

# A NEW NON-SYMMETRIC INFORMATION DIVERGENCE OF CSISZAR'S CLASS, PROPERTIES AND ITS BOUNDS

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## Abstract

Non-parametric measures give the amount of information supplied by the data for discriminating in favor of a probability distribution  $P$  against another  $Q$ , or for measuring the distance or affinity between  $P$  and  $Q$ .

There are several generalized functional divergences, such as: Csiszar divergence, Renyi-like divergence, Bregman divergence, Burbea- Rao divergence etc. all. In this paper, a non-parametric non symmetric measure of divergence which belongs to the family of Csiszar's  $f$ -divergence is proposed. Its properties are studied and get the bounds in terms of some well known divergence measures.

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## 1. INTRODUCTION

Let  $\Gamma_n = \left\{ P = (p_1, p_2, p_3, \dots, p_n) : p_i > 0, \sum_{i=1}^n p_i = 1 \right\}$ ,  $n \geq 2$

be the set of all complete finite discrete probability distributions. If we take  $p_i \geq 0$  for some  $i = 1, 2, 3, \dots, n$ ,

then we have to suppose that  $0f(0) = 0f\left(\frac{0}{0}\right) = 0$ .

Csiszar [2], given the generalized  $f$ -divergence measure, which is given by:

$$C_f(P, Q) = \sum_{i=1}^n q_i f\left(\frac{p_i}{q_i}\right) \quad (1.1)$$

Where  $f : (0, \infty) \rightarrow \mathbb{R}$  (set of real no.) is real, continuous and convex function and  $P = (p_1, p_2, p_3, \dots, p_n)$ ,  $Q = (q_1, q_2, q_3, \dots, q_n) \in \Gamma_n$ , where  $p_i$  and  $q_i$  are probability mass functions. Many known divergences can be obtained from this generalized measure by suitably defining the convex function  $f$ . Some of those are as follows:

$$\diamond \quad K(P, Q) = \sum_{i=1}^n p_i \log\left(\frac{p_i}{q_i}\right) = \text{Kullback-Leibler divergence measure [4]} \quad (1.2)$$

$$\diamond \quad \chi^2(P, Q) = \sum_{i=1}^n \frac{(p_i - q_i)^2}{q_i} = \text{Chi-Square divergence measure [5]} \quad (1.3)$$

$$\diamond \quad h(P, Q) = \sum_{i=1}^n \frac{(\sqrt{p_i} - \sqrt{q_i})^2}{2} = \text{Hellinger discrimination [3]} \quad (1.4)$$

$$\diamond \quad R_a(P, Q) = \sum_{i=1}^n \frac{p_i^a}{q_i^{a-1}}, a > 1 = \text{Renyi's "a" order entropy [6]} \quad (1.5)$$

$$\diamond \quad B(P, Q) = \sum_{i=1}^n \sqrt{p_i q_i} = \text{Bhattacharya divergence measure [1]} \quad (1.6)$$

**Relative information of type "s" [9]**

$$\diamond \quad \Phi_s(P, Q) = [s(s-1)]^{-1} \left[ \sum_{i=1}^n p_i^s q_i^{1-s} - 1 \right], s \neq 0, 1 \text{ and } s \in \mathbb{R} \quad (1.7)$$

Particularly

$$\lim_{s \rightarrow 1} \Phi_s(P, Q) = K(P, Q), \lim_{s \rightarrow 0} \Phi_s(P, Q) = K(Q, P) \tag{1.8}$$

Where  $K(P, Q)$  is given by (1.2).

$$\diamond G(P, Q) = \sum_{i=1}^n \left( \frac{p_i + q_i}{2} \right) \log \left( \frac{p_i + q_i}{2 p_i} \right) = \text{Relative AG Divergence [7]} \tag{1.9}$$

Similarly, we get many others divergences as well by defining suitable convex function

## 2. NEW INFORMATION DIVERGENCE MEASURE

In this section, we shall obtain a new divergence measure corresponding to new convex function, and will study the properties.

