

# THE JOINT DISTRIBUTION OF THE STANDARDIZED MAXIMUM AND STANDARDIZED MINIMUM FOR A NORMAL SAMPLE

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## Abstract

We find the joint distribution of Grubbs test statistics for a normal sample. Those statistics are standardized maximum and standardized minimum. We note some properties of the joint distribution function which can be applied for possible application of this function.

## 1 Introduction

Let  $X_1, X_2, \dots, X_{n-1}, X_n$  be a random sample from a normal  $N(a, \sigma^2)$  distribution with mean  $a$  and variance  $\sigma^2$ . Grubbs proposed the standardized maximum and standardized minimum [1]:

$$T_n^{(1)} = \frac{\max_{1 \leq i \leq n} \{X_i\} - \bar{X}}{S}; \quad T_n^{(1)} = \frac{\bar{X} - \min_{1 \leq i \leq n} \{X_i\}}{S},$$

where  $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$  is the sample mean and  $S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$  is the sample variance.

It is known that distributions of statistics  $T_n^{(1)}$  and  $T_{n,(1)}$  coincide, i.e.

$$P(T_n^{(1)} < t) = P(T_{n,(1)} < t). \quad (1)$$

Let  $F_n^{(1)}(t)$  be the distribution function of  $T_n^{(1)}$ , then [2]:

$$F_n^{(1)}(t) = P(T_n^{(1)} < t) = \begin{cases} 0, & t \leq \frac{1}{\sqrt{n}}, n \geq 2; \\ n \int_{\frac{1}{\sqrt{n}}}^t F_{n-1}^{(1)}(g_n(x)) f_{T_n}(x) dx, & \frac{1}{\sqrt{n}} < t \leq \frac{n-1}{\sqrt{n}}, n \geq 3; \\ 1, & t > \frac{n-1}{\sqrt{n}}, n \geq 2; \end{cases} \quad (2)$$

where

$$f_{T_n}(x) = \frac{1}{n-1} \sqrt{\frac{n}{\pi}} \Gamma\left(\frac{n-1}{2}\right) / \Gamma\left(\frac{n-2}{2}\right) \left(1 - \frac{n}{(n-1)^2} x^2\right)^{\frac{n-4}{2}}, \quad |x| < \frac{n-1}{\sqrt{n}}; \quad (3)$$

$$g_n(x) = \frac{n}{n-1} x / \sqrt{\frac{n-1}{n-2} \left(1 - \frac{n}{(n-1)^2} x^2\right)}, \quad |x| < \frac{n-1}{\sqrt{n}}, \quad n \geq 3. \quad (4)$$

We find the joint distribution of Grubbs test statistics  $T_{n,(1)}$  and  $T_n^{(1)}$  for a normal sample.

## 2 The joint distribution of Grubbs test statistics for a normal sample

Define

$$T_i = \frac{X_i - \bar{X}}{S}, \quad (i = 1, 2, \dots, n).$$

Random variables  $T_1, T_2, \dots, T_n$  are identically distributed. Let  $f_{i;n}(x)$  be density function of studentized deviation  $T_i$ ,  $i = \overline{1, n}$ , then [3]:

$$f_{i;n}(x) = f_{T_n}(x), \quad (i = 1, 2, \dots, n),$$

where  $f_{T_n}(x)$  can be calculated with using (3).

We have the following equalities for statistics  $T_n^{(1)}$  and  $T_{n,(1)}$  :

$$T_{n,(1)} = - \min_{1 \leq i \leq n} \{T_i\}, \quad (5)$$

$$T_n^{(1)} = \max_{1 \leq i \leq n} \{T_i\}. \quad (6)$$

It follows from relationships (1) and (2) that

$$\frac{1}{\sqrt{n}} \leq T_{n,(1)} \leq \frac{n-1}{\sqrt{n}}, \quad \frac{1}{\sqrt{n}} \leq T_n^{(1)} \leq \frac{n-1}{\sqrt{n}}. \quad (7)$$

Let  $\Lambda_n(t_1, t_2) = P(T_{n,(1)} < t_1, T_n^{(1)} < t_2)$  be the joint distribution function of Grubbs test statistics  $T_n^{(1)}$  and  $T_{n,(1)}$ . The following theorem describes our main result.

