

(18)

Date

Page

OM

Student Notebook

## Inequalities, Convergence and the Weak Law of Large Number

### ★ The Markov inequality

⇒ Use a bit of information about a distribution to learn something about probabilities of "extreme events".

Markov inequality: If  $X \geq 0$  &  $a > 0$

then  $P(X \geq a) \leq \frac{E[X]}{a}$

$$E[X] = \int_0^{\infty} x f_X(x) dx \geq \int_a^{\infty} x f_X(x) dx$$

$$\geq a \int_a^{\infty} f_X(x) dx = a P(X \geq a)$$

$$\Rightarrow P(X \geq a) \leq \frac{E[X]}{a}$$

or

$$\text{Let } Y = \begin{cases} 0 & \text{if } X < a \\ a & \text{if } X \geq a \end{cases}$$

$$\Rightarrow Y \leq X$$

$$E[Y] \leq E[X]$$

$$\Rightarrow a P(X \geq a) \leq E[X]$$

$$\Rightarrow P(X \geq a) \leq E[X]/a$$

## ★ The Chebyshev inequality

⇒ Random variable  $X$ , with finite mean  $\mu$  and variance  $\sigma^2$ .

"If the variance is small, then  $X$  is unlikely to be too far from mean"

Chebyshev inequality:  $P(|X - \mu| \geq c) \leq \frac{\sigma^2}{c^2}$

$$P(|X - \mu| \geq c)$$

$$\Rightarrow P((X - \mu)^2 \geq c^2)$$

$$\Rightarrow P((X - \mu)^2 \geq c^2) \leq \frac{E[(X - \mu)^2]}{c^2} \left\{ \begin{array}{l} \text{from} \\ \text{Markov} \\ \text{inequality} \end{array} \right\}$$

$$\Rightarrow P(|X - \mu| \geq c) \leq \frac{\sigma^2}{c^2}$$

$$\Rightarrow P(|X - \mu| \geq k\sigma) \leq \frac{\sigma^2}{k^2\sigma^2} = \frac{1}{k^2}$$



## \* The weak Law of Large Number (WLLN)

$\Rightarrow X_1, X_2, \dots$  are iid (independent & identically distributed)

$\Rightarrow$  finite mean  $\mu$  and variance  $\sigma^2$ .

Sample mean:  $M_n = \frac{X_1 + \dots + X_n}{n}$

$$E[M_n] = \frac{1}{n} (E[X_1] + \dots + E[X_n]) = \frac{n\mu}{n} = \mu$$

$$\text{Var}(M_n) = \frac{1}{n^2} \text{Var}(X_1 + \dots + X_n) = \frac{n\sigma^2}{n^2} = \frac{\sigma^2}{n}$$

$$P(|M_n - \mu| \geq \epsilon) \leq \frac{\text{Var}(M_n)}{\epsilon^2} = \frac{\sigma^2}{n\epsilon^2}$$

$\downarrow \lim_{n \rightarrow \infty}$   
 0

WLLN: For  $\epsilon > 0$

$$P(|M_n - \mu| \geq \epsilon) \Rightarrow 0 \text{ as } n \rightarrow \infty$$

## ★ The pollster's problem

⇒  $P$ : fraction of population that will vote "yes" in a referendum.

⇒  $i^{\text{th}}$  (randomly selected) person polled:

$$X_i = \begin{cases} 1 & \text{if yes} \\ 0 & \text{if no} \end{cases}$$

$$M_n = \frac{X_1 + \dots + X_n}{n} \quad \text{fraction of yes in our sample}$$

$$P(|M_{100000} - P| \geq 0.01) \leq \frac{\sigma^2}{n\varepsilon^2} = \frac{P(1-P)}{10^5 \times 10^{-4}}$$

$$\leq \frac{1}{4}$$

$$\frac{1/4}{n 10^{-4}} \leq \frac{5}{10^2} \Leftrightarrow n \geq \frac{10^6}{20} \approx 50,000$$

## ★ Convergence in Probability

Definition: A sequence  $Y_n$  converges in probability to a number  $a$  if:

$$\text{For any } \varepsilon > 0, \lim_{n \rightarrow \infty} P(|Y_n - a| \geq \varepsilon) = 0$$



⇒ Suppose that  $X_n \rightarrow a$   $Y_n \rightarrow b$  in probability

• (i) If  $g$  is continuous, then  
 $g(X_n) \rightarrow g(a)$

(ii)  $X_n + Y_n \rightarrow a + b$

⇒ But  $E[X_n]$  need not converge to  $a$ .

★ Comparing  $E[g(x)]$  to  $g(E[X])$   
(Jensen's Inequality)

⇒ Let  $g$  be convex

⇒ If  $0 \leq p \leq 1$ , then {definition of convex function}

$$g(px + (1-p)y) \leq pg(x) + (1-p)g(y)$$

$$g(E[X]) \leq E[g(X)]$$

★ Hoeffding's Inequality for  $P(X_1 + \dots + X_n > na)$

→  $X_i$  : iid

$$P(X_1 + \dots + X_n \geq na) \leq e^{-\frac{na^2}{2}}$$

