

# 1 The Ridge Regression Estimator

See my notes on ridge regression.

## 2 Entropy and Information

### Part (a)

$$\Pr(X = x) = (1 - b)^{(x-1)} \cdot b$$

$$\begin{aligned} H(X) &= - \sum_{x=1}^{\infty} ((1 - b)^{(x-1)} \cdot b) ((x - 1) \log(1 - b) + \log b) \\ &= -b \sum_{x=1}^{\infty} \left( \log(1 - b) \cdot (x - 1)(1 - b)^{(x-1)} \right) + \left( \log b \cdot (1 - b)^{(x-1)} \right) \\ &= -b \left( \log(1 - b) \cdot \frac{1 - b}{b^2} + \log b \cdot \frac{1}{b} \right) \end{aligned}$$

### Part (b)

If  $b = 1/2$ ,  $H(X) = 2$ . For every round  $k$ , we ask  $X = k$ ? This implies the answer was 0 in all previous rounds  $(1, \dots, k - 1)$ . Note that the probability of the number of questions that we need to ask equals  $k$  is  $(\frac{1}{2})^k$ . The expected number of questions to ask is

$$\sum_{k=1}^{\infty} k \cdot \left(\frac{1}{2}\right)^k = 2 = H(X)$$

## 3 Decision Trees

### Part (a)

False. The statement would be true if the splits were allowed to form complex boundaries (not binary and linear).

### Part (b)

False. XOR example.

### Part(d)

Deep  $\Rightarrow$  High variance & Low bias (overfit).

Shallow  $\Rightarrow$  Low variance & High bias (underfit).