omega^r$ be defined by the formula

$$h_1(\omega_1, \dots, \omega_r) = (\omega_1 \widehat{\omega}_1^{-1}, \dots, \omega_r \widehat{\omega}_r^{-1}).$$

Then we have that

$$\begin{aligned} & (f_{N_1, 1, n}(\sigma_1 + it, \widehat{\omega}_1), \dots, f_{N_r, 1, n}(\sigma_r + it, \widehat{\omega}_r)) = \\ & h(h_1((e^{it\lambda_{m1}} : m \in \mathbb{N}), \dots, (e^{it\lambda_{mr}} : m \in \mathbb{N}))). \end{aligned}$$

Similarly to the case of the measure $P_{T, N_1, \dots, N_r, n}$ we obtain that the probability measure $P_{T, N_1, \dots, N_r, n}$ converges weakly to the measure $m_{Hr} (hh_1)^{-1}$ as $T \rightarrow \infty$. The Haar measure m_{Hr} is invariant with respect to translations by points from Ω^r . Therefore,

$$m_{Hr} (hh_1)^{-1} = (m_{Hr} h_1^{-1}) h^{-1} = m_{Hr} h^{-1},$$

and the theorem is proved.

4. Limit theorems for absolutely convergent series

Define, for $\omega_j \in \Omega$ and $j = 1, \dots, r$,

$$f_{n,j}(s) = \sum_{m=1}^{\infty} a_{mj} v_j(m, n) e^{-\lambda_{mj} s}$$

and

$$f_{n,j}(s, \omega_j) = \sum_{m=1}^{\infty} a_{mj} \omega_j(m) v_j(m, n) e^{-\lambda_{mj} s}.$$

Then the latter series both converge absolutely for $\sigma > \sigma_{1j}$. The proof of this is given in [12], Lemma 4. In this section we consider the weak convergence of the probability measures

$$P_{T,n}(A) = \nu_T(((f_{n,1}(\sigma_1 + it), \dots, f_{n,r}(\sigma_r + it)) \in A)), \quad A \in \mathcal{B}(\mathbb{C}^r),$$

and

$$\widehat{P}_{T,n}(A) = \nu_T(((f_{n,1}(\sigma_1 + it, \omega_1), \dots, f_{n,r}(\sigma_r + it, \omega_r)) \in A)), \quad A \in \mathcal{B}(\mathbb{C}^r).$$

Theorem 6. *Let $\sigma_j > \sigma_{1j}$, $j = 1, \dots, r$. Then there exists a probability measure P_n on $(\mathbb{C}^r, \mathcal{B}(\mathbb{C}^r))$ such that the measures $P_{T,n}$ and $\widehat{P}_{T,n}$ both converge weakly to P_n as $T \rightarrow \infty$.*

Proof. We will apply Theorem 5. Without loss of generality we take $N_1 = \dots = N_r \stackrel{\text{def}}{=} N$. Then by Theorem 5 the measures $P_{T,N_1, \dots, N_r, n} \stackrel{\text{def}}{=} P_{T,N,n}$ and $\widehat{P}_{T,N_1, \dots, N_r, n} \stackrel{\text{def}}{=} \widehat{P}_{T,N,n}$ both converge weakly to the same measure $P_{N,n}$, say, as $T \rightarrow \infty$.

First we will prove that the family of probability measures $\{P_{N,n}\}$ is tight for fixed n . Let η be a random variable defined on a certain probability space $(\widehat{\Omega}, \mathcal{F}, \mathbb{P})$ and uniformly distributed on $[0, 1]$, and let, for $j = 1, \dots, r$,

$$X_{T,N,j,n} = X_{T,N,j,n}(\sigma_j) = f_{N,j,n}(\sigma_j + iT\eta).$$

Then we have that

$$\underline{X}_{T,N,n} \stackrel{\text{def}}{=} (X_{T,N,1,n}, \dots, X_{T,N,r,n}) \xrightarrow[T \rightarrow \infty]{\mathcal{D}} \underline{X}_{N,n}, \quad (12)$$

where $\underline{X}_{N,n}$ is a \mathbb{C}^r -valued random element with distribution $P_{N,n}$, and $\xrightarrow{\mathcal{D}}$ means the convergence in distribution.

