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Pascal Petit, Dominique J. Bicout

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1 **Health risk assessment with multiple reference indices**

2

3 Pascal Petit<sup>a,\*</sup>, Dominique J. Bicot<sup>a</sup>

4 <sup>a</sup> Univ. Grenoble Alpes, CNRS, Grenoble INP, VetAgro Sup, TIMC, 38000 Grenoble, France

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6 \* Corresponding authors at: Laboratoire TIMC-IMAG, Equipe EPSP, Faculté de Médecine de Grenoble,

7 Domaine de La Merci, 38706 La Tronche Cedex, France.

8 Tel: +33(0) 4 76 63 75 06

9 E-mail address: [pascal.petit@univ-grenoble-alpes.fr](mailto:pascal.petit@univ-grenoble-alpes.fr)

10 **ABSTRACT**

11 *Background:* Conducting a risk assessment can be challenging, especially when dealing with several  
12 reference indices, which could lead to conflicting conclusions between studies. The common approach  
13 is to use a reference index from a single source based on the risk assessor's preference.

14 *Objectives:* To propose an approach for constructing a multi-reference index-based aggregated risk  
15 estimate using mathematical objectivity to reflect all of the available information.

16 *Methods:* The aggregated risk estimate based on multiple reference indices (AREMRI) results from  
17 the weighted linear combination of risk distributions that were obtained with each reference index  
18 available. The weights were calculated using the degree of agreement among the reference index-  
19 based risk distributions. The approach is illustrated through three practical cases of benzene inhalation  
20 cancer risk assessment using inhalation unit risks (IURs) from six different regulatory agencies.

21 *Results:* The degrees of agreement between the reference index-based risk distribution, obtained with  
22 the six IURs, ranged from 0.7 to 92%. The highest weights were attributed to reference index-based  
23 risk distributions that had the highest degree of agreement with the maximum number of other  
24 reference index-based risk distributions. Regardless of the practical case considered, the AREMRI risk  
25 distribution resulted in the third highest risk compared to the six single risk distributions.

26 *Conclusion:* Our approach can be useful in the presence of several reference indices by providing  
27 useful insights, consistency and direct comparisons between studies to support better-informed risk  
28 assessment and management decisions. This approach can shed some light on some of the  
29 uncertainties associated with the toxicological reference values in a risk assessment associated with the  
30 toxicological reference values. If the uncertainty is large, more detailed evaluation of the toxicological  
31 reference values would be needed.

32

33 **KEYWORDS**

34 Health risk assessment; probabilistic risk assessment; risk aggregation; cancer risk

## 35 **1. Introduction**

36 Risk assessments are paramount in ensuring the protection of human health against environmental  
37 pollutants (Rotter et al. 2018). Performing a risk assessment requires the use of toxicity reference  
38 indices. Reference indices are derived using rigorous methodological weights of evidence approaches  
39 by regulatory agencies based on available data generated in experimental toxicology and exposure  
40 assessment. For a given absorption pathway (oral, respiratory, cutaneous), dates when the index values  
41 were derived, the method of derivation, and knowledge or assumptions about substances' mechanisms  
42 of action are often different from one organization/agency/study to another (Petit et al. 2020). These  
43 differences arise mainly from the use of different data sources/key studies (e.g., the results from  
44 animal studies or epidemiological studies), extrapolation methods, correction factors, and commonly  
45 accepted assumptions and uncertainties (ANSES 2015; Beck et al. 2016; Melnick et al. 2008; US EPA  
46 2014). Even when similar studies are considered for the same critical health effects and absorption  
47 pathway, organizations/agencies/studies can reach different conclusions due to different policies  
48 and/or expert judgments (Beck et al. 2016). Depending on the reference index chosen, the results of  
49 the risk assessment can thus differ and lead to different conclusions and decisions, which may have  
50 broad and profound implications regarding the prevention of adverse effects and the protection of  
51 human health and the environment.

52 The double question that arises when carrying out a risk assessment is knowing which reference index  
53 to use and how to take into account the information provided by the other indices. Indeed, there are no  
54 criteria or benchmark/gold standard approaches guiding the choice of an index, especially since, by  
55 definition, all reference indices available from health/regulatory agencies are legitimate. In practice,  
56 we tend to use either an index from our own home country agency (when available) or an index from  
57 an agency to which we subjectively attach a certain importance. Thus, in this paper, we explored and  
58 suggest an alternative approach, which would be to construct and use an aggregated risk estimate  
59 based on multiple reference indices (AREMRI) created from mathematical objectivity. Such an  
60 AREMRI would allow for direct comparisons among studies, promote transparency and ensure a  
61 consistency of analysis during risk assessment. The aim here is not to judge how researchers and/or

62 regulatory agencies derive reference indices or to criticize existing reference indices but rather to use  
 63 them as is, starting from the fact that each of the indices carries a part of the truth.

