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## On convexity preserving families of probability distribution

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## On convexity preserving families of probability distributions by W. R. VAN ZWET \*

Introduction

Let  $F_{\theta}$  be a family of probability distribution functions on  $R^1$  with parameter  $\theta \in T \subseteq R^1$ , and let X denote the union of the supports of these distributions. For  $k \ge 0$ , let  $\{g_0, g_1, \dots, g_{k+1}\}$  be a set of real-valued finite functions on X that are integrable with respect to  $F_{\theta}$  for all  $\theta \in T$  and define

$$\chi_i(\theta) = \int g_i(x) \, dF_{\theta}(x), \quad i = 0, 1, \dots, k+1. \tag{1.1}$$

Following S. KARLIN and W. J. STUDDEN in [3] with a minor modification, we shall say that  $\{g_0, g_1, \ldots, g_{k+1}\}$  constitute a weak complete Tchebycheff system (WCTsystem) if for each  $0 \le m \le k+1$  and all  $x_0 < x_1 < \ldots < x_m \in X$  the determinant

det 
$$[g_i(x_j)]_{i,j=0,...,m} \ge 0;$$
 (1.2)

the system is called a complete Tchebycheff system (CT-system) if the inequality is always strict. The difference between this definition of a WCT-system and the one given in [3] is that we retain the case where  $g_0, g_1, \ldots, g_m$  are linearly dependent on X for some  $m \leq k+1$ ; in that case any choice of  $g_{m+1}, \ldots, g_{k+1}$  will trivially satisfy definition (1.2). We shall also express inequalities (1.2) by saying that  $g_{k+1}$  is generalized convex with respect to the WCT-system  $\{g_0, ..., g_k\}$ .

Our discussion of WCT-systems will involve the related concept of total positivity (cf. [1]). A function  $f(x,\theta)$  on  $X \times T$  is said to be totally positive of order  $n(TP_n)$  if for every  $1 \le m \le n$ , all  $x_1 < x_2 < \ldots < x_m \in X$  and all  $\theta_1 < \theta_2 < \ldots < \theta_m \in T$ ,  $\det \left[f(x_i, \theta_i)\right]_{i, i=1, \dots, m} \ge 0.$ (1.3)

The first question that comes to mind in this context is whether one can find conditions on the family  $F_{\theta}$  that ensure that  $\{\chi_0, \chi_1, \dots, \chi_{k+1}\}$  will be a WCT-system on T whenever  $\{g_0, g_1, ..., g_{k+1}\}$  constitutes a WCT-system on X. If the family  $F_{\theta}$  possesses densities  $p(x, \theta)$  with respect to a  $\sigma$ -finite measure  $\mu$  with spectrum X and hence

 $\chi_i(\theta) = \int g_i(x) \, p(x,\theta) \, d\mu(x),$ 

this question is easily answered. We have for each  $0 \le m \le k+1$  (cf. [1])

 $\det \left[\chi_i(\theta_j)\right] = \int \ldots \int \det \left[g_i(x_j)\right] \det \left[p(x_i, \theta_j)\right] d\mu(x_0) \ldots d\mu(x_m)$  $x_0 < x_1 < \ldots < x_m$ 

where in each determinant i and j run from 0 to m. It follows that the condition that p is  $TP_{k+2}$  is certainly sufficient; since we require that  $\{\chi_0, \ldots, \chi_{k+1}\}$  will inherit the WCT-property for every WCT-system  $\{g_0, \ldots, g_{k+1}\}$ , the condition is essentially also

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necessary (by "essentially" is meant that for any  $\theta_1 < ... < \theta_m$  the defining inequality (1.3) need not hold on a set of product-measure 0). We note that the fact that  $F_{\theta}$ are probability distribution functions is not used in establishing the condition. In view of this general result it is hardly surprising that recent discussions of convexity preserving properties (cf. [1] and [2]) have been confined to families of den-

sities that are totally positive of the appropriate order. However, one usually does

not discuss the class of all WCT-systems of a given order but restricts attention to a relatively small subclass (e.g. the case where  $g_i = g_1^i$  for i = 0, 1, ..., k). Also one often imposes additional restrictions on the family  $F_{\theta}$  to ensure that for those systems  $\{g_0, ..., g_{k+1}\}$  that are considered,  $\{\chi_0, ..., \chi_{k+1}\}$  will also belong to some restricted class.

