

STATISTICAL CRITERIA TO ESTIMATE THE INTERNAL DOSES IN WORKERS EXPOSED TO RADIOACTIVE AIRBORNES

Guillermo Sánchez¹, Agustín Pérez².

¹ENUSA Industrias Avanzadas S.A. Fabrica de Juzbado. Apdo 328. E-37080-Salamanca. (Spain). gsl@fab.enusa.es

²ENUSA Industrias Avanzadas S.A. apf@fab.enusa.es

Abstract.- The regulations require that monitoring of intake of radioactive material should be applied depending on radiological hazards but they do not establish precise criteria how these radiological hazards should be evaluated. Using the data from Juzbado Uranium Plant, we have made an statistical study of air sampling concentration and of the intake of a group of workers exposed to inhalation of uranium aerosols. We have found that these can be fitted to the probability distributions. Applying these distributions we have used simulation techniques to evaluate the hazards from intake. All these methods can be useful for the design and the conduct of the air control monitoring and to classify the risk for a specific workshop.

1. INTRODUCCIÓN

The Juzbado Fuel Fabrication Plant uranium fuel assemblies to light water reactors are made. This process requires manipulating powder of enriched uranium oxides. Hood and glove boxes are used for handling the uranium, but small amounts of radioactivity may be released into the room air as airborne. "Ceramic Area" is the working area where there is a potential radiological hazard for internal contamination. The method applied in Juzbado Plant to estimate the intake is similar to those applied in other uranium facilities. The uranium radioisotopes air concentrations are periodically monitored with a Static Air Sampler" (SAS). The SAS aspirates the air from the environment using a pump. The air is led through a paper filter where the airborne is collected. Air samplers are fixed in locations P_i strategically located in the workplaces. Workers are moving in the area, staying some time close every point P_i . The uranium air concentration is changing (Fig. 1).

The filter at point P_i is collected when the workers change shift every working day j and its activity $A_i(j)$ is measured. For special operations where the maximum level of airborne uranium concentration could be exceeded (i.e. while cleaning equipments, individual respirators are used), a double system to cut the flow

through the filter is applied in this case. Hereafter we will not consider the concentration during these special periods of time.

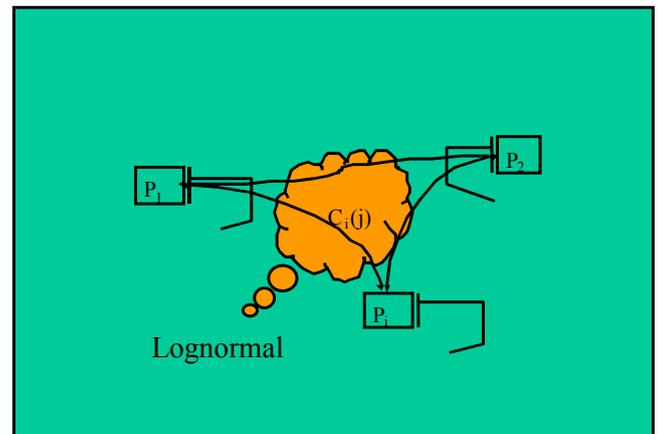


FIG. 1. Worker moving in an area with air sampling fixed at points P_i . Daily concentration $C_i(j)$ at point i on day j can be fitted to a lognormal distribution

The daily average concentration $C_i(j)$ for j at point P_i is given by $C_i(j) = A_i(j) / (\rho T)$, where T is the time during which the air flows, ρ is drawn from the filter the day j , usually a 8-hours shift. The worker stays every day, j , in N different points, i , for a period of time Δt_{ij} . The intake for this worker during day j is

$z_{1-\alpha}$ z factor of a normal standard $N(0,1)$ with one side level of signification α (e.g.: $\alpha=0.05$ $z_{0.95} = 1.649$).

3. ESTIMATING THE POTENTIAL RADIOLOGICAL HAZARDS IN WORKING AREAS

In many occasions we have only the daily intake $I^k(j)$ values (worker $k = 1, \dots, m$) of a few workers and a statistical requirement to apply (8) is that $m \geq 30$. Besides the quality of $A_i(j)$ (activity measured in filter) values is better than $I^k(j)$. Therefore we are interested in evaluating the hazard of an area and not of a specific worker.

