# 行政院國家科學委員會專題研究計畫成



## 多變量無母數檢定於臨床實驗之中間時期分析

### Multivariate Nonparametric Tests for Interim Analysis in Clinical Trial

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主持人:陳怡如 淡江大學統計系

計畫參與人員:章峻福 淡江大學統計系

#### 中文摘要

Mann-Whitney-Wilcoxon 檢定是一種 廣泛被使用在檢定有關兩個獨立樣本來自 兩個母體位置不同的檢定方法。在 1984 年 Hettmansperger 曾討論有關 Mann-Whitney-Wilcoxon 的多變量檢定方法,我們將使用 此多變量檢定方法發展出一個中間時期分 析之多變量檢定方法,同時我們也發展出 另一種分析方法,即應用標準化之 Mann-Whitney-Wilcoxon 多變量檢定統計量。對 於此標準化之多變量檢定統計量的分母部 分將使用 Hettmansperger 之估計值與拔靴 帶估計值。

**關鍵詞**: 中間時期分析、 臨床實驗、 Mann-Whitney-Wilcoxon 檢定、 一致性估 計量、自重抽法。

#### **Abstract**

The Mann-Whitney-Wilcoxon test is a very widely used procedure for testing the null hypothesis that two independent samples have been drawn from the population which are equal and is sensitive to different in A multivariate version of the Mann-Whitney-Wilcoxon test was proposed by Hettmansperger (1984). We use these multivariate methods to develop an interim analysis for this multivariate test. Also, we propose to develop an alternative analysis by employing the standardized multivariate version of Mann-Whitney-Wilcoxon test Two approaches for statistics. the unknown denominator of the standardized multivariate test statistics are employed by the Hettmasperger's estimate and bootstrap estimate.

**Keywords**: Interim analysis, clinical trial, multivariate version of Mann-Whitney-Wilcoxon, consistent estimate, bootstrap method.

#### 1. Introduction

The procedure for testing the null hypothesis of equal population location parameters was proposed by Mann, Whitney and Wilcoxon. Hettmansperger (1984) considered this test in multivariate case. We would like to develop an interim analysis for this multivariate test.

Suppose that we have two independent random samples  $\bar{X}' = (X_{1i}, X_{2i})$  and  $\bar{Y}' = (Y_{1j}, Y_{2j})$  for  $i = 1, \cdots, n$ ;  $j = 1, \cdots, m$  from the bivariate distribution with cdfs  $F(v_1, v_2)$  and  $F(v_1 - \Delta_1, v_2 - \Delta_2)$ , respectively. Let the parameter  $\bar{\Delta}' = (\Delta_1, \Delta_2)$  represent the shift from the  $\bar{X}$  distribution to the  $\bar{Y}$  distribution for the two components. We would like to employ the multivariate version of Mann-Whitney-Wilcoxon test to conduct one interim analysis to test  $H_0: \bar{\Delta} = \bar{0}$  versus  $H_1: \bar{\Delta} \neq \bar{0}$  at time  $T_1$  and continue following subjects until  $T_2$  if  $H_0$  is not rejected at time