The following theorem is well known in literature [2].

**Theorem 1:** If the function  $f$  is convex and normalized, i.e.,  $f(1) = 0$ , then  $C_f(P, Q)$  and its ad joint  $C_f(Q, P)$  are both non-negative and convex in the pair of probability distribution  $(P, Q) \in \Gamma_n \times \Gamma_n$ .

Let  $f: (0, \infty) \rightarrow \mathbb{R}$ , be a mapping, defined as:

$$f(t) = \frac{(t-1)^4}{t}, t \in (0, \infty) \tag{2.1}$$

And

$$f'(t) = \frac{(t-1)^3}{t^2} (3t+1), f''(t) = \frac{2(t-1)^2}{t^3} (3t^2+2t+1) \tag{2.2}$$

Properties of function defined by (2.1), are as follows:

- ❖ Since  $f''(t) \geq 0 \forall t \in (0, \infty) \Rightarrow f(t)$  is a convex function.
- ❖ Since  $f(1) = 0 \Rightarrow f(t)$  is a normalized function.
- ❖ Since  $f'(t) < 0$  at  $(0, 1)$  and  $f'(t) > 0$  at  $(1, \infty) \Rightarrow f(t)$  is monotonically decreasing in  $(0, 1)$  and monotonically increasing in  $(1, \infty)$ , and  $f'(1) = 0$ .

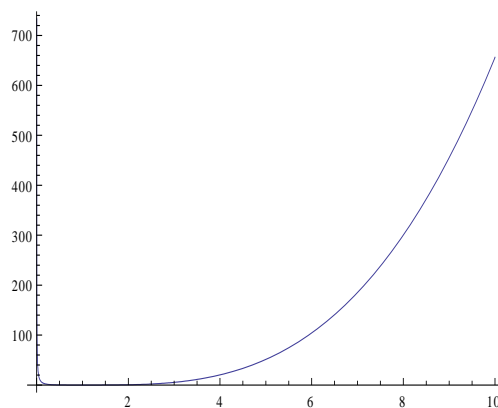


Fig1: Graph of convex function  $f(t)$

Now, put (2.1) in (1.1), we get the following new divergence:

$$C_f(P, Q) = V^*(P, Q) = \sum_{i=1}^n \frac{(p_i - q_i)^4}{p_i q_i^2} \tag{2.3}$$

Properties of divergence defined by (2.3), are as follows:

- ❖ In view of theorem 1, we can say that  $V^*(P, Q) > 0$  and convex in the pair of probability distribution  $(P, Q) \in \Gamma_n \times \Gamma_n$ .
- ❖  $V^*(P, Q) = 0$  if  $P = Q$  or  $p_i = q_i$  (Attains its minimum value).
- ❖ Since  $V^*(P, Q) \neq V^*(Q, P) \Rightarrow V^*(P, Q)$  is non-symmetric divergence measure w.r.t.  $P$  &  $Q$ .

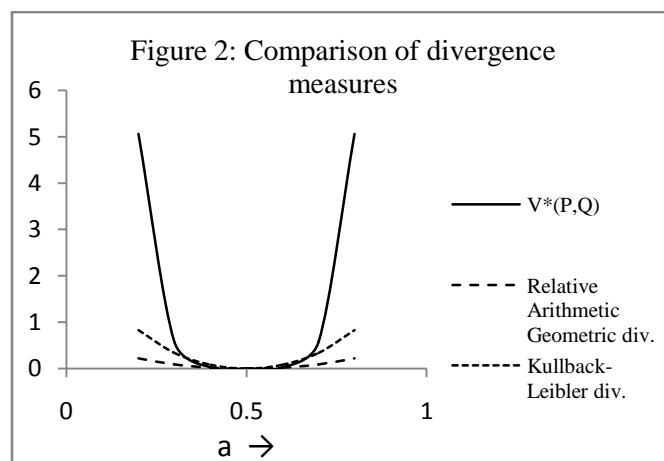


Figure 2 shows the behavior of  $V^*(P, Q)$ , Relative Arithmetic-Geometric divergence  $G(P, Q)$  and Kullback-Leibler divergence  $K(P, Q)$ . We have considered  $p_i = (a, 1-a)$  and  $q_i = (1-a, a)$  where  $a \in (0, 1)$ . It

is clear from figure 2 that the new measure  $V^*(P, Q)$  has a steeper slope than  $G(P, Q)$  and  $K(P, Q)$ .

