

this time: inference for p

read:

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LN pp. L-(161) \rightarrow L-(185)

AMS 7

next time: hypothesis testing

today:

10 May 18

LN pp. L-(157) \rightarrow

- when μ_0 is not in _____ interval,
 - 95%; diff. between theory (μ_0) & data (\bar{y}) is statistically significant (statsig)
 - 99%; diff. is highly statsig
- completely diff. question:
 - is diff. between μ_0 & \bar{y} practically significant (practsig)?

We should ask:

① Practsig?

- if the diff. is too small, then stop here... otherwise,

② statsig?

Q: Is a diff. that's statsig real?

A: Opposite of real \rightarrow not real

- easy to attribute to unlucky random sampling

Devil's Advocate: don't quickly assume that the theory is wrong because the diff. is statsig

$\hookrightarrow \mu_0$ really is 24.3; the reason why the data (\bar{y}) is so different is b/c the sample size is so small & may be affected by unlucky random sampling

* always take Devil's Advocate's POV seriously (logical possibility)

* Statistics = BS filter

* statsig (real) = hard to attribute to unlucky sampling

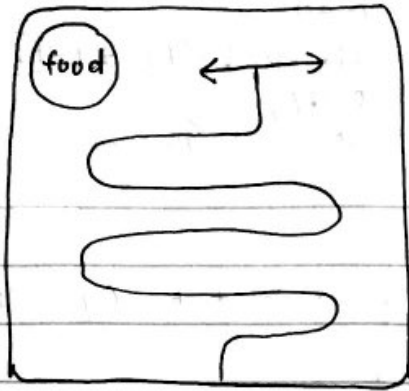
Q: Does confidence = probability?

$P_f(24.5^\circ\text{C} < \mu < 25.6^\circ\text{C}) = 95\%$?

A: In relative frequency version of probability - no; μ is a fixed unknown constant that doesn't change across repetitions.

Q: Okie then, what does the interval actually mean?

A: Across many repetitions of the whole process of getting random samples & building CIs, about 95% of these intervals should correctly include μ_0 .



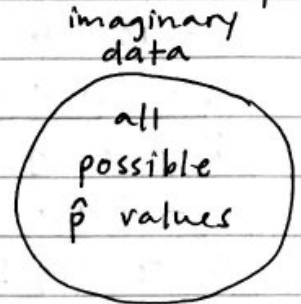
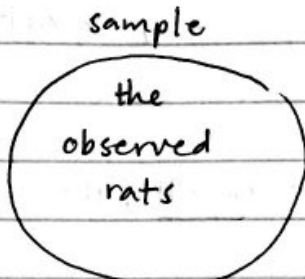
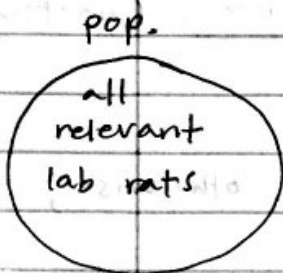
Lab rats

10 out of 12 rats turned left toward food

→ **83%** → practsig
→ calculate to determine if statsig

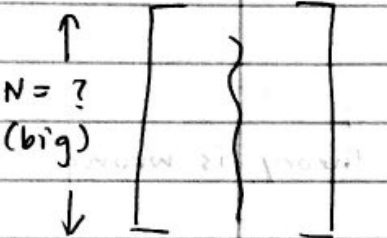
Devil's Advocate: You're supposed to get 50%, but got 83% instead due to unlucky random sampling.

1 = L
0 = R

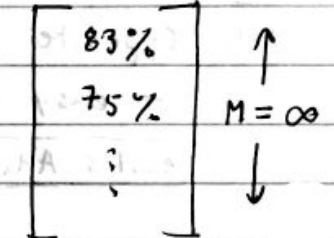
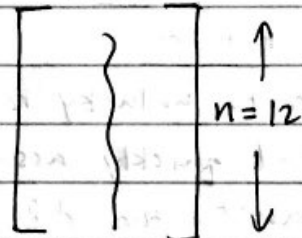


left?

left?



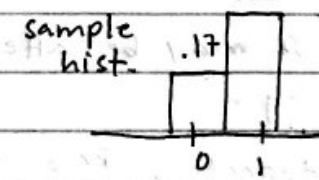
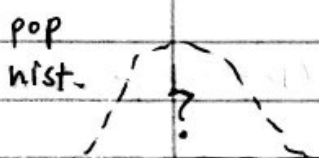
actual like SRS (IID)



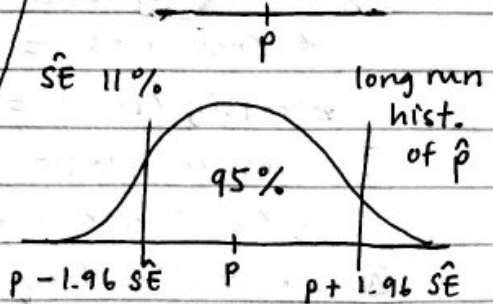
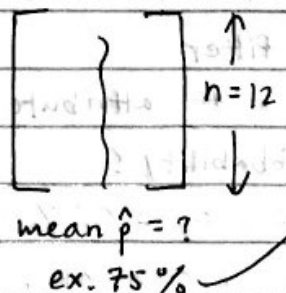
* mean $\mu = p = ?$
SD $\sigma = ?$
 $= \sqrt{p(1-p)}$

* mean $\bar{y} = \hat{p} = \frac{10}{12} = .83 = 83\%$

EV of $\hat{p} = p$
* \hat{SE} of $\hat{p} \doteq 11\%$



SE 11% long run hist.



* = inferential summary

σ (in disguise) → $SE_{IID}(\hat{p}) = \sqrt{\frac{p(1-p)}{n}} = \sqrt{\frac{(0.83)(0.17)}{12}} = 0.108 \doteq 11\%$

* don't agree w/ devil's advocate; diff. is **statsig**

• large-sample approximation: pretend n is large enough to use CLT

95% CI for p: $\hat{p} \pm 1.96 \hat{SE}(\hat{p}) = \hat{p} \pm 1.96 \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$