Let $\underline{z}_1 = (z_{11}, \dots, z_{1r})$, $\underline{z}_2 = (z_{21}, \dots, z_{2r}) \in \mathbb{C}^r$. Define a metric ρ in \mathbb{C}^r by

$$\rho(\underline{z}_1, \underline{z}_2) = \left(\sum_{j=1}^r |z_{1j} - z_{2j}|^2 \right)^{\frac{1}{2}}.$$

Then, clearly, this metric induces the topology of \mathbb{C}^r .

Since the series for $f_{n,j}$ converges absolutely for $\sigma > \sigma_{1j}$, $j = 1, \dots, r$, we obtain, for $M > 0$,

$$\begin{aligned} \limsup_{T \rightarrow \infty} \mathbb{P}(\rho(\underline{X}_{T,N,n}, \underline{0}) > M) &\leq \\ &\leq \frac{1}{M} \sup_{N \geq 1} \limsup_{T \rightarrow \infty} \frac{1}{T} \int_0^T \rho(f_{\underline{N},n}(\sigma_1, \dots, \sigma_r; t), \underline{0}) \, dt = \\ &= \frac{1}{M} \sup_{N \geq 1} \limsup_{T \rightarrow \infty} \frac{1}{T} \int_0^T \left(\sum_{j=1}^r |f_{N,j,n}(\sigma_j + it)|^2 \right)^{\frac{1}{2}} \, dt \leq \\ &\leq \frac{1}{M} \sup_{N \geq 1} \limsup_{T \rightarrow \infty} \left(\frac{1}{T} \sum_{j=1}^r \int_0^T |f_{N,j,n}(\sigma_j + it)|^2 \, dt \right)^{\frac{1}{2}} = \\ &= \frac{1}{M} \sup_{N \geq 1} \left(\sum_{j=1}^r \sum_{m=1}^N |a_{mj}|^2 v_j^2(m, n) e^{-2\lambda_{mj}\sigma_j} \right)^{\frac{1}{2}} \leq R < \infty, \end{aligned} \tag{13}$$

where

$$f_{\underline{N},n}(\sigma_1, \dots, \sigma_r; t) = (f_{N,1,n}(\sigma_1 + it), \dots, f_{N,r,n}(\sigma_r + it)).$$

Now we take $M = R\epsilon^{-1}$, where ϵ is an arbitrary positive number. Then (13) yields

$$\limsup_{T \rightarrow \infty} \mathbb{P}(\rho(\underline{X}_{T,N,n}, \underline{0}) > M) \leq \epsilon.$$

This and (12) imply the inequality

$$\mathbb{P}(\rho(\underline{X}_{T,N,n}, \underline{0}) > M) \leq \epsilon. \tag{14}$$

Now we define

$$K_\epsilon = \{\underline{z} \in \mathbb{C}^r : \rho(\underline{z}, \underline{0}) \leq M\}.$$

Then, obviously, K_ϵ is a compact subset of the space \mathbb{C}^r . In view of (14) and of the definition of $P_{N,n}$

$$P_{N,n}(K_\epsilon) \geq 1 - \epsilon$$

for all $N \in \mathbb{N}$. This shows that the tightness of the family $\{P_{N,n}\}$. Hence, by the Prokhorov theorem, see, for example, [1], the latter family is relatively compact.

By the definition of $f_{n,j}(s)$ and $f_{N,n,j}(s)$, for $\sigma > \sigma_{1j}$,

$$\lim_{N \rightarrow \infty} f_{N,j,n}(s) = f_{n,j}(s), \quad j = 1, \dots, r,$$

and the series for $f_{n,j}(s)$ absolutely converges. Therefore, denoting

$$\underline{f}_n(\sigma_1, \dots, \sigma_r; t) = (f_{n,1}(\sigma_1 + it), \dots, f_{n,r}(\sigma_r + it)),$$

we have, for every $\epsilon > 0$ and $\sigma_j > \sigma_{1j}$, $j = 1, \dots, r$, that

