Estimation of different entropies via Abel–Gontscharoff Green functions and Fink’s identity using Jensen type functionals

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Abstract
In this work, we estimated the different entropies like Shannon entropy, Rényi divergences, Csiszár divergence by using Jensen’s type functionals. The Zipf’s–Mandelbrot law and hybrid Zipf’s–Mandelbrot law are used to estimate the Shannon entropy. The Abel–Gontscharoff Green functions and Fink’s Identity are used to construct new inequalities and generalized them for m-convex function.

Keywords m-convex function, Jensen’s inequality, Shannon entropy, f- and Rényi divergence, Fink’s identity, Abel–Gontscharoff Green function, Entropy

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1. Introduction and preliminary results
In recent years many researchers generalized different inequalities using different identities involving green functions, for example in [24] Nasir et al. generalized the Popoviciu inequality using Montgomery identity along with the new green function. Also in [25] Niaz et al. used Fink’s identity along with new Abel–Gontscharoff type Green functions for ‘two point right focal’ to generalize the refinement of Jensen inequality.

The most commonly used words, the largest cities of countries, income of billionaire can be described in terms of Zipf’s law. The $f$-divergence means the distance between two probability distributions by making an average value, which is weighted by a specified function. As $f$-divergence, there are other probability distributions like Csiszár $f$-divergence [11,12], some special case of which is Kullback–Leibler-divergence used to find the appropriate distance between the probability distributions (see [20,21]). The notion of distance is stronger than divergence because it gives the properties of symmetry and triangle inequalities. Probability theory has application in many fields and the divergence between probability distribution has many applications in these fields.

Many natural phenomena like distribution of wealth and income in a society, distribution of face book likes, distribution of football goals follow power law distribution (Zipf’s Law). Like above phenomena, distribution of city sizes also follows power law distribution. Auerbach [3] first time gave the idea that the distribution of city size can be well approximated with the help of Pareto distribution (Power Law distribution). This idea was well refined by many researchers but Zipf [32] worked significantly in this field. The distribution of city sizes is investigated by many scholars of the urban economics, like Rosen and Resnick [29], Black and Henderson [4], Ioannides and Overman [19], Soo [30], Anderson and Ge [2] and Bosker et al. [5]. Zipf’s law states that: “The rank of cities with a certain number of inhabitants varies proportional to the city sizes with some negative exponent, say that is close to unity”. In other words, Zipf’s Law states that the product of city sizes and their ranks appear roughly constant. This indicates that the population of the second largest city is one half of the population of the largest city and the third largest city equal to the one third of the population of the largest city and the population of $n$th city is $\frac{1}{n}$ of the largest city population. This rule is called rank, size rule and also named as Zipf’s Law. Hence Zipf’s Law not only shows that the city size distribution follows the Pareto distribution, but also shows that the estimated value of the shape parameter is equal to unity.

In [18] L. Horváth et al. introduced some new functionals based on the $f$-divergence functionals and obtained some estimates for the new functionals. They obtained $f$-divergence and Rényi divergence by applying a cyclic refinement of Jensen’s inequality. They also construct some new inequalities for Rényi and Shannon entropies and used Zipf–Mandelbrot law to illustrate the results.

The inequalities involving higher order convexity are used by many physicists in higher dimension problems since the founding of higher order convexity by T. Popoviciu (see [27, p. 15]). It is quite interesting fact that there are some results that are true for convex functions but when we discuss them in higher order convexity they do not remain valid.

In [27, p. 16], the following criteria are given to check the $m$-convexity of the function.

If $f^{(m)}$ exists, then $f$ is $m$-convex if and only if $f^{(m)} \geq 0$.

In recent years many researchers have generalized the inequalities for $m$-convex functions; like S. I. Butt et al. generalized the Popoviciu inequality for $m$-convex function using Taylor's formula, Lidstone polynomial, Montgomery identity, Fink’s identity, Abel–Gontscharoff interpolation and Hermite interpolating polynomial (see [6–10]).

Since many years Jensen’s inequality has of great interest. The researchers have given the refinement of Jensen’s inequality by defining some new functions (see [16,17]). Like many researchers L. Horváth and J. Pečarić in [14,17], see also [15, p. 26], gave a refinement of Jensen’s inequality for convex function. They defined some essential notions to prove the refinement given as follows:
Let $X$ be a set, and:
$P(X) := \text{Power set of } X,$
$|X| := \text{Number of elements of } X,$
$\mathbb{N} := \text{Set of natural numbers with 0}.$

Consider $q \geq 1$ and $r \geq 2$ be fixed integers. Define the functions

$$F_{r,s} : \{1, \ldots, q\}^r \to \{1, \ldots, q\}^{r-1} \quad 1 \leq s \leq r,$$

$$F_r : \{1, \ldots, q\}^r \to P(\{1, \ldots, q\}^{r-1}),$$

and

$$T_r : P(\{1, \ldots, q\}^r) \to P(\{1, \ldots, q\}^{r-1}),$$

by

$$F_{r,s}(i_1, \ldots, i_r) := (i_1, i_2, \ldots, i_{s-1}, i_{s+1}, \ldots, i_r) \quad 1 \leq s \leq r,$$

$$F_r(i_1, \ldots, i_r) = \bigcup_{s=1}^{r} \{F_{r,s}(i_1, \ldots, i_r)\},$$

and

$$T_r(I) = \begin{cases} \phi, & I = \phi; \\ \bigcup_{(i_1, \ldots, i_r) \in I} F_r(i_1, \ldots, i_r), & I \neq \phi. \end{cases}$$

Next let the function

$$\alpha_{r,i} : \{1, \ldots, q\}^r \to \mathbb{N} \quad 1 \leq i \leq q$$

defined by

$$\alpha_{r,i}(i_1, \ldots, i_r) \text{ is the number of occurrences of } i \text{ in the sequence } (i_1, \ldots, i_r).$$

For each $I \in P(\{1, \ldots, q\}^r)$ let

$$\alpha_I := \sum_{(i_1, \ldots, i_r) \in I} \alpha_{r,i}(i_1, \ldots, i_r) \quad 1 \leq i \leq q.$$

$(H_1)$ Let $n, m$ be fixed positive integers such that $n \geq 1$, $m \geq 2$ and let $I_m$ be a subset of $\{1, \ldots, n\}^m$ such that

$$\alpha_{m,i} \geq 1 \quad 1 \leq i \leq n.$$

Introduce the sets $I_l \subset \{1, \ldots, n\}^l \ (m-1 \geq l \geq 1)$ inductively by

$$I_{l-1} := T_l(I_l) \quad m \geq l \geq 2.$$

Obviously the sets $I_1 = \{1, \ldots, n\}$, by $(H_1)$ and this insures that $\alpha_{1,i} = 1 (1 \leq i \leq n)$. From

$(H_1)$ we have $\alpha_{r,i} \geq (m-1 \geq l \geq 1, 1 \leq i \leq n).$

For $m \geq l \geq 2$ and for any $(j_1, \ldots, j_{l-1}) \in I_{l-1}$, let

$$\mathcal{H}_l(j_1, \ldots, j_{l-1}) := \{(i_1, \ldots, i_l) \times \{1, \ldots, l\} | F_{l,s}(i_1, \ldots, i_l) = (j_1, \ldots, j_{l-1})\}.$$
With the help of these sets they define the functions \( \eta_{m,l} : I_l \to \mathbb{N} \) by:

\[
\eta_{m,l}(i_1, \ldots, i_m) := \begin{cases} 1 & (i_1, \ldots, i_m) \in I_l; \\ 0 & \text{otherwise}. \end{cases}
\]

They define some special expressions for \( 1 \leq l \leq m \), as follows:

\[
\mathcal{A}_{m,l} = \mathcal{A}_{m,l}(I_m, x_1, \ldots, x_n, p_1, \ldots, p_n; f) := \frac{(m-1)!}{(l-1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{m,l}(i_1, \ldots, i_l)
\]

\[
\times \left( \sum_{j=1}^l \frac{p_{i_j}}{\mathcal{A}_{m,j}} \right) f \left( \frac{\sum_{j=1}^l \frac{p_{i_j}}{\mathcal{A}_{m,j}} x_{i_j}}{\sum_{j=1}^l \frac{p_{i_j}}{\mathcal{A}_{m,j}}} \right)
\]

and prove the following theorem.

