

**CONTROL OF DIRECTIONAL ERRORS IN FIXED  
SEQUENCE MULTIPLE TESTING**

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**Supplementary Material**

**S1 Proofs**

PROOF OF LEMMA 1. Let  $T$  and  $P$  denote the test statistic and the corresponding  $p$ -value for testing  $H$ , respectively. When testing  $H$ , a type 3 error occurs if  $H$  is rejected and  $\theta T < 0$ . Then, the type 3 error rate is given by  $Pr(P \leq \alpha, \theta T < 0)$ .

When  $\theta > 0$ , we have

$$\begin{aligned} Pr(P \leq \alpha, \theta T < 0) &= Pr(2F_0(T) \leq \alpha, T < 0) \\ &= Pr\left(T \leq F_0^{-1}\left(\frac{\alpha}{2}\right)\right) = F_\theta\left(F_0^{-1}\left(\frac{\alpha}{2}\right)\right) \\ &\leq F_0\left(F_0^{-1}\left(\frac{\alpha}{2}\right)\right) = \frac{\alpha}{2}. \end{aligned}$$

The inequality follows from the assumption that  $F_\theta$  is stochastically increasing in  $\theta$ . Similarly, when  $\theta < 0$ , we can also prove that  $Pr(P \leq \alpha, \theta T < 0) \leq \frac{\alpha}{2}$ .  $\square$

**PROOF OF THEOREM 1(i).** Induction will be used to show that Procedure 1 strongly controls the mdFWER at level  $\alpha$ . First consider the case of  $n = 2$ . We show control of the mdFWER of Procedure 1 in all possible combinations of true and false null hypotheses while testing two hypotheses  $H_1$  and  $H_2$ .

**Case I:  $H_1$  is true.** Type 1 or type 3 error occurs only when  $H_1$  is rejected.

$$\text{mdFWER} = Pr(P_1 \leq \alpha) \leq \alpha.$$

**Case II: Both  $H_1$  and  $H_2$  are false.** We have no type 1 errors but only type 3 errors.

$$\begin{aligned} & \text{mdFWER} \\ &= Pr(\{P_1 \leq \alpha, T_1\theta_1 < 0\} \cup \{P_1 \leq \alpha, T_1\theta_1 \geq 0, P_2 \leq \alpha/2, T_2\theta_2 < 0\}) \\ &\leq Pr(P_1 \leq \alpha, T_1\theta_1 < 0) + Pr(P_2 \leq \alpha/2, T_2\theta_2 < 0) \\ &\leq \frac{\alpha}{2} + \frac{\alpha}{4} = \frac{3\alpha}{4}. \end{aligned}$$

The first inequality follows from Bonferroni inequality and the second follows from Lemma 1.

**Case III:  $H_1$  is false and  $H_2$  is true.** The mdFWER is bounded above

by

$$\begin{aligned}
& Pr(\text{ make type 3 error when testing } H_1) \\
& \quad + Pr(\text{ make type 1 error when testing } H_2) \\
& \leq Pr(P_1 \leq \alpha, T_1\theta_1 < 0) + Pr(P_2 \leq \alpha/2) \\
& \leq \frac{\alpha}{2} + \frac{\alpha}{2} = \alpha.
\end{aligned}$$

The first inequality follows from Bonferroni inequality and the second follows from Lemma 1 and  $P_2 \sim U(0, 1)$  since  $H_2$  is true.

Now assume the inductive hypothesis that the mdFWER is bounded above by  $\alpha$  when testing at most  $n - 1$  hypotheses by using Procedure 1 at level  $\alpha$ . In the following, we prove the mdFWER is also bounded above by  $\alpha$  when testing  $n$  hypotheses  $H_1, \dots, H_n$ . Without loss of generality, assume  $H_1$  is a false null (if  $H_1$  is a true null, the desired result directly follows by using the same argument as in Case I of  $n = 2$ ). Then, the mdFWER is bounded above by

$$\begin{aligned}
& Pr(\text{ make type 3 error when testing } H_1) \\
& \quad + Pr(\text{ make at least one type 1 or type 3 errors when testing } H_2, \dots, H_n) \\
& \leq \frac{\alpha}{2} + \frac{\alpha}{2} = \alpha.
\end{aligned}$$

The inequality follows from the induction assumption, noticing that  $H_2, \dots, H_n$  are tested by using Procedure 1 at level  $\alpha/2$ . Thus, the desired result fol-

lows.

(ii). We now prove that the critical constants are unimprovable. For instance, when  $H_1$  is true, it is easy to see that the first critical constant,  $\alpha$ , is unimprovable. For each given  $k = 2, \dots, n$ , when  $\theta_i > 0, i = 1, \dots, k - 1$  and  $\theta_k = 0$ , that is,  $H_i, i = 1, \dots, k - 1$  are false and  $H_k$  is true, we present a simple joint distribution of the test statistics  $T_1, \dots, T_k$  to show that the  $k$ th critical constant of this procedure is also unimprovable.

