# Measures of Distance between Probability Distributions* 

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#### Abstract

In statistical estimation problems measures between probability distributions play significant roles. Hellinger coefficient, Jeffreys distance, Chernoff coefficient, directed divergence, and its symmetrization $J$-divergence are examples of such measures. Here these and like measures are characterized through a composition law and the sum form they possess. The functional equations


$$
f(p r, q s)+f(p s, q r)=(r+s) f(p, q)+(p+q) f(r, s)
$$

and

$$
f(p r, q s)+f(p s, q r)=f(p, q) f(r, s)
$$

are instrumental in their deduction. © 1989 Academic Press, Inc.

## 1. Introduction

Let $\Gamma_{n}^{0}=\left\{P=\left(p_{1}, p_{2}, \ldots, p_{n}\right) \mid 0<p_{k}<1, \sum_{k=1}^{n} p_{k}=1\right\}$ denote the set of all discrete $n$-ary complete positive probability distributions. Let $\left.I_{0}=\right] 0,1\left[, \mathbf{R}_{+}=\right] 0, \infty[, \mathbf{R}$ be the set of real numbers and $\mathbf{C}$ be the set of complex numbers. The following are some known measures between two probability distributions $P$ and $Q$ in $\Gamma_{n}^{0}$ :
(a) Hellinger coefficient $[4,9]$

$$
\begin{equation*}
H_{n}(P, Q)=\sum_{k=1}^{n} \sqrt{p_{k} q_{k}}, \tag{1.1}
\end{equation*}
$$

(b) Chernoff coefficient $[5,11]$

$$
\begin{equation*}
C_{n, \alpha}(P \| Q)=\sum_{k=1}^{n} p_{k}^{\alpha} q_{k}^{1-\alpha}, \quad \alpha \in I_{0} \tag{1.2}
\end{equation*}
$$

* This work is supported by NSERC of Canada grants.
(c) Jeffreys distance [10]

$$
\begin{equation*}
K_{n}(P, Q)=\sum_{k=1}^{n}\left(\sqrt{p_{k}}-\sqrt{q_{k}}\right)^{2} . \tag{1.3}
\end{equation*}
$$

These measures have many applications in statistics, pattern recognition, and numerical taxonomy. In information theory, for $P$ and $Q$ in $\Gamma_{n}^{0}$, the directed divergence [15] was defined as

$$
\begin{equation*}
I_{n}(P \| Q)=\sum_{k=1}^{n} p_{k} \log _{2} \frac{p_{k}}{q_{k}}, \tag{1.4}
\end{equation*}
$$

and was characterized in $[12,18]$. It serves as a separability measure and is frequently used in statistics [15] and pattern recognition [22, 24]. However, the directed divergence is not symmetric. In [15], symmetric divergence was introduced as

$$
\begin{equation*}
J_{n}(P, Q)=I_{n}(P \| Q)+I_{n}(Q \| P) \tag{1.5}
\end{equation*}
$$

to restore the symmetry. It has the explicit form

$$
\begin{equation*}
J_{n}(P, Q)=\sum_{k=1}^{n}\left(p_{k}-q_{k}\right) \log _{2} \frac{p_{k}}{q_{k}}, \tag{1.6}
\end{equation*}
$$

and is also called the $J$-divergence in honor of Jeffreys [10] who first used this measure in some estimation problems. It is non-negative and attains minimum when $P=Q$. Moreover, it satisfies the composition law

$$
\begin{equation*}
J_{n m}(P * R, Q * S)+J_{n m}(P * S, Q * R)=2 J_{n}(P, Q)+2 J_{m}(R, S) \tag{1.7}
\end{equation*}
$$

for all $P, Q \in \Gamma_{n}^{0}$ and $R, S \in \Gamma_{m}^{0}$ where

$$
P * R=\left(p_{1} r_{1}, p_{1} r_{2}, \ldots, p_{1} r_{m}, p_{2} r_{1}, \ldots, p_{2} r_{m}, \ldots, p_{n} r_{m}\right)
$$