**Theorem 1.1.** Assume \((H_1)\), and let \( f : I \to \mathbb{R} \) be a convex function where \( I \subset \mathbb{R} \) is an interval. If \( x_1, \ldots, x_n \in I \) and \( p_1, \ldots, p_n \) are positive real numbers such that \( \sum_{s=1}^n p_s = 1 \), then

\[
f \left( \sum_{s=1}^n p_s x_s \right) \leq \mathcal{A}_{m,m} \leq \mathcal{A}_{m,m-1} \leq \cdots \leq \mathcal{A}_{m,2} \leq \mathcal{A}_{m,1} = \sum_{s=1}^n p_s f(x_s).
\]

We define the following functionals by taking the differences of refinement of Jensen’s inequality given in (1).

\[
\Theta_1(f) = \mathcal{A}_{m,r} - f \left( \sum_{s=1}^n p_s x_s \right), \quad r = 1, \ldots, m,
\]

\[
\Theta_2(f) = \mathcal{A}_{m,r} - \mathcal{A}_{m,k}, \quad 1 \leq r < k \leq m.
\]

Under the assumptions of Theorem 1.1, we have

\[
\Theta_i(f) \geq 0, \quad i = 1, 2.
\]

Inequalities (4) are reversed if \( f \) is concave on \( I \).

In [26], the green function \( G : [\alpha_1, \alpha_2] \times [\alpha_1, \alpha_2] \to \mathbb{R} \) is defined as

\[
G(u, v) = \begin{cases} \frac{(u - \alpha_2)(v - \alpha_1)}{\alpha_2 - \alpha_1}, & \alpha_1 \leq v \leq u; \\ \frac{(v - \alpha_2)(u - \alpha_1)}{\alpha_2 - \alpha_1}, & u \leq v \leq \alpha_2. \end{cases}
\]

The function \( G \) is convex with respect to \( v \) and due to symmetry also convex with respect to \( u \). One can also note that \( G \) is continuous function.

In [31] it is given that any function \( f : [\alpha_1, \alpha_2] \to \mathbb{R} \) such that \( f \in C^2([\alpha_1, \alpha_2]) \) can be written as

\[
f(u) = \frac{\alpha_2 - u}{\alpha_2 - \alpha_1} f(\alpha_1) + \frac{u - \alpha_1}{\alpha_2 - \alpha_1} f(\alpha_2) + \int_{\alpha_1}^{\alpha_2} G(u, v) f''(v) dv.
\]
2. Inequalities for Csiszár divergence

In [11,12] Csiszár introduced the following notion.

**Definition 1.** Let \( f : \mathbb{R}^+ \to \mathbb{R}^+ \) be a convex function, let \( \mathbf{r} = (r_1, \ldots, r_n) \) and \( \mathbf{q} = (q_1, \ldots, q_n) \) be positive probability distributions. Then \( f \)-divergence functional is defined by

\[
I_f(\mathbf{r}, \mathbf{q}) := \sum_{i=1}^n q_i f\left( \frac{r_i}{q_i} \right).
\]  

And he stated that by defining

\[
f(0) := \lim_{x \to 0^+} f(x); \quad f\left( \frac{0}{0} \right) := 0; \quad f\left( \frac{a}{0} \right) := \lim_{x \to 0^+} x f\left( \frac{a}{x} \right), \quad a > 0,
\]

we can also use the nonnegative probability distributions as well.

In [18], L. Horváth, et al. gave the following functional based on the previous definition.

**Definition 2.** Let \( I \subset \mathbb{R} \) be an interval and let \( f : I \to \mathbb{R} \) be a function, let \( \mathbf{r} = (r_1, \ldots, r_n) \in \mathbb{R}^n \) and \( \mathbf{q} = (q_1, \ldots, q_n) \in (0, \infty)^n \) such that

\[
\frac{r_s}{q_s} \in I, \quad s = 1, \ldots, n.
\]

Then they define the sum \( \widehat{I}_f(\mathbf{r}, \mathbf{q}) \) as

\[
\widehat{I}_f(\mathbf{r}, \mathbf{q}) := \sum_{s=1}^n q_s f\left( \frac{r_s}{q_s} \right).
\]  

We apply Theorem 1.1 to \( \widehat{I}_f(\mathbf{r}, \mathbf{q}) \)

**Theorem 2.1.** Assume \((H_1)\), let \( I \subset \mathbb{R} \) be an interval and let \( \mathbf{r} = (r_1, \ldots, r_n) \) and \( \mathbf{q} = (q_1, \ldots, q_n) \) are in \((0, \infty)^n\) such that

\[
\frac{r_s}{q_s} \in I, \quad s = 1, \ldots, n.
\]

(i) If \( f : I \to \mathbb{R} \) is a convex function, then

\[
\widehat{I}_f(\mathbf{r}, \mathbf{q}) = \sum_{i=1}^n q_i f\left( \frac{r_i}{q_i} \right) = A_{m,1}^{[1]} \geq A_{m,2}^{[1]} \geq \cdots \geq A_{m,m-1}^{[1]} \geq A_{m,m}^{[1]}
\]

\[
\geq f\left( \frac{\sum_{s=1}^n r_s}{\sum_{s=1}^n q_s} \right) \sum_{s=1}^n q_s.
\]  

where

\[
A_{m,l}^{[1]} = \frac{(m-1)!}{(l-1)!} \sum_{(i_1, \ldots, i_l) \in \Pi_l} \eta_m(i_1, \ldots, i_l) \left( \sum_{j=1}^l q_{i_j} a_{m,i_j} \right) \left( \sum_{j=1}^l \frac{r_{i_j}}{a_{m,i_j}} \right).
\]  

If \( f \) is a concave function, then inequality signs in (10) are reversed.
(ii) If $f : I \to \mathbb{R}$ is a function such that $x \to xf(x)(x \in I)$ is convex, then

$$
\left( \sum_{s=1}^{n} r_s \right) f \left( \sum_{s=1}^{n} \frac{r_s}{q_s} \right) \leq A_{m,m}^{[2]} \leq A_{m,m-1}^{[2]} \leq \cdots \leq A_{m,1}^{[2]} \leq A_{m,1}^{[2]}
$$

where

$$
A_{m,l}^{[2]} = \frac{(m-1)!}{(l-1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{I_m,l}(i_1, \ldots, i_l) \left( \sum_{j=1}^{l} \frac{q_{i_j}}{\alpha_{m,i_j}} \right) \left( \sum_{j=1}^{l} \frac{r_{i_j}}{q_{i_j}} \right) \times f \left( \sum_{j=1}^{l} \frac{r_{i_j}}{q_{i_j}} \right)
$$

Proof. (i) Consider $p_s = \frac{q_s}{r_s}$ and $x_s = \frac{x}{q_s}$ in Theorem 1.1, we have

$$
f \left( \sum_{s=1}^{n} \frac{q_s}{r_s} \right) \leq \cdots \leq \frac{(m-1)!}{(l-1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{I_m,l}(i_1, \ldots, i_l)
$$

and taking the sum $\sum_{s=1}^{n} q_s$ we have (10).

(ii) Using $f := idf$ (where “idf” is the identity function) in Theorem 1.1, we have

$$
\sum_{s=1}^{n} p_s x_s f \left( \sum_{s=1}^{n} p_s x_s \right) \leq \cdots \leq \frac{(m-1)!}{(l-1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{I_m,l}(i_1, \ldots, i_l)
$$

and taking the sum $\sum_{s=1}^{n} q_s$ we have (10). Now on using $p_s = \frac{q_s}{\sum_{s=1}^{n} q_s}$ and $x_s = \frac{x}{q_s}$, $s = 1, \ldots, n$, we get
\[
\sum_{s=1}^{n} q_s \sum_{j=1}^{l} \frac{r_s}{q_s} f \left( \sum_{j=1}^{l} \frac{\eta_{is,j}}{\alpha_{is,j}} \right) \leq \cdots \leq \frac{(m-1)!}{(l-1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{is,l}(i_1, \ldots, i_l)
\]

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\[
\frac{n}{\sum_{s=1}^{n} q_s} r_s f \left( \sum_{s=1}^{n} \frac{q_s}{r_s} \right) \leq \frac{n}{\sum_{s=1}^{n} q_s} \frac{r_s}{q_s} f \frac{(r_s)}{q_s}
\]

On taking sum \( \sum_{s=1}^{n} q_s \) on both sides, we get (12). \( \square \)

3. Inequalities for Shannon Entropy

Definition 3 (See [18]). The Shannon entropy of positive probability distribution \( \mathbf{r} = (r_1, \ldots, r_n) \) is defined by

\[
S := - \sum_{s=1}^{n} r_s \log(r_s). \tag{16}
\]

Corollary 3.1. Assume \((H_l)\).