$$\begin{aligned} & \lim_{N \rightarrow \infty} \limsup_{T \rightarrow \infty} \nu(\rho(\underline{f}_{N,n}(\sigma_1, \dots, \sigma_r; t), \underline{f}_n(\sigma_1, \dots, \sigma_r; t)) \geq \epsilon) \leq \\ & \leq \lim_{N \rightarrow \infty} \limsup_{T \rightarrow \infty} \frac{1}{\epsilon T} \int_0^T \rho(\underline{f}_{N,n}(\sigma_1, \dots, \sigma_r; t), \underline{f}_n(\sigma_1, \dots, \sigma_r; t)) dt = 0. \end{aligned} \quad (15)$$

Define, for $\sigma_j > \sigma_{1j}$,

$$X_{T,j,n} = X_{T,n}(\sigma_j) = f_{n,j}(\sigma_j + iT\eta), \quad j = 1, \dots, r,$$

and put

$$\underline{X}_{T,n} = (X_{T,1,n}, \dots, X_{T,r,n}).$$

Then by (15)

$$\lim_{N \rightarrow \infty} \limsup_{T \rightarrow \infty} \mathbb{P}(\rho(\underline{X}_{T,N,n}, \underline{X}_{T,n}) \geq \epsilon) = 0. \quad (16)$$

The family $\{P_{N,n}\}$ is relatively compact. Therefore, there exists a subsequence $\{P_{N',n}\} \subset \{P_{N,n}\}$ which converges weakly to the probability measure P_n , say, as $N' \rightarrow \infty$. Then

$$\underline{X}_{N',n} \xrightarrow[N' \rightarrow \infty]{\mathcal{D}} P_n. \quad (17)$$

The space \mathbb{C}^r is separable. Therefore, (12), (16) and (17) show that the conditions of Theorem 4.2 from [1] are satisfied. Consequently,

$$\underline{X}_{T,n} \xrightarrow[T \rightarrow \infty]{\mathcal{D}} P_n, \quad (18)$$

i.e. the measure $P_{T,n}$ converges weakly to the probability measure P_n on $(\mathbb{C}^r, \mathcal{B}(\mathbb{C}^r))$ as $T \rightarrow \infty$.

In view of (18), the measure P_n is independent of the subsequence $\{P_{N',n}\}$. Therefore, by (17)

$$\underline{X}_{N,n} \xrightarrow[N \rightarrow \infty]{\mathcal{D}} P_n. \quad (19)$$

Now, repeating the above arguments for the random elements

$$\widehat{\underline{X}}_{T,N,n} = (\widehat{X}_{T,N,1,n}, \dots, \widehat{X}_{T,N,r,n})$$

and

$$\widehat{\underline{X}}_{T,n} = (\widehat{X}_{T,1,n}, \dots, \widehat{X}_{T,r,n}),$$

where

$$\widehat{X}_{T,N,j,n} = \widehat{X}_{T,N,j,n}(\sigma_j, \omega_j) = f_{N,j,n}(\sigma_j + iT\eta, \omega_j), \quad j = 1, \dots, r,$$

$$\widehat{X}_{T,j,n} = \widehat{X}_{T,j,n}(\sigma_j, \omega_j) = f_{j,n}(\sigma_j + iT\eta, \omega_j), \quad j = 1, \dots, r,$$

and taking into account (19), we obtain that the probability measure $\widehat{P}_{T,n}$ also converges weakly to P_n as $T \rightarrow \infty$. The theorem is proved.

5. Approximation in the mean

To pass from the functions $f_{n,j}(s)$ to $f_j(s)$ we need an approximation in the mean of $f_1(s), \dots, f_r(s)$ and of $f_1(s, \omega_1), \dots, f_r(s, \omega_r)$ by $f_{n,1}(s), \dots, f_{n,r}(s)$ and by $f_{n,1}(s, \omega_1), \dots, f_{n,r}(s, \omega_r)$, respectively. Let

$$\underline{f}(\sigma_1, \dots, \sigma_r; t) = (f_1(\sigma_1 + it), \dots, f_r(\sigma_r + it)),$$

and

$$\underline{f}(\sigma_1, \dots, \sigma_r; t, \underline{\omega}) = (f_1(\sigma_1 + it, \omega_1), \dots, f_r(\sigma_r + it, \omega_r)),$$

$$\underline{f}_n(\sigma_1, \dots, \sigma_r; t, \underline{\omega}) = (f_{n,1}(\sigma_1 + it, \omega_1), \dots, f_{n,r}(\sigma_r + it, \omega_r)).$$

