



## Bayesian estimation of the relative deviations between activities in the radionuclide standardization

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### ABSTRACT

The dissemination of the activity is performed from radionuclide sources prepared in a sequence of dilutions and weighing methods. In this process, the activity of the source can be estimated statistically from the deposited mass and the activity concentration of the master solution. After preparation, the activity is obtained from absolute or relative measurement methods. However, the methods of activity determinations used may not fulfill the necessary independence of the conventional statistical approach due to the presence of possible correlations between activities that arise with the use of the same standardized sources or with the same method of quantity measurements. In this paper, Bayesian estimates for the relative deviation of activities and their uncertainty were obtained in order to evaluate the performance of the main sources' preparation method. The estimate result ( $0.55 \pm 0.27$ ) % for a data set of radionuclide standardization performed between 2017 and 2018 at LNMRI, although close to zero, shows one should investigate possible effects affecting the preparation and measurement of the sources. This Bayesian estimate was validated by monte carlo simulation method.

*Keywords:* Radionuclide metrology; Radioactive source preparation; Bayesian statistic.

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## 1. INTRODUCTION

In the area of Radionuclide Metrology, the dissemination of activity [1] is carried out from radionuclide sources (standards for the dissemination of quantity) prepared from a sequence of dilutions and weighing [2] of a solution with high concentration of activity. Dilutions are carried out in glass ampoules with a stable isotope acid carrier solution of the radionuclide in order to maintain the stability of the standard and avoid the adsorption of radionuclides on the inner wall of the flask and heavy [3]. Weighing are performed by means of the several differential weighing methods (the pycnometer method, the substitution method, elimination method and modified elimination method), depends on the required uncertainty associated to activity per mass of the source [4]. After the source preparation, the activity is determined from absolute or relative measurement methods for liquid or solid sources [5].

In this process of preparation of sources, two estimates of activity are available, the first obtained from the multiplication of the mass deposited by the activity concentration of the master solution and the second one obtained from the measurement methods. As the activity measurement does not change the activity of the prepared source, the relative deviation between the two activity values can be used to evaluate the performance of the process of preparation of sources and possible effects between the preparation of solid and liquid sources and the measurement of the activity. In order to performing these evaluations in presence of set of relative deviation and its uncertainty a mean value that summarizes all information about deviation is required.

The problem to estimation of the mean value should be faced as that of the estimation of the key-comparison reference value (KCRV) for a data set. By this way a mean value for the relative deviation as the weighted mean based on classical statistical is recommended [6]. However, the presence of correlations between the relative deviation arisen from usage of the same standard sources and the activity measurement method does not comply with the independence required to this statistical approach. Thus, in order to get around this problem, in this work, a Bayesian estimate for the weighted mean of the relative deviation was determined for a data set of 45 radionuclide standardizations performed between 2017 and 2018 at LNMRI.

## 2. THEORETICAL BACKGROUND

### 2.1. Bayesian theory applied to relative deviation

The relative deviation of the activities is defined according to equation (1). In this work we consider the activity determined from the source preparation process  $A_d$  with uncertainty from the master solution, the measured activity of the radionuclide source  $A$  and its measurement uncertainty and, only as reference (without uncertainty) for the relative deviation, the average activity  $A_m$ .

$$D = \frac{A - A_d}{A_m} \quad (1)$$

The uncertainty of the relative deviation  $D$  is determined from the uncertainties of the activities  $A$  and  $A_d$ , all of activities were corrected to the same reference date.

The system of measurements of sources activities yields as results the activity  $A$  and associated uncertainty  $u$ , which are assumed to be based on a large number of counts after all corrections for systematic effects have been applied. It is considered that these results, when determined from the GUM [7], come from a normal probability distribution centered on  $\xi$ , which describes the measurement system, this is the likelihood function related to the parameter  $\xi$ , according (2) [8]. The same hypothesis is assumed for the activity determined after the preparation of the sources,  $A_d$ .

$$p(A, u|\xi) \propto \exp\left[-\frac{1}{2}\left(\frac{A - \xi}{u}\right)^2\right] \quad (2)$$