We will assume that the joint distribution of  $\bar{X}' = (X_{11}, X_{12}, X_{21}, X_{22})$  and  $\vec{Y}' = (Y_{11}, Y_{12}, Y_{21}, Y_{22})$  have the covariance matrix  $\Sigma = \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{bmatrix}$  , where  $\boldsymbol{\Sigma}_{11} = \boldsymbol{\Sigma}_{22} = \begin{bmatrix} \boldsymbol{\sigma}_1^2 & \boldsymbol{\sigma}_{12} \\ \boldsymbol{\sigma}_{21} & \boldsymbol{\sigma}_2^2 \end{bmatrix}, \quad \boldsymbol{\Sigma}_{12} = \boldsymbol{\Sigma}_{21} = \begin{bmatrix} \boldsymbol{\sigma}_3 & \boldsymbol{\sigma}_4 \\ \boldsymbol{\sigma}_4 & \boldsymbol{\sigma}_3 \end{bmatrix},$  $(X_{pli}, X_{p2i})$  and  $(Y_{pli}, Y_{p2i})$  are independent bivariate samples observed at time  $i = 1, \dots, n$ ;  $j = 1, \dots, m$  and p = 1, 2. Let  $\vec{U}' = (\vec{U}_1, \vec{U}_2) = (U_{11}, U_{12}, U_{21}, U_{22})$  denote the multivariate version of the Mann-Whitney-Wilcoxon test statistic, where  $\bar{U}_1' = (U_{11}, U_{12})$ ,  $\vec{U}_2' = (U_{21}, U_{22}), \qquad U_{1q} = \sum_{i=1}^m (\frac{R_{1qi}}{N+1} - \frac{1}{2})$ time  $T_{1}$  $U_{2q} = \sum_{i=1}^{m} \left( \frac{R_{2qy}}{N+1} - \frac{1}{2} \right)$  is computed at time  $T_2$ and  $R_{pqj}$  is the rank of  $Y_{pqj}$  in the combined sample of the  $q^{th}$  component at time  $T_n$ , q = 1, 2; p = 1, 2; N = n + m. Under  $H_0$ , it can be shown that  $ar{U}$  converges in distribution to a multivariate normal with zero means and covariance matrix, say  $\Sigma^*$ . Since  $\Sigma^*$  is unknown, Hettmansperger proposed a consistent estimate of  $\Sigma^*$ ,  $\Sigma^* = \begin{bmatrix} \hat{\Sigma}_{11}^* & \hat{\Sigma}_{12}^* \\ \hat{\Sigma}_{21}^* & \hat{\Sigma}_{22}^* \end{bmatrix}$ , where  $\hat{\Sigma}_{11}^* = \begin{bmatrix} \hat{\sigma}_1^{2^{(1)}} & \hat{\sigma}_{12}^{(1)} \\ \hat{\sigma}_{12}^{(1)} & \hat{\sigma}_{2}^{2^{(1)}} \end{bmatrix}$  is estimated by the ranks at time  $T_1$  and  $\hat{\Sigma}_{22}^* = \begin{bmatrix} \hat{\sigma}_1^{2^{(2)}} & \hat{\sigma}_{12}^{(2)} \\ \hat{\sigma}_{12}^{(2)} & \hat{\sigma}_2^{2^{(2)}} \end{bmatrix}$  is estimated by the ranks at time  $T_2$ . Define  $U_1^* = \frac{1}{N} \vec{U}_1' \hat{\Sigma}_{11}^{*-1} \vec{U}_1$  and  $U_2^* = \frac{1}{N} \vec{U}_2' \hat{\Sigma}_{22}^{*-1} \vec{U}_2$ . To conduct the multivariate test, we have to find out the joint distribution. Unfortunately, the exact distribution is unknown, but the joint

limiting distribution can be proved to follow

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Jensen's type bivariate distribution (1970).

class of bivariate chi distribution is relevant to some classical problems in statistical inference, such as the joint distribution of dependent quadratic forms which arise in connection with two interim analyses.  $\vec{V}'_j = (V_{1j}, V_{2j})$  be independent random vectors distributed as  $N_2(\vec{0}, \Sigma)$ , where  $\Sigma = \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix}$ . The distribution of  $\bar{V}$  has been addressed by Hewett and Tsutakawa (1972), Hewett and Supurrier (1979). Chin and Hewett (1988). Gunst and Webser (1973) discussed some distributional properties and applications of the baivariate chi square distribution. Jensen (1970) generalized the case of the bivariate chi square distribution when  $\Sigma = \begin{vmatrix} I & S' \\ S & I \end{vmatrix}$  is a nonsingular matrix. The previous bivaraite chi square distribution is a special case of the

Jensen's type bivaraite chi square distribution.

#### 2. One Interim Analysis

The testing procedure for one interim analysis is as follows:  $H_0$  is rejected at time  $T_1$  if  $U_1^* \ge b_1$ ; if  $U_1^* < b_1$ , continue following the subjects until time  $T_2$ . At time  $T_2$ , if  $U_2^* \ge b_2$ , reject  $H_0$ ; if  $U_2^* < b_2$ , do not reject  $H_0$ . We determine the  $\alpha$ ,  $\alpha_1$ ,  $\alpha_2$  such that  $\alpha = \alpha_1 + \alpha_2$  in advance. The values of  $b_1$  and  $b_2$  are obtained by solving the equations:  $\alpha_1 = P(U_1^* \ge b_1), \alpha_2 = P\{(U_1^* < b_1) \cap (U_2^* \ge b_2)\}$ where the probabilities are computed under  $H_0$ . Thus, to solve these equations, we have to know the joint distribution of  $U_1^*$  and  $U_2^*$ under  $H_0$ . The exact joint distribution is unknown, but we will show that the joint limiting distribution of  $(U_1^*, U_2^*)'$  follows the Jensen's type bivariate chi square distribution.