Define  $Z_k \sim N(0, 1)$  and  $Z_i = \Phi^{-1}(|2\Phi(Z_{i+1}) - 1|), i = 1, \dots, k - 1$ , where  $\Phi(\cdot)$  is the cdf of  $N(0, 1)$ . Let  $q_i$  denote  $Z_i$ 's upper  $\alpha/2^i$  quantile. It is easy to check that for each  $i = 1, \dots, k$ ,  $Z_i \sim N(0, 1)$ . Thus,  $-q_i$  is  $Z_i$ 's lower  $\alpha/2^i$  quantile. In addition, by the construction of  $Z_i$ 's, it is easy to see that the event  $Z_i \geq q_i$  is equivalent to the event  $Z_{i+1} \notin (-q_{i+1}, q_{i+1})$ .

Let  $T_i = Z_i + \theta_i, i = 1, \dots, k$ , thus  $T_i \sim N(\theta_i, 1)$ . Then, as  $\theta_i \rightarrow 0+$  for  $i = 1, \dots, k - 1$ , we have

$$\begin{aligned}
\text{mdFWER} &= \sum_{j=1}^{k-1} Pr(T_1 \geq q_1, \dots, T_{j-1} \geq q_{j-1}, T_j \leq -q_j) \\
&\quad + Pr(T_1 \geq q_1, \dots, T_{k-1} \geq q_{k-1}, T_k \notin (-q_k, q_k)) \\
&= \sum_{j=1}^{k-1} Pr(Z_1 \geq q_1, \dots, Z_{j-1} \geq q_{j-1}, Z_j \leq -q_j) \\
&\quad + Pr(Z_1 \geq q_1, \dots, Z_{k-1} \geq q_{k-1}, Z_k \notin (-q_k, q_k)) \\
&= \sum_{j=1}^{k-1} Pr(Z_j \leq -q_j) + Pr(Z_k \notin (-q_k, q_k))
\end{aligned}$$

$$= \sum_{j=1}^{k-1} \frac{\alpha}{2^j} + \frac{\alpha}{2^{(k-1)}} = \alpha.$$

Thus, the  $k$ th critical constant of Procedure 1 is unimprovable and hence each critical constant of Procedure 1 is unimprovable under arbitrary dependence.  $\square$

PROOF OF LEMMA 2. Note that when  $\theta_1 > 0$  and  $\theta_2 = 0$ , we have

$$\begin{aligned} & \text{mdFWER} \\ &= Pr(P_1 \leq \alpha, \theta_1 T_1 < 0) + Pr(P_1 \leq \alpha, \theta_1 T_1 \geq 0, P_2 \leq \alpha) \\ &= Pr(P_1 \leq \alpha, T_1 < 0) + Pr(P_1 \leq \alpha, T_1 \geq 0, P_2 \leq \alpha, T_2 > 0) \\ &\quad + Pr(P_1 \leq \alpha, T_1 \geq 0, P_2 \leq \alpha, T_2 \leq 0) \\ &= Pr(2F_0(T_1) \leq \alpha) + Pr(2(1 - F_0(T_1)) \leq \alpha, 2(1 - F_0(T_2)) \leq \alpha) \\ &\quad + Pr(2(1 - F_0(T_1)) \leq \alpha, 2F_0(T_2) \leq \alpha) \\ &= Pr(T_1 \leq c_1) + Pr(T_1 \geq c_2, T_2 \geq c_2) + Pr(T_1 \geq c_2, T_2 \leq c_1) \\ &= F_{\theta_1}(c_1) + 1 - F_{\theta_1}(c_2) - F_0(c_2) + F_{(\theta_1, 0)}(c_2, c_2) + F_0(c_1) - F_{(\theta_1, 0)}(c_2, c_1) \\ &= \alpha + F_{\theta_1}(c_1) - F_{\theta_1}(c_2) + F_{(\theta_1, 0)}(c_2, c_2) - F_{(\theta_1, 0)}(c_2, c_1). \end{aligned} \tag{S1.1}$$

Specifically, under Assumption 1 (independence), (S1.1) can be simplified as,

$$\begin{aligned} & \alpha + F_{\theta_1}(c_1) - F_{\theta_1}(c_2) + F_{\theta_1}(c_2)F_0(c_2) - F_{\theta_1}(c_2)F_0(c_1) \\ &= \alpha + F_{\theta_1}(c_1) - \alpha F_{\theta_1}(c_2). \end{aligned}$$

Similarly, when  $\theta_1 < 0$  and  $\theta_2 = 0$ , we can prove that

$$\text{mdFWER} = 1 + F_{\theta_1}(c_1) - F_{\theta_1}(c_2) + F_{(\theta_1,0)}(c_1, c_1) - F_{(\theta_1,0)}(c_1, c_2). \quad \square$$

**PROOF OF LEMMA 3.** By using the same arguments as in Theorem 1, we can easily prove control of the mdFWER of Procedure 2 in the case of  $n = 2$  when  $H_1$  is true or both  $H_1$  and  $H_2$  are false. In the following, we prove the desired result also holds when  $H_1$  is false and  $H_2$  is true.