As the relative deviation defined in (1) is a function of the difference between the measured activity and the determined activity, its measurement system also obeys a normal distribution and thus, by changes of variables in (2), the likelihood function can be rewritten as a function of the relative deviation  $D$ , the measurement uncertainty of the relative deviation  $u_D$  and the parameter  $\eta$ , as previously, the likelihood can be associated to a normal distribution, now centered on  $\eta$ , (3).

$$p(D, u_D|\eta) \propto \exp\left[-\frac{1}{2}\left(\frac{D - \eta}{u_D}\right)^2\right] \quad (3)$$

To determine the posterior probability distribution  $p(\eta | D, u_D)$  it is necessary to define the *a priori* distribution  $p(\eta)$  associated to the parameter [9]. As we do not have any information about the behavior of the parameter  $\eta$  that allows us to associate some probability distribution to it, it is

natural to use a reference non-informative prior (4), that is a constant when the estimation is based on a location parameter [10].

$$p(\eta) \propto 1 \quad (4)$$

Thus, after the first measurement of activity  $A_1$  and uncertainty  $u_1$ , the values  $D_1$  and  $u_{D1}$  are determined from (1). The posterior distribution  $p(\eta | D, u_D)$  can be obtained (5).

$$p(\eta | D_1, u_{D1}) \propto \exp \left[ -\frac{1}{2} \left( \frac{D_1 - \eta}{u_{D1}} \right)^2 \right] \quad (5)$$

After observing the measurements sequences, conditionally independent given the parameter  $\eta$ ,  $\mathbf{D} = \{D_1, \dots, D_n\}$  and  $\mathbf{u}_D = \{u_{D1}, \dots, u_{Dn}\}$ , the posterior distribution becomes (6).

$$p(\eta | \mathbf{D}, \mathbf{u}_D) \propto \exp \left[ -\frac{1}{2} \sum_{i=1}^n \left( \frac{D_i - \eta}{u_{Di}} \right)^2 \right] \quad (6)$$

The exponent of the *à posteriori* distribution can be rewritten in order to represent it as a normal distribution.

$$-\frac{1}{2} \sum_{i=1}^n \left( \frac{D_i - \eta}{u_{Di}} \right)^2 = -\frac{1}{2} \left[ \sum_{i=1}^n \frac{D_i^2}{u_{Di}^2} - 2\eta \sum_{i=1}^n \frac{D_i}{u_{Di}^2} + \eta^2 \sum_{i=1}^n \frac{1}{u_{Di}^2} \right] \quad (7)$$

The first term in (7) does not depend on the parameter  $\eta$  and only has known values, so it can be included in the normalization factor of the *à posteriori* (8).

$$p(\eta | \mathbf{D}, \mathbf{u}_D) \propto \exp \left\{ -\frac{1}{2} \left[ \frac{1}{u_p^2} (\eta^2 - 2\eta u_p^2 b) \right] \right\} \cdot \frac{1}{u_p^2} \equiv \sum_{i=1}^n \frac{1}{u_{Di}^2}; \quad b \equiv \sum_{i=1}^n \frac{D_i}{u_{Di}^2} \quad (8)$$

In (8), the square of the term between brackets is completed by adding and subtracting  $(bu_p^2)^2$ , where the last term in does not depend on the parameter  $\eta$ , so it can be included in the normalization factor. Then, the expression for *à posteriori* becomes (9).

$$p(\eta | \mathbf{D}, \mathbf{u}_D) \propto \exp \left\{ -\frac{1}{2} \left[ \frac{1}{u_p^2} (\eta - bu_p^2)^2 \right] \right\} \quad (9)$$

This expression for *à posteriori* distribution is the same of a random variable obeying a normal distribution with mean  $bu_p^2$  and variance  $u_p^2$ . Thus, from the definitions of  $b$  and  $u_p^2$ , it can be concluded that the expected value and the parameter variance are, respectively, the weighted mean and variance obtained from the measured values, which pertain to the sequences  $\mathbf{D}$  and  $\mathbf{u}_D$  (10).

$$E(\eta|\mathbf{D}, \mathbf{u}_D) = u_p^2 \sum_{i=1}^n \frac{D_i}{u_{Di}^2} \quad e \quad V(\eta|\mathbf{D}, \mathbf{u}_D) = u_p^2 = \left( \sum_{i=1}^n \frac{1}{u_{Di}^2} \right)^{-1} \quad (10)$$

