

STT 207 10-26-09

SUMMARY OF CI INFORMATION

SETUP
 WITH REPL n_x
 (EQUAL PROBABILITY)

PARAM
 Pop MEAN
 μ_x

POINT ESTIMATOR
 \bar{x} (RULE)
 (MOST SURELY WRONG)

CI (\approx) $(1-\alpha)$ CLAIM
 $\bar{x} \pm 1.96 \frac{s_x}{\sqrt{n_x}}$

CI \equiv WIDE BODIED GUESS

"
 "

p_x

eg FRACTION
 OF INDIVIDUALS
 IN POPULATION
 HAVING SOME
 PARTICULAR
 CHARACTERISTIC

POINT ESTIMATOR
 \hat{p}_x (or \bar{p}_x)
 SAMPLE PROPORTION

$\hat{p}_x \pm 1.96 \sqrt{\hat{p}_x(1-\hat{p}_x)}$

SCORING

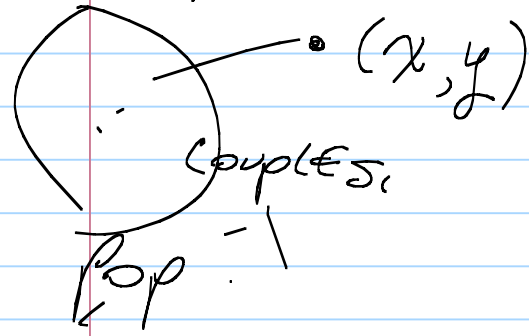
1 PENCIL USER
 0 OTHER

$\hat{p}_x = \bar{x}$

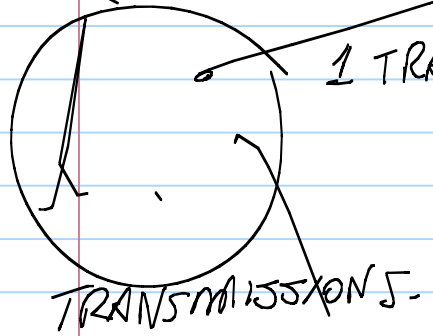
NOTE:
 $\sqrt{\hat{p}_x(1-\hat{p}_x)}$

$= \sqrt{\sum_{i=1}^n (x_i - \bar{x})^2 / n}$

Setup:



OR



$x = \text{viscosity FLUID}$
1 TRANS.
 $y = \text{TEMP. FLUID}$

Setup is SPECIAL.
KNOW M_x

PARAM

M_y

POINT EST

~~y~~

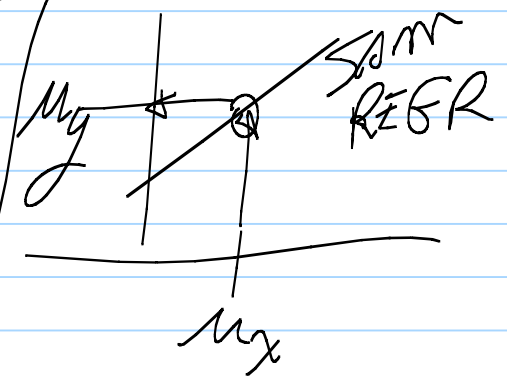
REGR-BASED ESTIMATOR

OF M_y
THINKING

\Rightarrow REGR ESTIMATOR OF M_y
IS $\boxed{\bar{y} + (M_x - \bar{x}) R \frac{dy}{dx}}$

$$\frac{RSE}{RUN} = \frac{y - \bar{y}}{x - \bar{x}} = R \frac{dy}{dx}$$

(x, y) ON REGR LINE IF AND ONLY IF



SO REGR BASED EST OF μ_y (CONT)

ESTIMATOR $\bar{y} + (M_x - \bar{x}) R \frac{dy}{dx}$

MODIFIES \bar{y} TO TAKE
ACCOUNT OF HOW FAR

\bar{x} IS FROM KNOWN M_x .