Theorem 7. *Let $\sigma_j > \sigma_{1j}$, $j = 1, \dots, r$. Then*

$$\lim_{N \rightarrow \infty} \limsup_{T \rightarrow \infty} \frac{1}{T} \int_0^T \rho(\underline{f}(\sigma_1, \dots, \sigma_r; t), \underline{f}_n(\sigma_1, \dots, \sigma_r; t)) dt = 0$$

and

$$\lim_{N \rightarrow \infty} \limsup_{T \rightarrow \infty} \frac{1}{T} \int_0^T \rho(\underline{f}(\sigma_1, \dots, \sigma_r; t, \underline{\omega}), \underline{f}_n(\sigma_1, \dots, \sigma_r; t, \underline{\omega})) dt = 0$$

for almost all $(\omega_1, \dots, \omega_r)$.

Proof. Suppose that the function $f(s)$ satisfies the conditions of Theorem A, and for $\sigma > \sigma_1$,

$$f_n(s) = \sum_{m=1}^{\infty} a_m v(m, n) e^{-\lambda_m s},$$

$$f_n(s, \omega) = \sum_{m=1}^{\infty} a_m \omega(m) v(m, n) e^{-\lambda_m s},$$

where $v(m, n) = \exp\{-e^{-(\lambda_n - \lambda_m)\sigma_2}\}$ with $\sigma_2 > \sigma_a - \sigma_1$, and $\omega \in \Omega$. Then in [12] it was obtained that, for $\sigma > \sigma_1$,

$$\lim_{N \rightarrow \infty} \limsup_{T \rightarrow \infty} \frac{1}{T} \int_0^T |f(\sigma + it) - f_n(\sigma + it)| dt = 0$$

and

$$\lim_{N \rightarrow \infty} \limsup_{T \rightarrow \infty} \frac{1}{T} \int_0^T |f(\sigma + it, \omega) - f_n(\sigma + it, \omega)| dt = 0$$

for almost all $\omega \in \Omega$. Since

$$\rho(\underline{z}_1, \underline{z}_2) \leq \sum_{j=1}^r |z_{1j} - z_{2j}|,$$

hence the theorem follows.

6. Joint limit theorems for $f_j(s)$ and $f_j(s, \omega)$

In this section we begin to prove Theorem 1. We will prove limit theorems for the vectors $\underline{f}(\sigma_1, \dots, \sigma_r; t)$ and $\underline{f}(\sigma_1, \dots, \sigma_r; t, \omega)$ defined in Section 5.

Theorem 8. *Let $\sigma_j > \sigma_{1j}$, $j = 1, \dots, r$. Then the probability measures P_T and*

$$\widehat{P}_T(A) = \nu_T(\underline{f}(\sigma_1, \dots, \sigma_r; t, \omega) \in A), \quad A \in \mathcal{B}(\mathbb{C}^r),$$

both converge weakly to the same probability measure on $(\mathbb{C}^r, \mathcal{B}(\mathbb{C}^r))$ as $T \rightarrow \infty$.

Proof. We argue similarly to the proof of Theorem 6. By Theorem 6 the probability measures $P_{T,n}$ and $\widehat{P}_{T,n}$ converge weakly to the same measure P_n on $(\mathbb{C}^r, \mathcal{B}(\mathbb{C}^r))$ as $T \rightarrow \infty$. We will show that the family of probability measures $\{P_n : n \in \mathbb{N}\}$ is tight. For this, we will preserve the notation of previous sections.