### 3. CSISZAR'S FUNCTIONAL DIVERGENCE AND INEQUALITIES

The following theorem is well known in literature [8].

**Theorem 2:** Let  $f : I \subset R_+ \rightarrow R$  (I is an open interval) be a mapping which is normalized, i.e.,  $f(1) = 0$  and suppose that

I.  $f$  is twice differentiable on  $(\alpha, \beta)$ ,  $0 < \alpha \leq 1 \leq \beta < \infty$  with  $\alpha \neq \beta$ .

II. There exist real constants  $m, M$  such that  $m < M$  and  $m \leq t^{2-s} f''(t) \leq M \forall t \in (\alpha, \beta)$  and  $s \in R$  and

If  $P, Q \in \Gamma_n$  with  $0 < \alpha \leq \frac{P_i}{Q_i} \leq \beta < \infty \forall i = 1, 2, 3, \dots, n$ , then

$$m \Phi_s(P, Q) \leq C_f(P, Q) \leq M \Phi_s(P, Q) \quad (3.1)$$

And

$$m[\eta_s(P, Q) - \Phi_s(P, Q)] \leq C_\rho(P, Q) - C_f(P, Q) \leq M[\eta_s(P, Q) - \Phi_s(P, Q)] \quad (3.2)$$

Where

$$C_\rho(P, Q) = C_{f'}\left(\frac{P^2}{Q}, P\right) - C_{f'}(P, Q) = \sum_{i=1}^n (p_i - q_i) f'\left(\frac{p_i}{q_i}\right) \quad (3.3)$$

$$\eta_s(P, Q) = C_{\Phi_s'}\left(\frac{P^2}{Q}, P\right) - C_{\Phi_s'}(P, Q) = (s-1)^{-1} \sum_{i=1}^n (p_i - q_i) \left(\frac{p_i}{q_i}\right)^{s-1}, s \neq 1 \quad (3.4)$$

And  $C_f(P, Q), \Phi_s(P, Q)$  are given by (1.1) and (1.7) respectively

### 4. BOUNDS OF NEW INFORMATION DIVERGENCE MEASURE

In this section, we derive bounds for  $V^*(P, Q)$  in terms of the well known divergences in the following propositions at  $s = 2, 1, 1/2, 0$  and  $-1$ , by using the theorem 2.

#### 4.1 Proposition 4.1(at s=2)

Let  $\chi^2(P, Q)$  and  $V^*(P, Q)$  be defined as in (1.3) and (2.3) respectively. Then, we have

i. If  $0 < \alpha < 1$ , then

$$0 \leq V^*(P, Q) \leq \max\left\{\frac{(\alpha-1)^2}{\alpha^3}(3\alpha^2+2\alpha+1), \frac{(\beta-1)^2}{\beta^3}(3\beta^2+2\beta+1)\right\} \chi^2(P, Q)$$

$$0 \leq V_\rho^*(P, Q) - V^*(P, Q) \quad (4.1)$$

$$\leq \max\left\{\frac{(\alpha-1)^2}{\alpha^3}(3\alpha^2+2\alpha+1), \frac{(\beta-1)^2}{\beta^3}(3\beta^2+2\beta+1)\right\} \chi^2(P, Q) \quad (4.2)$$

ii. If  $\alpha = 1$ , then

$$0 \leq V^*(P, Q) \leq \frac{(\beta-1)^2}{\beta^3}(3\beta^2+2\beta+1) \chi^2(P, Q) \quad (4.3)$$

$$0 \leq V_\rho^*(P, Q) - V^*(P, Q) \leq \frac{(\beta-1)^2}{\beta^3}(3\beta^2+2\beta+1) \chi^2(P, Q) \quad (4.4)$$