**Theorem 1.** *If  $X_1, X_2, \dots, X_n$  is a random sample from a normal  $N(a, \sigma^2)$  distribution, then the joint distribution function of Grubbs test statistics  $T_n^{(1)}$  and  $T_{n,(1)}$  for the case  $n = 2$  is given by*

$$\Lambda_2(t_1, t_2) = \begin{cases} 1, & (t_1, t_2) \in \Delta_2, \quad \Delta_2 = [\frac{\sqrt{2}}{2} < t_1 < \infty; \frac{\sqrt{2}}{2} < t_2 < \infty]; \\ 0, & (t_1, t_2) \notin \Delta_2, \end{cases} \quad (8)$$

and for the case  $n > 2$

$$\Lambda_n(t_1, t_2) = \begin{cases} F_n^{(1)}(t_2), & t_1 \geq \frac{n-1}{\sqrt{n}}; \\ F_n^{(1)}(t_1), & t_2 \geq \frac{n-1}{\sqrt{n}}; \\ n \int_{\frac{1}{\sqrt{n}}}^{t_2} \Lambda_{n-1}(\rho_n(t_1, -x), g_n(x)) f_{T_n}(x) dx, & (t_1, t_2) \in \Delta_n; \\ 0, & (t_1, t_2) \notin \Delta_n, \quad t_1 < \frac{n-1}{\sqrt{n}}, \quad t_2 < \frac{n-1}{\sqrt{n}}, \end{cases} \quad (9)$$

where distribution function  $F_n^{(1)}(t)$  can be calculated with using (2);

$$\rho_n(u, v) = \left(u + \frac{v}{n-1}\right) / \sqrt{\frac{n-1}{n-2} \left(1 - \frac{n}{(n-1)^2} v^2\right)}, \quad |v| < \frac{n-1}{\sqrt{n}}; \quad (10)$$

functions  $g_n(x)$  and  $f_{T_n}(x)$  can be calculated with using (4) and (3) correspondingly;  $\Delta_n = [1/\sqrt{n} < t_1 < (n-1)/\sqrt{n}; 1/\sqrt{n} < t_2 < (n-1)/\sqrt{n}]$ , if  $n > 2$ .

*Proof.* Suppose that  $n = 2$ . In that case  $T_{2,(1)} = T_2^{(1)} = \frac{1}{\sqrt{2}}$ . We have  $\Lambda_2(t_1, t_2) = P\{(1/\sqrt{2} < t_1) \cap (1/\sqrt{2} < t_2)\}$ , that proves the formula (8).

Suppose that  $n > 2$ . Then for  $t_1 \geq \frac{n-1}{\sqrt{n}}$  by (7) we have

$$\Lambda_n(t_1, t_2) = P\{(T_{n,(1)} < (n-1)/\sqrt{n}) \cap (T_n^{(1)} < t_2)\} = P(T_n^{(1)} < t_2).$$

Therefore,

$$\Lambda_n(t_1, t_2) = F_n^{(1)}(t_2), \quad t_1 \geq \frac{n-1}{\sqrt{n}}, \quad (11)$$

where  $F_n^{(1)}(t_2)$  is defined by (2).

Similarly, in the case  $t_2 \geq (n-1)/\sqrt{n}$  we have  $\Lambda_n(t_1, t_2) = P(T_{n,(1)} < t_1)$ . It follows from (1) that

$$\Lambda_n(t_1, t_2) = F_n^{(1)}(t_1), \quad t_2 \geq (n-1)/\sqrt{n}. \quad (12)$$

Next, it follows from conditions (7), that

$$\Lambda_n(t_1, t_2) = 0, \quad \text{if } (t_1 \leq 1/\sqrt{n}) \text{ or } (t_2 \leq 1/\sqrt{n}). \quad (13)$$

Let be  $\Delta_n = [1/\sqrt{n} < t_1 < (n-1)/\sqrt{n}; 1/\sqrt{n} < t_2 < (n-1)/\sqrt{n}]$ . We have with use (5) and (6) that  $\Lambda_n(t_1, t_2) = P\{(\min_{1 \leq j \leq n} \{T_j\} > -t_1) \cap (\max_{1 \leq j \leq n} \{T_j\} < t_2)\}$ . Hence, the joint distribution function of statistics  $T_{n,(1)}$  and  $T_n^{(1)}$  is defined by

$$\Lambda_n(t_1, t_2) = P\{\cap_{i=1}^n (-t_1 < T_i < t_2)\}. \quad (14)$$

Next, using formulas (14), (6) and (7), we have for  $(t_1, t_2) \in \Delta_n$

$$\Lambda_n(t_1, t_2) = P\left\{\cap_{i=1}^n \left(-t_1 < T_i \leq \max_{1 \leq j \leq n} \{T_j\}\right) \cap \left(\frac{1}{\sqrt{n}} < \max_{1 \leq j \leq n} \{T_j\} < t_2\right)\right\}.$$

