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**RESEARCH ARTICLE**

**A NON-SYMMETRIC DIVERGENCE AND KULLBACK-LEIBLER DIVERGENCE MEASURE**

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**ABSTRACT**

Information divergence measures and their bounds are well known in the literature of Information Theory. In this research article, we shall consider a new non-symmetric information divergence measure. Upper and lower bounds of non-symmetric divergence measure in terms of Kullback-Leibler divergence measure have been studied. Numerical bounds of new divergence measures are also discussed.

**Key words:**

Csiszar's f-Divergence Measure,  
Kullback-Leibler Divergence Measure,  
Information Inequalities etc.

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**INTRODUCTION**

Let

$$\Gamma_n = \left\{ P = (p_1, p_2, \dots, p_n) \mid p_i \geq 0, \sum_{i=1}^n p_i = 1 \right\}, n \geq 2 \tag{1.1}$$

be the set of all complete finite discrete probability distributions. There are many information and divergence measures exist in the literature on Information Theory. Csiszar (Csiszar, 1961) and (Csiszar, 1978) introduced a generalized measure of information using f-divergence measure is given by

$$I_f(P, Q) = \sum_{i=1}^n q_i f\left(\frac{p_i}{q_i}\right) \tag{1.2}$$

where  $f: \mathbf{R}_+ \rightarrow \mathbf{R}_+$  is a convex function and  $P, Q \in \Gamma_n$

Here we list existing divergence measure which is the category of Csiszar's f-divergence measure, together with the suitable generating function f.

**Kullback-Leibler divergence measure (Kullback and Leibler, 1951)**

(i) If  $f(t) = -\log t$  then kullback and Leibler divergence measure is given by

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$$I_f(P, Q) = D(Q, P) = \sum_{i=1}^n q_i \log \frac{q_i}{p_i} \dots\dots\dots (1.3)$$

(ii) If  $f(t) = t \log t$  then kullback and Leibler divergence measure is given by

$$I_f(P, Q) = D(P, Q) = \sum_{i=1}^n p_i \log \frac{p_i}{q_i} \dots\dots\dots (1.4)$$

In whole paper, in the section 2, we have introduced information inequalities. New non-symmetric information divergence measure has derived in section 3. Bounds of new information divergence measure in terms of Kullback-Leibler divergence measure have studied in section 4. In section 5, give the numerical bounds of new non-symmetric information divergence in terms of Kullback-Leibler divergence measure.

**INEQUALITIES RELATED WITH CSISZAR’S F-DIVERGENCE MEASURES**

The following proposition is one of the results of the theorem given in (Taneja and Kumar, 2004) and similar line to (Dragomir, 2001; Jain and Saraswat, 2013 and Jain and Saraswat, 2013).

**Proposition 2.1:-** Let  $f : (0, \infty) \rightarrow \mathbf{R}$  be a mapping which is normalized i.e.  $f(1) = 0$  and satisfies the assumptions.

(i) f is twice differentiable on  $(r, R)$  .where  $0 \leq r \leq 1 \leq R \leq \infty$

(ii) there exist the real Constants  $m, M$  such that  $m < M$

$$m \leq t^{2-s} f''(t) \leq M, \forall t \in (r, R), s \in \mathbf{R} \dots\dots\dots (2.1)$$

If  $P, Q \in \Gamma_n$  are discrete probability distributions satisfying assumption

$$0 < r \leq \frac{p_i}{q_i} \leq R < \infty, \forall i \in \{1, 2, 3, \dots, n\} \dots\dots\dots (2.1)$$

then we have the inequality

$$m \Phi_s(P, Q) \leq I_f(P, Q) \leq M \Phi_s(P, Q) \dots\dots\dots (2.3)$$

The case  $s=0, s=1$  of proposition (2.1) gives

**Proposition 2.2:-**Let  $f : (0, \infty) \rightarrow \mathbf{R}$  is normalized i.e.  $f(1) = 0$  and satisfies the assumptions.

(i) f is twice differentiable on  $(r, R)$  .where  $0 \leq r \leq 1 \leq R \leq \infty$

(ii) there exist Constant  $m, M$  such that  $m < M$

$$m \leq t^2 f''(t) \leq M, \forall t \in (r, R) \dots\dots\dots (2.4)$$

$$m \leq t f''(t) \leq M, \forall t \in (r, R) \dots\dots\dots (2.5)$$

If  $P, Q \in \Gamma_n$  are discrete probability distributions satisfying assumption

$$0 < r \leq \frac{p_i}{q_i} \leq R < \infty, \forall i \in \{1, 2, 3, \dots, n\}$$

then we have the inequality corresponding to  $s=0$  &  $s=1$

$$m D(Q, P) \leq I_f(P, Q) \leq M D(Q, P) \dots\dots\dots (2.6)$$

$$m D(P, Q) \leq I_f(P, Q) \leq M D(P, Q) \dots\dots\dots (2.7)$$

In view of proposition (4.1) we can state the following results.

