

Probability of detecting co-clusters and setting parameters

1 Notation

- $A \in \mathbb{R}^{M \times N}$ is a matrix with K co-clusters (co-cluster set $C = \{C_k\}_{k=1}^K$);
- A is partitioned into $m \times n$ blocks, each block has size $\phi_i \times \psi_j$, that is, $M = \sum_{i=1}^m \phi_i$ and $N = \sum_{j=1}^n \psi_j$;
- thus block set $B = \{B_{(i,j)}\}_{i=1}^m, j=1}^n$;
- the size of co-cluster $C_k \in \mathbb{R}^{M^{(k)} \times N^{(k)}}$ that falls into block $B_{(i,j)}$ is $M_{(i,j)}^{(k)} \times N_{(i,j)}^{(k)}$;
- T_m is the minimum number of rows, T_n is the minimum number of columns.

2 Probability

Consider co-cluster C_k ,

$$P(M_{(i,j)}^{(k)} = \alpha) = \frac{\binom{M^{(k)}}{\alpha} \binom{M - M^{(k)}}{\phi_i - \alpha}}{\binom{M}{\phi_i}}$$

$$P(N_{(i,j)}^{(k)} = \beta) = \frac{\binom{N^{(k)}}{\beta} \binom{N - N^{(k)}}{\psi_j - \beta}}{\binom{N}{\psi_j}}$$

The tail probability of $M_{(i,j)}^{(k)}$ and $N_{(i,j)}^{(k)}$ are

$$P(M_{(i,j)}^{(k)} < T_m) = \sum_{\alpha=1}^{T_m-1} P(M_{(i,j)}^{(k)} = \alpha)$$

$$\leq \exp(-2(s_i^{(k)})^2 \phi_i)$$

where $s_i^{(k)} = \frac{M^{(k)}}{M} - \frac{T_m - 1}{\phi_i}$, and

$$P(N_{(i,j)}^{(k)} < T_n) = \sum_{\beta=1}^{T_n-1} P(N_{(i,j)}^{(k)} = \beta)$$

$$\leq \exp(-2(t_j^{(k)})^2 \psi_j)$$

where $t_j^{(k)} = \frac{N^{(k)}}{N} - \frac{T_n - 1}{\psi_j}$.

The joint probability of $M_{(i,j)}^{(k)}$ and $N_{(i,j)}^{(k)}$ are

$$\begin{aligned} P(M_{(i,j)}^{(k)} < T_m, N_{(i,j)}^{(k)} < T_n) &= \sum_{\alpha=1}^{T_m-1} \sum_{\beta=1}^{T_n-1} P(M_{(i,j)}^{(k)} = \alpha) P(N_{(i,j)}^{(k)} = \beta) \\ &\leq \exp[-2(s_i^{(k)})^2 \phi_i + -2(t_j^{(k)})^2 \psi_j] \end{aligned}$$

If $\phi_i = p$ and $\psi_j = q$ for all i and j , then

Suppose event ω_k is that co-cluster C_k can't be find in any block $B_{(i,j)}$, then

$$\begin{aligned} P(\omega_k) &= \prod_{i=1}^m \prod_{j=1}^n P(M_{(i,j)}^{(k)} < T_m, N_{(i,j)}^{(k)} < T_n) \\ &\leq \prod_{i=1}^m \prod_{j=1}^n \exp\{-2[(s_i^{(k)})^2 \phi_i + (t_j^{(k)})^2 \psi_j]\} \\ &= \exp\{-2 \sum_{i=1}^m \sum_{j=1}^n [(s_i^{(k)})^2 \phi_i + (t_j^{(k)})^2 \psi_j]\} \end{aligned}$$

If $\phi_i = \phi$ and $\psi_j = \psi$ for all i and j , then

$$\begin{aligned} s_i^{(k)} &= s^{(k)} = \frac{M^{(k)}}{M} - \frac{T_m - 1}{\phi} \\ t_j^{(k)} &= t^{(k)} = \frac{N^{(k)}}{N} - \frac{T_n - 1}{\psi} \end{aligned}$$

$$P(\omega_k) \leq \exp\{-2[\phi m (s^{(k)})^2 + \psi n (t^{(k)})^2]\}$$

And if we do T_p times of random sampling, the Probability of detecting the co-cluster is

$$\begin{aligned} P &= 1 - P(\omega_k)^{T_p} \\ &\geq 1 - \exp\{-2T_p[\phi m (s^{(k)})^2 + \psi n (t^{(k)})^2]\} \end{aligned}$$

according to which, we can set $m, n, \phi, \psi, T_m, T_n$ and T_p to ensure the probability of detecting the co-cluster is larger than a given threshold.

