Overview

Suppose $A \in \mathbb{R}^{m \times n}$ and $\tilde{A} = A + E$. Also, $U^{\top}AV = \begin{bmatrix} D \\ 0 \end{bmatrix}$ and $\tilde{U}^{\top}\tilde{A}\tilde{V} = \begin{bmatrix} \tilde{D} \\ 0 \end{bmatrix}$. There is a main theorem for bounding the distance between D and D:

Theorem 1.1 (Mirsky).

$$\|\tilde{D} - D\| \le \|E\|$$

where the norm can be any unitarily equivalent norm (e.g. $\|\cdot\|_2$ or $\|\cdot\|_F$).

Our purposes

Let the covariance matrix for X be

$$\Sigma = \begin{bmatrix} \Sigma_1 & 0 \\ 0 & \Sigma_2 \end{bmatrix}$$

where

$$\Sigma_1 = \begin{bmatrix} \Sigma_{11} & 0\\ 0 & \Sigma_{12} \end{bmatrix} \in \mathbb{R}^{p_1 \times p_1}$$

and $\Sigma_{1s} \in \mathbb{R}^{p_{1s} \times p_{1s}}$ for s = 1, 2. Let $S = \{1, \dots, p_1\}$ and $S^c = \{1, \dots, p\} \setminus S$. Lastly, we define $\Sigma_{\mathcal{S}} = [\Sigma_1, 0]^{\top}$ to be the first p_1 columns of Σ .

The goal here is to extend the results of the SPCA paper by including the possibility in Σ_1 that we have missed some important features (hence their Σ_1 corresponds to our Σ_{11} . The model for X is then (here I'm writing/thinking about a single latent factor model, I'm presuming that complexifying that to multi-factor will be a matter of notation):

$$X_{ij} = v_i \rho_j + \sigma z_{ij}$$

where v_i, z_{ij} are all mutually independent standard normals and $\rho_j \neq 0$ iff $j \in \mathcal{S}$.

2.1The result

As $F = \mathbb{X}^{\top} \mathbb{X}_1$, then¹

$$F = \left[\sum_{i=1}^{n} (v_i \rho_j + \sigma z_{ij}) (v_i \rho_k + \sigma z_{ik}) \right]_{1 \le j \le p, k \in \mathcal{S}}$$
 (1)

$$= \begin{bmatrix} \sum_{i=1}^{n} (v_i \rho_j + \sigma z_{ij})(v_i \rho_k + \sigma z_{ik}) \\ \sum_{i=1}^{n} (\sigma z_{ij})(v_i \rho_k + \sigma z_{ik}) \end{bmatrix}$$
(2)

$$= \begin{bmatrix} \sum_{i=1}^{n} (v_i \rho_j + \sigma z_{ij})(v_i \rho_k + \sigma z_{ik}) \\ \sum_{i=1}^{n} (\sigma z_{ij})(v_i \rho_k + \sigma z_{ik}) \end{bmatrix}$$

$$= \begin{bmatrix} \sum_{i=1}^{n} (v_i^2 \rho_j \rho_k + \sigma z_{ij} v_i \rho_k + \sigma z_{ik} v_i \rho_j + \sigma^2 z_{ij} z_{ik}) \\ \sum_{i=1}^{n} (\sigma z_{ij} v_i \rho_k + \sigma^2 z_{ij} z_{ik}) \end{bmatrix},$$

$$(3)$$

where the top block has $j \in S$ and the bottom block has $j \in S^c$ (this convention will persist for the rest of the document).

¹We will need to divide through by some function of n.

Using the result from the previous theorem, write $A = \Sigma_{\mathcal{S}}$ and $E = F - \Sigma_{\mathcal{S}}$. Hence, the singular values of $\Sigma_{\mathcal{S}}$ and F will be close if $F - \Sigma_{\mathcal{S}}$ is small.

Note the expectation of F:

$$\mathbb{E}F = \begin{bmatrix} n\rho\rho^{\top} + n\sigma^2 I \\ 0 \end{bmatrix}.$$

Hence,

$$E = [F - \mathbb{E}F] \cdot - \begin{bmatrix} n\rho\rho^{\top} + n\sigma^{2}I \\ 0 \end{bmatrix}.$$

So, up to the $n\sigma^2I$ factor, we have to bound the norm difference between a random matrix and it's expectation.

$$F - \mathbb{E}F - \begin{bmatrix} n\rho\rho^{\top} + n\sigma^{2}I \\ 0 \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^{n} v_{i}^{2}\rho_{j}\rho_{k} - n\rho_{j}\rho_{k} \\ 0 \end{bmatrix} + \begin{bmatrix} \sigma\sum_{i=1}^{n} v_{i}(z_{ij}\rho_{k} + z_{ik}\rho_{j}) \\ \sigma\sum_{i=1}^{n} v_{i}z_{ij}\rho_{k} + \sigma^{2}\sum_{i=1}^{n} z_{ij}z_{ik} \end{bmatrix} + (4)$$

$$+ \begin{bmatrix} \sigma^2 \sum_{i=1}^n z_{ij} z_{ik} - n\sigma^2 I \\ 0 \end{bmatrix}$$
 (5)

$$= (i) + (ii) + (iii). \tag{6}$$

Now,

$$(i) = \left(\sum_{i=1}^{n} v_i^2 - n\right) \begin{bmatrix} \rho \rho^{\mathsf{T}} \\ 0 \end{bmatrix}$$

and

$$(iii) = \sigma^2 \begin{bmatrix} \sum_{i=1}^n Z_{i\mathcal{S}} Z_{i\mathcal{S}}^\top - nI \\ 0 \end{bmatrix},$$

where $Z_{i\mathcal{S}} = [z_{ij}]_{j \in \mathcal{S}}$. Similarly, $Z_{i\mathcal{S}^c} = [z_{ij}]_{j \in \mathcal{S}^c}$. Hence,

$$(ii) = \sigma \sum_{i=1}^{n} v_i \begin{bmatrix} Z_{i\mathcal{S}} \rho^{\top} \\ Z_{i\mathcal{S}^c} \rho^{\top} \end{bmatrix} + \sigma \sum_{i=1}^{n} v_i \begin{bmatrix} \rho Z_{i\mathcal{S}}^{\top} \\ 0 \end{bmatrix} + \sigma^2 \sum_{i=1}^{n} \begin{bmatrix} 0 \\ Z_{i\mathcal{S}} Z_{i\mathcal{S}^c}^{\top} \end{bmatrix}.$$

Now, concentration results can be used to show E is small in Frobenius norm. Presumably, we can find some results about spectral norm as well (which would probably be more useful as it would allow us to say $|d_j - \tilde{d}_j| \leq ||E||_2$ for all j).