(i) If \( \mathbf{q} = (q_1, \ldots, q_n) \in (0, \infty)^n \), and the base of \( \log \) is greater than 1, then

\[
S \leq A_{m,m}^{[3]} \leq A_{m, m-1}^{[3]} \leq \cdots \leq A_{m,2}^{[3]} \leq A_{m,1}^{[3]} = \log \left( \frac{n}{\sum_{s=1}^{n} q_s} \right) \sum_{s=1}^{n} q_s, \tag{17}
\]

where

\[
A_{m,j}^{[3]} = \frac{(m-1)!}{(l-1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{m,l}(i_1, \ldots, i_l) \left( \sum_{j=1}^{l} \frac{q_i}{\alpha_{is,j}} \right) \log \left( \sum_{j=1}^{l} \frac{q_i}{\alpha_{is,j}} \right). \tag{18}
\]

If the base of \( \log \) is between 0 and 1, then inequality signs in (17) are reversed.

(ii) If \( \mathbf{q} = (q_1, \ldots, q_n) \) is a positive probability distribution and the base of \( \log \) is greater than 1, then we have the estimates for the Shannon entropy of \( \mathbf{q} \)

\[
S \leq A_{m,m}^{[4]} \leq A_{m, m-1}^{[4]} \leq \cdots \leq A_{m,2}^{[4]} \leq A_{m,1}^{[4]} = \log(n), \tag{19}
\]

where

\[
A_{m,j}^{[4]} = \frac{(m-1)!}{(l-1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{m,l}(i_1, \ldots, i_l) \left( \sum_{j=1}^{l} \frac{q_i}{\alpha_{is,j}} \right) \log \left( \sum_{j=1}^{l} \frac{q_i}{\alpha_{is,j}} \right). \]
Proof. (i) Using \( f := \log \) and \( r = (1, \ldots, 1) \) in Theorem 2.1 (i), we get (17).

(ii) It is the special case of (i). \( \square \)

Definition 4 (See [18])
The Kullback–Leibler divergence between the positive probability distribution \( r = (r_1, \ldots, r_n) \) and \( q = (q_1, \ldots, q_n) \) is defined by

\[
D(r, q) := \sum_{s=1}^{n} r_s \log \left( \frac{r_s}{q_s} \right). \tag{20}
\]

Corollary 3.2. Assume (H_1).

(i) Let \( r = (r_1, \ldots, r_n) \in (0, \infty)^n \) and \( q := (q_1, \ldots, q_n) \in (0, \infty)^n \). If the base of log is greater than 1, then

\[
\sum_{s=1}^{n} r_s \log \left( \frac{\sum_{s=1}^{n} r_s}{\sum_{s=1}^{n} q_s} \right) \leq A_{m,m}^{[5]} \leq A_{m,m-1}^{[5]} \leq \cdots \leq A_{m,2}^{[5]} \leq A_{m,1}^{[5]} \leq \cdots \leq A_{m,2}^{[5]} \leq A_{m,1}^{[5]} \leq \cdots \leq A_{m,2}^{[5]} \leq A_{m,1}^{[5]}
\]

\[
= \sum_{s=1}^{n} r_s \log \left( \frac{r_s}{q_s} \right) = D(r, q), \tag{21}
\]

where

\[
A_{m,l}^{[5]} = \frac{(m-1)!}{(l-1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{l,m}(i_1, \ldots, i_l) \left( \sum_{j=1}^{l} \frac{q_{i_j}}{a_{l,m,i_j}} \right) \times \log \left( \sum_{j=1}^{l} \frac{r_{i_j}}{a_{l,m,i_j}} \right). \]

If the base of \( \log \) is between 0 and 1, then inequality in (21) is reversed.

(ii) If \( r \) and \( q \) are positive probability distributions, and the base of \( l \) is greater than 1, then we have

\[
D(r, q) = A_{m,1}^{[6]} \geq A_{m,2}^{[6]} \geq \cdots \geq A_{m,m-1}^{[6]} \geq A_{m,m}^{[6]} \geq 0, \tag{22}
\]

where

\[
A_{m,l}^{[6]} = \frac{(m-1)!}{(l-1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{l,m}(i_1, \ldots, i_l) \left( \sum_{j=1}^{l} \frac{q_{i_j}}{a_{l,m,i_j}} \right) \times \log \left( \sum_{j=1}^{l} \frac{r_{i_j}}{a_{l,m,i_j}} \right)
\]

If the base of \( \log \) is between 0 and 1, then inequality signs in (22) are reversed.

Proof. (i) On taking \( f := \log \) in Theorem 2.1 (ii), we get (21).

(ii) Since \( r \) and \( q \) are positive probability distributions therefore \( \sum_{s=1}^{n} r_s = \sum_{s=1}^{n} q_s = 1 \), so the smallest term in (21) is given as

\[
\sum_{s=1}^{n} r_s \log \left( \frac{\sum_{s=1}^{n} r_s}{\sum_{s=1}^{n} q_s} \right) = 0. \tag{23}
\]

Hence for positive probability distribution \( r \) and \( q \) the (21) will become (22). \( \square \)
4. Inequalities for Rényi Divergence and Entropy

The Rényi divergence and entropy come from [28].

Definition 5. Let $\mathbf{r} = (r_1, \ldots, r_n)$ and $\mathbf{q} = (q_1, \ldots, q_n)$ be positive probability distributions, and let $\lambda \geq 0$, $\lambda \neq 1$.

(a) The Rényi divergence of order $\lambda$ is defined by

$$D_\lambda(\mathbf{r}, \mathbf{q}) := \frac{1}{\lambda - 1} \log \left( \sum_{j=1}^{n} q_j \left( \frac{r_j}{q_j} \right)^\lambda \right).$$  \hspace{1cm} (24)

(b) The Rényi entropy of order $\lambda$ of $\mathbf{r}$ is defined by

$$H_\lambda(\mathbf{r}) := \frac{1}{1 - \lambda} \log \left( \sum_{i=1}^{n} r_i^\lambda \right).$$  \hspace{1cm} (25)

The Rényi divergence and the Rényi entropy can also be extended to non-negative probability distributions. If $\lambda \to 1$ in (24), we have the Kullback–Leibler divergence, and if $\lambda \to 1$ in (25), then we have the Shannon entropy. In the next two results, inequalities can be found for the Rényi divergence.

Theorem 4.1. Assume $(H_1)$, let $\mathbf{r} = (r_1, \ldots, r_n)$ and $\mathbf{q} = (q_1, \ldots, q_n)$ are probability distributions.

(i) If $0 \leq \lambda \leq \mu$ such that $\lambda, \mu \neq 1$, and the base of $\log$ is greater than 1, then

$$D_\lambda(\mathbf{r}, \mathbf{q}) \leq A_{m,m}^{[7]} \leq A_{m,m-1}^{[7]} \leq \cdots \leq A_{m,2}^{[7]} \leq A_{m,1}^{[7]} = D_\mu(\mathbf{r}, \mathbf{q}),$$  \hspace{1cm} (26)

where

$$A_{m,j}^{[7]} = \frac{1}{\mu - 1} \log \left( \sum_{i_1, \ldots, i_j} \eta_{m,j}(i_1, \ldots, i_j) \left( \sum_{j=1}^{j} \frac{r_i}{\alpha_{m,i}} \right)^{\lambda - 1} \left( \sum_{j=1}^{j} \frac{r_j}{\alpha_{m,j}} \right)^{\mu - 1} \right).$$

The reverse inequalities hold in (26) if the base of $\log$ is between 0 and 1.