Evidently $J_{n}$ has the sum form [12,16, 18]. It was characterized in [14] through recursivity, symmetry, differentiability, and some normalization conditions. Applications of $J_{n}$ to statistics may be found in [10, 11, 15, 19,21], to pattern recognition in [22,24], and to questionnaire analysis in [8]. The directed divergence of degree $\alpha$ defined as

$$
\begin{equation*}
I_{n, \alpha}(P \| Q)=\left(2^{\alpha-1}-1\right)^{-1}\left(\sum_{k=1}^{n} p_{k}^{x} q_{k}^{1 \cdot x}-1\right), \tag{1.8}
\end{equation*}
$$

where $\alpha \in \mathbf{R}-\{1\}$, was characterized in [17]. Like directed divergence, this
measure is not symmetric either. In [21], thus the symmetric divergence ( $J$-divergence) of degree $\alpha$ was defined as

$$
J_{n, \alpha}(P, Q)=I_{n, \alpha}(P \| Q)+I_{n, \alpha}(Q \| P)
$$

which indeed by (1.8) equals

$$
\begin{equation*}
J_{n, \alpha}(P, Q)=\frac{\sum_{k=1}^{n}\left(p_{k}^{\alpha} q_{k}^{1-\alpha}+p_{k}^{1-\alpha} q_{k}^{\alpha}\right)-2}{2^{\alpha-1}-1} \tag{1.9}
\end{equation*}
$$

It is a one parameter generalization of (1.6) since (1.9) tends to (1.6) as $\alpha \rightarrow 1$. Some of its properties can be found in [21]. In addition to having the sum form it also satisfies the following composition law

$$
\begin{align*}
& J_{n m, \alpha}(P * R, Q * S)+J_{n m, \alpha}(P * S, Q * R) \\
& \quad=2 J_{n, \alpha}(P, Q)+2 J_{m, \alpha}(R, S)+\lambda J_{n, \alpha}(P, Q) J_{m, \alpha}(R, S), \tag{1.10}
\end{align*}
$$

where $\lambda=2^{\alpha-1}-1$. This property will be called the symmetric compositivity of $J_{n, \alpha}$. If $\alpha \rightarrow 1$, then $\lambda \rightarrow 0$ and (1.10) yields (1.7).

Let $\mu_{n}: \Gamma_{n}^{0} \times \Gamma_{n}^{0} \rightarrow \mathbf{R}(n \geqslant 2)$ be symmetric, that is, $\mu_{n}(P, Q)=\mu_{n}(Q, P)$.
Definition 1. A sequence of symmetric measures $\left\{\mu_{n}\right\}$ is said to be symmetrically compositive if for some $\lambda \in \mathbf{R}$,

$$
\begin{align*}
& \mu_{n m}(P * R, Q * S)+\mu_{n m}(P * S, Q * R) \\
& \quad=2 \mu_{n}(P, Q)+2 \mu_{m}(R, S)+\lambda \mu_{n}(P, Q) \mu_{m}(R, S) \tag{1.11}
\end{align*}
$$

for all $P, Q \in \Gamma_{n}^{0}, S, R \in \Gamma_{m}^{0}$. If $\lambda=0$, then $\left\{\mu_{n}\right\}$ is said to be symmetrically additive.

Definition 2. A sequence of symmetric measures $\mu_{n}: \Gamma_{n}^{0} \times \Gamma_{n}^{0} \rightarrow \mathbf{R}$ ( $n \geqslant 2$ ) is said to have the sum form [16,18] if there exists a symmetric function $f: I_{0}^{2} \rightarrow \mathbf{R}$ such that

$$
\begin{equation*}
\mu_{n}(P, Q)=\sum_{k=1}^{n} f\left(p_{k}, q_{k}\right) \tag{1.12}
\end{equation*}
$$

for all $P, Q \in \Gamma_{n}^{0}$ (symmetry of $f$ means $f(p, q)=f(q, p)$ ).
The function $f$ in (1.12) is called a generating function of the sequence $\left\{\mu_{n}\right\}$. For details regarding the existence of $f$, interested readers should refer to [16].

The sum form and symmetric compositivity lead to the study of the functional equation

$$
\begin{align*}
& \sum_{i=1}^{n} \sum_{j=1}^{m} f\left(p_{i} r_{j}, q_{i} s_{j}\right)+\sum_{i=1}^{n} \sum_{i=1}^{m} f\left(p_{i} s_{j}, q_{i} r_{j}\right) \\
& \quad=2 \sum_{i=1}^{n} f\left(p_{i}, q_{i}\right)+2 \sum_{j=1}^{m} f\left(r_{j}, s_{j}\right)+\lambda \sum_{i=1}^{n} f\left(p_{i}, q_{i}\right) \sum_{j=1}^{m} f\left(r_{j}, s_{j}\right), \tag{1.13}
\end{align*}
$$

