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## On the mean values of derivatives of entire functions of several complex variables represented by multiple Dirichlet series

**Dr. Vinita Vijai****Abstract**

In this paper we try to obtain lower bounds of  $m_{2,k}(\sigma_1, \sigma_2; f^{(r)})$ ,  $r = 1, 2$  in terms of  $m_{2,k}(\sigma_1, \sigma_2; f)$ , where  $I_2(\sigma_1, \sigma_2)$  and  $m_{2,k}(\sigma_1, \sigma_2)$  are defined as mean values of an entire function of two complex variables.

**Keywords:** Entire functions, several complex variables represented, multiple Dirichlet series

**Introduction**

Let us consider

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

$((s_j = \sigma_j + it_j), j = 1, 2)$  where  $a_{m,n} \in \mathbb{C}$ , the field of complex numbers,  $\lambda'_m s, \mu'_n s$  are real, and

$$0 < \lambda_1 < \lambda_2 < \dots < \lambda_m \rightarrow \infty;$$

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

It has been proved [1] that if

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

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

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

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  $\mathbb{C}^2$ . 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, pp.100]

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

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$$\mu(\sigma) = \mu_f(\sigma_1, \sigma_2) = \max_{(m,n) \in \mathbb{N}^2} \{|a_{m,n}| \exp(\sigma_1 \lambda_m + \sigma_2 \mu_n)\}$$

where  $\mathbb{N}$  is the set of natural numbers.

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

$$(1.4) I_2(\sigma_1, \sigma_2; f) = I_2(\sigma_1, \sigma_2) = \lim_{T_1, T_2 \rightarrow \infty} \frac{1}{4T_1 T_2} \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} |f(\sigma_1 + it_1, \sigma_2 + it_2)|^2 dt_1 dt_2$$

Now we define the mean value

$m_{2,k}(\sigma_1, \sigma_2)$  of  $|f(s_1, s_2)|$  as

$$(1.5) m_{2,k}(\sigma_1, \sigma_2; f) = m_{2,k}(\sigma_1, \sigma_2) \\ = \lim_{T_1, T_2 \rightarrow \infty} \frac{1}{T_1 T_2 e^{k\sigma_1} e^{k\sigma_2}} \int_0^{\sigma_1} \int_0^{\sigma_2} \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} \{|f(x_1 + it_1, x_2 + jt_2)|^2 e^{kx_1} e^{kx_2}\} dx_1 dx_2 dt_1 dt_2$$

where  $k$  is any positive number.

These mean values defined by (1.4) and (1.5) are increasing functions of  $\sigma_1$  and  $\sigma_2$

For

$$|f(s_1, s_2)|^2 = f(s_1, s_2) \overline{f(s_1, s_2)} \\ = \sum_{m_1, n=1}^{\infty} |a_{m,n}|^2 \exp\{2(\sigma_1 \lambda_m + \sigma_2 \mu_n)\} \\ + \sum_{m \neq M} \sum_{n \neq N} a_{m,n} \bar{a}_{M,N} \exp\{\sigma_1(\lambda_m + \lambda_M) + \sigma_2(\mu_n + \mu_N) + it_1(\lambda_m - \lambda_M) + it_2(\mu_n - \mu_N)\}$$

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  $m \neq M$ ,  $n \neq N$ , vanish as  $T_1, T_2 \rightarrow \infty$  and we get

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

Hence  $I_2(\sigma_1, \sigma_2)$  is an increasing function of  $\sigma_1$  and  $\sigma_2$ .

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

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

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

We now obtain a result on lower bound of  $m_{2,k}(\sigma_1, \sigma_2; f^{(r)})$ ,  $r = 1, 2$  in terms of

$m_{2,k}(\sigma_1, \sigma_2; f)$ , and any one of  $\sigma_1$  and  $\sigma_2$  while keeping the other fixed, where

$m_{2,k}(\sigma_1, \sigma_2; f^{(r)})$ , is the mean value of  $\left| \frac{\partial}{\partial s_r} f(s_1, s_2) \right|$ ,  $r = 1, 2$ ; that is

$$(2.1) m_{2,k}(\sigma_1, \sigma_2; f^{(1)}) = \lim_{T_1, T_2 \rightarrow \infty} \frac{1}{T_1 T_2 e^{k\sigma_1} e^{k\sigma_2}} \times$$

$$\int_0^{\sigma_1} \int_0^{\sigma_2} \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} \left\{ \left| \frac{\partial}{\partial x_1} f(x_1 + it_1, x_2 + it_2) \right|^2 \right\} \times \\ \times e^{kx_1} e^{kx_2} dx_1 dx_2 dt_1 dt_2$$

and similarly for  $m_{2,k}(\sigma_1, \sigma_2; f^{(2)})$

3. Lemma 1:  $e^{k\sigma_1} I_2(\sigma_1, \sigma_2)$  is a convex function of  $e^{k\sigma_1} m_{2,k}(\sigma_1, \sigma_2)$  for a fixed  $\sigma_2$