64 The remainder of this paper is organized as follows. The general approach for constructing an  
 65 AREMRI risk distribution (uncertain risks) is presented in Section 2. Section 3 is devoted to an  
 66 illustrative application of the developed approach to inhalation cancer risks. Finally, the last section,  
 67 Section 4, discusses the proposed approach and considers the practical implications and  
 68 implementation of the developed analysis in an interactive web application.

69

## 70 **2. Methods**

### 71 *2.1 Aggregated risk distribution construction based on multiple reference indices (AREMRI)*

72 Consider a discrete ensemble of  $n$  indicators of uncertain risk  $r$ , each based on a reference index from  
 73 a different source (e.g., organization/agency/study), and each being described by a distribution  
 74 function normalized to one,  $R_i(r)$ , with  $i = 1, 2, \dots, n$ . It is assumed that all reference indices from  $n$   
 75 sources are legitimate and relevant (developed for the same purpose) and that there is no objective  
 76 and/or impartial reason to choose one over another for the risk assessment. As a first step, the  
 77 ensemble of  $n$  indicators can be characterized by the matrix of degrees of agreement,  $\mathbf{A}$ , given by (Eq.  
 78 1):

$$79 \quad \mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{pmatrix} = \begin{pmatrix} 1 & a_{12} & \cdots & a_{1n} \\ a_{21} & 1 & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & 1 \end{pmatrix} \quad (1)$$

80 where  $a_{ij} = a_{ji}$  is the degree of agreement between two reference index-based distributions  
 81 (indicators)  $i$  and  $j$  such that  $0 \leq a_{ij} = a_{ji} \leq 1$  and  $a_{ii} = 1$ .  $a_{ij} = 0$  refers to no agreement, while  
 82  $a_{ij} = 1$  indicates total agreement between the two reference index-based risk distributions. A  
 83 reference index-based distribution is defined as a risk distribution resulting from the use of a  
 84 given/source-specific reference index (e.g., oral slope factor from US EPA). Mathematically,  $a_{ij} =$   
 85  $\text{overlap}[R_i, R_j]$  is obtained by calculating the overlap between two (normalized to one) distribution

86 functions  $R_i(r)$  and  $R_j(r)$ ; a general method for calculating the overlap (or overlapping area) between  
 87 two functions is described elsewhere (Petit et al. 2020).

88 The content of  $\mathbf{A}$  is very informative to compare reference index-based risk distribution with each  
 89 other and also to guide the development of a pragmatic choice of one reference index over the others,  
 90 in possible combination with other techniques such as stochastic dominance or expected utility  
 91 (Verteramo Chiu et al. 2020). Whichever approach is used, the result would be to choose the indicator  
 92 (i.e., reference index) from a single source (e.g., regulatory agency/study). Consequently, we are rather  
 93 in favor of constructing a composite indicator resulting from the aggregation of all indicators to reflect  
 94 all of the available information. Our aim is to develop such an aggregation based on the content of  $\mathbf{A}$   
 95 and, in this case, there will be no aggregation when  $\mathbf{A} = \mathbf{1}$  (identity matrix), i.e., when none of the  
 96 reference index-based distributions would agree with the others.

97 To this end, we propose to construct an aggregated risk distribution based on multiple reference  
 98 indices (AREMRI),  $R_a(r)$ , as a weighted linear combination of reference index-based risk  
 99 distributions  $R_i(r)$ , as follows (Eq. 2):

$$100 \quad R_a(r) = \sum_{i=1}^n \omega_i R_i(r) \quad (2)$$

101 where  $\omega_i$  ( $0 \leq \omega_i \leq 1$  such that  $\sum_{i=1}^n \omega_i = 1$ ) is a contributing weight of the risk distribution  
 102 generated with the reference index “ $i$ ” and  $R_i(r)$  is a risk distribution based on a single given  
 103 reference index. Next, to determine the  $\omega_i$  coefficients and obtain an objective aggregated distribution,  
 104 we require that  $R_a(r)$  has a maximum agreement with all risk distributions generated with all  
 105 reference indices available  $R_i(r)$ , i.e., to maximize total agreement  $\theta$  between  $R_a(r)$  and all  $R_i(r)$ .