(ii) If $1 < \mu$ and the base of $\log$ is greater than 1, then

$$D_1(\mathbf{r}, \mathbf{q}) = D(\mathbf{r}, \mathbf{q}) = \sum_{i=1}^{n} r_i \log \left( \frac{r_i}{q_i} \right) \leq A_{m,m}^{[8]} \leq A_{m,m-1}^{[8]} \leq \cdots \leq A_{m,2}^{[8]} \leq A_{m,1}^{[8]} = D_\mu(\mathbf{r}, \mathbf{q}),$$  \hspace{1cm} (27)

where

$$A_{m,j}^{[8]} = \frac{1}{\mu - 1} \log \left( \sum_{i_1, \ldots, i_j} \eta_{m,j}(i_1, \ldots, i_j) \left( \sum_{j=1}^{j} \frac{r_i}{\alpha_{m,i}} \right)^{\mu - 1} \right).$$  \hspace{1cm} (28)

Here the base of $\exp$ is the same as the base of $\log$, and the reverse inequalities hold if the base of $\log$ is between 0 and 1.

(iii) If $0 \leq \lambda < 1$, and the base of $\log$ is greater than 1, then
\[ D_h(\mathbf{r}, \mathbf{q}) \leq A^{[9]}_{m,m} \leq A^{[9]}_{m,m-1} \leq \cdots \leq A^{[9]}_{m,2} \leq A^{[9]}_{m,1} = D_1(\mathbf{r}, \mathbf{q}), \]  

where

\[ A^{[9]}_{m,l} = \frac{1}{\lambda - 1} \frac{(m - 1)!}{(l - 1)!} \sum_{i_1, \ldots, i_l \in I} \eta_{m,l}(i_1, \ldots, i_l) \left( \sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}} \right)^\lambda \times \log \left( \frac{\sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}}}{\sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}}} \right)^{\lambda - 1} \]  

\[ D_1(\mathbf{r}, \mathbf{q}) = \frac{1}{\beta - 1} \sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}} \]  

**Proof.** By applying Theorem 1.1 with \( I = (0, \infty), f : (0, \infty) \to \mathbb{R}, f(t) = t^{\lambda-1} \),

\[ p_s := r_s, \quad x_s := \left( \frac{r_s}{q_s} \right)^\lambda, \quad s = 1, \ldots, n, \]

we have

\[ \left( \sum_{s=1}^n q_s \left( \frac{r_s}{q_s} \right)^\lambda \right)^{\frac{1}{\lambda}} \leq \left( \sum_{s=1}^n r_s \left( \frac{r_s}{q_s} \right)^\lambda \right)^{\frac{1}{\lambda - 1}} \leq \frac{\sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}}}{\sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}}} \left( \frac{\sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}}}{\sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}}} \right)^{\lambda - 1} \leq \cdots \leq \sum_{s=1}^n \left( \frac{r_s}{q_s} \right)^\lambda \leq \left( \sum_{s=1}^n q_s \left( \frac{r_s}{q_s} \right)^\lambda \right)^{\frac{1}{\lambda}} \]

if either \( 0 < \lambda < 1 < \beta \) or \( 1 < \lambda \leq \mu \), and the reverse inequality in (30) holds if \( 0 \leq \lambda \leq \beta < 1 \). By raising to power \( \frac{1}{\mu - \lambda} \), we have from all

\[ \left( \sum_{s=1}^n q_s \left( \frac{r_s}{q_s} \right)^\lambda \right)^{\frac{1}{\lambda}} \leq \left( \sum_{s=1}^n r_s \left( \frac{r_s}{q_s} \right)^\lambda \right)^{\frac{1}{\lambda - 1}} \leq \frac{\sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}}}{\sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}}} \left( \frac{\sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}}}{\sum_{j=1}^l \frac{r_{i_j}}{\alpha_{m,j}}} \right)^{\lambda - 1} \leq \cdots \leq \left( \sum_{s=1}^n r_s \left( \frac{r_s}{q_s} \right)^\lambda \right)^{\frac{1}{\lambda - 1}} \leq \left( \sum_{s=1}^n q_s \left( \frac{r_s}{q_s} \right)^\lambda \right)^{\frac{1}{\lambda}} \]

Since log is increasing if the base of log is greater than 1, it now follows (26). If the base of log is between 0 and 1, then log is decreasing and therefore inequality in (26) is reversed. If \( \lambda = 1 \) and \( \beta = 1 \), we have (ii) and (iii) respectively by taking limit, when \( \lambda \) goes to 1. \( \square \)

**Theorem 4.2.** Assume \((H_1)\), let \( \mathbf{r} = (r_1, \ldots, r_n) \) and \( \mathbf{q} = (q_1, \ldots, q_n) \) are probability distributions. If either \( 0 \leq \lambda < 1 \) and the base of log is greater than 1, or \( 1 < \lambda \) and the base of log is between 0 and 1, then
other cases can be proved similarly.

Since

\[
A_{m,1}^{[10]} \leq A_{m,2}^{[10]} \leq \cdots \leq A_{m,m-1}^{[10]} \leq A_{m,m}^{[10]} \leq D_{\lambda}(r, q) \leq A_{m,m}^{[11]} \]

where

\[
A_{m,m}^{[10]} = \frac{1}{(\lambda - 1)/2} \left( \frac{m - 1}{(l - 1)!} \sum_{(i_1, \ldots, i_l) \in I_i} \eta_{l,m,i}(i_1, \ldots, i_l) \right) \times \left( \sum_{j=1}^{l} r_{ij} \left( \frac{r_{ij}}{q_{ij}} \right)^{\lambda - 1} \right) \left( \sum_{j=1}^{l} r_{ij} \frac{r_{ij}}{q_{ij}} \right) \left( \sum_{j=1}^{l} r_{ij} \frac{r_{ij}}{q_{ij}} \right)
\]

and

\[
A_{m,m}^{[11]} = \frac{1}{\lambda - 1} \left( \frac{m - 1}{(l - 1)!} \sum_{(i_1, \ldots, i_l) \in I_i} \eta_{l,m,i}(i_1, \ldots, i_l) \right) \left( \sum_{j=1}^{l} r_{ij} \frac{r_{ij}}{q_{ij}} \right) \left( \sum_{j=1}^{l} r_{ij} \frac{r_{ij}}{q_{ij}} \right) \left( \sum_{j=1}^{l} r_{ij} \frac{r_{ij}}{q_{ij}} \right)
\]

The inequalities in (32) are reversed if either 0 \leq \lambda < 1 and the base of log is between 0 and 1, or 1 < \lambda and the base of l is greater than 1.

**Proof.** We prove only the case when 0 \leq \lambda < 1 and the base of log is greater than 1 and the other cases can be proved similarly. Since \( \frac{1}{\lambda - 1} < 0 \) and the function \( \log \) is concave then choose \( I = (0, \infty), f := \log, p_s = r_s, x_s := (r_s)^{\lambda - 1} \) in Theorem 1.1, we have

\[
D_{\lambda}(r, q) = \left( \frac{n}{\lambda - 1} \log \left( \sum_{s=1}^{n} q_s \left( \frac{r_s}{q_s} \right)^{\lambda} \right) \right) = \left( \frac{n}{\lambda - 1} \log \left( \sum_{s=1}^{n} r_s \left( \frac{r_s}{q_s} \right)^{\lambda - 1} \right) \right)
\]

\[
\leq \cdots \leq \left( \frac{n}{\lambda - 1} \log \left( \sum_{s=1}^{n} r_s \left( \frac{r_s}{q_s} \right)^{\lambda - 1} \right) \right)
\]

\[
\leq \cdots \leq \sum_{s=1}^{n} r_s \log \left( \frac{r_s}{q_s} \right) = D_1(r, q)
\]

and this gives the upper bound for \( D_{\lambda}(r, q) \).
Since the base of log is greater than 1, the function \( x \mapsto xf(x) \) \((x > 0)\) is convex therefore \( \frac{1}{1-x} < 0 \) and Theorem 1.1 gives

\[
D_h(\mathbf{r}, \mathbf{q}) = \frac{1}{\lambda - 1} \log \left( \sum_{s=1}^{n} q_s \left( \frac{r_s}{q_s} \right)^{\lambda} \right)
\]

\[
= \frac{1}{\lambda - 1} \log \left( \sum_{s=1}^{n} q_s \left( \frac{r_s}{q_s} \right)^{\lambda} \right) \log \left( \sum_{s=1}^{n} q_s \left( \frac{r_s}{q_s} \right)^{\lambda} \right)
\]

\[
\geq \cdots \geq \frac{1}{\lambda - 1} \left( \sum_{s=1}^{n} q_s \left( \frac{r_s}{q_s} \right)^{\lambda} \right) \left( \sum_{s=1}^{n} q_s \left( \frac{r_s}{q_s} \right)^{\lambda} \right)
\]

\[
= \frac{1}{\lambda - 1} \left( \sum_{s=1}^{n} q_s \left( \frac{r_s}{q_s} \right)^{\lambda} \right) \left( \sum_{s=1}^{n} q_s \left( \frac{r_s}{q_s} \right)^{\lambda} \right)
\]

which give the lower bound of \( D_h(\mathbf{r}, \mathbf{q}) \). \( \square \)

By using Theorems 4.1, 4.2 and Definition 5, some inequalities of Rényi entropy are obtained. Let \( \frac{1}{n} = \left( \frac{1}{n}, \ldots, \frac{1}{n} \right) \) be a discrete probability distribution.

