

# Homework 4 - due Wed. Feb. 26th

## Mathematical and Computational Foundations of Data Science

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**Homework Policies.** As in the first homework set.

### Assignment

**SVD and Least Squares.** Study the materials presented in class on Singular Value Decomposition and its applications as Principal Component Analysis and the solution of Least Squares problems.

### Exercises

*Exercise 1* (Vandermonde Matrix, revisited). Back to the Vandermonde matrix discussed in the first homework set. Fix  $\{x_i\}_{i=1}^m$  points in an interval  $I$  (e.g.  $I = [-1, 1]$ ). Consider the map  $\mathbb{R}^{d+1} \rightarrow \mathbb{R}^m$  defined by  $c = (c_0, \dots, c_d) \mapsto (p_c(x_1), \dots, p_c(x_m))$ , where  $p_c = \sum_{l=0}^d c_l x^l$ , i.e. from vectors of coefficients of polynomials of degree  $\leq d$  to vectors  $(p_c(x_i))_{i=1}^m$  of values of such polynomials. From the first homework set, we know that:

- this map is linear
- this map is represented, upon choosing the standard basis in  $\mathbb{R}^{d+1}$  and  $\mathbb{R}^m$ , by the  $m \times d$  Vandermonde matrix

$$A = \begin{bmatrix} 1 & x_1 & x_1^2 & \dots & x_1^d \\ 1 & x_2 & x_2^2 & \dots & x_2^d \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_m & x_m^2 & \dots & x_m^d \end{bmatrix}.$$

Therefore, if  $c = [c_0; c_1; \dots; c_d]$  is the column vector of coefficients of  $p$ , i.e.  $p(x) = c_0 + c_1x + c_2x^2 + \dots + c_dx^d$ , then  $Ac$  gives the sampled polynomial values, i.e.  $(Ac)_i = p(x_i)$ , for  $i = 1, \dots, m$ .

Suppose there is a (say, continuous) function  $f : I \rightarrow \mathbb{R}$ , and we have access to  $(x_i, y_i)_{i=1, \dots, m}$ , where  $y_i = f(x_i) + \sigma\eta_i$ , where  $\eta_i \sim \mathcal{N}(0, 1)$ , and  $\sigma$  is a (fixed) positive number (standard deviation of the noise  $\eta_i$ ).

We want to construct a polynomial  $\hat{f}$  of degree  $\leq d$  that approximates  $f$  in the least squares sense.

- Write down a least squares problem for  $\hat{f}$  given the data  $\{(x_i, y_i)\}_{i=1}^m$ , as we did in class, in the form  $Ac = y$  (what are  $c$  and  $y$ ? which dimension do they have and what do they represent?)
- Write code implementing this computation, that solves the least squares problem using SVD, in the case of  $d < m$ .
- Let  $f(x) = \sin(x)$ , and the interval  $I = [-1, 1]$ . Fix  $d = 10$ , and let  $x_1, \dots, x_m$  be drawn uniformly at random on  $I$ . Fix  $\sigma = 0.05$ . Compute  $\hat{f}$ , and note that it is random, because of the randomness in the points  $x_i$  and in the noise  $\eta_i$  (to be sure, run the code a few times and visualize the corresponding  $\hat{f}$ 's obtained with each run). In fact, let's denote  $\hat{f}_m := \hat{f}$ , as the dependency on  $m$ , the number of points, will be in focus. Plot the mean and standard deviation of  $\|\hat{f}_m - f\|_{L^2([-1, 1])}$  as a function of  $m$ , and comment on their behavior as  $m$  grows (to compute these  $L^2$ -norms, use a function to compute the relevant integral already available in the software application you are using). [Note that at this point you are using the true  $f$ , but you have not used anything but the  $y_i$ 's to produce  $\hat{f}_m$ .]