**Proof:**

Firstly, put  $s=2$  in (1.7) and (3.4) respectively, we get

$$\Phi_s(P, Q) = \frac{1}{2} \sum_{i=1}^n \frac{p_i^2}{q_i} - 1 = \frac{1}{2} \sum_{i=1}^n \frac{p_i^2}{q_i} - 2p_i + q_i = \frac{1}{2} \sum_{i=1}^n \frac{(p_i - q_i)^2}{q_i} = \frac{1}{2} \chi^2(P, Q) \quad (4.5)$$

$$\eta_s(P, Q) = \sum_{i=1}^n (p_i - q_i) \frac{p_i}{q_i} = \sum_{i=1}^n \frac{p_i^2}{q_i} - p_i = \sum_{i=1}^n \frac{p_i^2}{q_i} - 1$$

$$= \sum_{i=1}^n \frac{p_i^2}{q_i} - 2p_i + q_i = \sum_{i=1}^n \frac{(p_i - q_i)^2}{q_i} = \chi^2(P, Q) \quad (4.6)$$

And by putting  $f'(t)$  in (3.3), we get

$$V_\rho^*(P, Q) = V_{f'}^*\left(\frac{P^2}{Q}, P\right) - V_{f'}^*(P, Q) = \sum_{i=1}^n \frac{(p_i - q_i)^4}{(p_i q_i)^2} (3p_i + q_i) \quad (4.7)$$

Let  $g(t) = f''(t) = \frac{2(t-1)^2}{t^3}(3t^2+2t+1)$  (After putting  $s=2$  in  $t^{2-s} f''(t)$ )

Then  $g'(t) = \frac{6(t^4 - 1)}{t^4}, g''(t) = \frac{24}{t^5}$

If  $g'(t) = 0 \Rightarrow t^4 - 1 = 0 \Rightarrow t = 1, -1$

It is clear that  $g(t)$  is monotonic decreasing on  $(0, 1)$  and monotonic increasing on  $[1, \infty)$ .

Also  $g(t)$  has minimum value at  $t=1$ , since  $g''(1) = 24 > 0$  so

$$m = \inf_{t \in (0, \infty)} g(t) = g(1) = 0 \tag{4.8}$$

Now, we have two cases:

i. If  $0 < \alpha < 1$ , then

$$M = \sup_{t \in (\alpha, \beta)} g(t) = \max \{g(\alpha), g(\beta)\} \\ = \max \left\{ \frac{2(\alpha-1)^2}{\alpha^3} (3\alpha^2 + 2\alpha + 1), \frac{2(\beta-1)^2}{\beta^3} (3\beta^2 + 2\beta + 1) \right\} \tag{4.9}$$

ii. If  $\alpha = 1$ , then

$$M = \sup_{t \in [1, \beta)} g(t) = \frac{2(\beta-1)^2}{\beta^3} (3\beta^2 + 2\beta + 1) \tag{4.10}$$

The results (4.1), (4.2), (4.3) and (4.4) are obtained by using (2.3), (4.5), (4.6), (4.7), (4.8), (4.9) and (4.10) in 3.1 and 3.2.

**4.2 Proposition 4.2(at s=1)**

Let  $K(P, Q)$  and  $V^*(P, Q)$  be defined as in (1.2) and (2.3) respectively. Then, we have

i. If  $0 < \alpha < 1$ , then

$$0 \leq V^*(P, Q) \leq \max \left\{ \frac{2(\alpha-1)^2}{\alpha^2} (3\alpha^2 + 2\alpha + 1), \frac{2(\beta-1)^2}{\beta^2} (3\beta^2 + 2\beta + 1) \right\} K(P, Q) \tag{4.11}$$

