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## A Note on Characterizations of Entropies

J. S. CHAWLA

In this note we discuss and characterize four entropies each involving a parameter, by using the concept of generalized probability distributions. We shall call them entropies of order  $\alpha$ ,  $\beta$ ,  $\gamma$  and  $\delta$ .

#### 1. GENERALIZED PROBABILITY DISTRIBUTIONS

Let  $(\Omega, \mathcal{R}, \mathbf{P})$  be a probability space, that is,  $\Omega$  an arbitrary non-empty set, called the set of elementary events,  $\mathcal{R}$  a  $\sigma$ -field of subsets of  $\Omega$  and  $\mathbf{P}$  a probability measure defined on  $\mathcal{R}$ . Let us call a function  $\xi = \xi(w)$  which is defined for  $w \in \Omega_1$  where  $\Omega_1 \in \mathcal{R}$  and  $\mathbf{P}(\Omega_1) > 0$ , and which is measurable w.r.t.  $\mathcal{R}$ , a generalized random variable. If  $\mathbf{P}(\Omega_1) = 1$  we call  $\xi$  an ordinary (or complete) random variable, while if  $0 < \mathbf{P}(\Omega_1) < 1$  we call  $\xi$  an incomplete random variable. Clearly, an incomplete random variable can be interpreted as a quantity describing the result of an experiment depending on chance which is not always observable. The distribution of a generalized random variable will be called a generalized probability distribution. In particular, in the case when  $\xi$  takes on only a finite number of different values  $x_1, x_2, \ldots, x_n$  the distribution of  $\xi$  consists of the set of numbers  $p_k = \mathbf{P}\{\xi = x_k\}$  for  $k = 1, 2, \ldots, n$ . Thus a finite discrete generalized probability distribution is simply a sequence  $p_1, p_2, \ldots, p_n$  of non-negative numbers such that putting  $\mathbf{P} = (p_1, p_2, \ldots, p_n)$  and  $\mathbf{W}(\mathbf{P}) = \sum_{k=1}^n p_k$ , we have,  $0 < \mathbf{W}(\mathbf{P}) \le 1$ .

We shall call W(P) the weight of the distribution. Thus the weight of an ordinary distribution is equal to 1. A distribution which has a weight less than 1 will be called an incomplete distribution. If  $P = (p_1, p_2, \ldots, p_n)$  and  $Q = (q_1, q_2, \ldots, q_m)$  are two finite discrete generalized probability distributions, then P \* Q is the direct product of the distributions P and Q and is a finite discrete generalized probability distribution consisting of the numbers  $p_1q_k$  with  $j = 1, 2, \ldots, n$ ;  $k = 1, 2, \ldots, m$ .

$$P \cup Q = (p_1, p_2, ..., p_n, q_1, q_2, ..., q_m).$$

### 2. CHARACTERIZATION OF ENTROPIES

Let  $\Delta$  denote the set of all finite discrete generalized probability distributions, that is,  $\Delta$  is the set of all sequences  $P=(p_1,p_2,\ldots,p_n)$  of non-negative numbers such that  $0<\sum_{k=1}^n p_k \leq 1$ . It is assumed that the entropy H[P] of a finite discrete generalized probability distribution is defined for all  $P \in \Delta$ . We define the following postulates:

Postulate 1. H[P] is a symmetric function of the elements of P.

**Postulate 2.** If  $\{p\}$  denotes the generalized probability distribution consisting of the probability p, then  $H[\{p\}]$  is a continuous function of p in the interval 0 .

Postulate 3.  $H\left[\left\{\frac{1}{2}\right\}\right] = 1$ .

Postulate 4.  $H[\{1\}] = 0$ .