Note that  $H_1$  is false and  $H_2$  is true imply  $\theta_1 \neq 0$  and  $\theta_2 = 0$ . To show that the mdFWER is controlled for  $\theta_1 > 0$  and  $\theta_2 = 0$ , we only need to show by Lemma 2 that  $\alpha + F_{\theta_1}(c_1) - \alpha F_{\theta_1}(c_2) \leq \alpha$ . This is equivalent to show

$$F_{\theta_1}(c_2) (F_0(c_2) - F_0(c_1)) \leq F_{\theta_1}(c_2) - F_{\theta_1}(c_1). \quad (\text{S1.2})$$

For proving (S1.2), it is enough to prove the following, as  $0 \leq F_0(c_2) \leq 1$ ,

$$F_{\theta_1}(c_2) (F_0(c_2) - F_0(c_1)) \leq F_0(c_2) (F_{\theta_1}(c_2) - F_{\theta_1}(c_1)). \quad (\text{S1.3})$$

Dividing both sides of (S1.3) by  $F_{\theta_1}(c_2)F_0(c_2)$ , we see that we only need to prove,

$$1 - \frac{F_0(c_1)}{F_0(c_2)} \leq 1 - \frac{F_{\theta_1}(c_1)}{F_{\theta_1}(c_2)},$$

which follows directly from (3.5) and Assumption 2 (MLR).

Similarly, to show that the mdFWER is controlled for  $\theta_1 < 0$  and  $\theta_2 = 0$ , we only need to show by Lemma 2 that  $1 + \alpha F_{\theta_1}(c_1) - F_{\theta_1}(c_2) \leq \alpha$ . This is equivalent to showing

$$(1 - \alpha)(1 - F_{\theta_1}(c_1)) \leq F_{\theta_1}(c_2) - F_{\theta_1}(c_1).$$

Writing  $1 - \alpha$  as  $(1 - F_0(c_1)) - (1 - F_0(c_2))$  and writing  $F_{\theta_1}(c_2) - F_{\theta_1}(c_1)$  as

$(1 - F_{\theta_1}(c_1)) - (1 - F_{\theta_1}(c_2))$ , we get that it is equivalent to prove

$$[(1 - F_0(c_1)) - (1 - F_0(c_2))](1 - F_{\theta_1}(c_1)) \leq (1 - F_{\theta_1}(c_1)) - (1 - F_{\theta_1}(c_2)) \quad (\text{S1.4})$$

Since  $0 \leq 1 - F_0(c_1) \leq 1$ , to prove inequality (S1.4), it is enough to prove the following,

$$\begin{aligned} & (1 - F_{\theta_1}(c_1)) [(1 - F_0(c_1)) - (1 - F_0(c_2))] \\ & \leq (1 - F_0(c_1)) [1 - F_{\theta_1}(c_1)] - [1 - F_{\theta_1}(c_2)]. \end{aligned} \quad (\text{S1.5})$$

Dividing both sides of (S1.5) by  $(1 - F_{\theta_1}(c_1))(1 - F_0(c_1))$ , we see that proving (S1.4) is equivalent to showing

$$\frac{1 - F_{\theta_1}(c_2)}{1 - F_{\theta_1}(c_1)} \leq \frac{1 - F_0(c_2)}{1 - F_0(c_1)}, \quad (\text{S1.6})$$

which follows directly from (3.6) and Assumption 2 (MLR). By combining the arguments of the above two cases, the desired result follows.  $\square$

PROOF OF THEOREM 2. The proof is by induction on number of hypotheses  $n$ . We already proved strong control of the mdFWER for  $n = 2$  in Lemma 3. Let us assume the result holds for testing any  $n = k$  hypotheses, that is,  $\text{mdFWER} \leq \alpha$  while testing any  $k$  pre-ordered hypotheses. We now argue that it will hold for  $n = k + 1$  hypotheses. Without loss of generality, assume  $H_1$  is a false null, as in the proof of Theorem 1.

Let  $V_{k+1}^{(-1)}$  denote the total number of type 1 or type 3 errors committed while testing  $H_2, \dots, H_{k+1}$  and excluding  $H_1$ . Then, by the inductive hypothesis, the mdFWER while testing the  $k$  hypotheses  $H_2, \dots, H_{k+1}$  is  $Pr(V_{k+1}^{(-1)} > 0) \leq \alpha$ . Then, the mdFWER of testing  $k + 1$  hypotheses  $H_1, \dots, H_{k+1}$  is defined by

$$\begin{aligned}
& Pr \left( \{P_1 \leq \alpha, T_1\theta_1 < 0\} \cup \{P_1 \leq \alpha, T_1\theta_1 \geq 0, V_{k+1}^{(-1)} > 0\} \right) \\
&= Pr(P_1 \leq \alpha, T_1\theta_1 < 0) + Pr(P_1 \leq \alpha, T_1\theta_1 \geq 0) \cdot Pr(V_{k+1}^{(-1)} > 0) \\
&\leq Pr(P_1 \leq \alpha, T_1\theta_1 < 0) + \alpha Pr(P_1 \leq \alpha, T_1\theta_1 \geq 0). \tag{S1.7}
\end{aligned}$$