$$0 \leq V_\rho^*(P, Q) - V^*(P, Q) \\ \leq \max \left\{ \frac{2(\alpha-1)^2}{\alpha^2} (3\alpha^2 + 2\alpha + 1), \frac{2(\beta-1)^2}{\beta^2} (3\beta^2 + 2\beta + 1) \right\} K(Q, P) \tag{4.12}$$

ii. If  $\alpha = 1$ , then

$$0 \leq V^*(P, Q) \leq \frac{2(\beta-1)^2}{\beta^2} (3\beta^2 + 2\beta + 1) K(P, Q) \tag{4.13}$$

$$0 \leq V_\rho^*(P, Q) - V^*(P, Q) \leq \frac{2(\beta-1)^2}{\beta^2} (3\beta^2 + 2\beta + 1) K(Q, P) \tag{4.14}$$

**Proof:**

Firstly, put  $s=1$  in (1.7) and (3.4) respectively, we get

$$\lim_{s \rightarrow 1} \Phi_s(P, Q) = \sum_{i=1}^n p_i \log \left( \frac{p_i}{q_i} \right) = K(P, Q) \tag{4.15}$$

$$\lim_{s \rightarrow 1} \eta_s(P, Q) = \sum_{i=1}^n p_i \log \left( \frac{p_i}{q_i} \right) + q_i \log \left( \frac{q_i}{p_i} \right) = K(P, Q) + K(Q, P) \tag{4.16}$$

Let  $g(t) = t f''(t) = \frac{2(t-1)^2}{t^2} (3t^2 + 2t + 1)$  (After putting  $s=1$  in  $t^{2-s} f''(t)$ )

Then

$$g'(t) = \frac{4(t-1)}{t^3} (3t^3 + t^2 + t + 1), g''(t) = \frac{12(t^4 + 1)}{t^4}$$

If

$$g'(t) = 0 \Rightarrow (t-1)(3t^3 + t^2 + t + 1) = 0 \Rightarrow t = 1, -0.63$$

It is clear that  $g(t)$  is monotonic decreasing on  $(0, 1)$  and monotonic increasing on  $[1, \infty)$ .

Also  $g(t)$  has minimum value at  $t=1$ , since  $g''(1) = 24 > 0$  so

$$m = \inf_{t \in (0, \infty)} g(t) = g(1) = 0 \tag{4.17}$$

Now, we have two cases:

$$M = \sup_{t \in (\alpha, \beta)} g(t) = \max \{g(\alpha), g(\beta)\}$$

$$= \max \left\{ \frac{2(\alpha-1)^2}{\alpha^2} (3\alpha^2 + 2\alpha + 1), \frac{2(\beta-1)^2}{\beta^2} (3\beta^2 + 2\beta + 1) \right\} \tag{4.18}$$

ii. If  $\alpha = 1$ , then

$$M = \sup_{t \in [1, \beta)} g(t) = \frac{2(\beta-1)^2}{\beta^2} (3\beta^2 + 2\beta + 1) \tag{4.19}$$

The results (4.11), (4.12), (4.13) and (4.14) are obtained by using (2.3), (4.7), (4.15), (4.16), (4.17), (4.18), and (4.19) in 3.1 and 3.2.

#### 4.3 Proposition 4.3(at s=1/2)

Let  $h(P, Q), R_a(P, Q), B(P, Q)$  and  $V^*(P, Q)$  be defined as in (1.4), (1.5), (1.6) and (2.3) respectively. Then, we have

i. If  $0 < \alpha < 1$ , then

$$0 \leq V^*(P, Q) \leq \max \left\{ \frac{8(\alpha-1)^2}{\alpha^{3/2}} (3\alpha^2 + 2\alpha + 1), \frac{8(\beta-1)^2}{\beta^{3/2}} (3\beta^2 + 2\beta + 1) \right\} h(P, Q) \quad (4.20)$$

$$0 \leq V_{\rho}^*(P, Q) - V^*(P, Q)$$

$$\leq \max \left\{ \frac{8(\alpha-1)^2}{\alpha^{3/2}} (3\alpha^2 + 2\alpha + 1), \frac{8(\beta-1)^2}{\beta^{3/2}} (3\beta^2 + 2\beta + 1) \right\} \left[ \frac{1}{2} \{R_{3/2}(Q, P) - B(P, Q)\} - h(P, Q) \right] \quad (4.21)$$