**Proof**

Using (1.7) we have

$$\frac{\partial \{e^{k\sigma_1} I_2(\sigma_1, \sigma_2)\}}{\partial \{e^{k\sigma_1} m_{2,k}(\sigma_1, \sigma_2)\}} \\ = \frac{\frac{\partial}{\partial \sigma_1} \{e^{k\sigma_1} I_2(\sigma_1, \sigma_2)\}}{\frac{\partial}{\partial \sigma_1} \{e^{k\sigma_1} m_{2,k}(\sigma_1, \sigma_2)\}} \\ = \frac{ke^{k\sigma_1} I_2(\sigma_1, \sigma_2) + e^{k\sigma_1} \frac{\partial}{\partial \sigma_1} I_2(\sigma_1, \sigma_2)}{\frac{\partial}{\partial \sigma_1} \left\{ \frac{4}{e^{k\sigma_1} e^{k\sigma_2}} \cdot e^{k\sigma_1} \int_0^{\sigma_1} \int_0^{\sigma_2} I_2(x_1, x_2) \{e^{kx_1} e^{kx_2} dx_1 dx_2\} \right\}} \\ = \frac{ke^{k\sigma_1} I_2(\sigma_1, \sigma_2) + e^{k\sigma_1} \frac{\partial}{\partial \sigma_1} I_2(\sigma_1, \sigma_2)}{\frac{4}{e^{k\sigma_2}} \cdot e^{k\sigma_1} \int_0^{\sigma_2} I_2(\sigma_1, x_2) e^{kx_2} dx_2}$$

Since  $I_2(\sigma_1, x_2)e^{kx_2}$  is an increasing function of  $\sigma_1, x_2$  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  $0 < \xi < \sigma_2$

$$= \frac{kI_2(\sigma_1, \sigma_2) + \frac{\partial}{\partial \sigma_1} I_2(\sigma_1, \sigma_2)}{4I_2(\sigma_1, \sigma_2)(\sigma_2 - \xi)} \\ = \frac{1}{4(\sigma_2 - \xi)} \left\{ k + \frac{\left(\frac{\partial}{\partial \sigma_1}\right)(I_2(\sigma_1, \sigma_2))}{I_2(\sigma_1, \sigma_2)} \right\}$$

The right hand side increases with  $\sigma_1$  for a fixed  $\sigma_2$ , from (1.6). Therefore

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

4. Lemma 2:  $\log m_{2,k}(\sigma_1, \sigma_2)$  is a convex function of  $\sigma_1$  for a fixed value of  $\sigma_2$  and vice-versa

**Proof:** Using (1.7), we have

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

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

$$\frac{\partial \{\log m_{2,k}(\sigma_1, \sigma_2)\}}{\partial (\sigma_1)}$$

$$\begin{aligned}
&= \frac{1}{m_{2,k}(\sigma_1, \sigma_2)} \left\{ -km_{2,k}(\sigma_1, \sigma_2) + \frac{4I_2(\sigma_1, \sigma_2)e^{k\sigma_2}}{e^{k\sigma_2}} \int_{\xi}^{\sigma_2} dx_2 \right\}, 0 < \xi < \sigma_2 \\
&= \frac{1}{m_{2,k}(\sigma_1, \sigma_2)} \left\{ -km_{2,k}(\sigma_1, \sigma_2) + 4(\sigma_2 - \xi)I_2(\sigma_1, \sigma_2) \right\} \\
&= \left\{ 4(\sigma_2 - \xi) \frac{I_2(\sigma_1, \sigma_2)}{m_{2,k}(\sigma_1, \sigma_2)} - k \right\}
\end{aligned}$$

The right-hand side increases with  $\sigma_1$  for a fixed  $\sigma_2$ , follows from Lemma

1. Therefore,

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

2. **Theorem 1:** If  $m_{2,k}(\sigma_1, \sigma_2; f^{(1)})$  is the mean value of  $|f^{(1)}(s_1, s_2)|$ , where  $f^{(1)}(s_1, s_2) = \frac{\partial}{\partial s_1} f(s_1, s_2)$ , then for a fixed value  $\sigma_2$ ,

$$(4.1) \quad m_{2,k}(\sigma_1, \sigma_2; f^{(1)}) \geq \frac{m_{2,k}(\sigma_1, \sigma_2)}{2^2} \left\{ \frac{\log m_{2,k}(\sigma_1, \sigma_2) - 1}{\sigma_1^2} \right\}^2$$

for  $\sigma_1 \geq \sigma_1^0$ , where  $\sigma_1^0$  is depending on the function of  $f$ .