106 The total agreement  $\theta$  is related to the agreement matrix  $\mathbf{A}$  as (Eq. 3):

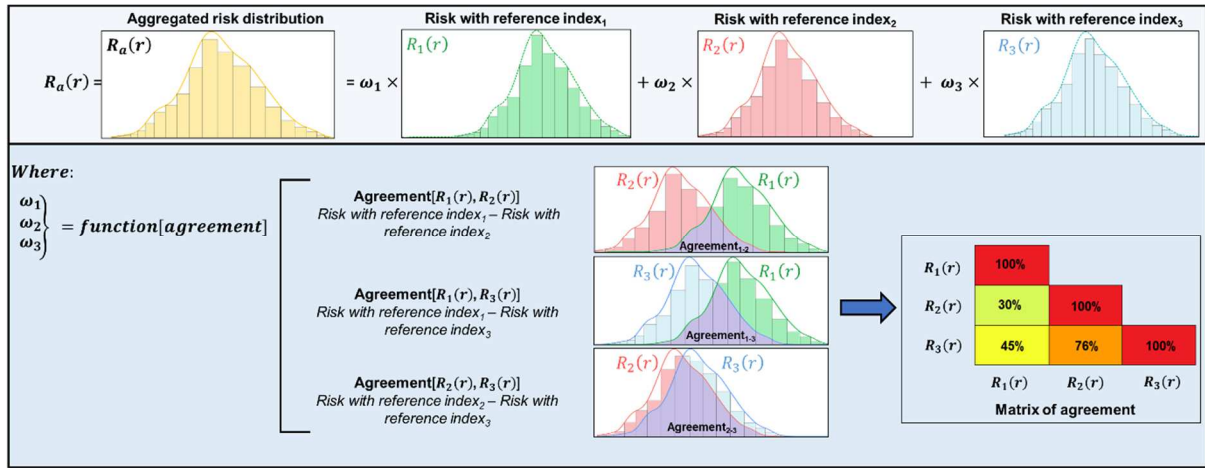
$$107 \quad \theta = \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix}^T \underbrace{\begin{pmatrix} 1 & a_{12} & \cdots & a_{1n} \\ a_{21} & 1 & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & 1 \end{pmatrix}}_{\mathbf{A}} \begin{pmatrix} \omega_1 \\ \omega_2 \\ \vdots \\ \omega_n \end{pmatrix} \quad (3)$$

108 The agreement matrix  $\mathbf{A}$  is symmetric with the trace,  $\text{Tr}(\mathbf{A}) = \sum_{k=1}^n \lambda_k = n$ , where  $\lambda_k$  are the  
 109 eigenvalues associated with (normalized to one) eigenvectors  $\mathbf{v}_k$  of matrix  $\mathbf{A}$  such that  $0 \leq \lambda_1 < \lambda_2 <$

110  $\dots < \lambda_n \leq n$ . When there is no agreement among all reference index-based risk distributions  $R_i(r)$ ,  $\mathbf{A}$   
 111 reduces to the identity matrix with  $a_{ij} = 0$  for  $i \neq j$ , and the total agreement is always  $\theta = 1$   
 112 regardless of the weights  $\omega_i$ . In this case, no aggregation based on the agreement matrix  $\mathbf{A}$  is possible  
 113 as there is no objective way of choosing one reference index over another. In the opposite situation  
 114 when all  $a_{ij} = 1$ , i.e., perfect agreement among all reference index-based risk distributions, the total  
 115 agreement is maximum,  $\theta = n$ . We will show below that, in this case,  $\omega_i = 1/n$  (arithmetic mean).  
 116 Finally, when  $0 < a_{ij} < 1$ , the maximum of the total agreement,  $\theta$ , is obtained by setting the  $\omega_i$   
 117 coefficients equal to eigenvector  $\mathbf{v}_n$  corresponding to the largest eigenvalue  $\lambda_n$  as shown in (Eq. 4):

$$118 \quad \begin{pmatrix} \omega_1 \\ \omega_2 \\ \vdots \\ \omega_n \end{pmatrix} = \frac{1}{[\sum_{i=1}^n v_{n,i}]} \begin{pmatrix} v_{n,1} \\ v_{n,2} \\ \vdots \\ v_{n,n} \end{pmatrix} \Rightarrow \theta = \lambda_n \frac{1}{[\sum_{i=1}^n v_{n,i}]} \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix}^T \begin{pmatrix} v_{n,1} \\ v_{n,2} \\ \vdots \\ v_{n,n} \end{pmatrix} = \lambda_n \quad (4)$$

119 This determines the weights of  $R_a(r)$  in Eq. 2. It is paramount to ensure that all  $v_{n,i}$  have the same  
 120 sign. Fig. 1 presents a schematic summary of the different steps of the construction of the aggregated  
 121 risk distribution based on multiple reference indices.



