# CS391 Assignment 2

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#### Exercise 1

a) If the data comes from a Gaussian Distribution we can use the Z-scores to find outliers. Anything that falls 3 standard deviation away from the mean is an outlier.

28, 21.111968065340623, 11.5578708791414, -1.6009047634840101, 21.700777435927332, 11.395997104576299, 11.35797838900 0063, 20.04886141800474, -10.9700334130475, -1.6772466232641587, -10.072341609825347, -9.586953797092942] There are 17 outliers

b) I don't know

#### Exercise 2

a) For a matrix M with vectors u and v, it follows that:  $Mu \cdot v = u \cdot M^Tv$ Now if M is symmetric and the vectors are eigenvectors of M with corresponding distinct eigenvalues  $\lambda_1$  and  $\lambda_2$ , such that  $Mu = \lambda_1 u$  and  $Mv = \lambda_2 v$ then we obtain the following:

 $\lambda_1(u \cdot v) = (\lambda_1 u \cdot v) = (Mu \cdot v) = (u \cdot M^T v) = (u \cdot Mv) = (u \cdot \lambda_2 v) = \lambda_2(u \cdot v)$ Thus,  $(\lambda_1 - \lambda_2)(u \cdot v) = 0$ . Since  $(\lambda_1 - \lambda_2) \neq 0$ , it must mean that  $(u \cdot v) = 0$ . Making the eigenvectors orthogonal.

b) If matrix A is symmetric it follows that  $A = A^T$ , we also know that  $(AB)^T = B^T A^T$  for any matrix A and B. Therefore:

 $(AA^T)^T = (A^T)^T A^T = AA^T \implies AA^T$  is symmetric. Since by definition a symmetric matrix is equal to its transpose.

c) To begin we know that a symmetric matrix is always diagonizable. Let  $A = U\Sigma V^T$  be the SVD. Then  $AA^T = U\Sigma^2 U^T$  and  $A^TA = V\Sigma^2 V^T$ . Since in part (a) we proved that U and V are orthogonal. The inverse of an orthogonal matrix is equal to it's inverse.  $A^TA = I$ , we have  $(A^TA)A^{-1} = IA^{-1} = A^{-1}$ . Matrix multiplication is associative  $(A^T A)A^{-1} = A^T (AA^{-1}) \implies A^T = A^{-1}$ . The same would apply for U and V.

d) 
$$A = \begin{bmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{bmatrix}$$
  
First we need to find  $AA^T$ 

$$AA^{T} = \begin{bmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{bmatrix} \begin{bmatrix} 3 & 2 \\ 2 & 3 \\ 2 & -2 \end{bmatrix} = \begin{bmatrix} 17 & 8 \\ 8 & 17 \end{bmatrix}$$

$$det(AA^T - \lambda I) = \lambda^2 - 34\lambda + 225 = (\lambda - 25)(\lambda - 9)$$
. The eigenvalues of  $A^TA$  are 25, 9, 0.

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$$\Sigma = \begin{bmatrix} 5 & 0 & 0 \\ 0 & 3 & 0 \end{bmatrix}$$

For 
$$\lambda = 25$$
,  $A^T A - 25I = \begin{bmatrix} -12 & 12 & 2 \\ 12 & -12 & -2 \\ 2 & -2 & -17 \end{bmatrix}$  which row-reduces to  $\begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \Longrightarrow$ 

$$v_1 = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{bmatrix}$$

For 
$$\lambda = 9$$
,  $A^{T}A - 9I = \begin{bmatrix} 4 & 12 & 2 \\ 12 & 4 & -2 \\ 2 & -2 & -1 \end{bmatrix}$  which row-reduces to  $\begin{bmatrix} 1 & 0 & \frac{-1}{4} \\ 0 & 1 & \frac{1}{4} \\ 0 & 0 & 0 \end{bmatrix} \Longrightarrow$ 

$$v_2 = \begin{bmatrix} \frac{1}{\sqrt{18}} \\ \frac{-1}{\sqrt{18}} \\ \frac{4}{\sqrt{18}} \end{bmatrix}$$

For the last vector we need 
$$v_2^T v_3 = 0 \implies v_3 = \begin{bmatrix} \frac{2}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{-1}{\sqrt{3}} \end{bmatrix}$$

To get you we just need find the vectors using  $u_i = \frac{1}{\sqrt{\lambda}} A v_i \implies U = \frac{1}{\sqrt{\lambda}} A v_i$ 

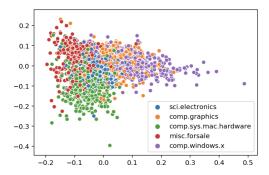
$$\begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \end{bmatrix}$$

The SVD of A = 
$$U\Sigma V^T = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} 5 & 0 & 0 \\ 0 & 3 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{18}} & \frac{-1}{\sqrt{18}} & \frac{4}{\sqrt{18}} \\ \frac{2}{\sqrt{3}} & \frac{-2}{\sqrt{3}} & \frac{-1}{\sqrt{3}} \end{bmatrix}$$

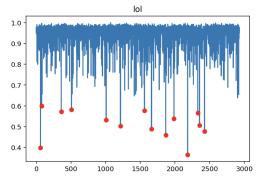
### Exercise 3

I limited to my graph to a rank of 25. In class we discussed that when looking at the singular value graph we want to choose a value of K at around the knee. I approximate my K to be 25 when look at the singular value graph.

I picked the 0.6 as the limit for outliers is because most data points fall between 1 and 1. Any values below 0.6 seems to be an extremity. With this there are 14 outliers.



 $[ \quad 61 \quad \ 77 \quad 360 \quad 508 \ 1008 \ 1214 \ 1560 \ 1662 \ 1867 \ 1986 \ 2183 \ 2329 \ 2354 \ 2426]$ 



The graph above shows where a lot of the points are clustered. Anything below 0.6 is declared an outlier.

The graph below show a scatter plot of the data points with an estimated best fine line. Points that are really far away from the green line are considered outliers.