ii. If  $\alpha = 1$ , then

$$0 \leq V^*(P, Q) \leq \frac{8(\beta-1)^2}{\beta^{3/2}} (3\beta^2 + 2\beta + 1) h(P, Q) \quad (4.22)$$

$$0 \leq V_{\rho}^*(P, Q) - V^*(P, Q)$$

$$\leq \frac{8(\beta-1)^2}{\beta^{3/2}} (3\beta^2 + 2\beta + 1) \left[ \frac{1}{2} \{R_{3/2}(Q, P) - B(P, Q)\} - h(P, Q) \right] \quad (4.23)$$

#### Proof:

Firstly, put  $s=1/2$  in (1.7) and (3.4) respectively, we get

$$\begin{aligned} \Phi_s(P, Q) &= 4 \sum_{i=1}^n 1 - \sqrt{p_i q_i} = 2 \sum_{i=1}^n 2 - 2\sqrt{p_i q_i} \\ &= 2 \sum_{i=1}^n p_i + q_i - 2\sqrt{p_i q_i} = 4 \sum_{i=1}^n \frac{(\sqrt{p_i} - \sqrt{q_i})^2}{2} = 4h(P, Q) \end{aligned} \quad (4.24)$$

$$\eta_s(P, Q) = 2 \sum_{i=1}^n (q_i - p_i) \sqrt{\frac{q_i}{p_i}} = 2 \sum_{i=1}^n \left( \frac{q_i^{3/2}}{p_i^{1/2}} - \sqrt{p_i q_i} \right) = 2 [R_{3/2}(Q, P) - B(P, Q)] \quad (4.25)$$

$$\text{Let } g(t) = t^{\frac{3}{2}} f''(t) = \frac{2(t-1)^2}{t^{\frac{3}{2}}} (3t^2 + 2t + 1) \quad (\text{After}$$

putting  $s=1/2$  in  $t^{2-s} f''(t)$ )

$$\text{Then } g'(t) = \frac{3(t-1)}{t^{\frac{5}{2}}} (5t^3 + t^2 + t + 1), g''(t) = \frac{3}{7} (15t^4 - 4t^3 + 5)$$

$$\text{If } g'(t) = 0 \Rightarrow (t-1)(5t^3 + t^2 + t + 1) = 0 \Rightarrow t = 1, -0.53$$

It is clear that  $g(t)$  is monotonic decreasing on  $(0, 1)$  and monotonic increasing on  $[1, \infty)$ .

Also  $g(t)$  has minimum value at  $t=1$ , since  $g''(1) = 24 > 0$  so

$$m = \inf_{t \in (0, \infty)} g(t) = g(1) = 0 \quad (4.26)$$

Now, we have two cases:

i. If  $0 < \alpha < 1$ , then

$$\begin{aligned} M &= \sup_{t \in (\alpha, \beta)} g(t) = \max \{g(\alpha), g(\beta)\} \\ &= \max \left\{ \frac{2(\alpha-1)^2}{\alpha^{\frac{3}{2}}} (3\alpha^2 + 2\alpha + 1), \frac{2(\beta-1)^2}{\beta^{\frac{3}{2}}} (3\beta^2 + 2\beta + 1) \right\} \end{aligned} \quad (4.27)$$

ii. If  $\alpha = 1$ , then

$$M = \sup_{t \in [1, \beta)} g(t) = \frac{2(\beta-1)^2}{\beta^{\frac{3}{2}}} (3\beta^2 + 2\beta + 1) \quad (4.28)$$

The results (4.20), (4.21), (4.22) and (4.23) are obtained by using (2.3), (4.7), (4.24), (4.25), (4.26), (4.27), and (4.28) in 3.1 and 3.2.