By Theorem 6

$$\underline{X}_{T,n} \xrightarrow[T \rightarrow \infty]{\mathcal{D}} \underline{X}_n, \tag{20}$$

where \underline{X}_n is a \mathbb{C}^r -valued random element with distribution P_n . Since the series (11) converges and the series for each $f_{n,j}$ converges absolutely, we

have, for $M > 0$,

$$\begin{aligned}
\limsup_{T \rightarrow \infty} \mathbb{P}(\rho(\underline{X}_{T,n}, \underline{0}) > M) &\leq \\
&\leq \frac{1}{M} \sup_{n \geq 1} \limsup_{T \rightarrow \infty} \frac{1}{T} \int_0^T \rho(\underline{f}_n(\sigma_1, \dots, \sigma_r; t), \underline{0}) \, dt = \\
&= \frac{1}{M} \sup_{n \geq 1} \limsup_{T \rightarrow \infty} \frac{1}{T} \int_0^T \left(\sum_{j=1}^r |f_{n,j}(\sigma_j + it)|^2 \right)^{\frac{1}{2}} \, dt \leq \\
&\leq \frac{1}{M} \sup_{n \geq 1} \limsup_{T \rightarrow \infty} \left(\frac{1}{T} \sum_{j=1}^r \int_0^T |f_{n,j}(\sigma_j + it)|^2 \, dt \right)^{\frac{1}{2}} = \\
&= \frac{1}{M} \sup_{n \geq 1} \left(\sum_{j=1}^r \sum_{m=1}^{\infty} |a_{mj}|^2 v_j^2(m, n) e^{-2\lambda_{mj}\sigma_j} \right)^{\frac{1}{2}} \leq \\
&\leq \frac{1}{M} \left(\sum_{j=1}^r \sum_{m=1}^{\infty} |a_{mj}|^2 e^{-2\lambda_{mj}\sigma_j} \right)^{\frac{1}{2}} \leq R < \infty.
\end{aligned}$$

Hence, taking $M = R\epsilon^{-1}$, we find that

$$\limsup_{T \rightarrow \infty} \mathbb{P}(\rho(\underline{X}_{T,n}, \underline{0}) > M) \leq \epsilon.$$

Consequently, in view of (20)

$$\mathbb{P}(\rho(\underline{X}_n, \underline{0}) > M) \leq \epsilon.$$

This shows that

$$P_n(K_\epsilon) \geq 1 - \epsilon$$

for all $n \in \mathbb{N}$, i.e. the family $\{P_n\}$ is tight. Hence, by the Prokhorov theorem, it is relatively compact. Therefore, there exists a subsequence $\{P_{n_1}\} \subset \{P_n\}$ which converges weakly to the probability measure P , say, on $(\mathbb{C}^r, \mathcal{B}(\mathbb{C}^r))$ as $n_1 \rightarrow \infty$. Then

$$\underline{X}_{n_1} \xrightarrow[n_1 \rightarrow \infty]{\mathcal{D}} P. \quad (21)$$

Let, for $\sigma_j > \sigma_{1j}$

$$X_{T,j} = X_{T,j}(\sigma_j) = f_j(\sigma_j + iT\eta), \quad j = 1, \dots, r,$$

and

$$\underline{X}_T = (X_{T,1}, \dots, X_{T,r}).$$

Then by the first assertion of Theorem 7

$$\lim_{n \rightarrow \infty} \limsup_{T \rightarrow \infty} \mathbb{P}(\rho(\underline{X}_{T,n}, \underline{X}_T) \geq \epsilon) \leq$$

$$\lim_{n \rightarrow \infty} \limsup_{T \rightarrow \infty} \frac{1}{\epsilon T} \int_0^T \rho(\underline{f}_n(\sigma_1, \dots, \sigma_r; t), \underline{f}(\sigma_1, \dots, \sigma_r; t)) = 0.$$

This, (20), (21) and Theorem 4.2 of [1] show that

$$\underline{X}_T \xrightarrow[T \rightarrow \infty]{\mathcal{D}} P. \tag{23}$$

Now let, for $\sigma_j > \sigma_{1j}$,

$$\widehat{X}_{T,j} = \widehat{X}_{T,j}(\sigma_j) = f_j(\sigma_j + iT\eta, \omega_j), \quad j = 1, \dots, r,$$

and

$$\widehat{X}_T = (\widehat{X}_{T,1}, \dots, \widehat{X}_{T,r}).$$

Then, reasoning similarly as above for the vectors $\widehat{X}_{T,n}$ and \widehat{X}_T , and using (23) and the second assertion of Theorem 7, we obtain that the probability measure \widehat{P}_T also converges to P as $T \rightarrow \infty$. The theorem is proved.

