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# Course of Numerical Methods for Engineering

## Lab 12

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**PHYS-ENG, A.Y. 2020-21**

**30/11-1/12/2020**



**Topic of this session:**

- ▶ **Applications of the Singular Value Decomposition**



# Frobenius norm of a matrix

## Definition

For any  $m \times n$  matrix  $A$  its **Frobenius norm** is defined by

$$\|A\|_F = \sqrt{\sum_{i=1}^m \sum_{j=1}^n |a_{i,j}|^2}$$

## Theorem

For any  $m \times n$  matrix  $A$  with  $m \geq n$ , it holds

$$\|A\|_F = \sqrt{\sum_{i=1}^n \sigma_i^2(A)},$$

where  $\sigma_i(A)$  denote the **singular values** of  $(A)$ .



## Theorem

Given a  $m \times n$  matrix  $A$  with  $m > n$  and with rank  $r(A) = r \leq n$ , let  $A = U\Sigma V^T$  be its singular value decomposition and let  $\hat{A} = U\hat{\Sigma}V^T$ , where  $\hat{\Sigma}$  is obtained from  $\Sigma$  setting  $\sigma_{p+1}(A) = \dots = \sigma_r(A) = 0$  for some  $p < r$ . One has then that

$$\|A - \hat{A}\|_F = \sqrt{\sum_{i=p+1}^r \sigma_i^2(A)}.$$

**Application to ill-conditioned problems:** when  $\sigma_1/\sigma_n \gg 1$ , take  $\hat{A} = U\hat{\Sigma}V^T$ , where  $\hat{\Sigma}$  is obtained by dropping a sufficient number of singular values so as to make  $\sigma_1/\sigma_p$  sufficiently small, but such that

$$\|A - \hat{A}\|_F = \sqrt{\sum_{i=p+1}^n \sigma_i^2}$$

is still sufficiently small.



## Exercise 1

Let  $N = 400$  and consider the linear functional  $f(y) = v^T y$ , where  $y \in \mathbb{R}^N$  and  $v = [1, 1, \dots, 1, 1]^T \in \mathbb{R}^N$ . Let the vectors  $e_i$  denote the canonical basis of  $\mathbb{R}^N$  and consider the vector  $b \in \mathbb{R}^{2N-1}$  such that

$$b_i = f(e_i) + \epsilon_\sigma, \quad i = 1, \dots, N \quad b_i = f(e_i + e_N) + \epsilon_\sigma \quad i = N + 1, \dots, 2N - 1,$$

where  $\epsilon_\sigma$  denotes a random variable with Gaussian distribution with zero mean and standard deviation  $\sigma$ . Build the matrix of the overdetermined system whose  $i$ -th equation is given by  $e_i^T x = b_i \quad i = 1, \dots, N \quad (e_i + e_N)^T x = b_i \quad i = N + 1, \dots, 2N - 1$  in the case  $\sigma = 0.01$  (use the Octave/MATLAB command `randn`).

Check that the associated normal equations have a unique solution. Compute the solution of the overdetermined system:

- (a) solving the normal equations by the Cholesky method;
- (b) using the SVD decomposition approach.

Compare all the numerical solutions obtained with the reference solution  $x_{\text{ex}} = v$ , by computing relative  $l_2$  and  $l_\infty$  errors. Repeat the computation for  $\sigma = 2$  and discuss the differences between the two cases.



## Exercise 2

Repeat the previous exercise for the underdetermined system whose data are given by the vector  $\mathbf{b} \in \mathbb{R}^{N/2}$  such that

$$b_i = f(\mathbf{e}_i) + \epsilon_\sigma, \quad i = 1, \dots, N/2.$$



## Exercise 3

Consider the  $100 \times 100$  tridiagonal matrix  $B$  with values equal to 2 on the main diagonal and to -1 on the first sub and superdiagonal. Build the matrix

$$A = \exp(-5B) = \sum_{k=0}^{\infty} \frac{(-5B)^k}{k!}$$

using the MATLAB command `expm`. Using the SVD of  $A$ , build an approximation  $\hat{A}$  of rank 40 and compute the relative error in the Frobenius norm. Display the singular values of  $A$  in a semi-logarithmic plot.



## Exercise 4

Consider the vector  $v = [0, 1/N, \dots, (N-1)/N, 1]^T$  with  $N = 50$  and build the Vandermonde matrix  $A$  associated to  $v$  using the Octave/MATLAB command `vander`.

- (a) Solve the linear system  $Ax = b$  whose right-hand side is such that the exact solution is  $x_{ex} = [1, 1, \dots, 1, 1]^T \in \mathbb{R}^{N+1}$  by the LU factorization approach with pivoting. Check that the *a priori* error estimate in the  $l_2$  and  $l_\infty$  norms are verified.
- (b) Using the SVD of  $A$ , build an approximation  $\hat{A}$  of rank  $p < N + 1$  such that  $\|\hat{A} - A\|_F / \|A\|_F < 10^{-7}$  and  $\sigma_1(\hat{A}) / \sigma_p(\hat{A}) < 10^7$ . Display the singular values of  $A$  in a semi-logarithmic plot. Solve the resulting singular system in the minimum norm least square sense using the SVD of  $\hat{A}$ . Compute the relative error in the  $l_2$  norm.





## Exercise 5

**Check whether it is possible to solve exercise 4 when starting from the vector**

$$v = [0, 1/N, \dots, (N-1)/N, 3]^T.$$



## Exercise 6

**Plot filled contours (Octave/MATLAB: `contourf`) of the function  $f(x, y) = \sin(6\pi(x + y^2))$  in  $[0, 1] \times [0, 1]$  using an equispaced grid with  $\Delta x = \Delta y = 0.01$ . Encode the values of  $f$  in a matrix  $A$ . Using the SVD of  $A$ , build an approximation  $\hat{A}$  of rank 3, compute the relative error in the Frobenius norm, and visually compare the contour plots of  $A$  and  $\hat{A}$ .**