**Theorem 1**: Let  $\bar{U}' = (\bar{U}_1, \bar{U}_2)$ , where  $\bar{U}_1' = (U_{11}, U_{12})$ ,  $\bar{U}_2' = (U_{21}, U_{22})$ , are the multivariate version of the Mann-Whitney-Wilcoxon test statistics at two time periods. Suppose that  $n, m \to \infty$ ,  $\frac{n}{N} \to \lambda$ ,  $0 < \lambda < 1$  and under  $H_0$ ,  $\frac{1}{\sqrt{N}} \bar{U}' \xrightarrow{D} N_4(\bar{0}, \Sigma)$ , where  $\Sigma$  is positive definite matrix and  $\Sigma^*$  is Hettmansperger's consistent estimate of  $\Sigma$ . Then  $\frac{1}{N} (\bar{U}_1' \hat{\Sigma}_{11}^{-1} \bar{U}_1, \bar{U}_2' \hat{\Sigma}_{22}^{-1} \bar{U}_2) \xrightarrow{D}$  Jensen's type bivariate chi square distribution.

By the spectral decomposition, there exists the orthogonal matrices  $P = (\vec{e}_1, \vec{e}_2)$ and  $Q = (\vec{f}_1, \vec{f}_2)$ , where  $\vec{e}_1$  and  $\vec{e}_2$  are the eigenvectors  $\Sigma_{11}^{-1/2}\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}\Sigma_{11}^{-1/2} \quad \text{and} \quad \bar{f}_1 \quad \text{and} \quad \bar{f}_2 \quad \text{are the}$ eigenvectors  $\Sigma_{22}^{-1/2} \Sigma_{21} \Sigma_{11}^{-1} \Sigma_{12} \Sigma_{22}^{-1/2}$ . Define  $\vec{U}^* = \frac{1}{\sqrt{N}} A \vec{U}$ , where  $A = Diag(P'\Sigma_{11}^{-1/2}, Q'\Sigma_{22}^{-1/2})$ . It can be that the  $\vec{U}^* = \frac{1}{\sqrt{N}} (\vec{e}_1' \Sigma_{11}^{-1/2} U_{11})$ ,  $\bar{e}_{2} \Sigma_{11}^{-1/2} U_{12}, \bar{f}_{1} \Sigma_{22}^{-1/2} U_{21}, \bar{f}_{2} \Sigma_{22}^{-1/2} U_{22})'$  $\bar{U}^* \xrightarrow{D} N_4(\hat{0}, \Sigma^*), \text{ where } \Sigma^* = \begin{vmatrix} I & \Sigma_{12}^* \\ \Sigma_{21}^* & I \end{vmatrix},$  $\Sigma_{12}^* = \Sigma_{21}^* = \begin{bmatrix} \rho_1 & 0 \\ 0 & \rho_2 \end{bmatrix}$  where  $\rho_1$  and  $\rho_2$  are the square roots of the eigenvalues of  $\Sigma_{11}^{-1/2}\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}\Sigma_{11}^{-1/2}\Sigma_{22}^{-1/2}\Sigma_{21}\Sigma_{11}\Sigma_{12}\Sigma_{22}^{-1/2}$  $\frac{1}{N}(\bar{U}_{1}^{\prime}\Sigma_{11}^{-1}\bar{U}_{1},\bar{U}_{2}^{\prime}\Sigma_{22}^{-1}\bar{U}_{2}) = (U_{11}^{*} + U_{12}^{*},U_{21}^{*} + U_{22}^{*})^{\prime},$ then  $\frac{1}{N}(\vec{U}_1'\hat{\Sigma}_{11}^{-1}\vec{U}_1,\vec{U}_2'\hat{\Sigma}_{22}^{-1}\vec{U}_2) \stackrel{D}{\longrightarrow} \text{Jensen's}$ type bivariate chi square distribution. The

proof is completed.