where $P, Q \in \Gamma_{n}^{0}, R, Q \in \Gamma_{m}^{0}$, and $\lambda \in \mathbf{R}$.
A function $A$ from an interval to $\mathbf{R}$ is said to be additive if $A(x+y)=A(x)+A(y)$. An additive $A: I_{0} \rightarrow \mathbf{R}$ has a unique extension to additive $\bar{A}: \mathbf{R} \rightarrow \mathbf{R}$ [7, Theorem 4.3]. A mapping $M: I_{0} \rightarrow \mathbf{R}$ is multiplicative if $M(x y)=M(x) M(y)$. A multiplicative $M: I_{0} \rightarrow \mathbf{R}$ has a unique extension to multiplicative $\bar{M}: \mathbf{R}_{+} \rightarrow \mathbf{R}[2]$. A function $L: I_{0} \rightarrow \mathbf{R}$ is said to be logarithmic provided $L(x y)=L(x)+L(y)$. A logarithmic $L: I_{0} \rightarrow \mathbf{R}$ has a unique extension $\bar{L}: \mathbf{R}_{+} \rightarrow \mathbf{R}[2]$.
The aim of this paper is to find all the symmetrically compositive sequences $\mu_{n}: \Gamma_{n}^{0} \times \Gamma_{n}^{0} \rightarrow \mathbf{R}$ having the sum form with a measurable symmetric generating function $f: I_{0}^{2} \rightarrow \mathbf{R}$. The measures we obtain include such measures as Hellinger coefficient [9], Chernoff coefficient [5], Jeffreys distance [10], $J$-divergence [15], $J$-divergence of degree $\alpha$ [21], and more. The concept of measuring similarity between two distributions is dual to that of distance and the measures obtained here cover both.

This paper is organized as follows: In Section 2, we present the Lebesgue measurable solution of the functional equation (1.13). In Section 3, we display all the regular symmetric measures which have the sum form and the compositivity. Section 4 contains discussion about a new measure there contained and its relationship to Hellinger coefficient and Jeffreys distance.

## 2. Solutions of the Functional Equation (1.13)

We make use of the following two auxiliary results to obtain the solutions of the functional equations (1.13).

Lemma 1. A function $f: I_{0}^{2} \rightarrow \mathbf{R}$ satisfies the functional equation

$$
\begin{equation*}
f(p r, q s)+f(p s, q r)=(r+s) f(p, q)+(p+q) f(r, s) \tag{2.1}
\end{equation*}
$$

for all $p, q, r, s \in I_{0}$ if, and only if,

$$
\begin{equation*}
f(p, q)=p\left[L_{1}(q)-\Sigma_{2}(p)\right]+q\left[L_{1}(p)-L_{2}(q)\right] \tag{2.2}
\end{equation*}
$$

where $L_{1}, L_{2}$ are logarithmic on $\mathbf{R}_{+}$.

Proof. It is easy to verify that $f$ given by (2.2) satisfies (2.1). Obviously, $f=0$ is a solution of (2.1) and is of the form (2.2). We now suppose $f \neq 0$. First, we will show that $f$ satisfying (2.1) for all $p, q, r, s \in I_{0}$ can be extended (uniquely) from $I_{0}^{2}$ to $\mathbf{R}_{+}^{2}$. Setting $r=s=\lambda$ in (2.1), we get

$$
\begin{equation*}
f(\lambda p, \lambda q)=\lambda f(p, q)+\lambda(p+q) l(\lambda) \tag{2.3}
\end{equation*}
$$

where $l(\lambda):=(2 \lambda)^{-1} f(\lambda, \lambda)$ is logarithmic (replace $\lambda$ by $\lambda_{1} \lambda_{2}$ in (2.3) and use (2.3) three times). Now, we extend $f$ to $\hat{f}$, as follows. For any $p, q \in \mathbf{R}_{+}$, choose positive $\lambda$ sufficiently small such that $\lambda, \lambda p, \lambda q \in I_{0}$. Define

$$
\begin{equation*}
\bar{f}(p, q)=\frac{1}{\lambda} f(\lambda p, \lambda q)-(p+q) l(\lambda) \tag{2.4}
\end{equation*}
$$

