

Part A: Theory and Foundations

Problem 1

Show that the optimization problem

$$\text{maximize } u^T v$$

subject to $u \in \mathcal{R}(X), v \in \mathcal{R}(Y)$ given the vectors u, v which subtend the smallest angles between the subspaces $\mathcal{R}(X)$ and $\mathcal{R}(Y)$.

Problem 2

Using the notation from the CCA lecture, show that

$$a = R_X^\dagger \phi, \quad b = R_Y^\dagger \psi$$

are solutions for the transformation vectors a, b in CCA.

Problem 3

In class we showed that the solution to the angles between subspaces optimization problem led to

$$Q_X^T Q_Y = \Phi \Sigma \Psi^T$$

Show that the elements of Σ are the cosines of the angles between the vector pairs $u^{(i)}, v^{(i)}$ where

$$u^{(i)} = Q_X \phi^{(i)}, \quad v^{(i)} = Q_Y \psi^{(i)}.$$

Problem 4

The GSVD equation can be written

$$\alpha_j X^T X \psi^{(j)} = \beta_j Q Q^T \psi^{(j)}$$

Show that the solutions $\psi^{(j)}$ to the generalized singular value problem are orthogonal in both of the following senses:

$$(\psi^{(i)})^T X^T X \psi^{(j)} = \lambda_i \delta_{ij}$$

and

$$(\psi^{(i)})^T Q^T Q \psi^{(j)} = \lambda_i \delta_{ij}$$

What can you conclude about the maximum noise fraction basis from this?

Problem 5

Show that we can approximate the covariance matrix of the noise by

$$N^T N = \frac{1}{2} dX^T dX$$

Keep careful track of your assumptions.

Part B: Computing

Download the open book/closed book data set from canvas. Take X to be the $P \times 2$ matrix of open book exam results and Y to be the $P \times 3$ matrix of closed book results.

Problem 1.

- Compute the canonical vectors $\{a_1, a_2\}$ and $\{b_1, b_2\}$ for X and Y respectively.
- Plot $\alpha_i = a_1^T x^{(i)}$ and $\beta_i = b_1^T y^{(i)}$. What conclusions can you draw?
- Compute the vectors $u^{(1)}, u^{(2)}$ and $v^{(1)}, v^{(2)}$; we showed this explicitly for $u^{(1)}, v^{(1)}$ in class. Argue why the same equations produce $u^{(2)}, v^{(2)}$.
- Verify that the angle between $u^{(1)}$ and $v^{(1)}$ is $\arccos \sigma_1$.

Problem 2.

Download the data set MNFdata.mat from canvas. This is a 629×4 data matrix X . This problem concerns the implementation of the maximum noise fraction algorithm and the SVD on this data set.

- Compute the maximum noise fraction basis ϕ where $\phi = X\psi$ by solving the following problems

- the eigenvector problem

$$(dX^T dX/2)^{-1} X^T X \psi = \lambda \psi$$

- the generalized eigenvector problem

$$C\psi = D\lambda\psi$$

where $D = dX^T dX/2$ and $C = X^T X$

- the GSVD of $A = dX$ and $B = X$

- Compare your results in part a).
- Compute the SVD of this data matrix and compare the resulting basis vectors with MNF.
- Project the data (columns of X) onto the first 2 MNF basis vectors and the first 2 SVD left singular vectors and compare.