In sections 3 and 4 of this paper we investigate how far the  $TP_{k+2}$  condition for p can be relaxed in two such restricted cases that seem to be important in practice. Like section 1, the second section is of an expository character.

#### 2. Convexity of order k

Let f be a real-valued finite function defined on an arbitrary set  $Y \subseteq R^1$ . For  $k \ge 0$ we shall say that f is convex of order k  $(C_k)$  if f is generalized convex with respect to the CT-system  $\{1, y, y^2, ..., y^k\}$ , i.e. if for all  $y_1 < y_2 < ... < y_{k+2}$ 

$$D_{f}(y_{1}, ..., y_{k+2}) = \begin{vmatrix} 1 & y_{1} & y_{1}^{2} & \dots & y_{1}^{k} & f(y_{1}) \\ 1 & y_{2} & y_{2}^{2} & \dots & y_{2}^{k} & f(y_{2}) \\ \vdots & & & \vdots \\ 1 & y_{k+2} & y_{k+2}^{2} & \dots & y_{k+2}^{k} & f(y_{k+2}) \end{vmatrix} \ge 0.$$
(2.1)

For k = 0,1, (2.1) reduces to the ordinary definitions of non-decreasing or (measurable) convex functions. Generally speaking (2.1) is an extension of the concept of non-negative (k+1)-th derivative.  $C_k$  functions were extensively studied by T. POPO-VICIU in [6]. We note that S. KARLIN [1] refers to  $C_k$  functions as convex of order (k+1).

If  $P_m$  denotes a polynomial of degree at most m, then equivalent definitions of the  $C_k$  property are obviously:

- (A) (cf. [1]). For every P<sub>k</sub>, f−P<sub>k</sub> changes sign at most (k+1) times on Y. If it does have (k+1) sign-changes, the signs occur in the order (−)<sup>k+1</sup>, (−)<sup>k</sup>, ..., +, −, + for increasing values of the argument.
- (B) For every  $y_1 < y_2 < ... < y_{k+2} \in Y$ , the  $P_{k+1}$  having  $P_{k+1}(y_i) = f(y_i)$ , i = 1, 2, ..., k+2, has non-negative coefficient for its (k+1)-th degree term.

There is also a close connection with differences. Let

$$\Delta_{h}^{1} f(y) = f(y+h) - f(y)$$
  
$$\Delta_{h}^{m} f(y) = \Delta_{h}^{1} \Delta_{h}^{m-1} f(y) = \sum_{j=0}^{m} (-1)^{m-j} {m \choose j} f(y+jh)$$
(2.2)

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# and generally $\Delta_{h_1,\ldots,h_m}^m f(y) = \Delta_{h_m}^1 \Delta_{h_1,\ldots,h_{m-1}}^{m-1} f(y) = \sum_{j=0}^m (-1)^{m-j} \sum_{1 \le i_1 < i_2 < \ldots < i_j \le m} f(y + \sum_{\nu=1}^{j} h_{i_\nu}).$ (2.3) Furthermore let

$$D_{f}^{*}(y_{1}, ..., y_{k+2}) = \frac{D_{f}(y_{1}, ..., y_{k+2})}{\prod (y_{j} - y_{i})}; \qquad (2.4)$$

 $1 \leq i < j \leq k+2$ 

since the denominator is positive for  $y_1 < y_2 < ... < y_{k+2}$ ,  $D_f$  may be replaced by  $D_f^*$  in definition (1.1). The following relation between  $D_f^*$  and differences may be proved by induction on k.

#### Lemma 2.1

If  $\Pi$  denotes the set of permutations  $\pi = [\pi(1), \pi(2), ..., \pi(k+1)]$  of the numbers 1, 2, ..., k+1, then

$$\Delta_{h_1,\ldots,h_{k+1}}^{k+1} f(y) = \prod_{i=1}^{k+1} h_i \sum_{\pi \in \Pi} D_f^* \left( y, y + h_{\pi(1)}, \ldots, y + \sum_{\nu=1}^{k+1} h_{\pi(\nu)} \right).$$
(2.5)

We note that for  $h_1 = h_2 = \ldots = h_{k+1} = h$ , (2.5) reduces to  $\Delta_h^{k+1} f(y) = (k+1)! h^{k+1} D_f^*[y, y+h, ..., y+(k+1)h].$ (2.6)

It follows from lemma 2.1 that if f is  $C_k$  on Y, then for all  $h_1, h_2, \ldots, h_{k+1} > 0$ ,  $\Delta_{h_1,\ldots,h_{k+1}}^{k+1}f(y) \ge 0$ (2.7)

whenever defined, i.e. whenever all  $y + \sum_{v=1}^{j} h_{i_v} \in Y$ .