In presence of previous knowledge about the correlation between each relative deviation  $D_i$ , the presented theoretical description should be rewritten in matrix form. In such case, the probability distribution associated to the vector of deviations  $\mathbf{D}^*$  is a multivariate normal distribution with a covariance matrix  $\mathbf{W}$ , equation (11), where  $\mathbf{1}$  is a unitary column vector and the superscript refers to transpose vector [11].

$$p(\mathbf{D}^*, \mathbf{W}|\eta) \propto \exp \left[ -\frac{1}{2} (\mathbf{D}^* - \mathbf{1}\eta)^T \mathbf{W}^{-1} (\mathbf{D}^* - \mathbf{1}\eta) \right] \quad (11)$$

Due to the lack of conditional independence, the likelihood factorization in equation (6) is not possible. However the Bayes theorem and the consideration about a reference non-informative prior remains valid (12).

$$p(\eta|\mathbf{D}^*, \mathbf{W}) \propto \exp \left[ -\frac{1}{2} (\mathbf{D}^* - \mathbf{1}\eta)^T \mathbf{W}^{-1} (\mathbf{D}^* - \mathbf{1}\eta) \right] \quad (12)$$

As before, bracketed terms can be rearranged from a few multiplications and keep only the factors that contain the parameter (13).

$$p(\eta|\mathbf{D}^*, \mathbf{W}) \propto \exp \left\{ -\frac{1}{2} \left[ (\mathbf{1}^T \mathbf{W}^{-1} \mathbf{1}) \left( \eta^2 - 2\eta \left( \frac{\mathbf{1}^T \mathbf{W}^{-1} \mathbf{D}^*}{\mathbf{1}^T \mathbf{W}^{-1} \mathbf{1}} \right) \right) \right] \right\} \quad (13)$$

A normal distribution for  $\eta$  can be obtained in analogy to the procedure used to obtain equations (8) and (9) (14).

$$p(\eta|\mathbf{D}^*, \mathbf{W}) \propto \exp \left\{ -\frac{1}{2} \left[ (\mathbf{1}^T \mathbf{W}^{-1} \mathbf{1}) \left( \eta - \left( \frac{\mathbf{1}^T \mathbf{W}^{-1} \mathbf{D}^*}{\mathbf{1}^T \mathbf{W}^{-1} \mathbf{1}} \right)^2 \right) \right] \right\} \quad (14)$$

The expected posterior value and variance are obtained, as before, from the terms in the normal distribution (15) [12].

$$E(\eta|\mathbf{D}^*, \mathbf{W}) = \frac{\mathbf{1}^T \mathbf{W}^{-1} \mathbf{D}^*}{\mathbf{1}^T \mathbf{W}^{-1} \mathbf{1}} \quad e \quad V(\eta|\mathbf{D}^*, \mathbf{W}) = \frac{1}{\mathbf{1}^T \mathbf{W}^{-1} \mathbf{1}} \quad (15)$$

A coverage interval can be established from probability distribution, the same way as in classical statistics, by converting the normal distribution into a standardized normal distribution

$\Phi(z)$  and by defining a random variable  $Z$  as a function of the mean and standard deviation of a posteriori (16).

$$Z = [\eta - E(\eta|\mathbf{D}^*, \mathbf{W})] / \sqrt{V(\eta|\mathbf{D}^*, \mathbf{W})} \quad (16)$$

Assuming two quantiles of the reduced normal  $-z_\gamma$  and  $z_\beta$  that contain a coverage probability  $(1 - \alpha)$ , the probability of finding  $Z$  in the coverage interval  $[-z_\gamma, z_\beta]$ , which is given by equation (17).

$$P\left(-z_\gamma \leq [\eta - E(\eta|\mathbf{D}^*, \mathbf{W})] / \sqrt{V(\eta|\mathbf{D}^*, \mathbf{W})} \leq z_\beta\right) = \Phi(z_\beta) - \Phi(-z_\gamma) = 1 - \alpha \quad (17)$$

The interval covered with the quantiles (13) has length of  $(z_\beta - z_\gamma)\sqrt{V(\eta|\mathbf{D}, \mathbf{W})}$ , (18).

$$E(\eta|\mathbf{D}^*, \mathbf{W}) - z_\gamma \sqrt{V(\eta|\mathbf{D}^*, \mathbf{W})} \leq \eta \leq E(\eta|\mathbf{D}^*, \mathbf{W}) + z_\beta \sqrt{V(\eta|\mathbf{D}^*, \mathbf{W})} \quad (18)$$

It is important to highlight that in Bayesian statistics the coverage interval, also called the credibility interval, is determined for the random variable, the parameter  $\eta$ , and the interval is deterministic [11], unlike in the classical statistics, where the coverage interval is given for the deterministic parameter and the interval is a random variable [13].