**Proof:** We have

$$\begin{aligned}
&m_{2,k}(\sigma_1, \sigma_2; f^{(1)}) = \\
&\lim_{T_1, T_2 \rightarrow \infty} \frac{1}{T_1 T_2 e^{k\sigma_1} e^{k\sigma_2}} \int_0^{\sigma_1} \int_0^{\sigma_2} \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} |f^{(1)}(x_1 + it_1, x_2 + it_2)|^2 e^{kx_1} e^{kx_2} dx_1 dx_2 dt_1 dt_2 \\
&= \lim_{T_1, T_2 \rightarrow \infty} \frac{1}{T_1 T_2 e^{k\sigma_1} e^{k\sigma_2}} \times \\
&\int_0^{\sigma_1} \int_0^{\sigma_2} \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} \left| \lim_{\varepsilon \rightarrow 0} \frac{f(x_1 + it_1, x_2 + it_2) - f\{x_1(1-\varepsilon) + it_1, x_2 + it_2\}}{x_1 \varepsilon} \right|^2 \times \\
&\times e^{kx_1} e^{kx_2} dx_1 dx_2 dt_1 dt_2 \\
&\geq \lim_{\varepsilon \rightarrow 0} \lim_{T_1, T_2 \rightarrow \infty} \frac{1}{\varepsilon^2 \sigma_1^2 T_1 T_2 e^{k\sigma_1} e^{k\sigma_2}} \times \\
&\int_0^{\sigma_1} \int_0^{\sigma_2} \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} \left[ |f(x_1 + it_1, x_2 + it_2)| - |f\{x_1(1-\varepsilon) + it_1, x_2 + it_2\}| \right]^2 \times \\
&\times e^{kx_1} e^{kx_2} dx_1 dx_2 dt_1 dt_2
\end{aligned}$$

Applying Minkowski's inequality, we get

$$\begin{aligned}
&\left[ \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} \{ |f(x_1 + it_1, x_2 + it_2)| - |f(x_1(1-\varepsilon) + it_1, x_2 + it_2)| \}^2 dt_1 dt_2 \right]^2 \\
&\geq \left\{ \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} [|f(x_1 + it_1, x_2 + it_2)|^2 dt_1 dt_2] \right\}^{1/2} \\
&- \left\{ \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} [|f(x_1(1-\varepsilon) + it_1, x_2 + it_2)|^2 dt_1 dt_2] \right\}^{1/2}
\end{aligned}$$

Therefore,

$$\begin{aligned}
m_{2,k}(\sigma_1, \sigma_2; f^{(1)}) &\geq \lim_{\varepsilon \rightarrow 0} \lim_{T_1, T_2 \rightarrow \infty} \frac{1}{\varepsilon^2 \sigma_1^2 T_1 T_2 e^{k\sigma_1} e^{k\sigma_2}} \times \\
&\int_0^{\sigma_1} \int_0^{\sigma_2} \left\{ \left[ \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} |f(x_1 + it_1, x_2 + it_2)|^2 dt_1 dt_2 \right]^{1/2} \right. \\
&\left. - \left[ \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} |f(x_1(1-\varepsilon) + it_1, x_2 + it_2)|^2 dt_1 dt_2 \right]^{1/2} \right\}^2 e^{kx_1} e^{kx_2} dx_1 dx_2 \\
&= \lim_{\varepsilon \rightarrow 0} \lim_{T_1, T_2 \rightarrow \infty} \frac{1}{\varepsilon^2 \sigma_1^2 T_1 T_2 e^{k\sigma_1} e^{k\sigma_2}} \times
\end{aligned}$$

$$x \int_0^{\sigma_1} \int_0^{\sigma_2} [\{ e^{kx_1} e^{kx_2} \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} |f(x_1 + it_1, x_2 + it_2)|^2 dt_1 dt_2 \}^{1/2} - \{ e^{kx_1} e^{kx_2} \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} |f(x_1(1 - \varepsilon) + it_1, x_2 + it_2)|^2 dt_1 dt_2 \}^{1/2}]^2 dx_1 dx_2$$