From $f(\lambda \mu p, \lambda \mu q)=f(\lambda . \mu p, \lambda . \mu q)=f(\mu . \lambda p, \mu . \lambda q)$, using (2.3), it follows that the right side of (2.4) is independent of $\lambda$ and thus $f$ is well defined. For $p$, $q, r, s \in \mathbf{R}_{+}$, choose $\lambda \in I_{0}$ such that $\lambda p, \lambda q, \lambda r, \lambda s \in I_{0}$. Then

$$
\begin{aligned}
& f(p r, q s)+f(p s, q r) \\
&= \frac{1}{\lambda^{2}}\left\{f\left(\lambda^{2} p r, \lambda^{2} q s\right)+f\left(\lambda^{2} p s, \lambda^{2} q r\right)\right\}-(p r+q s) l\left(\lambda^{2}\right)-(p s+q r) l\left(\lambda^{2}\right) \\
&= \frac{1}{\lambda^{2}}\{f(\lambda p \lambda r, \lambda q \lambda s)+f(\lambda p \lambda s, \lambda q \lambda r)\}-(p+q)(r+s) l\left(\lambda^{2}\right) \\
&= \frac{1}{\lambda}\{(r+s) f(\lambda p, \lambda q)+(p+q) f(\lambda r, \lambda s)\} \\
&-2(p+q)(r+s) l(\lambda) \quad \text { (by using }(2.1) \text { and } l \text { logarithmic }) \\
&=(r+s) f(p, q)+(p+q) f(r, s) .
\end{aligned}
$$

Thus $f$ satisfies (2.1) on $\mathbf{R}_{+}$. From here on, let us simply assume that $f$ satisfies (2.1) for $p, q, r, s \in \mathbf{R}_{+}$. Set $p=q, r=s$ in (2.1) to get

$$
f(p r, p r)=r f(p, p)+p f(r, r)
$$

From this it follows that

$$
\begin{equation*}
L(p):=\frac{1}{p} f(p, p) \tag{2.5}
\end{equation*}
$$

is logarithmic on $\mathbf{R}_{+}$. Setting $q=s=1$ in (2.1), we get

$$
\begin{equation*}
f(p, r)=(1+r) g(p)+(1+p) g(r)-g(p r) \tag{2.6}
\end{equation*}
$$

where $g(p):=f(1, p)$. Note that (2.1) implies $f(1,1)=0$ and $f$ is symmetric. Now (2.5) and (2.6) give

$$
\begin{equation*}
g\left(p^{2}\right)=2(1+p) g(p)-p L(p), \quad p \in \mathbf{R}_{+} \tag{2.7}
\end{equation*}
$$

With $p=r, q=s$ (2.1) yields

$$
\begin{equation*}
f\left(p^{2}, q^{2}\right)+f(p q, p q)=2(p+q) f(p, q) \tag{2.8}
\end{equation*}
$$

Putting (2.5), (2.6), and (2.7) into (2.8), we have

$$
\begin{gathered}
(1-p)(1-q)[2 g(p q)-L(p q)]=(1-p)(1-p q)[2 g(q)-L(q)] \\
+(1-q)(1-p q)[2 g(p)-L(p)]
\end{gathered}
$$

Defining

$$
2 L_{2}(x)= \begin{cases}(1-x)^{-1}(2 g(x)-L(x)), & x \neq 1 \\ 0, & x=1\end{cases}
$$

we get from this definition and $g(1)=0$ that

$$
g(p)=\frac{1}{2} L(p)+(1-p) L_{2}(p), \quad p \in \mathbf{R}_{+}
$$

and from the above equation $L_{2}(p q)=L_{2}(p)+L_{2}(q)$ follows whenever $p \neq 1, q \neq 1$, and $p q \neq 1 . L_{2}$ evidently satisfies $L_{2}(p q)=L_{2}(p)+L_{2}(q)$ when $p=1$ or $q=1$. To check that this equation is also true for the case $p \neq 1$, $q \neq 1$ but $p q=1$ we have to show that $L_{2}\left(p^{-1}\right)=-L_{2}(p)$ for $p \neq 1$. The latter is equivalent to

$$
\left(1+\frac{1}{p}\right) f(p, p)+2 p f\left(1, \frac{1}{p}\right)=2 f(1, p)
$$

which can be obtained by putting $q=p, r=1$, and $s=1 / p$ in (2.1). Thus, $L_{2}$ is logarithmic on $\mathbf{R}_{+}$. Now using (2.6), we get (2.2), where $L_{1}=\frac{1}{2} L+L_{2}$. This proves Lemma 1.