#### 4.4 Proposition 4.4(at s=0)

Let  $K(P, Q), \chi^2(P, Q)$  and  $V^*(P, Q)$  be defined as in (1.2), (1.3) and (2.3) respectively. Then, we have

i. If  $0 < \alpha < 1$ , then

$$0 \leq V^*(P, Q) \leq \max \left\{ \frac{2(\alpha-1)^2}{\alpha} (3\alpha^2 + 2\alpha + 1), \frac{2(\beta-1)^2}{\beta} (3\beta^2 + 2\beta + 1) \right\} K(Q, P) \quad (4.29)$$

$$0 \leq V_{\rho}^*(P, Q) - V^*(P, Q) \leq \max \left\{ \frac{2(\alpha-1)^2}{\alpha} (3\alpha^2 + 2\alpha + 1), \frac{2(\beta-1)^2}{\beta} (3\beta^2 + 2\beta + 1) \right\} \{ \chi^2(Q, P) - K(Q, P) \} \tag{4.30}$$

ii. If  $\alpha = 1$ , then

$$0 \leq V^*(P, Q) \leq \frac{2(\beta-1)^2}{\beta} (3\beta^2 + 2\beta + 1) K(Q, P) \tag{4.31}$$

$$0 \leq V_{\rho}^*(P, Q) - V^*(P, Q) \leq \frac{2(\beta-1)^2}{\beta} (3\beta^2 + 2\beta + 1) \{ \chi^2(Q, P) - K(Q, P) \} \tag{4.32}$$

**Proof:**

Firstly, put  $s=0$  in (1.7) and (3.4) respectively, we get

$$\lim_{s \rightarrow 0} \Phi_s(P, Q) = \sum_{i=1}^n q_i \log \left( \frac{q_i}{p_i} \right) = K(Q, P) \tag{4.33}$$

$$\eta_s(P, Q) = \sum_{i=1}^n \frac{q_i^2}{p_i} - q_i = \sum_{i=1}^n \frac{q_i^2}{p_i} - 1 = \sum_{i=1}^n \frac{q_i^2}{p_i} - 2q_i + p_i = \sum_{i=1}^n \frac{(p_i - q_i)^2}{p_i} = \chi^2(Q, P) \tag{4.34}$$

Let  $g(t) = t^2 f''(t) = \frac{2(t-1)^2}{t} (3t^2 + 2t + 1)$  (After putting

$s=0$  in  $t^{2-s} f''(t)$ )

Then

$$g'(t) = \frac{2(t-1)}{t^2} (9t^3 + t^2 + t + 1), g''(t) = \frac{1}{t^3} (36t^4 - 16t^3 + 4) \tag{4.35}$$

If

$$g'(t) = 0 \Rightarrow (t-1)(9t^3 + t^2 + t + 1) = 0 \Rightarrow t = 1, -0.43$$

It is clear that  $g(t)$  is monotonic decreasing on  $(0, 1)$  and monotonic increasing on  $[1, \infty)$ .

Also  $g(t)$  has minimum value at  $t=1$ , since  $g''(1) = 24 > 0$  so

$$m = \inf_{t \in (0, \infty)} g(t) = g(1) = 0 \tag{4.35}$$

Now, we have two cases:

i. If  $0 < \alpha < 1$ , then

$$M = \sup_{t \in (\alpha, \beta)} g(t) = \max \{ g(\alpha), g(\beta) \} = \max \left\{ \frac{2(\alpha-1)^2}{\alpha} (3\alpha^2 + 2\alpha + 1), \frac{2(\beta-1)^2}{\beta} (3\beta^2 + 2\beta + 1) \right\} \tag{4.36}$$

ii. If  $\alpha = 1$ , then

$$M = \sup_{t \in [1, \beta)} g(t) = \frac{2(\beta-1)^2}{\beta} (3\beta^2 + 2\beta + 1) \tag{4.37}$$

The results (4.29), (4.30), (4.31) and (4.32) are obtained by using (2.3), (4.7), (4.33), (4.34), (4.35), (4.36), and (4.37) in 3.1 and 3.2.

