

# Learning From Data

## Exercise 1.13

### Proof

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## Task

### Exercise 1.13

Consider the bin model for a hypothesis  $h$  that makes an error with probability  $\mu$  in approximating a deterministic target function  $f$  (both  $h$  and  $f$  are binary functions). If we use the same  $h$  to approximate a noisy version of  $f$  given by

$$P(y \mid \mathbf{x}) = \begin{cases} \lambda & y = f(\mathbf{x}), \\ 1 - \lambda & y \neq f(\mathbf{x}). \end{cases}$$

- (a) What is the probability of error that  $h$  makes in approximating  $y$ ?
- (b) At what value of  $\lambda$  will the performance of  $h$  be independent of  $\mu$ ?  
[Hint: The noisy target will look completely random.]

## Solution

### 0.1 (a)

In deterministic case the target function was giving the results like this  $f(x) = y$  with probability 100%. Now instead of  $y = f(x)$  we take an output  $y$  to be a random variable that is affected by  $x$ . We should consider two types of errors:

- $h(x) = f(x)$  and  $f(x) \neq y \rightarrow (1 - \mu) * (1 - \lambda)$
- $h(x) \neq f(x)$  and  $f(x) = y \rightarrow \mu * \lambda$

This can be presented in table form.

	$f(x)$	+	-
$h(x)$			
+		correct	$(1 - \mu) * (1 - \lambda)$
-		$\mu\lambda$	correct

Now the error will be the sum of errors which is:

$$(1 - \mu) * (1 - \lambda) + \mu * \lambda = 1 - \lambda - \mu + 2\mu\lambda$$

**0.2 (b)**

In order to make  $h$  independent of  $\mu$  we should get rid of it. We can achieve this if  $\lambda = \frac{1}{2}$

$$1 - \lambda - \mu + 2\mu\lambda = 1 - \frac{1}{2} - \mu + 2\mu\frac{1}{2} = \frac{1}{2}$$