

The Concept of Measure of M-Divergence

Nidhi Sharma, P. Jha and C.L. Dewangan

*Department of Mathematics,
Govt. J.Y. Chhattisgarh College Raipur, India
E-mail: nsharma630@gmail.com*

Abstract

M-divergence is a measure of the directed divergence of a probability distribution $P = (p_1, p_2, \dots, p_n)$ from another probability distribution $Q = (q_1, q_2, \dots, q_n)$ when the probabilities in both distributions are monotonic increasing or monotonic decreasing.

Keywords: entropy, directed divergence, monotonicity.

Mathematical Classification No.: 94A17.

Introduction

In 1999, Kapur and Sharma [2] introduced M-entropy and later the same was discussed in Ph.D. Thesis of S. Sharma [4]. In 1951 Kullback-Leibler [3] introduced the measure of directed divergence

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

subject to

$$\sum_{i=1}^n p_i = 1$$

Kapur [1] defined measure of directed divergence

$$D_a(P:Q) = \sum_{i=1}^n (1 + ap_i) \ln \frac{(1 + ap_i)}{(1 + aq_i)}, \quad a > 0.$$

subject to

$$\sum_{i=1}^n p_i = 1$$

Here in the present paper we have introduced the concept of measures of M-Divergence corresponding to above measures of directed divergence.

The First Measure of M-Divergence

The first such measure is defined by

$$D_1(P:Q) = p_1 \ln \frac{p_1}{q_1} + (p_2 - p_1) \ln \frac{p_2 - p_1}{q_2 - q_1} + \dots + (p_n - p_{n-1}) \ln \frac{p_n - p_{n-1}}{q_n - q_{n-1}} + (1 - p_n) \ln \frac{1 - p_n}{1 - q_n} \tag{1}$$

subject to

$$q_1 < q_2 < \dots < q_n; p_1 < p_2 < \dots < p_n. \tag{2}$$

Now

$$\begin{aligned} \frac{\partial D_1}{\partial p_1} &= \ln \frac{p_1}{q_1} - \ln \frac{p_2 - p_1}{q_2 - q_1}, \quad \frac{\partial^2 D_1}{\partial p_1^2} = \frac{1}{p_1} + \frac{1}{p_2 - p_1} > 0 \\ \frac{\partial D_1}{\partial p_2} &= \ln \frac{p_2 - p_1}{q_2 - q_1} - \ln \frac{p_3 - p_2}{q_3 - q_2}, \quad \frac{\partial^2 D_1}{\partial p_2^2} = \frac{1}{p_2 - p_1} + \frac{1}{p_3 - p_2} > 0 \\ &\dots \dots \dots \\ \frac{\partial D_1}{\partial p_n} &= \ln \frac{p_n - p_{n-1}}{q_n - q_{n-1}} - \ln \frac{1 - p_n}{1 - q_n}, \quad \frac{\partial^2 D_1}{\partial p_n^2} = \frac{1}{p_n - p_{n-1}} + \frac{1}{1 - p_n} > 0 \\ \frac{\partial^2 D_1}{\partial p_i \partial p_{i+1}} &= -\frac{1}{p_{i+1} - p_i} \\ \therefore \frac{\partial^2 D_1}{\partial p_i^2} \cdot \frac{\partial^2 D_1}{\partial p_{i+1}^2} - \left(\frac{\partial^2 D_1}{\partial p_i \partial p_{i+1}} \right)^2 &> 0 \end{aligned}$$

Thus $D_1(P:Q)$ is a convex function of p_1, p_2, \dots, p_n .

Its minimum value subject to

$$\sum_{i=1}^n p_i = 1$$

is given by

$$\frac{p_2 - p_1}{p_1} = \frac{q_2 - q_1}{q_1}, \frac{p_3 - p_2}{p_2 - p_1} = \frac{q_3 - q_2}{q_2 - q_1}, \dots$$

$$\frac{P_n - P_{n-1}}{P_{n-1} - P_{n-2}} = \frac{q_n - q_{n-1}}{q_{n-1} - q_{n-2}}, \frac{P_n - P_{n-1}}{1 - P_n} = \frac{q_n - q_{n-1}}{1 - q_n}$$

which is satisfied if

$$p_1 = q_1, p_2 = q_2, \dots, p_n = q_n.$$

i.e. $P = Q$ (3)

so that the minimum value of $D_1(P:Q)$ arises when $P = Q$ and $D_1(P;Q) \geq 0$.