↖ SIMPLE CORRELATION

JAHEN ENTITLED TO USE 95% Z-BASED CI FOR μ_y

FORM $\left(\bar{y} + (M_x - \bar{x}) R \frac{dy}{dx} \right) \pm 1.96 \frac{dy}{dx} \sqrt{1 - R^2}$

DIFF. OF
POP MEANS

PARAMM ESTIMATOR CI
 $\mu_x - \mu_y$ $\bar{x} - \bar{y}$ $(\bar{x} - \bar{y}) \pm 1.96$

μ_x
 $\sigma_x \Rightarrow$

μ_y
 $\sigma_y \Rightarrow$

x = HT GAIN
ON PLACEBO

y = HT GAIN
ON PROTEIN

MINUS

IF WISH

68% CI $z=1.0$

$$\pm 1.96 \sqrt{\frac{\sigma_x^2}{n_x} + \frac{\sigma_y^2}{n_y}}$$

DIFF OF

POP PROPORTIONS

$p_x - p_y$

$\hat{p}_x - \hat{p}_y$

$\hat{p}_x - \hat{p}_y \pm 1.96$

± 1.96

$\sqrt{\frac{\hat{p}_x(1-\hat{p}_x)}{n_x} + \frac{\hat{p}_y(1-\hat{p}_y)}{n_y}}$

$\hat{p}_x(1-\hat{p}_x)$

$\hat{p}_y(1-\hat{p}_y)$

p_x = POPN FRACTION
W/ FEVER DROPPING
AFTER PLACEBO.

p_y = POPN FRACTION
W/ MED.

NEG

POS.

STAT 200 10-26-09

Note Title

6/2009

Assignment due in recitation 10-27-09.

1-6. Regression-based CI for μ_y when we have data (x_i, y_i) for which μ_x is known. Each of the students in a with-replacement and equal-probability random sample of 20 students from the class is scored x = exam 1 raw score, y = exam 2 raw score. Suppose we know the class mean $\mu_x = 17.3$ for exam 1 and seek to estimate the class mean raw score on exam 2. The following data are from the random sample of 20 students:

$\bar{x} = 16.8$ (lower than the class exam 1 average)
 $\bar{y} = 12.22$ $R =$ sample correlation of $(x, y) = 0.64$
 $s_x = 1.1$ $s_y = 2.73$

We will ignore FPC issues for the present.

1. Point estimate of μ_y ignoring x data.
2. Regression-based point estimator of μ_y ("improved" estimator).
3. Usual z -based 95% CI for μ_y using estimator of (1).
4. 95% z -based CI for μ_y using estimator of (2).
5. Is the regression estimator (2) raised from (1) and if so why?

$\mu_x =$
 KNOWN
 17.3

Obs (x_i, y_i)
 $\bar{x} = 16.8$
 $\bar{y} = 12.22$
 $s_x = 1.1$ $s_y = 2.73$
 $R = 0.64$

$\approx R^2 = .42$ of σ_y^2
 15 EXPLAINED BY REGR
 ON X

POINT ESTIM

6. Is the CI (4) narrower than that of (3) and if so why?

imp ESTOR : $\bar{y} + (\mu_x - \bar{x}) \frac{R s_y}{s_x} = 12.22$

\circledast $12.22 + (17.3 - 16.8) \cdot 0.64 \frac{2.73}{1.1}$

95% CI $\circledast \pm 1.96 \frac{s_y}{\sqrt{n}} \sqrt{1 - R^2}$

7-10. z-based CI for $\mu_x - \mu_y$ when independently sampling each of populations x, y with equal-probability and with-replacement. Illustrated by sampling each of exam 1 score and exam 2 scores separately. Suppose we do not know any of the population parameters and

$\bar{x} = 17.4$
 $\bar{y} = 14.8$
 $n_x = 30$
 $n_y = 40$
 $s_x = 1.9$
 $s_y = 3.8$

$(14.8 - 17.4) \pm 1.96 \sqrt{\frac{1.9^2}{30} + \frac{3.8^2}{40}}$

EST $\mu_x - \mu_y$

μ_x

#7.

Remember, the samples from x are entirely unrelated to ("independent" of) those from y. Unlike the setting of problems 1-7 where we selected students and took their exam 1 and exam 2 scores (paired data) we now sample potentially different students from each of exam 1 scores and exam 2 scores (unpaired data).

7. Give a 95% z-based CI for $\mu_x - \mu_y$ based on the above data.

8. Now that we have estimates of $s_x = 1.9$ and $s_y = 3.8$ what fraction of our 70 observations should we have allocated to population x in order to make the resulting CI narrowest?