From the definition and symmetry of the reduced normal distribution, the probabilities of the quantiles, respectively,  $\beta$  and  $\gamma$  are related to the coverage probability, (19).

$$\Phi(z_\beta) - \Phi(-z_\gamma) = 1 - (\beta + \gamma) = 1 - \alpha \Rightarrow \alpha = \beta + \gamma \quad (19)$$

According to the range probability value  $(1 - \alpha)$ , several quantum  $z_\gamma$  and  $z_\beta$  can be established for different probabilities  $\beta$  and  $\gamma$ ; however, in Bayesian theory the confidence interval for the parameter is determined by the pair of quantiles that form the shortest length  $(z_\beta - z_\gamma)\sqrt{V(\eta|\mathbf{D}, \mathbf{W})}$ , subject to restriction (19). This minimum length interval may be different from the probabilistically symmetric range (usually used in classical statistics) for asymmetric or multimodal (multiple peaks) probability distributions. However, this is not the case for the normal distribution and other symmetric and unimodal distributions where the minimum and probabilistically symmetric length confidence intervals coincide, and  $\beta = \gamma = \alpha / 2$ , (20).

$$E(\eta|\mathbf{D}^*, \mathbf{W}) - z_{\alpha/2} \sqrt{V(\eta|\mathbf{D}^*, \mathbf{W})} \leq \eta \leq E(\eta|\mathbf{D}^*, \mathbf{W}) + z_{\alpha/2} \sqrt{V(\eta|\mathbf{D}^*, \mathbf{W})} \quad (20)$$

In this paper, the expanded uncertainty  $U$  that forms the confidence interval for the parameter  $\eta$  is defined, for a 95% coverage probability, by (21).

$$U = z_{2,5\%} \sqrt{V(\eta|\mathbf{D}^*, \mathbf{W})} \quad (21)$$

### 3. CALCULATION PROCEDURE

#### 3.1. Input data

The data set obtained from 45 sources prepared between 2017 and 2018 are shown in Tables 1 and 2. In these Tables, the expanded uncertainty  $U$  is a relative one. The terms GMX, CI and AC refers to respectively gamma spectrometry, ionizing chamber and  $4\pi\beta\text{-}\gamma$  anti-coincidence counting methods. In standard ID column are indicated the ionizing chamber calibration factor (different for each nuclide) or the standard code. Solid (S) and Liquid (L) are the physical states of the source.

The covariance matrix  $\mathbf{W}$  required in equation (15) was obtained from covariance matrix  $\Psi_A$  between pairs of measurement activities, from covariance between pairs of estimated activities  $\Psi_{Ad}$  and the crossed covariance between measurement and estimated activities  $\Psi_{Ad,A}$  (22).

$$\mathbf{W} = \mathbf{M}^{-1}(\Psi_A + \Psi_{Ad} - \Psi_{A,Ad} - \Psi_{A,Ad}^T)\mathbf{M}^{-1} \quad (22)$$

In equation (22) the diagonal matrix  $\mathbf{M}$  is composed by the mean of the activities. The elements of each covariance matrix were calculated considering the square of the lowest standard uncertainty between equal radionuclides, measured with the same measurement method and radionuclide standard. This approach is justified because this kind of covariance is related to B-type uncertainties which should have the same value for the pairs of activities uncertainties.