**Postulate 5.** If  $P = (p_1, p_2, ..., p_n)$  and  $Q = (q_1, q_2, ..., q_m)$ , then H[P \* Q] = H[P] + H[Q].

Postulate 6. If  $P = (p_1, p_2, ..., p_n)$  and  $Q = (q_1, q_2, ..., q_m)$ , such that  $W(P) + W(Q) \le 1$ , then

$$H[P \cup Q] = \frac{W(P) H(P] + W(Q) H[Q]}{W(P) + W(Q)}.$$

It has been shown by Renyi [8] that if H[P] satisfies the Postulates 1, 2, 3, 5 and 6, then

$$H[P] = \frac{\sum\limits_{k=1}^{n} p_k \log_2 (1/p_k)}{\sum\limits_{k=1}^{n} p_k},$$

which is a well known Shannon's entropy. We define another postulate as follows:

**Postulate 7.** There exists a strictly monotonic and continuous function y = g(x) such that if  $P = (p_1, p_2, \dots, p_n)$ ,  $Q = (q_1, q_2, \dots, q_m)$  and  $W(P) + W(Q) \le 1$ , then

$$H[P \cup Q] = g^{-1} \left| \frac{W(P) g(H[P]) + W(Q) g(H[Q])}{W(P) + W(Q)} \right|.$$

It was also shown by Renyi [8] that if  $g(x) = g_a(x) = 2^{(1-\alpha)x}$  where  $\alpha > 0$ ,  $\alpha \neq 1$ , then Postulates 1, 2, 3, 5 and 7 characterize Renyi's entropy of order  $\alpha$ . In other words if H[P] satisfies postulates 1, 2, 3, 5 and 7 with  $g(x) = g_a(x) = 2^{(1-\alpha)x}$ ,  $\alpha > 0$  and  $\alpha \neq 1$ , then, for  $P = (p_1, p_2, \dots, p_n)$ , we have

$$H_{a}[P] = \frac{1}{1-\alpha} \log_{2} \left[ \left( \sum_{k=1}^{n} p_{k}^{\alpha} \right) \middle/ \left( \sum_{k=1}^{n} p_{k} \right) \right],$$

which is well known Renyi's entropy.

M. Behara and P. Nath [3] introduced the following postulate for characterizing a new class of entropies.

Postulate 5'. For every 
$$P = (p_1, p_2, ..., p_n) \in \Lambda$$
,  $n = 1, 2, 3, ...$  and  $Q = \{q\}$ 

$$H[(p_1, p_2, ..., p_n) * q] = H[p_1q, p_2q, ..., p_nq] = a H[p_1, p_2, ..., p_n] H[q] + b H[p_1, p_2, ..., p_n] + b H[q] + c$$
, where  $ac = b^2 - b$  and  $a \neq 0$ .

They proved that if H[P] is defined for all  $P \in \Lambda$  and satisfies the Postulates 2, 3, 4, 5' and 6, then

$$H[P] = H_{\beta}[P] = \frac{1 - (\sum_{k=1}^{n} p_{k}^{\beta} / \sum_{k=1}^{n} p_{k})}{1 - 2^{1-\beta}}, \ \beta \neq 1 \text{ and } \beta > 0.$$

This entropy was first of all proposed by Havrda and Charvát [7]. Later on, Daroczy [6] and M. Behara and P. Nath [2, 3] studied these entropies. Vajda [9] also characterized this entropy for finite discrete generalized probability distributions.

In the following theorems we characterize two more entropies and to avoid confusion we call them entropies of order  $\gamma$  and  $\delta$ .

**Theorem 2.1.** If H[P] is defined for all  $P \in \Delta$  and satisfies Postulates 2, 3, 4 and 7 with  $g(x) = g_{\gamma}(x) = [1 - x(1 - 2^{\gamma - 1})]^{1/\gamma}$ ,  $\gamma \neq 1$ ,  $\gamma > 0$  and the following postulate **Postulate 5**". H[pq] = aH[p]H[q] + bH[p] + bH[q] + c, where  $ac = b^2 - b$  and  $a \neq 0$ , then  $H[P] = H_{\gamma}[P]$ , where, putting  $P = (p_1, p_2, \ldots, p_n)$ ; we have

$$H_{\gamma}[P] = \left\{1 - \left[\left(\sum_{k=1}^{n} p_{k}^{1/\gamma}\right) / \left(\sum_{k=1}^{n} p_{k}\right)\right]^{\gamma}\right\} / (1 - 2^{\gamma - 1}).$$

