

# ON MIXTURES OF DISTRIBUTIONS<sup>1</sup>

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**1. Introduction.** Let  $F_\theta$ ,  $\theta \in T \subseteq \mathbb{R}^1$ , be a family of distribution functions on  $\mathbb{R}^1$ , where  $F_\theta$  is measurable in  $\theta$ . For an arbitrary non-degenerate distribution function  $H$  that assigns probability 1 to  $T$  we consider the  $H$ -mixture of the family  $F_\theta$ , i.e. the distribution function

$$F_H(x) = \int_T F_\theta(x) dH(\theta).$$

In 1943 W. Feller [1] proved that if  $F_\theta$  is Poisson, then the variance of  $F_H$  is always larger than the variance of the Poisson distribution with the same expectation. This note is an attempt to generalize this and related results.

**2. Convexity arguments.** If  $g_1$  and  $g_2$  are integrable with respect to  $F_\theta$  for all  $\theta \in T$  and with respect to  $F_H$  we define for  $i = 1, 2$

$$\begin{aligned}\chi_i(\theta) &= \int g_i(x) dF_\theta(x) \\ \chi_i(H) &= \int g_i(x) dF_H(x) = \int_T \chi_i(\theta) dH(\theta).\end{aligned}$$

We note that the same symbols  $\chi_i$  are used to denote both the functions  $\chi_i(\theta)$  and the functionals  $\chi_i(H)$ .

We shall say that the function  $\chi_2$  is convex, concave, or linear with respect to  $\chi_1$  on  $T$  if there exists a convex, concave, or linear function  $\varphi$  on  $\chi_1(T)$  such that

$$\chi_2(\theta) = \varphi(\chi_1(\theta)) \quad \text{for all } \theta \in T.$$

We note that this definition differs from the one usually given in that there is no monotonicity requirement for  $\chi_1$  involved.

The following version of Jensen's inequality will be needed in the sequel (cf. [2], p. 75).

LEMMA 1. Let  $\mathfrak{H}$  denote the class of distribution functions  $H$  on  $T$  having

$$\chi_1(H) = \chi_1(\theta_H) \quad \text{for some } \theta_H \in T.$$

Then a necessary and sufficient condition for  $\chi_2(H) \geq \chi_2(\theta_H)$  to hold for all  $H \in \mathfrak{H}$  is that  $\chi_2$  is convex with respect to  $\chi_1$  on  $T$ .

The inequality is strict for all  $H \in \mathfrak{H}$  for which  $\chi_1$  is not constant a.e.  $[H]$  on  $T$ , if and only if the convexity is everywhere strict; there is equality for all  $H \in \mathfrak{H}$  if and only if  $\chi_2$  is linear with respect to  $\chi_1$  on  $T$ . If in the above "convex" is replaced by "concave" the inequality is reversed.

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PROOF. If  $\chi_2 = \varphi\chi_1$  and  $\varphi$  is convex we have  $\chi_2(H) = \int_T \chi_2(\theta) dH(\theta) = \int_T \varphi(\chi_1(\theta)) dH(\theta) \geq \varphi(\int_T \chi_1(\theta) dH(\theta)) = \varphi(\chi_1(H)) = \varphi(\chi_1(\theta_H)) = \chi_2(\theta_H)$  by Jensen's inequality, which also yields the sufficiency of the conditions for strictness and equality. The converse follows by considering distributions  $H$  that concentrate at two points.

To apply Lemma 1 to a comparison of the variances of the distributions  $F_H$  and  $F_{\theta_H}$  where  $F_H$  and  $F_{\theta_H}$  have the same expectation, we set  $g_1(x) = x, g_2(x) = x^2$  and define

$$\begin{aligned} \mu_1(\theta) &= \int x dF_\theta(x), \\ \mu_1(H) &= \int x dF_H(x), \\ \mu_2(\theta) &= \int x^2 dF_\theta(x), \\ \mu_2(H) &= \int x^2 dF_H(x), \\ \sigma^2(\theta) &= \mu_2(\theta) - \mu_1^2(\theta), \end{aligned}$$

and 
$$\sigma^2(H) = \mu_2(H) - \mu_1^2(H).$$

COROLLARY 1. A necessary and sufficient condition for  $\sigma^2(H) \geq \sigma^2(\theta_H)$  to hold for all  $H$  having  $\mu_1(H) = \mu_1(\theta_H), \theta_H \in T$ , is that  $\mu_2$  is convex with respect to  $\mu_1$  on  $T$ .

We note that if  $\mu_1(\theta)$  is linear, the above condition reduces to convexity of  $\mu_2(\theta)$  on  $T$ . The result is then a direct generalization of Feller's theorem.

**3. Total positivity.** It turns out that for most well-known families  $F_\theta, \mu_2(\theta)$  is convex in  $\mu_1(\theta)$  and hence mixing increases the variance. As will be seen the reason for this is that many of these families possess totally positive densities, a concept that was extensively investigated by S. Karlin et al. (cf. e.g. [3], [4], [5]).

