

ISSN: 2456-1452 Maths 2017; 2(2): 60-63 © 2017 Stats & Maths www.mathsjournal.com Received: 25-01-2017 Accepted: 27-02-2017

Vinita Vijai Associate Professor, Department of Mathematics, Isabella Thoburn College, Lucknow, Uttar Pradesh, India

A study of convex properties of mean values of entire functions of several complex variables represented by multiple Dirichlet series

Vinita Vijai

Abstract

In this paper, an attempt is made to study some convex problems relating mean values $I_2(\sigma_1, \sigma_2)$ and $m_{2,k}(\sigma_1, \sigma_2)$ of an entire function $f(s_1, s_2)$ of two complex variables.

Keywords: Convex properties, mean values, entire functions, several complex variables represented, multiple Dirichlet series

Introduction

Let us consider

(1.1)
$$f(s_1, s_2) = \sum_{m, n=1}^{\infty} a_{m,n} \exp(s_1 \lambda_m + s_2 \mu_n),$$

$$((s_j = \sigma_j + it_j), j = 1,2 \text{ where } a_{m,n} \in C,$$

the field of complex numbers, $\lambda'_m s$, $\mu'_n s$ are real, and

$$0 < \lambda_1 < \lambda_2 < \cdots \lambda_m \to \infty;$$

$$0 < \mu_1 < \mu_2 < \dots < \mu_n \rightarrow \infty$$

It has been proved [1] that if

$$(1.2) \lim_{m \to \infty} \frac{\log m}{\lambda_m} = 0, \lim_{n \to \infty} \frac{\log n}{\mu_n} = 0,$$

The domain of convergence of the series (1.1) coincides with its domain of absolute convergence. Also, Sarkar [2, pp.99] has shown that the necessary and sufficient condition that the series

(1.1) satisfying (1.2) to be entire is that

$$(1.3) \lim_{(m,n)\to\infty} \frac{\log|a_{m,n}|}{\lambda_m + \mu_n} = -\infty$$

Let the family of all double Dirichlet series of the form (1.1) satisfying (1.2) and (1.3) be denoted by F.

Then $f \in F$ denotes an entire function over C^2 . The results can be extended to several complex variables.

Corresponding to an $f \in F$, the maximum modulus $M = M_f$ and the maximum term $\mu = \mu_f$ on R^2 Are defined as [2, pp100]

Correspondence Vinita Vijai Associate Professor, Department of Mathematics, Isabella Thoburn College, Lucknow, Uttar Pradesh, India International Journal of Statistics and Applied Mathematics

$$M(\sigma) = M_f(\sigma_1, \sigma_2) = \max\{|f(s_1, s_2)| : s_1, s_2 \in C, \text{Re } s_1 = \sigma_1, \text{Re } s_2 = \sigma_2\}$$

where N is the set of natural numbers.

The mean value $I_2(\sigma_1, \sigma_2)$ of $|f(s_1, s_2)|$ is defined as

$$(1.4)I_2(\sigma_{1},\sigma_{2};f) = I_2(\sigma_{1},\sigma_{2}) = \lim_{T_{11}T_{2}\to\infty} \frac{1}{4T_{1}T_{2}} \int_{-T_{1}}^{T_{1}} \int_{-T_{2}}^{T_{2}} |f(\sigma_{1}+it_{1},\sigma_{2}+it_{2})|^{2} dt_{1} dt_{2}$$

Now we define the mean value $m_{2,k}(\sigma_1, \sigma_2)$ of $|f(s_1, s_2)|$ as

$$(1.5) m_{2,k}(\sigma_1, \sigma_2; f) = m_{2,k}(\sigma_1, \sigma_2)$$

$$=\lim_{T_{1,T_{2}}\to\infty}\frac{1}{T_{1}T_{2}e^{k\sigma_{1}}e^{k\sigma_{2}}}\int_{0}^{\sigma_{1}}\int_{0}^{\sigma_{2}}\int_{-T_{1}}^{T_{1}}\int_{-T_{2}}^{T_{2}}\{|f(x_{1}+it_{1},x_{2}+jt_{2})|^{2}e^{kx_{1}}e^{kx_{2}}\}dx_{1}dx_{2}dt_{1}dt_{2}$$

Where k is any positive number.

These mean values defined by (1.4) and (1.5) are increasing functions of σ_1 and σ_2

$$|f(s_1, s_2)|^2 = f(s_1, s_2)\overline{f(s_1, s_2)}$$

$$= \sum_{m,n=1}^{\infty} \left| a_{m,n} \right|^2 \exp\{2(\sigma_1 \lambda m + \sigma_2 \mu_n)\}$$

$$+\sum\nolimits_{m\neq M}\sum\nolimits_{n\neq N}a_{m,n}\bar{a}_{M,N}\exp\{\sigma_{1}(\lambda_{m}+\lambda_{M})+\sigma_{2}(\mu_{n}+\mu_{N})+it_{1}(\lambda_{m}-\lambda_{M})+it_{2}(\mu_{n}-\mu_{N})\}$$

