

This time: measurement error; prob. models for means
Next time: statistical models for means; interval estimation

Read: DD (B) ch.11 LN: pp. L-(137)-(156) This time: LN pp. L-(127) →

3.9 is more useful than # is > 3.5

weighing butter:

16 oz
16 oz
16 oz
⋮

16.0
16.0
16.0
⋮

16.01
15.98
⋮
15.99

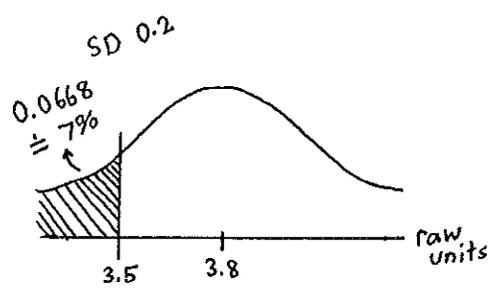
↖ deterministic ↗
probabilistic
(same value everytime)
(stochastic)
(different value everytime)

Basic Model

$$\begin{aligned}
 (\text{obs. } 1) \quad y_1 &= (\text{true value}) + (\text{bias}) + (\text{random error } 1) \\
 (\text{obs. } 2) \quad y_2 &= (\text{true value}) + (\text{bias}) + (\text{random error } 2) \\
 &\vdots
 \end{aligned}$$

mean 0
 Independent
 IID

chart on other page



pop. hist.

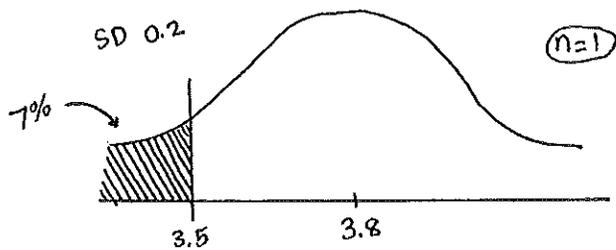
(1 reading at a time)

saying you're hypo. when not true

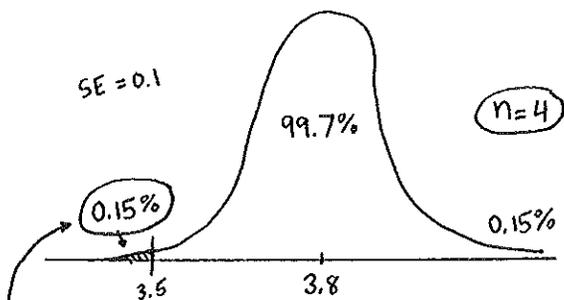
$$P(\text{misclassification}) = P(y_1 > 3.5)$$

SV

$$\frac{3.5 - 3.8}{0.2} = -\frac{0.3}{0.2} = -1.50 \quad \text{check L-34/35}$$



long run hist. of \bar{y}



← less spread out

$$\frac{3.5 - 3.8}{0.1} = -3$$

$P(\text{misdiagnosis w/n} = 4) = 0.15\%$

n	error rate	cost
1	7%	\$25
4	0.15%	\$100

(big n leads to larger cost)

↑ benefit (small error rate with big n)

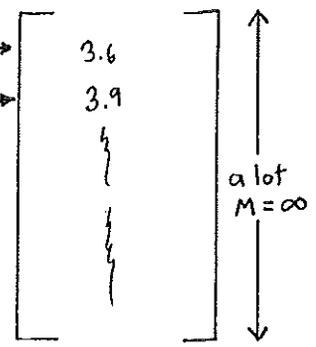
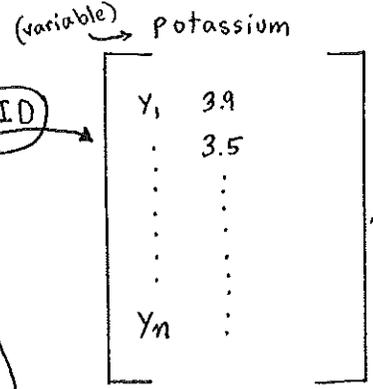
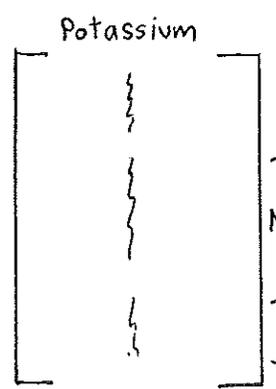
Would you pay \$25 for a 7% chance to eat more bananas than you needed to?

Conceptual

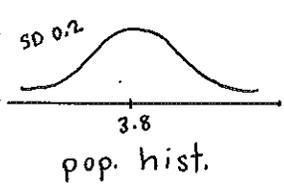
Step 2 pop data set
All possible blood sample readings

Step 1 Sample data set
the observed blood samples

Step 3 imag. data set
possible \bar{y} s

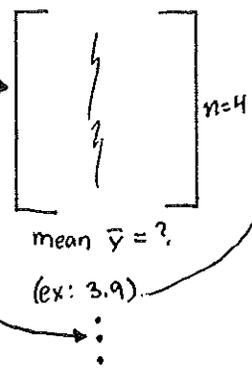


mean $\mu = 3.8$
SD $\sigma = 0.2$



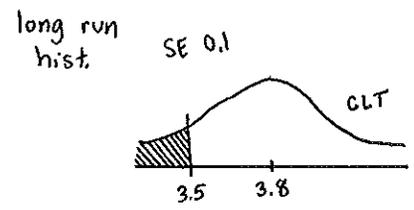
IID
 $N = \infty$
Imagine

mean $\bar{y} = ?$
(ex: 3.6)



long run mean $\text{EV of } \bar{y} = 3.8$

long run SD $\text{SE of } \bar{y} = 0.1$



- * expected value of \bar{y}
- = EV of \bar{y}
- = $E(\bar{y})$
- = $E_{\text{IID}}(\bar{y}) = ?$

math fact: $E_{\text{IID}}(\bar{y}) = \mu$

- ** long run SD of \bar{y} = standard error of \bar{y}
- = SE of \bar{y} = $SE(\bar{y}) = SE_{\text{IID}}(\bar{y}) = ?$

N	no
μ	no
σ	$SE(\bar{y}) \uparrow$
n	$SE(\bar{y}) \downarrow$

math fact: $SE_{\text{IID}}(\bar{y}) = \frac{\sigma}{\sqrt{n}}$

Square root law: to cut the SE in half, you have to quadruple the sample size.

here $SE(\bar{y}) = \frac{0.2}{\sqrt{4}} = 0.1$