Introduction

This coursework covers material in the first three weeks of the module. It is divided into three parts which roughly will correspond to the material in each week; however you can carry out the assignment in the time that suits you, with all parts being handed in as a single submission. The assignments will be closely related to the support in the lab classes which you should attend to get help with completing them.

Tasks

Week 1: 14. January

1. Solving Underdetermined Problems (30%)

As explained in lectures the linear equation $x_1 + 2x_2 = 5$ is used to introduce the concept of minimum norm solutions of underdetermined problems:

$$\mbox{minimize} \quad \Phi = \sum_i |x_i|^p, \qquad \mbox{subject to} \quad Ax = b,$$
 where in our case, $A = (1\ 2),\, b = 5.$

- a.) Write a function of two variables, x and p, where x is a vector of length 2 and p is a scalar; this function will compute the value of Φ as given above.
- b.) Use library functions to compute solutions of the above optimization problem, for $p = \frac{1}{2}$ 1, 1.5, 2, 2.5, 3, 3.5, 4.
- c.) Plot the solutions you have obtained as points on a 2D graph together with the line representing the constraint equation $x_1 + 2x_2 = 5$. The result should look like Figure 1.
- d.) Another solution method is to use the Moore-Penrose generalised inverse

$$A^{\dagger} := A^{\mathsf{T}} (AA^{\mathsf{T}})^{-1} \,,$$

and then apply it to get $x_{MP} = A^{\dagger}b$. Implement this solution and plot it on the same graph as in the previous question. What value of p does this correspond to and why?

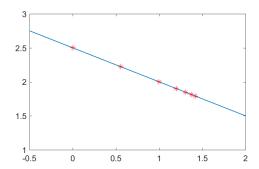


Figure 1: Example for a line plot for Exercise 1

In parts 2 and 3 of the assignment we will implement discrete convolution of a 1D function by an explict matrix vector multiplication, analyse its behaviour and compare to other methods for performing convolution.

Week 2: 21. January

2. Singular Value Decomposition (30%)

In this part we analyse the matrix representation of convolution using Singular Value Decomposition (SVD).

- a.) Set up a spatial grid on the interval [-1,1] in n equally spaced steps of size δn . The grid represents the values $[x_1,\ldots x_n]$ with $x_i=-1+(i-1)\delta n$ for $i=1,\ldots,n$ and $\delta n=2/(n-1)$. Make sure that $x_1=-1$ and $x_n=1$.
- b.) Create a vector of values of the Gaussian function centred at $\mu = 0$ with $\sigma = 0.2$, given by

$$G(x) = \frac{\delta n}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right).$$

You should evaluate this function at the grid points you create in part a).

c.) Create the convolution matrix of size $n \times n$ with entries

$$A_{i,j} = G(x_i - x_j) = \frac{\delta n}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(x_i - x_j)^2}{2\sigma^2}\right)$$
 (1)

- d.) Plot the matrix A as an image, where A is considered as a 2D-array for the case n = 100.
- e.) Compute the SVD of matrix A using library functions. You should obtain three matrices U, W, V, where W is a diagonal matrix of same size as A containing the singular values. Verify that the equation $A = UWV^{\mathsf{T}}$ is satisfied.

- f.) Compute the pseudoinverse A^{\dagger} of A by using the formula $A^{\dagger} = VW^{\dagger}U^{\mathsf{T}}$ as given in lectures. This requires you to create a method for constructing W^{\dagger} . For the case n=10, check that this has the property $WW^{\dagger} = W^{\dagger}W = Id_n$ where Id_n is the $n \times n$ Identity matrix. Check also that $AA^{\dagger} = A^{\dagger}A = Id_n$.
- g.) Repeat the last two steps for n=20. What do you observe? Choose n=100 again and plot the first 9 columns of V, the last 9 columns of V, and the singular values on a logarithmic scale, i.e. $\log(\operatorname{diag}(W))$.

Week 3: 28. January

3. Convolutions and Fourier transform (40%)

In the following we want to examine the convolution of a 1D signal; we define a step function on the interval [-1,1] by

$$f(x) = \chi_{(-0.95, -0.6]}(x) + 0.2\chi_{(-0.6, -0.2]}(x) - 0.5\chi_{(-0.2, 0.2]}(x) + 0.7\chi_{(0.4, 0.6]}(x) - 0.7\chi_{(0.6, 1]}(x),$$

where the characteristic function of an interval (a, b] is defined as

$$\chi_{(a,b]}(x) = \begin{cases} 1 & \text{for } a < x \le b \\ 0 & \text{otherwise.} \end{cases}$$

Here's a plot of f in Figure 2.

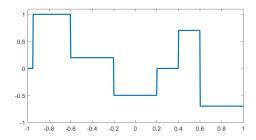


Figure 2: The step function f

- a.) Create a function for f as above and plot it on a grid on the interval [-1,1]. Choose a sufficiently large number of grid points n to resolve the jumps.
- b.) Compute the matrix A as in part 2 for $\sigma = 0.05$, 0.1, 0.2 and plot the singular values.
- c.) Verify that the plot of the singular values follows (half) a Gaussian function and determine the variance of this Gaussian in each case.
- d.) Perform the convolution of the function f with the matrix A (by matrix multiplication) for all three choices of σ and plot the result.

- e.) Since convolution is equivalent to multiplication in Fourier space, perform convolution by multiplication in Fourier space for the three choices of σ and plot the result (remember to take the inverse Fourier transform); comment on any differences that you observe.
- f.) Repeat the convolution with the matrix A using periodic boundary conditions when assembling A.

Report

Write one report for all 3 parts. Explain your method and present your results and figures. Make sure that you provide an answer to all questions. The total length of the report would normally be between 6-10 pages. Submit your report using Moodle. Code can be uploaded separately.