## LARSON—MATH 511—CLASSROOM WORKSHEET 11 Gilbert Strang Lecture 7.

## More on Strang's Lectures

- 1. We showed that  $A^TA$  is symmetric. It is a **key fact** that if A has linearly independent columns then  $A^TA$  is positive definite (and thus had positive eigenvalues). Why?
- 2. We showed that  $A^TA$  is symmetric. It is a **key fact** that  $A^TA$  is positive semi-definite (and thus has non-negative eigenvalues). Why?

**SVD Algorithm** (general case, A has rank r)

Suppose A is any  $m \times n$  matrix.

- Find  $A^T A$  (its  $n \times n$ )
  This matrix has rank r, is symmetric and is positive semi-definite.
- Find the positive eigenvalues and corresponding (unit) eigenvectors  $\lambda$ ,  $\hat{v}$  (So  $\lambda_1 \geq \lambda_2 \geq \ldots \geq \lambda_r > 0$ ). The remaining eigenvalues  $\lambda_{r+1}, \ldots, \lambda_n$  are 0. The remaining (unit) eigenvectors  $\hat{v}_{r+1}, \ldots, \hat{v}_n$  are a basis for the nullspace of  $A^T A$ . Let  $\Lambda$  be the diagonal matrix with the  $\lambda$ s on the diagonal.

Vectors corresponding to different eigenvalues are orthogonal. For each eigenspace with dimension greater than 1, Gram-Schmidt can be used to find an orthogonal basis.

- Let  $V = [\hat{v}_1 \dots \hat{v}_r \dots \hat{v}_n]$ . V is orthogonal.
- Let  $\sigma_i = \sqrt{\lambda_i}$  and let  $\Sigma$  be the  $m \times n$  matrix with diagonal entries  $\sigma_1, \ldots, \sigma_r$
- Find  $AA^T$  (its  $m \times m$ )
  This matrix has rank r, is symmetric and is positive semi-definite.
- Check that  $A\hat{v}_i$  is an eigenvector for  $AA^T$  with corresponding eigenvalue  $\lambda_i$ . This means that these are all the non-zero eigenvalues for  $AA^T$ . Let  $u_i = \frac{1}{\sigma_i}A\hat{v}_i$  for  $i = 1 \dots r$ . Check that these  $\hat{u}_i$  are unit.
- The remaining eigenvalues of  $AA^T$ ,  $\lambda_{r+1}, \ldots, \lambda_m$  are 0. The remaining (unit) eigenvectors  $\hat{u}_{r+1}, \ldots, \hat{u}_m$  are a basis for the nullspace of  $AA^T$ .
- Let  $U = [\hat{u}_1 \dots \hat{u}_r \dots \hat{u}_m]$ . U is orthogonal.
- $AV = U\Sigma$  and  $A = U\Sigma V^T$ .

3. Low Rank Approximation. Why does  $A = \sigma_1 \hat{u}_1 \hat{v}_1^T + \ldots + \sigma_r \hat{u}_r \hat{v}_r^T$ ?

## Sage/CoCalc

- (a) Start the Chrome browser.
- (b) Go to http://cocalc.com
- (c) Login (likely using your VCU email address).
- (d) You should see an existing Project for our class. Click on that.
- (e) Click "New", then "Sage Worksheet", then call it c11.

We will find the singular values  $\sigma_1 \geq \dots \sigma_r > 0$  of a matrix A with rank r in three ways: (1) via the eigenvalues of  $A^T A$ , (2) via the eigenvalues of  $AA^T$ , and (3) using Sage's built-in SVD method.

- 4. Input  $A = \begin{bmatrix} 3 & 4 \\ 0 & 5 \end{bmatrix}$  (remember to inform Sage you mean for the entries to be interpreted as elements of a Real Double Field (RDF).
- 5. Find the rank of A.
- 6. Find  $A^T A$ , its eigenvalues, and the singular values of A.
- 7. Find  $AA^T$ , its eigenvalues, and the singular values of A.
- 8. Try: A.SVD(). See the help in order to interpret the output.
- 9. Check that  $A = \sigma_1 \hat{u}_1 \hat{v}_1^T + \ldots + \sigma_r \hat{u}_r \hat{v}_r^T$ .
- 10. Repeat for the matrix  $A = \begin{bmatrix} 3 & 4 & 0 \\ 0 & 5 & 1 \end{bmatrix}$ .

## Getting your classwork recorded

When you are done, before you leave class...

- 1. Click the "Make pdf" (Adobe symbol) icon and make a pdf of this worksheet. (If CoCalc hangs, click the printer icon, then "Open", then print or make a pdf using your browser).
- 2. Send me an email with an informative header like "Math 511—c11 worksheet attached" (so that it will be properly recorded).
- 3. Remember to attach today's classroom worksheet!