The equality follows by Assumption 1 (independence) and the inequality follows by the inductive hypothesis. Note that (S1.7) is the same as (3.8) under independence, which is equal to the mdFWER of Procedure 2 in the case of two hypotheses. So again by applying Lemma 3, we get that  $\text{mdFWER} \leq \alpha$  for  $n = k + 1$ . Hence, the proof follows by induction.  $\square$



PROOF OF THEOREM 3 . Without loss of generality, we assume  $\theta_i > 0$  if  $\theta_i \neq 0$  for  $i = 1, \dots, n$ . Also, if there exists an  $i$  with  $\theta_i = 0$ , by induction, we can simply assume  $i_0 = n$ . Thus, to prove the mdFWER control of Procedure 2, we only need to consider two cases:

- (i)  $\theta_i > 0$  for  $i = 1, \dots, n$ ;
- (ii)  $\theta_i > 0$  for  $i = 1, \dots, n - 1$  and  $\theta_n = 0$ .

**Case (i).** Consider the general case of  $\theta_i > 0, i = 1, \dots, n$ . By Assumption 3, the test statistics  $T_1, \dots, T_n$  are positively regression dependent. For  $j = 1, \dots, n - 1$ , let  $E_{n-j}$  denote the event of making at least one type 3 error when testing  $H_{j+1}, \dots, H_n$  using Procedure 2 at level  $\alpha$ . By using induction, we prove the following two lemmas hold.

**Lemma 1.** *Assume the conditions of Theorem 3. For  $j = 1, \dots, n - 1$ , the following inequality holds.*

$$Pr(E_{n-j} | T_1 > c_2, \dots, T_j > c_2) \leq \alpha. \tag{S1.8}$$

PROOF OF LEMMA 1. We prove the result by using reverse induction.

When  $j = n - 1$ , we have

$$\begin{aligned}
& Pr(E_{n-j}|T_1 > c_2, \dots, T_j > c_2) \\
&= Pr(T_n < c_1|T_1 > c_2, \dots, T_{n-1} > c_2) \\
&= \frac{Pr(T_n < c_1)Pr(T_1 > c_2, \dots, T_{n-1} > c_2|T_n < c_1)}{Pr(T_1 > c_2, \dots, T_{n-1} > c_2)} \\
&\leq Pr(T_n < c_1) \leq \alpha.
\end{aligned}$$

The inequality follows from Assumption 3.

Assume the inequality (S1.8) holds for  $j = m$ . In the following, we prove that it also holds for  $j = m - 1$ . Note that

$$\begin{aligned}
& Pr(E_{n-m+1}|T_1 > c_2, \dots, T_{m-1} > c_2) \\
&= Pr\left(\{T_m < c_1\} \cup \left(\{T_m > c_2\} \cap E_{n-m}\right) | T_1 > c_2, \dots, T_{m-1} > c_2\right) \\
&= Pr(T_m < c_1|T_1 > c_2, \dots, T_{m-1} > c_2) \\
&\quad + Pr\left(\{T_m > c_2\} \cap E_{n-m} | T_1 > c_2, \dots, T_{m-1} > c_2\right) \\
&= Pr(T_m < c_1|T_1 > c_2, \dots, T_{m-1} > c_2) \\
&\quad + Pr(T_m > c_2|T_1 > c_2, \dots, T_{m-1} > c_2) Pr(E_{n-m}|T_1 > c_2, \dots, T_m > c_2) \\
&\leq Pr(T_m < c_1|T_1 > c_2, \dots, T_{m-1} > c_2) \\
&\quad + \alpha Pr(T_m > c_2|T_1 > c_2, \dots, T_{m-1} > c_2) \\
&\leq \alpha.
\end{aligned}$$

Therefore, the desired result follows. Here, the first inequality follows from

the assumption of induction and the second follows from Lemma 2 below.

□

**Lemma 2.** *Assume the conditions of Theorem 3. For  $j = 1, \dots, n-1$ , the following inequality holds:*

$$\begin{aligned} & Pr(T_j < c_1 | T_1 > c_2, \dots, T_{j-1} > c_2) \\ & + \alpha Pr(T_j > c_2 | T_1 > c_2, \dots, T_{j-1} > c_2) \leq \alpha. \end{aligned} \quad (\text{S1.9})$$

Specifically, for  $j = 1$ , we have

$$Pr(T_1 < c_1) + \alpha Pr(T_1 > c_2) \leq \alpha.$$

PROOF OF LEMMA 2. To prove the inequality (S1.9), it is enough to show that

$$Pr(T_j < c_1 | T_1 > c_2, \dots, T_{j-1} > c_2) \leq \alpha Pr(T_j < c_2 | T_1 > c_2, \dots, T_{j-1} > c_2),$$

which is equivalent to

$$\begin{aligned} & (1 - \alpha) Pr(T_j < c_2 | T_1 > c_2, \dots, T_{j-1} > c_2) \\ & \leq Pr(T_j < c_2 | T_1 > c_2, \dots, T_{j-1} > c_2) - Pr(T_j < c_1 | T_1 > c_2, \dots, T_{j-1} > c_2). \end{aligned}$$