**Corollary 4.3.** Assume \((H_1)\), let \( \mathbf{r} = (r_1, \ldots, r_n) \) and \( \mathbf{q} = (q_1, \ldots, q_n) \) are positive probability distributions.

(i) If \( 0 \leq \lambda \leq \mu, \lambda, \mu \neq 1 \), and the base of \( \log \) is greater than 1, then

\[
H_\lambda(\mathbf{r}) = \log(n) - D_h\left( \mathbf{r}, \frac{1}{n} \right) \geq A_{m,m}^{[12]} \geq A_{m,m}^{[12]} \geq \cdots A_{m,2}^{[12]} \geq A_{m,1}^{[12]} = H_\mu(\mathbf{r}),
\]

(35)
where
\[ A_{m,l}^{[12]} = \frac{1}{1 - \mu} \log \left( \frac{(m - 1)!}{(l - 1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{m,l}(i_1, \ldots, i_l) \times \left( \frac{\sum_{j=1}^{l} r_{ij}}{\sum_{j=1}^{l} \alpha_{m,i_j}} \right) \times \left( \frac{\sum_{j=1}^{l} r_{ij}^{\mu}}{\sum_{j=1}^{l} \alpha_{m,i_j}^{\mu}} \right)^{\frac{\mu-1}{\mu}} \right). \]

The reverse inequalities hold in (35) if the base of \( \log \) is between 0 and 1.

(ii) If \( 1 < \mu \) and base of \( \log \) is greater than 1, then
\[ S = -\sum_{i=1}^{n} p_i \log(p_i) \geq A_{m,m}^{[13]} \geq A_{m,m-1}^{[13]} \geq \cdots \geq A_{m,2}^{[13]} \geq A_{m,1}^{[13]} = H_\mu(r) \quad (36) \]
where
\[ A_{m,l}^{[13]} = \log(n) + \frac{1}{1 - \mu} \log \left( \frac{(m - 1)!}{(l - 1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{m,l}(i_1, \ldots, i_l) \times \left( \frac{\sum_{j=1}^{l} r_{ij}}{\sum_{j=1}^{l} \alpha_{m,i_j}} \right) \times \exp \left( \frac{(\mu - 1) \sum_{j=1}^{l} r_{ij} \log(nr_{ij})}{\sum_{j=1}^{l} r_{ij}} \right) \right). \]

The base of \( \exp \) is the same as the base of \( \log \). The inequalities in (36) are reversed if the base of \( \log \) is between 0 and 1.

(iii) If \( 0 \leq \lambda < 1 \), and the base of \( \log \) is greater than 1, then
\[ H_\lambda(r) \geq A_{m,m}^{[14]} \geq A_{m,m-1}^{[14]} \geq \cdots \geq A_{m,2}^{[14]} \leq A_{m,1}^{[14]} = S, \quad (37) \]
where
\[ A_{m,m}^{[14]} = \frac{1}{1 - \lambda} \frac{(m - 1)!}{(l - 1)!} \sum_{(i_1, \ldots, i_l) \in I_l} \eta_{m,l}(i_1, \ldots, i_l) \times \left( \frac{\sum_{j=1}^{l} r_{ij}}{\sum_{j=1}^{l} \alpha_{m,i_j}} \right) \times \log \left( \frac{\sum_{j=1}^{l} r_{ij}^{\lambda}}{\sum_{j=1}^{l} \alpha_{m,i_j}^{\lambda}} \right). \quad (38) \]

The inequalities in (37) are reversed if the base of \( \log \) is between 0 and 1.

**Proof.** (i) Suppose \( q = \frac{1}{n} \) then from (24), we have
\[ D_h(r, q) = \frac{1}{\lambda - 1} \log \left( \sum_{s=1}^{n} n^{\lambda - 1} \rho_s^{\lambda} \right) = \log(n) + \frac{1}{\lambda - 1} \log \left( \sum_{s=1}^{n} \rho_s^{\lambda} \right), \quad (39) \]

therefore we have
\[ H_h(r) = \log(n) - D_h \left( r, \frac{1}{n} \right). \quad (40) \]
Now using Theorem 4.1 (i) and (40), we get
\[ H_\lambda(r) = \log(n) - D_\lambda \left( r, \frac{1}{n} \right) \geq \cdots \geq \log(n) - \frac{1}{\mu - 1} \]

\[ \times \log \left( \frac{n^{\mu - 1}(m - 1)!}{(\mu - 1)!} \sum_{(i_1, \ldots, i_l) \in \mathcal{L}} \eta_{m,l}(i_1, \ldots, i_l) \times \left( \sum_{j=1}^{l} \frac{r_{ij}}{\alpha_{m,j}} \right) \left( \frac{\sum_{j=1}^{l} r_{ij}}{\sum_{j=1}^{l} \alpha_{m,j}} \right) \right) \]

\[ \geq \cdots \geq \log(n) - D_\mu(r, q) = H_\mu(r), \]

(ii) and (iii) can be proved similarly. □

**Corollary 4.4.** Assume (H₁) and let \( r = (r_1, \ldots, r_n) \) and \( q = (q_1, \ldots, q_n) \) are positive probability distributions.

If either \( 0 \leq \lambda < 1 \) and the base of log is greater than 1, or \( 1 < \lambda \) and the base of log is between 0 and 1, then

\[ -\frac{1}{\lambda} \sum_{j=1}^{n} r_j^\lambda \log(r_j) = A[^{15}]_{m,1} \geq A[^{15}]_{m,2} \geq \cdots \geq A[^{15}]_{m, m-1} \geq A[^{15}]_{m, m} \]

\[ \geq H_\lambda(r) \geq A[^{16}]_{m, m} \geq A[^{16}]_{m, m-1} \geq \cdots A[^{16}]_{m, 2} \geq A[^{16}]_{m, 1} = H(r), \]

where

\[ A[^{15}]_{m,j} = \frac{1}{(\lambda - 1) \sum_{s=1}^{n} r_s^{\lambda} (m - 1)!} \sum_{(i_1, \ldots, i_l) \in \mathcal{L}_j} \eta_{m,l}(i_1, \ldots, i_l) \left( \sum_{j=1}^{l} \frac{r_{ij}}{\alpha_{m,j}} \right) \log \left( \frac{\sum_{j=1}^{l} r_{ij}}{\sum_{j=1}^{l} \alpha_{m,j}} \right) \]

and

\[ A[^{16}]_{m,1} = \frac{1}{1 - \lambda} \frac{(m - 1)!}{(l - 1)!} \sum_{(i_1, \ldots, i_l) \in \mathcal{L}_1} \eta_{m,l}(i_1, \ldots, i_l) \left( \sum_{j=1}^{l} \frac{r_{ij}}{\alpha_{m,j}} \right) \log \left( \frac{\sum_{j=1}^{l} r_{ij}}{\sum_{j=1}^{l} \alpha_{m,j}} \right). \]

The inequalities in (42) are reversed if either \( 0 \leq \lambda < 1 \) and the base of log is between 0 and 1, or \( 1 < \lambda \) and the base of log is greater than 1.