Lemma 2. Suppose $f: I_{0}^{2} \rightarrow \mathbf{R}$ satisfies the functional equation

$$
\begin{equation*}
f(p r, q s)+f(p s, q r)=f(p, q) f(r, s) \tag{2.9}
\end{equation*}
$$

for all $p, q, r, s \in I_{0}$. Then

$$
\begin{equation*}
f(p, q)=M_{1}(p) M_{2}(q)+M_{1}(q) M_{2}(p) \tag{2.10}
\end{equation*}
$$

where $M_{1}, M_{2}: I_{0} \rightarrow \mathbf{C}$ are multiplicative. Further, either $M_{1}$ and $M_{2}$ are both real or $M_{2}$ is the complex conjugate of $M_{1}$. The converse is also true.

Proof. Without loss of generality, let us assume that $f \neq 0$. Observe first that with $r=s=\lambda$ in (2.9) we get the $M$-homogeneity law

$$
\begin{equation*}
f(\lambda p, \lambda q)=\frac{1}{2} f(\lambda, \lambda) f(p, q)=M(\lambda) f(p, q) \tag{2.11}
\end{equation*}
$$

where $M: I_{0} \rightarrow \mathbf{R}, M(\lambda):=\frac{1}{2} f(\lambda, \lambda)$. Since $f \neq 0, M$ is multiplicative and can be uniquely extended to a multiplicative $\bar{M}(\neq 0)$ on $\mathbf{R}_{+}$[2]. As in Lemma 1, it is easy to show that

$$
\bar{f}(p, q)=\bar{M}\left(\frac{1}{\lambda}\right) f(\lambda p, \lambda q)
$$

for $p, q \in \mathbf{R}_{+}\left(\lambda\right.$ small such that $\left.\lambda p, \lambda q \in I_{0}\right)$ extends $f$ and Eq. (2.9) to $\mathbf{R}_{+}^{2}$. We assume from here on that $f$ has this extended meaning.

We then fix $s=q=1$ in (2.9) and define $g(p)=f(p, 1)$ to get

$$
\begin{equation*}
f(p, q)=g(p) g(q)-g(p q) \tag{2.12}
\end{equation*}
$$

The assumption $f \neq 0$ implies that $g$ is not multiplicative, and $f(1,1)=2$. Setting $s=1$ in (2.9) and using (2.12), we obtain

$$
\begin{equation*}
g(p r) g(q)-2 g(p q r)=g(p) g(q) g(r)-g(p) g(q r)-g(r) g(p q) \tag{2.13}
\end{equation*}
$$

We fix $q$ in (2.13) to get

$$
\begin{equation*}
F(p r)=F(p) \frac{g(r)}{2}+F(r) \frac{g(p)}{2} \tag{2.14}
\end{equation*}
$$

where $F(p):=g(p) g(q)-2 g(p q)$.
If $F$ is identically 0 , then $M_{1}:=2^{-1} g$ is multiplicative and $f(p, q)=2 M_{1}(p q)$, and $f$ thus is of the form (2.10) with $M_{2}=M_{1}$. If $F \neq 0$, then from (2.14) [6, Lemma 5, Remark 4] and since $g$ is not multiplicative we have $g=M_{1}+M_{2}$ where $M_{1}, M_{2}: \mathbf{R}_{+} \rightarrow \mathrm{C}$ are multiplicative; that is, $f$ is of the form (2.10) where $M_{1}, M_{2}$ are complex valued. Suppose $M_{k}(p)=a_{k}(p)+i b_{k}(p)$ with real $a_{k}(p), b_{k}(p)(k=1,2)$. Since $f$ is real valued, letting $q=1$ in (2.12), we get $b_{1}(p)=-b_{2}(p)=b(p)$ say. Again using $f$ real, that is, the imaginary part of $f=0$, we get

$$
\begin{equation*}
\left[a_{2}(p)-a_{1}(p)\right] b(q)+\left[a_{2}(q)-a_{1}(q)\right] b(p)=0 \tag{2.15}
\end{equation*}
$$

Suppose there is a $p_{0}$ such that $a_{1}\left(p_{0}\right) \neq a_{2}\left(p_{0}\right)$. Then from (2.15), we have $b(q)=c\left(a_{2}(q)-a_{1}(q)\right)$, that is,

$$
c\left(a_{2}(p)-a_{1}(p)\right)\left(a_{2}(q)-a_{1}(q)\right)=0
$$

for all $p, q$. Then $c=0$, so $b(q)=0$ and $M_{1}$ and $M_{2}$ are real. Otherwise $a_{1}=a_{2}$ and $M_{2}$ equals the conjugate of $M_{1}$ as desired. The converse part is straightforward. This completes the proof of Lemma 2.

