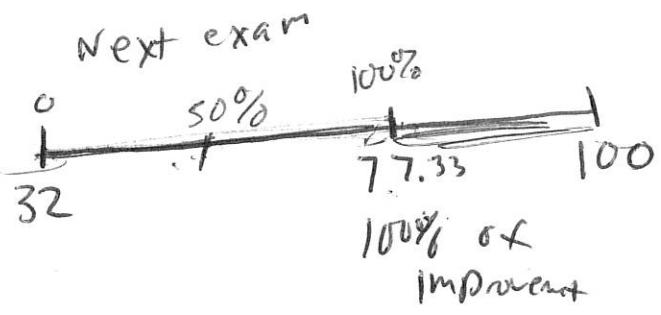


Stat 381 4/3/2015

1st exam score 32

2x



improve by 25%

(37.5%)

$X = \#$ length time until next birth

$$P(X \leq 1) = \int_0^1 \frac{1}{6} e^{-x/6} dx = 1 - e^{-1/6}$$

Let $Y = \#$ babies in 1 hour

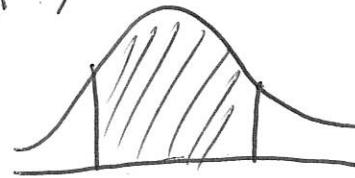
$$Y \sim \text{Poisson}\left(\frac{1}{6}\right) \quad P(Y \geq 1) = 1 - P(Y=0)$$

$$1 - e^{-1/6}$$

$$\bar{X} \sim N\left(\mu=50, \sigma_{\bar{x}}=\frac{\sigma}{\sqrt{n}}\right)$$

$$\bar{X} \sim N(50, \sigma_{\bar{x}}=5)$$

$$\text{normalcdf}(44.58, 58, 50, 5)$$



$$T \sim \text{Poisson}(200) \quad H = \frac{T}{40} \quad 40H = T$$

$$P(H > 5.5) = P\left(\frac{T}{40} > 5.5\right) = P(T > 5.5 \cdot 40)$$

$$P(T > 220)$$

$$= 1 - P(T \leq 220)$$

$$1 - \text{poissondcf}(200, 220)$$

~~F has $\mu = 5$~~ X # in 1 hour
 $\mu_X = 5$ $\sigma_X^2 = 5$
 $\Rightarrow \sigma_X = \sqrt{5}$

$H = \bar{X}$ has $\mu_{\bar{X}} = 5$.

$$\sigma_{\bar{X}} = \frac{\sqrt{5}}{\sqrt{40}}$$

Chapter 9 - Statistical Inference

use sample data to infer properties of the population.

1) estimation (intervals)

2) Hypothesis testing / Decision Making

Estimation

def point estimate of population parameter θ is a single value $\hat{\theta}$ (theta hat) of \hat{H} the statistic.

ex) \bar{X} is a value of \bar{X} estimating μ
 \hat{P} is a value of \hat{P} estimating P

def a point estimate of statistic $\hat{\theta}$ is unbiased estimator of θ if $E(\hat{\theta}) = \theta$.

to estimate μ X_1, \dots, X_n iid population

$$\circ X_1 \quad E(X_1) = \mu$$

$$\therefore \frac{X_1 + X_n}{2} \quad E\left(\frac{X_1 + X_n}{2}\right) = \frac{1}{2}(E(X_1) + E(X_n))$$

$$\circ \bar{X} = E(\bar{X}) = \mu = \frac{1}{2}(\mu + \mu) = \mu$$

biased $\max(X_1, \dots, X_n)$

Given $\hat{\theta}_1$ and $\hat{\theta}_2$ unbiased for θ ,

the "better" statistic is the one with the smaller variance. ("Efficiency")

def If we consider all possible unbiased estimators of θ , the one with the least variance is called the best, or most efficient estimator of θ .

$$\sigma_{\bar{X}}^2 = \frac{\sigma^2}{n}$$

is it possible to have B
 σ_B^2 $E(B) = \mu$
and $\sigma_B^2 = \frac{\sigma^2}{n+1}$

def If as sample size increases (to infinity) if the estimator converges (in some sense) to θ , we say $\hat{\theta}$ is consistent.