Suppose that the family  $F_\theta$  possesses densities  $p(x, \theta)$  with respect to a  $\sigma$ -finite measure  $\nu$  with spectrum  $X \subseteq R^1$ , i.e.

$$F_\theta(x) = \int_{-\infty}^x p(u, \theta) d\nu(u).$$

The density  $p(x, \theta)$  (or the family  $F_\theta$ ) is called totally positive of order  $k(TP_k)$ , if for all  $x_1 < x_2 < \dots < x_m$  in  $X, \theta_1 < \theta_2 < \dots < \theta_m$  in  $T$  and all  $1 \leq m \leq k$ , the determinant

$$\begin{vmatrix} p(x_1, \theta_1) & p(x_1, \theta_2) & \dots & p(x_1, \theta_m) \\ p(x_2, \theta_1) & p(x_2, \theta_2) & \dots & p(x_2, \theta_m) \\ \vdots & \vdots & \ddots & \vdots \\ p(x_m, \theta_1) & p(x_m, \theta_2) & \dots & p(x_m, \theta_m) \end{vmatrix} \geq 0.$$

Densities of this type possess the following variation diminishing property (cf. [5] for a statement of the general result as it is mentioned here; cf. [3] for a proof of a special case. The determinantal inequality on which this proof is based may also be used to provide a direct proof of the general result). Let  $V(g)$  denote the number of changes of sign of  $g$ . If  $\chi$  is given by the absolutely convergent integral

$$\chi(\theta) = \int g(x)p(x, \theta) d\nu(x) = \int g(x) dF_\theta(x),$$

where  $p(x, \theta)$  is  $TP_k$  and  $V(g) \leq k - 1$  then  $V(\chi) \leq V(g)$ . If  $V(\chi) = V(g)$  then  $g$  and  $\chi$  change sign in the same order.

The following lemma may be obtained by exploiting this variation diminishing property. It is a slightly different formulation of a result obtained by Karlin in [4], p. 343.

**LEMMA 2.** *If  $p(x, \theta)$  is  $TP_3$ ,  $g_1$  is monotone on  $X$ , and  $g_2$  is convex with respect to  $g_1$  on  $X$ , then  $\chi_2$  is convex with respect to  $\chi_1$  on  $T$ .*

Karlin proved the lemma for the case where  $g_1(x) = x$  and  $\chi_1$  is linear. He also showed (cf. [4], pp. 342–343) that if  $g_1$  is non-decreasing (non-increasing) on  $X$ , then so is  $\chi_1$  on  $T$ . Since the property of total positivity is preserved under non-decreasing (non-increasing) transformations  $g_1$  and  $\chi_1$  of the random variable and the parameter simultaneously, the lemma may be proved by reducing it to the special case considered by Karlin.

Combining Lemmata 1 and 2, we have

**THEOREM 1.** *If  $F_\theta$  is  $TP_3$ ,  $g_1$  is monotone on  $X$ ,  $g_2$  is convex with respect to  $g_1$  on  $X$ , and  $\chi_1(H) = \chi_1(\theta_H)$  for some  $\theta_H \in T$ , then  $\chi_2(H) \geq \chi_2(\theta_H)$ .*

Since  $g_1(x) = x$  is monotone and  $g_2(x) = x^2$  is convex, we have from Corollary 1

**COROLLARY 2.** *If  $F_\theta$  is  $TP_3$  and  $\mu_1(H) = \mu_1(\theta_H)$ ,  $\theta_H \in T$ , then  $\sigma^2(H) \geq \sigma^2(\theta_H)$ .*

Finally we remark that the conclusions of Lemma 2, Theorem 1 and Corollary 2 are obviously independent of a particular parametrization of the family  $F_\theta$ . It is therefore sufficient to require that there exists a parametrization such that the family is  $TP_3$ .

#### REFERENCES

- [1] FELLER, W. (1943). On a general class of contagious distributions. *Ann. Math. Statist.* **14** 389–400.
- [2] HARDY, G. H., LITTLEWOOD, J. E., and PÓLYA, G. (1934). *Inequalities*. Cambridge Univ. Press.
- [3] KARLIN, S. (1955). Decision theory for Pólya type distributions. Case of two actions, I. *Proc. Third Berkeley Symp. Math. Statist. Prob.* **1** 115–128, Univ. of California, Berkeley.
- [4] KARLIN, S. (1963). Total positivity and convexity preserving transformations. *Proc. Symp. Pure Math.* **7** 329–347, Amer. Math. Soc., Providence.
- [5] KARLIN, S., PROSCHAN, F., and BARLOW, R. E. (1961). Moment inequalities of Pólya frequency functions. *Pacific J. Math.* **11** 1023–1033.