

## 1 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  $\tilde{D}$ :

**Theorem 1.1** (Mirsky).

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

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

## 2 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  $\mathcal{S} = \{1, \dots, p_1\}$  and  $\mathcal{S}^c = \{1, \dots, p\} \setminus \mathcal{S}$ . Lastly, we define  $\Sigma_{\mathcal{S}} = [\Sigma_1, 0]^\top$  to be the first  $p_1$  columns of  $\Sigma$ .

The goal here is to extend the results of the SPCA paper by including the possibility in  $\Sigma_1$  that we have missed some important features (hence their  $\Sigma_1$  corresponds to our  $\Sigma_{11}$ . 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  $j \in \mathcal{S}$ .

### 2.1 The result

As  $F = \mathbb{X}^\top \mathbb{X}_1$ , then<sup>1</sup>

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

$$= \left[ \frac{\sum_{i=1}^n (v_i \rho_j + \sigma z_{ij})(v_i \rho_k + \sigma z_{ik})}{\sum_{i=1}^n (\sigma z_{ij})(\sigma z_{ik})} \right] \quad (2)$$

$$= \left[ \frac{\sum_{i=1}^n (v_i^2 \rho_j \rho_k + \sigma z_{ij} v_i \rho_k + \sigma z_{ik} v_i \rho_j + \sigma^2 z_{ij} z_{ik})}{\sum_{i=1}^n (\sigma z_{ij} v_i \rho_k + \sigma^2 z_{ij} z_{ik})} \right], \quad (3)$$

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

<sup>1</sup>We 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,

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

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

$$F - \mathbb{E}F - \begin{bmatrix} n\rho\rho^\top + n\sigma^2 I \\ 0 \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^n v_i^2 \rho_j \rho_k - n\rho_j \rho_k \\ 0 \end{bmatrix} + \begin{bmatrix} \sigma \sum_{i=1}^n v_i (z_{ij} \rho_k + z_{ik} \rho_j) \\ \sigma \sum_{i=1}^n v_i z_{ij} \rho_k + \sigma^2 \sum_{i=1}^n z_{ij} z_{ik} \end{bmatrix} + \quad (4)$$

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

$$= (i) + (ii) + (iii). \quad (6)$$

Now,

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

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  $j$ ).