In the special case that Y is an interval there is also a converse result and the following definition of the  $C_k$  property is equivalent to (2.1) in this case:

(C) f is (Lebesgue)-measurable and for  $h > 0, y \in Y, y+(k+1)h \in Y$ ,

$$\Delta_h^{k+1} f(y) \ge 0. \tag{2.8}$$

In this case, however, the  $C_k$  property is hardly a generalization of non-negative (k+1)-th derivative at all. In fact, if Y is an open interval and  $k \ge 1$ , definition (2.1) ensures continuity of f on Y and is equivalent to

(D) f is (k-1) times continuously differentiable and  $f^{(k-1)}$  is convex on Y.

Finally we consider the special case where Y is a set of consecutive integers. For integer h > 0

$$\Delta_{h}^{k+1} f(y) = \sum_{h_{1}=0}^{h-1} \dots \sum_{h_{k+1}=0}^{h-1} \Delta_{1}^{k+1} f\left(y + \sum_{j=1}^{k+1} h_{j}\right).$$
(2.9)

Combining (2.6) and (2.9) we find that the  $C_k$  property may be defined in this case by

- For all  $y, y+k+1 \in Y$ **(E)** 
  - $\Delta_1^{k+1} f(y) \ge 0.$ (2.10)

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For further details concerning the definitions given above the reader is referred to [6]. Let  $f_1$  and  $f_2$  be real-valued finite functions on Y. We shall say that  $f_2$  is  $C_k$  with respect to  $f_1$  on Y if there exists a  $C_k$  function f on  $f_1(Y)$  such that  $f_2 = f(f_1)$  on Y. If  $f_1$  is non-decreasing on Y and  $f_2$  is constant on any set where  $f_1$  is constant, this reduces to

 $\begin{vmatrix} 1 & f_1(y_1) & f_1^2(y_1) & \dots & f_1^k(y_1) & f_2(y_1) \\ 1 & f_1(y_2) & f_1^2(y_2) & \dots & f_1^k(y_2) & f_2(y_2) \\ \vdots & & & \vdots \\ 1 & f_1(y_{k+2}) & f_1^2(y_{k+2}) & \dots & f_1^k(y_{k+2}) & f_2(y_{k+2}) \end{vmatrix} \ge 0$ (2.11)

for all  $y_1 < y_2 < ... < y_{k+2} \in Y$ .

#### 3. Preserving convexity of order k

Returning to the setup of section 1, we let g be a real-valued finite function on X that is integrable with respect to  $F_{\theta}$  for all  $\theta \in T$  and define

 $\chi(\theta) = \int g(x) \, dF_{\theta}(x).$ 

We shall say that the family  $F_{\theta}$  preserves convexity of order k if  $\chi$  is  $C_k$  on T whenever g is  $C_k$  on X, i.e. whenever g is generalized convex with respect to  $\{1, x, ..., x^k\}$  then  $\chi$  is generalized convex with respect to  $\{1, \theta, ..., \theta^k\}$ . In [1] S. KARLIN has shown that if densities  $p(x, \theta)$  with respect to  $\mu$  exist, then a sufficient condition for  $F_{\theta}$  to preserve

convexity of order k is that p is  $TP_{k+2}$  and that whenever g is a polynomial of exact degree  $m \leq k$ , then  $\chi$  is also a polynomial of exact degree m. According to the result of section 1 the first part of this condition ensures that  $\chi$  is generalized convex with respect to the WCT-system

 $\int x^i dF_{\theta}(x), \qquad i=0, 1, ..., k,$ 

whereas the second part ensures that this is equivalent to generalized convexity with respect to  $\{1, \theta, ..., \theta^k\}$ .

