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HW1 Problem 2

Let SVD of  $A = USV^T$ ,

Where  $U$  and  $V$  are orthonormal matrices and  $S$  is diagonal.

$$A^T A = VSU^T USV^T = VS^2 V^T \text{ -- (1)}$$

Let the Eigen decomposition of  $A^T A$  be represented as:  $A^T A = E \Sigma E^{-1}$

**Conclusion:** the orthonormal matrix  $V$  from SVD of  $A$  is the eigen vector matrix of Eigen decomposition of  $A^T A$  ( $V = E$ ), and the square of the diagonal matrix from SVD of  $A$  is the Eigen value matrix of  $A^T A$  ( $\Sigma = S^2$ ). The proof is shown as follows:

**Proof:**

Now let's prove that (1) is also an Eigen decomposition of the square matrix  $A^T A$ .

In equation (1), multiply on both sides on the right by  $V$ :

$$A^T A V = VS^2 V^T V = VS^2 \text{ -- (2) (apply the property of orthonormal matrix that } V^T V = 1)$$

To continue with eqn (2):

$A^T A [v_1 \ v_2 \ v_3 \ \dots \ v_n] = [v_1 \ v_2 \ v_3 \ \dots \ v_n] S^2 = B$ , Let  $v_1, v_2, v_n$  be vectors of each column of  $V$ , Let Matrix  $S^2$ 's diagonal be:  $s_1^2, s_2^2, \dots, s_n^2$ . On the left hand side, the matrix multiplication can be viewed as  $A^T A$  dot with  $v_1, v_2, \dots, v_n$ . By definition, the dot product of these result in each column of  $B$ , call them  $b_1, b_2, \dots, b_n$ . In equations:  $A^T A \cdot v_i = b_i, i \in (1, n)$  -- (3)

Now consider the left hand side, the matrix multiplication can be viewed as the outer product of  $v_1, v_2, \dots, v_n$  and each row of matrix  $S$ , and the sum of the matrices created by the outer products is exactly  $B$ . In equations:

$$v_1 \cdot s_1^2 + v_2 \cdot s_2^2 + \dots + v_n \cdot s_n^2 \text{ --- (4)}$$

Notice that  $S$  is a diagonal matrix, which means that the  $i$ th term in (4) results in a matrix with all zeros except for the  $i$ th column. Thus, the net effect of each term in (4) is to produce a column in matrix  $B$ .

Therefore, from (3) and (4), we have  $b_i = v_i \cdot s_i^2 \rightarrow$

$$A^T A \cdot v_i = v_i \cdot s_i^2 = s_i^2 \cdot v_i, \text{ -- (6) as } s_i^2 \text{ is simply a scalar.}$$

In (6), make the observation that that  $v_i$  is the eigen vector of matrix  $A^T A$ , and  $s_i^2$  is the associating eigen value. The conclusion holds for any  $i$  in range  $(1, n)$ , thus, matrix  $V$  is indeed an eigen vectors matrix, and matrix  $S$  is indeed an eigen value matrix. Therefore, let's restate the conclusion: (1) is an eigen decomposition of  $A^T A$  with  $V = E$  and  $\Sigma = S^2$ .