

FYS-STK4155 Week 36

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Exercise 1

Show that

$$\mathbb{E}[(\mathbf{y} - \tilde{\mathbf{y}})^2] = \text{Bias}[\tilde{\mathbf{y}}] + \text{Var}[\tilde{\mathbf{y}}] + \sigma^2$$

where \mathbf{y} is defined by $\mathbf{y} = f(\mathbf{x}) + \varepsilon$, $\varepsilon \sim N(0, \sigma^2)$ is a normal distributed error and $f(\mathbf{x})$ is the approximated function given our model $\tilde{\mathbf{y}}$ obtained by minimizing $(\mathbf{y} - \tilde{\mathbf{y}})^2$ with $\tilde{\mathbf{y}} = \mathbf{X}\boldsymbol{\beta}$. Explain what the terms mean and discuss their interpretations. Perform then a bias-variance analysis of a simple one-dimensional function by studying the MSE value as a function of the complexity of your model. Use OLS only. Discuss the bias and variance trade-off as function of your model complexity (the degree of the polynomial) and the number of data points. **Solution:**

For ease of notation we write $f(\mathbf{x}) = f$ and simply ignore vector notation since everything is a scalar in the end. Then we have

$$\begin{aligned}\mathbb{E}[(\mathbf{y} - \tilde{\mathbf{y}})^2] &= \mathbb{E}[(f + \varepsilon - \tilde{\mathbf{y}})^2] = \mathbb{E}[(f - \tilde{\mathbf{y}})^2] + 2 \underbrace{\mathbb{E}[(f - \tilde{\mathbf{y}})\varepsilon]}_{=0} + \underbrace{\mathbb{E}[\varepsilon^2]}_{=\sigma^2} \\ &= \mathbb{E}[((f - \mathbb{E}[\tilde{\mathbf{y}}]) - (\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}]))^2] + \sigma^2 \\ &= \mathbb{E}[(f - \mathbb{E}[\tilde{\mathbf{y}}])^2] + \mathbb{E}[(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])^2] - 2 \mathbb{E}[(f - \mathbb{E}[\tilde{\mathbf{y}}])(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])] + \sigma^2 \\ &= \text{Bias}[\tilde{\mathbf{y}}] + \text{Var}[\tilde{\mathbf{y}}] + \sigma^2 - 2 \mathbb{E}[(f - \mathbb{E}[\tilde{\mathbf{y}}])(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])]\end{aligned}$$

where $\mathbb{E}[(f - \tilde{\mathbf{y}})\varepsilon] = 0$ is justified by ε being independent and we note that the wrong definition of the Bias is given in the problem text (with that definition σ^2 gets put into the 'Bias'). All that remains is to show that the last term is 0. Since $\mathbb{E}[f] = f$ and $\mathbb{E}[f \mathbb{E}[\tilde{\mathbf{y}}]] = f \mathbb{E}[\mathbb{E}[\tilde{\mathbf{y}}]] = f \mathbb{E}[\tilde{\mathbf{y}}]$ then

$$\begin{aligned}\mathbb{E}[(f - \mathbb{E}[\tilde{\mathbf{y}}])(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])] &= \mathbb{E}[f\tilde{\mathbf{y}} - f\mathbb{E}[\tilde{\mathbf{y}}] - \tilde{\mathbf{y}}\mathbb{E}[\tilde{\mathbf{y}}] + \mathbb{E}^2[\tilde{\mathbf{y}}]] \\ &= f \mathbb{E}[\tilde{\mathbf{y}}] - f \mathbb{E}[\tilde{\mathbf{y}}] - \mathbb{E}^2[\tilde{\mathbf{y}}] + \mathbb{E}^2[\tilde{\mathbf{y}}] = 0\end{aligned}$$

Since I have already written it in down I just want to quickly show why (I believe at least) the definition given in the problem text is wrong:

$$\begin{aligned}\mathbb{E}[(\mathbf{y} - \tilde{\mathbf{y}})^2] &= \mathbb{E}[((\mathbf{y} - \mathbb{E}[\tilde{\mathbf{y}}]) - (\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}]))^2] \\ &= \mathbb{E}[(\mathbf{y} - \mathbb{E}[\tilde{\mathbf{y}}])^2] + \mathbb{E}[(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])^2] - 2 \mathbb{E}[(\mathbf{y} - \mathbb{E}[\tilde{\mathbf{y}}])(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])] \\ &\quad \underbrace{=}_{\text{wrong}} \text{Bias}[\tilde{\mathbf{y}}] + \text{Var}[\tilde{\mathbf{y}}] - 2 \mathbb{E}[(\mathbf{y} - \mathbb{E}[\tilde{\mathbf{y}}])(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])]\end{aligned}$$

Then we have

$$\begin{aligned}\mathbb{E}[(\mathbf{y} - \mathbb{E}[\tilde{\mathbf{y}}])(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])] &= \mathbb{E}[(f + \varepsilon - \mathbb{E}[\tilde{\mathbf{y}}])(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])] \\ &= \mathbb{E}[(f - \mathbb{E}[\tilde{\mathbf{y}}])(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])] + \mathbb{E}[\varepsilon(\tilde{\mathbf{y}} - \mathbb{E}[\tilde{\mathbf{y}}])]\end{aligned}$$

I have already shown explicitly that the first term is 0 and the second term is 0 due to the same reasons as above. So with this definition we would get the wrong result that

$$\mathbb{E}[(\mathbf{y} - \tilde{\mathbf{y}})^2] = \text{Bias}[\tilde{\mathbf{y}}] + \text{Var}[\tilde{\mathbf{y}}]$$