

A REMARK ON A PAPER OF TRAWINSKI AND DAVID ENTITLED:  
 "SELECTION OF THE BEST TREATMENT IN A PAIRED-  
 COMPARISON EXPERIMENT"<sup>1</sup>

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The present remark is concerned with the asymptotic behavior of the highest score in a paired-comparison experiment, if the number  $t$  of treatments is very large. It answers a question implicitly posed in Fig. 1 A of [1]. Besides, a useful inequality for the joint cumulative distribution function of the scores will be derived.

Assume that the treatments  $T_i (i = 1, 2, \dots, t)$  are all equal, with the exception of a single "outlier"  $T_t$ , which will be preferred with probability  $p > \frac{1}{2}$  when compared with any other treatment.<sup>2</sup> Assume that each pair of treatments is compared exactly once, and declare best the treatment with the highest score  $a_i$ , that is, with the highest number of preferences. Let  $P_{p,t}$  be the probability of selecting the actually best treatment  $T_t$  by this procedure.

What happens if  $t$  tends to infinity? Fig. 1 A of [1] seems to indicate that  $P_{p,t}$  tends to 1 if  $p$  is near 1, and to 0 if  $p$  is near  $\frac{1}{2}$ .

Rather surprisingly, this conjecture is false; actually we have  $\lim_{t \rightarrow \infty} P_{p,t} = 1$  for all fixed  $p > \frac{1}{2}$ .

Consider first the case where all treatments are equal (no outlier). Then each score  $a_i$  is binomially distributed, and the reduced score  $a_i^* = \{a_i - [(t-1)/2]\} / [(t-1)/4]^{\frac{1}{2}}$  is asymptotically normal with mean 0 and variance 1. In particular (see, e.g., W. Feller, *An Introduction to Probability Theory*, 2nd ed. p. 178).

$$(1) \quad P[a_i^* > x_{t-1}] \sim [(2\pi)^{\frac{1}{2}} x_{t-1}]^{-1} e^{-\frac{1}{2}x_{t-1}^2}$$

provided  $t \rightarrow \infty$ ,  $x_{t-1} \rightarrow \infty$ ,  $x_{t-1}^3 / [(t-1)/4]^{\frac{1}{2}} \rightarrow 0$ . The sign  $\sim$  denotes that the ratio of the two sides tends to 1.

In particular, let  $\epsilon > 0$  and put

$$(2) \quad x_{t-1}^{\pm} = [2 \log(t-1) - (1 \pm \epsilon) \log \log(t-1)]^{\frac{1}{2}};$$

one obtains

$$(3) \quad P[a_i^* > x_{t-1}^{\pm}] \sim \log(t-1)^{\pm \epsilon/2} / (4\pi)^{\frac{1}{2}} (t-1).$$

Now replace treatment  $T_t$  by an outlier,  $p > \frac{1}{2}$ . This stochastically decreases the reduced scores  $a_i^*$ ,  $1 \leq i < t$ , by an amount less than  $1/[(t-1)/4]^{\frac{1}{2}}$ . But a straightforward calculation shows that we may add a term of the order

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<sup>2</sup> To avoid confusion with the number  $\pi = 3.14 \dots$ , I changed the  $\pi$  of  $T$ . and  $D$ . into  $p$ .



$o[\log(t - 1)]^{-\frac{1}{2}}$  to  $x_{t-1}^{\pm}$  without destroying (3). Hence, (3) holds for  $1 \leq i < t$ , even if  $T_t$  is an outlier.

Let  $c'' < 1/(4\pi)^{\frac{1}{2}} < c'$ ; if the factor  $1/(4\pi)^{\frac{1}{2}}$  in (3) is replaced by  $c'$  or  $c''$ , one obtains strict inequalities for sufficiently large  $t$ , and the general inequality  $P[\cup A_i] \leq \Sigma P[A_i]$  then yields

$$(4) \quad P[\max_{1 \leq i < t} a_i^* > x_{t-1}^-] < c' \cdot (\log(t - 1))^{-\epsilon'/2}$$

for sufficiently large  $t$ .

With the aid of the Lemma below, one gets

$$(5) \quad P[\max_{1 \leq i < t} a_i^* < x_{t-1}^+] \leq \prod_{i=1}^{t-1} P[a_i^* < x_{t-1}^+] < \{1 - c''[(\log(t - 1))^{\epsilon/2}/(t - 1)]\}^{t-1} \leq \exp \{-c'' \cdot (\log(t - 1))^{\epsilon/2}\}$$

for sufficiently large  $t$ .

