

Tutorial 9 solution

1. In the original data space the pooled covariance estimate Σ_p has at most a rank $N-g$. Therefore the scatter matrix $S_w = (N-g) \Sigma_p$ similarly has maximum rank of $N-g$, thus the PCA projection can have at most $N-g$ non-zero eigenvectors. Therefore the dimension of P is (at best) $(n \times N - g)$.
2. $Xp = X_{N \times n} \cdot P_{n \times (N-g)}$. Xp therefore has dimension $N \times (N-g)$. The LDA transformation matrix is limited by the number of groups because the rank of S_b is $g - 1$. Therefore the dimension of L is $(N - g \times g - 1)$.
3. $Xf = Xp_{N \times (N-g)} \cdot L_{(N-g) \times (g-1)}$. The final dimension of the data matrix is the dimension of Xf, that is, $(N \times g - 1)$.
4. $Y = Xf_{N \times (g-1)} \cdot L_{(g-1) \times (N-g)}^T \cdot P_{(N-g) \times n}^T$. No, Y is an approximation of X because in the PCA step we have had to discard some principal components to overcome the singularity of S_w . We have used $(N - g)$ principal components rather than $(N - 1)$.