Discussion #9

Cross Validation

1. After running 5-fold cross validation, we get the following mean squared errors for each fold and value of λ when using Ridge regularization:

Fold Num	$\lambda = 0.1$	$\lambda = 0.2$	$\lambda = 0.3$	$\lambda = 0.4$	Row Avg
1	80.2	70.2	91.2	91.8	83.4
2	76.8	66.8	88.8	98.8	82.8
3	81.5	71.5	86.5	88.5	82.0
4	79.4	68.4	92.3	92.4	83.1
5	77.3	67.3	93.4	94.3	83.0
Col Avg	79.0	68.8	90.4	93.2	

How do we use the information above to choose our model? Do we pick a specific fold? a specific λ ? or a specific fold- λ pair? Explain.

- 2. In the typical setup of k-fold cross validation, we use a different parameter value on each fold, compute the mean squared error of each fold and choose the parameter whose fold has the lowest loss.
 - A. True
 - B. False

Discussion #9

Guessing at Random

3. A multiple choice test has 100 questions, each with five possible answers of which one is right. The grading scheme is as follows:

- 4 points are awarded for each right answer.
- For each other answer (wrong, missing, etc), one point is taken off; that is, -1 points are awarded.

A student hasn't studied at all and therefore guesses each answer uniformly at random, independently of all the other answers.

Define the following random variables:

- R: the number of answers the student gets right
- W: the number of answers the student does not get right
- S: the student's score on the test

We analyze the random variable R, which denotes the number of answers the student got right.

(a) What is the distribution of R? Provide the name and parameters of the appropriate distribution. Explain your answer.

- (b) Find $\mathbb{E}(R)$.
- 4. True or False: $\mathbb{SD}(R) = \mathbb{SD}(W)$.
- 5. Find $\mathbb{E}(S)$, the student's expected score on the test.
- 6. Find SD(S).

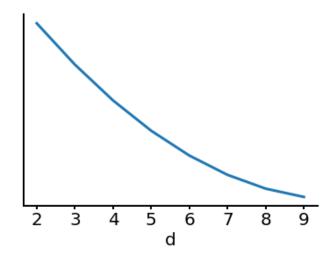
Discussion #9

Bias-Variance Trade-Off

7. Your team would like to train a machine learning model in order to predict the next YouTube video that a user will click on based on the videos the user has watched in the past. We extract m attributes (such as length of video, view count etc) from each video and our model will be based on the previous d videos watched by that user.

Hence the number of features for each data point for the model is $m \cdot d$. Currently, you're not sure how many videos to consider.

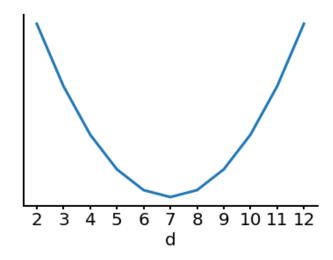
(a) Your colleague generates the following plot, where the value d is on the x-axis. However, they forgot to label the y-axis.



Which of the following could the y-axis represent? Select all that apply.

- ☐ A. Training Error
- ☐ B. Validation Error
- ☐ C. Bias
- ☐ D. Variance
- (b) Your colleague generates the following plot, where the value d is on the x-axis. However, they forgot to label the y-axis again.

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Which of the following could the y axis represent? Select all that apply.

- \square A. Training Error
- ☐ B. Validation Error
- ☐ C. Bias
- ☐ D. Variance
- 8. We randomly sample some data $(x_i, y_i)_{i=1}^n$ and use it to fit a model $f_{\hat{\theta}}(x)$ according to some procedure (e.g. OLS, Ridge, LASSO). We then sample a new point that is independent from our existing points, but sampled from the same underlying truth as our data. Furthermore, assume that we have a function g(x) and some noise generation process that produces ϵ such that $\mathbb{E}\left[\epsilon\right]=0$ and $\mathrm{var}(\epsilon)=\sigma^2$. Every time we query mother nature for Y at a given a x, she gives us $Y=g(x)+\epsilon$. (The true function for our data is $Y=g(x)+\epsilon$.) A new ϵ is generated each time, independent of the last. In class, we showed that

$$\underbrace{\mathbb{E}\left[(Y - f_{\hat{\theta}}(x))^2\right]}_{} = \underbrace{\sigma^2}_{} + \underbrace{\left(g(x) - \mathbb{E}\left[f_{\hat{\theta}}(x)\right]\right)^2}_{} + \underbrace{\mathbb{E}\left[\left(f_{\hat{\theta}}(x) - \mathbb{E}\left[f_{\hat{\theta}}(x)\right]\right)^2\right]}_{}$$

- (a) Label each of the terms above.
 - Word Bank: observation variance, model variance, observation bias², model bias², model risk, empirical mean square error.
- (b) What is random in the equation above? Where does the randomness come from?

(c) Calculate the value of $\mathbb{E}\left[\epsilon f_{\hat{\theta}}(x)\right]$.