The inequalities (4) and (5) constitute the main result of this note; an immediate consequence of them is, for instance, the

COROLLARY. *If  $t \rightarrow \infty$ , then  $\max_{1 \leq i < t} a_i^* - [2 \log(t - 1)]^{\frac{1}{2}} \rightarrow 0$  in probability.*

Inequality (4) or the corollary imply that  $\max_{1 \leq i < t} a_i/(t - 1)$  tends in probability to  $\frac{1}{2}$ . If  $T_t$  is an outlier, then  $a_t/(t - 1)$  tends to  $p > \frac{1}{2}$ , and it follows immediately that  $\lim_{t \rightarrow \infty} P_{p,t} = 1$ .

Furthermore, one may conclude from the corollary that  $a_{\max}/(t - 1) = \max_{1 \leq i < t} a_i/(t - 1)$  clusters around  $m_{t-1} = \frac{1}{2} + [\log(t - 1)/2(t - 1)]^{\frac{1}{2}}$ , with a dispersion of the order  $o[1/(t - 1)^{\frac{1}{2}}]$  (more precisely, there exist intervals centered at  $m_{t-1}$ , of length  $o[1/(t - 1)^{\frac{1}{2}}]$ , which contain  $a_{\max}/(t - 1)$  with arbitrarily high preassigned probability). Similarly,  $a_i/(t - 1)$  clusters around  $p$  and has a dispersion of the order  $O[1/(t - 1)^{\frac{1}{2}}]$ . Both dispersions are of smaller order than  $m_{t-1} - \frac{1}{2}$ . Now assume that  $p$  is very near to  $\frac{1}{2}$ ; then this implies that  $P_{p,t}$  will first decrease, until the value of  $m_{t-1} - \frac{1}{2}$  becomes comparable with  $p - \frac{1}{2}$ . Already for moderate  $p$  this leads to large values of  $t$ ; since, for example,  $m_{1000} \approx 0.56$ ,  $m_{50,000} \approx 0.51$ ,  $m_{10,000,000} \approx 0.501$ .

We shall now prove the inequality used in establishing (5). Let  $p_{ij}$  be the probability that  $T_i$  is preferred to  $T_j$ .

LEMMA. *For any probability matrix  $(p_{ij})$  and any numbers  $(k_1, \dots, k_m)$ ,  $m \leq t$ , the joint cumulative distribution function of the scores  $a_1, \dots, a_m$  satisfies*

$$F(k_1, \dots, k_m) = P[a_1 < k_1, \dots, a_m < k_m] \leq P[a_1 < k_1] \cdot \dots \cdot P[a_m < k_m].$$

PROOF. Any two scores  $a_i, a_j$  are dependent only through the result of the comparison between the respective treatments  $T_i, T_j$ . Put  $a_i = a'_i + w_i$ ,  $a_j = a'_j + w_j$ ;  $w_i = 1 - w_j$  being 1 or 0, according as  $T_i$  is preferred to  $T_j$  or not. Replace  $w_i$  and  $w_j$  by independent variables without changing the marginal distributions. This destroys the dependence between  $a_i$  and  $a_j$  and changes  $F$  into a new joint distribution function  $F'$ . An explicit computation yields that  $F'$

majorizes  $F$ :

$$\begin{aligned}
& F'(k_1, \dots, k_m) - F(k_1, \dots, k_m) \\
&= \sum_{c_i, c_j} P[a_1 < k_1, \dots, a'_i < k_i - c_i, \dots, a'_j < k_j - c_j, \dots] \\
&\quad \cdot \{P[w_i = c_i] \cdot P[w_j = c_j] - P[w_i = c_i, w_j = c_j]\} \\
&= p_{ij} \cdot p_{ji} \cdot \{P[\dots, a'_i < k_i, \dots, a'_j < k_j, \dots] \\
&\quad - P[\dots, a'_i < k_i, \dots, a'_j < k_j - 1, \dots] \\
&\quad - P[\dots, a'_i < k_i - 1, \dots, a'_j < k_j, \dots] \\
&\quad + P[\dots, a'_i < k_i - 1, \dots, a'_j < k_j - 1, \dots]\} \\
&= p_{ij} p_{ji} P[a_1 < k_1, \dots, k_i - 1 \leq a'_i < k_i, \dots, \\
&\quad k_j - 1 \leq a'_j < k_j, \dots, a_m < k_m] \geq 0.
\end{aligned}$$

One repeats now this procedure for other pairs of treatments; eventually one obtains the distribution function corresponding to independent scores, thus proving that it majorizes the original distribution function.

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#### REFERENCE

- [1] TRAWINSKI, B. J. and DAVID, H. A. (1963). Selection of the best treatment in a paired-comparison experiment. *Ann. Math. Statist.* **34** 75-91.