

Problem Set 9, Nov 14, 2019 (Theory Questions SVD).

Goals. The goal of this exercise is to

- familiarize yourself with the theory related to SVD.
- have time to discuss Project 2 with the assistants and teammates.

1 Theory Questions

Problem 1 (How to compute U and S efficiently):

In class, we saw that solving the eigenvector/value problem for the matrix $\mathbf{X}\mathbf{X}^\top$ gives us a way to compute U and S . But in some instances $D \gg N$. In those cases, is there a way to accomplish this computation more efficiently?

Problem 2 (Positive semi-definite):

Show that if \mathbf{X} is a $N \times N$ symmetric matrix then the SVD has the form $\mathbf{U}\mathbf{S}\mathbf{V}^\top$, where \mathbf{U} is a $N \times N$ unitary matrix and \mathbf{S} is a $N \times N$ diagonal matrix with non-necessarily positive entries. Show that if \mathbf{X} is positive semi-definite, then all entries of \mathbf{S} are non-negative.

P1:

$$\mathbf{X} = \mathbf{U}\mathbf{S}\mathbf{V}^\top$$

$$\mathbf{X}\mathbf{X}^\top = \mathbf{U}\mathbf{S}\mathbf{V}^\top \mathbf{V}\mathbf{S}^\top\mathbf{U}^\top = \mathbf{U}\mathbf{S}\mathbf{S}^\top\mathbf{U}^\top = \mathbf{U}\mathbf{S}^2\mathbf{U}^\top$$

$$\mathbf{X} : D \begin{array}{|c|c|c|}\hline & & \\ \hline & N & \\ \hline & & D \end{array} = D \begin{array}{|c|c|c|}\hline & & \\ \hline & & \\ \hline & & D \end{array}$$

$$\mathbf{X}^\top\mathbf{X} = \mathbf{V}\mathbf{S}^\top\mathbf{U}^\top \underbrace{\mathbf{U}\mathbf{S}\mathbf{V}^\top}_{\Sigma} \mathbf{V}\mathbf{S}^\top\mathbf{U}^\top = \mathbf{V}\mathbf{S}^\top\mathbf{S}\mathbf{V}^\top$$

let v_i denote the columns of V . Then:

$$X^T X v_j = \lambda_j^2 v_j$$

The j -th column of V is an eigenvector of $X^T X$ with eigen value λ_j^2 .

\Rightarrow Solving the eigenvalue/vector problem for the matrix $X^T X$ gives us a way to compute V and S .

$$X^T X (X v_j) = \lambda_j^2 X v_j$$

$$X^T X u_j = \lambda_j^2 u_j$$

\Rightarrow we can compute the desired eigenvectors u_j from v_j without solving the eigenvector/value problem for $X^T X$ (DAD).

P2:-

$$X = (N \times N)$$

$$X = U S V^T \Rightarrow \begin{aligned} U &= N \times N \\ V &= N \times N \\ S &= N \times N. \end{aligned}$$

* The columns of U are the eigenvectors of $A = X X^T$
* The columns of V are the eigenvectors of $B = X^T X$

Since X is $N \times N$ symmetric $XX^T = X^TX$

\Rightarrow the eigenspace associated to each distinct eigenvalue of A is the same eigenspace associated to the same distinct eigenvalue of B .

$$\underline{U = V}.$$

$$X = USV^T \Rightarrow U^T X V = UU^T S V^T V = S$$