9-12. z-based CI for $\mu_x - \mu_y$ when independently sampling each of populations x, y with equal-probability and with-replacement. Illustrated by an example discussed in lecture of 10-21-09.

9. Toss a coin until you see the pattern HH. For example, the sequence TTHH-HTTTT first finds the HH pattern after 4 tosses. Repeat the experiment 20 times, each time recording the score x = number of tosses required to get HH. Calculate sample mean \bar{x} and sample standard deviation s_x .

IF WE MINIMIZE $\sqrt{s_x^2/n_x + s_y^2/n_y}$ FOR $n_x + n_y$ FIXED \implies OPT'L ALLOCATION $f_x = \frac{s_x}{s_x + s_y}$

TWO SAMPLE PROBLEM

Twenty rows each of 22 coin tosses.

MatrixForm[Table[Flatten[Table[c[[Random[Integer, 1] + 1]], {1, 1, 22}]], {3, 1, 20}]]

$x_1 = 10$
 $x_2 = 12$

→

H	T	T	T	T	H	T	T	H	H	T	H	H	H	T	H	T	H	T	H	T
H	T	H	T	H	T	T	H	T	H	H	T	H	T	H	T	T	T	T	H	H
T	T	T	H	T	T	H	H	H	T	H	T	T	H	T	H	H	T	T	H	H
T	T	H	H	H	T	H	H	T	H	H	H	T	H	T	H	T	T	T	T	H
T	T	T	T	H	T	H	H	H	H	H	T	H	H	H	H	T	T	H	H	T
H	H	H	H	T	T	H	T	T	T	H	T	T	H	T	H	T	H	T	H	T
H	T	T	T	T	H	T	H	H	T	T	H	H	H	T	T	T	H	T	T	T
H	H	T	T	H	H	T	T	H	T	T	H	T	H	H	T	H	H	H	H	H
T	T	H	H	H	T	T	T	T	T	T	T	T	T	T	T	H	T	H	T	H
H	T	H	H	T	H	T	H	H	H	H	T	H	T	H	H	T	H	T	H	T
H	H	H	H	T	T	H	T	H	T	H	T	T	H	T	H	H	T	T	H	T
H	T	T	T	H	T	T	H	T	T	T	H	H	T	T	H	H	T	H	H	H
H	T	T	H	H	H	T	H	T	T	T	H	H	H	H	T	H	T	T	T	T
T	H	H	T	H	H	H	H	T	T	T	H	H	H	T	T	T	T	T	T	H
H	H	T	H	H	H	H	H	T	H	T	H	H	H	T	T	T	T	T	H	H
T	H	T	T	T	H	H	H	H	T	T	H	T	H	H	T	T	T	T	H	T
H	H	T	H	T	T	T	H	H	T	T	T	T	H	T	H	T	H	H	T	H
H	H	T	H	T	T	H	T	H	H	H	T	T	H	H	T	H	H	T	H	T
T	T	T	T	T	H	T	H	H	H	T	T	T	H	H	T	H	T	H	T	H

Toss coin + look for pattern HH.

\bar{x} for $n_x = 20 \sim M_x?$

I find $\bar{x} = 4.8$ $s_x = 3.9$ $n_x = 20$

CI for M_x : $4.8 \pm 1.96 \frac{3.9}{\sqrt{20}}$

$n_y = 30$ "HT"

Thirty rows each of 22 coin tosses.

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MatrixForm[Table[Flatten[Table[c[[Random[Integer, 1] + 1]], {i, 1, 22}]], {j, 1, 30}]]
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25