We can use this $D_1(P:Q)$ as a measure of M-divergence when both p_i 's and q_i 's are monotonic increasing.

Thus when there are no constraints except the natural constraint $\sum_{i=1}^n p_i = 1$ and the inequality constraints $p_i \geq 0, 1 \geq p_i \geq p_{i-1}, i = 1, 2, \dots, n$, the minimum M-divergence probability distribution is given by (3) and is same as the apriori distribution.

The Second Measure of M-Divergence

The second such measure is defined by

$$D_2(P:Q) = (1 + ap_i) \ln \frac{1 + ap_i}{1 + aq_i} + a(p_2 - p_1) \ln \frac{p_2 - p_1}{q_2 - q_1} + \dots + \dots \tag{4}$$

subject to

$$q_1 < q_2 < \dots < q_n ; p_1 < p_2 < \dots < p_n. \tag{5}$$

Now

$$\begin{aligned} \frac{\partial D_2}{\partial p_1} &= a \ln \frac{1 + ap_1}{1 + aq_1} - a \ln \frac{p_2 - p_1}{q_2 - q_1}, \frac{\partial^2 D_2}{\partial p_1^2} = \frac{a^2}{1 + ap_1} + \frac{a}{p_2 - p_1} > 0 \\ \frac{\partial D_2}{\partial p_2} &= a \ln \frac{p_2 - p_1}{q_2 - q_1} - a \ln \frac{p_3 - p_2}{q_3 - q_2}, \frac{\partial^2 D_2}{\partial p_2^2} = \frac{a}{p_2 - p_1} + \frac{a}{p_3 - p_2} > 0 \\ &\dots \dots \dots \\ &\dots \dots \dots \\ \frac{\partial D_2}{\partial p_n} &= a \ln \frac{p_n - p_{n-1}}{q_n - q_{n-1}} - a \ln \frac{p_{n+1} - p_n}{q_{n+1} - q_n}, \frac{\partial^2 D_2}{\partial p_n^2} = \frac{a}{p_n - p_{n-1}} + \frac{a}{p_{n+1} - p_n} > 0 \\ \frac{\partial^2 D_2}{\partial p_i \partial p_{i+1}} &= - \frac{a}{p_{i+1} - p_i} \\ \therefore \frac{\partial^2 D_2}{\partial p_i^2} \frac{\partial^2 D_2}{\partial p_{i+1}^2} - \left(\frac{\partial^2 D_2}{\partial p_i \partial p_{i+1}} \right)^2 &> 0 \end{aligned}$$

Thus $D_2(P:Q)$ is a convex function of p_1, p_2, \dots, p_n .

Its minimum value subject to

$$\sum_{i=1}^n p_i = 1$$

is given by

$$\frac{p_2 - p_1}{1 + ap_1} = \frac{q_2 - q_1}{1 + aq_1}, \frac{p_3 - p_2}{p_2 - p_1} = \frac{q_3 - q_2}{q_2 - q_1}, \dots, \frac{p_{n+1} - p_n}{p_n - p_{n-1}} = \frac{q_{n+1} - q_n}{q_n - q_{n-1}}$$

which is satisfied if

$$p_1 = q_1, p_2 = q_2, \dots, p_n = q_n.$$

i.e. $P = Q$ (6)

so that the minimum value of $D_2(P:Q)$ arises when $P = Q$ and $D_2(P:Q) \geq 0$.

We can use this $D_2(P:Q)$ as a measure of M-divergence when both p_i 's and q_i 's are monotonic increasing.

Thus when there are no constraints except the natural constraint $\sum_{i=1}^n p_i = 1$ and the inequality constraints $p_i \geq 0, 1 \geq p_i \geq p_{i-1}, i = 1, 2, \dots, n$, the minimum M-divergence probability distribution is given by (6) and is same as the apriori distribution.

References

- [1] Kapur, J.N. (1994). "Measure Of Information And Their Application" Wiley Eastern Limited, New Age International Limited.
- [2] Kapur, J.N. and Sharma S. (1999). "On Measures of M-Entropy" Indian J. Pure Appl. Math., 30(2): 129-145.
- [3] Kullback, S. and Leibler, R. A. (1951). "On Information and Sufficiency" Ann. Math. Stat. Vol. 22, pp. 79-86.
- [4] Sharma Sandeep(2000). Ph. D. Thesis, Jamia Millia Islamia University.