**Table 1.** First 27 data used to determine activity deviation and *a posteriori* parameters.

Radionuclide	Master Solution				Source				
	A <sub>d</sub> (kBq)	U (%)	Method	Stand. ID	A (kBq)	U (%)	Method	Stand. ID	Code
<sup>137</sup> Cs	20.7	0.92	GMX	Factor	21.1	1.3	GMX	24S99	14S03
<sup>137</sup> Cs	20.7	0.98	GMX	Factor	21.4	1.4	GMX	24S99	67S08
<sup>137</sup> Cs	228.9	0.73	GMX	Factor	221.3	1.4	GMX	24S99	74S14
<sup>137</sup> Cs	57.1	0.74	GMX	Factor	60.0	1.3	GMX	24S99	193S14
<sup>137</sup> Cs	57.6	0.74	GMX	Factor	56.7	1.4	GMX	24S99	214S14
<sup>241</sup> Am	0.106	1.4	GMX	32	0.11	9.2	GMX	32	14L17
<sup>152</sup> Eu	49.9	1.5	CI3	Factor	48.9	3.5	GMX	36	45L17
<sup>241</sup> Am	0.69	1.4	GMX	32	0.69	1.5	GMX	32	55L17
<sup>60</sup> Co	0.64	0.84	GMX	67	0.63	1.0	GMX	47	55L17
<sup>137</sup> Cs	1.02	2.7	GMX	113	1.04	2.8	GMX	79	55L17
<sup>210</sup> Pb	1.06	2.5	GMX	17	1.00	3.5	GMX	8	55L17
<sup>57</sup> Co	1.06	0.76	CI3	Factor	1.08	1.7	GMX	35	55L17
<sup>133</sup> Ba	1.41	1.0	GMX	24	1.49	1.3	GMX	24	55L17
<sup>134</sup> Cs	1.66	0.85	CI3	Factor	1.71	2.8	GMX	60	55L17
<sup>54</sup> Mn	1.85	1.2	CI3	Factor	1.77	1.3	GMX	43	55L17
<sup>65</sup> Zn	2.12	1.8	CI3	Factor	2.33	2.0	GMX	36	55L17
<sup>226</sup> Ra	1.11	2.6	GMX	03	1.10	2.5	GMX	47	66L17
<sup>60</sup> Co	21.7	0.50	CI3	Factor	21.7	2.5	GMX	47	87L17
<sup>137</sup> Cs	11.3	1.3	CI3	Factor	11.4	2.8	GMX	79	88L17
<sup>60</sup> Co	20.8	0.50	CI3	Factor	21.0	1.2	CI3	Factor	87L17
<sup>241</sup> Am	6.2	1.4	GMX	32	5.8	1.3	GMX	52	18L18
<sup>241</sup> Am	5.8	1.4	GMX	32	5.5	1.4	GMX	52	17L18
<sup>60</sup> Co	2.48	1.3	NaI	10	2.47	1.3	NaI	107	47
<sup>60</sup> Co	2.48	1.3	NaI	10	2.5	6.2	CI3	Factor	47
<sup>226</sup> Ra	0.132	2.6	GMX	03	0.134	2.6	GMX	47	29L18
<sup>22</sup> Na	365	0.69	CI3	Factor	337	2.2	GMX	132S08	13S17
<sup>60</sup> Co	12.2	0.49	CI3	Factor	12.2	1.2	GMX	67	32L18



**Table 2.** Last 18 data used to determine activity deviation and *a posteriori* parameters.

Radionuclide	Master Solution				Source				
	A <sub>d</sub> (kBq)	U (%)	Method	Stand. ID	A (kBq)	U (%)	Method	Stand. ID	Code
<sup>60</sup> Co	13.1	0.49	CI3	Factor	13.0	1.0	GMX	67	33L18
<sup>241</sup> Am	3.84	0.49	CI1	Factor	3.9	1.5	GMX	52	34L18
<sup>241</sup> Am	4.14	0.49	CI1	Factor	4.1	1.3	GMX	52	35L18
<sup>137</sup> Cs	399	0.73	CI3	Factor	453	1.6	CI3	Factor	43S18
<sup>137</sup> Cs	317	0.73	CI3	Factor	362	1.7	CI3	Factor	44S18
<sup>137</sup> Cs	399	0.73	CI3	Factor	446	1.4	GMX	24S99	43S18
<sup>137</sup> Cs	317	0.73	CI3	Factor	352	1.4	GMX	24S99	44S18
<sup>137</sup> Cs	316	0.73	CI3	Factor	367	1.5	NaI	41S16	44S18
<sup>137</sup> Cs	399	0.73	CI3	Factor	464	3.3	NaI	24S99	43S18
<sup>241</sup> Am	27.8	1.2	CI3	Factor	28.2	1.9	GMX	22S95	146S18
<sup>133</sup> Ba	19.6	0.51	CI3	Factor	18.6	0.90	GMX	129S15	147S18
<sup>60</sup> Co	45.4	0.48	CI3	Factor	46.1	1.3	GMX	130S08	148S18
<sup>152</sup> Eu	4.9	1.7	CI3	Factor	4.8	2.5	GMX	121S00	149S18
<sup>166</sup> Ho	10.7	0.50	AC	Factor	10.9	2.8	GMX	34S00	150S18
<sup>137</sup> Cs	22.9	0.76	CI3	Factor	23.3	1.4	GMX	37S97	151S18
<sup>137</sup> Cs	57.9	0.76	CI3	Factor	57.8	1.4	GMX	37S97	152S18
<sup>137</sup> Cs	58.4	0.76	CI3	Factor	57.1	1.5	GMX	37S97	153S18
<sup>137</sup> Cs	59.5	0.76	CI3	Factor	60.1	1.5	GMX	37S97	154S18