We have found that $\{A_i(j)\}$ values, in each point i , has a good fit to many points to the lognormal $LN(\mu_i, \sigma_i)$ distribution given by

$$LN(\mu_i, \sigma_i) \equiv \begin{cases} f_i(x) = \frac{1}{\sigma_i x \sqrt{2\pi}} \text{Exp} \left[-\frac{1}{2} \left(\frac{\ln x - \mu_i}{\sigma_i} \right)^2 \right] & \text{if } x > 0, \\ f_i(x) = 0, & \text{in otherwise} \end{cases}$$

The mean and variance of the lognormal distribution are

$$\mu_X = e^{\mu + \sigma^2/2}, \quad \sigma_X^2 = e^{2\mu + 2\sigma^2} \left(e^{\sigma^2} - \sum_i w_i e^{\mu_i + \sigma_i^2/2} \right), \quad (10)$$

The following estimators derived from the transformation to the Normal distribution are widely used in practice (we have called $X_i = A_i(j)$)

$$\hat{\mu}_i = \frac{1}{N} \sum_i \ln A_i(j),$$

$$\hat{\sigma}_i^2 = \frac{1}{N} \sum_i (\ln A_i(j) - \hat{\mu}_i)^2 \quad (11)$$

where N is the total number of data of the point i .

In some occasions $A_i(j)$ are bellow lower level detection (LLD). An improved method for normal and lognormal censured

distribution [3] is utilized to take into account LLD values of $A_i(j)$.

We have considered two situations to evaluate the hazard of an area: a) Scenery A.- It is assumed that w_{ij} is known being more or less the same all days for people who work in the same area, that is $w_{ij} \approx w_i$, b) Scenery B.- It is assumed that w_{ij} changes. In both situations we want to know, for each area, the average daily intake I and their uncertainties.

Scenery A .- They are made the assumptions that follow: a) w_i takes, in each point i , a constant value all days, b) $A_i(j)$ values can be fitted to a lognormal distribution.

The sum of lognormal variables is not a known distribution, but there are some approximations to lognormal distributions [4]. One of the simplest is Fenton-Wilkinson's approximation (FW) [5,6]. This approximation has been proven accurate under certain conditions (see e.g.[7]) as is the case here for independent random variable (r.v.), means not too spread and similar variances. In particular, Fenton-Wilkinson's approximation for I in eqn (3) provides a lognormal distribution $LN(\mu(t), \sigma^2(t))$ with mean μ_I and variance σ_I^2 ,

$$\mu_I = e^{\mu + \sigma^2/2}, \quad \sigma_I^2 = e^{2\mu + 2\sigma^2} \left(e^{\sigma^2} - \sum_i w_i e^{\mu_i + \sigma_i^2/2} \right), \quad (10)$$

$$\sigma_I^2 \approx \sum_i w_i^2 e^{2\mu_i + \sigma_i^2} (e^{\sigma_i^2} - 1) \quad (12)$$

where μ_I and σ_I^2 are the mean and variance of the lognormal distribution of the intake of a person who usually works in the area. Then the daily intake and their uncertainties can be evaluated that follow.

$$I \approx \mu_I \pm z_{\frac{\gamma+1}{2}} \sqrt{\sum_i w_i^2 e^{2\mu_i + \sigma_i^2} (e^{\sigma_i^2} - 1)} \quad (13)$$

To evaluate the intake and their limits for a long period can be used eqn(7) and (8).

If it is not verified that means are not too spread and variances are similar, then mean μ_I and variance σ_I^2 can be estimated using simulation techniques. We can simulate the daily intake $I_j^{(k)}$ (we denote the worker by

(k) instead of k to indicate that it is a simulation) $I_j^{(k)}$ substituting in eqn.(3) $A_i(j)$ for $Random[LN(\mu_i, \sigma_i)]$ where $Random[LN(\mu_i, \sigma_i)]$ represent pseudorandom elements distributed according a Lognormal distribution

$$I_j^{(k)} = \sum_{i=1}^m \bar{w}_i Random[LN(\mu_i, \sigma_i)]^{(k)} \quad (14)$$

then, mean and variance of the daily intake is evaluated as usual

$$\hat{\mu}_I = \frac{\sum_{k=1}^n I^{(k)}}{n} \quad \hat{\sigma}_I^2 = \frac{\sum_{k=1}^n (I^{(k)} - \hat{\mu}_I)^2}{n-1} \quad (15)$$

where n are the number of simulation (it should be taken $n > 1000$)

Scenario B .- They are made the assumptions that follow:

a) w_{ij} changes, but the \hat{w}_i average values is known: We have represented the daily stay of every worker in each sampling point applying the uniform distribution, $U_i(0,1)$, with a density function given by:

$$U_i(0,1) \equiv \begin{cases} 0 & \text{if } x < 0, x \geq 1 \\ 1 & \text{if } 0 \leq x < 1 \end{cases} \quad (16)$$

b) $A_i(j)$ values can be fitted to a lognormal distribution $LN(\mu_i, \sigma_i)$ where μ_i , and σ_i , are the average and deviation of daily (or shift) the activities, at point i , $A_i(j)$ calculated with eqn (11) using experimental data.