122  
 123 **Fig. 1:** Scheme of the general approach to construct an aggregated risk distribution based on multiple  
 124 reference indices

125  $\omega_i$ : contributing weight of agency “ $i$ ” with  $0 \leq \omega_i \leq 1$  such that  $\sum_{i=1}^n \omega_i = 1$ ;  $R_a(r)$ : aggregated risk  
 126 distribution function based on multiple reference indices;  $R_i(r)$ : risk distribution function generated  
 127 with reference index “ $i$ ”.

128

129 In case of perfect agreement among all reference index-based risk distributions (indicators), i.e., all  
 130  $a_{ij} = 1$ , the highest eigenvalue of the agreement matrix  $\mathbf{A}$  is  $\lambda_n = n$ , associated with the eigenvector  
 131  $\mathbf{v}_n^T = (1, 1, \dots, 1)^T$ ; this leads to  $\omega_1 = \omega_2 = \dots = \omega_n = 1/n$ .

132 When dealing with only two reference indices (indicators), i.e.,  $n = 2$  and  $a_{12} = a_{21} \neq 0$ , the highest  
 133 eigenvalue of  $\mathbf{A}$  is  $\lambda_2 = 1 + a_{12}$ , associated with the eigenvector  $\mathbf{v}_2^T = (1, 1)^T$ . In this case, the  
 134 weights are  $\omega_1 = \omega_2 = 1/2$ , independent of  $a_{12}$ , provided that  $a_{12} \neq 0$ .

135 To gain more understanding of these weights, consider an example of  $n = 3$  for which weights can be  
 136 calculated explicitly as:

$$137 \quad \mathbf{A} = \begin{pmatrix} 1 & a_{12} & 0 \\ a_{21} & 1 & a_{23} \\ 0 & a_{32} & 1 \end{pmatrix} \Rightarrow \begin{pmatrix} \omega_1 \\ \omega_2 \\ \omega_3 \end{pmatrix} = \frac{1}{[a_{12} + \sqrt{a_{12}^2 + a_{23}^2 + a_{23}}]} \begin{pmatrix} a_{12} \\ \sqrt{a_{12}^2 + a_{23}^2} \\ a_{23} \end{pmatrix} \quad (5)$$

138 where  $a_{12} = a_{21} < 1$  and  $a_{23} = a_{32} < 1$  such that  $a_{13} = a_{31} = 0$ , and the highest eigenvalue of  $\mathbf{A}$  is  
 139  $\lambda_3 = 1 + \sqrt{a_{12}^2 + a_{23}^2}$ . Clearly, Eq. 5 shows how the weights are “proportional or a function” of the  
 140 respective agreements. For instance, setting  $a_{23} = 0$  leads to  $\omega_3 = 0$ , i.e., the contribution of a  
 141 reference index-based risk distribution (indicator) is zero when the reference index-based risk  
 142 distribution does not agree with any other. In addition, the weight of a reference index-based risk  
 143 distribution (indicator) increases with its total agreement with the other reference index-based risk  
 144 distributions (indicators). Indeed, we have  $\omega_3 < \omega_1 < \omega_2$  when  $a_{23} < a_{12}$ , and  $\omega_1 < \omega_3 < \omega_2$  for  
 145  $a_{23} > a_{12}$ . These observations remain valid for  $n > 3$ .

146

## 147 *2.2 Application to inhalation cancer risks*

148

149 To illustrate the usefulness of using an aggregated risk distribution based on multiple reference  
 150 indices, we consider three practical cases/scenarios of risk assessment of benzene exposure where each  
 151 exposure scenario is characterized by a distribution of benzene concentrations assumed constant over  
 152 time and following a lognormal distribution with the geometric mean (GM) and geometric standard  
 153 deviation (GSD) parameters given in Table 1 and an exposure duration of 4 hours a day, 120 days a



154 year for 50 years. Regardless of the reference index considered, the concentration distributions used  
155 for a given scenario were exactly the same (same geometric mean and geometric standard deviation).  
156 The concentration distribution parameters differed from one practical case to another with a GM set to  
157 253, 253 and 58.4  $\mu\text{g}/\text{m}^3$  and a GSD set to 2.80, 1.03 and 2.80 for cases 1, 2 and 3, respectively. For  
158 all scenarios, both the GM and GSD were set at pragmatic random values generated with R software  
159 4.0.5® (R Core Team, Vienna, Austria) for Windows 10©.