The series on the right being absolutely convergent, the resulting series is uniformly convergent for any finite t_1 and t_2 range, therefore integrating term by term, the terms for which $m \ne M$, $n \ne N$, vanish as $T_1, T_2 \to \infty$ and we get

$$(1.6)I_2(\sigma_1, \sigma_2) = \sum_{m,n=1}^{\infty} |a_{m,n}|^2 \exp\{2(\sigma, \lambda_m + \sigma_2 \mu_n)\}$$

Hence $I_2(\sigma_1, \sigma_2)$ is an increasing function of σ_1 and σ_2 . Also from (1.4) and (1.5)

$$(1.7) \ m_{2,k}(\sigma_1,\sigma_2) = \frac{4}{e^{k\sigma_1}e^{k\sigma_2}} \int_0^{\sigma_1} \int_0^{\sigma_2} I_2(x_1,x_2) e^{kx_1} e^{kx_2} dx_1 dx_2$$

Using (1.6) we obtain

$$m_{2,k}(\sigma_1,\sigma_2) = \frac{4}{e^{k\sigma_1}e^{k\sigma_2}} \int_0^{\sigma_1} \int_0^{\sigma_2} \left[\sum_{m,n=1}^{\infty} \left| a_{m,n} \right|^2 \exp\{2(\sigma_1\lambda_m + \sigma_2\mu_n)\} e^{kx_1}e^{kx_2}dx_1 dx_2 \right]$$

$$=4\sum_{m\,n=1}^{\infty}\frac{\left|a_{m,n}\right|^{2}\left(e^{2\lambda}m^{\sigma_{1}}-e^{-k\sigma_{1}}\right)\left(e^{2\mu_{n}\sigma_{2}}-e^{-k\sigma_{2}}\right)}{(2\lambda_{m}+k)(2\mu_{n}+k)}$$

Thus $m_{2,k}(\sigma_1, \sigma_2)$ is also an increasing function of σ_1 and σ_2 .

We now prove following Theorems.

Theorem 1

 $Log I_2(\sigma_1, \sigma_2)$ is a convex function of σ_1 for a fixed value of σ_2 and Vice-versa.

Proof

Let I_2' , I_2'' denote partial derivatives of I_2 with respect to σ_1 . Then

$$\frac{\partial^2}{\partial \sigma_1^2}(\log I_2) = \frac{I_2 I_2'' - I_2' 2}{I_2^2},$$

And by Schwarz's Inequality

International Journal of Statistics and Applied Mathematics

$$I_2'^2 = \left[\Sigma |a_{m,n}|^2 2\lambda_m \exp\{2(\sigma_1 \lambda_m + \sigma_2 \mu_n)\} \right]$$

$$\leq \left[\sum |a_{m,n}|^2 \exp[2\{\sigma_1 \lambda_m + \sigma_2 \mu_n\}] X \right]$$

$$\left[\sum |a_{m,n}|^2 4\lambda_m^2 \exp[2\{\sigma_1 \lambda_m + \sigma_2 \mu_n\}] \right]$$

$$= I_2 I_2''$$

Hence the result and vice-versa.

Theorem 2

 $e^{k\sigma_1}I_2(\sigma_1,\sigma_2)$ is a convex function of $e^{k\sigma_1}m_{2,k}(\sigma_1,\sigma_2)$ for a fixed σ_2

Proof

Using (1.7) we have

$$\frac{\partial \{e^{k\sigma_1}I_2(\sigma_1,\sigma_2)\}}{\partial \{e^{k\sigma_1}m_{2,k}(\sigma_1,\sigma_2)\}} = \frac{\frac{\partial}{\partial \sigma_1}\{e^{k\sigma_1}I_2(\sigma_1,\sigma_2)\}}{\frac{\partial}{\partial \sigma_1}\{e^{k\sigma_1}m_{2,k}(\sigma_1,\sigma_2)\}}$$

$$= \frac{ke^{k\sigma_1}I_2(\sigma_1,\sigma_2) + e^{k\sigma_1}\frac{\partial}{\partial \sigma_1}\bar{I}_2(\sigma_{11}\sigma_2)}{\frac{\partial}{\partial \sigma_1}\Big[_{e^{k\sigma_1}e^{k\sigma_2}} \cdot e^{k\sigma_1}\int_0^{\sigma_1} \int_0^{\sigma_2} I_2(x_1,x_2)\Big\{e^{kx_1}e^{kx_2} dx_1 dx_2\Big\}}$$

$$=\frac{ke^{k\sigma_1}\overline{\bot}_2(\sigma_1,\sigma_2)+e^{k\sigma_1}\frac{\partial}{\partial\sigma_1}I_2(\sigma_1,\sigma_2)}{\frac{4}{e^{k\sigma_2}}e^{k\sigma_1}\int_0^{\sigma_2}I_2(\sigma_1,x_2)e^{kx_2}\,dx_2}$$

