If A is a complex  $n \times n$  matrix, the eigenvalues  $\lambda_1, \ldots, \lambda_n$  and the singular values  $\sigma_1 \geq \sigma_2 \geq \cdots \geq \sigma_n$  of A are not unrelated, since

$$\sigma_1^2 \cdots \sigma_n^2 = \det(A^*A) = |\det(A)|^2$$

and

$$|\lambda_1| \cdots |\lambda_n| = |\det(A)|,$$

so we have

$$|\lambda_1|\cdots|\lambda_n|=\sigma_1\cdots\sigma_n.$$

More generally, Hermann Weyl proved the following remarkable theorem:

**Theorem 22.6.** (Weyl's inequalities, 1949) For any complex  $n \times n$  matrix, A, if  $\lambda_1, \ldots, \lambda_n \in \mathbb{C}$  are the eigenvalues of A and  $\sigma_1, \ldots, \sigma_n \in \mathbb{R}_+$  are the singular values of A, listed so that  $|\lambda_1| \geq \cdots \geq |\lambda_n|$  and  $\sigma_1 \geq \cdots \geq \sigma_n \geq 0$ , then

$$|\lambda_1| \cdots |\lambda_n| = \sigma_1 \cdots \sigma_n$$
 and  $|\lambda_1| \cdots |\lambda_k| \le \sigma_1 \cdots \sigma_k$ , for  $k = 1, \dots, n-1$ .

A proof of Theorem 22.6 can be found in Horn and Johnson [96], Chapter 3, Section 3.3, where more inequalities relating the eigenvalues and the singular values of a matrix are given.

Theorem 22.5 can be easily extended to rectangular  $m \times n$  matrices, as we show in the next section. For various versions of the SVD for rectangular matrices, see Strang [170] Golub and Van Loan [80], Demmel [48], and Trefethen and Bau [176].

## 22.4 Singular Value Decomposition for Rectangular Matrices

Here is the generalization of Theorem 22.5 to rectangular matrices.

**Theorem 22.7.** (Singular value decomposition) For every real  $m \times n$  matrix A, there are two orthogonal matrices U ( $n \times n$ ) and V ( $m \times m$ ) and a diagonal  $m \times n$  matrix D such that  $A = VDU^{\top}$ , where D is of the form

$$D = \begin{pmatrix} \sigma_1 & \dots & & & \\ & \sigma_2 & \dots & & \\ \vdots & \vdots & \ddots & \vdots & & \\ & & \dots & \sigma_n \\ 0 & \vdots & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \vdots & \dots & 0 \end{pmatrix} \quad or \quad D = \begin{pmatrix} \sigma_1 & \dots & 0 & \dots & 0 \\ & \sigma_2 & \dots & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & 0 & \vdots & 0 \\ & & \dots & \sigma_m & 0 & \dots & 0 \end{pmatrix},$$