However, this condition is not necessary. For k = 0 a condition that is necessary as well as sufficient was given by J. KRZYZ in [4]:

#### Lemma 3.1

 $\chi$  is non-decreasing on T whenever g is non-decreasing on X if and only if the family  $F_{\theta}$  is stochastically increasing (i.e.  $F_{\theta}(x)$  is non-increasing in  $\theta$  for every fixed x). Since the  $TP_2$  property of p is equivalent to monotone likelihood ratio, KRZYZ's condition is weaker than KARLIN's for k = 0 (cf. [5]).

For general k it is also easy to find a necessary and sufficient condition, provided that we restrict attention to those  $C_k$  functions g that can be extended to a  $C_k$  function on an open interval containing X. Since the convex functions constitute a convex cone spanned by the linear functions and functions of the form

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$$h(x) = \begin{cases} 0 & \text{for } x \leq x_0 \\ x - x_0 & \text{for } x > x_0 \end{cases}$$

we find from definition D of section 2 that the convex cone of  $C_k$  functions is spanned by the polynomials  $P_k$  of degree at most k and functions of the form

$$h_k(x) = \begin{cases} 0 & \text{for } x \leq x_0 \\ (x - x_k)^k & \text{for } x > x \end{cases}$$

#### 101 $( \land )$ $\lambda_0$ $x > x_0$ .

For k = 0 this is obviously also true. It follows that it is sufficient as well as necessary to require that  $\chi$  be  $C_k$  whenever g is of one of the forms mentioned above. However, if g is a polynomial of degree at most k, then so is -g and as a result both  $\chi$  and  $-\chi$ are required to be  $C_k$ , which implies that  $\chi$  is also a polynomial of degree at most k. Hence we have proved

#### Lemma 3.2

 $\chi$  is  $C_k$  on T whenever g is  $C_k$  on an open interval containing X, if and only if for every  $x_0$ 

 $\int (x - x_0)^k dF_{\theta}(x)$ 

is  $C_k$  on T and whenever g is a polynomial of degree at most k the same holds for  $\chi$ . We note that for  $k \leq 1$  the condition that the  $C_k$  function g can be extended to a  $C_k$  function on an open interval containing X is always satisfied. For k = 0 the lemma reduces to lemma 3.1.

Although for  $k \ge 1$  lemma 3.2 seems to be fairly useless for practical purposes, the results obtained so far do seem to indicate that there exists a large class of  $C_k$  preserving families that do not possess any total positivity properties. The results in the remainder of this section exhibit a number of these families.

#### Theorem 3.1

Let  $F_0$  and F be distribution functions with characteristic functions  $\varphi_0$  and  $\varphi$  respectively, and suppose that F is infinitely divisible and has F(-0) = 0. If for  $t \ge 0$ ,  $F_t$ denotes the distribution function corresponding to  $\varphi_0.\varphi^t$ , then the family  $F_t$ ,  $0 \leq t < \infty$ , preserves convexity of all orders.

#### Proof

Let  $G_t$  denote the distribution function corresponding to  $\varphi^t$  and let  $X_t$  ( $t \ge 0$ ) be a stochastic process with non-negative stationary independent increments for which  $X_0, X_{s+t} - X_s$  and  $X_t$  (s,  $t \ge 0$ ) have distribution functions  $F_0, G_t$  and  $F_t$  respectively. For fixed  $t \ge 0$  and h > 0 define

$$Z_i = X_{t+ih} - X_{t+(i-1)h}, \quad i = 1, 2, ..., k+1.$$

 $Z_1, Z_2, \ldots, Z_{k+1}$  are independent and identically distributed random variables that are also independent of  $X_t$ . Hence, because of the exchangeability of  $Z_1, \ldots, Z_{k+1}$ ,

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$$E\left[\sum_{j=0}^{k+1} (-1)^{k+1-j} {k+1 \choose j} g(X_{t+jh}) \mid X_t = x\right] =$$

$$= E\left[\sum_{j=0}^{k+1} (-1)^{k+1-j} {k+1 \choose j} g(x+Z_1+\ldots+Z_j)\right] =$$

$$= E\left[\sum_{j=0}^{k+1} (-1)^{k+1-j} \sum_{1 \le i_1 < \ldots < i_j \le k+1} g(x+Z_{i_1}+\ldots+Z_{i_j})\right] =$$

$$= E\left[\varDelta_{Z_1,\ldots,Z_{k+1}}^{k+1} g(x)\right].$$

Since  $Z_1, ..., Z_{k+1} \ge 0$  with probability 1, the last expression is non-negative for every  $C_k$  function g and all x by (2.7). As a result

$$\Delta_h^{k+1} \chi(t) = E\left[\sum_{j=0}^{k+1} (-1)^{k+1-j} {k+1 \choose j} g(X_{t+jh})\right] \ge 0$$

for all  $t \ge 0$  and h > 0. As  $\chi$  is a measurable function defined on the interval  $[0, \infty)$ , it is  $C_k$  by definition C of section 2.