Note that

$$1 - \alpha = Pr_{\theta_j=0}(T_j < c_2) - Pr_{\theta_j=0}(T_j < c_1).$$

Thus, the above inequality is equivalent to

$$Pr_{\theta_j=0}(T_j < c_2) - Pr_{\theta_j=0}(T_j < c_1) \leq 1 - \frac{Pr(T_j < c_1 | T_1 > c_2, \dots, T_{j-1} > c_2)}{Pr(T_j < c_2 | T_1 > c_2, \dots, T_{j-1} > c_2)},$$

which in turn is implied by

$$\begin{aligned}
& 1 - \frac{Pr_{\theta_j=0}(T_j < c_1)}{Pr_{\theta_j=0}(T_j < c_2)} \\
\leq & 1 - \frac{Pr(T_j < c_1 | T_1 > c_2, \dots, T_{j-1} > c_2)}{Pr(T_j < c_2 | T_1 > c_2, \dots, T_{j-1} > c_2)}. \tag{S1.10}
\end{aligned}$$

Note that by Assumption 2, we have

$$\frac{Pr(T_j < c_1)}{Pr(T_j < c_2)} \leq \frac{Pr_{\theta_j=0}(T_j < c_1)}{Pr_{\theta_j=0}(T_j < c_2)}.$$

Thus, to prove the inequality (S1.10), we only need to show that

$$\frac{Pr(T_j < c_1 | T_1 > c_2, \dots, T_{j-1} > c_2)}{Pr(T_j < c_2 | T_1 > c_2, \dots, T_{j-1} > c_2)} \leq \frac{Pr(T_j < c_1)}{Pr(T_j < c_2)},$$

which is equivalent to

$$Pr(T_1 > c_2, \dots, T_{j-1} > c_2 | T_j < c_1) \leq Pr(T_1 > c_2, \dots, T_{j-1} > c_2 | T_j < c_2),$$

which follows from Assumption 3. Therefore, the desired result follows.  $\square$

Based on Lemmas 1 and 2, we have

$$\begin{aligned}
\text{mdFWER} &= Pr(T_1 < c_1) + \sum_{j=2}^n Pr(T_1 > c_2, \dots, T_{j-1} > c_2, T_j < c_1) \\
&= Pr(T_1 < c_1) + Pr(T_1 > c_2) \sum_{j=2}^n Pr(T_2 > c_2, \dots, T_{j-1} > c_2, T_j < c_1 | T_1 > c_2) \\
&= Pr(T_1 < c_1) + Pr(T_1 > c_2) Pr(E_{n-1} | T_1 > c_2) \\
&\leq Pr(T_1 < c_1) + \alpha Pr(T_1 > c_2) \\
&\leq \alpha.
\end{aligned}$$

Therefore, the mdFWER is controlled at level  $\alpha$  for Case (i). Here, the first inequality follows from Lemma 1 and the second follows from Lemma 2.

**Case (ii).** Consider the general case of  $\theta_i > 0, i = 1, \dots, n - 1$  and  $\theta_n = 0$ . Under Assumption 3,  $T_i, i = 1, \dots, n - 1$  are positively regression dependent and under Assumption 4,  $T_n$  is independent of  $T_i$ 's . Note that

$$\begin{aligned}
& \text{mdFWER} \\
&= \sum_{j=1}^{n-1} Pr(T_1 > c_2, \dots, T_{j-1} > c_2, T_j < c_1) \\
&\quad + Pr(T_1 > c_2, \dots, T_{n-1} > c_2, T_n < c_1) + Pr(T_1 > c_2, \dots, T_n > c_2) \\
&= \sum_{j=1}^{n-1} Pr(T_1 > c_2, \dots, T_{j-1} > c_2, T_j < c_1) + \alpha Pr(T_1 > c_2, \dots, T_{n-1} > c_2).
\end{aligned}$$

The second equality follows from Assumption 4.

For  $m = 1, \dots, n - 1$ , define

$$\Delta_m = \sum_{j=1}^m Pr(T_1 > c_2, \dots, T_{j-1} > c_2, T_j < c_1) + \alpha Pr(T_1 > c_2, \dots, T_m > c_2).$$

Thus, mdFWER =  $\Delta_{n-1}$ . By using induction, we prove below that  $\Delta_m \leq \alpha$  for  $m = 1, \dots, n - 1$ .

For  $m = 1$ , by using Lemma 2, we have

$$\Delta_1 = Pr(T_1 < c_1) + \alpha Pr(T_1 > c_2) \leq \alpha.$$

Assume  $\Delta_m \leq \alpha$ . In the following, we show  $\Delta_{m+1} \leq \alpha$ . Note that

$$\begin{aligned}
& \Delta_{m+1} \\
= & \sum_{j=1}^{m+1} Pr(T_1 > c_2, \dots, T_{j-1} > c_2, T_j < c_1) \\
& + \alpha Pr(T_1 > c_2, \dots, T_m > c_2, T_{m+1} > c_2) \\
= & \sum_{j=1}^m Pr(T_1 > c_2, \dots, T_{j-1} > c_2, T_j < c_1) \\
& + Pr(T_1 > c_2, \dots, T_m > c_2) [Pr(T_{m+1} < c_1 | T_1 > c_2, \dots, T_m > c_2) \\
& + \alpha Pr(T_{m+1} > c_2 | T_1 > c_2, \dots, T_m > c_2)] \\
\leq & \sum_{j=1}^m Pr(T_1 > c_2, \dots, T_{j-1} > c_2, T_j < c_1) + \alpha Pr(T_1 > c_2, \dots, T_m > c_2) \\
= & \Delta_m \leq \alpha. \tag{S1.11}
\end{aligned}$$