Since $I_2(\sigma_1, x_2)e^{kx_2}$ is an increasing function of σ_1, x_2 and therefore applying the second mean value theorem, we have

$$\frac{\partial \{e^{k\sigma_1}I_2(\sigma_1,\sigma_2)\}}{\partial \{e^{k\sigma_1}m_{2,k}(\sigma_1,\sigma_2)\}} = \frac{ke^{k\sigma_1}I_2(\sigma_1,\sigma_2) + e^{k\sigma_1}\frac{\partial}{\partial \sigma_1}I_2(\sigma_1,\sigma_2)}{\frac{4}{e^{k\sigma_2}}e^{k\sigma_1}I_2(\sigma_{1'}\sigma_2)e^{k\sigma_2}\int_{\xi}^{\sigma_2}dx_2}$$

Where $0 < \xi < \sigma_2$

$$=\frac{kI_2(\sigma_1,\sigma_2)+\frac{\partial}{\partial\sigma_1}I_2(\sigma_1,\sigma_2)}{4I_2(\sigma_1,\sigma_2)(\sigma_2-\xi)}$$

$$= \frac{1}{4(\sigma_2 - \xi)} \left\{ k + \frac{\left(\frac{\partial}{\partial \sigma_1}\right) \left(I_2(\sigma_1, \sigma_2)\right)}{I_2(\sigma_1, \sigma_2)} \right\}$$

The right hand side increases with σ_1 for a fixed σ_2 , follows from Theorem 1.

Therefore

$$\frac{\partial^2 \{e^{k\sigma_1}I_2(\sigma_1,\sigma_2)\}}{\partial^2 \{e^{k\sigma_1}m_2, \mu(\sigma_1,\sigma_2)\}} > 0$$
, for a fixed σ_2 . Hence the result

Theorem 3

 $e^{k\sigma_2} I_2(\sigma_1, \sigma_2)$ is a convex function of $e^{k\sigma_2} m_{2,k}(\sigma_1, \sigma_2)$ for a fixed σ_1 .

The proof is similar to Theorem 2.

Theorem 4

 $Log \ m_{2,k}(\sigma_1,\sigma_2)$ is a convex function of σ_1 for a fixed value of σ_2 and vice-versa

Proof

International Journal of Statistics and Applied Mathematics

Using (1.7), we have

$$\begin{split} \frac{\partial \{\log m_{2,k}(\sigma_1,\sigma_2)\}}{\partial \sigma_1} \\ &= \frac{\partial}{\partial \sigma_1} \{\log m_{2,k} \ (\sigma_1,\sigma_2)\} \\ &= \frac{4}{m_{2,k}(\sigma_1,\sigma_2)} [\frac{1}{e^{k\sigma_2}} \{-ke^{-k\sigma_1} \int\limits_0^{\sigma_1} \int\limits_0^{\sigma_2} I_2(x_1,x_2)e^{kx_1+kx_2} \, dx_1 \, dx_2 + \frac{1}{e^{k\sigma_1}} e^{k\sigma_1} \int\limits_0^{\sigma_2} I_2(\sigma_1,x_2)e^{kx_2} \, dx_2\}] \\ &= \frac{1}{m_{2,k}(\sigma_1,\sigma_2)} \{-km_{2,k}(\sigma_1,\sigma_2) + \frac{4}{e^{k\sigma_2}} \int_0^{\sigma_2} I_2(\sigma_1,x_2)e^{kx_2} \, dx_2\} \end{split}$$

Since $I_2(\sigma_1, x_2)e^{kx_2}$ is an increasing function of σ_1, x_2 and therefore applying the second mean-value theorem, we get

$$\begin{split} &\frac{\partial \{\log m_{2,k}(\sigma_{1},\sigma_{2})\}}{\partial(\sigma_{1})} \\ &= \frac{1}{m_{2,k}(\sigma_{1},\sigma_{2})} \Big\{ -km_{2,k}(\sigma_{1},\sigma_{2}) + \frac{4I_{2}(\sigma_{1},\sigma_{2})e^{k\sigma_{2}}}{e^{k\sigma_{2}}} \int_{\xi}^{\sigma_{2}} dx_{2} \}, \, 0 < \xi < \sigma_{2} \\ &= \frac{1}{m_{2,k}(\sigma_{1},\sigma_{2})} \Big\{ -km_{2,k}(\sigma_{1},\sigma_{2}) + 4(\sigma_{2} - \xi)I_{2}(\sigma_{1},\sigma_{2}) \Big\} \\ &= \Big\{ 4(\sigma_{2} - \xi) \frac{I_{2}(\sigma_{1},\sigma_{2})}{m_{2,k}(\sigma_{1},\sigma_{2})} - k \Big\} \end{split}$$

The right-hand side increases with σ_1 for a fixed σ_2 , follows from Theorem

Therefore,

 $\frac{\partial^2 \{log \, m_{2,k}(\sigma_1,\sigma_2)\}}{\partial \sigma_1^2} > 0$, for a fixed value of σ_2 . Hence the result and vice-versa.

References

- 1. Janusauskas AI. Elementary theorem on the convergence of doubleDirichlet Series Dokl. Akad. Nauk SSSR. 1977;234:1.
- 2. Sarkar PK. On order and type of an entire function in several complex variables represented by Dirichlet Series Bull. Cal. Math. Soc. 1982;74:99-11.