Proof. From Postulate 5", we have

$$\begin{split} a \; H[pq] &= a^2 \; H[p] \; H[q] \; + \; ab \; H[p] \; + \; ab \; H[q] \; + \; ac \; \Rightarrow \\ \Rightarrow a \; H[pq] &= \; a^2 \; H[p] \; H[q] \; + \; ab \; H[p] \; + \; ab \; H[q] \; + \; b^2 \; - \; b \; \Rightarrow \\ \Rightarrow \left(a \; H[pq] \; + \; b\right) \; = \left(a \; H[p] \; + \; b\right) \left(a \; H[q] \; + \; b\right). \end{split}$$

Replacing a H[p] + b by h[p], we obtain

(1) 
$$h\lceil pq \rceil = h\lceil p \rceil h\lceil q \rceil.$$

The continuity of  $H[p] \Rightarrow$  continuity of h[p]. Hence non-identically vanishing continuous solutions of (1) are of the form

$$h\lceil p \rceil = p^{1-\gamma}, \quad \gamma \neq 1$$

so that

$$H[p] = \frac{p^{1-\gamma} - b}{a}, \quad \gamma \neq 1.$$

Using Postulates 3 and 4, it can be easily seen that b=1 and  $a=2^{\gamma-1}-1, \gamma \neq 1$ , so that

$$H[p] = H_{\gamma}[p] = \frac{1 - p^{1-\gamma}}{1 - 2^{\gamma-1}}.$$

From Postulate 7, we have

$$H[p_{1}, p_{2}, \dots, p_{n}] = g^{-1} \left| \frac{\sum_{k=1}^{n} p_{k} g(H[p_{k}])}{\sum_{k=1}^{n} p_{k}} \right| =$$

$$= g^{-1} \left| \frac{\sum_{k=1}^{n} p_{k} g\left(\frac{1-p_{k}^{1-\gamma}}{1-2^{\gamma-1}}\right)}{\sum_{k=1}^{n} p_{k}} \right| = g^{-1} \left| \frac{\sum_{k=1}^{n} p_{k} \left\{1-\left(1-2^{\gamma-1}\right)\frac{1-p_{k}^{1-\gamma}}{1-2^{\gamma-1}}\right\}^{1/\gamma}}{\sum_{k=1}^{n} p_{k}} \right| =$$

$$= g^{-1} \left| \frac{\sum_{k=1}^{n} p_{k} p_{k}^{(1-\gamma)/\gamma}}{\sum_{k=1}^{n} p_{k}} \right| = g^{-1} \left| \frac{\sum_{k=1}^{n} p_{k} p_{k}^{1/\gamma}}{\sum_{k=1}^{n} p_{k}} \right| =$$

$$= \left\{1 - \left(\frac{\sum_{k=1}^{n} p_{k}^{1/\gamma}}{\sum_{k=1}^{n} p_{k}}\right)^{\gamma}\right\} / (1-2^{\gamma-1}); \gamma \neq 1, \gamma > 0.$$

Because, if  $y = g(x) = [1 - x(1 - 2^{\gamma - 1})]^{1/\gamma}$ , then

$$y^{\gamma} = 1 - x(1 - 2^{\gamma - 1}) \Rightarrow x(1 - 2^{\gamma - 1}) = 1 - y^{\gamma} \Rightarrow x = \frac{1 - y^{\gamma}}{1 - 2^{\gamma - 1}} \Rightarrow$$
$$\Rightarrow g^{-1}(y) = \frac{1 - y^{\gamma}}{1 - 2^{\gamma - 1}}.$$

The above entropy of order γ has been studied by M. Behara and J. S. Chawla [4]. Suguru Arimoto [1] while investigating the finite parameter estimation problem deduced the entropy of order γ using the concept of generalized information mea-

deduced the entropy of order  $\gamma$  using the concept of generalized information measure and discussed its relationship with Renyi's entropy of order  $\alpha$ .

