

On Generalized Measure of Weighted Symmetric Directed Divergence; Information Improvement

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Abstract: In the present paper we have obtained weighted directed divergence and symmetric directed divergence, information improvement, generalized information improvement corresponding to Sharma and Taneja⁵ measure of entropy.

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1. Introduction

Let $P = (p_1, p_2, \dots, p_n)$ be a probability distribution and $W = (w_1, w_2, \dots, w_n)$ be a set of weights associated with the n - outcomes, then corresponding to Shannon's¹ measure of entropy,

$$(1.1) \quad S(P) = -\sum_{i=1}^n p_i \ln p_i,$$

Guinasu⁴ defined a measure of weighted entropy

$$(1.2) \quad S(P, W) = -\sum_{i=1}^w w_i p_i \ln p_i,$$

and characterized it axiomatically. Now if $Q = (q_1, q_2, \dots, q_n)$ is another probability distribution then Kullback Leibler² gave a measure of directed

divergence of P from Q , as

$$(1.3) \quad D(P:Q) = \sum_{i=1}^n p_i \ln \frac{p_i}{q_i},$$

($\sum$ stands for summation over i from 1 to n through this paper) Kapur⁷ has defined weighted directed divergence axiomatically. A measure $D(P:Q,W)$ will be said to be an appropriate measure of weighted directed divergence if

- (i) It is a continuous function of $(p_1, p_2, \dots, p_n); (q_1, q_2, \dots, q_n); (w_1, w_2, \dots, w_n)$.
- (ii) It is permutationally symmetric, i.e. it does not change when the triplets $(p_1, q_1, w_1), (p_2, q_2, w_2), \dots, (p_n, q_n, w_n)$, are permuted amongst themselves.
- (iii) It is always ≥ 0 and vanishes when $p_i = q_i, i = 1, 2, \dots, n$.
- (iv) It is a convex function of $(p_1, p_2, \dots, p_n)$ which has its minimum value zero when $p_i = q_i, i = 1, 2, \dots, n$.
- (v) It reduces to a positive multiple of an ordinary measure of directed divergence when all the weights are equal.

Taneja and Tuteja⁶ gave the following weighted directed divergence corresponding to $D(P:Q)$ as

$$(1.4) \quad D_0(P:Q,W) = \sum_{i=1}^n w_i p_i \ln \frac{p_i}{q_i},$$

But very soon Kapur⁷ has pointed out that [1.4] is not a correct weighted directed divergence since it does not always satisfy (iii). Kapur⁷ has suggested the following weighted directed divergence corresponding to Kullback Leibler's² measure

$$(1.5) \quad D_0(P:Q,W) = \sum_{i=1}^n w_i \left(p_i \ln \frac{p_i}{q_i} - p_i + q_i \right),$$

Kapur⁷ also gave the following weighted divergence corresponding to Havrda-Charvat's³ directed divergence as

(1.6)

$$D_{\alpha}(P:Q,W) = \frac{1}{\alpha(\alpha-1)} \sum_{i=1}^n w_i \left\{ p_i^{\alpha} q_i^{1-\alpha} - \alpha p_i + \alpha q_i - q_i \right\}, \alpha > 0.$$

It is obvious that

$$(1.7) \quad \lim_{\alpha \rightarrow 1} D_{\alpha}(P:Q,W) = D_0(P:Q,W),$$

$$(1.8) \quad \text{It is well known that } D(P:Q) \neq D(Q:P),$$

$$(1.9) \quad \text{So naturally } D(P:Q,W) \neq D(Q:P,W),$$

Thus a weighted symmetric directed divergence has been defined as

$$(1.10) \quad J(P:Q,W) = D(P:Q,W) + D(Q:P,W),$$

$$(1.11) \quad \text{Therefore } J(P:Q,W) = J(Q:P,W),$$

The measure $I(P:Q:R,W)$ of weighted information improvement in going from Q to R when the true distribution is P and the weighted function is $W(x)$ is defined by

$$(1.12) \quad I(P:Q:R,W) = D(P:Q,W) - D(P:R,W)$$

This has the following properties:

$$(1.13) \quad \text{when } Q = R, I(P:Q:R,W) = 0,$$

i.e. no improvement, intuitionally correct

$$(1.14) \quad \text{when } Q = P, I(P:Q:R,W) < 0,$$

i.e., there cannot be any distribution better than true distribution.

