#### INTRACLASS CORRELATION FORMULAS

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## 1. Fisher's intraclass correlation coefficient

In the simplest scenario of clustering, we dispose of a sample z of a random variable Z. The sample of n realization is grouped into N groups also referred as clusters. The group sizes are  $n_g$ , where g is the group index, and where by definition  $\sum_{g=1}^{g=N} n_g = n$ . For convenience, we note the i-th realization in group g as  $z_{ig}$ . For this one-level hierarchy Fisher's intraclass correlation coefficient can be written as follows[1]:

(1) 
$$\rho_{\text{Fisher}} = \frac{\sum_{g} \sum_{i \neq j} (z_{ig} - \bar{z}) (z_{jg} - \bar{z})}{V(z_{ig}) \cdot \sum_{g} n_g (n_g - 1)}$$

where  $z_{ig}$  denotes the *i*-th value in group (cluster) g.  $V(z_{ig})$  denotes the variance of the sample values regardless of the group structure:

(2) 
$$V(z_{ig}) = \frac{\sum_{g} \sum_{i} (z_{ig} - \bar{z})^{2}}{n}$$

Likewise, the sample mean  $\bar{z}$  is calculated regardless of the group structure as:

(3) 
$$\bar{z} = \frac{\sum_{g} \sum_{i} z_{ig}}{n}$$

In eq. 1, the squared terms are explicitly excluded by the condition  $i \neq j$ . For calculation purposes, and particularly with the aggregation methods in R, it is more convenient to sum over all combinations and subtract squared terms instead. In this sense, one notes that  $\sum_{g} \sum_{i \neq j} (z_{ig} - \bar{z}) (z_{jg} - \bar{z}) = \sum_{g} \left[ \sum_{i} (z_{ig} - \bar{z}) \sum_{j} (z_{jg} - \bar{z}) \right] - \sum_{g} \sum_{i} (z_{ig} - \bar{z})^{2}.$  With the group means  $\overline{z_g} = \frac{\sum_{i} z_{ig}}{n_g}$ , one further notes that  $\sum_{i} (z_{ig} - \bar{z}) = n_g (\overline{z_g} - \bar{z})$ . By the definition of the overall variance in eq. 2, we also have  $\sum_{g} \sum_{i} (z_{ig} - \bar{z})^{2} = n \cdot V(z_{ig})$ , and so eq. 1 can be equivalently rewritten as:

(4) 
$$\rho_{\text{Fisher}} = \frac{1}{\sum_{g} n_g (n_g - 1)} \cdot \left[ \frac{\sum_{g} n_g^2 (\overline{z_g} - \overline{z})^2}{V(z_{ig})} - n \right]$$

As it easier to evaluate in R that eq. 1, we used eq. 4 to facilitate the evaluation of  $\rho_{\text{Fisher}}$  (i.e. R script ICC.R, method "unbiased").

### 2. ANOVA-BASED INTRACLASS CORRELATION COEFFICIENTS

For equal group sizes  $n_q = K = \text{const}$ , eq. 4 can be further simplified:

(5) 
$$\rho_{\text{Harris}} = \frac{1}{(K-1)} \cdot \left[ \frac{K \sum_{g} (\overline{z_g} - \overline{z})^2}{N \cdot V(z_{ig})} - 1 \right]$$

where we have used of n = NK for equal group sizes. Eq. 5 is well-known[2]. Indeed, aside from the biais correction term  $\frac{1}{K-1}$ , it explicitly relates intraclass variance (the  $(\overline{z_g} - \overline{z})^2$  terms) to overall residual variance  $V(z_{ig})$ . This is why in analysis of variance ANOVAapproaches, the intraclass correlation coefficients are typically derived from intra-and intergroup variance, as in[1]:

(6) 
$$\rho_{\text{ANOVA}} = \frac{\sigma_{\text{inter-group}}^2}{\sigma_{\text{inter-group}}^2 + \sigma_{\text{intra-group}}^2}$$

The statistical properties such as expectation values and confidence intervals of such ANOVA-based intraclass correlation coefficients have been studied in great detail[3]. For completeness, the intraclass correlation calculation method ICC in the moultonTools package (file ICC.R) allows evaluation of such ANOVA-based intraclass correlation by invoking functions of the linear modelling package "nlme" on which moultonTools depends for this purpose. However, the primary aim of the moultonTools package is not to replace such ANOVA-based approaches, it is generally better to remain within an ANOVA framework when relying on ANOVA for intraclass correlation.