Remark 1. The general solutions of functional Eqs. 2.1) and (2.9) are obtaincd in Lemmas 1 and 2 without assuming any regularity conditions on $f$.

Now we proceed to find the Lebesgue measurable and symmetric solution of the functional equation (1.13). We first treat the case $\lambda=0$.

Theorem 1. Let $f: I_{0}^{2} \rightarrow \mathbf{R}$ be symmetric and measurable in each variable. Then $f$ satisfies the functional equation (1.13) with $\lambda=0$ for all pairs of positive integers $m, n(\geqslant 2)$ and for all $P, Q \in \Gamma_{n}^{0}, R, S \in \Gamma_{m}^{0}$ if, and only if, $f$ is given by

$$
\begin{equation*}
f(p, q)=a\left(p \log _{2} p+q \log _{2} q\right)+b\left(p \log _{2} q+q \log _{2} p\right) \tag{2.16}
\end{equation*}
$$

where $a, b$ are arbitrary constants.
Proof. For fixed $P, Q \in \Gamma_{n}^{0}$, define

$$
\begin{align*}
F(r, s)= & \sum_{i=1}^{n}\left[f\left(p_{i} r, q_{i} s\right)+f\left(p_{i} s, q_{i} r\right)\right. \\
& \left.-(r+s) f\left(p_{i}, q_{i}\right)-\left(p_{i}+q_{i}\right) f(r, s)\right] \tag{2.17}
\end{align*}
$$

First, (2.17) in (1.13) with $\lambda=0$ gives $\sum_{j=1}^{m} F\left(r_{j}, s_{j}\right)=0$ and then the measurability of $f$ implies (refer to [3]) $F(r, s)=a_{1} r+a_{2} s+a_{3}$, where $a_{1}, a_{2}, a_{3}$ are functions of $P$ and $Q$ with $a_{1}+a_{2}+m a_{3}=0$. Using the symmetry of $f$ and noting the fact that (1.13) holds for all $m$ (enough for two values of $m$ ) we get $a_{1}=a_{2}=a_{3}=0$, that is, the right hand side of (2.17) is zero. Keeping $r$ and $s$ constant, defining

$$
G(p, q)=f(p r, q s)+f(p s, q r)-(r+s) f(p, q)-(p+q) f(r, s)
$$

and applying once again the above procedure to $G$ in place of $F$, we get (2.1). Since $f$ is measurable in each variable, so are $L_{1}$ and $L_{2}$ derived in Lemma 1, giving $L_{1}(p)=a \log _{2} p$ and $L_{2}(p)=b \log _{2} p$, where $a$ and $b$ are real constants (see [1]). Thus from (2.2) follows (2.16). The converse part is easy to verify. This proves the theorem.

Now we consider the case $\lambda \neq 0$ and prove the following theorem.

THEOREM 2. Let $f: I_{0}^{2} \rightarrow \mathbf{K}$ be symmetric and measurable in each variable. Then $f$ satisfies the functional equation (1.13) for all pairs of positive integers
$m, n(\geqslant 2)$ and for all $P, Q \in \Gamma_{n}^{0}, R, S \in \Gamma_{m}^{0}$ with $\lambda \neq 0$ if, and only if, $f$ has the form

$$
\begin{equation*}
f(p, q)=-\frac{1}{\lambda}(p+q) \tag{2.18}
\end{equation*}
$$

or

$$
\begin{equation*}
f(p, q)=\frac{1}{\lambda}\left[p^{\alpha} q^{\beta}+p^{\beta} q^{\alpha}-p-q\right] \tag{2.19}
\end{equation*}
$$

or

$$
\begin{equation*}
f(p, q)=\frac{1}{\lambda}\left[2 p^{\beta} q^{\beta} \cos \left(\alpha \log _{2} \frac{p}{q}\right)-p-q\right] \tag{2.20}
\end{equation*}
$$

where $\alpha, \beta$ are arbitrary real constants.
Proof. First, define $h: I_{0}^{2} \rightarrow \mathbf{R}$ by

$$
\begin{equation*}
h(p, q)=p+q+\lambda f(p, q) \tag{2.21}
\end{equation*}
$$

and rewrite (1.13) as

$$
\begin{equation*}
\sum_{i-1}^{n} \sum_{j-1}^{m}\left[h\left(p_{i} r_{j}, q_{i} s_{j}\right)+h\left(p_{i} s_{j}, q_{i} r_{j}\right)\right]=\sum_{i=1}^{n} h\left(p_{i}, q_{i}\right) \sum_{j-1}^{m} h\left(r_{j}, s_{j}\right) \tag{2.22}
\end{equation*}
$$