**Proof.** The proof is similar to Corollary 4.3 by using Theorem 4.2. □

5. Inequalities by using Zipf–Mandelbrot law

In probability theory and statistics, the Zipf–Mandelbrot law is a distribution. It is a power law distribution on ranked data, named after the linguist G. K. Zipf who suggests a simpler distribution called Zipf’s law. The Zipf’s law is defined as follows (see [32]).
**Definition 6.** Let $N$ be a number of elements, $s$ be their rank and $t$ be the value of exponent characterizing the distribution. Zipf’s law then predicts that out of a population of $N$ elements, the normalized frequency of element of rank $s$, $f(s, N, t)$ is

$$f(s, N, t) = \frac{1}{\sum_{j=1}^{N} \frac{1}{j^t}}.$$  

(43)

The Zipf–Mandelbrot law is defined as follows (see [22]).

**Definition 7.** Zipf–Mandelbrot law is a discrete probability distribution depending on three parameters $N \in \{1, 2, \ldots, \}$, $q \in [0, \infty)$ and $t > 0$, and is defined by

$$f(s; N, q, t) := \frac{1}{(s + q)^t H_{N,q,t}}, \quad s = 1, \ldots, N,$$

where

$$H_{N,q,t} = \sum_{j=1}^{N} \frac{1}{(j + q)^t}. \quad (45)$$

If the total mass of the law is taken over all $N$, then for $q \geq 0$, $t > 1$, $s \in \mathbb{N}$, density function of Zipf–Mandelbrot law becomes

$$f(s; q, t) = \frac{1}{(s + q)^t H_{q,t}},$$

where

$$H_{q,t} = \sum_{j=1}^{\infty} \frac{1}{(j + q)^t}. \quad (47)$$

For $q = 0$, the Zipf–Mandelbrot law (44) becomes Zipf’s law (43).

**Conclusion 5.1.** Assume $(H_1)$, let $\mathbf{r}$ be a Zipf–Mandelbrot law, by Corollary 4.3 (iii), we get: If $0 \leq \lambda < 1$, and the base of $\log$ is greater than 1, then

$$H_\lambda(\mathbf{r}) = \frac{1}{1 - \lambda} \log \left( \frac{1}{H_{N,q,t}^{\lambda}} \sum_{s=1}^{N} \frac{1}{(s + q)^{\lambda s}} \right) \geq \cdots \geq$$

$$\frac{1}{1 - \lambda} \frac{(m - 1)!}{(l - 1)!} \sum_{(i_1, \ldots, i_l) \in \mathbf{r}} \eta_{m,j}(i_1, \ldots, i_l) \left( \sum_{j=1}^{l} \frac{1}{\alpha_{m,j}(i_j + q) H_{N,q,t}} \right)$$

$$\times \log \left( \frac{1}{H_{N,q,t}^{\lambda - 1}} \sum_{j=1}^{l} \frac{1}{\alpha_{m,j}(i_j + q)^{\lambda - 1}} \right) \geq \cdots \geq$$

$$\frac{t}{H_{N,q,t}} \sum_{s=1}^{N} \frac{\log(s + q)}{(s + q)^t} + \log(H_{N,q,t}) = S.$$

The inequalities in (48) are reversed if the base of $\log$ is between 0 and 1.
Conclusion 5.2. Assume \( (H_1) \), let \( r_1 \) and \( r_2 \) be the Zipf–Mandelbrot law with parameters \( N \in \{1, 2, \ldots \}, q_1, q_2 \in [0, \infty) \) and \( s_1, s_2 > 0 \), respectively, then from Corollary 3.2 (ii), we have if the base of \( l \) is greater than 1, then

\[
\mathcal{D}(r_1, r_2) = \frac{1}{(s + q_1)^{\frac{1}{l}} H_{N, q_1, l_1}} \log \frac{(s + q_2)^{\frac{1}{l}} H_{N, q_2, l_1}}{(s + q_1)^{\frac{1}{l}} H_{N, q_2, l_1}} \geq \ldots
\]

\[
\geq \frac{(m - 1)!}{(l - 1)!} \sum_{(i_1, \ldots, i_l) \in \mathcal{I}} \eta_{l_1}(i_1, \ldots, i_l)
\]

\[
\times \left( \sum_{j=1}^{l} \frac{1}{(i_j + q_2)^{\frac{1}{l}} H_{N, q_2, l_1}} a_{i_1 j} \right) \left( \sum_{j=1}^{l} \frac{1}{(i_j + q_2)^{\frac{1}{l}} H_{N, q_2, l_1}} a_{i_2 j} \right) \left( \sum_{j=1}^{l} \frac{1}{(i_j + q_2)^{\frac{1}{l}} H_{N, q_2, l_1}} a_{i_3 j} \right) \ldots \geq \ldots \geq 0.
\]

(49)

The inequalities in (49) are reversed if the base of \( l \) is between 0 and 1.

6. Shannon entropy, Zipf–Mandelbrot law and hybrid Zipf–Mandelbrot law
Here we maximize the Shannon entropy using method of Lagrange multiplier under some equations constraints and get the Zipf–Mandelbrot law.

Theorem 6.1. If \( J = \{1, 2, \ldots, N\} \), for a given \( q \geq 0 \) a probability distribution that maximizes the Shannon entropy under the constraints

\[
\sum_{s \in J} r_s = 1, \sum_{s \in J} r_s (\ln(s + q)) = \psi,
\]

is Zipf–Mandelbrot law.

Proof. If \( J = \{1, 2, \ldots, N\} \), we set the Lagrange multipliers \( \lambda \) and \( t \) and consider the expression

\[
\hat{S} = -\sum_{s=1}^{N} r_s \ln r_s - \lambda \left( \sum_{s=1}^{N} r_s - 1 \right) - t \left( \sum_{s=1}^{N} r_s \ln(s + q) - \psi \right)
\]

Just for the sake of convenience, replace \( \lambda \) by \( \ln \lambda - 1 \), thus the last expression gives

\[
\hat{S} = -\sum_{s=1}^{N} r_s \ln r_s - (\ln \lambda - 1) \left( \sum_{s=1}^{N} r_s - 1 \right) - t \left( \sum_{s=1}^{N} r_s \ln(s + q) - \psi \right)
\]

From \( \hat{S}_{r_s} = 0 \), for \( s = 1, 2, \ldots, N \), we get

\[
r_s = \frac{1}{\lambda(s + q)^{l_1}}
\]

and on using the constraint \( \sum_{s=1}^{N} r_s = 1 \), we have

...
\[ \lambda = \sum_{s=1}^{N} \left( \frac{1}{(s+1)^{t}} \right) \]

where \( t > 0 \), concluding that

\[ r_s = \frac{1}{(s+q)^{1}} H_{N,q,t}, \quad s = 1, 2, \ldots, N. \]

**Remark 6.2.** Observe that the Zipf–Mandelbrot law and Shannon Entropy can be bounded from above (see [23]).

\[ S = -\sum_{s=1}^{N} f(s, N, q, t) \ln f(s, N, q, t) \leq -\sum_{s=1}^{N} f(s, N, q, t) \ln q_s \]

where \((q_1, \ldots, q_N)\) is a positive \(N\)-tuple such that \( \sum_{s=1}^{N} q_s = 1 \).

**Theorem 6.3.** If \( J = \{1, \ldots, N\} \), then probability distribution that maximizes Shannon entropy under constraints

\[ \sum_{s \in J} r_s := 1, \quad \sum_{s \in J} r_s \ln(s + q) := \Psi, \quad \sum_{s \in J} sr_s := \eta \]

is hybrid Zipf–Mandelbrot law given as

\[ r_s = \frac{w^s}{(s+q)^k \Phi^*(k, q, w)}, \quad s \in J, \]

where

\[ \Phi_J(k, q, w) = \sum_{s \in J} \frac{w^s}{(s+q)^k}. \]

**Proof.** First consider \( J = \{1, \ldots, N\} \), we set the Lagrange multiplier and consider the expression

\[ \tilde{S} = -\sum_{s=1}^{N} r_s \ln r_s + \ln w \left( \sum_{s=1}^{N} sr_s - \eta \right) - (\ln \lambda - 1) \left( \sum_{s=1}^{N} r_s - 1 \right) - k \left( \sum_{s=1}^{N} r_s \ln(s + q) - \Psi \right). \]

On setting \( \tilde{S}_{r_s} = 0 \), for \( s = 1, \ldots, N \), we get

\[ -\ln r_s + s \ln w - \ln \lambda - k \ln(s + q) = 0, \]

after solving for \( r_s \), we get \( \lambda = \sum_{s=1}^{N} \frac{w^s}{(s+q)^k} \), and we recognize this as the partial sum of Lerch’s transcendent that we will denote by

\[ \Phi^*_N(k, q, w) = \sum_{s=1}^{N} \frac{w^s}{(s+q)^k} \text{ with } w \geq 0, k > 0. \]

**Remark 6.4.** Observe that for Zipf–Mandelbrot law, Shannon entropy can be bounded from above (see [23]).
where \((q_1, \ldots, q_N)\) is any positive \(N\)-tuple such that \(\sum_{i=1}^{N} q_i = 1\).

