

457

①

5-31

Recall

$$M_x(t) = E[e^{tX}]$$

Let X_1, X_2, \dots, X_n be independent random variables, with unknown distribution, with mean μ and standard deviation σ .

$$\text{Let } Z_i = \frac{X_i - \mu}{\sigma}$$

$$\text{So } E[Z_i] = 0 \text{ and } \text{Var}[Z_i] = 1$$

$$\text{Let } Y = \frac{1}{\sqrt{n}} \sum_{i=1}^n Z_i$$

$$M_Y(t) = M_{\frac{1}{\sqrt{n}} \sum Z_i}(t) = M_{\sum Z_i}\left(\frac{t}{\sqrt{n}}\right) \quad (2)$$

$$= M_{Z_1}\left(\frac{t}{\sqrt{n}}\right) M_{Z_2}\left(\frac{t}{\sqrt{n}}\right) \cdots M_{Z_n}\left(\frac{t}{\sqrt{n}}\right)$$

$$= [M_{Z_1}\left(\frac{t}{\sqrt{n}}\right)]^n$$

$$= \left(E\left[e^{\frac{t}{\sqrt{n}} Z_1}\right]\right)^n$$

$$= \left(E\left[1 + \frac{t}{\sqrt{n}} Z_1 + \frac{t^2 Z_1^2}{2n} + O\left(\frac{1}{n^{3/2}}\right)\right]\right)^n$$

$$= \left[1 + 0 + \frac{t^2}{2n} + O\left(\frac{1}{n^{3/2}}\right)\right]^n$$

$$M_Y(t) = \left[1 + \frac{t^2}{2n} + O\left(\frac{1}{n^{3/2}}\right)\right]^n$$

Find the limit as $n \rightarrow \infty$ (3)

$$\begin{aligned} \ln M_Y(t) &= n \ln \left[1 + \frac{t^2}{2n} + O\left(\frac{1}{n^{3/2}}\right) \right] \\ &= \frac{\ln \left[1 + \frac{t^2}{2n} + O\left(\frac{1}{n^{3/2}}\right) \right]}{\frac{1}{n}} \quad \left(\frac{0}{0}\right) \end{aligned}$$

$$\lim_{n \rightarrow \infty} \ln M_Y(t) =$$

$$\begin{aligned} &\lim_{n \rightarrow \infty} \frac{1}{1 + \frac{t^2}{2n} + O\left(\frac{1}{n^{3/2}}\right)} \cdot \left[\frac{-t^2}{2n^2} + O\left(\frac{1}{n^{3/2}}\right) \right] \\ &= \lim_{n \rightarrow \infty} \frac{1}{1 + \frac{t^2}{2n} + O\left(\frac{1}{n^{3/2}}\right)} \left(\frac{t^2}{2} + O\left(\frac{1}{\sqrt{n}}\right) \right) \end{aligned}$$

$$= \frac{t^2}{2} \quad (4)$$

$$\text{So } \lim_{n \rightarrow \infty} M_Y(t) = e^{\frac{t^2}{2}}$$

This is the m.g.f. of the standard normal distribution.

$$Y = \frac{1}{\sqrt{n}} \sum_{i=1}^n Z_i$$

$$= \frac{1}{\sqrt{n}} \sum_{i=1}^n \frac{X_i - \mu}{\sigma}$$

$$= \frac{1}{\sigma\sqrt{n}} \left(\sum_{i=1}^n X_i - \sum_{i=1}^n \mu \right)$$

$$= \frac{1}{\sigma\sqrt{n}} (n\bar{X} - n\mu)$$

$$= \frac{n}{\sigma\sqrt{n}} (\bar{X} - \mu)$$

$$Y = \frac{\bar{X} - \mu}{(\sigma/\sqrt{n})}$$

Regardless of the original distribution of X_1, \dots, X_n ,

Y has a distribution that approaches the standard normal

as $n \rightarrow \infty$

Central Limit Theorem

⑤

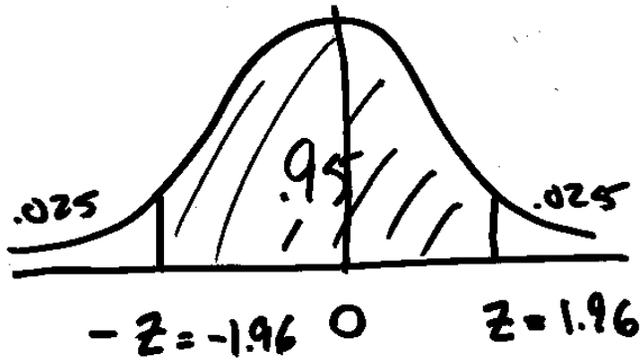
Regardless of the original distribution of X_1, \dots, X_n ,

\bar{X} will have an approximate normal distribution with mean μ and standard deviation $\frac{\sigma}{\sqrt{n}}$,

provided that n is large.

⑥

Goal: Use this theorem to estimate a population mean and give a margin of error.



(7)

$$P(-1.96 < Z < 1.96) = .95$$

$$P\left(-1.96 < \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} < 1.96\right) \approx .95$$

$$-1.96 < \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} < 1.96 \quad \text{true with 95\% Confidence}$$

(8)

$$-1.96 \frac{\sigma}{\sqrt{n}} < \bar{x} - \mu < 1.96 \frac{\sigma}{\sqrt{n}}$$

$$-\bar{x} - 1.96 \frac{\sigma}{\sqrt{n}} < -\mu < -\bar{x} + 1.96 \frac{\sigma}{\sqrt{n}}$$

$$\bar{x} + 1.96 \frac{\sigma}{\sqrt{n}} > \mu > \bar{x} - 1.96 \frac{\sigma}{\sqrt{n}}$$

With 95% confidence, we can state that the population mean, μ , will fall within $\bar{x} \pm 1.96 \frac{\sigma}{\sqrt{n}}$

↑
point estimate

↑
margin of error

Problem: σ is usually unknown. ⑨
What happens if s is used in
its place?