Note that $h$ in (2.21) is symmetric and measurable in each variable. If $h=0$, then $f$ has the form (2.18). From here on, let $h \neq 0$. For fixed $P, Q \in \Gamma_{n}^{0}$, defining $F: I_{0}^{2} \rightarrow \mathbf{R}$ by

$$
F(r, s)=\sum_{i=1}^{n}\left[h\left(p_{i} r, q_{i} s\right)+h\left(p_{i} s, q_{i} r\right)-h(r, s) h\left(p_{i}, q_{i}\right)\right]
$$

as in Theorem 1, we first obtain $F(r, s)=0$. Applying once again the same procedure, we obtain the functional equation (2.9) satisfied by $h$. It can be shown that the measurability of $h$ implies that of $M_{1}$ and $M_{2}$ in (2.10). Thus for real $M_{1}, M_{2}$, we get $M_{1}(p)=p^{\alpha}, M_{2}(q)=q^{\beta}$ [1], and from (2.10) we obtain (2.19). In the case where $M_{1}, M_{2}$ are complex, we get $M_{1}(p)=p^{\beta+i \alpha}=p^{\beta} e^{i \alpha \log _{e} p}$ and $M_{2}(q)=q^{\beta} e^{-i \alpha \log _{e} q} \quad$ [13], resulting in (2.20). The converse is easy to verify. This completes the proof of the theorem.

## 3. Characterization of Measures

In this section we display all regular symmetric measures having the sum form and the symmetric compositivity.

Theorem 3. Suppose that a sequence of measures $\left\{\mu_{n}\right\}$ has the sum form (1.12) with a measurable symmetric generating function $f: I_{0}^{2} \rightarrow \mathbf{R}$, and is symmetrically compositive for all pairs of positive integers $m, n(\leqslant 2)$. Then
$\mu_{n}(P, Q)=\sum_{k=1}^{n}\left[p_{k}\left(a \log _{2} p_{k}+b \log _{2} q_{k}\right)+q_{k}\left(a \log _{2} q_{k}+b \log _{2} p_{k}\right)\right]$
if $\lambda=0$, or else

$$
\begin{equation*}
\mu_{n}(P, Q)=\frac{1}{\lambda}\left[\sum_{k=1}^{n}\left(p_{k}^{\alpha} q_{k}^{\beta}+p_{k}^{\beta} q_{k}^{\alpha}\right)-2\right] \tag{3.2}
\end{equation*}
$$

or

$$
\begin{equation*}
\mu_{n}(P, Q)=-\frac{2}{\lambda}\left[1-\sum_{k=1}^{n}\left(p_{k} q_{k}\right)^{\beta} \cos \left(\alpha \log _{2} \frac{p_{k}}{q_{k}}\right)\right], \tag{3.3}
\end{equation*}
$$

or

$$
\begin{equation*}
\mu_{n}(P, Q)=-\frac{2}{\vec{\lambda}} \tag{3.4}
\end{equation*}
$$

where $a, b, \alpha$, and $\beta$ are arbitrary real constants.
Proof. Since $\mu_{n}$ satisfies the sum form and (1.11), we obtain the functional equation (1.13) for all pairs of positive integers $m, n$. Thus, using Theorem 1 (when $\lambda=0$ ) we obtain (3.1); using Theorem 2 (when $\lambda \neq 0$ ) we obtain (3.2), (3.3), or (3.4). This completes the proof of the theorem.

If the sequence $\left\{\mu_{n}\right\}$ is measuring "distance" between $P$ and $Q$, it is pleasant to assume that

$$
\mu_{n}(P, P)=0
$$

Corollary 1. A measure of distance $\left\{\mu_{n}\right\}$ satisfies the hypotheses in Theorem 3 and $\mu_{n}(P, P)=0$ if, and only if,

$$
\begin{equation*}
\mu_{n}(P, Q)=a J_{n}(P, Q) \tag{3.5}
\end{equation*}
$$

or

$$
\begin{equation*}
\mu_{n}(P, Q)=\frac{\left(2^{\alpha-1}-1\right)}{\lambda} J_{n, \alpha}(P, Q) \tag{3.6}
\end{equation*}
$$

or

$$
\begin{equation*}
\mu_{n}(P, Q)=-\frac{2}{\lambda}\left[1-\sum_{k=1}^{n}\left(p_{k} q_{k}\right)^{1 / 2} \cos \left(\alpha \log _{2} \frac{p_{k}}{q_{k}}\right)\right] . \tag{3.7}
\end{equation*}
$$

