

DATA 100: Vitamin 9 Solutions

April 9, 2019

1 Risk Optimization

Let $\hat{\theta}_n(x_0)$ be an estimator for $\theta(x_0) = E[Y|X = x_0]$ using a randomly sampled learning set X_n . The mean squared error (MSE) of this estimator is equal to:

- ☒ $E[(\hat{\theta}_n(x_0) - \theta(x_0))^2|X_n]$
- ☐ $Var(\hat{\theta}_n(x_0)|X_n) + E[\hat{\theta}_n(x_0)|X_n]$
- ☐ $E[(E[\hat{\theta}_n(x_0)|X_n] - \theta(x_0))^2]$
- ☒ $Var(\hat{\theta}_n(x_0)|X_n) + Bias(\hat{\theta}_n(x_0)|X_n)^2$

Explanation: See the proof in lecture 17's slides. Always remember to condition on \mathbf{X}_n !

2 Bias-Variance Trade-Off

2.1 Underfitting

Fill in the blanks:

Underfitting occurs when a model is too ____ to represent the _____ of the learning set.

- ☒ simple, data generating mechanism
- ☐ complex, data generating mechanism
- ☐ simple, bias
- ☐ complex, bias

2.2 Underfitting

Overfitting occurs when a model is too ____ and fits the ____ from sampling in the learning set.

- ☐ simple, variance
- ☐ complex, variance
- ☐ simple, noise
- ☒ complex, noise

Explanation (Q2.1 and Q2.2): Underfitting consists of fitting a model that is too simple, and overfitting consists of fitting a model that is too complex. When a model is overly complex, it tends to fit to the noise in the data instead of the signal (the true data generating mechanism). Striking a balance between simplicity and complexity is not an easy task.

2.3 Complexity

More complex models can reduce bias, but can increase variance. Which of the following are example ways to measure model complexity?

- ☒ the number of parameters
- ☒ the length of the parameter vector
- ☒ the number of covariates in a linear regression function
- ☒ the degree of the polynomial in a polynomial regression function
- ☐ None of these

Explanation: See the lecture slides for a list of models and their corresponding measures of complexity.

3 Cross Validation

K-fold cross-validation (CV) consists of partitioning the ____ set data into K disjoint subsets of roughly equal size, which are then used to train estimators and evaluate their performance. If using CV for model selection, the model with the ____ CV risk is selected. Using CV for model selection will help you avoid selecting a model that overfits to the noise in the learning data.

- ☐ test, highest
- ☐ learning, highest

☐ test, lowest

☒ learning, lowest

Explanation: CV consists of partitioning the learning set. When using CV for model selection, the model with the lowest CV risk is chosen.