

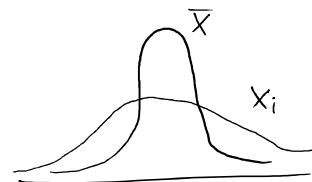
Lec 11/28

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Recall: If X_1, X_2, \dots, X_n are a random sample from a population w/ mean μ and variance σ^2 then

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

$$\text{Var}(\bar{X}) = \frac{\text{Var}(X_i)}{n} = \frac{\sigma^2}{n}$$



If $X_i \stackrel{iid}{\sim} N(\mu, \sigma^2)$ then $\bar{X} \sim N(\mu, \frac{\sigma^2}{n})$

for any Dist.

As $n \rightarrow \infty$, $\text{Var}(\bar{X}) = \frac{\sigma^2}{n} \rightarrow 0$

As $n \rightarrow \infty$ $\bar{X}_n \xrightarrow{P} \mu$

Let $\epsilon > 0$. Then $\lim_{n \rightarrow \infty} P(|\bar{X}_n - \mu| > \epsilon)$ (Chebyshev's Ineq: $P(|X - \mu| > k\sigma) \leq \frac{1}{k^2}$)

$$= \lim_{n \rightarrow \infty} \left(P(|\bar{X}_n - \mu| > k\sigma) \leq \frac{\sigma^2}{\epsilon^2 n} \right) = 0$$

let $\epsilon = \frac{k\sigma}{\sqrt{n}}$
 $\Rightarrow k = \frac{\epsilon\sqrt{n}}{\sigma}$
 $k^2 = \frac{\epsilon^2 n}{\sigma^2}$

$\rightarrow 0$.

\bar{X} is consistent for μ (as $n \rightarrow \infty$, it converges in probability to the parameter).

For large n , $\bar{X} \approx N(\mu, \frac{\sigma^2}{n})$ (Central Limit Theorem). (p 234-235).

as $n \rightarrow \infty$, $\bar{X} \approx N(\mu, \frac{\sigma^2}{n})$

for any ϵ , $\int_{\mathbb{R}} |f_{\bar{X}_n}(x) - f_{\mu}(x)| dx < \epsilon$ for $n >$ some M .

V. ind r (o)

$$X_i \stackrel{\text{i.i.d.}}{\sim} \text{Gamma}(\alpha, \beta)$$

$$\sum_{i=1}^n X_i \sim \text{Gamma}(n\alpha, \beta)$$

$$\bar{X} \sim \text{something}$$