Proof. Since $\mu_{n}(P, P)=0$, (3.1) yields $a=-b$ and thus for $\lambda=0$ we get (3.5). Now using $\mu_{n}(P, P)=0$ in (3.2) we obtain $\beta=1-\alpha$ and hence (3.6). Again $\mu_{n}(P, P)=0$ in (3.3) yields $\beta=\frac{1}{2}$. Thus, one obtains (3.7). Since (3.4) does not satisfy the condition $\mu_{n}(P, P)=0, \mu_{n}$ corresponding to (3.4) does not mature. This completes the proof of the corollary.

## 4. Cosine Divergence

While $\mu_{n}$ in (3.5) and (3.6) are constant multiples of known measures such as $J$-divergence and $J$-divergence of degree $\alpha$, the measures $\mu_{n}$ in (3.7) seem to be new. Their standardization

$$
\begin{equation*}
D_{n, \alpha}(P, Q)=\frac{1}{2}\left[1-\sum_{k=1}^{n}\left(p_{k} q_{k}\right)^{1 / 2} \cos \left(\alpha \log _{2} \frac{p_{k}}{q_{k}}\right)\right], \tag{4.1}
\end{equation*}
$$

which we call the cosine divergence, has the following nice and desirable properties:
(a) Positive definite; $D_{n, \alpha}(P, Q) \geqslant 0$ and it is zero if, and only if, $P=Q$.
(b) Symmetric; $D_{n, \alpha}(P, Q)=D_{n, \alpha}(Q, P)$.
(c) Bounded from above; $D_{n, \alpha}(P, Q) \leqslant 1$.
(d) Subadditive; $D_{n m, \alpha}(P * R, Q * S)+D_{n m, \alpha}(P * S, Q * R) \leqslant$ $2 D_{n, \alpha}(P, Q)+2 D_{m, \alpha}(R, S)$.
(e) $D_{n, \alpha}$ can be extended continuously to the square of the space of complete distributions $\Gamma_{n}=\left\{P=\left(p_{1}, p_{2}, \ldots, p_{n}\right) \mid p_{k} \geqslant 0, \sum_{k=1}^{n} p_{k}=1\right\}$.

Properties (a) and (c) follow from Holder incquality and (d) follows from (1.11), (3.3), and (4.1). For fixed $Q=\left(\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right)$, we draw the level curves of $D_{3, \alpha}(P, Q)$ in the variables $p_{1}, p_{2}$ (Fig. 1).
Notice that if $\alpha=0, D_{n, \alpha}(P, Q)$ reduces to

$$
D_{n, \mathrm{a}}(P, Q)=\frac{1}{4} \sum_{k=1}^{n}\left(\sqrt{p_{k}}-\sqrt{q_{k}}\right)^{2}
$$

which is a multiple of the formerly known Jeffreys distance.
Measuring similarity between $P$ and $Q$ can be conceived as a dual to that of measuring distance. If a distance measure $\mu_{n}$ between probability distribution is bounded from above by a constant $b$ then $b-\mu_{n}$ is naturally a similarity measure. Applying this principle to $D_{n, \alpha}$, the similarity measure

$$
\begin{equation*}
S_{n, \alpha}(P, Q)=\frac{1}{2}\left[1+\sum_{k=1}^{n}\left(p_{k} q_{k}\right)^{1 / 2} \cos \left(\alpha \log _{2} \frac{p_{k}}{q_{k}}\right)\right] \tag{4.2}
\end{equation*}
$$



Figure 1
is obtained. This similarity measure has the following properties: (a) $S_{n, \alpha}(P, Q) \leqslant 1$ and equality holds if, and only if $P=Q$; (b) $S_{n, \alpha}(P, Q)-S_{n, \alpha}(Q, P)$; (c) $S_{n, \alpha}(P, Q) \geqslant 0$; (d) $S_{n, \alpha}$ is superadditive. This measure also satisfies most of the properties of a similarity measure listed in [23]. If $\alpha=0$, then

$$
S_{n, 0}(P, Q)=\frac{1}{2}\left[1+H_{n}(P, Q)\right]
$$

where $H_{n}(P, Q)$ is the Hellinger (or Bhattacharyya) coefficient (refer to $[4,9]$ ).

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