Model Evaluation through a Monte Carlo Experiment Machine Learning Concepts CEU, Winter 2024

- 1. The goal of this problem is to conduct Monte Carlo simulations about the mean squared error of various regression models.
 - a) Let X_1 , X_2 and X_3 be independent uniform[0,2] variables and let $\epsilon \sim N(0,1)$, independent of $X = (X_1, X_2, X_3)$. Generate n = 500 observations on these variables and compute the corresponding outcomes from the model

$$Y_i = X_{1i}^3 - 3.5X_{1i}^2 + 3X_{1i} + \epsilon_i, \qquad i = 1, \dots, n.$$

Note that $f(X) = f(X_1) = X_1^3 - 3.5X_1^2 + 3X_1$, i.e., the predictors X_2 and X_3 are useless (irrelevant).

- b) Estimate three different regression models using the data generated under part a) as the training sample:
 - Model 1: The OLS regression of Y on a constant and X_1
 - Model 2: The OLS regression of Y on a constant, X_1 , X_1^2 and X_1^3
 - Model 3: The OLS regression of Y on a constant, X_1 , X_2 , X_3 , and all powers and cross-products of order 2 and 3. (This means all terms of the form $X_1^a X_2^b X_3^c$ where $a, b, c \in \{0, 1, 2\}$ and a + b + c = 2 or a + b + c = 3.)
- c) Compute the estimate $\hat{f}_m(x)$ from each model m = 1, 2, 3 for x = (0.1, 0, 0) and x = (1.1, 0, 0). Compute the true value f(x).
- d) Repeat steps a) through c) many times, say, R = 5000. Then you'll have R possible estimates $\hat{f}_m(x)$ from each model for the two value of x. Compute $bias[\hat{f}_m(x)]$, $Var[\hat{f}_m(x)]$ and $MSE[\hat{f}_m(x)] = bias^2[\hat{f}_m(x)] + Var[\hat{f}_m(x)]$ for all models and x. Organize your results into a table.
- e) Pick 10 of your training samples and produce three separate figures where you plot the 10 estimates $x_1 \mapsto \hat{f}_m(x_1, 0, 0)$ over a fine grid of x_1 values (go from 0 to

- 2 in small steps) for m = 1, 2, 3. Add the plot of $f(x_1, 0, 0)$ to each figure. This will help you visualize and understand the results under part d).
- f) Interpret the results you have obtained under part d) and e). Discuss bias-variance tradeoff, underfit, overfit, etc.