7. Proof of Theorem 1

It remains to identify the limit measure P in Theorem 8. For this, we will apply some elements of the ergodic theory.

Let $a_{t,j} = \{e^{-i\lambda_m j t} : m \in \mathbb{N}\}$ for $t \in \mathbb{R}$, $j = 1, \dots, r$. Then, for each j , $\{a_{t,j} : t \in \mathbb{R}\}$ is a one-parameter group. We define the one-parameter family $\{\varphi_{t,j} : t \in \mathbb{R}\}$ of transformations on Ω_j by $\varphi_{t,j} = a_{t,j}\omega_j$ for $\omega_j \in \Omega_j$, $j = 1, \dots, r$. Then we obtain a one parameter group $\{\varphi_{t,j} : t \in \mathbb{R}\}$ of measurable transformations on Ω_j , $j = 1, \dots, r$.

Define $\{\Phi_t : t \in \mathbb{R}\} = \{\varphi_{t,1} : t \in \mathbb{R}\} \times \dots \times \{\varphi_{t,r} : t \in \mathbb{R}\}$. Then $\{\Phi_t : t \in \mathbb{R}\}$ is a one-parameter group of measurable transformations on Ω^r .

Lemma 9. *The one-parameter group $\{\Phi_t : t \in \mathbb{R}\}$ is ergodic.*

Proof. In [18] it was proved that $\{\varphi_{t,j} : t \in \mathbb{R}\}$ for each $j = 1, \dots, r$ is an ergodic one-parameter group. Hence the lemma follows.

Proof of Theorem 1. Let $A \in \mathcal{B}(\mathbb{C}^r)$ be a continuity set of the measure P in Theorem 8. Then, by Theorem 10, for $\sigma_1 > \sigma_{11}, \dots, \sigma_r > \sigma_{1r}$,

$$\lim_{T \rightarrow \infty} \nu_T(\underline{f}(\sigma_1, \dots, \sigma_r; t, \underline{\omega}) \in A) = P(A) \tag{24}$$

for almost all $\underline{\omega} \in \Omega^r$. Now we fix the set A and define a random variable θ on $(\Omega^r, \mathcal{B}(\Omega^r), m_{Hr})$ by

$$\theta(\underline{\omega}) = \begin{cases} 1 & \text{if } F(\sigma_1, \dots, \sigma_r; \underline{\omega}) \in A, \\ 0 & \text{if } \underline{F}(\sigma_1, \dots, \sigma_r; \underline{\omega}) \notin A. \end{cases}$$

Then

$$\mathbb{E}(\theta) = \int_{\Omega^r} \theta d m_{Hr} = m_{Hr}(\omega \in \Omega : \underline{F}(\sigma_1, \dots, \sigma_r; \underline{\omega}) \in A) \stackrel{\text{def}}{=} P_F$$

is the distribution of the random element \underline{F} . Since by Lemma 9 the one-parameter group $\{\Phi_t : t \in \mathbb{R}\}$ is ergodic, the random process $\theta(\Phi_t(\underline{\omega}))$ is also ergodic. Therefore, by the Birkhoff-Khinchine theorem

$$\lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T \theta(\Phi_t(\underline{\omega})) dt = \mathbb{E}(\theta) \quad (26)$$

for almost all $\underline{\omega} \in \Omega^r$. The definitions of θ and of $\{\Phi_t : t \in \mathbb{R}\}$ yield

$$\frac{1}{T} \int_0^T \theta(\Phi_t(\underline{\omega})) dt = \nu_T(\underline{f}(\sigma_1, \dots, \sigma_r; t, \underline{\omega}) \in A).$$

Hence and from (25), (26), we deduce that

$$\lim_{T \rightarrow \infty} \nu_T(\underline{f}(\sigma_1, \dots, \sigma_r; t, \underline{\omega}) \in A) = P_F(A)$$

for almost all $\underline{\omega} \in \Omega^r$. Consequently, by (24)

$$P(A) = P_F(A)$$

for any continuity set A of the measure P . It is well known that all continuity sets constitute the determining class. Therefore,

$$P(A) = P_F(A)$$

for all $A \in \mathcal{B}(\mathbb{C}^r)$, and the theorem is proved.