The first inequality follows from Lemma 2 and the second follows from the inductive hypothesis. Thus,  $\Delta_m \leq \alpha$  for  $m = 1, \dots, n - 1$ . Therefore,  $\text{mdFWER} = \Delta_{n-1} \leq \alpha$ , the desired result.

Combining the arguments of Cases (i) and (ii), the proof of Theorem 3 is complete.  $\square$

**PROOF OF PROPOSITION 2.** From the proof of Theorem 1 and by Lemma 1, it is easy to see that we only need to prove the mdFWER control of Procedure 2 when  $H_1$  is false and  $H_2$  is true, i.e.,  $\theta_1 \neq 0$  and  $\theta_2 = 0$ .

**Case I:  $\theta_1 > 0$  and  $\theta_2 = 0$ .** By Lemma 2, the mdFWER of Procedure 2 is

controlled at level  $\alpha$  if we have the following:

$$F_{\theta_1}(c_1) - F_{\theta_1}(c_2) + F_{(\theta_1,0)}(c_2, c_2) - F_{(\theta_1,0)}(c_2, c_1) \leq 0.$$

After rewriting  $F_{(\theta_1,0)}(x, y)$  as  $Pr(T_1 \leq x, T_2 \leq y)$  and then dividing through by  $Pr(T_1 \leq c_2)$ , we get,

$$Pr(T_2 \leq c_2 | T_1 \leq c_2) - Pr(T_2 \leq c_1 | T_1 \leq c_2) \leq 1 - \frac{Pr(T_1 \leq c_1)}{Pr(T_1 \leq c_2)}.$$

Dividing by  $Pr(T_2 \leq c_2 | T_1 \leq c_2)$ , we get,

$$\begin{aligned} & 1 - \frac{Pr(T_2 \leq c_1 | T_1 \leq c_2)}{Pr(T_2 \leq c_2 | T_1 \leq c_2)} \\ & \leq \frac{1}{Pr(T_2 \leq c_2 | T_1 \leq c_2)} \left( 1 - \frac{Pr(T_1 \leq c_1)}{Pr(T_1 \leq c_2)} \right). \end{aligned} \quad (\text{S1.12})$$

For proving (S1.12), it is enough to prove the following inequality, as

$$\frac{1}{Pr(T_2 \leq c_2 | T_1 \leq c_2)} \geq 1.$$

$$1 - \frac{Pr(T_2 \leq c_1 | T_1 \leq c_2)}{Pr(T_2 \leq c_2 | T_1 \leq c_2)} \leq 1 - \frac{Pr(T_1 \leq c_1)}{Pr(T_1 \leq c_2)}. \quad (\text{S1.13})$$

By Assumption 2 and (3.5), it follows that  $\frac{F_0(c_2)}{F_0(c_1)} \leq \frac{F_{\theta_1}(c_2)}{F_{\theta_1}(c_1)}$ , which is equivalent to,

$$1 - \frac{Pr(T_2 \leq c_1)}{Pr(T_2 \leq c_2)} \leq 1 - \frac{Pr(T_1 \leq c_1)}{Pr(T_1 \leq c_2)}.$$

Thus for proving (S1.12), it is enough to prove the following:

$$1 - \frac{Pr(T_2 \leq c_1 | T_1 \leq c_2)}{Pr(T_2 \leq c_2 | T_1 \leq c_2)} \leq 1 - \frac{Pr(T_2 \leq c_1)}{Pr(T_2 \leq c_2)}. \quad (\text{S1.14})$$

But, (S1.14) is equivalent to showing

$$Pr(T_1 \leq c_2 | T_2 \leq c_1) \geq Pr(T_1 \leq c_2 | T_2 \leq c_2),$$

which follows directly from Assumption 5.

**Case II:**  $\theta_1 < 0$  and  $\theta_2 = 0$ . Similarly, by Lemma 2, the mdFWER of Procedure 2 is controlled at level  $\alpha$  if we have the following:

$$1 + F_{\theta_1}(c_1) - F_{\theta_1}(c_2) + F_{(\theta_1,0)}(c_1, c_1) - F_{(\theta_1,0)}(c_1, c_2) \leq \alpha, \quad (\text{S1.15})$$

which after some rearrangement and rewriting  $1 - \alpha$  as  $F_0(c_2) - F_0(c_1)$  gives,

$$\begin{aligned} & (F_0(c_2) - F_{(\theta_1,0)}(c_1, c_2)) - (F_0(c_1) - F_{(\theta_1,0)}(c_1, c_1)) \\ & \leq (1 - F_{\theta_1}(c_1)) - (1 - F_{\theta_1}(c_2)). \end{aligned} \quad (\text{S1.16})$$