#### 3. Alternative Analysis

Instead of using quadratic form of test statistics  $U_1^*$  and  $U_2^*$ , we propose to develop an alternative interim analysis for testing  $H_0: \bar{\Delta} = \bar{0}$  versus  $H_1: \bar{\Delta} \neq \bar{0}$ . Let  $\bar{V}' = (V_{11}, V_{12}, V_{21}, V_{22})$  be the standardized multivariate version of Mann-Whitney-Wilcoxon test statistics, where  $V_{ij} = \frac{U_{ij}}{\sqrt{\hat{\sigma}_j^{2(i)}}}$  and  $\sigma_j^{2(i)}$  is the

Hettmansperger's estimated variance of the  $j^{th}$  variable at time  $T_i$  for  $i \ j=1,2$ .  $H_0$  is rejected at time  $T_1$  if  $V_{11} \ge c_1$  or  $V_{12} \ge c_1$ . Otherwise we continue the test until time  $T_2$ . At time  $T_2$ , if  $V_{21} \ge c_2$  or  $V_{22} \ge c_2$ , reject  $H_0$ ; if  $V_{21} < c_2$  and  $V_{22} < c_2$ , do not reject  $H_0$ .

**Theorem** 2: If  $\vec{V}' = (V_{11}, V_{12}, V_{21}, V_{22})$  is the standardized multivariate version of Mann-Whitney-Wilcoxon test statistics for testing  $H_0: \vec{\Delta} = \vec{0}$ , then  $\vec{V}' = (V_{11}, V_{12}, V_{21}, V_{22})$  converges in distribution to a multivariate normal with zero means and covariance matrix, say  $\Sigma_{(\vec{V})}$ .

The proof is quite straightforward. According to Hettmansperger's result, we can obtain that  $\bar{U}' = (U_{11}, U_{12}, U_{21}, U_{22})$  is converges to the multivariate normal. Then the limiting distribution of the standardized random vector  $\bar{V}$  with the consistent estimator of the variance is also a multivariate normal.

Under  $H_0$ , the probabilities of  $\alpha_1$ , and  $\alpha_2$  are as follows:

$$\alpha_1 = P(\{V_{11} \ge c_1\}) \cup \{V_{12} \ge c_1\})$$

$$= P(\{V_{11} \ge c_1\}) + P(\{V_{12} \ge c_1\})$$

$$- P(\{V_{11} \ge c_1\}) \cap \{V_{12} \ge c_1\})$$

$$\alpha_2 = P[(\{V_{11} < c_1\} \cap \{V_{12} < c_1\}) \cap (\{V_{21} \ge c_2\} \cup \{V_{22} \ge c_2\})]$$

$$= P[(\{V_{11} < c_1\} \cap \{V_{12} < c_1\} \cap (\{V_{21} \ge c_2\}) \cup \{V_{22} \ge c_2\})]$$

$$\begin{split} & (\{V_{11} < c_1\} \cap \{V_{12} < c_1\}) \cap \{V_{22} \ge c_2\})] \\ &= P[(\{V_{11} < c_1\} \cap \{V_{12} < c_1\} \cap (\{V_{21} \ge c_2\})] + \\ & P[(\{V_{11} < c_1\} \cap \{V_{12} < c_1\}) \cap \{V_{22} \ge c_2\})] - \\ & P(\{V_{11} < c_1\} \cap \{V_{12} < c_1\} \cap (\{V_{21} \ge c_2\}) \cap \{V_{22} \ge c_2\}) \end{split}$$

Since the joint limiting distribution of  $V_{11}$  and  $V_{12}$  is bivariate normal, we may use IMSL subroutines ANORDF. BNRDF and ZBREN to solve the critical value  $c_1$ . We use the fact that the joint limiting distribution of both  $(V_{11}, V_{12}, V_{21})'$  and  $(V_{11}, V_{12}, V_{22})'$  are trivariate normal distributions and  $(V_{11}, V_{12}, V_{21}, V_{22})'$  has a limiting multivariate normal distribution and IMSL subroutines MULNOA and ZBREN to solve the value  $c_2$ . We find that the results for the estimated variance in the denominator of the test statistics using either by the Hettmansperger consistent estimate or by the bootstrap estimate are very similar.

#### 4. Evaluation of Project

The main parts of this project has been completed. We have derived the joint limiting distribution of the proposed test for one interim analysis. The theoretical part is done, but the application has not completed yet due to the complicated form of the density. The possible future research is to compute the probabilities of Jensen's bivariate chi square distribution. For the second method, we also obtain the joint limiting distribution and find the behavior of limiting distribution is really close to multivariate normal. The other possible future research is to show that the consistency of the bootstrap estimate.

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