

MACHINE LEARNING 2 (MAP 541/DSB). PC5

EXERCISE 1 (SVD AND THE RECONSTRUCTION PROBLEM)

Let X be a $n \times p$ matrix. Show that a solution to the following optimization problem

$$\min_{u \in \mathbb{R}^n, v \in \mathbb{R}^p} J(u, v) \quad \text{with} \quad J(u, v) = \|X - uv^\top\|_F^2$$

with $\|v^*\| = 1$, is given by

$$v^* \quad \text{and} \quad u^* = X v^*,$$

where v^* is the normalized eigen vector associated with λ the largest eigen value of $X^\top X$. Furthermore, $\|u^*\| = \sqrt{\lambda}$.

EXERCISE 2 (ITERATED SVD)

The weighted SVD of order q problem is

$$\min_{U \in \mathbb{R}^{n \times q}, V \in \mathbb{R}^{p \times q}} \|X - UV^\top\|_W^2$$

where

$$\|M\|_W^2 = \sum_{i,j} W_{i,j} M_{i,j}^2. \quad (1)$$

We assume here that W is mask that is $W_{i,j} \in \{0, 1\}$.

Let \odot be the component-wise multiplication, the iterated SVD algorithm is given by

- Start by an initial factorization $U_0 V_0^\top$.
- Iterate T time:
 - Compute the completed matrix $R_t = W \odot X + (1 - W) \odot (U_t V_t^\top)$ where 1 is a matrix filled with 1 everywhere.
 - Use the SVD to obtain a rank q factorization of R_t by $U_{t+1} V_{t+1}^\top$
- Use the last factorization $U_T V_T^\top$.

We let $\mathcal{L}(U, V) = \|X - UV^\top\|_W^2$ and define

$$\mathcal{L}^+(U, V, U', V) = \|W \odot X + (1 - W) \odot U' V'^\top - UV^\top\|^2$$

1. Verify that $\mathcal{L}(U, V) \leq \mathcal{L}^+(U, V, U', V')$ and $\mathcal{L}(U, V) = \mathcal{L}^+(U, V, U, V)$
2. Prove that $\mathcal{L}(U_{t+1}, V_{t+1}) \leq \mathcal{L}(U_t, V_t)$. You may use the fact that the SVD can be used to minimize the problem

$$\min_{U \in \mathbb{R}^{n \times q}, V \in \mathbb{R}^{p \times q}} \|X - UV^\top\|^2.$$

EXERCISE 3 (ALTERNATED LEAST SQUARE)

The penalized weighted SVD of order q problem is

$$\min_{U \in \mathbb{R}^{n \times q}, V \in \mathbb{R}^{p \times q}} \|X - UV^\top\|_W^2 + \lambda \|U\|^2 + \lambda \|V\|^2$$

with $\lambda \geq 0$ where

$$\|M\|_W^2 = \sum_{i,j} W_{i,j} M_{i,j}^2 \quad (2)$$

with $W_{i,j} \geq 0$.

To solve it using the penalized Alternating Least Square(ALS) the following sequence has to be implemented: initialize U and V with the SVD on the full matrix

loop

compute U that $\min_U \|X - UV^\top\|_W^2 + \lambda \|U\|^2$ with a fix V
 compute V that $\min_V \|X - UV^\top\|_W^2 + \lambda \|V\|^2$ with a fix U

1. Show that

$$\min_U \|X - UV^\top\|_W^2 + \lambda \|U\|^2 \text{ with a fix } V$$

boils down to solving n least square problems of the form

$$\min_{u_i \in \mathbb{R}^q} \|x_i - Vu_i\|_{D_i}^2 + \lambda \|u_i\|^2, \quad \text{for } i = 1, \dots, n$$

with $x_i = X_{i,:}$ a \mathbb{R}^p vectors, $D_i = \text{diag}(W_{i,:})$ the diagonal matrix with vector $W_{i,j}, j = 1, \dots, p$ on the diagonal, $\|x\|_D^2 = \sum_j D_j x_j^2$, and whose solution is

$$u_i = (V^\top D_i V + \lambda I)^{-1} V^\top D_i x_i,$$

where I is the identity matrix of dimension q .

2. Show that

$$\min_V \|X - UV^\top\|_W^2 + \lambda \|V\|^2 \text{ with a fix } U$$

boils down at solving p least square problems of the form

$$\min_{v_j \in \mathbb{R}^q} \|x_j - Uv_j\|_{D_j}^2 + \lambda \|v_j\|^2, \quad \text{for } j = 1, \dots, p$$

with $x_j = X_{:,j}$ a \mathbb{R}^n vectors and $D_j = \text{diag}(W_{:,j})$ the diagonal matrix with vector $W(i,j), i = 1, \dots, n$ and whose solution is

$$v_j = (U^\top D_j U + \lambda I)^{-1} U^\top D_j x_j,$$

where I is the identity matrix of dimension q .