### 3.2. Validation by Monte Carlo method

The introduced theory was validated by the propagation of distributions principle as stated in JCGM 101 [8]. It was performed  $7 \times 10^5$  simulations for each random variable normally distributed A and A<sub>d</sub> in Table 1. The mean of the normal distribution is the activity while the standard deviation is considered the relative standard uncertainty. The covariance matrix **W** also took part in this simulation which was implemented by Monte Carlo method (MCM) using a properly spreadsheet plugin.

#### 4. RESULTS AND DISCUSSION

Table 2 shows the expected value and the expanded uncertainty for the parameter  $\eta$  of the posterior distribution of the relative deviation  $D$  obtained by Bayesian theory and Monte Carlo method. These results are very compatible so one can consider the theory is validated.

**Table 2.** A posteriori estimates for parameter  $\eta$  .

	$E[\eta D,up]$ (%)	U (%)
<b>Bayesian</b>	0.5514	0.27
<b>MCM</b>	0.5512	0.26

**Figure 1:** MCM results for parameter  $\eta$ .

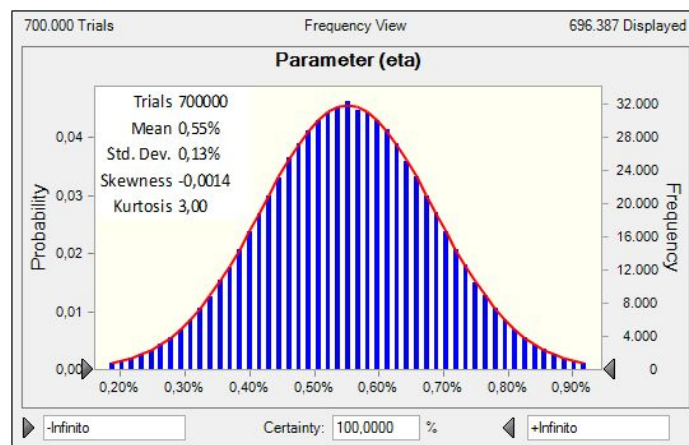
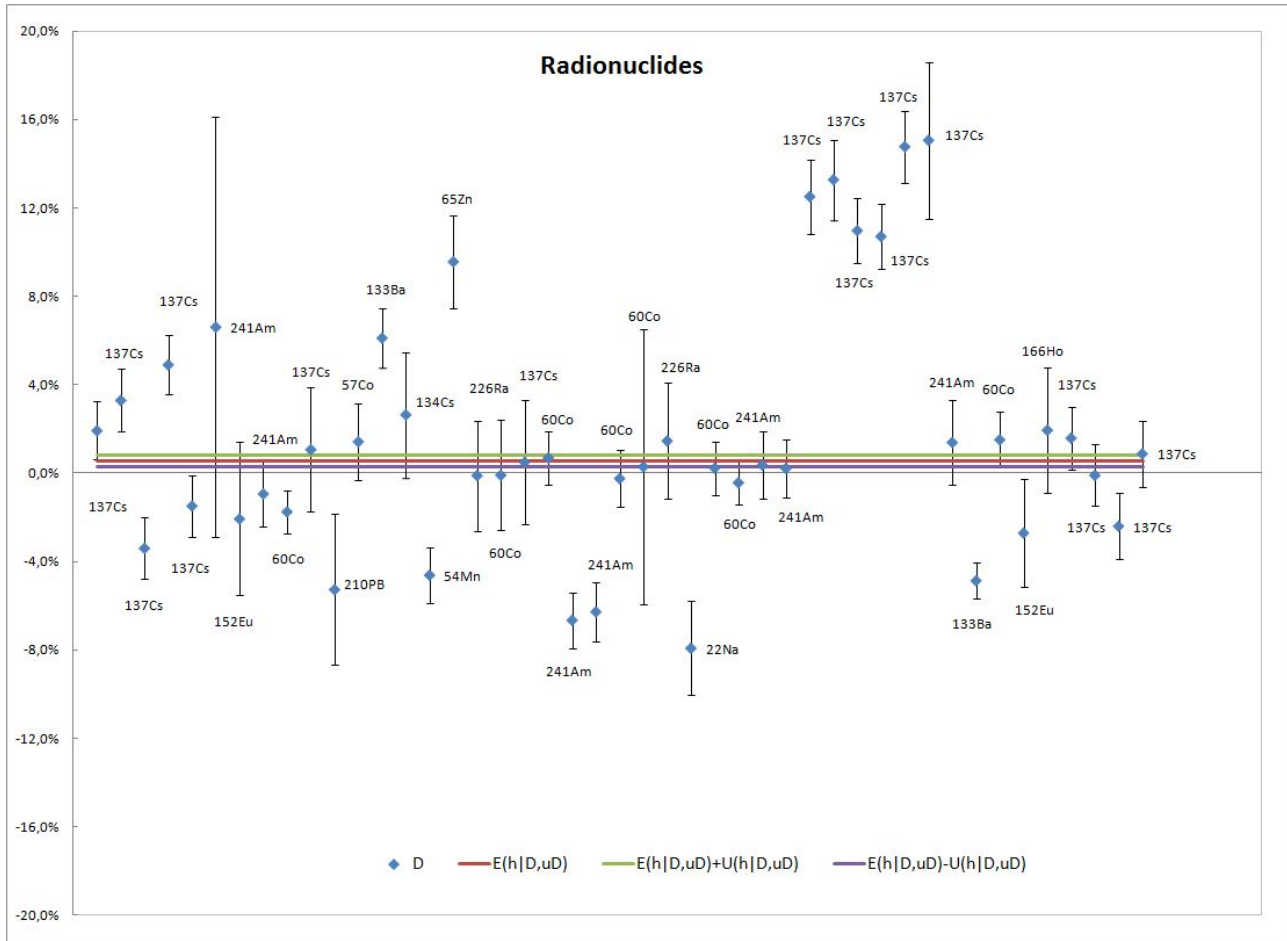


Figure 1 shows the simulation results by MCM which confirms the normal distribution for the parameter  $\eta$ . Besides, the parameters kurtosis and skewness emphasizes the convergence for a normal distribution to the simulation of the experimental data.

Figure 2 shows the comparison between the estimated values and the values of the individual relative deviations  $D_i$  for the radionuclide sources.

**Figure 2:** Individual relative deviations  $D_i$ , expected posterior value  $E[\eta | \mathbf{D}, \mathbf{u}_D]$  and confidence interval limits for parameter  $\eta$ .



It is important to note that if the covariance was not considered the expected value would be almost the double 1.16% with the same uncertainty so it would be affected by the greater values which are responsible for the dispersion between the individual deviations. It can be clearly observed with respect to the set of  $^{137}\text{Cs}$  sources at right upper corner.

## 5. CONCLUSIONS

From the radioactive decay law, it is expected that a source of radionuclides will naturally reduce activity, and that with the application of corrections to these effects, as was done in this paper, the variation of the activity with time is zero. However, the result (of this work) indicates that the variation of the activity between the preparation of the sources and the measurement is positive, although it is very close to zero, but should be of order 1% if the lack of independence was not considered. This result may be a consequence of effects on source preparations, such as residual evaporation and correction of the balance indication, or effects arising on activity measurements resulting, for example, from calibration factors. A further research will be done to test these hypotheses and adequately conclude on the cause of the positive relative deviation between activities.

One can conclude as a performance characteristic to preparing process that the accuracy to prepared sources from a nominal activity value previously specified is not better than  $(0.55 \pm 0.27) \%$ .

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