

ECE 313: Lecture 12
Maximum likelihood parameter estimation
Markov and Chebychev inequalities

Ex 1. Suppose we have a coin $\begin{cases} 1 \text{ (H)} & \text{w.p. } p \\ 0 \text{ (T)} & \text{w.p. } 1-p \end{cases} \sim \text{Bern}(p)$

Throw 10 trials, record outcome

$$\{X_i\}_{i=1}^{10} = (0, 0, 1, 0, 0, 0, 1, 1, 0, 0)$$

Q: ① Find an estimate of p

② Find / Determine the confidence of this $\hat{p}$

$\begin{cases} \text{Ex 1(b)} & \text{"Polling"} \\ \text{Ex 1(c)} & \end{cases}$

System reliability testing

Go back to Ex 1: $I \{ \cdot \} = P\{X_1=0\} \cdot P\{X_2=0\} \cdot P\{X_3=1\} \dots$

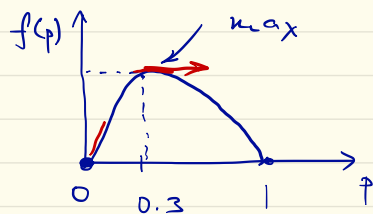
$$= (1-p) \cdot (1-p) \cdot p \dots (1-p) \dots$$

$$= p^{\# \text{ of } 1's} (1-p)^{\# \text{ of } 0's} = p^3 (1-p)^7$$

Let $f(p) = \mathbb{P}_p \{ (X_1, X_2, \dots, X_n) = (0, 0, 1, \dots) \}$
 is the like hood func. $= \boxed{p^3 (1-p)^7}$

Maximum likelihood estimator

$$\hat{p}_{ML} = \arg \max_{0 \leq p \leq 1} f(p)$$



$$f(0) = 0$$

$$f(1) = 0$$

$$f(0.3) = (0.3)^3 (0.7)^7$$

Consider a more general form $\left\{ \begin{array}{l} n \text{ trials} \\ \Rightarrow k \text{ of '1'} \end{array} \right.$

$$f(p) = p^k (1-p)^{n-k}$$

$$\frac{df}{dp} = k p^{k-1} (1-p)^{n-k} + p^k (n-k) (1-p)^{n-k-1} \cdot (-1)$$

$$\frac{df}{dp} = p^{k-1} (1-p)^{n-k-1} \left[\underbrace{k(1-p) - p(n-k)}_{=0} \right]$$

$$\frac{df}{dp} = 0$$

$\Rightarrow$

$$\frac{df}{dp} = 0 \quad (\Rightarrow) \quad k(1-p) - p(n-k) = 0$$

$$(\Rightarrow) \quad k - kp - (n-k)p = 0$$

$$(\Rightarrow) \quad np = k$$

$$(\Rightarrow) \quad p = \frac{k}{n}$$

Hence $\hat{p}_{ML} = \frac{k}{n}$

Ex 2: Throw a coin n times ← prob of "H" is p (unknown)

Observe $C = k$ heads

$$\hat{p} = ?$$

$$C \sim \text{Binom}(n, p) \quad \text{unknown } p$$

$$g(p) = \mathbb{P}\{C = k\} = \binom{n}{k} p^k (1-p)^{n-k}$$

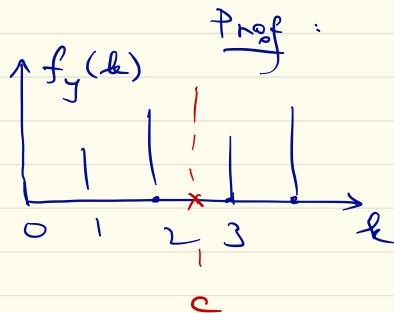
$$\arg \max_p g(p) = \arg \max_p p^k (1-p)^{n-k} = \frac{k}{n}$$

* Markov inequality:

Let Y be a non negative discrete random variable (rv)

For a fixed $c > 0$:

$$P\{Y \geq c\} \leq \frac{E[Y]}{c}$$



$$E[Y] = \sum_{k \geq 0} k \cdot P\{Y = k\}$$

$$= \underbrace{\sum_{0 \leq k < c} k P\{Y = k\}}_{\geq 0} + \underbrace{\sum_{k \geq c} k P\{Y = k\}}_{\geq c \cdot \sum_{k \geq c} P\{Y = k\}}$$

$$\geq 0$$

$$\geq c \cdot \underbrace{\sum_{k \geq c} P\{Y = k\}}_{P\{Y \geq c\}}$$

* Chebyshev inequality:

Let $Y = |X - E[X]|^2$ nonneg. r.v.

$$P \{ Y \geq d^2 \}$$

$$= P \{ |X - E[X]| > d \}$$

from Markov

$$\leq \frac{E[|X - E[X]|^2]}{d^2} = \frac{\text{Var}(X)}{d^2}$$

$$E_X : \quad X \text{ r.v. with } E[X] = 3.5 \\ \text{Var}[X] = 0.01$$

$$\Rightarrow P \{ |X - 3.5| > 0.2 \} \leq \frac{0.01}{0.2^2}$$

$$= \frac{1}{4}$$