

Last time we assumed A was invertible so the λ 's are all non-zero. Then we defined

$$z_i = \frac{y_i}{\sqrt{\lambda_i}} \quad \text{for } i=1, \dots, m$$

and obtained an orthonormal basis.

If A is not invertible then some of the λ 's are zero. which means it's not possible to define m vectors by renormalizing the y 's.

Group the indices i where z_i can be defined together and the indices where $\lambda_i = 0$ in the other group.

Now, since the indices themselves were arbitrary, I can make the assumption

$$\lambda_i \neq 0 \quad \text{for } i=1, \dots, m$$

$$\lambda_i = 0 \quad \text{for } i=m+1, \dots, n$$

Traditionally we order the eigenvalues of B so that $\lambda_n \geq \lambda_{n-1} \geq \dots \geq \lambda_m \geq \lambda_{m-1} \geq \dots \geq \lambda_1$

Recall

the $\sqrt{\lambda_i}$ are called singular values of the matrix A .

$$\Sigma = \begin{bmatrix} \sqrt{\lambda_1} & & & 0 \\ & \sqrt{\lambda_2} & & \\ & & \ddots & \\ & & & \sqrt{\lambda_m} \\ & & & & 0 \\ & & & & & \ddots \\ & & & & & & 0 \end{bmatrix} = \begin{bmatrix} \sigma_1 & & & 0 \\ & \sigma_2 & & \\ & & \ddots & \\ & & & \sigma_m \\ & & & & 0 \\ & & & & & \ddots \\ & & & & & & 0 \end{bmatrix}$$

The square roots of the eigenvalues of $B = A^T A$ are by definition the singular values of A .

Standard notation $\sigma_i = \sqrt{\lambda_i}$
Singular values...

When A is not invertible some of the σ 's are zero.

$$\Sigma = \begin{bmatrix} \sigma_1 & & & \\ & \sigma_2 & & \\ & & \ddots & \\ & & & \sigma_m \\ & & & & 0 \\ & & & & & \ddots \\ & & & & & & 0 \end{bmatrix}$$

$n-m$ zeros along the diagonal

also recall the V matrix from last time

$$V = \begin{bmatrix} | & | & & | \\ z_1 & z_2 & \dots & z_m \\ | & | & & | \end{bmatrix}$$

ends at m since some z 's are missing.
is not square

To make it square need to add some vectors

$$V = \left[\begin{array}{c|c|c|c} z_1 & z_2 & \dots & z_m \\ \hline & & & \text{missing vectors} \end{array} \right]$$

Since V is supposed to be an orthogonal matrix the missing vectors need to be unit length and orthogonal to all the other vectors so in the end V has n orthonormal vectors as columns.

Use Gram-Schmidt process to find the missing vectors.

$$V = \left[\begin{array}{c|c|c|c|c|c} z_1 & z_2 & \dots & z_m & z_{m+1} & \dots & z_n \\ \hline & & & & & & \end{array} \right]$$

just made these up so V is orthogonal.

Only thing left is to check that $A = V \Sigma U^T$ still...

Need to check AU is the same as $V \Sigma$.

$$AU = V \Sigma$$

Now...

$$\begin{aligned} AU &= A \left[\begin{array}{c|c|c|c} x_1 & x_2 & \dots & x_n \\ \hline & & & \end{array} \right] = \left[\begin{array}{c|c|c|c} Ax_1 & Ax_2 & \dots & Ax_n \\ \hline & & & \end{array} \right] = \left[\begin{array}{c|c|c|c} y_1 & y_2 & \dots & y_n \\ \hline & & & \end{array} \right] \\ \text{also} \quad & \approx \left[\begin{array}{c|c|c|c} y_1 & \dots & y_m & \underbrace{0 \quad \dots \quad 0}_{n-m \text{ zero columns}} \\ \hline & & & \end{array} \right] \end{aligned}$$

$$V\Sigma = \left[\begin{array}{c|c|c|c|c} |v_1\rangle & \dots & |v_m\rangle & |v_{m+1}\rangle & \dots & |v_n\rangle \end{array} \right] \left[\begin{array}{c} \sigma_1 \\ \vdots \\ \sigma_m \\ 0 \\ \vdots \\ 0 \end{array} \right]$$

$$= \left[\begin{array}{c|c|c|c|c} \sqrt{\lambda_1}v_1 & \dots & \sqrt{\lambda_m}v_m & 0 & \dots & 0 \end{array} \right] \approx \left[\begin{array}{c|c|c|c|c} y_1 & \dots & y_m & 0 & \dots & 0 \end{array} \right]$$

Since $v_i = \frac{y_i}{\sqrt{\lambda_i}}$

These are the same so we are done...