In this case we have used a Monte Carlo simulation to estimate the daily intake $I_j^{(k)}$. We can simulate $I_j^{(k)}$ substituting in eqn.(3) $A_i(j)$ for $Random[LN(\mu_i, \sigma_i)]$ and w_{ij} for $\hat{w}_i Random[U_i(0,1)]$ where $Random[LN(\mu_i, \sigma_i)]$ and $Random[U_i(0,1)]$ represent pseudorandom elements distributed according a Uniform and Lognormal distribution (it can be generated using standard program such as *Mathematica* or Excel). Also, every working day (shift) j must be verified $\sum_{i=1}^m w_{ij} = 1$, then

$$I^{(k)} = \frac{1}{\sum_{i=1}^m \bar{w}_i Random[U_i(0,1)]^{(k)}} \sum_{i=1}^m \bar{w}_i Random[U_i(0,1)]^{(k)} Random[LN(\mu_i, \sigma_i)]^{(k)} \quad (17)$$

Eqn (14) can be no conservative. One option too restrictive is to choose the point i with the highest average concentration.

$$I^{(k)} = \max_{1 \leq i \leq m} \{ LN[\mu_i, \sigma_i]^{(k)} \} \quad (18)$$

4. APPLICATION

Here is included an example describing how these method can be used in the real world. We have applied this procedure to a grinding workshop in Ceramic Area. The intake for workers moving in this workshop is estimated mainly by using three sampling points (P_1, P_2, P_3) near a equipment. We have also used two additional points: one represents the grinding area concentration, P_A , an other, P_F , represents the time that the worker is out of the Ceramic Area.

We have used the real A_i values of a long period (>400 working days). We have found that daily activity in these points (in mBq/d) can be fitted to the following probability function: $f_1(a) = LN(5.56, 1.26)$, $f_2(a) = LN(5.83, 1.55)$, $f_3(a) = LN(5.01, 0.88)$, $f_A(a) = LN(4.43, 0.68)$, and $f_F(a) = LN(4.1, 0.71)$. The average time of staying in each point are: $w_1 = 0.20$, $w_2 = 0.30$, $w_3 = 0.25$, $w_A = 0.15$, $w_F = 0.10$. We have simulated the daily intake I substituting these functions in eqn(14) making 200 simulation (the number of working days of a year), using and $m = 30$ in each simulation. The solution is shown in figure 2.

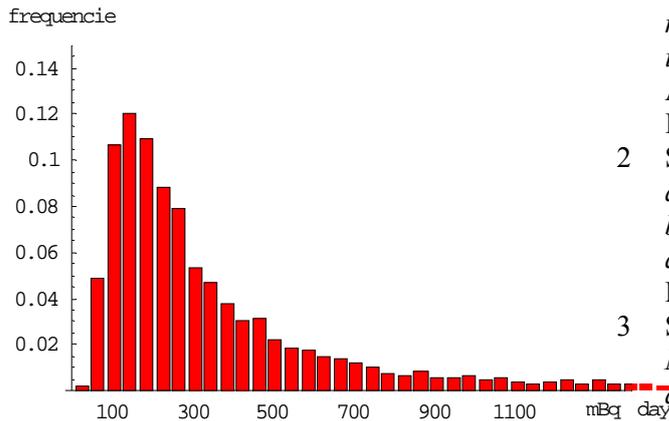


Fig 2 Distribution of daily intake of workers exposed to random concentration

The average annual intake, \bar{I}^D , is 102.8 Bq and the standard deviation, s_D , 17.7 Bq. Substituting in (8) $I_L = 132$ Bq and choosing $L = 150$ Bq (10% ALI for uranium with low enrichment) we obtain $I_L < L$, and therefore the probability to exceed of 10% ALI in this area is $<0.4\%$. According to 10CFR20.1502(b) neither individuals intake estimation are required..

5. CONCLUSIONS

The regulations require that monitoring of intake of radioactive material should be applied depending on radiological hazards but they do not establish precise criteria how these radiological hazards should be evaluated. Using the data from Juzbado Uranium Plant, we have made an statistical study of air sampling concentration and of the intake of a group of workers exposed to inhalation of uranium aerosols. We have found that these can be fitted to the probability distributions. An improved method for normal and lognormal censured distribution is used to take into account lower level detection (LLD) values. Applying these distributions we have used simulation techniques to evaluate the hazards from intake. All these methods can be useful for the design and the conduct of the air control monitoring and to classify the risk for an specific workshop.

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