

$$\begin{aligned}
 Y_1 &= \theta + b + e_1 \quad \leftarrow \text{IID with mean 0} \\
 Y_2 &= \theta + b + e_2 \\
 &\vdots \\
 Y_n &= \theta + b + e_n \\
 \hline
 \bar{Y} &= \theta + b + \bar{e}
 \end{aligned}$$

* \bar{e} will (with high prob.) be closer to 0 than any of the e_i (mean of n IID draws, each with mean 0) ex. $\frac{(+0.03) + (-0.01) + \dots + (-0.05)}{n}$

* as $n \uparrow$, $\bar{e} \downarrow 0$ with high prob.; therefore, \bar{Y} will be close (when n is large) to $(\theta + b)$
(truth + bias)

* to make \bar{Y} get arbitrarily close to θ , we need 2 things:

- ① n should get \uparrow
- ② $b = 0$ (measuring process is unbiased)

* you cannot make bias $\downarrow 0$ as n increases

1936	Literary Digest
Roosevelt (D)	- 24,000 letters (pre-stamped postcards)
Landon (R)	- got back 16,000 letters
	- LD est. that $\left(\begin{array}{l} \text{Landon } 60\% \\ \text{Roosevelt } 40\% \end{array} \right)$

truth:

$\left(\begin{array}{l} \text{Roosevelt } 60\% \\ \text{Landon } 40\% \end{array} \right) \rightarrow 20\text{-percentage point error}$

George Gallup

$\left(\begin{array}{l} \text{Iowa} \\ \text{state} \end{array} \right) \left(\begin{array}{l} 1,000 \\ \text{people} \end{array} \right)$

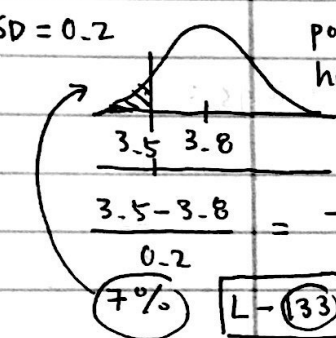
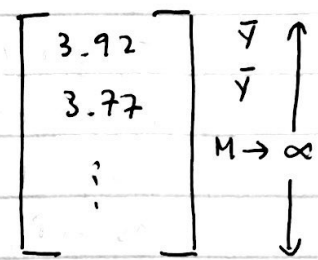
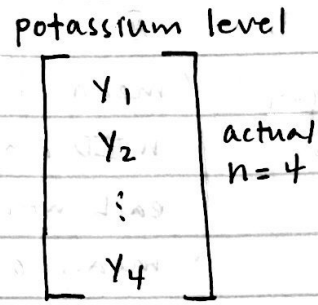
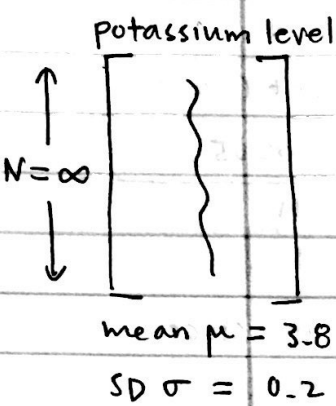
addresses:

- landowner records
 - telephone books
 - club membership
- } rich

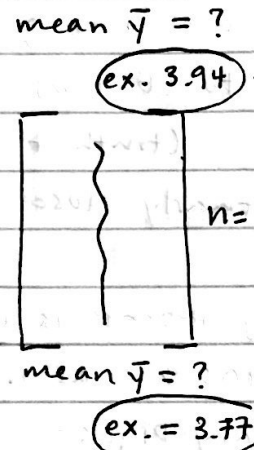
population (conceptual)
all possible measurements

sample the observed measurements

imaginary dataset



like IID
hyp IID



long run mean } expected value of \bar{y}
long run SD } standard error of \bar{y}
long run hist. }
3.8
EV

$P(\text{misdiagnosis with } n=1) \approx 7\% \leftarrow \text{error rate too high}$
 $P(\text{misdiagnosis with } n=4) = P(\bar{y}(n=4) < 3.5) = ?$

(expected value of \bar{y}) = (EV of \bar{y}) = $E_{\text{IID}}(\bar{y}) = \mu$

(standard error of \bar{y}) = (SE of \bar{y}) = $SE_{\text{IID}}(\bar{y}) = \frac{\sigma}{\sqrt{n}}$

ingredient	belong?
N	X
μ	X
σ	$\sigma \uparrow SE(\bar{y}) \uparrow$
n	$n \uparrow SE(\bar{y}) \downarrow$

Square Root Law:

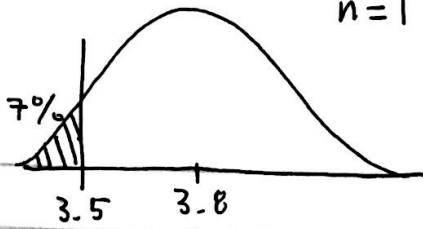
- uncertainty goes down with n, but only at a \sqrt{n} rate
- to cut $SE(\bar{y})$ in half, need to quadruple n

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

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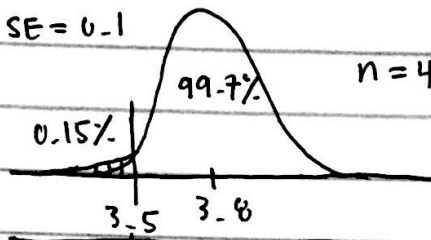
SE = 0.2 (CLT)

n = 1



SE = 0.1

n = 4



$$\frac{3.5 - 3.8}{0.1} = -\frac{0.3}{0.1}$$

n	P(misdiagnosis)	cost
1	7%	\$25
4	0.15%	\$100

↙ ↘
cost - benefit tradeoff