If we consider only integer values of t in theorem 3.1, we may drop the assumption that F is infinitely divisible without affecting the proof. The  $C_k$  character of  $\chi$  on the integers now follows from  $\Delta_1^{k+1} \chi \ge 0$  by definition E of section 2. Specializing to the case where  $F_0$  is degenerate at 0 we obtain

#### Corollary 3.1

Every family  $F_n$ , n = 1, 2, ..., of *n*-fold convolutions of a distribution function  $F_1$ 

having  $F_1(-0) = 0$  preserves convexity of every order.

We note that the fact that  $F_n$  preserves convexity or order k was proved by S. KAR-LIN and F. PROSCHAN in [2] under the additional assumption that  $F_1$  possesses a density p that is a Pólya frequency density of order k+2 (i.e. p(x-y) is  $TP_{k+2}$  in x and y).

Another special case of theorem 3.1 is obtained by assuming F to be degenerate at 1, in which case the theorem reduces to:

Every location parameter family  $F_{\theta}(x) = G(x-\theta), -\infty < \theta < \infty$ , preserves convexity of every order.

This result is of course rather trivial. Without invoking theorem 3.1, it follows at once from

$$\Delta_h^{k+1}\chi(\theta) = \Delta_h^{k+1}\int g(x+\theta)\,dG(x) = \int \Delta_h^{k+1}g(x+\theta)\,dG(x).$$

In the same manner one easily verifies:

Every scale parameter family  $F_{\theta}(x) = G(x/\theta)$ ,  $0 < \theta < \infty$ , preserves convexity of every odd order. If moreover G(-0) = 0, then the family preserves convexity of all orders.

## 4. Invariant convexity preserving families

Let  $g_1, g_2, \chi_1$  and  $\chi_2$  be defined as in section 1. We shall say that the family  $F_{\theta}$  is invariant convexity preserving if, whenever  $g_1$  is non-decreasing and  $g_2$  is convex with

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respect to  $g_1$  on X, then  $\chi_1$  is non-decreasing and  $\chi_2$  is convex with respect to  $\chi_1$  on T. In terms of WCT-systems we may express this property by requiring that for every WCT-system of the form  $\{1, g_1, g_2\}$  the corresponding system  $\{1, \chi_1, \chi_2\}$  is also a WCT-system.

In the first place this definition asserts that the family  $F_{\theta}$  preserves the monotonicity of  $g_1$  and hence by lemma 1 the family is stochastically increasing;  $F_{\theta}$  also preserves

convexity (of order 1) provided that the parameter is subjected to a suitable nondecreasing transformation

 $\eta = \eta(\theta) = \int x \, dF_{\theta}(x).$ 

Moreover, this convexity preserving property is invariant under non-decreasing transformations  $g_1$  of the random variable, the appropriate monotone transformation of  $\theta$  then becoming  $\chi_1$ . It is precisely because of this invariance that we do not require that  $F_{\theta}$  be convexity preserving with respect to  $\theta$  itself, i.e. that  $\eta$  be linear in  $\theta$ . This property would be destroyed by non-linear transformations  $g_1$  anyway and would only result in fixing a possibly awkward parametrization.

From the general result of section 1 it follows that  $F_{\theta}$  is invariant convexity preserving if the density p is  $TP_3$ . The following theorem provides a necessary and sufficient condition.

#### Theorem 4.1

Define  $\overline{F}_{\theta}(x) = 1 - F_{\theta}(x)$ . The family  $F_{\theta}$  is invariant convexity preserving if and only if  $\{1, \overline{F}_{\theta}(x_1), \overline{F}_{\theta}(x_2)\}$  is a WCT-system on T for every fixed pair  $x_1 < x_2$ .

