

## Appendix

### Proof of theorem 1

Let  $G_j^{(r)} \sim N\left(\frac{\sum_{i=1}^{(n_1 < \Lambda < n_r)(m_1 < \Lambda < m_r)} R_{ij}}{(n_1 < \Lambda < n_r)(m_1 < \Lambda < m_r)}\right)$  and  $T_j^{(r)} \sim N\left(\sqrt{n_1 < \Lambda < n_r} [G_j^{(r)} > E(G_j^{(r)})]\right)$  for  $r \in \{1, 2, \dots, K\}$ ,  $s \in \{1, 2, \dots, p\}$  ;  $j \in \{1, 2, \dots, K, p\}$ .

By the conditional expectation, it can be shown that

$$\begin{aligned} \text{Var}(T_j^{(r)}) &\sim \frac{n_1 < \Lambda < n_r}{m_1 < \Lambda < m_r} t_{j2}^2 < \frac{1}{m_1 < \Lambda < m_r} \left(\frac{p^2 > 1}{12}\right) > t_{j1}^2 > t_{j2}^2 \\ \text{Var}(T_p^{(r)}) &\sim \frac{1}{\sum_{j=0}^{p-1}} [t_{j1}^2 < \frac{n_1 < \Lambda < n_r}{m_1 < \Lambda < m_r} t_{j2}^2] < \frac{1}{\sum_{j=0}^{p-1}} [t_{j3}^2 < \frac{n_1 < \Lambda < n_r}{m_1 < \Lambda < m_r} t_{j4}^2] \\ &< \frac{1}{m_1 < \Lambda < m_r} \frac{p^2 > 1}{12} > \sum_{j=0}^{p-1} (t_{j1}^2 < t_{j2}^2) > \sum_{j=0}^{p-1} (t_{j3}^2 < t_{j4}^2) \end{aligned}$$

and

$$\begin{aligned} \text{Var}\left(\sum_{j=1}^p j T_j^{(r)}\right) &\sim \sum_{j=1}^{p-1} (p-j)^2 [t_{j1}^2 < \frac{n_1 < \Lambda < n_r}{(m_1 < \Lambda < m_r)} t_{j2}^2] \\ &< \sum_{j=0}^{p-1} (p-j)(p-j) \lambda t_{j3}^2 < \frac{n_1 < \Lambda < n_r}{m_1 < \Lambda < m_r} t_{j4}^2 \\ &< \frac{1}{m_1 < \Lambda < m_r} \left(\frac{p^2 > 1}{12}\right) > \sum_{j=1}^{p-1} (p-j)^2 (t_{j1}^2 < t_{j2}^2) \\ &> \frac{1}{m_1 < \Lambda < m_r} \sum_{j=0}^{p-1} (p-j)(p-j) (t_{j3}^2 < t_{j4}^2) \end{aligned}$$

where

$$\begin{aligned} t_{j1}^2 &\sim \text{Var}(E[(R_{ij} > \frac{p-1}{2}) | \bar{X}_i]) , \quad t_{j2}^2 \sim \text{Var}(E[(R_{ij} > \frac{p-1}{2}) | \bar{Y}_k]) , \\ t_{j3}^2 &\sim \text{Cov}(E[(R_{ij} > \frac{p-1}{2}) | \bar{X}_i], E[(R_{ij} > \frac{p-1}{2}) | \bar{X}_i]) , \\ t_{j4}^2 &\sim \text{Cov}(E[(R_{ij} > \frac{p-1}{2}) | \bar{Y}_k], E[(R_{ij} > \frac{p-1}{2}) | \bar{Y}_k]) \end{aligned}$$

for  $j \in \{0, 1, \dots, j\}$ ;  $j \in \{1, 2, \dots, K, p\}$

Let  $T_j^{(r)*}$  be the projection of  $T_j^{(r)}$ , then  $T_j^{(r)*} \sim N\left(\sqrt{n_1 < \Lambda < n_r} [G_j^{(r)*} > \frac{p-1}{2}]\right)$ . By the projection theorem,

$$\begin{aligned} \sum_{j=1}^k j T_j^{(r)*} &\sim \sqrt{n_1 < \Lambda < n_r} \frac{1}{n_1 < \Lambda < n_r} \sum_{i=1}^{(n_1 < \Lambda < n_r)} V(\bar{X}_i) \\ &< \sqrt{n_1 < \Lambda < n_r} \frac{1}{m_1 < \Lambda < m_r} \sum_{k=1}^{(m_1 < \Lambda < m_r)} U(\bar{Y}_k) \end{aligned}$$

where

$$\begin{aligned} V(\bar{X}_i) &\sim \sum_{j=1}^{p-1} (p-j) E[(R_{ij} > E(R_{ij})) | \bar{X}_i] . \\ U(\bar{Y}_k) &\sim \sum_{j=1}^{p-1} (p-j) E[(R_{ij} > E(R_{ij})) | \bar{Y}_k] , \quad \sum_{i=1}^{(n_1 < \Lambda < n_r)} V(\bar{X}_i) \quad \text{and} \end{aligned}$$

