

Recall the singular valued decomposition of a matrix  $A = U\Sigma V'$ . Let  $u_i$  and  $v_i$  denote the  $i^{th}$  columns of  $U$  and  $V$  respectively and  $\sigma_i$  the  $i^{th}$  singular value ( $\Sigma_{ii} = \sigma_i$ ). Assume the singular values are enumerated in decreasing order so that  $\sigma_i \geq \sigma_{i+1}$ . With these conventions, the SVD takes the form

$$A = \sum_i \sigma_i u_i v_i'.$$

Notice that each term in this sum is a rank-1 matrix. We can approximate the matrix  $A$  by truncating this sum, effectively ignoring the small singular values. Let's define  $A_n$  to be the matrix formed by retaining the first  $n$  terms of the singular value decomposition:

$$A_n = \sum_{i=1}^n \sigma_i u_i v_i'.$$

Import the grey-scale image, *weathered-face.jpg* as a matrix,  $A$ , and compute the singular value decomposition of  $A = U\Sigma V'$ .

- How many non-zero singular values does  $A$  have?
- What are the values of the five largest singular values?
- Construct the approximations  $A_{10}$ ,  $A_{20}$  and  $A_{40}$  and plot their corresponding images.
- How many terms  $n$  do you do need to retain in the sum to recognize that the approximation  $A_n$  is a face?