T	H	T	H	H	T	H	T	T	T	H	H	T	T	H	H	T	H	T	T	T	T
H	H	H	H	T	H	T	H	H	T	H	H	H	H	H	T	T	H	H	H	T	T
T	T	H	T	H	H	H	H	H	H	H	T	T	H	T	T	H	T	H	H	T	T
H	T	T	H	T	H	T	T	T	H	H	T	T	H	T	H	H	T	T	T	T	H
H	T	H	H	H	T	H	T	H	T	H	T	H	H	H	T	H	H	T	T	H	T
H	T	H	T	T	T	T	T	T	T	T	T	T	T	T	T	T	T	T	T	T	H
T	T	H	T	T	H	H	H	T	H	H	H	T	H	T	H	H	T	T	T	T	H
H	T	T	T	H	H	T	H	H	H	H	T	H	T	H	H	H	T	T	T	T	H
T	H	T	H	H	H	T	T	T	T	T	H	H	T	T	T	H	H	T	H	T	H
H	H	T	H	H	H	T	T	T	T	T	H	H	T	T	T	H	H	T	H	T	H
H	H	H	T	T	H	H	T	T	T	T	H	T	T	T	H	H	T	H	H	T	H
H	H	T	H	T	T	H	T	T	T	H	T	T	T	H	H	T	H	H	T	H	H
T	T	T	T	T	H	T	T	T	T	T	H	T	T	H	H	T	H	H	T	T	H
T	T	H	T	T	H	H	H	T	H	T	H	H	H	T	T	H	T	H	H	T	H
H	H	H	H	H	H	T	T	H	H	H	H	T	H	T	H	T	T	T	H	H	H
H	T	H	T	H	H	T	H	H	T	H	T	H	H	T	H	T	T	T	H	H	T
H	H	T	T	H	H	T	H	T	T	T	H	T	H	T	T	T	T	T	H	H	T
H	H	T	H	T	H	T	T	H	T	H	H	H	T	H	H	H	T	H	T	H	H
T	H	T	H	H	T	H	T	H	T	H	H	H	T	H	H	H	T	T	T	H	T
T	H	H	T	H	H	H	H	T	H	H	H	T	H	H	H	T	H	T	T	T	H
T	T	T	H	H	H	T	H	T	H	H	H	T	T	T	T	T	H	H	H	T	H

IF WANT
HT +
FAIL.
YOU GET H/H

IF YOU
WANT HH
& FAIL.
YOU GET HT

NOW YOU
NEED WN/H

ACTUALLY,
 $M_x = (HH) 6$
 $M_y(HT) = 4$

$\bar{y}(HT) = \boxed{6}$ $s_y = 1.47$

CI USING p_9 HH FOR ORIGINAL ASSIGNMENT "SOLUTION"
+ p_9 HT FOR EXTRA COINTOSSINGS POSTED.

4. $\mu_x - \mu_y$ CI for 6 $(4.8 - 6) \pm 1.96 \sqrt{\frac{3.9^2}{20} + \frac{1.47^2}{30}}$

≈ 2 IN THEORY

WITH THIS DATA OPT'L $f_x = \frac{3.9}{3.9 + 1.47}$

10. As in (9) but instead look for the pattern HT and repeat 30 times. Calculate sample mean \bar{y} and sample standard deviation s_y .

11. Does it seem to you that the population mean time μ_x to get HH should be the same as the population mean time μ_y to get HT? If they differ which should be the larger (population) mean time? Why?

12. Give a 95% z-based CI for $\mu_x - \mu_y$. Does the CI fall entirely to one side of 0? If so we might take this as some evidence that the population means are not exactly the same, especially if the CI falls far from 0. **Note: In the lecture 10-21-09 we had a REAL FLUKE with a similar example using patterns HTH and HTT when our sample means (obtained from students flipping coins) happened to both EXACTLY equal their theoretical population means for that example (which I know and shared without proof). As you look over the lecture notes do not be confused by the fluke that happened there. Probably ours is the only class ever to have it happen for this example. To flukes!**

Unrelated to above.

13. z-based 95% CI for difference of proportions $p_x - p_y$ when independently sampling populations x, y. See page 561. This will be covered 10-26-09. From a very large population of patients we sample 30 (equal probability with replacement) to receive a placebo. Of these 30 there are 20 who feel better 2 hours later. From the same population we sample 40 (equal probability with replacement) to receive a prescription medication. Of these 40 there are 32 who feel better 2 hours later. The population is so very large that we consider these samples without-replacement. Give a 95% z-based CI for the difference $p_x - p_y$.

CI for $p_x - p_y$

$$\hat{p}_x \ominus \hat{p}_y \pm 1.96 \sqrt{\frac{\hat{p}_x(1-\hat{p}_x)}{n_x} \oplus \frac{\hat{p}_y(1-\hat{p}_y)}{n_y}}$$

$$\frac{20}{30} - \frac{32}{40} \pm 1.96 \sqrt{\frac{20/30 \quad 10/30}{30} \oplus \left(\frac{32}{40} \quad 8/40 \quad /40\right) \frac{600-40}{600-1}}$$

*N_y = 600
5% SE w/o REPL*