

## APPLIED SIGNAL PROCESSING 2011

### EXERCISE 3

#### a) Image compression using PCA

Consider an  $N_r \times N_c$  array  $\mathbf{X}(i, j)$  representing an image. Determine the deviation from the mean  $\mathbf{W}(i, j) = \mathbf{X}(i, j) - m$  where  $m$  is the mean value

$$m = \frac{1}{N_r N_c} \sum_{i,j} \mathbf{X}(i, j)$$

In order to compress the array by PCA, determine the singular value decomposition

$$\mathbf{W} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T = \sum_i^p \sigma_i \mathbf{u}_i \mathbf{v}_i^T$$

where  $p = \min(N_r, N_c)$ .

Plot the singular values  $\sigma_i$ . Introduce next a compressed approximation  $\mathbf{W}_r$  of  $\mathbf{W}$  using only the  $r$  first dominant principal components,

$$\mathbf{W}_r = \sum_i^r \sigma_i^r \mathbf{u}_i \mathbf{v}_i^T = \mathbf{U}_r \mathbf{\Sigma}_r \mathbf{V}_r^T$$

where  $\mathbf{U}_r = [\mathbf{u}_1 \cdots \mathbf{u}_r]$ ,  $\mathbf{\Sigma}_r = \text{diag}(\sigma_1, \dots, \sigma_r)$  and  $\mathbf{V}_r = [\mathbf{v}_1 \cdots \mathbf{v}_r]$ . Next present the compressed image  $\mathbf{X}_r(i, j) = \mathbf{W}_r(i, j) + m$  and compare with the original.

Calculate the amount of data needed to represent the compressed image. Compute the average root-mean-square approximation error

$$e(\mathbf{W} - \mathbf{W}_r) = \left( \sum_{i,j} \frac{1}{N_r N_c} (\mathbf{W}(i, j) - \mathbf{W}_r(i, j))^2 \right)^{1/2}$$

and determine the relative root-mean-square approximation error  $e(\mathbf{W} - \mathbf{W}_r)/e(\mathbf{W})$ .

#### b) Blind source signal separation using ICA

Take  $M (\geq 2)$  signals  $y_i(n)$  of equal lengths and mix them randomly to produce  $M$  mixed signals  $w_i(n)$ . Then use the FastICA algorithm to unmix the signals  $w_i(n)$  to find independent signal components  $s_i(n)$ . Compare the calculated independent components with the original signals  $y_i(n)$ .

The calculations can be performed using the FastICA program package, which can be downloaded from the web.

**Note:** Observe that FastICA determines the independent components  $s_i(n)$  in arbitrary order, the components are normalized to have unit variance (whereas there is no such assumption on the original signals  $y_i(n)$ ), and they are arbitrary with respect to sign ( $-s_i(n)$  and  $s_i(n)$  are equally valid independent components). For comparison,  $y_i(n)$  and  $s_i(n)$  should therefore be scaled to have equal variances, and possible sign differences should be observed.