References

- [1] P. Billingsley, *Convergence of Probability Measures*, John Wiley, New York, 1968.
- [2] H. Bohr, B. Jessen, Über die Wertverteilung der Riemannsches Zetafunktion, Erste Mitteilung, *Acta Math.* **54** (1930), 1-35.
- [3] H. Bohr, B. Jessen, Über die Wertverteilung der Riemannsches Zetafunktion, Zweite Mitteilung, *Acta Math.* **58** (1932), 1-55.
- [4] J. Genys, A. Laurinčikas, Value distribution of general Dirichlet series. IV, *Lith.Math.J.* **43**(3)(2003), 281-294.

- [5] J. Genys, A. Laurinčikas, On joint limit theorem for general Dirichlet series, *Nonlinear Analysis: Modelling and Control*, **8**(2) (2003), 27-39.
- [6] J. Genys, A. Laurinčikas, A joint limit theorem for general Dirichlet series, *Lith.Math.J.* **44**(1)(2004), 145-156.
- [7] H. Heyer, *Probability Measures on Locally Compact Groups*, Springer-Verlag, Berlin, Heidelberg, New York, 1977.
- [8] D. Joyner, *Distribution Theorems for L -functions*, Longman Scientific and Technical, Harlow, 1986.
- [9] A. Laurinčikas, *Limit Theorems for the Riemann Zeta-Function*, Kluwer Academic Publishers, Dordrecht, Boston, London, 1996.
- [10] A. Laurinčikas, Value-distribution of general Dirichlet series, in : *Probability Theory and Math. Statistics: Proceedings of the Seventh Vilnius Conference (1998)*, B. Grigelionis et al (Eds) VSP/Utrecht, TEV/Vilnius (1999), 405-419.
- [11] A. Laurinčikas, Value-distribution of general Dirichlet series. II, *Lith.Math.J.* **41**(4)(2001), 351-360.
- [12] A. Laurinčikas, Limit theorems for general Dirichlet series, *Theory Stoch. Processes*, **8**(24) No 3-4 (2002), 256-269.
- [13] A. Laurinčikas, A joint limit theorem on the complex plane for general Dirichlet series, *Lith.Math.J.* **44**(3)(2004), 225-231.
- [14] A. Laurinčikas, Value-distribution of general Dirichlet series. IV, *Nonlinear Analysis: Modelling and Control*, **10**(3) (2005), 1-13.
- [15] A. Laurinčikas, R. Garunkštis, *The Lerch Zeta-Function*, Kluwer Academic Publishers, Dordrecht, Boston, London, 2002.
- [16] A. Laurinčikas, K. Matsumoto, The joint universality and the functional independence for Lerch zeta-functions, *Nagoya Math.J.* **157** (2000), 211-227.
- [17] A. Laurinčikas, K. Matsumoto, The joint universality of twisted automorphic L -functions, *J.Math.Soc. Japan*, **56**(3) (2004), 923-939.
- [18] A. Laurinčikas, W. Schwarz and J. Steuding, Value distribution of general Dirichlet series. III, in: *Analytic and Probab. Methods in Number Theory, Proceedings of the Third Intern. Conference in honour of J. Kubilius, Palanga (2001)*, A. Dubickas et al (Eds), TEV, Vilnius (2002), 137-156.
- [19] A. Laurinčikas, J. Steuding, A joint limit theorem for general Dirichlet series, *Lith. Math. J.* **42**(2) (2002), 163-173.
- [20] M. Loève, *Probability Theory*, Van Nostrand Company, Toronto, New York, London 1955.

CONTACT INFORMATION

A. Laurinčikas Vilnius University, Naugarduko 24, 03225
Vilnius, Lithuania
E-Mail: antanas.laurincikas@maf.vu.lt

Received by the editors: 11.05.2003
and in final form 06.04.2007.