

# 1 To Split or Not To Split?

This is some section!

**Question 1:** In this case, our hypothesis space  $\mathcal{H} = \{h_1, \dots, h_M\}$  is finite with  $|\mathcal{H}| = M$ , which means that we can use **Theorem 3.2** to conclude that with probability  $1 - \delta$  for all  $h \in \mathcal{H}$

$$L(\hat{h}^*) \leq \hat{L}(\hat{h}^*, S_{val}) + \sqrt{\frac{\ln \frac{M}{\delta}}{2n}} \quad (1)$$

where  $n = |S_{val}|$ .

**Question 2:** Let  $S_{val}^*$  be the validation set on which we are testing the hypothesis  $\hat{h}^*$  that we end up choosing. In the setup proposed by our fellow student, we are only testing a single hypothesis, namely  $\hat{h}^*$ , on  $S_{val}^*$ , and  $|S_{val}^*| = \frac{n}{M}$ . Therefore, we can use **Theorem 3.1** to conclude that with probability  $1 - \delta$  for all  $h \in \mathcal{H}$

$$L(\hat{h}^*) \leq \hat{L}(\hat{h}^*, S_{val}) + \sqrt{\frac{\ln \frac{1}{\delta}}{2 \frac{n}{M}}} = \hat{L}(\hat{h}^*, S_{val}) + \sqrt{\frac{M \ln \frac{1}{\delta}}{2n}} \quad (2)$$

Our fellow student has therefore made a bad proposal, since bound is now growing linearly with  $M$  instead of logarithmically.

**Question 3:** Again we only test a single hypothesis, namely  $\hat{h}^*$ , on  $S_{val}^2$ . This time we have that  $|S_{val}^2| = \frac{n}{2}$ . Therefore, we can use **Theorem 3.1** to conclude that with probability  $1 - \delta$  for all  $h \in \mathcal{H}$

$$L(\hat{h}^*) \leq \hat{L}(\hat{h}^*, S_{val}) + \sqrt{\frac{\ln \frac{1}{\delta}}{2 \frac{n}{2}}} = \hat{L}(\hat{h}^*, S_{val}) + \sqrt{\frac{\ln \frac{1}{\delta}}{n}} \quad (3)$$

Assume that my fellow student followed this procedure, and I followed the procedure in question 1. Let  $\hat{h}^*$  be the hypothesis that I end up choosing, and let  $\tilde{h}^*$  be the hypothesis my fellow student chooses. Where I am using the full  $S_{val}$  to choose  $\hat{h}^*$ , my fellow student is only using  $S_{val}^1$  to choose  $\tilde{h}^*$ . Therefore, we cannot not be sure that  $\hat{h}^* = \tilde{h}^*$ . Apart from not knowing whether we choose the same hypothesis, we also do not test our chosen hypothesis on the same set. Where I am using  $S_{val}$ , my fellow student is using  $S_{val}^2$ . All in all, it is therefore not very easy to tell how close my empirical error  $\hat{L}(\hat{h}^*, S_{val})$  is to the empirical error  $\hat{L}(\tilde{h}^*, S_{val}^2)$  of my fellow student. However, we can say that I have a higher probability of choosing the hypothesis  $h_i$  in  $\mathcal{H}$  with the lowest expected loss  $L(h_i)$ , since I am using a bigger validation set to inform my decision.

If we assume that  $\hat{L}(\hat{h}^*, S_{val}) = \hat{L}(\tilde{h}^*, S_{val}^2)$ , then we know that my bound is tighter than my fellow student's, if and only if

$$\sqrt{\frac{\ln \frac{M}{\delta}}{2n}} < \sqrt{\frac{\ln \frac{1}{\delta}}{n}} \quad (4)$$

This is equivalent to

$$\ln \frac{M}{\delta} < 2 \ln \frac{1}{\delta} \quad (5)$$

which is equivalent to

$$\frac{M}{\delta} < \left(\frac{1}{\delta}\right)^2 \quad (6)$$

which is equivalent to

$$M\delta < 1 \quad (7)$$

This means that under the assumption that  $\hat{L}(\hat{h}^*, S_{val}) = \hat{L}(\tilde{h}^*, S_{val}^2)$ , then if we for instance wanted a certainty  $1 - \delta = 0.95$ , then my procedure would have a tighter bound, if and only if  $M < 20$ .

As I had already said, then even if we had a big  $M$ , my fellow student would still be less certain than me of picking the best hypothesis in  $\mathcal{H}$ , which is a drawback of his method.

**Question 4:** As I have already explained in question 3, then choosing a large  $\alpha$  - and thereby a large validation set  $S_{val}^1$  - means having a better chance of choosing the hypothesis in  $\mathcal{H}$ , which actually has the lowest expected loss, as  $\hat{h}^*$ . This also means that we should expect a lower empirical loss  $L(\hat{h}^*, S_{val}^2)$  on the test set  $S_{val}^2$  than if we had used a smaller validation set to choose  $\hat{h}^*$ . However, a large  $\alpha$  also means a small test set. Therefore, we also get more uncertain how well the empirical loss  $L(\hat{h}^*, S_{val}^2)$  on the test set reflects the true expected loss  $L(\hat{h}^*)$ . This can be seen by the fact that the term

$$\sqrt{\frac{\ln \frac{1}{\delta}}{2(1 - \alpha)n}} \quad (8)$$

in our bound

$$L(\hat{h}^*) \leq \hat{L}(\hat{h}^*, S_{val}^2) + \sqrt{\frac{\ln \frac{1}{\delta}}{2(1 - \alpha)n}} \quad (9)$$

grows when  $\alpha$  becomes larger. All in all, it therefore not clear whether a larger  $\alpha$  will make us choose  $\hat{h}^*$ , such that the resulting bound on  $L(\hat{h}^*)$  becomes larger or smaller. In general, the larger  $M$  becomes, the larger I would also choose  $\alpha$ , since a large hypothesis space also means a large probability of accidentally choosing a bad hypothesis as  $\hat{h}^*$ , if the validation set is too small.

## 2 Occam's Razor

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## 3 Kernels

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