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# ASYMPTOTIC INDEPENDENCE BETWEEN LARGEST AND SMALLEST OF A SET OF INDEPENDENT OBSERVATIONS

by

John E. Walsh

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DEPARTMENT OF STATISTICS
Southern Methodist University

# ASYMPTOTIC INDEPENDENCE BETWEEN LARGEST AND SMALLEST OF A SET OF INDEPENDENT OBSERVATIONS

John E. Walsh

Southern Methodist University\*

### ABSTRACT

Let  $X_n$  and  $X_1$  be the largest and smallest order statistics, respectively, of a set of n independent univariate observations. Under rather general conditions, with respect to the distributions of the individual observations,  $X_n$  and  $X_1$  are asymptotically independent. That is, the maximum difference between  $P(X_1 \le x_1, X_n \le x_n)$  and  $P(X_1 \le x_1)P(X_n \le x_n)$  tends to zero as  $n \to \infty$ . However, asymptotic independence does not occur for all cases.

### INTRODUCTION AND RESULTS

Asymptotic independence of the largest and smallest order statistics of a random univariate sample is well known. The question arises as to what extent this asymptotic independence remains when the observations are still required to be independent but can have arbitrarily different distributions. That is, let  $\mathbf{X}_n$  and  $\mathbf{X}_l$  be the largest and smallest, respectively of a set of n independent observations. When does the maximum of

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$$P(X_{1} \leq x_{1})P(X_{n} \leq x_{n}) - P(X_{1} \leq x_{1}, X_{n} \leq x_{n}), \tag{1}$$

over  $x_1$  and  $x_n$ , tend to zero as  $n \to \infty$ ?

The interest is in the range of  $x_1$  values that are meaningful for  $P(X_1 \le x_1)$  and range of  $x_n$  values that are meaningful for  $P(X_n \le x_n)$ . That is, the analysis is made in terms of (attainable) percentiles for  $P(X_1 \le x_1)$  and for  $P(X_n \le x_n)$ . Let

$$P(X_n \le x_n) = e^{-a}, P(X_1 \le x_1) = 1 - e^{-b},$$

where a and b are arbitrary but fixed. Then,

$$\prod_{i=1}^{n} F_{i}(x_{n}) = e^{-a}, \qquad \prod_{i=1}^{n} [1 - F_{i}(x_{1})] = e^{-b},$$

where  $F_i(x)$  is the cumulative distribution function (cdf) for the i-th observation (i = 1, . . . , n).

Now consider the  $\mathbf{F_i}(\mathbf{x_n})$  and  $\mathbf{F_i}(\mathbf{x_l}).$  Let these cdf's be expressed as

$$F_{i}(x_{n}) = e^{-a_{i}/n}$$
,  $F_{i}(x_{1}) = 1 - e^{-b_{i}/n}$ 

where  $a_i = a_i(n)$  and  $b_i = b_i(n)$ . Asymptotic independence between  $X_n$  and  $X_1$  always occurs if

$$a_i \le A(n), \qquad b_i \le B(n)$$

for all i. Here, A(n) and B(n) are O(n) and at least one of them is o(n). That is both A(n)/n and B(n)/n tend to constants as  $n \to \infty$ , and at least one of these constants is zero. For example, the forms  $C_1 n/\log n$  and  $C_2 n^{1-\epsilon}$  (with  $\epsilon > 0$  and fixed but as small as desired) yield zero constants.

Examination shows that asymptotic independence fails to occur only when, for one or more values of i, both  $a_i$  and  $b_i$  are O(n) and neither is o(n). For these observations, the values of  $1 - F_i(x_n)$  and  $F_i(x_1)$ , representing the "tail" probabilities, are relatively much larger than these values for the other observations (ratio becomes infinite as  $n \to \infty$ ). Thus, for large n, approximate independence of  $X_n$  and  $X_1$  should not be accepted when a few of the distributions seem to have a much wider spread (in both tails) than the others.

The next and final section contains derivations of the results that are stated in this section.

#### **DERIVATIONS**

The difference (1) can be written

$$\begin{split} \prod_{i=1}^{n} \ F_{i}(x_{n}) \big[ 1 - F_{i}(x_{1}) \big] - \prod_{i=1}^{n} \big[ F_{i}(x_{n}) - F_{i}(x_{1}) \big] \\ &= e^{-(a+b)} - e^{-a} \prod_{i=1}^{n} \bigg[ 1 - e^{a_{i}/n} + e^{-(b_{i} - a_{i})/n} \bigg] \\ &= e^{-(a+b)} - e^{-a} \prod_{i=1}^{n} \bigg[ 1 - \frac{b_{i}}{n} - \sum_{k=2}^{\infty} \frac{a_{i}^{k} + (-1)^{k+1}(b_{i} - a_{i})^{k}}{k! \ n^{k}} \bigg] \\ &= e^{-(a+b)} - e^{-a} \exp \left\{ \sum_{i=1}^{n} \log_{e} \left[ 1 - \frac{b_{i}}{n} - \sum_{k=2}^{\infty} \frac{a_{i}^{k} + (-1)^{k+1}(b_{i} - a_{i})^{k}}{k! \ n^{k}} \right] \right\} \\ &= e^{-(a+b)} \left( 1 - \exp \left\{ - \sum_{i=1}^{n} \frac{a_{i}b_{i}}{n^{2}} \left[ 1 + \frac{a_{i} + b_{i}}{2n} + \sum_{k=4}^{\infty} G_{k-2} \left( \frac{a_{i}}{n} + \frac{b_{i}}{n} \right) \right] \right\} \right), \end{split}$$

where  $G_{k-2}$  ( $a_i/n$ ,  $b_i/n$ ) is a mixed polynomial of degree k-2 that is symmetrical in  $a_i/n$  and  $b_i/n$ . The value of

$$\sum_{i=1}^{n} \frac{a_{i}b_{i}}{n^{2}} \left[ 1 + \frac{a_{i}+b_{i}}{2n} + \sum_{k=4}^{\infty} G_{k-2} \left( a_{i}/n, b_{i}/n \right) \right]$$
 (2)

is largest when some of the  $a_i$  have their maximum value, some of the  $b_i$  have their maximum value, and the others are zero. Also, the i values are such that the summation of  $a_i b_i / n^2$  is largest and, say, A(n) is O(n) with nonzero constant (equivalent results would be obtained if B(n) had the nonzero constant).

Let the values of i such that  $a_i = A(n)$  be  $i = 1, \ldots, r_n$ , while the values such that  $b_i = B(n)$  are  $i = 1, \ldots, r_1$ . Then, since,

$$\sum_{i=1}^{n} a_i = na, \qquad \sum_{i=1}^{n} b_i = nb,$$

 $r_n$  is O(1) and  $r_1$  is O[n/B(n)]. With these substitutions, (2) becomes

$$r_n[A(n)/n][B(n)/n][1 + O(1)]$$

and tends to zero as  $n \to \infty$ . Thus, the exponential of the negative of (2) tends to unity and the difference (1) tends to zero.

If both A(n) and B(n) had nonzero constants, the value of (2) would be

$$\min(r_1, r_n)[A(n)/n][B(n)/n][1 + O(1)]$$

and would not tend to zero as  $n \to \infty$ . Hence the difference (1) would not tend to zero and asymptotic independence does not occur.