## CONFIDENCE INTERVALS FOR VARIANCE COMPONENTS (Section 17.3.5)

1

These play the role for random effects that confidence intervals for contrasts play for fixed effects.

Confidence intervals for  $\sigma^2$ : Constructed just as for fixed effects; see Section 3.4.6 or the class notes *Choosing Sample Sizes*.

## Confidence intervals involving $\sigma_T^2$ :

Three types of are of interest:

- For  $\sigma_T^2$
- For  $\sigma_T^2/\sigma^2$
- For  $\sigma_T^2/(\sigma_T^2 + \sigma^2)$

The first cannot be done exactly, so we'll take that last.

Confidence intervals for  $\sigma_T^2/\sigma^2$ :

We use the fact (see notes *Testing for Treatment Effect as a Proportion of Error Variance*) that

$$\frac{MST/\left(c\sigma_T^2 + \sigma^2\right)}{MSE/\left(\sigma^2\right)} \sim \text{F(v-1, n-v)},$$

where c is a certain constant defined in terms of n,v, and the  $r_i$ ; c = r if the design is balanced) (See notes *Random Effects Models* or Section 17.3)

If we want a  $(1-\alpha)100\%$  CI for  $\sigma_T^2/\sigma^2$ , take

$$f_1 = F(v-1, n-v, 1-\alpha/2)$$
  
(so that there is area  $\alpha/2$  to the left of  $f_1$  in the  $F(v-1, n-v)$  distribution), and

$$f_2 = F(v-1, n-v, \alpha/2)$$
 (so that there is area  $\alpha/2$  to the right of  $f_2$  in the  $F(v-1, n-v)$  distribution).

[Picture!]

Then

Prob 
$$(f_1 \le \frac{\frac{MST}{(c\sigma_T^2 + \sigma^2)}}{\frac{MSE}{(\sigma^2)}} \le f_2) = 1 - \alpha$$

Equivalently:

Prob 
$$(f_1 \le [MST/MSE] [\sigma^2/(c\sigma_T^2 + \sigma^2)] \le f_2)$$
  
= 1-  $\alpha$ ,

Working with the left inequality:

$$(c{\sigma_T}^2\!+\sigma^2)\!/\!\sigma^2\!\leq\!(MST/MSE)(1/f_1)$$

$$c(\sigma_T^2/\sigma^2) + 1 \le (MST/MSE)(1/f_1)$$

$$c(\sigma_T^2/\sigma^2) \le (MST/MSE)(1/f_1) - 1$$

Working with the right inequality: is equivalent to

$$(MST/MSE)(1/f_2) \le (c\sigma_T^2 + \sigma^2)/\sigma^2$$

$$= c(\sigma_T^2/\sigma^2) + 1$$

$$(MST/MSE)(1/f_2) - 1 \le c(\sigma_T^2/\sigma^2)$$

So

3

Prob ((1/c)[ (MST/MSE)(1/f<sub>2</sub>) - 1] 
$$\leq \sigma_T^2/\sigma^2$$
  
 $\leq$  (1/c)[ (MST/MSE)(1/f<sub>1</sub>) - 1]) = 1 -  $\alpha$ .

Thus desired confidence interval has left endpoint

$$(1/c)[(msT/msE)(1/f_2) - 1]$$

and right endpoint

$$(1/c) (msT/msE)(1/f_1) - 1)$$

Here, "(a, b) is a 95% confidence interval for  $\sigma_T^2/\sigma^2$ " Means:

*Note*: Conceivably the left hand endpoint could be less than 0, which is unrealistic. If it is < 0, do *not* give in to the temptation to replace it by zero; that would give the false impression of a smaller confidence interval than warranted.

Example: Use the loom data to find a 95% confidence interval for  $\sigma_T^2/\sigma^2$ .

Confidence intervals for  $\sigma_T^2/(\sigma_T^2 + \sigma^2) = the$  proportion of the total variance if the response attributable to the treatment level:

Such confidence intervals are readily obtained from confidence intervals for  $\sigma_T^2/\sigma^2$  as follows.

Divide both numerator and denominator of  $\sigma_T^2/(\sigma_T^2 + \sigma^2)$  by  $\sigma^2$  to obtain

$$\sigma_{T}^{2}/(\sigma_{T}^{2}+\sigma^{2}) = \frac{\sigma_{T}^{2}/\sigma^{2}}{\left(\sigma_{T}^{2}/\sigma^{2}\right)+1} = f\left(\sigma_{T}^{2}/\sigma^{2}\right),$$
where  $f(x) = x/(x+1) = \frac{1}{1+\frac{1}{x}}$ .

From the last formula for f(x), we can see that f(x) is an increasing function of x. i.e.,  $\sigma_T^2/(\sigma_T^2 + \sigma^2)$  is an increasing function of  $\sigma_T^2/\sigma^2$ .

Thus if (a,b) is a (1- $\alpha$ )100% confidence interval for  $\sigma_T^2/\sigma^2$ , then (f(a), f(b)) = (a/(a + 1), b/(b + 1)) is a (1- $\alpha$ )100% confidence interval for  $\sigma_T^2/(\sigma_T^2 + \sigma^2)$ .

*Note*:  $\sigma_T^2/(\sigma_T^2 + \sigma^2)$  is sometimes called the "population intraclass correlation coefficient"

*Caution*: The phrase "intraclass correlation coefficient" is also used to refer to other things.

6

Example: With the loom data, find a 95% confidence interval for  $\sigma_T^2/(\sigma_T^2 + \sigma^2)$ .

Confidence intervals for  $\sigma_T^2$ : There is no exact method. There are several approximate methods. Here's one. It's useful if  $\sigma_T^2$  is not too small, and is adaptable to more complicated models.

Recall that U = (1/c)(MST - MSE) is an unbiased estimator of  $\sigma_T^2$ . If we knew its distribution, we could use that to get confidence intervals for  $\sigma_T^2$  in the usual way.

However, U does not have a tractable distribution.

But it is true that

U/
$$\sigma_T^2 \approx \chi^2(x)/x$$
, where  

$$x \approx \frac{(msT - msE)^2}{(msT)^2/(v-1)^+ (msE)^2/(n-v)}$$

(Note: This formula is given correctly on p. 605 of the text, but incorrectly on p. 600.)

x from this formula is not usually an integer, so we need to interpret degrees of freedom in the  $\chi^2$  distribution as the parameter in a formula for the pdf. (This is analogous to the two-sample, unequal variance t-test.)

Thus (Draw a picture!)

P(
$$\chi^2(x, 1-\alpha/2) \prec xU/\sigma_T^2 \prec \chi^2(x, \alpha/2)$$
)  $\approx 1-\alpha$ ,

where  $\prec$  means "is less than or approximately equal to", and  $\chi^2(x,\beta)$  is the value with proportion  $\beta$  of the  $\chi^2(x)$  distribution to its *right*.

The left and right approximate inequalities are, respectively, equivalent to

$$\sigma_T^2 \prec xU/\chi^2(x, 1-\alpha/2)$$
 and  $\sigma_T^2 \succ xU/\chi^2(x, \alpha/2)$ .

Thus if

$$u = (1/c)(msT - msE)$$
  
(which is our estimate for  $\sigma_T^2$ ),

then

$$(xu/\chi^{2}(x, \alpha/2), xu/\chi^{2}(x, 1-\alpha/2))$$

is an approximate  $(1-\alpha)100\%$  confidence interval for  $\sigma_T^2$ .

Example: With the loom data, find a 95% confidence interval for  $\sigma_T^2$ .