**NON-SYMMETRIC DIVERGENCE MEASURE**

In this section we introduce a new information divergence measure which is the category of Csiszar’s f-divergence measure. Let us consider the function  $f : (0, \infty) \rightarrow \mathbf{R}$

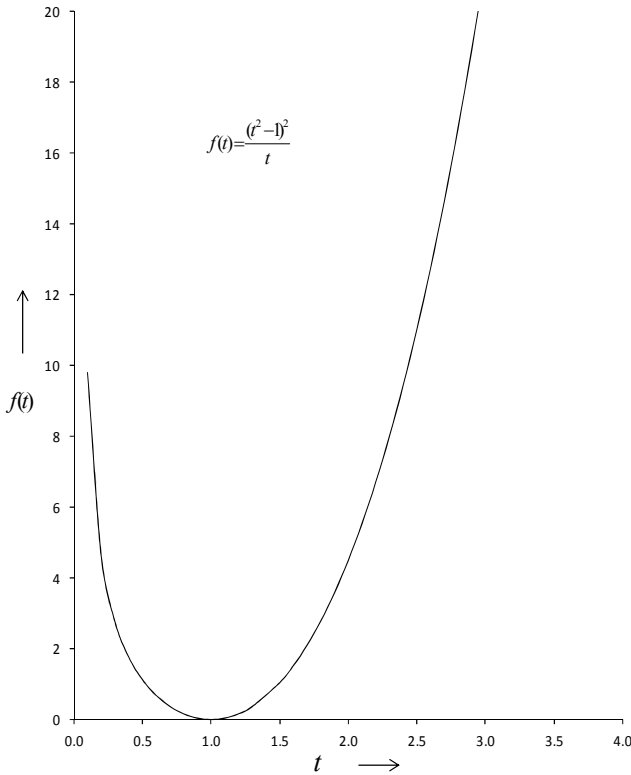
$$f(t) = \frac{(t^2 - 1)^2}{t}, f'(t) = \frac{3t^4 - 2t^2 - 1}{t^2}, f''(t) = \frac{6t^4 + 2}{t^3} > 0, \forall t > 0 \tag{3.1}$$

Hence function  $f(t)$  is convex from equation (3.1) and figure 3.1, and  $f(1) = 0$  i.e. normalized.

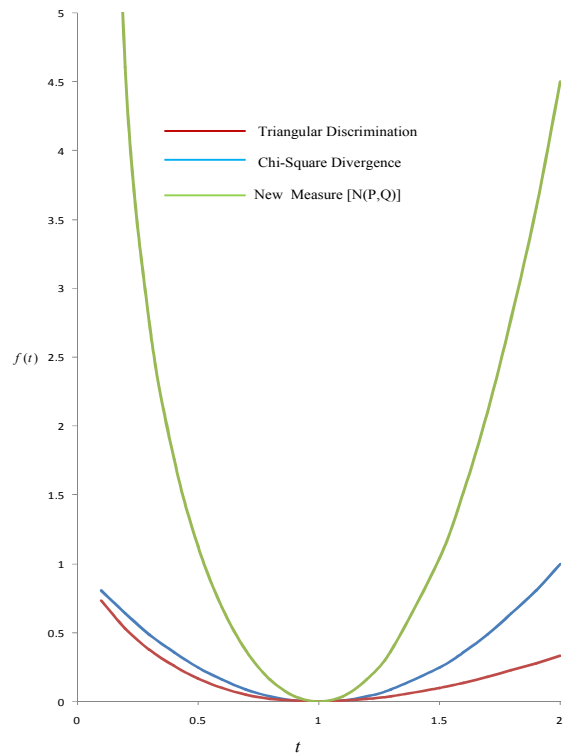
The f-divergence measure corresponding to function (3.1) is given by

$$I_f(P, Q) = \sum_{i=1}^n q_i f\left(\frac{p_i}{q_i}\right) = \sum_{i=1}^n \frac{(p_i^2 - q_i^2)^2}{q_i^2 p_i} = 4 \sum_{i=1}^n \left(\frac{1}{2} \frac{2p_i q_i}{p_i + q_i}\right) \frac{(p_i + q_i)(p_i - q_i)^2}{2 q_i} = N(P, Q) \tag{3.2}$$

where “ $N(P, Q)$ ” is may be combination of Harmonic , Arithmetic and  $\chi^2$ -divergence measure.