## 3. Range of values and biais of intraclass correlation coefficients

An intraclass correlation coefficient should ideally range from 0 (for random sample values that are not correlated to group identity g) to 1 for full clustering (i.e. when the  $z_{ig}$  values are constant in each group such that  $z_{ig} = \overline{z_g}$ ).

For the ANOVA-based approaches, according to eq. 6, full correlation leads indeed to  $\rho_{\text{ANOVA}} = 1$  since with full correlation,  $\sigma_{\text{intra-group}}^2 = 0$ . However, since the  $\sigma^2$  values are never negative,  $\rho_{\text{ANOVA}} \geq 0$  always holds; due to non-zero variability of the sampling and thus  $\rho_{\text{ANOVA}}$ , the expectation value is generally positive:  $E\left(\rho_{\text{ANOVA}}\right) > 0$ . This introduces known biais in the ANOVA approach[1], in is the reason why especially for smaller samples, unbiased formulae such as eq. 1, eq. 4 or eq. 5 or continue to be used despite their more complicated definitions[1].

It is also worthwhile to consider the value range for the unbiased estimators given by eq. 1, eq. 4 and eq. 5. Eq. 5 being a specialisation of eq. 1 or equivalently eq. 4, we shall limit our analysis to eq. 4. Designed to minimize or eliminate bias for the uncorrelated case[1], we shall here consider the fully correlated case.

As above, in the fully correlated case, we have  $z_{ig} = \overline{z_g}$ . Therefore, the variance  $V_{\text{correlated}}(z_{ig})$  becomes:

(7) 
$$V(z_{ig}) = \frac{\sum_{g} n_g (\overline{z_g} - \overline{z})^2}{n}$$

and we have:

(8) 
$$\rho_{\text{Fisher, correlated}} = \frac{n}{\sum_{g} n_g (n_g - 1)} \cdot \left[ \frac{\sum_{g} n_g^2 (\overline{z_g} - \overline{z})^2}{\sum_{g} n_g (\overline{z_g} - \overline{z})^2} - 1 \right] \stackrel{?}{=} 1$$

Ideally,  $\rho_{\text{Fisher, correlated}} = 1$  regardless of the detailed  $z_{ig}$  values. This is unfortunately not generally the case, as the example  $z = \{\{0,0\},\{1,1,1\}\}$  shows. Indeed, in this case, the total number of samples is n = 5, the group sizes are  $n_g = \{n_1, n_2\} = \{2, 3\}$ , and the group means are  $\overline{z_1} = 0$ ,  $\overline{z_2} = 1$ , with a global mean of  $\overline{z} = \frac{3}{5}$ . Numerically, eq. 8 evaluates to:

(9) 
$$\rho_{\text{Fisher, correlated, example}} = \frac{5}{2 \cdot 1 + 3 \cdot 2} \cdot \left[ \frac{2^2 \left( -\frac{3}{5} \right)^2 + 3^2 \left( \frac{2}{5} \right)^2}{2 \cdot \left( -\frac{3}{5} \right)^2 + 3 \cdot \left( \frac{2}{5} \right)^2} - 1 \right] = \frac{7}{8} \neq 1$$

While close to 1 in the present example, this shows that the desired equality to 1 in eq. 8 does not generally hold.

Upon inspection, the cause seems to be the unequal group sizes. Indeed, for equal group sizes  $n_g = K = \text{const}$ , eq. 8 can be simplified, achieving the exact theoretical value of 1 regardless of the actual  $\overline{z_q}$ 

(10) 
$$\rho_{\text{Fisher, correlated}} = \frac{NK}{NK(K-1)} \cdot \left[ \frac{K^2 \sum_g (\overline{z_g} - \overline{z})^2}{K \sum_g (\overline{z_g} - \overline{z})^2} - 1 \right] = 1$$

Hence, we conclude that for unequal group sizes, Fisher's formula may be unbiased, but it doesn't produce  $\rho = 1$  for the fully correlated case as desired. On the opposite, the ANOVA approach produces  $\rho = 1$  for the fully correlated case, but is biaised.