Now work an example from the book that we didn't have time for in class

$$A = \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix}$$

First find the eigenvectors and eigenvalues of $B = A^T A$.

$$A^T A = \begin{bmatrix} 2 & 2 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ 2 & 2 \end{bmatrix} = \begin{bmatrix} 8 & 2 \\ 2 & 5 \end{bmatrix}$$

$$\det(B - \lambda I) = \det \begin{bmatrix} 8-\lambda & 2 \\ 2 & 5-\lambda \end{bmatrix} = (8-\lambda)(5-\lambda) - 4 = \lambda^2 - 13\lambda + 36 = (\lambda-9)(\lambda-4)$$

so the eigenvalues are $\lambda_1 = 9$ and $\lambda_2 = 4$

Note the eigenvalues are non-negative and we chose them in order so $\lambda_1 \geq \lambda_2$.

The corresponding eigenvectors:

$\lambda = 9$
 $x \in \text{Nul} \begin{bmatrix} -1 & 2 \\ 2 & -4 \end{bmatrix}$

so

$-x_1 + 2x_2 = 0$ $x_1 = 2x_2$ $x_2 = \text{free}$
 and $x = \begin{bmatrix} 2 \\ 1 \end{bmatrix} x_2$ represent all the eigenvectors

To make a vector of unit length normalize. Thus,

$$\lambda_1 = 9 \quad \text{and} \quad x_1 = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$\underline{\underline{A=H}}$ $x \in \text{Nul} \begin{bmatrix} 4 & 2 \\ 2 & 1 \end{bmatrix} \Leftrightarrow 2x_1 + x_2 = 0 \quad x_2 = -2x_1 \quad x_1 = \text{free}$
 and $x = \begin{bmatrix} 1 \\ -2 \end{bmatrix} x_1$ represent all the eigenvectors

To make a vector of unit length normalize. Thus,

$$\lambda_2 = 4 \quad \text{and} \quad x_2 = \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

In summary, B has an orthonormal basis of eigenvectors given by

$$\{x_1, x_2\} = \left\{ \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ -2 \end{bmatrix} \right\}$$

so the matrix U is

$$U = \begin{bmatrix} | & | \\ x_1 & x_2 \\ | & | \end{bmatrix} = \begin{bmatrix} 2/\sqrt{5} & 1/\sqrt{5} \\ 1/\sqrt{5} & -2/\sqrt{5} \end{bmatrix} \quad \rightarrow \quad U^{-1} = U^T = \begin{bmatrix} 2/\sqrt{5} & 1/\sqrt{5} \\ 1/\sqrt{5} & -2/\sqrt{5} \end{bmatrix}$$

By chance it turned out that U was symmetric
 \downarrow

and

$$\Sigma = \begin{bmatrix} \sqrt{\lambda_1} & 0 \\ 0 & \sqrt{\lambda_2} \end{bmatrix} = \begin{bmatrix} \sqrt{9} & 0 \\ 0 & \sqrt{4} \end{bmatrix} = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix}$$

What's left is to find the matrix V . Recall $y_i = Ax_i$

$$y_1 = Ax_1 = \begin{bmatrix} 2 & -1 \\ 2 & 2 \end{bmatrix} \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \frac{1}{\sqrt{5}} \begin{bmatrix} 3 \\ 6 \end{bmatrix} = \frac{3}{\sqrt{5}} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

and

$$y_2 = Ax_2 = \begin{bmatrix} 2 & -1 \\ 2 & 2 \end{bmatrix} \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ -2 \end{bmatrix} = \frac{1}{\sqrt{5}} \begin{bmatrix} 4 \\ -2 \end{bmatrix} = \frac{2}{\sqrt{5}} \begin{bmatrix} 2 \\ -1 \end{bmatrix}$$

Consequently, $v_1 = \frac{y_1}{\sqrt{\lambda_1}} = \frac{\frac{3}{\sqrt{5}} \begin{bmatrix} 1 \\ 2 \end{bmatrix}}{3} = \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

and $v_2 = \frac{y_2}{\sqrt{\lambda_2}} = \frac{\frac{2}{\sqrt{5}} \begin{bmatrix} 2 \\ -1 \end{bmatrix}}{2} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ -1 \end{bmatrix}$

These should be unit vectors.
 If they are not unit vectors then something went wrong.

It follows that

$$V = \begin{bmatrix} | & | \\ v_1 & v_2 \\ | & | \end{bmatrix} = \begin{bmatrix} 1/\sqrt{5} & 2/\sqrt{5} \\ 2/\sqrt{5} & -1/\sqrt{5} \end{bmatrix}$$

and the singular value decomposition of A is

$$\begin{bmatrix} 2 & -1 \\ 2 & 2 \end{bmatrix} = A = V \Sigma U^T = \begin{bmatrix} 1/\sqrt{5} & 2/\sqrt{5} \\ 2/\sqrt{5} & -1/\sqrt{5} \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 2/\sqrt{5} & 1/\sqrt{5} \\ 1/\sqrt{5} & -2/\sqrt{5} \end{bmatrix}$$

orthogonal matrix *diagonal matrix* *orthogonal matrix*

We'll go over the above computation on Monday, do one more example and discuss the review sheet for the final exam. Please check for the review sheet over the weekend.