#### Proof

The condition asserts that for  $x_1 < x_2$  and  $\theta_0 < \theta_1 < \theta_2$ ,

$$\begin{vmatrix} 1 & \overline{F}_{\theta_{0}}(x_{1}) \\ 1 & \overline{F}_{\theta_{1}}(x_{1}) \end{vmatrix} \geq 0, \begin{vmatrix} 1 & \overline{F}_{\theta_{0}}(x_{1}) & \overline{F}_{\theta_{0}}(x_{2}) \\ 1 & \overline{F}_{\theta_{1}}(x_{1}) & \overline{F}_{\theta_{1}}(x_{2}) \\ 1 & \overline{F}_{\theta_{2}}(x_{1}) & \overline{F}_{\theta_{2}}(x_{2}) \end{vmatrix} \geq 0.$$
(4.1)

The first inequality means that  $F_{\theta}$  is stochastically increasing and we have already remarked that this is necessary and sufficient for  $\chi_1$  to be non-decreasing whenever  $g_1$ is. We may therefore assume that  $\overline{F}_{\theta}(x)$  is non-decreasing in  $\theta$  for every fixed x and restrict attention to the second inequality.

Let  $g_1$  be non-decreasing and let  $g_2 = f(g_1)$  where f is convex on  $g_1(X)$ . Since a convex function can be extended to a convex function on an interval, the same reasoning that we used in the proof of lemma 3.2 shows that we need only be concerned with functions f that are linear and functions f of the form

$$f_{\alpha}$$
  $f_{\alpha}$   $f_{\alpha}$   $f_{\alpha}$ 

$$f(y) = \begin{cases} 0 & \text{for } y \le y_0 \\ y - y_0 & \text{for } y > y_0. \end{cases}$$
(4.2)

Without loss of generality we may assume that  $y_0 = g_1(x_0) \in g_1(X)$ . For linear  $f, \chi_2$  is linear and hence convex with respect to  $\chi_1$ . Only functions f of the form (4.2)

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remain to be considered and as a result we have the following necessary and sufficient condition for a stochastically increasing family  $F_{\theta}$  to be invariant convexity preserving.

For every non-decreasing  $g_1$  and every  $x_0 \in X$ ,

 $\chi_2(\theta) = \int_{x_0}^{\infty} \left[ g_1(x) - g_1(x_0) \right] dF_{\theta}(x)$ 

is convex with respect to  $\chi_1(\theta)$ .

By an approximation argument one shows that it is sufficient to consider only those functions  $g_1$  that are left-continuous, non-decreasing step-functions assuming finitely many values. But then the above condition becomes:

For all  $m = 1, 2, ..., all x_1 < x_2 < ... < x_m$ , all  $\alpha_i > 0$ , i = 1, 2, ..., m, all  $1 \leq i_0 \leq m$  and all c,

$$\sum_{i=i_0}^{m} \alpha_i F_0(x_i)$$
(4.3)

is convex with respect to

 $\sum_{i=1}^{m} \alpha_i F_0(x_i) + c.$ (4.4)

Since (4.4) is non-decreasing in  $\theta$  and (4.3) is constant on any set where (4.4) is con-

stant, the determinantal convexity definition (2.11) for k = 1 applies. By subtracting from the second column in this determinant we find that convexity of (4.3) with respect to (4.4) is equivalent to

$$\begin{vmatrix} 1 & \sum_{i=1}^{i_0-1} \alpha_i \overline{F}_{\theta_0}(x_i) & \sum_{i=i_0}^m \alpha_i \overline{F}_{\theta_0}(x_i) \\ \vdots & \vdots \\ 1 & \sum_{i=1}^{i_0-1} \alpha_i \overline{F}_{\theta_2}(x_i) & \sum_{i=i_0}^m \alpha_i \overline{F}_{\theta_2}(x_i) \end{vmatrix} = \\ = \sum_{i=1}^{i_0-1} \sum_{j=i_0}^m \alpha_i \alpha_j \begin{vmatrix} 1 & \overline{F}_{\theta_0}(x_i) & \overline{F}_{\theta_0}(x_j) \\ \vdots & \vdots \\ 1 & \overline{F}_{\theta_2}(x_i) & \overline{F}_{\theta_2}(x_j) \end{vmatrix} \ge 0.$$

By choosing  $i_0 = m = 2$  we find that condition (4.1) is necessary; since every term in (4.5) has  $x_i < x_j$  it is also sufficient. This completes the proof of the theorem.

