## Analytical Methods



View Article Online

## AMC TECHNICAL BRIEFS



Cite this: DOI: 10.1039/c4ay90062f

# Unbalanced robust ANOVA for the estimation of measurement uncertainty at reduced cost

Analytical Methods Committee, AMCTB No 64

Received 10th July 2014

DOI: 10.1039/c4ay90062f

www.rsc.org/methods

An empirical estimation of the random components of measurement uncertainty arising from the sampling and analytical processes can be made *via* an experiment involving replication. Analysis of variance (ANOVA) is then used to estimate the variance components. However, the fully balanced design is unduly costly, requiring four analyses per sampling target. This cost can be reduced by 25% by using an unbalanced design. Robust ANOVA in this context is often a useful tool, as it gives a more representative estimate of the separate variances than classical ANOVA when outlying results are encountered, but software for executing unbalanced robust ANOVA has hitherto been unavailable. This Technical Brief introduces the unbalanced design and the corresponding software provided by the AMC.

#### Balanced design—classical ANOVA

Estimation of uncertainty in measurement is essential for the reliable interpretation of results. The method recommended by the Eurachem/CITAC guide as being the simplest to apply is the duplicate method.<sup>1</sup> This requires randomised duplication both at the sampling and the analytical stages of an investigation. A schematic diagram of the full balanced design is shown in Fig. 1(A).

For this purpose a minimum of eight targets are randomly assigned for the duplication.<sup>1</sup> (A 'target' could be the material at a sampling location, a batch of produce, *etc.*, as appropriate.) At each of these targets, a duplicate sample is acquired according to a predetermined but suitably re-randomised protocol. Both of the samples from each target are then analysed twice. The separate contributions made by sampling and analysis to the total uncertainty budget can be estimated by analysis of variance (ANOVA). A



simulated dataset has been generated for the purpose of demonstration and is shown in Table 1 and Fig. 2.

Table 2 includes the output from the application of classical ANOVA to these data. This naive analysis suggests that very high levels of analytical uncertainty are prevalent. However, environmental measurements are often affected by a proportion of extreme values. In this case, the difference between the



**Fig. 1** Design for the estimation of the random components of uncertainty arising from both the sampling and the analytical procedures. (A) Balanced design. (B) Unbalanced design.

Table 1 Simulated data drawn from a population with mean = 10, sampling relative standard deviation = 20%, analytical relative standard deviation = 5%. A single outlying result has been substituted for the original result at location 7 (boldface type). The column heading 'S1A1' indicates the first result on the first sample, 'S1A2' the second result on the first sample, and so on

Location	S1A1	S1A2	S2A1	S2A2
4	10.7	14.6	12.0	145
1	13./	14.6	13.8	14.5
2	10.1	9.4	5.7	5.7
3	5.2	5.4	10.9	9.2
4	11.7	12.5	10.4	9.9
5	9.5	10.0	9.3	8.2
6	11.8	11.0	12.0	12.7
7	18.2	8.7	13.0	12.4
8	13.4	12.1	14.9	15.8
9	5.3	5.3	5.1	5.4
10	12.6	12.9	17.2	18.6

Table 2	Estimates of sampling and analytical	uncertainties obtained by	y applying various ANOVA	procedures to the data in Table 1
---------	--------------------------------------	---------------------------	--------------------------	-----------------------------------

	Seed value (before outlier is included)	Balanced (classical)	Balanced (robust)	Unbalanced (classical)	Unbalanced (robust)
Mean	10	11	11	12	11
Relative sampling uncertainty (2s)	40%	30%	38%	8%	34%
Relative analytical uncertainty (2s)	10%	30%	13%	37%	12%



Fig. 2 Simulated data from a duplicated experiment to estimate uncertainty from sampling and analysis.

measurements made on Analysis 1 and Analysis 2 of Sample 7 is noticeably greater than the norm (Table 1). A visual presentation (Fig. 2) shows that nine out of the ten targets show much smaller differences between the analytical duplicates.

#### Balanced design-robust ANOVA

Classical ANOVA has provided an idea of the sampling and analytical uncertainties that are atypical of the majority of the targets, because of the outlying result. Where such instances occur, robust ANOVA can give more realistic estimates of the variance components, by downweighting<sup>1</sup> the influence of extreme values. Calculation of robust variances by the AMC method requires an iterative approach, which is best carried out by a computer program.<sup>2,3</sup> Application of this method to the simulated data is shown in Table 2.

Here, for the balanced design, the estimate of analytical uncertainty has diminished from 30% to 13% while the sampling uncertainty has increased from 30% to 38%. In both instances these robust estimates can be considered more representative of the typical differences between duplicated analyses and samples in the underlying population.

#### The unbalanced experimental design

A drawback of the balanced design is the cost incurred by the four analyses on each target. An alternative method, described in the Eurachem/CITAC guide, is to use an unbalanced experimental design (Fig. 1(B)). In this design, only one of the two samples per target is subjected to duplicated analysis. Thus the cost of analysis for the experiment is reduced by 25%. However,

unbalanced robust nested ANOVA is complicated to execute and hitherto no software has been available. A new program has now been written to perform the robust calculations required, and is available gratis from the AMC website. Tests on real and simulated datasets show good comparisons between estimates made using the balanced and the unbalanced designs.<sup>4</sup>

Analysis of the simulated data in Table 1 using an unbalanced design (*i.e.*, omitting data from column S2A2) is also shown in Table 2. The unbalanced estimates are only marginally different from those obtained by the full balanced design. Again, these outcomes can be considered to be better estimates of the typical variances in the overall population than those given by classical ANOVA. In practice, the random components of sampling and analytical uncertainty could be subsumed into the total measurement uncertainty for data interpretation purposes.<sup>1</sup>

#### Software

The program U-RANOVA, which has been written in Visual Basic for Applications (Excel) is now available on the AMC website, together with a file of the example data. It is a stand-alone program to execute robust ANOVA with nested data, with options for the use of balanced or unbalanced designs. It will run in Windows or, with slight changes, in a Mac environment.

### References

- 1 Eurachem/CITAC, Measurement uncertainty arising from sampling, 2007.
- 2 Analytical Methods Committee, Analyst, 1989, 114, 1699.
- 3 AMC Technical Brief, No 6.
- 4 P. D. Rostron and M. H. Ramsey, *Accredit. Qual. Assur.*, 2012, 17, 7.

#### Peter Rostron [University of Sussex]

This Technical Brief was drafted on behalf of the Subcommittee for Uncertainty from Sampling and approved by the Analytical Methods Committee on 04/07/14.

<b>CPD Certification</b> I certify that I have studied this document as a contribution to Continuing Professional Development.
Name SignatureDate
Name of supervisorDateDate