3 Nosi case

3.1 Assumption

Assume each noise n_{ij} complies with a normal distribution $N(0, \sigma^2)$, i.i.d. for all i and j . Suppose $\exists \lambda > 0$, such that

$$\lambda \leq \max(\|B\|_1, \|B^\top\|_1) / \sigma^2.$$

3.2 Score

The score of a submatrix $A_{I,J}$ is defined as

$$S(I, J) = \min(S_{row}(I, J), S_{col}(I, J)) \quad (1)$$

$$S_{row}(I, J) = \min_{i_1, i_2 \in I} \left(1 - \frac{1}{|I| - 1} \sum_{i_2 \in I, i_2 \neq i_1} \langle x_{i_1, J}, x_{i_2, J} \rangle \right) \quad (2)$$

$$S_{col}(I, J) = \min_{j_1, j_2 \in J} \left(1 - \frac{1}{|J| - 1} \sum_{j_2 \in J, j_2 \neq j_1} \langle x_{I, j_1}, x_{I, j_2} \rangle \right) \quad (3)$$

where $x_{i,J}$ is the i -th row of $A_{I,J}$. Here we define the inner product of two vectors x and y as

$$\langle x, y \rangle = \exp\left(-\frac{\|x - y\|_1^2}{2\alpha\|x\|_1\|y\|_1}\right)$$

Suppose A is a hidden co-cluster matrix, and E is the noise matrix. Then the observed matrix is $B = A + E$. Consider co-cluster $B_{I,J}$, denote $1 - \frac{1}{|I|-1} \sum_{i_2 \in I, i_2 \neq i_1} \langle x_{i_1, J}, x_{i_2, J} \rangle$ as $s_{row}(i_1, i_2, J)$, then

$$\begin{aligned} \mathbb{E}(s_{row}(i_1, i_2, J)) &= 1 - \frac{1}{|I| - 1} \sum_{i_2 \in I, i_2 \neq i_1} \mathbb{E}(\langle x_{i_1, J}, x_{i_2, J} \rangle) \\ &= 1 - \frac{1}{|I| - 1} \sum_{i_2 \in I, i_2 \neq i_1} \exp\left(-\frac{\|x_{i_1, J} - x_{i_2, J}\|_1^2}{2\alpha\|x_{i_1, J}\|_1\|x_{i_2, J}\|_1}\right) \\ &\geq 1 - \exp\left(-\frac{2}{\alpha \min(\|x_{i_1, J}\|_1, \|x_{i_2, J}\|_1)}\right) \\ &\geq 1 - \exp\left(-\frac{2}{\alpha \max(\|B\|_1, \|B^\top\|_1)}\right) \\ \sigma^2(s_{row}(i_1, i_2, J)) &= \frac{1}{|I| - 1} \sum_{i_2 \in I, i_2 \neq i_1} \sigma^2(\langle x_{i_1, J}, x_{i_2, J} \rangle) \\ &= \frac{1}{|I| - 1} \sum_{i_2 \in I, i_2 \neq i_1} \exp\left(-\frac{\|x_{i_1, J} - x_{i_2, J}\|_1^2}{2\alpha\|x_{i_1, J}\|_1\|x_{i_2, J}\|_1}\right) \\ &\leq \sigma^2 \exp\left(-\frac{2}{\alpha \min(\|x_{i_1, J}\|_1, \|x_{i_2, J}\|_1)}\right) \\ &\leq \sigma^2 \exp\left(-\frac{2}{\alpha \max(\|B\|_1, \|B^\top\|_1)}\right) \end{aligned}$$

Thus the expected value of $S_{row}(I, J)$ satisfies

$$\begin{aligned} \mathbb{E}(S_{row}(I, J)) &\geq |J||I| \left(1 - \exp\left(-\frac{2}{\alpha \max(\|B\|_1, \|B^\top\|_1)}\right) \right) \\ \sigma^2(S_{row}(I, J)) &\leq |I||J| \sigma^2 \exp\left(-\frac{2}{\alpha \max(\|B\|_1, \|B^\top\|_1)}\right) \end{aligned}$$

Similarly, we can get

$$\begin{aligned}\mathbb{E}(S_{col}(I, J)) &\geq |J||I| \left(1 - \exp\left(-\frac{2}{\alpha \max(\|B\|_1, \|B^\top\|_1)}\right)\right) \\ \sigma^2(S_{col}(I, J)) &\leq |I||J|\sigma^2 \exp\left(-\frac{2}{\alpha \max(\|B\|_1, \|B^\top\|_1)}\right)\end{aligned}$$

Then since $x^2 \leq x$ for $x \in (0, 1)$, we have

$$\begin{aligned}\mathbb{E}(S(I, J)) &\geq 1 - \exp\left(-\frac{2}{\alpha \max(\|B\|_1, \|B^\top\|_1)}\right) \\ \sigma^2(S(I, J)) &\leq |I||J|\sigma^2 \exp\left(-\frac{2}{\alpha \max(\|B\|_1, \|B^\top\|_1)}\right)\end{aligned}$$

According to the Chernoff bound, we have

$$\begin{aligned}P(S(I, J) \leq \mathbb{E}(S(I, J)) - \epsilon) &\leq \exp\left(-\frac{\epsilon^2}{2\sigma^2(S(I, J))}\right) \\ &\leq \exp\left(-\frac{\epsilon^2}{2|I||J|\sigma^2 \exp\left(-\frac{2}{\alpha \max(\|B\|_1, \|B^\top\|_1)}\right)}\right) \\ &\leq \exp\left(-\frac{\epsilon^2}{2|I||J|\sigma^2 \exp\left(-\frac{2}{\alpha\lambda}\right)}\right)\end{aligned}$$

Thus if we set $\epsilon = \sqrt{2|I||J|\sigma^2 \exp\left(-\frac{2}{\alpha\lambda}\right) \log(1/\delta)}$, then

$$P(S(I, J) \leq \mathbb{E}(S(I, J)) - \epsilon) \leq \delta$$

Combine with the probability control of T_p , we can select parameters to ensure the probability of detecting the co-cluster is larger than a given threshold.