$$(1.15) \quad \text{when } R = P, I(P:Q:R,W) > 0,$$

i.e. definitely improvement if we reach to true distribution. Some times during the course of investigation true distribution P also changes to another distribution S . In this case generalized measure of weighted information improvement $G(P:S/Q/R,W)$ is defined as

$$(1.16) \quad G(P:S/Q/R,W) = \frac{I(P:Q:R,W) + I(S:Q:R,W)}{2},$$

when there is no change in true distribution i.e. $S = P$.

$G(P:S/Q/R,W)=I(P:Q:R,W)$, This is again intuitively correct.

In section 1.2 we will obtain a measure of weighted directed divergence with 2 parameters, which generalizes (1.6) and enumerate its certain properties. In section 1.3 we will obtain measure of weighted symmetric directed divergence, weighted information improvement and weighted generalized information improvement.

2. Generalized Weighted Directed Divergence

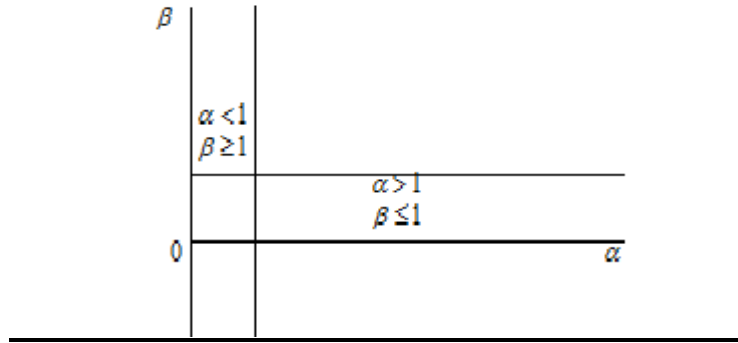
Now consider

$$(1.17) \quad D_{\alpha,\beta}(P:Q,W) = \frac{\sum w_i \left\{ p_i^\alpha q_i^{1-\alpha} - p_i^\beta q_i^{1-\beta} - (\alpha-\beta)(p_i - q_i) \right\}}{(\alpha-\beta)(\alpha+\beta-1)},$$

where $\alpha, \beta \geq 0$ and $\alpha > 1, \beta \leq 1$ or $\alpha < 1, \beta \geq 1$

$\alpha = \beta = t$ not admissible except at $t=1$

Range of validity is given in the following figure.



Properties of $D_{\alpha,\beta}(P:Q,W)$

$$(1.18) \quad (P_1) \quad D_{\alpha,\beta}(P:Q,W) = D_{\beta,\alpha}(P:Q,W)$$

$$(P_2) \quad D_{\alpha,1}(P:Q,W) = D_{\alpha}(P:Q,W);$$

$$(1.19) \quad D_{1,\beta}(P:Q,W) = D_{\beta}(P:Q,W)$$

$$(P_3) \quad D_{\alpha,\beta}(P:Q,W) = 0 \text{ when } P = Q,$$

$$(P_4) \quad D_{\alpha,\beta}(P:Q,W)$$

$$(1.20) \quad = \frac{\alpha(\alpha-1)D_{\alpha}(P:Q,W) + \beta(1-\beta)D_{\beta}(P:Q,W)}{(\alpha-\beta)(\alpha+\beta-1)}$$

