

Homework set 4 — APPM4720/5720, Spring 2016

Problem 1: Suppose that \mathbf{A} is an $m \times n$ matrix of *approximate* rank k , and that we have identified two index sets I_s and J_s such that the matrices

$$(1) \quad \mathbf{C} := \mathbf{A}(:, J_s)$$

$$(2) \quad \mathbf{R} := \mathbf{A}(I_s, :)$$

hold k columns/rows that approximately span the column/row space of \mathbf{A} . You may assume that \mathbf{C} and \mathbf{R} both have rank k (in other words, the index vectors J_s and I_s are not *very* bad). Then

$$\mathbf{A} \approx \mathbf{C}\mathbf{C}^\dagger \mathbf{A}\mathbf{R}^\dagger \mathbf{R},$$

and the optimal choice for the “U” factor in the CUR decomposition is

$$\mathbf{U} := \mathbf{C}^\dagger \mathbf{A}\mathbf{R}^\dagger.$$

Set $\mathbf{X} = \mathbf{C}\mathbf{C}^\dagger$.

(a) Suppose that \mathbf{C} has the SVD

$$\begin{array}{cccc} \mathbf{C} & = & \mathbf{U} & \mathbf{D} & \mathbf{V}^*. \\ m \times k & & m \times k & k \times k & k \times k \end{array}$$

Prove that $\mathbf{X} = \mathbf{U}\mathbf{U}^*$.

(b) Suppose that \mathbf{C} has the QR factorization

$$\begin{array}{ccc} \mathbf{C} & \mathbf{P} & = & \mathbf{Q} & \mathbf{S}. \\ m \times k & k \times k & & m \times k & k \times k \end{array}$$

Prove that $\mathbf{X} = \mathbf{Q}\mathbf{Q}^*$. (Observe that \mathbf{S} is necessarily invertible, since \mathbf{C} has rank k . You can then prove that $\mathbf{C}^\dagger = \mathbf{P}\mathbf{S}^{-1}\mathbf{Q}^*$.)

(c) Prove that \mathbf{X} is the orthogonal projection onto $\text{Col}(\mathbf{C})$.

(d) Suppose that \mathbf{A} has precisely rank k and that \mathbf{C} and \mathbf{R} are both of rank k . Prove that then

$$\mathbf{C}^\dagger \mathbf{A}\mathbf{R}^\dagger = (\mathbf{A}(I_s, J_s))^{-1}.$$

Problem 2: Let \mathbf{A} be an $n \times n$ matrix and suppose (i) that $\mathbf{A}(i, j) > 0$ for every i, j and (ii) that $\sum_{i=1}^n \mathbf{A}(i, j) = 1$ for every j (each column sums to one).

(a) Let \mathbf{p} be a vector of non-negative numbers such that $\sum_{j=1}^n \mathbf{p}(j) = 1$. Set $\mathbf{p}' = \mathbf{A}\mathbf{p}$. Prove that $\sum_{j=1}^n \mathbf{p}'(j) = 1$. (In other words, the matrix \mathbf{A} maps every probability distribution on the set $\{1, 2, \dots, n\}$ to another probability distribution.)

(b) Prove that \mathbf{A} has an eigenvector with corresponding eigenvalue 1.

Problem 3: On the course webpage, you can download a file `testmatrices.mat` that holds three test matrices \mathbf{A} , \mathbf{B} , and \mathbf{C} . Each matrix is of size $m \times 1000$ and contains a thousand samples from a multivariate normal distribution on \mathbb{R}^m . Use PCA to estimate the mean and the co-variance matrices of these distributions. It is sufficient to hand in your numerical answers. (The person doing the reference homework should also hand in code, and a brief description of the solution process.)