**Theorem 2.2.** If H[P] is defined for all  $P \in A$  and satisfies the Postulates 2, 3, 6 and the following:

Postulate 5".  $H^{1/\delta}[pq] = H^{1/\delta}[p] + H^{1/\delta}[q]; \delta \neq 1, \delta > 0$ , then, for  $P = (p_1, p_2, \dots, p_n)$ 

$$H[P] = H_{\delta}[P] = \sum_{k=1}^{n} p_{k} |\log_{2} p_{k}|^{\delta} / \sum_{k=1}^{n} p_{k}.$$

Proof. Replacing  $H^{1/\delta}[p]$  by h[p] in Postulate 5", we obtain

(2) 
$$h[pq] = h[p] + h[q].$$

Now, by Postulate 2, H[p] is a continuous function of p,  $p \in (0, 1]$  and hence h[p] is a continuous function of p,  $p \in (0, 1]$ . Thus the only continuous solutions of (2) are of the form  $h[p] = c \log_2 p$  and hence  $H[p] = c^{\delta}(\log_2 p)^{\delta}$ .

Now,  $H_{\left[\frac{1}{2}\right]} = 1 \Rightarrow 1 = c^{\delta}(\log_2 \frac{1}{2})^{\delta} \Rightarrow 1 = c^{\delta}(-\log_2 2)^{\delta} \Rightarrow 1 = c^{\delta}(-1)^{\delta} \Rightarrow (-c)^{\delta} = 1 \Rightarrow c = -1$ . Thus  $H^{1/\delta}[p] = -\log_2 p \Rightarrow H[p] = |\log_2 p|^{\delta}$ . For  $(p_1, p_2) \in \Delta$ , we have

$$H[p_1, p_2] = \frac{p_1 H[p_1] + p_2 H[p_2]}{p_1 + p_2} = \frac{p_1 |\log_2 p_1|^{\delta} + p_2 |\log_2 p_2|^{\delta}}{p_1 + p_2},$$

$$H[p_1, p_2, p_3] = \frac{(p_1 + p_2) H[p_1, p_2] + p_3 H[p_3]}{p_1 + p_2 + p_3} =$$

$$= \frac{(p_1 + p_2) \frac{p_1 H[p_1] + p_2 H[p_2]}{(p_1 + p_2)} + p_3 H[p_3]}{(p_1 + p_2 + p_3)} = \frac{\sum_{i=1}^{3} p_i |\log_2 p_i|^{\delta}}{\sum_{i=1}^{3} p_i}.$$

Using the method of mathematical induction, we obtain

$$H[p_1, p_2, \ldots, p_n] = \frac{\sum_{i=1}^{n} p_i |\log_2 p_i|^{\delta}}{\sum_{i=1}^{n} p_i}.$$

This entropy has been studied by the author [5] for complete probability distributions. It has also been shown in [5] that the entropy of order  $\delta$  like Shannon entropy is also an invariant for isomorphic Bernoulli shifts.

It is important to note that if  $P = (p_1, p_2, \ldots, p_n)$ , then

$$\lim_{\alpha \to 1} H_{\alpha}[\mathsf{P}] = \lim_{\beta \to 1} H_{\beta}[\mathsf{P}] = \lim_{\gamma \to 1} H_{\gamma}[\mathsf{P}] = \lim_{\delta \to 1} H_{\delta}[\mathsf{P}] = \frac{-\sum\limits_{k=1}^{n} p_{k} \log_{2} p_{k}}{\sum\limits_{k=1}^{n} p_{k}}$$

which is Shannon's entropy for a finite discrete generalized probability distribution.

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