

To prove (a), note that for  $|t_i| < \alpha_i, x_1 \geq 0$ , we have

$$(4) \quad \iint_{u_i \geq x_i} dF_n(u_1, u_2) \leq \exp(-|t_1|x_1 - |t_2|x_2)$$

$$\iint_{u_i \geq x_i} \exp(|t_1|u_1 + |t_2|u_2) dF_n(u_1, u_2) \leq M_0 \exp(-|t_1|x_1 - |t_2|x_2),$$

where  $\varphi_n(|t_1|, |t_2|) \leq M_0$ . Such a number  $M_0 = M_0(t_1, t_2)$  exists since  $\{\varphi_n(|t_1|, |t_2|)\}$  converges for  $|t_i| < \alpha_i$ . This gives an estimate for  $M_n(x_1, x_2)$ , which shows that (a) holds. The Helly selection principle ([2], pp. 60-62 and 83) leads to (b). The relations (c) and (d) follow immediately from Theorem 1.

From Theorems 1 and 2 we obtain

**THEOREM 3.** *Let  $\{F_n(x_1, x_2)\}$  be a sequence of df's and let  $\{\varphi_n(t_1, t_2)\}$  be the corresponding sequence of mgf's which are all assumed to exist for  $|t_i| < \alpha_i$ . Then the necessary and sufficient condition for the convergence of  $\{\varphi_n(t_1, t_2)\}$  for  $|t_i| < \alpha_i$  is that the relations (a) and (b) of Theorem 2 be satisfied.*

REFERENCES

[1] W. KOZAKIEWICZ, "On the convergence of sequences of moment generating functions," *Annals of Math. Stat.*, Vol. 18 (1947), pp. 61-69.  
 [2] H. CRAMÉR, *Mathematical Methods of Statistics*, Princeton University Press, 1946.

A NOTE ON THE MAXIMUM VALUE OF KURTOSIS

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In "A note on skewness and kurtosis," J. E. Wilkins (*Annals of Math. Stat.* Vol. 15 (1944), pp. 333-335) gave a short and elegant proof of the inequality for skewness and kurtosis

$$(1) \quad \beta_2 \geq \beta_1^2 + 1.$$

Then he gave an upper bound, depending on the population size  $N$ , for the skewness:

$$(2) \quad \max \beta_1 = (N - 2)/(N - 1)^{1/2}.$$

Now we shall derive an upper bound for the kurtosis. It will appear that the sign "=" in (1) is valid for the upper bounds, and the two maximum values indeed arise in the same "extreme" population.

To find the maximum value of the kurtosis  $\beta_2$  we consider the function  $\Sigma x_i^4$  in the  $x$ -space, where  $\Sigma x_i^2 = N$  and  $\Sigma x_i = 0$ . We have to maximize  $\Sigma x_i^4 - \lambda \Sigma x_i^2 - \mu \Sigma x_i$ . The maximizing values are given by the  $N$  equations, found by differentiation with respect to  $x_i$

$$(3) \quad 4x_i^3 - 2\lambda x_i - \mu = 0,$$

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together with the two relations  $\Sigma x_i^2 = N$  and  $\Sigma x_i = 0$ . Multiplication with  $x_i$  and summation over all  $N$  equations give

$$4N\beta_2 - 2N\lambda = 0;$$

hence

$$\max \beta_2 = \frac{1}{2}\lambda.$$

Since it is not possible that all values of  $x$  in the population are equal, the equation (3) of third degree must have at least two different real roots, and hence it has three real roots, which we may represent by

$$\sqrt{\frac{2\lambda}{3}} \cos \alpha, \quad \sqrt{\frac{2\lambda}{3}} \cos \left(\alpha + \frac{2}{3}\pi\right), \quad \sqrt{\frac{2\lambda}{3}} \cos \left(\alpha + \frac{4}{3}\pi\right),$$

where  $\cos 3\alpha = 4 \cos^3 \alpha - 3 \cos \alpha = \mu / (2\lambda/3)^{3/2}$ . Suppose the number of those roots is, respectively,  $k$ ,  $l$ , and  $m$ , with sum  $N$ . Writing  $v$  for  $l - m$ , we have, since not all values in the population can be the same,

$$\left\{ \begin{array}{l} 1 \leq k \leq N - 1 \\ |v| \leq N - k \end{array} \right\} \quad \text{or} \quad \left\{ \begin{array}{l} k = 0 \\ |v| \leq N - 2 \end{array} \right\},$$

