

## WASP course – Machine Learning

**Exercise 1.** Consider the following generative linear model:  $y_i = a.x_i + n_i, i = 1, \dots, n$ , where  $a$  is an unknown scalar, and the noise process  $n_i$  is such that  $E(n_i) = 0$  and  $Cov(n_i, n_j) = \sigma^2\delta_{ij}$ .

- a. Derive the LSE estimator  $\hat{a}$ .
- b. We propose a different estimator  $\tilde{a}$  constructed by considering the straight line going through the origin and the centre of gravity of the data. Compute  $\tilde{a}$ .
- c. Are  $\tilde{a}$  and  $\hat{a}$  biased?
- \*d. Compare the variance of the two estimators.

**Exercise 2.** Reinforcement learning. An agent has an asset to sell. She sequentially receives  $N$  offers  $w_1, \dots, w_N$  i.i.d. with density  $f(w)$ . At time  $i$ , after receiving the offer  $w_i$ , the decision maker has to decide whether to accept the offer or to reject it. If the offer is accepted, the reward is  $(1 + r)^{N-i}w_i$ , where  $r > 0$  denotes the interest rate. Once she accepted an offer, the subsequent offers do not matter. The problem is to define a strategy maximizing the expected reward

- a. Formulate the problem of identifying the best policy as a Markov Decision Process.
- b. Compute the best threshold-based policy using dynamic programming. Note: at step  $k$ , a threshold based policy accepts the offer if and only if the offer is greater than a threshold  $\alpha_k$ .
- c. The agent has now many identical assets to sell sequentially. For each asset, as before, she receives i.i.d. offers with unknown distribution (the density  $f$  is unknown). Propose a simple algorithm whose strategy converges to the optimal strategy identified in b.