

Homework 2

1. *Predicting outcomes*

Consider the pmf

$$p = \frac{1}{1000} [7 \ 9 \ 10 \ 36 \ 41 \ 43 \ 55 \ 73 \ 72 \ 71 \ 72 \ 73 \ 72 \ 72 \ 72 \ 72 \ 73 \ 77]$$

on $\Omega = \{1, \dots, 18\}$. Let x be the identity random variable, i.e., $x : \Omega \rightarrow \Omega$ defined by $x(\omega) = \omega$.

- (a) Compute the mean and covariance of x .
- (b) We would like to pick an *estimate* of x . Find the estimate $\hat{x}_{\text{abs}} \in \Omega$ that minimizes the cost function

$$\mathbb{E}|x - \hat{x}|$$

Also find the minimum possible cost.

- (c) Find the estimate $\hat{x}_{\text{min-error}} \in \Omega$ that minimizes the probability of error

$$\text{Prob}(x \neq \hat{x})$$

Also find the minimum probability of error.

- (d) Find the estimate $\hat{x}_{\text{squared}} \in \Omega$ that minimizes the cost function

$$\mathbb{E}((x - \hat{x})^2)$$

What is the minimum cost? Show that, if we relax the constraint that \hat{x} lie in Ω , the \hat{x} that minimizes this cost function is

$$\hat{x} = \mathbb{E} x$$

Is it true that \hat{x}_{squared} is the element of Ω which is closest in 2-norm to $\mathbb{E} x$?

- (e) Let's simulate the random variable x (using your code from another question). Collect 10,000 samples from x , and compute the error achieved by your estimates $\hat{x}_{\text{min-error}}$, \hat{x}_{squared} and \hat{x}_{abs} using the appropriate error measure. Also compute the mean square error achieved by using the mean as an estimate.

2. *Chernoff Bounds*

- (a) The Chernoff bound is similar to the Markov and Chebyshev bounds, as follows. Suppose $x : \Omega \rightarrow \mathbb{R}$ is a random variable, and let λ be any positive real number. Then for all $\varepsilon > 0$,

$$\text{Prob}(x \geq \varepsilon) \leq e^{f(\lambda) - \lambda\varepsilon} \tag{1}$$

where the function f is

$$f(\lambda) = \log \mathbb{E}(e^{\lambda x})$$

Prove this.

- (b) Suppose we have n IID discrete random variables y_1, \dots, y_n . Each y_i is a *Bernoulli* random variable, that is it takes value either 0 or 1 with probability $\frac{1}{2}$. Define the random variable x by

$$x = \sum_{i=1}^n y_i$$

What is the pmf of x ?

- (c) When x is as in part (b), what is $f(\lambda)$? Hint; use the fact that x is a sum of IID random variables.

- (d) By choosing λ to minimize the right-hand side of equation (1), we can make the bound as tight as possible. Show that the optimal λ is

$$\lambda = \log\left(\frac{\varepsilon}{n - \varepsilon}\right)$$

and hence when $\varepsilon \geq \mathbb{E}x$, the best bound possible is

$$\text{Prob}(x \geq \varepsilon) \leq \frac{1}{2^n} \left(\frac{n}{n - \varepsilon}\right)^n \left(\frac{\varepsilon}{n - \varepsilon}\right)^{-\varepsilon}$$

What is the bound when $\varepsilon = \mathbb{E}x$?

- (e) For $n = 10$, plot

$$\text{Prob}(x \geq \varepsilon)$$

against ε , and also plot the Chernoff bound.

We can also use the Chebyshev bound to give an upper bound on this probability, since if $\varepsilon \geq \mathbb{E}x$ we have

$$\begin{aligned} \text{Prob}(x \geq \varepsilon) &= \text{Prob}(x - \mathbb{E}x \geq \varepsilon - \mathbb{E}x) \\ &\leq \text{Prob}(|x - \mathbb{E}x| \geq \varepsilon - \mathbb{E}x) \\ &\leq \frac{\text{cov}(x)}{(\varepsilon - \mathbb{E}x)^2} \end{aligned}$$

Also plot the Chebyshev bound on your plot. Notice that the Chernoff bound is tighter than the Chebyshev bound.