Under the assumption of Theorem 2.1 (i), define the non-negative functionals as follows:
\[
\Theta_3(f) = A_{m,r}^{[1]} - f \left( \sum_{s=1}^{N} \frac{r_s}{\sum_{s=1}^{N} q_s} \right) \sum_{s=1}^{N} q_s, \quad r = 1, \ldots, m, \quad (50)
\]
\[
\Theta_4(f) = A_{m,r}^{[1]} - A_{m,k}^{[1]}, \quad 1 \leq r < k \leq m. \quad (51)
\]

Under the assumption of Theorem 2.1 (ii), define the non-negative functionals as follows:
\[
\Theta_5(f) = A_{m,r}^{[2]} - f \left( \sum_{s=1}^{N} \frac{r_s}{\sum_{s=1}^{N} q_s} \right) \sum_{s=1}^{N} q_s, \quad r = 1, \ldots, m, \quad (52)
\]
\[
\Theta_6(f) = A_{m,r}^{[2]} - A_{m,k}^{[2]}, \quad 1 \leq r < k \leq m. \quad (53)
\]

Under the assumption of Corollary 3.1 (i), define the following non-negative functionals
\[
\Theta_7(f) = A_{m,r}^{[3]} + \sum_{i=1}^{N} q_i \log(q_i), \quad r = 1, \ldots, n \quad (54)
\]
\[
\Theta_8(f) = A_{m,r}^{[3]} - A_{m,k}^{[3]}, \quad 1 \leq r < k \leq m. \quad (55)
\]

Under the assumption of Corollary 3.1 (ii), define the following non-negative functionals as
\[
\Theta_9(f) = A_{m,r}^{[4]} - S, \quad r = 1, \ldots, m \quad (56)
\]
\[
\Theta_{10}(f) = A_{m,r}^{[4]} - A_{m,k}^{[4]}, \quad 1 \leq r < k \leq m. \quad (57)
\]

Under the assumption of Corollary 3.2 (i), let us define the non-negative functionals as follows:
\[
\Theta_{11}(f) = A_{m,r}^{[5]} - \sum_{s=1}^{N} r_s \log \left( \sum_{i=1}^{N} \frac{r_n}{\sum_{s=1}^{N} q_s} \right), \quad r = 1, \ldots, m \quad (58)
\]
\[
\Theta_{12}(f) = A_{m,r}^{[5]} - A_{m,k}^{[5]}, \quad 1 \leq r < k \leq m. \quad (59)
\]

Under the assumption of Corollary 3.2 (ii), define the non-negative functionals as follows
\[
\Theta_{13}(f) = A_{m,r}^{[6]} - A_{m,k}^{[5]}, \quad 1 \leq r < k \leq m. \quad (60)
\]

Under the assumption of Theorem 4.1 (i), consider the following functionals
\[
\Theta_{14}(f) = A_{m,r}^{[7]} - D_r(r, q), \quad r = 1, \ldots, m \quad (61)
\]
\[
\Theta_{15}(f) = A_{m,r}^{[7]} - A_{m,k}^{[5]}, \quad 1 \leq r < k \leq m. \quad (62)
\]
Under the assumption of Theorem 4.1 (ii), consider the following functionals:

\[ \Theta_{16}(f) = A_{m,r}^{[8]} - D_{k}(r, q), \quad r = 1, \ldots, m \]  
(63)

\[ \Theta_{17}(f) = A_{m,r}^{[8]} - A_{m,k}^{[8]}, \quad 1 \leq r < k \leq m. \]  
(64)

Under the assumption of Theorem 4.1 (iii), consider the following functionals:

\[ \Theta_{18}(f) = A_{m,r}^{[9]} - D_{k}(r, q), \quad r = 1, \ldots, m \]  
(65)

\[ \Theta_{19}(f) = A_{m,r}^{[9]} - A_{m,k}^{[9]}, \quad 1 \leq r < k \leq m. \]  
(66)

Under the assumption of Theorem 4.2 consider the following non-negative functionals

\[ \Theta_{20}(f) = D_{k}(r, q) - A_{m,r}^{[10]}, \quad r = 1, \ldots, m \]  
(67)

\[ \Theta_{21}(f) = A_{m,k}^{[10]} - A_{m,r}^{[10]}, \quad 1 \leq r < k \leq m. \]  
(68)

\[ \Theta_{22}(f) = A_{m,r}^{[11]} - D_{k}(r, q), \quad r = 1, \ldots, m \]  
(69)

\[ \Theta_{23}(f) = A_{m,r}^{[11]} - A_{m,k}^{[11]}, \quad 1 \leq r < k \leq m. \]  
(70)

\[ \Theta_{24}(f) = A_{m,r}^{[11]} - A_{m,k}^{[10]}, \quad r = 1, \ldots, m, \quad k = 1, \ldots, m. \]  
(71)

Under the assumption of Corollary 4.3 (i), consider the following non-negative functionals

\[ \Theta_{25}(f) = H_{k}(r) - A_{m,r}^{[12]}, \quad r = 1, \ldots, m \]  
(72)

\[ \Theta_{26}(f) = A_{m,k}^{[12]} - A_{m,r}^{[12]}, \quad 1 \leq r < k \leq m. \]  
(73)

Under the assumption of Corollary 4.3 (ii), consider the following functionals

\[ \Theta_{27}(f) = S - A_{m,r}^{[13]}, \quad r = 1, \ldots, m \]  
(74)

\[ \Theta_{28}(f) = A_{m,k}^{[13]} - A_{m,r}^{[13]}, \quad 1 \leq r < k \leq m. \]  
(75)

Under the assumption of Corollary 4.3 (iii), consider the following functionals

\[ \Theta_{29}(f) = H_{k}(r) - A_{m,r}^{[14]}, \quad r = 1, \ldots, m \]  
(76)

\[ \Theta_{30}(f) = A_{m,k}^{[14]} - A_{m,r}^{[14]}, \quad 1 \leq r < k \leq m. \]  
(77)

Under the assumption of Corollary 4.4, define the following functionals

\[ \Theta_{31} = A_{m,r}^{[15]} - H_{k}(r), \quad r = 1, \ldots, m \]  
(78)

\[ \Theta_{32} = A_{m,r}^{[15]} - A_{m,k}^{[15]}, \quad 1 \leq r < k \leq m. \]  
(79)

\[ \Theta_{33} = H_{k}(r) - A_{m,r}^{[16]}, \quad r = 1, \ldots, m \]  
(80)
Figure 1 shows the graph of Green functions \( G_i(z, w), i = 1, 2, 3, 4 \) defined in (86)–(89) respectively for fixed value of \( w \). They also introduced some new Abel–Gontscharoff type identities by using these new Green functions in the following lemma.

