# Errata of Algorithm 2 in the Paper "A Novel Incremental Principal Component Analysis and Its Application for Face Recognition"

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Algorithm 2—Proposed SVDU-IPCA Algorithm: Given the original data  $X_1 = [x_1, x_2, ..., x_m]$  and the rank-k eigendecomposition of  $\Sigma_1 = P^T P$ , for the newly added data  $[x_{m+1}, x_{m+2}, ..., x_{m+r}]$ , do the following.

- 1) Compute the matrix  $\Sigma_2$  and  $\Sigma_3$  according to (2).
- 2) Obtain the best rank-k approximation of  $P_{l\times m}$ , which is

$$\tilde{P}^{(1)} = \begin{bmatrix} I_k \\ 0 \end{bmatrix}_{I \times k} \Lambda_k V_k^{\mathrm{T}}. \blacktriangleleft$$

- Compute Q<sub>1</sub> according to (3) and Q<sub>3</sub> as the square root of Σ<sub>3</sub> Q<sub>1</sub><sup>T</sup>Q<sub>1</sub> in (5).
- 4) Obtain the QR decomposition  $(I_{(l+r)\times(l+r)} \begin{bmatrix} I_k \\ 0 \end{bmatrix}[I_k \ 0])\begin{bmatrix} Q_1 \\ Q_3 \end{bmatrix} = JK$ , i.e.,

$$\begin{bmatrix} 0_{k\times k} & \\ & I \end{bmatrix}_{(l+r)\times (l+r)} \begin{bmatrix} Q_1 \\ Q_3 \end{bmatrix} = JK.$$

5) Obtain the SVD of the smaller matrix

$$\begin{bmatrix} \Lambda_k & I_k & 0 \end{bmatrix} \begin{bmatrix} Q_1 \\ Q_3 \end{bmatrix} = \hat{U} \hat{\Lambda} \hat{V}^{\mathrm{T}}.$$

6) Obtain the best rank-k approximation of  $\begin{bmatrix} P & Q_1 \\ Q_2 & Q_3 \end{bmatrix}$ , which is

$$\begin{bmatrix} P & Q_1 \\ Q_2 & Q_3 \end{bmatrix} = \begin{pmatrix} \begin{bmatrix} I_k & J \end{bmatrix} \hat{U} \end{pmatrix} \hat{\Lambda} \begin{pmatrix} \begin{bmatrix} V_k & 0 \\ 0 & I \end{bmatrix} \hat{V} \end{pmatrix}^{\mathrm{T}}.$$

7) Obtain the best rank-k approximation or

$$\Sigma = \begin{bmatrix} P & Q_1 \\ Q_2 & Q_3 \end{bmatrix}^{\mathrm{T}} \begin{bmatrix} P & Q_1 \\ Q_2 & Q_3 \end{bmatrix}.$$

#### Error 1:

Here, there should be only  $V_k$  instead of Transpose  $(V_k)$ .

#### Error 2:

We should not take Q<sub>3</sub> as square root of  $\Sigma_3 - Q_1^{\rm T}Q_1$ , instead find Eigen decomposition of  $\Sigma_3 - Q_1^{\rm T}Q_1$  as R\* $\theta$ \*R' and obtain Q<sub>3</sub> = R\*sqrt( $\theta$ )\*R'.

## Important Caution 1:

Here,

$$\Lambda_k = (\operatorname{diag}(\lambda_1, \lambda_2, \ldots, \lambda_k))^{1/2}$$

### Error 3:

Writing as  $[I_kJ]$  will give dimension error as dimensions of  $I_k$  and J would not match. So, append zeros below  $I_k$  to match the dimensions and then compute like  $\begin{bmatrix} I_k & J \\ \Omega & \end{bmatrix}$