Thus, proving (S1.15) is equivalent to proving that

$$Pr(T_1 \geq c_1, T_2 \leq c_2) - Pr(T_1 \geq c_1, T_2 \leq c_1) \leq Pr(T_1 \geq c_1) - Pr(T_1 \geq c_2).$$

Dividing through by  $Pr(T_1 \geq c_1)$ , we get

$$\begin{aligned} & Pr(T_2 \geq c_1 | T_1 \geq c_1) - Pr(T_2 \geq c_2 | T_1 \geq c_1) \\ & \leq 1 - \frac{Pr(T_1 \geq c_2)}{Pr(T_1 \geq c_1)}. \end{aligned} \quad (\text{S1.17})$$

Thus to prove (S1.15), it is enough to prove the following,

$$1 - \frac{Pr(T_2 \geq c_2 | T_1 \geq c_1)}{Pr(T_2 \geq c_1 | T_1 \geq c_1)} \leq 1 - \frac{Pr(T_1 \geq c_2)}{Pr(T_1 \geq c_1)},$$

which is equivalent to proving,

$$\frac{Pr(T_2 \geq c_2 | T_1 \geq c_1)}{Pr(T_2 \geq c_1 | T_1 \geq c_1)} \geq \frac{Pr(T_1 \geq c_2)}{Pr(T_1 \geq c_1)}. \quad (\text{S1.18})$$



By Assumption 2 and (3.6), it follows that for  $\theta_1 < 0$ ,  $\frac{Pr(T_1 \geq c_2)}{Pr(T_1 \geq c_1)} \leq \frac{Pr(T_2 \geq c_2)}{Pr(T_2 \geq c_1)}$ .

Thus to prove (S1.15), it is enough to prove the following,

$$\frac{Pr(T_2 \geq c_2 | T_1 \geq c_1)}{Pr(T_2 \geq c_1 | T_1 \geq c_1)} \geq \frac{Pr(T_2 \geq c_2)}{Pr(T_2 \geq c_1)}. \quad (\text{S1.19})$$

But (S1.19) is equivalent to showing

$$Pr(T_1 \geq c_1 | T_2 \geq c_2) \geq Pr(T_1 \geq c_1 | T_2 \geq c_1), \quad (\text{S1.20})$$

which follows directly from Assumption 5. By combining the arguments of the above two cases, the desired result follows.  $\square$

PROOF OF PROPOSITION 3. By Corollary 1, without loss of generality, assume that  $\theta_i > 0, i = 1, 2$  and  $\theta_3 = 0$ , that is,  $H_1$  and  $H_2$  are false and  $H_3$  is true. Note that

$$\begin{aligned} & \text{mdFWER} \\ &= Pr(T_1 \leq c_1) + Pr(T_1 \geq c_2, T_2 \leq c_1) \\ & \quad + Pr(T_1 \geq c_2, T_2 \geq c_2, T_3 \notin (c_1, c_2)). \end{aligned} \quad (\text{S1.21})$$

In the following, we prove that

$$\begin{aligned} & Pr(T_1 \geq c_2, T_2 \leq c_1) + Pr(T_1 \geq c_2, T_2 \geq c_2, T_3 \notin (c_1, c_2)) \\ & \leq Pr(T_1 \geq c_2, T_3 \notin (c_1, c_2)). \end{aligned} \quad (\text{S1.22})$$

To prove (S1.22), it is enough to show the following inequality:

$$\begin{aligned} & Pr(T_2 \leq c_1|T_1) + Pr(T_2 \geq c_2, T_3 \notin (c_1, c_2)|T_1) \\ \leq & Pr(T_3 \notin (c_1, c_2)|T_1). \end{aligned} \quad (\text{S1.23})$$

Note that

$$\begin{aligned} & Pr(T_2 \geq c_2, T_3 \leq c_1|T_1) \\ = & Pr(T_3 \leq c_1|T_1) - Pr(T_2 < c_2, T_3 \leq c_1|T_1) \end{aligned} \quad (\text{S1.24})$$

and

$$\begin{aligned} & Pr(T_2 \geq c_2, T_3 \geq c_2|T_1) \\ = & 1 - Pr(T_2 < c_2|T_1) - Pr(T_3 < c_2|T_1) + Pr(T_2 < c_2, T_3 < c_2|T_1). \end{aligned} \quad (\text{S1.25})$$

In addition, we have

$$Pr(T_3 \notin (c_1, c_2)|T_1) = 1 + Pr(T_3 \leq c_1|T_1) - Pr(T_3 < c_2|T_1). \quad (\text{S1.26})$$

Thus, in order to show (S1.23), by combining (S1.24)-(S1.26), we only need to prove the following inequality:

$$\begin{aligned} & Pr(T_2 < c_2, T_3 < c_2|T_1) - Pr(T_2 < c_2, T_3 \leq c_1|T_1) \\ \leq & Pr(T_2 < c_2|T_1) - Pr(T_2 \leq c_1|T_1). \end{aligned} \quad (\text{S1.27})$$