Since $\frac{\alpha(\alpha-1)}{(\alpha-\beta)(\alpha+\beta-1)}$ and $\frac{\beta(1-\beta)}{(\alpha-\beta)(\alpha+\beta-1)}$,

both are > 0 for $\alpha < 1, \beta \geq 1$ or $\alpha > 1, \beta \leq 1$, so $D_{\alpha,\beta}(P:Q,W)$ will be convex function of $p_1, p_2, \dots, p_n$ (being sum of two convex functions of $p_1, p_2, \dots, p_n$ with positive coefficients) (P_5) $D_{\alpha,\beta}(P:Q,W)$ has minimum value at $p_i = q_i \forall i=1, 2, \dots, n$, its minimum value is zero, so $D_{\alpha,\beta}(P:Q,W) \geq 0$, (P_6) $D_{\alpha,\beta}(P:Q,W)$ Reduces to ordinary directed divergence corresponding to measure of entropy given by Sharma and Taneja⁵.

3. Generalized Weighted Symmetric Directed Divergence, Weighted Information Improvement and Weighted Generalized Information Improvement

$$(1.21) \quad J_{\alpha,\beta}(P:Q,W) = D_{\alpha,\beta}(P:Q,W) + D_{\alpha,\beta}(Q:P,W),$$

$$= \frac{\sum_i w \left\{ p_i^{\alpha} q_i^{1-\alpha} + q_i^{\alpha} p_i^{1-\alpha} - p_i^{\beta} q_i^{1-\beta} - q_i^{\beta} p_i^{1-\beta} \right\}}{(\alpha-\beta)(\alpha+\beta-1)}$$

$$(1.22) \quad I_{a,b}(P:Q:R,W) = D_{a,b}(P:Q,W) - D_{a,b}(P:R,W),$$

$$= \frac{\sum_i w \left\{ p_i^{\alpha} \left(q_i^{1-\alpha} - r_i^{1-\alpha} \right) - p_i^{\beta} \left(q_i^{1-\beta} - r_i^{1-\beta} \right) - (\alpha-\beta)(r_i - q_i) \right\}}{(\alpha-\beta)(\alpha+\beta-1)}$$

$$G_{\alpha,\beta}(P:S/Q/R,W) = (1/2)[I_{\alpha,\beta}(P:Q:R,W) + I_{\alpha,\beta}(S:Q:R,W)]$$

$$\begin{aligned}
& \sum w_i \left\{ \left(\frac{p_i^{\alpha+s_i} \alpha}{2} \right) (q_i^{1-\alpha} - r_i^{1-\alpha}) \right. \\
& \left. - \left(\frac{p_i^{\beta+s_i} \beta}{2} \right) \frac{(q_i^{1-\beta} - r_i^{1-\beta}) - (\alpha-\beta)(r_i - q_i)}{(\alpha-\beta)(\alpha+\beta-1)} \right\}, \\
& = \sum_{i=1}^n w_i \frac{\left\{ \left(\frac{p_i^{\alpha+s_i} \alpha}{2} \right) (q_i^{1-\alpha} - r_i^{1-\alpha}) - \left(\frac{p_i^{\beta+s_i} \beta}{2} \right) \right.}{(\alpha-\beta)(\alpha+\beta-1)} \left. (q_i^{1-\beta} - r_i^{1-\beta}) - (\alpha-\beta)(r_i - q_i) \right\}.
\end{aligned}$$

References

1. C. E. Shannon, A Mathematical Theory of Communication, *Bell System Tech. J.*, **27** (1948) 623-659.
2. S. Kullback, and R. A. Leibler, on Information and Sufficiency, *Ann Math Stat.*, **22** (1951) 79-86.
3. J. H. Havarda and F.C. Charvat, Quantification method of classification processes, Concept of Structural and α - entropy, *Kybernetika*, **3** (1967) 3-11.
4. S. Guiasu, Weighted Entropy Report on Math, *Physics*, **2** (1971) 165-179.
5. B. D. Sharma and I. J. Taneja Entropy of type α, β and other Generalized measures of information theory, *Metrika*, **22** (1975) 208-218.
6. H. C. Taneja, and R. K. Tuteja Characterization of a qualitative and quantitative measures of relative information, *Information Sciences*, **33** (1984) 106.
7. J. N. Kapur, *Measures of Information and their Applications* Wiley Eastern Ltd., New Delhi, 1994.