Then

$$\Lambda_n(t_1, t_2) = nP\left\{\cap_{i=1}^{n-1} (-t_1 < T_i < T_n) \cap \left(\frac{1}{\sqrt{n}} < T_n < t_2\right)\right\}. \quad (15)$$

Define

$$T_i^* = \frac{X_i - \bar{X}^*}{S^*}, \quad (i = 1, 2, \dots, n-1),$$

where  $\bar{X}^* = \frac{1}{n-1} \sum_{k \neq n} X_k$ ,  $S^{*2} = \frac{1}{n-2} \sum_{k \neq n} (X_k - \bar{X}^*)^2$ .

It is valid that [3]

$$T_i = T_i^* \sqrt{\frac{n-1}{n-2} \left(1 - \frac{n}{(n-1)^2} T_n^2\right)} - \frac{1}{n-1} T_n, \quad i = \overline{1, n-1}, \quad |T_n| < \frac{n-1}{\sqrt{n}}.$$

Therefore the relationship (15) can be written as:

$$\Lambda_n(t_1, t_2) = nP\left\{\cap_{i=1}^{n-1} (\rho_n(-t_1, T_n) < T_i^* < g_n(T_n)) \cap (1/\sqrt{n} < T_n < t_2)\right\}, \quad (16)$$

where the function  $g_n(x)$  is defined by (4) and  $\rho_n(u, v)$  — by (10).

It is well known that random variable  $T_n$  is the function of  $\bar{X}^*$ ,  $S^*$  and  $X_n$  [3]. Besides, random variables  $\{T_1^*, T_2^*, \dots, T_{n-1}^*\}$  is independent of  $\{\bar{X}^*, S^*, X_n\}$ . Hence, the random variable  $T_n$  is independent of  $\{T_1^*, T_2^*, \dots, T_{n-1}^*\}$  [3]. Therefore the relationship (16) can be written as

$$\Lambda_n(t_1, t_2) = n \int_{1/\sqrt{n}}^{t_2} P\{\cap_{i=1}^{n-1} \{\rho_n(-t_1, x)\} < T_i^* < g_n(x)\} f_{T_n}(x) dx.$$

It follows from (14) that

$$\Lambda_n(t_1, t_2) = n \int_{1/\sqrt{n}}^{t_2} \Lambda_{n-1}\{\rho_n(t_1, -x), g_n(x)\} f_{T_n}(x) dx, \quad (t_1, t_2) \in \Delta_n. \quad (17)$$

If we combine relationships (11), (12), (13) and (17), then we receive (9), that proves the theorem.

Note some properties of the joint distribution function  $\Lambda_n(t_1, t_2)$  which can be derived from (9).

1. Function  $\Lambda_n(t_1, t_2)$  is symmetrical, i.e.  $\Lambda_n(t_1, t_2) = \Lambda_n(t_2, t_1)$ .
2. It is valid for  $(t_1, t_2) \in D_n$  and  $n > 2$  :

$$\Lambda_n(t_1, t_2) = F_n^{(1)}(t_1) - n \int_{t_2}^{\frac{n-1}{\sqrt{n}}} F_{n-1}^{(1)}(\rho_n(t_1, -x)) f_{T_n}(x) dx,$$

where  $D_n = [\frac{1}{\sqrt{n}} < t_1 < \frac{n-1}{\sqrt{n}}; \tau_n^* \leq t_2 < \frac{n-1}{\sqrt{n}}]; \tau_n^* = \sqrt{\frac{(n-1)(n-2)}{2n}}$ .

3. It is valid for  $(t_1, t_2) \in \Xi_n$  and  $n > 2$ :

$$\Lambda_n(t_1, t_2) = F_n^{(1)}(t_1) + F_n^{(1)}(t_2) - 1, \quad (18)$$

where  $\Xi_n = [t_n^* \leq t_1 < \frac{n-1}{\sqrt{n}}; t_n^* \leq t_2 < \frac{n-1}{\sqrt{n}}]; t_n^* = \sqrt{\frac{n-1}{2}}$ ;

$$F_n^{(1)}(t) = 1 - n \int_t^{\frac{n-1}{\sqrt{n}}} f_{T_n}(x) dx, \quad t \geq t_n^*.$$

## References

- [1] Grubbs F. (1950). Sample Criteria for Testing Outlying observations *Ann. Math. Statist.* Vol. **21**, pp. 27–58.
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