Note that (S1.27) can be rewritten as

$$\begin{aligned} & Pr(T_2 < c_2, T_3 < c_2 | T_1) \left[ 1 - \frac{Pr(T_2 < c_2, T_3 \leq c_1 | T_1)}{Pr(T_2 < c_2, T_3 < c_2 | T_1)} \right] \\ & \leq Pr(T_2 < c_2 | T_1) \left[ 1 - \frac{Pr(T_2 \leq c_1 | T_1)}{Pr(T_2 < c_2 | T_1)} \right]. \end{aligned} \quad (\text{S1.28})$$

Thus, to prove (S1.27), it is enough to show

$$1 - \frac{Pr(T_2 < c_2, T_3 \leq c_1 | T_1)}{Pr(T_2 < c_2, T_3 < c_2 | T_1)} \leq 1 - \frac{Pr(T_2 \leq c_1 | T_1)}{Pr(T_2 < c_2 | T_1)}. \quad (\text{S1.29})$$

That is,

$$\frac{Pr(T_2 \leq c_1 | T_1)}{Pr(T_2 < c_2 | T_1)} \leq \frac{Pr(T_2 < c_2, T_3 \leq c_1 | T_1)}{Pr(T_2 < c_2, T_3 < c_2 | T_1)}. \quad (\text{S1.30})$$

By Assumption 6 (BMLR), we have

$$\frac{Pr(T_2 \leq x_2 | T_1)}{Pr(T_3 \leq x_2 | T_1)} \geq \frac{Pr(T_2 \leq x_1 | T_1)}{Pr(T_3 \leq x_1 | T_1)}. \quad (\text{S1.31})$$

By (S1.31), to prove (S1.30), it is enough to show

$$\frac{Pr(T_3 \leq c_1 | T_1)}{Pr(T_3 < c_2 | T_1)} \leq \frac{Pr(T_2 < c_2, T_3 \leq c_1 | T_1)}{Pr(T_2 < c_2, T_3 < c_2 | T_1)}. \quad (\text{S1.32})$$

That is,

$$Pr(T_2 < c_2 | T_3 < c_2, T_1) \leq Pr(T_2 < c_2 | T_3 < c_1, T_1). \quad (\text{S1.33})$$

The inequality (S1.33) holds under Assumption 5. Therefore, the inequality (S1.22) holds.

Based on (S1.21)-(S1.22) and Proposition 1, we have

$$\text{mdFWER} = Pr(T_1 \leq c_1) + Pr(T_1 \geq c_2, T_3 \notin (c_1, c_2)) \leq \alpha.$$

Thus, the desired result follows.  $\square$

PROOF OF THEOREM 4. By Corollary 1, without loss of generality, assume that  $\theta_i > 0, i = 1, \dots, n - 1$  and  $\theta_n = 0$ , that is,  $H_i, i = 1, \dots, n - 1$  are false and  $H_n$  is true. Note that

$$\begin{aligned} & \text{mdFWER} \\ = & \sum_{j=1}^{n-1} Pr(T_1 \geq c_2, \dots, T_{j-1} \geq c_2, T_j \leq c_1) \\ & + Pr(T_1 \geq c_2, \dots, T_{n-1} \geq c_2, T_n \notin (c_1, c_2)). \end{aligned} \quad (\text{S1.34})$$

In the following, we prove that

$$\begin{aligned} & Pr(T_1 \geq c_2, \dots, T_{n-2} \geq c_2, T_{n-1} \leq c_1) \\ & + Pr(T_1 \geq c_2, \dots, T_{n-1} \geq c_2, T_n \notin (c_1, c_2)) \\ \leq & Pr(T_1 \geq c_2, \dots, T_{n-2} \geq c_2, T_n \notin (c_1, c_2)). \end{aligned} \quad (\text{S1.35})$$

To prove (S1.35), it is enough to show the following inequality:

$$\begin{aligned} & Pr(T_{n-1} \leq c_1 | T_1, \dots, T_{n-2}) \\ & + Pr(T_{n-1} \geq c_2, T_n \notin (c_1, c_2) | T_1, \dots, T_{n-2}) \\ \leq & Pr(T_n \notin (c_1, c_2) | T_1, \dots, T_{n-2}). \end{aligned} \quad (\text{S1.36})$$

By using the same argument as in proving (S1.23) in the case of three hypotheses, we can prove that the inequality (S1.36) holds under Assumptions

5 and 7. Then, by combining (S1.34) and (S1.35), we have

$$\begin{aligned}
 & \text{mdFWER} \\
 & \leq \sum_{j=1}^{n-2} Pr(T_1 \geq c_2, \dots, T_{j-1} \geq c_2, T_j \leq c_1) \quad (\text{S1.37}) \\
 & \quad + Pr(T_1 \geq c_2, \dots, T_{n-2} \geq c_2, T_n \notin (c_1, c_2)).
 \end{aligned}$$

Note that the right-hand side of (S1.37) is the mdFWER of Procedure 2 when testing  $H_1, \dots, H_{n-2}, H_n$ . By induction and Proposition 1, the mdFWER is bounded above by  $\alpha$ , the desired result.  $\square$