

# COL726: Minor-2 Take Home

Haroun Habeeb

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We know that  $A = X^t X$ . Let the  $svd(X) = U\Sigma V^*$  where  $U, V$  and unitary matrices and  $\Sigma$  is a diagonal matrix. Hence,

$$A = V^{*t} \Sigma^t U^t U \Sigma V^* = V^{*t} \Sigma^t I \Sigma V^* = V^{*t} \Sigma^t \Sigma V^* = V^{*t} \Sigma^2 V^*$$

Where  $\Sigma^2$  is a diagonal matrix with only positive entries (since  $\Sigma$  is real.) Notice that  $V^*$  is unitary, hence  $V^* = V^{*t^{-1}}$ . Hence,  $A = Q \Sigma^2 Q^{-1}$  with  $Q = V^{*t}$ .

Then,  $A = (\Sigma V^*)^t (\Sigma V^*)$ , which is of the form  $X^t X$ .

Hence, we can simply compute the Eigenvalue decomposition of  $A$ , and get both  $V^*$  and  $\Sigma = \text{sqr}t(\Sigma^2)$ .

(a) I'm not sure what  $n$ -dimensional vector means here... But I suppose that we could look at the entries of  $\Sigma$ . If there are  $n$  non-zero diagonal entries, then we have  $n$  vectors.

(b) If  $(\Sigma V^*)$  has  $\text{rank} = 2$ , then  $X$  could be transformed into a  $2 \times n$  matrix.

(c) Read them from  $\Sigma V^*$ .

(d) Let's say our approximation for  $X$  is  $\hat{X}$ . We know  $A = X^t X$  and can estimate  $\hat{A} = \hat{X}^t \hat{X}$ . The residual error would be  $X - \hat{X}$ . However, we could also make  $\text{norm}(A - \hat{A})$  the residual error.

Since  $V^*$  is already orthonormal, really all we must do is to consider only the first two singular values of  $\Sigma$ .

## Extra Light:

1.

2.

Positive definite condition is that  $\sum_{i,j} z_i z_j A_{ij} > 0$

which means that  $\sum_{i,j} z_i z_j (D_{ij}^2 - S_i - S_j + T) < 0$

3.

4.

5.