$\sum_{k=1}^{(m_1 < \Lambda < m_r)} U(\bar{Y}_k)$  are independent, and  $\sum_{j=1}^p j T_j^{(r)*} \stackrel{D}{\sim} N(0, \sum_{j=1}^{p-1} (p-j)^2 (t_{j1}^2 < 3t_{j2}^2) < \sum_{j=0}^{p-1} (p-j)(p-j) \lambda t_{j3}^2 < 3t_{j4}^2)$ .

Thus,

$$\begin{aligned} \text{Var}\left(\sum_{j=1}^p j T_j^{(r)*}\right) &\sim \sum_{j=1}^{p-1} (p-j)^2 (t_{j1}^2 < 3t_{j2}^2) < \sum_{j=0}^{p-1} (p-j)(p-j) \lambda t_{j3}^2 < 3t_{j4}^2 \quad \text{for} \\ n_1 &\sim m_1, n_2 \sim m_2, \Lambda, n_r \sim m_r, \quad \frac{n_1 < \Lambda < n_r}{m_1 < \Lambda < m_r} \in \mathcal{J} \quad \text{as} \\ n_1, \Lambda, n_r, m_1, \Lambda, m_r &\in \mathcal{L}. \quad \text{Thus, the null limiting} \\ \text{distribution of} & \\ \sum_{r=1}^s a_r W_r &\sim \sum_{r=1}^s a_r \left[ \frac{L_{1r} > E(L_{1r})}{\sqrt{\text{Var}(L_{1r})}} \right] \\ &\sim \sum_{r=1}^s a_r \frac{\sum_{j=1}^p j T_j^{(r)*}}{\sqrt{\text{Var}\left(\sum_{j=1}^p j T_j^{(r)*}\right)}} < a_1 \frac{o_p\left(\frac{1}{\sqrt{n_1}}\right)}{\frac{1}{\sqrt{n_1}} \sqrt{\text{Var}\left(\sum_{j=1}^p j T_j^{(1)*}\right)}} \\ &< \Lambda < a_s \frac{o_p\left(\frac{1}{\sqrt{n_1 < \Lambda < n_r}}\right)}{\frac{1}{\sqrt{n_1 < \Lambda < n_r}} \sqrt{\text{Var}\left(\sum_{j=1}^p j T_j^{(s)*}\right)}} \end{aligned}$$

is normal. By the result of the null limiting distribution for any arbitrary linear combination  $\sum_{r=1}^s a_r W_r$  is normal, we can obtain that the joint null limiting distribution of  $\sum_{r=1}^s W_r$  is multivariate normal.

### Proof of the theorem 2

By the previous result, it can be shown that

$$\begin{aligned} W_{2r} &\sim \frac{L_{2r} > E(L_{2r})}{\sqrt{\text{Var}(L_{2r})}} \sim \frac{\sum_{j=1}^{p_r} j T_j^{(r)*}}{\sqrt{\text{Var}\left(\sum_{j=1}^{p_r} j T_j^{(r)*}\right)}} < \frac{o_p\left(\frac{1}{\sqrt{n}}\right)}{\frac{1}{\sqrt{n}} \sqrt{\text{Var}\left(\sum_{j=1}^{p_r} j T_j^{(r)*}\right)}} , \quad \text{where} \\ T_j &\sim \sqrt{n} \frac{\sum_{i=1}^{nm} R_{ij} > E}{nm} . \quad \text{Since } T_j^* \text{ is the project of} \end{aligned}$$

$T_j$  and can be written as  $T_j^* \sim \sqrt{n} \frac{1}{n} \sum_{i=1}^n E(R_{ij} > \frac{p-1}{2} | \bar{X}_i) < \frac{1}{m} \sum_{k=1}^m E(R_{ij} > \frac{p-1}{2} | \bar{Y}_k)$ . For any constants  $b_1, b_2, b_3$ , it shows that the null limiting distribution of  $\sum_{r=1}^3 b_r W_{2r}$  is normal as  $n \sim m$  and  $\frac{n}{m} \in \mathcal{L}$ . Hence, the proof is completed.