

## Appendix C

### Propagation of Error, and Standard Errors for Derived Quantities

A reminder about how we get approximate standard errors for functions of quantities which are themselves estimated with error.

Suppose we are trying to estimate some quantity  $\theta$ . We compute an estimate  $\hat{\theta}$ , based on our data. Since our data is more or less random, so is  $\hat{\theta}$ . One convenient way of measuring the purely statistical noise or uncertainty in  $\hat{\theta}$  is its standard deviation. This is the **standard error** of our estimate of  $\theta$ <sup>[1]</sup>. Standard errors are not the only way of summarizing this noise, nor a completely sufficient way, but they are often useful.

Suppose that our estimate  $\hat{\theta}$  is a function of some intermediate quantities  $\hat{\psi}_1, \hat{\psi}_2, \dots, \hat{\psi}_p$ , which are also estimated:

$$\hat{\theta} = f(\hat{\psi}_1, \hat{\psi}_2, \dots, \hat{\psi}_p) \quad (\text{C.1})$$

For instance,  $\theta$  might be the difference in expected values between two groups, with  $\psi_1$  and  $\psi_2$  the expected values in the two groups, and  $f(\psi_1, \psi_2) = \psi_1 - \psi_2$ . If we have a standard error for each of the original quantities  $\psi_i$ , it would seem like we should be able to get a standard error for the **derived quantity**  $\theta$ . There is in fact a simple if approximate way of doing so, which is called **propagation of error**<sup>[2]</sup>.

We start with (what else?) a Taylor expansion (App. **B**). We'll write  $\psi_i^*$  for the true (ensemble or population) value which is estimated by  $\hat{\psi}_i$ .

$$f(\psi_1^*, \psi_2^*, \dots, \psi_p^*) \approx f(\hat{\psi}_1, \hat{\psi}_2, \dots, \hat{\psi}_p) + \sum_{i=1}^p (\psi_i^* - \hat{\psi}_i) \left. \frac{\partial f}{\partial \psi_i} \right|_{\psi=\hat{\psi}} \quad (\text{C.2})$$

$$f(\hat{\psi}_1, \hat{\psi}_2, \dots, \hat{\psi}_p) \approx f(\psi_1^*, \psi_2^*, \dots, \psi_p^*) + \sum_{i=1}^p (\hat{\psi}_i - \psi_i^*) \left. \frac{\partial f}{\partial \psi_i} \right|_{\psi=\hat{\psi}} \quad (\text{C.3})$$

$$\hat{\theta} \approx \theta^* + \sum_{i=1}^p (\hat{\psi}_i - \psi_i^*) f'_i(\hat{\psi}) \quad (\text{C.4})$$

introducing  $f'_i$  as an abbreviation for  $\left. \frac{\partial f}{\partial \psi_i} \right|_{\psi=\hat{\psi}}$ . The left-hand side is now the quantity

<sup>1</sup> It is not, of course, to be confused with the standard deviation of the data. It is not even to be confused with the standard error of the mean, unless  $\theta$  is the expected value of the data and  $\hat{\theta}$  is the sample mean.

<sup>2</sup> Or, sometimes, the **delta method**.

whose standard error we want. I have done this manipulation because now  $\hat{\theta}$  is a linear function (approximately!) of some random quantities whose variances we know, and some derivatives which we can calculate.

Remember the rules for arithmetic with variances: if  $X$  and  $Y$  are random variables, and  $a$ ,  $b$  and  $c$  are constants,

$$\mathbb{V}[a] = 0 \tag{C.5}$$

$$\mathbb{V}[a + bX] = b^2\mathbb{V}[X] \tag{C.6}$$

$$\mathbb{V}[a + bX + cY] = b^2\mathbb{V}[X] + c^2\mathbb{V}[Y] + 2bc\text{Cov}[X, Y] \tag{C.7}$$

While we don't know  $f(\psi_1^*, \psi_2^*, \dots, \psi_p^*)$ , it's constant, so it has variance 0. Similarly,  $\mathbb{V}[\hat{\psi}_i - \psi_i^*] = \mathbb{V}[\hat{\psi}_i]$ . Repeatedly applying these rules to Eq. [C.4](#)

$$\mathbb{V}[\hat{\theta}] \approx \sum_{i=1}^p (f'_i(\hat{\psi}))^2 \mathbb{V}[\hat{\psi}_i] + 2 \sum_{i=1}^{p-1} \sum_{j=i+1}^p f'_i(\hat{\psi}) f'_j(\hat{\psi}) \text{Cov}[\hat{\psi}_i, \hat{\psi}_j] \tag{C.8}$$

The standard error for  $\hat{\theta}$  would then be the square root of this.

If we follow this rule for the simple case of group differences,  $f(\psi_1, \psi_2) = \psi_1 - \psi_2$ , we find that

$$\mathbb{V}[\hat{\theta}] = \mathbb{V}[\hat{\psi}_1] + \mathbb{V}[\hat{\psi}_2] - 2\text{Cov}[\hat{\psi}_1, \hat{\psi}_2] \tag{C.9}$$

just as we would find from the basic rules for arithmetic with variances. The approximation in Eq. [C.8](#) comes from the nonlinearities in  $f$ .

If the estimates of the initial quantities are uncorrelated, Eq. [C.8](#) simplifies to

$$\mathbb{V}[\hat{\theta}] \approx \sum_{i=1}^p (f'_i(\hat{\psi}))^2 \mathbb{V}[\hat{\psi}_i] \tag{C.10}$$

and, again, the standard error of  $\hat{\theta}$  would be the square root of this. The special case of Eq. [C.10](#) is sometimes called *the* propagation of error formula, but I think it's better to use that name for the more general Eq. [C.8](#)