It may be of interest to compare the sufficient condition that  $F_{\theta}$  possesses a  $TP_3$ density  $p(x,\theta)$  with the necessary and sufficient condition of the theorem. One easily shows that the TP<sub>3</sub> assumption for p implies that  $\overline{F}_{\theta}(x)$  is TP<sub>3</sub>, or

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(4.5)

# $\frac{\overline{F}_{\theta_{0}}(x_{0})}{\overline{F}_{\theta_{1}}(x_{0})} \quad \frac{\overline{F}_{\theta_{0}}(x_{1})}{\overline{F}_{\theta_{1}}(x_{1})} \geq 0, \quad \frac{\overline{F}_{\theta_{0}}(x_{0})}{\overline{F}_{\theta_{1}}(x_{0})} \quad \frac{\overline{F}_{\theta_{0}}(x_{1})}{\overline{F}_{\theta_{1}}(x_{1})} \quad \frac{\overline{F}_{\theta_{0}}(x_{2})}{\overline{F}_{\theta_{2}}(x_{1})} \geq 0, \quad (4.6)$

for  $x_0 < x_1 < x_2$  and  $\theta_0 < \theta_1 < \theta_2$ . By letting  $x_0$  tend to  $-\infty$  we see that (4.6) implies (4.1). Hence the condition that  $\overline{F}_{\theta}(x)$  be  $TP_3$  is also sufficient for  $F_{\theta}$  to be in-

variant convexity preserving.

If we restrict ourselves to the special case where the parameter set T is an interval and  $\overline{F}_{\theta}(x)$  is differentiable with respect to  $\theta$ , it turns out that theorem 4.1 involves a  $TP_2$  instead of a  $TP_3$  condition.

#### Theorem 4.2

Let T be an interval and let  $q(x,\theta) = (\partial/\partial\theta) F_{\theta}(x)$  be defined on T for all x. Then the family  $F_{\theta}$  is invariant convexity preserving if and only if q is  $TP_2$ .

#### Proof

The first inequality in (4.1) is equivalent to  $q \ge 0$ . Since  $\overline{F}_{\theta}(x_2)$  is constant on any set where  $\overline{F}_{\theta}(x_1) + \overline{F}_{\theta}(x_2)$  is constant and the latter is non-decreasing in  $\theta$ , the second inequality of (4.1) asserts that  $\overline{F}_{\theta}(x_2)$  is convex with respect to  $\overline{F}_{\theta}(x_1) + \overline{F}_{\theta}(x_2)$ . This in turn is equivalent to  $q(x_1, \theta_1) q(x_2, \theta_2) - q(x_1, \theta_2) q(x_2, \theta_1) \ge 0$  for  $x_1 < x_2$  and  $\theta_1 < \theta_2$ . It is tempting to ask whether theorem 4.2 can be generalized. One conceivable generalization would deal with invariant  $C_k$  preserving families  $F_{\theta}$ , i.e. families for

which  $\chi_1$  is non-decreasing and  $\chi_2$  is  $C_k$  with respect to  $\chi_1$  whenever  $g_1$  is non-decreasing and  $g_2$  is  $C_k$  with respect to  $g_1$ . However, even a cursory inspection shows that only trivial examples of such families exist. The necessary requirement that  $\chi_2$  be a polynomial in  $\chi_1$  of degree at most k whenever  $g_2$  is a polynomial in  $g_1$  of degree at most k, can not be satisfied for every non-decreasing  $g_1$  except in a trivial manner.

A more promising generalization is to consider families  $F_{\theta}$  that transform WCTsystems  $\{1, g_1, ..., g_{k+1}\}$  into WCT-systems  $\{1, \chi_1, ..., \chi_{k+1}\}$ . If one restricts attention to the case where X and T are intervals and  $g_i$  and  $F_{\theta}$  satisfy certain regularity conditions, one shows in a fairly straightforward manner that a necessary and sufficient condition on  $F_{\theta}$  is that q be  $TP_{k+1}$ , thus generalizing lemma 3.1 and theorem 4.2 to the case where  $k \ge 2$ . We may conclude that although something may be lost for  $k \ge 2$ , the basic reason that theorems 4.1 and 4.2 work is not the fact that k = 1in that case, but that  $g_0 \equiv 1$  and that  $F_{\theta}$  are probability distribution functions.

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