with  $v$  even if  $N - k$  is even and  $v$  odd if  $N - k$  is odd.  $\Sigma x_i = 0$  gives, since  $\lambda \neq 0$ ,

$$(k - \frac{1}{2}l - \frac{1}{2}m) \cos \alpha + (\frac{1}{2}\sqrt{3}l - \frac{1}{2}\sqrt{3}m) \sin \alpha = 0,$$

$$(\frac{3}{2}k - \frac{1}{2}N) \cos \alpha + \frac{1}{2}\sqrt{3}v \sin \alpha = 0,$$

$$\tan \alpha = - \frac{3k - N}{v\sqrt{3}},$$

$$\sin \alpha = \frac{3k - N}{\sqrt{(3k - N)^2 + 3v^2}},$$

$$\cos \alpha = \frac{-v\sqrt{3}}{\sqrt{(3k - N)^2 + 3v^2}}.$$

Hence our second relation  $\Sigma x_i^2 = N$  gives

$$\begin{aligned} \frac{3N}{2\lambda} &= k \cos^2 \alpha + l(-\frac{1}{2} \cos \alpha + \frac{1}{2}\sqrt{3} \sin \alpha)^2 + m(-\frac{1}{2} \cos \alpha - \frac{1}{2}\sqrt{3} \sin \alpha)^2 \\ &= k \cos^2 \alpha + (l + m)(\frac{1}{4} \cos^2 \alpha + \frac{3}{4} \sin^2 \alpha) - (l - m) \cdot \frac{1}{2}\sqrt{3} \sin \alpha \cos \alpha \\ &= k \frac{3v^2}{(3k - N)^2 + 3v^2} + (N - k) \frac{\frac{3}{4}v^2 + \frac{3}{4}(3k - N)^2}{(3k - N)^2 + 3v^2} \\ (4) \qquad &+ v \cdot \frac{1}{2}\sqrt{3} \frac{(3k - N) v\sqrt{3}}{(3k - N)^2 + 3v^2} \\ &= \frac{9}{4}k - \frac{1}{4}N - \frac{(3k - N)^3}{(3k - N)^2 + 3v^2}. \end{aligned}$$

To find the maximum kurtosis we have to find the minimum possible value of the expression (4). Partial differentiation of (4)  $\frac{\partial}{\partial(3k - N)}$  gives

$$\begin{aligned} & \frac{3}{4} - \frac{3(3k - N)^2}{(3k - N)^2 + 3v^2} + \frac{2(3k - N)^4}{\{(3k - N)^2 + 3v^2\}^2} \\ &= \frac{-\frac{1}{4}(3k - N)^4 - \frac{3}{2}v^2(3k - N)^2 + \frac{27}{4}v^4}{\{(3k - N)^2 + 3v^2\}^2} = \frac{-\frac{1}{4}\{(3k - N)^2 + 9v^2\}^2 + \frac{27}{2}v^4}{\{(3k - N)^2 + 3v^2\}^2} \\ & \leq \frac{-\frac{81}{4}v^4 + \frac{27}{2}v^4}{\{(3k - N)^2 + 3v^2\}^2} \leq 0. \end{aligned}$$

Hence  $3N/2\lambda$  for every  $v$ , is decreasing with increasing  $k$ , so that we have to take the greatest possible value of  $k$ . In that case  $3k - N$  is certainly positive and we have to take the smallest possible value of  $|v|$  to minimize  $3N/2\lambda$ . In virtue of the conditions of  $k$  and  $v$  we have the "extreme" combinations  $k = N - 1, |v| = 1$  and  $k = N - 2, v = 0$ . Substituting in (4) gives for  $\lambda$ , respectively,  $2(N^2 - 3N + 3)/(N - 1)$  and  $N$ . Since those values are equal only for  $N = 2$  or  $3$ , and

$$2 \frac{N^2 - 3N + 3}{N - 1} > N$$

if  $N \geq 4$ , we find for our upper bound

$$\max \beta_2 = \frac{1}{2}\lambda = \frac{N^2 - 3N + 3}{N - 1}.$$

And indeed

$$\max \beta_2 = \max \beta_1^2 + 1.$$

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<sup>2</sup> After writing this paper, the author derived the more general formulae:

$$\max \beta_{2n} = \frac{(N - 1)^{2n+1} + 1}{N(N - 1)^n}$$

and

$$\max \beta_{2n+1} = \frac{(N - 1) \{ (N - 1)^{2n+2} - 1 \}}{N(N - 1)^{n+1}}.$$

Their proof will be published shortly in the Dutch mathematics and physics periodical *Simon Stevin*.