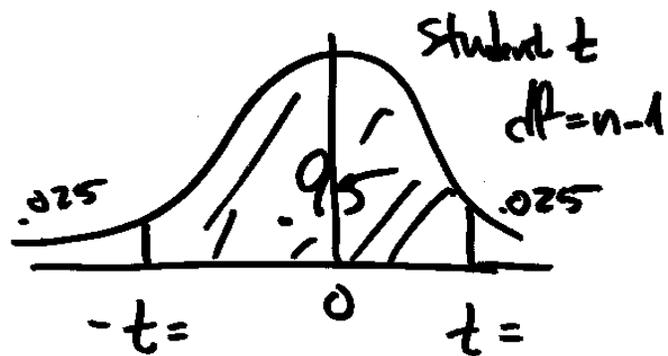
$$\frac{\bar{x} - \mu}{s/\sqrt{n}} = \frac{\left[\frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \right] \leftarrow N(0,1)}{\frac{s}{\sigma} \leftarrow ?}$$

Facts: ① The random variables \bar{x}
and s are independent

② $\frac{(n-1)s^2}{\sigma^2}$ has a χ^2 distribution
with $n-1$ d.f.

③ $\frac{Z}{\sqrt{\frac{\chi^2}{df}}}$ has a t ⑩
distribution

$$\frac{\frac{\bar{x} - \mu}{\sigma/\sqrt{n}}}{\sqrt{\frac{(n-1)s^2}{\sigma^2} \frac{1}{n-1}}} = \frac{\frac{\bar{x} - \mu}{\sigma/\sqrt{n}}}{\frac{s}{\sigma}}$$
$$= \frac{\bar{x} - \mu}{s/\sqrt{n}}$$



⑪

$$\bar{x} \pm t \frac{s}{\sqrt{n}}$$

HW #9 p. 226 # 18
p. 252 # 24, 26

Stat 451 HW #8 p. 193 # 24, p. 205 # 42, 46, 52

24. Bino ($n=400, p=.5$) $\mu=np=200$ $\sigma=\sqrt{npq}=\sqrt{100}=10$

$$\begin{aligned} \text{a) } P(185 \leq x \leq 210) &= P(184.5 \leq x \leq 210.5) \\ &= P(-1.55 \leq Z \leq 1.05) = .8531 - .0606 = \underline{.7925} \end{aligned}$$

$$\begin{aligned} \text{b) } P(X=205) &= P(204.5 \leq x \leq 205.5) \\ &= P(.45 \leq Z \leq .55) = .7088 - .6736 = \underline{.0352} \end{aligned}$$

$$\begin{aligned} \text{c) } P(X < 176) + P(X > 227) \\ &= P(X \leq 175) + P(X \geq 228) \\ &= P(X \leq 175.5) + P(X \geq 227.5) \\ &= P(Z \leq -2.45) + P(X \geq 227.5) \\ &= .0071 + [1 - .9970] = \underline{.0101} \end{aligned}$$

42. $T \sim \text{Gamma}(\alpha=2, \beta=\frac{1}{2})$

$$\lambda = \frac{1}{\beta} = 2$$

$$\begin{aligned} \text{a) } P(T > 1) &= P(X \leq 1), \text{ where } X \sim \text{Poisson}(\mu = \lambda t = 2) \\ &= p(0) + p(1) = e^{-2} \left(\frac{2^0}{0!} + \frac{2^1}{1!} \right) = 3e^{-2} = .4060 \end{aligned}$$

$$\text{so } P(T \leq 1) = 1 - .4060 = \underline{.5940}$$

$$\begin{aligned} \text{b) } P(T \geq 2) &= P(X \leq 1), \text{ where } X \sim \text{Poisson}(\mu = 4) \\ &= p(0) + p(1) = e^{-4} \left(\frac{4^0}{0!} + \frac{4^1}{1!} \right) = 5e^{-4} = \underline{.0916} \end{aligned}$$

46. For a single switch, $T \sim \text{Exp}(\beta=2)$

$$P(T < 1) = F(1) = 1 - e^{-1/2}$$

$$= .3935$$

Then use $\text{Bin}(n=100, p=.3935)$

+ find $P(X \leq 30)$ Answer from Excel: .033471

$$[= \text{BINOMDIST}(30, 100, 1 - \text{EXP}(-.5), 1)]$$

52. $\mu = \alpha\beta = 10$ $\sigma = \sqrt{\alpha\beta^2} = \sqrt{2} \beta = \sqrt{50}$

$$\Rightarrow \alpha = 2, \beta = 5$$

a) $P(T > 50) = P(X \leq 1)$ $X \sim \text{Poisson}(\lambda t = \frac{1}{5} \cdot 50 = 10)$

$$= e^{-10} \left(\frac{10^0}{0!} + \frac{10^1}{1!} \right) = 11e^{-10} = .000499$$

So $P(T \leq 50) = 1 - P(T > 50) = \underline{.9995}$

b) $P(T > 10) = P(X \leq 1)$ $X \sim \text{Poisson}(\lambda t = \frac{1}{5} \cdot 10 = 2)$

$$= e^{-2} \left(\frac{2^0}{0!} + \frac{2^1}{1!} \right) = 3e^{-2} = .4060$$

So $P(T > 10) = 1 - .4060 = \underline{.5940}$