**4.5 Proposition 4.5(at s =-1)**

Let  $\chi^2(P, Q), R_{\alpha}(P, Q)$  and  $V^*(P, Q)$  be defined as in (1.3), (1.5) and (2.3) respectively. Then, we have

i. If  $0 < \alpha < 1$ , then

$$0 \leq V^*(P, Q) \leq \max \{ (\alpha-1)^2 (3\alpha^2 + 2\alpha + 1), (\beta-1)^2 (3\beta^2 + 2\beta + 1) \} \chi^2(Q, P)$$

$$0 \leq V_{\rho}^*(P, Q) - V^*(P, Q) \tag{4.38}$$

$$\leq \max \{ (\alpha-1)^2 (3\alpha^2 + 2\alpha + 1), (\beta-1)^2 (3\beta^2 + 2\beta + 1) \} \{ R_2(Q, P) - R_3(Q, P) - \chi^2(Q, P) \} \tag{4.39}$$

If  $\alpha = 1$ , then

$$0 \leq V^*(P, Q) \leq (\beta-1)^2 (3\beta^2 + 2\beta + 1) \chi^2(Q, P) \tag{4.40}$$

$$0 \leq V_{\rho}^*(P, Q) - V^*(P, Q) \leq (\beta-1)^2 (3\beta^2 + 2\beta + 1) \{ R_3(Q, P) - R_2(Q, P) - \chi^2(Q, P) \} \tag{4.41}$$

**Proof:**

Firstly, put  $s=-1$  in (1.7) and (3.4) respectively, we get

$$\Phi_s(P, Q) = \frac{1}{2} \sum_{i=1}^n \frac{q_i^2}{p_i} - 1 = \frac{1}{2} \sum_{i=1}^n \frac{q_i^2}{p_i} - 2q_i + p_i = \frac{1}{2} \sum_{i=1}^n \frac{(p_i - q_i)^2}{p_i} = \frac{1}{2} \chi^2(Q, P) \tag{4.42}$$

$$\eta_s(P, Q) = \frac{1}{2} \sum_{i=1}^n (q_i - p_i) \frac{q_i^2}{p_i^2} = \frac{1}{2} \sum_{i=1}^n \left( \frac{q_i^3}{p_i^2} - \frac{q_i^2}{p_i} \right) = \frac{1}{2} [R_3(Q, P) - R_2(Q, P)] \quad (4.43)$$

Let  $g(t) = t^3 f''(t) = 2(t-1)^2 (3t^2 + 2t + 1)$  (After putting  $s=-1$  in  $t^{2-s} f''(t)$ )

Then  $g'(t) = 24t^2(t-1)$ ,  $g''(t) = 72t^2 - 48t$

If  $g'(t) = 0 \Rightarrow t = 0, 1$

It is clear that  $g(t)$  is monotonic decreasing on  $(0, 1)$  and monotonic increasing on  $[1, \infty)$ .

Also  $g(t)$  has minimum value at  $t=1$ , since  $g''(1) = 24 > 0$  so

$$m = \inf_{t \in (0, \infty)} g(t) = g(1) = 0 \quad (4.44)$$

Now, we have two cases:

i. If  $0 < \alpha < 1$ , then

$$\begin{aligned} M &= \sup_{t \in (\alpha, \beta)} g(t) = \max \{g(\alpha), g(\beta)\} \\ &= \max \{2(\alpha-1)^2(3\alpha^2 + 2\alpha + 1), 2(\beta-1)^2(3\beta^2 + 2\beta + 1)\} \end{aligned} \quad (4.45)$$

ii. If  $\alpha = 1$ , then

$$M = \sup_{t \in [1, \beta)} g(t) = 2(\beta-1)^2(3\beta^2 + 2\beta + 1) \quad (4.46)$$

The results (4.38), (4.39), (4.40) and (4.41) are obtained by using (2.3), (4.7), (4.42), (4.43), (4.44), (4.45), and (4.46) in 3.1 and 3.2.

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