

# EE16B: DESIGNING INFORMATION DEVICES AND SYSTEMS II

## LECTURE NOTES

### SINGULAR VALUE DECOMPOSITION (SVD)

Let  $A$  be an  $m \times n$  matrix,

$$A = U[\Lambda|0]V^T$$

where  $V$  and  $U$  have orthonormal columns and  $\Lambda$  is a diagonal matrix.

Let,

$$\begin{aligned} S &= AA^T \\ T &= A^T A \\ S &= U\Lambda_S U^T \end{aligned}$$

( $\Lambda_S$  contains eigenvalues of  $S$ )

$$T = V\Lambda_T V^T$$

( $\Lambda_T$  contains eigenvalues of  $T$ )

$$\Lambda_T = \begin{bmatrix} \lambda_{T,i} & & 0 \\ & \ddots & 0 \\ 0 & 0 & \ddots \end{bmatrix} (n \times n)$$

We've proved that symmetric matrices have orthogonal eigenspaces.

On the next homework, you will prove that symmetric matrices have a full complement of eigenvectors.

You now know enough to prove that if there exists a real symmetric matrix, all of its eigenvalues must be real.

Start with  $T$ . Consider

$$\vec{u}_i = \frac{A\vec{v}_i}{\sqrt{\lambda_{T,i}}} \text{ if } \lambda_{T,i} > 0$$

We can show that  $\sqrt{\lambda_{T,i}}$  must be real, but how do we prove they are non-negative?

Consider

$$\vec{v}_i^T T \vec{v}_i = \lambda_{T,i}$$

$$\vec{v}_i^T T \vec{v}_i = (\vec{v}_i^T A^T)(A\vec{v}_i) = \|A\vec{v}_i\|^2 = \lambda_{T,i}$$

All  $\lambda_T$ 's must be real and non-negative, since the norm is always greater than or equal to zero.

Assert

$$\tilde{U} \left[ \begin{array}{c|c} \sqrt{\lambda_{T,i}} & 0 \\ \cdot & \cdot \\ \cdot & \cdot \\ \hline & 0 \end{array} \right] V^T = A$$

$$A\vec{v}_i = \tilde{U}[\Lambda|0] \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \leftarrow i^{\text{th}} \text{ position}$$

$$\tilde{U} \begin{bmatrix} 0 \\ 0 \\ \sqrt{\lambda_{T,i}} \\ 0 \\ 0 \end{bmatrix} = A\vec{v}_i = \vec{u}_i \cdot \sqrt{\lambda_{T,i}}$$

Check orthonormality of  $[\vec{u}_i]$ : (using the substitutions  $T = A^T A$  and  $T = V \Lambda_T V^T$ )

$$\vec{u}_i^T \vec{u}_j = \frac{\vec{v}_i^T A^T}{\sqrt{\lambda_{T,i}}} \cdot \frac{A\vec{v}_j}{\sqrt{\lambda_{T,j}}} = \frac{\vec{v}_i^T A^T A \vec{v}_j}{\sqrt{\lambda_{T,i} \lambda_{T,j}}} = \frac{\vec{v}_i^T T \vec{v}_j}{\sqrt{\lambda_{T,i} \lambda_{T,j}}} = \frac{\vec{v}_i^T V \Lambda_T V^T \vec{v}_j}{\sqrt{\lambda_{T,i} \cdot \lambda_{T,j}}} = \begin{cases} 0 & i \neq j \\ 1 & i = j \end{cases}$$

$$S = AA^T = \tilde{U}[\Lambda|0]V^T V \begin{bmatrix} \Lambda \\ 0 \end{bmatrix} \tilde{U}^T = \tilde{U} \Lambda^2 \tilde{U}^T$$

$\rightarrow AA^T$  is diagonalized by  $\vec{u}$ , so  $\vec{u}$  must be  $U$ .