

May 8, 2017

This time: prob. models for means

Next time: statistical inference

Draper extra office hour Wed. 1:15-2:15 pm Jack's Lounge

$$P(\text{coming out ahead}) = P(S' > \$0) = 39\%$$

(B) $n=1000$

split

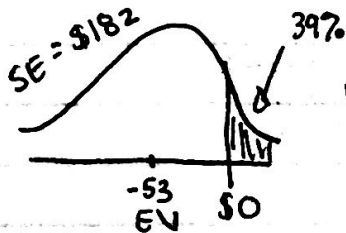
125

$n=35$ single #

$$E(S') = 35(-0.05) = \$1.84$$

$$\mu = (\$+17) + (\$+17) + 36(\$-1) = \$-\frac{2}{38}$$

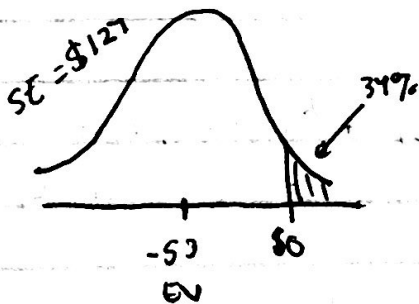
$$SE(S') = \sigma\sqrt{n} = \$4.02\sqrt{1000} = \$127$$



long run histogram

single #

$n=1,000$



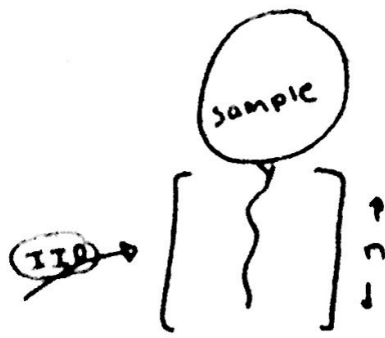
ditto split

Q: which strategy is better?

A: Depending on your degree of risk aversion or risk seeking

(B)

(A)



sum S $n\bar{y}$
 mean $\bar{y} = \frac{S}{n}$

p. R-(53)
 CLT

examples

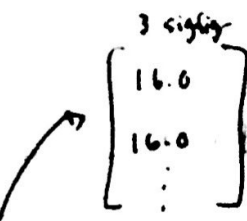
R-55 hypocalcemia
 case study

L-(127)

(prob. model for mean) + (measurement error)

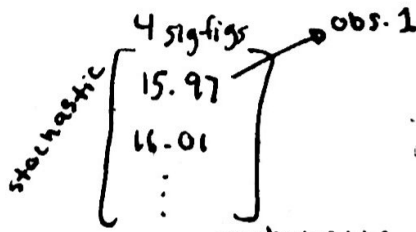


deterministic:



no measurement error

you always get the same answer



probabilistic
 probabilistic:
 meas. error

basic measurement

error model: ^{obs. n}

$$\begin{aligned} (\text{observation } 1) &= (\text{true value}) + (\text{bias}) + (\text{"random error 1"}) \\ (\text{obs. } 2) &= (\text{true value}) + (\text{bias}) + (\text{random error } 2) \\ &\vdots \end{aligned}$$

$$(\text{obs. } n) = (\text{true value}) + (\text{bias}) + (\text{random error } n)$$

$$\begin{bmatrix} y_i \end{bmatrix} = \begin{bmatrix} \theta \end{bmatrix} + \begin{bmatrix} b \end{bmatrix} + \begin{bmatrix} e_i \end{bmatrix}$$