#### 4. Unbiased full-range estimator

Hence, the final question is whether it is possible to provide an unbiased estimator for the intraclass correlation coefficient that evaluates to exactly 1 for the fully clustered case, yet is unbiased for a totally uncorrelated sample.

The answer to this question is yes, but only under the assumption of normality of at least the  $\overline{z}_g$  and by consequence the  $\overline{z}$  values. This is a relatively reasonable assumption due to the contribution of multiple terms to these averages, which by the central theorem of statistics tend to converge towards normality if the  $z_{ig}$  are identically distributed.

Under the assumption of normality, the sums of squares follow known  $\chi^2$  statistics, and the ratios follow F-statistics[3]. These have generally known expectation values such that biais can be corrected for. Hence, by suitably adjusting the expectation value to 0 while maintaining the return value for the fully correlated case, one obtains:

(11) 
$$\rho_{\text{ANOVA, unbiased}} = \frac{n-3}{n-N-2} \cdot \left[ \frac{\sum_{g} n_g \left( \overline{z_g} - \overline{z} \right)^2}{n \cdot V\left( z_{ig} \right)} - \frac{N-1}{n-3} \right]$$

Indeed, for the totally correlated case,  $\frac{\sum_g n_g(\overline{z_g}-\overline{z})^2}{n\cdot V(z_{ig})}=1$  and so eq. 11 reduces to  $\rho_{\text{ANOVA, unbiased, fully clustered}}=1$  as it ought to be. Under the hypothesis of normally (theoretical variance:  $\sigma^2$ ), identically distributed, independent and thus uncorrelated  $z_{ig}$ , the variance properties of the means imply that  $\frac{\sum_g n_g(\overline{z_g}-\overline{z})^2}{\sigma^2}$  is a random variable with a  $\chi^2$  distribution with N-1 degrees of freedem. Similarly, in the denominator,  $\frac{n\cdot V(z_{ig})}{\sigma^2}$  is a  $\chi^2$  variable with n-1 degrees of freedom. As in other F-statistics, numerator and denominator in eq. 11 are only approximately independent since they involve the same  $z_{ig}$  values grouped differently, but nevertheless, one expects the expression  $\frac{n-1}{N-1}\frac{\sum_g n_g(\overline{z_g}-\overline{z})^2}{n\cdot V(z_{ig})}$  to be reasonable approximated by an F-statistic with N-1 degrees of freedom in the numerator and n-1 degrees of freedom in the denominator.

The expectation value of F-statistics is well known, and so:

(12) 
$$E\left(\frac{n-1}{N-1} \cdot \frac{\sum_{g} n_{g} (\overline{z_{g}} - \overline{z})^{2}}{n \cdot V(z_{ig})}\right) = E(F_{N-1,n-1}) = \frac{n-1}{n-3} \Rightarrow$$

$$E\left(\frac{\sum_{g} n_{g} (\overline{z_{g}} - \overline{z})^{2}}{n \cdot V(z_{ig})}\right) = \frac{N-1}{n-3}$$

So that ultimately:

(13) 
$$E\left(\rho_{\text{ANOVA, unbiased, uncorrelated}}\right) = \frac{n-3}{n-N-2} \cdot \left[\frac{N-1}{n-3} - \frac{N-1}{n-3}\right] = 0$$
 as it ought to be.

# References

- [1] J. Angrist and J.-S. Pischke, Mostly Harmless Econometrics: An Empiricist's Companion. Princeton, Oxford: Princeton University Press, 2009.
- [2] J. A. HARRIS, "ON THE CALCULATION OF INTRA-CLASS AND INTER-CLASS COEFFICIENTS OF CORRELATION FROM CLASS MOMENTS WHEN THE NUMBER OF POSSIBLE COMBINATIONS IS LARGE," *Biometrika*, vol. 9, pp. 446–472, Oct. 1913.
- [3] J. L. Fleiss and P. E. Shrout, "Approximate interval estimation for a certain intraclass correlation coefficient," *Psychometrika*, vol. 43, pp. 259–262, June 1978.