**Figure 3.1. Graphical presentation of New Measure  $N(P, Q)$**



**Figure 3.2. Comparison of New Measure  $N(P, Q)$  with some well-known divergence measures**

It is clear that from the figure 3.1 and 3.2 the convex function  $f(t)$  gives a steeper slope. Further  $f(1) = 0$  , so that  $N(P, P) = 0$  and the convexity of the function  $f(t)$  ensure that the measure (3.2) is non-negative.

In following sections 4 we present particular cases of the proposition (2.1) using the measure  $N(P, Q)$  given in equation (3.2).

**BOUNDS IN TERMS OF KULLBACK-LEIBLER DIVERGENCE MEASURE**

**RESULTS**

**Result 4.1.1:-** Let  $P, Q \in \Gamma_n$  and  $s = 0$ . Let there exists  $r, R$  such that  $r < R$  and

$$0 < r \leq \frac{P_i}{q_i} \leq R < \infty, \forall i \in \{1, 2, 3, \dots, n\}$$

(i) If  $r \in (0, \frac{1}{\sqrt{3}})$  then

$$\frac{8\sqrt{3}}{3} D(Q, P) \approx 4.61 D(Q, P) \leq N(P, Q) \leq \max \left\{ \frac{6r^4 + 2}{r}, \frac{6R^4 + 2}{R} \right\} D(Q, P) \dots \dots \dots (4.1)$$

(ii) If  $r \in (\frac{1}{\sqrt{3}}, \infty)$

$$\frac{6R^4 + 2}{R} D(Q, P) \leq N(P, Q) \leq \frac{6r^4 + 2}{r} D(Q, P) \dots \dots \dots (4.2)$$

**Proof:** - From equations (3.1), (3.2) & (4.1), we get

$$g(t) = t^2 f''(t) = \frac{6t^4 + 2}{t}, \forall t > 0$$

$$\text{we have } g'(t) = 18t^2 - \frac{2}{t^2} = 0,$$

$$g'(t) = 0 \text{ gives } t_0 = \frac{1}{\sqrt{3}} \approx .58$$

$$g''(t) = 36t + \frac{4}{t^3},$$

and

$$g''(.58) = 36(.58) + \frac{4}{(.58)^3} = 49.3 \text{ (Positive)}$$

which shows that function  $g(t)$  has minimum realized at  $t_0 = .58$  and  $\min g(t_0) = m$

we have two cases:-

(i) If  $0 < r \leq \frac{1}{\sqrt{3}}$ , then

$$\left. \begin{aligned} m &= \inf_{t \in [r, R]} g(t) = g(t_0) = \frac{8\sqrt{3}}{3} \approx 4.61 \\ M &= \sup_{t \in [r, R]} g(t) = \max \left\{ \frac{6r^4 + 2}{r}, \frac{6R^4 + 2}{R} \right\} \end{aligned} \right\} \dots \dots \dots (4.3)$$

(ii) If  $\frac{1}{\sqrt{3}} < r < \infty$ , then

$$m = \inf_{t \in [r, R]} g(t) = \frac{6r^4 + 2}{r}, M = \sup_{t \in [r, R]} g(t) = \frac{6R^4 + 2}{R} \dots \dots \dots (4.4)$$

Equations (2.6) & (2.7) of proposition (2.2) using equations (3.2), (4.3) and (4.4) gives the result (4.1) & (4.2).

**Result 4.1.2:-**

Let  $P, Q \in \Gamma_n$  and  $s = 1$ . Let there exists  $r, R$  such that  $r < R$  and  $0 < r \leq \frac{P_i}{q_i} \leq R < \infty, \forall i \in \{1, 2, 3, \dots, n\}$ .

(i) If  $0 < r \leq 0.76$ , then

$$4\sqrt{3}D(P,Q) \leq N(P,Q) \leq \max \left[ \frac{6r^4 + 2}{r^2}, \frac{6R^4 + 2}{R^2} \right] D(P,Q) \dots\dots\dots (4.5)$$

(ii) If  $0.76 < r \leq \infty$ , then

$$\frac{6r^4 + 2}{r^2} D(P,Q) \leq N(P,Q) \leq \frac{6R^4 + 2}{R^2} D(P,Q) \dots\dots\dots (4.6)$$

**Proof:** - From equations (3.1), (3.2) & (4.2), we get

$$g(t) = tf''(t) = \frac{6t^4 + 2}{t^2}, \forall t > 0$$

we have  $g'(t) = 12t - \frac{4}{t^3} = 0$ ,

$$g'(t) = 0 \text{ gives } t_0 = \left(\frac{1}{3}\right)^{1/4} \approx 0.76$$

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

and  $g''(0.76) = (\text{Positive})$

Which shows that function  $g(t)$  has minimum realized at  $t_0 = \left(\frac{1}{3}\right)^{1/4} \approx 0.76$ .

