

Math 3235

Probability Theory  
6/25/23

$$X_i \quad i=1, \dots, n$$

i.i.d.

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$$

$$\bar{X}_n \xrightarrow{P} E(X) = \mu$$

$$\lim_{n \rightarrow \infty} \mathbb{P}(|\bar{X}_n - \mu| > \delta) = 0 \quad \forall \delta$$

$$E(\bar{X}_n) = \mu$$

$$\mathbb{P}(|\bar{X}_n - \mu| > \delta) \leq \frac{\text{Var}(\bar{X}_n)}{\delta^2}$$

$$\text{Var}(\bar{X}_n) = E\left(\left(\frac{1}{n} \sum (X_i - \mu)\right)^2\right)$$

$$\left(\sum_i a_i\right)^2 = \sum_{i,j} a_i a_j$$

$$\rightarrow \frac{1}{N^2} \mathbb{E} \left( \left( \sum_i (X_i - \mu) \right)^2 \right) =$$

Use the formula with  
 $a_i = X_i - \mu$

$$= \frac{1}{N^2} \sum_{i,j} \mathbb{E} \left( (X_i - \mu) (X_j - \mu) \right) =$$

if  $i \neq j$

$$\mathbb{E} \left( (X_i - \mu) (X_j - \mu) \right) =$$

$$\mathbb{E} (X_i - \mu) \mathbb{E} (X_j - \mu) = 0$$

$$\mathbb{E} (X_i X_j) = \mathbb{E} (X_i) \mathbb{E} (X_j)$$

$$\mathbb{E} (X_i - \mu) = 0$$

$$= \frac{1}{N^2} \sum_i \text{Var} (X_i)$$

$$\text{Var} (\bar{X}) = \frac{\text{Var} (X_i)}{N}$$

Thus

$$\mathbb{P} \left( |\bar{X}_n - \mu| > \delta \right) \leq \frac{\text{Var}(X_1)}{N \delta^2}$$

All we used is that

$$\text{cov}(X_i, X_j) =$$

$$\mathbb{E}(X_i X_j) - \mathbb{E}(X_i) \mathbb{E}(X_j) = 0$$

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$$\mathbb{P} \left( |X - \mu| > \delta \right) \leq \frac{\sigma^2}{\delta^2}$$

$$\mu = \mathbb{E}(X) \quad \sigma^2 = \text{Var}(X)$$

$$\delta = \sigma t$$

$$\mathbb{P} \left( |X - \mu| > \sigma t \right) \leq \frac{\sigma^2}{\sigma^2 t^2} = \frac{1}{t^2}$$

$$M_X(t) = \mathbb{E}(e^{tX})$$

$$M_{X+Y}(t) = M_X(t) M_Y(t) \quad X \perp Y$$

$$M_X(0) = \mathbb{E}(1) = 1$$

$$M_X(t) = \int_{-\infty}^{\infty} e^{tx} f(x) dx$$



$$X_1 \quad \dots \quad X_n$$

$$x_1 \quad \dots \quad x_n$$

$$0, 1, 0, 1, 0, 1$$

$$\bar{X}_n = \frac{1}{N} \sum_{i=1}^n X_i \quad \xrightarrow{p} p$$

$$\hat{p} = \bar{X}_n \quad \text{estimator}$$

I want  $N$  to be large enough such that

$$P(|\hat{p} - p| \leq 0.005) = 0.95$$

$$P\left(\left|\frac{\sum X_i}{N} - p\right| \leq 0.005\right)$$

If True prob is  $p$

$$E(X_i) = p \quad \text{Var}(X_i) = p(1-p)$$

$$\frac{\sum X_i - Np}{\sqrt{N(p(1-p))}} \approx N(0,1)$$

$$P\left(-0.005 \leq \frac{\sum X_i - Np}{N} \leq 0.005\right) =$$

$$P\left(\frac{-0.005 \sqrt{N}}{\sqrt{p(1-p)}} \leq \frac{\sum X_i - Np}{\sqrt{N(p(1-p))}} \leq \frac{0.005 \sqrt{N}}{\sqrt{p(1-p)}}\right)$$

Find  $N$  such that

$$P\left(\frac{-0.005\sqrt{N}}{\sqrt{p(1-p)}} \leq Z \leq \frac{0.005\sqrt{N}}{\sqrt{p(1-p)}}\right) > 0.95$$

$$1 - 2\Phi\left(\frac{0.005\sqrt{N}}{\sqrt{p(1-p)}}\right) > 0.95$$

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$N_k$  is geometric r.v. with  
par  $\frac{\lambda}{k}$

$$P(N_k = n) = (1 - p_k)^{n-1} p_k$$

$$Z_k = \frac{N_k}{k} \xrightarrow{d} Y$$

where  $Y$  is exponential with  
par  $\lambda$ .

$$P(Z_k \leq x) = P(N_k \leq kx) =$$

$$\sum_{i=1}^{\lfloor kx \rfloor} (1 - p_k)^{i-1} p_k = 1 - (1 - p_k)^{\lfloor kx \rfloor} =$$

$$\lfloor x \rfloor = \text{floor of } x$$

largest integer smaller than  $x$ .

$$= 1 - \left(1 - \frac{\lambda}{k}\right)^{\lfloor kx \rfloor} \xrightarrow{k \rightarrow \infty} 1 - e^{-\lambda x} \quad \text{Q.E.D.}$$