Unit #3

Jan. 27, 2023

## **PROBLEMS**

## Due Friday Feb. 3, 2023 @ 4pm

- 1. For each of the following pairs of "true" functions  $f(\mathbf{x})$  and model functions  $\hat{f}(\mathbf{x};\boldsymbol{\beta})$ , determine: (i) whether the model function is linear in the parameters  $\beta$ ; (ii) whether there is underfitting in the model; and (iii) if there is no underfitting, what are the true model parameters  $\beta$ ?
  - (a)  $f(x) = (1+x)^2$  and  $\hat{f}(x; \beta) = \beta_0 + \beta_1 x + \beta_2 x^2$ .
  - (b) f(x) = 1/(2-x) 2 and  $\hat{f}(x; a_0, a_1, b_0, b_1) = (a_0 + a_1 x)/(b_0 + b_1 x)$ .
  - (c)  $f(\mathbf{x}) = (x_1 x_2)^2$  and  $\widehat{f}(\mathbf{x}; a, b_1, b_2, c_1, c_2) = a + b_1 x_1 + b_2 x_2 + c_1 x_1^2 + c_2 x_2^2$ . (d)  $f(\mathbf{x}) = 3 \ln(x_1 x_2 / x_3)$  and  $\widehat{f}(\mathbf{x}; \boldsymbol{\beta}) = \sum_{j=1}^3 \beta_j \ln(x_j)$
- 2. A medical researcher is trying to choose between three model orders d. To evaluate the models, she uses 10-fold cross validation, which gave the following results.

Model	Training	Test	Test
order	$\overline{ ext{RSS}}_d$	$\overline{\mathrm{RSS}}_d$	$stdv(RSS_d)$
d=1	2.0	2.01	0.25
d=2	0.7	0.72	0.06
d=3	0.65	0.70	0.07

Note that the last column is the *unbiased* version of the sample standard deviation, i.e.,  $\operatorname{stdv}(\operatorname{RSS}_d) = \sqrt{\frac{1}{K-1}\sum_{k=1}^K (\operatorname{RSS}_{d,k} - \overline{\operatorname{RSS}}_d)^2}$ , not the standard error. Which model should be selected based on the "one standard error rule"?

- 3. Consider a random variable x with  $\mathbb{E}\{x\} = \mu$  and  $\operatorname{var}\{x\} = v$ , and another random variable  $\epsilon$  with  $\mathbb{E}\{\epsilon\}=0$  and  $\operatorname{var}\{\epsilon\}=\sigma^2$ , where  $\epsilon$  and x are independent. Finally, suppose  $y=x+\epsilon$ . Evaluate the following expectations:
  - (a)  $\mathbb{E}\{x^2\}$
  - (b)  $\mathbb{E}\{xy\}$
  - (c)  $\mathbb{E}\{y^2\}$
  - (d)  $\mathbb{E}\{y|x\}$
  - (e)  $\mathbb{E}\{y^2|x\}$

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4. Suppose that some training data  $\{(x_i, y_i)\}_{i=1}^n$  was generated from a noisy linear model

$$y_i = f(x_i) + \epsilon_i, \quad f(x_i) = \beta x_i.$$

For prediction, we assume the linear model

$$\widehat{y} = \widehat{f}(x; \widehat{\beta}) = \widehat{\beta}x,$$

with  $\widehat{\beta}$  computed using a least-squares (LS) fit to the training data. Finally, we wonder how our prediction will fare on test data (x, y) generated via

$$y = f(x) + \epsilon$$
.

You can assume that  $\{\epsilon_i\}$  and  $\epsilon$  are mutually independent, zero-mean,  $\sigma^2$ -variance random variables that are independent of  $\{x_i\}$  and x.

- (a) Derive an expression for the LS  $\widehat{\beta}$  in terms of  $\{(x_i, y_i)\}_{i=1}^n$  by first writing an expression for RSS and then zeroing its partial derivative with respect to  $\widehat{\beta}$ .
- (b) Write  $\widehat{\beta}$  in terms of  $\beta$  and  $\{(x_i, \epsilon_i)\}_{i=1}^n$ .
- (c) What is the bias  $\mathbb{E}\{\hat{y} y \mid x\}$ ? Is the predictor biased or unbiased? For this and remaining parts, you can assume that the training  $x_i = 1$  for all i, but please keep the test x generic.
- (d) What is the mean-squared error  $\mathbb{E}\{(\hat{y}-y)^2 \mid x\}$ ?
- (e) What is the variance of the prediction,  $\operatorname{var}\{\widehat{y} \mid x\}$ ?
- (f) Compute the irreducible error by subtracting the prediction variance and the squared bias from the MSE. Does it turn out as expected?