We have two cases:-

(i) If  $0 < r \leq 0.76$ , then

$$m = \inf_{t \in [r, R]} g(t) = g(t_0) = 4\sqrt{3} \dots\dots\dots (4.7)$$

$$M = \sup_{t \in [r, R]} = \max \left[ \frac{6r^4 + 2}{r^2}, \frac{6R^4 + 2}{R^2} \right] \dots\dots\dots (4.8)$$

(ii) If  $0.76 < r < \infty$ , then

$$m = \inf_{t \in [r, R]} g(t) = \frac{6r^4 + 2}{r^2}, M = \sup_{t \in [r, R]} = \frac{6R^4 + 2}{R^2} \dots\dots\dots (4.9)$$

Equations (2.6) and (2.7) of proposition (2.2) using equation (3.2), (4.7), (4.8) & (4.9) give the results (4.5) & (4.6).

**NUMERICAL ILLUSTRATIONS**

Let P be the binomial probability distribution for the random valuable X with parameter (n=8 p=0.5) and Q its approximated normal probability distribution. The following table have also discussed by Pranesh Kumar and Andrew Johnson in 2005.

**Table 5.1**

| x         | 0       | 1       | 2       | 3       | 4      | 5       | 6       | 7       | 8      |
|-----------|---------|---------|---------|---------|--------|---------|---------|---------|--------|
| p(x)      | 0.004   | 0.031   | 0.109   | 0.219   | 0.274  | 0.219   | 0.109   | 0.031   | 0.004  |
| q(x)      | 0.005   | 0.030   | 0.104   | 0.220   | 0.282  | 0.220   | 0.104   | 0.030   | 0.005  |
| p(x)/q(x) | 0.774   | 1.042   | 1.0503  | 0.997   | 0.968  | 0.997   | 1.0503  | 1.042   | 0.774  |
| N(P, Q)   | 0.00081 | 0.00013 | 0.00096 | 0.00018 | 0.0009 | 0.00018 | 0.00096 | 0.00013 | 0.0008 |

Here  $r = 0.77$  and  $R = 1.05$  are the lower and upper bounds. Now, we shall discuss the numerical bounds of new non-symmetric information divergence measure in terms of Kullback-Leibler divergence measure using equation (4.1), (4.2), (4.5) and (4.6) and the above table, then we get,

(i) If  $r \in (0, \frac{1}{\sqrt{3}})$  then

$$\frac{8\sqrt{3}}{3} D(Q, P) \approx 4.61 D(Q, P) \leq N(P, Q) \leq \max \left\{ \frac{6r^4 + 2}{r}, \frac{6R^4 + 2}{R} \right\} D(Q, P)$$

$$\frac{8\sqrt{3}}{3} D(Q, P) \approx 4.61 D(Q, P) \leq N(P, Q) \leq \max \left\{ \frac{6(.77)^4 + 2}{(.77)}, \frac{6(1.05)^4 + 2}{(1.05)} \right\} D(Q, P)$$

$$\frac{8\sqrt{3}}{3} D(Q, P) \approx 4.61 D(Q, P) \leq N(P, Q) \leq \max \{5.336, 8.850\} D(Q, P)$$

$$4.61 D(Q, P) \leq N(P, Q) \leq [8.850] D(Q, P)$$

(ii) If  $r \in (\frac{1}{\sqrt{3}}, \infty)$

$$\frac{6R^4 + 2}{R} D(Q, P) \leq N(P, Q) \leq \frac{6r^4 + 2}{r} D(Q, P)$$

$$[8.850] D(Q, P) \leq N(P, Q) \leq [5.336] D(Q, P)$$

(iii) If  $0 < r \leq 0.76$ , then

$$4\sqrt{3} D(P, Q) \leq N(P, Q) \leq \max \left[ \frac{6r^4 + 2}{r^2}, \frac{6R^4 + 2}{R^2} \right] D(P, Q)$$

$$4\sqrt{3} D(P, Q) \leq N(P, Q) \leq \max \left[ \frac{6(.77)^4 + 2}{(.77)^2}, \frac{6(1.05)^4 + 2}{(1.05)^2} \right] D(P, Q)$$

$$4\sqrt{3} D(P, Q) \leq N(P, Q) \leq \max [6.930, 8.4290] D(P, Q)$$

$$4\sqrt{3} D(P, Q) \leq N(P, Q) \leq [8.4290] D(P, Q)$$

(iv) If  $0.76 < r \leq \infty$ , then

$$\frac{6r^4 + 2}{r^2} D(P, Q) \leq N(P, Q) \leq \frac{6R^4 + 2}{R^2} D(P, Q)$$

$$[6.930] D(P, Q) \leq N(P, Q) \leq [8.4290] D(P, Q)$$

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