**Lemma A.** Let \( f : [\alpha_1, \alpha_2] \to \mathbb{R} \), where \( [\alpha_1, \alpha_2] \) be an interval, is a function such that \( f^{(n-1)} \) is absolutely continuous then the following identity holds

\[
f(z) = f(\alpha_1) + (z - \alpha_1)f'(\alpha_2) + \int_{\alpha_1}^{\alpha_2} G_1(z, w)f''(w)dw,
\]

where

\[
G_1(z, w) = \begin{cases} 
\alpha_1 - w, & \alpha_1 \leq w \leq \alpha_2; \\
\alpha_1 - z, & z \leq w \leq \alpha_2.
\end{cases}
\]

In [8], S. I. Butt et al. gave some new types of Green functions defined as

\[
G_2(z, w) = \begin{cases} 
\alpha_2 - z, & \alpha_1 \leq w \leq \alpha_2; \\
\alpha_2 - w, & z \leq w \leq \alpha_2,
\end{cases}
\]

\[
G_3(z, w) = \begin{cases} 
z - \alpha_1, & \alpha_1 \leq w \leq \alpha_2; \\
w - \alpha_1, & z \leq w \leq \alpha_2,
\end{cases}
\]

\[
G_4(z, w) = \begin{cases} 
\alpha_2 - w, & \alpha_1 \leq w \leq \alpha_2; \\
\alpha_2 - z, & z \leq w \leq \alpha_2,
\end{cases}
\]
Theorem 7.1. Assume (H1), and let $f : I = [\alpha_1, \alpha_2] \to \mathbb{R}$ be a function such that for $m \geq 3$ (an integer) $f^{(m-1)}$ is absolutely continuous. Also, let $x_1, \ldots, x_n \in I, p_1, \ldots, p_n$ be positive real numbers such that $\sum_{i=1}^n p_i = 1$. Assume that $F_{\alpha_i}^m, G_k (k = 1, 2, 3, 4)$ and $\Theta_i (i = 1, \ldots, 35)$ are the same as defined in (84), (86)–(89), (2), (3), (50)–(82) respectively. Then:

1. For $k = 1, 3, 4$ we have the following identities:

\[
\Theta_i(f) = (m - 2) \left( \frac{f'(\alpha_2) - f'(\alpha_1)}{\alpha_2 - \alpha_1} \right) \int_{\alpha_1}^{\alpha_2} \Theta_i(G_k(\cdot, w)) dw + \frac{1}{\alpha_2 - \alpha_1} \int_{\alpha_1}^{\alpha_2} \Theta_i(G_k(\cdot, w)) dw \\
\times \sum_{\lambda=1}^{m-3} \left( \frac{m - 2 - \lambda}{\lambda!} \right) (f^{(\lambda+1)}(\alpha_2)(w - \alpha_2)^{\lambda} - f^{(\lambda+1)}(\alpha_1)(w - \alpha_1)^{\lambda}) dw \\
+ \frac{1}{(m-3)!} (\alpha_2 - \alpha_1) \int_{\alpha_1}^{\alpha_2} f^{(m)}(\zeta) \\
\times \left( \int_{\alpha_1}^{\alpha_2} \Theta_i(G_k(\cdot, w))(w - \zeta)^{m-3} F_{\alpha_i}^m(G_k(\cdot, w)) dw \right) d\zeta, \quad i = 1, \ldots, 35.
\]
(2) For $k = 2$ we have

\[
\Theta_i(f) = (-1)(m - 2) \left( \frac{f'(a_2) - f'(a_1)}{a_2 - a_1} \right) \int_{a_1}^{a_2} \Theta_i(G_k(\cdot, w)) \, dw \\
+ \frac{(-1)}{m} \int_{a_1}^{a_2} \Theta_i(G_k(\cdot, w)) \times \sum_{k=1}^{m-3} \left( \frac{m - 2 - \lambda}{\lambda!} \right) (f^{(k+1)}(a_2)(w - a_2)^k \\
- f^{(k+1)}(a_1)(w - a_1)^k) \, dw \frac{(-1)}{(m-3)!} (a_2 - a_1) \int_{a_1}^{a_2} f^{(m)}(\zeta) \\
\times \left( \int_{a_1}^{a_2} \Theta_i(G_k(\cdot, w))(w - \zeta)^{m-3} F_{a_1}^{a_2}(\zeta, w) \, d\zeta \right) d\zeta,
\]

for $i = 1, \ldots, 35$.

**Proof.** (i) Using Abel–Gontsharoff-type identities (85), (91), (92) in $\Theta_i(f)$, $i = 1, \ldots, 35$, and using properties of $\Theta_i(f')$, we get

\[
\Theta_i(f) = \int_{a_1}^{a_2} \Theta_i(G_k(\cdot, w)) f''(w) \, dw, \quad i = 1, 2. \tag{95}
\]

From identity (83), we get

\[
f'(w) = (m - 2) \left( \frac{f'(a_2) - f'(a_1)}{a_2 - a_1} \right) + \sum_{k=1}^{m-3} \left( \frac{m - 2 - \lambda}{\lambda!} \right) \\
\times \left( \frac{f^{(k)}(a_2)(w - a_2)^{k-1} - f^{(k)}(a_1)(w - a_2)^{k-1}}{\lambda!} \right) \left( \frac{1}{a_2 - a_1} \right) \int_{a_1}^{a_2} (w - \zeta)^{m-3} F_{a_1}^{a_2}(\zeta, w) f^{(m)}(\zeta) \, d\zeta. \tag{96}
\]

Using (95) and (96) and applying Fubini’s theorem we get the result (93) for $k = 1, 3, 4$.

(ii) Substituting Abel–Gontsharoff-type inequality (90) in $\Theta_i(f)$, $i = 1, \ldots, 35$, and following similar steps to (i), we get (94). \(\square\)

**Theorem 7.2.** Assume (H1), and let $f : I = [a_1, a_2] \to \mathbb{R}$ be a function such that for $m \geq 3$ (an integer) $f^{(m-1)}$ is absolutely continuous. Also, let $x_1, \ldots, x_n \in I$, $p_1, \ldots, p_n$ are positive real numbers such that $\sum_{i=1}^{n} p_i = 1$. Assume that $F_{a_1}^{a_2}, G_k (k = 1, 2, 3, 4)$ and $\Theta_i (i = 1, 2)$ are the same as defined in (84), (86)–(89), (2), (3), (50)–(52) respectively. For $m \geq 3$ assume that

\[
\int_{a_1}^{a_2} \Theta_i(G_k(\cdot, \zeta))(w - \zeta)^{m-3} F_{a_1}^{a_2}(\zeta, w) \, dw \geq 0, \zeta \in [a_1, a_2], \quad i = 1, \ldots, 35, \tag{97}
\]

for $k = 1, 3, 4$. If $f$ is an $m$-convex function, then
(i) For $k = 1, 3, 4$, the following holds:

$$
\Theta_i(f) \geq (m - 2) \left( \frac{f''(\alpha_2) - f''(\alpha_1)}{\alpha_2 - \alpha_1} \right) \int_{a_i}^{a_2} \Theta_i(G_k(\cdot, w))dw \\
+ \frac{1}{\alpha_2 - \alpha_1} \int_{a_i}^{a_2} \Theta_i(G_k(\cdot, w)) \times \sum_{\lambda=1}^{m-3} \left( \frac{m - 2 - \lambda}{\lambda!} \right) f^{(\lambda+1)}(\alpha_2)(w - \alpha_2)^\lambda \\
- f^{(\lambda+1)}(\alpha_1)(w - \alpha_1)^\lambda dw, \\
i = 1, \ldots, 35. 
$$

(98)

(ii) For $k = 2$, we have

$$
\Theta_i(f) \leq (-1)(m - 2) \left( \frac{f''(\alpha_2) - f''(\alpha_1)}{\alpha_2 - \alpha_1} \right) \int_{a_i}^{a_2} \Theta_i(G_k(\cdot, w))dw \\
+ \frac{(-1)}{\alpha_2 - \alpha_1} \int_{a_i}^{a_2} \Theta_i(G_k(\cdot, w)) \times \sum_{\lambda=1}^{m-3} \left( \frac{m - 2 - \lambda}{\lambda!} \right) f^{(\lambda+1)}(\alpha_2)(w - \alpha_2)^\lambda \\
- f^{(\lambda+1)}(\alpha_1)(w - \alpha_1)^\lambda dw, \\
i = 1, \ldots, 35. 
$$

(99)

**Proof.** (i) Since $f^{(m-1)}$ is absolutely continuous on $[\alpha_1, \alpha_2]$, $f^{(m)}$ exists almost everywhere. Also, since $f$ is $m$-convex therefore we have $f^{(m)}(\zeta) \geq 0$ for a.e. on $[\alpha_1, \alpha_2]$. So, applying Theorem 1.1, we obtain (98).

(ii) Similar to (i). $\square$

**Remark A.** We can investigate the bounds for the identities related to the generalization of refinement of Jensen inequality using inequalities for the Čebyšev functional and some results relating to the Grüss and Ostrowski type inequalities can be constructed as given in Section 3 of [6]. Also we can construct the non-negative functionals from inequalities (98)–(99) and give related mean value theorems and we can construct the new families of $m$-exponentially convex functions and Cauchy means related to these functionals as given in Section 4 of [6].

**References**


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