What I have got.

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We mainly consider two cases: 1) without intercept but with scales, i.e., $\Omega_k = \sum_{r=1}^R u_{kr} \Theta_r$; 2) with intercept and the scales, i.e, $\Omega_k = \sum_{r=1}^R \Theta_0 + u_{kr} \Theta_r$.

1 Combined estimation error

- 1. Case 1:
 - I have found the accuracy for the simple case with out the penalty, i.e, R=1 in Note 0324. The objective function is

$$\min_{U,\Theta} \quad \sum_{k=1}^{K} \langle S_k, u_k \Theta \rangle - \log \det(\Theta),$$

where $u_k \ge a > 0$.

Lemma 1. Suppose $\Omega_k = u_k^* \Theta^*$ for $u_k \ge a > 0$. Then, there exists a local minimizer $(\hat{U}, \hat{\Theta})$ such that

$$\sum_{k=1}^{K} \left\| \hat{u}_k \hat{\Theta} - u_k^* \Theta^* \right\|_F \le C \sqrt{\frac{p^2 \log pK}{n}}.$$

• I have found the accuracy for the complex case with sparse penalty but with known membership matrix in 0611. The objective function is

$$\min_{\Theta_r} \quad \sum_{r=1}^{R} \sum_{k \in I_r} \langle S_k, u_{kr}^* \Theta_r \rangle - \log \det(\Theta_r) + \lambda |I_r| \|\Theta_r\|_1,$$

where $u_k^* \ge a > 0$.

Lemma 2. Suppose $\|\Theta_r^*\|_0 \leq s$ and $\lambda \geq \max_r C\sqrt{\frac{\log p}{|I_r|}}$, then there exists an optimal solution $\hat{\Theta}_r$ such that

$$\sum_{r=1}^{R} \sum_{k \in I} u_{kr}^* \left\| \hat{\Theta}_r - \Theta_r^* \right\| \le CR \sqrt{\frac{s \log p}{n}}.$$

2. Case 2:

• I have found the accuracy for the complex case with sparse penalty but with known membership matrix in 0611.

The objective function is

$$\min_{\Theta_r} \sum_{r=1}^{R} \sum_{k \in I_r} \langle S_k, \Theta_0 + u_{kr}^* \Theta_r \rangle - \log \det(\Theta_0 + u_{kr}^* \Theta_r) + \lambda |I_r| \|\Theta_r\|_1 + K \|\Theta_0\|_1,$$

where $\sum_{k \in [K]} u_{kr} = 0$.

Lemma 3. Suppose $\|\Theta_r^*\|_0 \leq s$ and $\lambda \geq \max_r C\sqrt{\frac{\log p}{n|I_r|}}$, then there exists an optimal solution $\hat{\Theta}_r$ such that

$$\sum_{r=1}^{R} \sum_{k \in I_r} \left\| \hat{\Theta}_0 - \Theta_0^* + u_{kr}^* [\hat{\Theta}_r - \Theta_r^*] \right\| \le C \sqrt{\frac{s \log pK}{n}}$$

2 Misclassification rate

Let $\ell(U,\Theta)$ denote the population-based objective function with penalty, i.e., replacing S_k by true Σ_k in objective. Let $\tilde{\Theta} = \arg\min \ell(\Theta)$ with given U.

- 1. Case 1:
 - I have found the perturbation misclassification rate in Note 0622.

Lemma 4. Suppose $MCR(U, U^*) \ge \epsilon$ and the minimal gap between Θ_r^* is delta. For λ small enough, we have

$$G(U^*) - G(U) \le \epsilon \delta \left\{ -\frac{1}{8\tau^2} \delta + \left(\frac{1}{2\tau^2} e + \lambda \sqrt{p} \right) \right\}, \tag{1}$$

where $e = \max_{k \in I_{ar} \cup I_{a'r}} \left\| \left(1 - \frac{u_{kr}}{u_{ka}^*} \right) \tilde{\Theta}_r \right\|_F$ and δ is the minimal gap between different precision matrices. For $e \leq \frac{1}{4}\delta$, and

$$\lambda \leq \min \left\{ \frac{1}{\sqrt{p}} \left[\frac{1}{8\tau^2} \delta - \frac{1}{2\tau^2} e \right], \min_{k \in [K], a, r \in [R]} C \frac{\|\Delta_{k, ar}\|_F}{\sqrt{p}\tau^2} \right\},$$

the right hand side of inequality (1) is negative.

References