## **DATA.ML.200 Pattern Recognition and Machine Learning**

Homework 1: Estimation theory

This homework prepares you for the next week exercises.

## 1. pen&paper LS estimator.

You have  $n=0,\dots,N-1$  samples from a sensor that can be modeled using a sinusoid

$$x[n] = A\cos\left(2\pi f_0 n + \phi\right) . \tag{1}$$

The estimation problem is to estimate the model parameters, one, two or all three of them,  $\theta = (A, f_0, \phi)$ .

The LS estimator  $\theta_{LSE}$  estimates the model parameters by finding the values that minimise the sum of squared errors (SSE) "loss"  $\ell_{SSE}$ . Define how the loss is computed for N samples x[n].

## 2. **pen&paper** *ML estimator*.

LS estimator does not explicitly model the noise in samples, but the noise w[n] is intrinsic property of real measurements. The model with the noise term is defined as

$$x[n] = A\cos(2\pi f_0 n + \phi) + w[n] , \qquad (2)$$

where  $w[n] \sim \mathcal{N}(0, \sigma^2)$ .

The ML estimator  $\theta_{ML}$  estimates the model parameters by finding the values that maximise the probability of obtaining them. This is done by computing the product of likelihood of each sample  $p(x[n];\theta)$ . Define the total likelihood of N samples.