Again, using Minkowski's inequality, we obtain

$$\begin{aligned} & m_{2,k}(\sigma_1, \sigma_2; f^{(1)}) \\ & \geq \lim_{\varepsilon \rightarrow 0} \lim_{T_1, T_2 \rightarrow \infty} \frac{1}{\varepsilon^2 \sigma_1^2 T_1 T_2 e^{k\sigma_1} e^{k\sigma_2}} \times \\ & x [\{ \int_0^{\sigma_1} \int_0^{\sigma_2} e^{kx_1} e^{kx_2} \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} |f(x_1 + it_1, x_2 + it_2)|^2 dx_1 dx_2 dt_1 dt_2 \}^{1/2} \\ & - \{ \int_0^{\sigma_1} \int_0^{\sigma_2} e^{kx_1} e^{kx_2} \int_{-T_1}^{T_1} \int_{-T_2}^{T_2} |f(x_1(1 - \varepsilon) + it_1, x_2 + it_2)|^2 dx_1 dx_2 dt_1 dt_2 \}^{1/2}]^2 \\ & \geq \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon^2 \sigma_1^2} [\{ m_{2,k}(\sigma_1, \sigma_2) \}^{1/2} \{ \frac{e^{-\varepsilon k \sigma_1}}{(1-\varepsilon)} e^{k(\sigma_1 - \sigma_1 \varepsilon)} (\frac{\varepsilon}{1-\varepsilon}) m_{2,k}(\sigma_1 - \sigma_1 \varepsilon, \sigma_2) \}^{1/2}]^2 \\ & = \lim_{\varepsilon \rightarrow 0} [ \frac{\{ m_{2,k}(\sigma_1, \sigma_2) \}^{1/2} - \{ (1-\varepsilon)^{-1} m_{2,k}(\sigma_1 - \sigma_1 \varepsilon, \sigma_2) \}^{1/2}}{\sigma_1 \varepsilon} ]^2 \end{aligned}$$

Let us now take,  $g(\sigma_1, \sigma_2) = \frac{\log m_{2,k}(\sigma_1, \sigma_2)}{\sigma_1}$

Then from Lemma 2,  $g(\sigma_1, \sigma_2)$  is an increasing function of  $\sigma_1$  for a fixed value of  $\sigma_2$  and so we have

$$\begin{aligned} m_{2,k}(\sigma_1, \sigma_2; f^{(1)}) & \geq \lim_{\varepsilon \rightarrow 0} [ \frac{e^{\sigma_1 g(\sigma_1, \sigma_2)/2 - (1-\varepsilon)^{-1/2} e^{(\sigma_1 - \sigma_1 \varepsilon) g(\sigma_1 - \sigma_1 \varepsilon, \sigma_2)/2}}{\sigma_1 \varepsilon} ]^2 \\ & \geq \lim_{\varepsilon \rightarrow 0} [ \frac{e^{\sigma_1 g(\sigma_1, \sigma_2)/2 - (1 + \frac{\varepsilon}{2} + \frac{3}{8} \varepsilon^2 + \dots) e^{(\sigma_1 - \sigma_1 \varepsilon) g(\sigma_1, \sigma_2)/2}}{\sigma_1 \varepsilon} ]^2 \\ & = \{ \frac{g(\sigma_1, \sigma_2)}{2} e^{\sigma_1 g(\sigma_1, \sigma_2)/2} - \frac{1}{2\sigma_1} e^{\sigma_1 g(\sigma_1, \sigma_2)/2} \}^2 \\ & = \frac{1}{2^2} [ \{ \sigma_1 g(\sigma_1, \sigma_2) e^{\sigma_1 g(\sigma_1, \sigma_2)/2} - e^{\sigma_1 g(\sigma_1, \sigma_2)/2} \}^2 ] \\ & = \frac{m_{2,k}(\sigma_1, \sigma_2)}{2^2} \left\{ \frac{\log m_{2,k}(\sigma_1, \sigma_2) - 1}{\sigma_1} \right\}^2 \end{aligned}$$

5. **Corollary 1:** If  $m_{2,k}(\sigma_1, \sigma_2; f^{(2)})$  is the mean value of  $|f^{(2)}(s_1, s_2)|$

where

$$f^{(2)}(s_1, s_2) = \frac{\partial}{\partial s_1} f(s_1, s_2) \text{ then for a fixed value of } \sigma_1$$

$$(5.1) \quad m_{2,k}(\sigma_1, \sigma_2; f^{(2)}) \geq \frac{m_{2,k}(\sigma_1, \sigma_2)}{2^2} \left\{ \frac{\log m_{2,k}(\sigma_1, \sigma_2) - 1}{\sigma_2^2} \right\}^2$$

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