

and since  $f_j = 0$  for  $j = r + 1, \dots, n$ ,

$$\langle v_i, f_j \rangle = 0 \quad 1 \leq i \leq n, \quad r + 1 \leq j \leq n.$$

If  $V$  is the matrix whose columns are  $v_1, \dots, v_n$ , then  $V$  is orthogonal and the above equations prove that

$$V^\top AU = D,$$

which yields  $A = VDU^\top$ , as required.

The equation  $A = VDU^\top$  implies that

$$A^\top A = UD^2U^\top, \quad AA^\top = VD^2V^\top,$$

which shows that  $A^\top A$  and  $AA^\top$  have the same eigenvalues, that the columns of  $U$  are eigenvectors of  $A^\top A$ , and that the columns of  $V$  are eigenvectors of  $AA^\top$ .  $\square$

**Example 22.4.** Here is a simple example of how to use the proof of Theorem 22.5 to obtain an SVD decomposition. Let  $A = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}$ . Then  $A^\top = \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}$ ,  $A^\top A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$ , and  $AA^\top = \begin{pmatrix} 2 & 0 \\ 0 & 0 \end{pmatrix}$ . A simple calculation shows that the eigenvalues of  $A^\top A$  are 2 and 0, and for the eigenvalue 2, a unit eigenvector is  $\begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix}$ , while a unit eigenvector for the eigenvalue 0 is  $\begin{pmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{pmatrix}$ . Observe that the singular values are  $\sigma_1 = \sqrt{2}$  and  $\sigma_2 = 0$ . Furthermore,  $U = \begin{pmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{pmatrix} = U^\top$ . To determine  $V$ , the proof of Theorem 22.5 tells us to first calculate

$$AU = \begin{pmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{pmatrix},$$

and then set

$$v_1 = (1/\sqrt{2}) \begin{pmatrix} \sqrt{2} \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}.$$

Once  $v_1$  is determined, since  $\sigma_2 = 0$ , we have the freedom to choose  $v_2$  such that  $(v_1, v_2)$  forms an orthonormal basis for  $\mathbb{R}^2$ . Naturally, we chose  $v_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$  and set  $V = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ . The columns of  $V$  are unit eigenvectors of  $AA^\top$ , but finding  $V$  by computing unit eigenvectors of  $AA^\top$  does not guarantee that these vectors are consistent with  $U$  so that  $A = V\Sigma U^\top$ . Thus one has to use  $AU$  instead. We leave it to the reader to check that

$$A = V \begin{pmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{pmatrix} U^\top.$$

Theorem 22.5 suggests the following definition.