160 We used the *InCaRisk* app (Petit et al. 2020) with 1,000 simulations runs to generate all of the  
161 reference index-based risk distributions  $R_i(r)$ . Briefly, the *InCaRisk* app is a free user-friendly and  
162 interactive web application that allows for cancer risk estimation following inhalation exposure  
163 (<https://exporisk-timc.imag.fr/InCaRisk/>). A total of 305 substances/chemicals with reference values  
164 from up to eight agencies are available. *InCaRisk* requires no programming knowledge to use. This  
165 app also incorporates a variety of features and options to make it easy to use, with the possibility of  
166 configuring the exposure settings (exposure scenario), the concentration distribution settings (choice  
167 of the type of distribution, distribution parameters, uploading an exposure concentration file) and to  
168 export the results as a report. This app offers the chance to have an immediate glimpse into the results  
169 and to see how the results change according to different setting configurations with interactive easy-to-  
170 read graphs, which are downloadable.

171 The reference index-based risk distributions  $R_i(r)$  were generated for six sanitary agencies for which  
172 the tumor site (leukemia) and tumor type (hematologic/immune) used in the inhalation unit risk (IUR)  
173 establishment were the same: ANSES (French Agency for Food, Environmental and Occupational  
174 Health & Safety), Health Canada, OEHHA (California Office of Environmental Health Hazard  
175 Assessment), RIVM (National Dutch Institute for Public Health and the Environment), US EPA (US  
176 Environmental Protection Agency) and WHO (World Health Organization). The IURs for benzene set  
177 by these agencies regarding the risk of leukemia for lifetime exposure were  $2.6 \times 10^{-5}$  (ANSES 2019),  
178  $3.3 \times 10^{-6}$  (Health Canada 2010),  $2.9 \times 10^{-5}$  (OEHHA 2019),  $5 \times 10^{-6}$  (RIVM 2001),  $7.8 \times 10^{-6}$  (US EPA  
179 2018) and  $6 \times 10^{-6} \text{ m}^3/\mu\text{g}$  (WHO 2000).

180

181 Table 1 here (see Table 1 after the Reference section).

182 **3. Results**

183 For a given practical case, each IUR-based risk distribution resulted in different cancer estimates even  
184 though the exposure concentration distributions were the same. These differences came solely from  
185 the IUR values that differed from one agency to another (Table 1). Indeed, each IUR-based risk  
186 distribution had the same shape (height and width) because of similar concentration parameters (Table  
187 1) but was shifted from one another (different mean) by a factor that depended on the ratio  
188  $IUR_{agency_i}/IUR_{agency_j}$  (Fig. 2C, 3C and 4C).

189 As seen in the matrix of agreement between sources (agencies) (Fig. 2A, 3A and 4A), the degrees of  
190 agreement were heterogeneous between IUR-based risk distributions, regardless of the practical case  
191 considered. The results were similar for cases 1 and 3. Degrees of agreement were lower for the  
192 second scenario than for the first and third scenarios because the GSDs were the lowest (narrower  
193 distributions) for case 2. For all scenarios, the highest degrees of agreement, 92% for cases 1 and 3  
194 and 81% for case 2, were found between IUR-based risk distributions from RIVM and WHO, while  
195 the lowest degrees of agreement were found between IUR-based risk distributions from Health Canada  
196 and OEHHA (30% for cases 1 and 3, 0.7% for case 2). Regardless of the practical case considered,  
197 IUR-based risk distributions from OEHHA and ANSES had a high degree of agreement with each  
198 other (84% for cases 1 & 3, 80% for case 2), a medium degree of agreement with all other IUR-based  
199 risk distributions for cases 1 & 3 (30 to 56%), and a low degree of agreement with all other IUR-based  
200 risk distributions for case 2 (0.7 to 15%). IUR-based risk distributions from the WHO, US EPA,  
201 Health Canada and RIVM had a medium to high degree of agreement with each other (66 to 92% for  
202 cases 1 & 3 and 30 to 81% for case 2).

203  
204 The weights were similar for cases 1 and 3, ranging from 0.136 to 0.189, and they ranged from 0.053  
205 to 0.247 for case 2 (Fig. 2B, 3B and 4B). The highest weights were attributed to IUR-based risk  
206 distributions that had the highest degree of agreement with the maximum number of other IUR-based  
207 risk distributions. Conversely, IUR-based risk distributions that shared a high degree of agreement  
208 with the smallest number of other IUR-based risk distributions had the smallest weights.

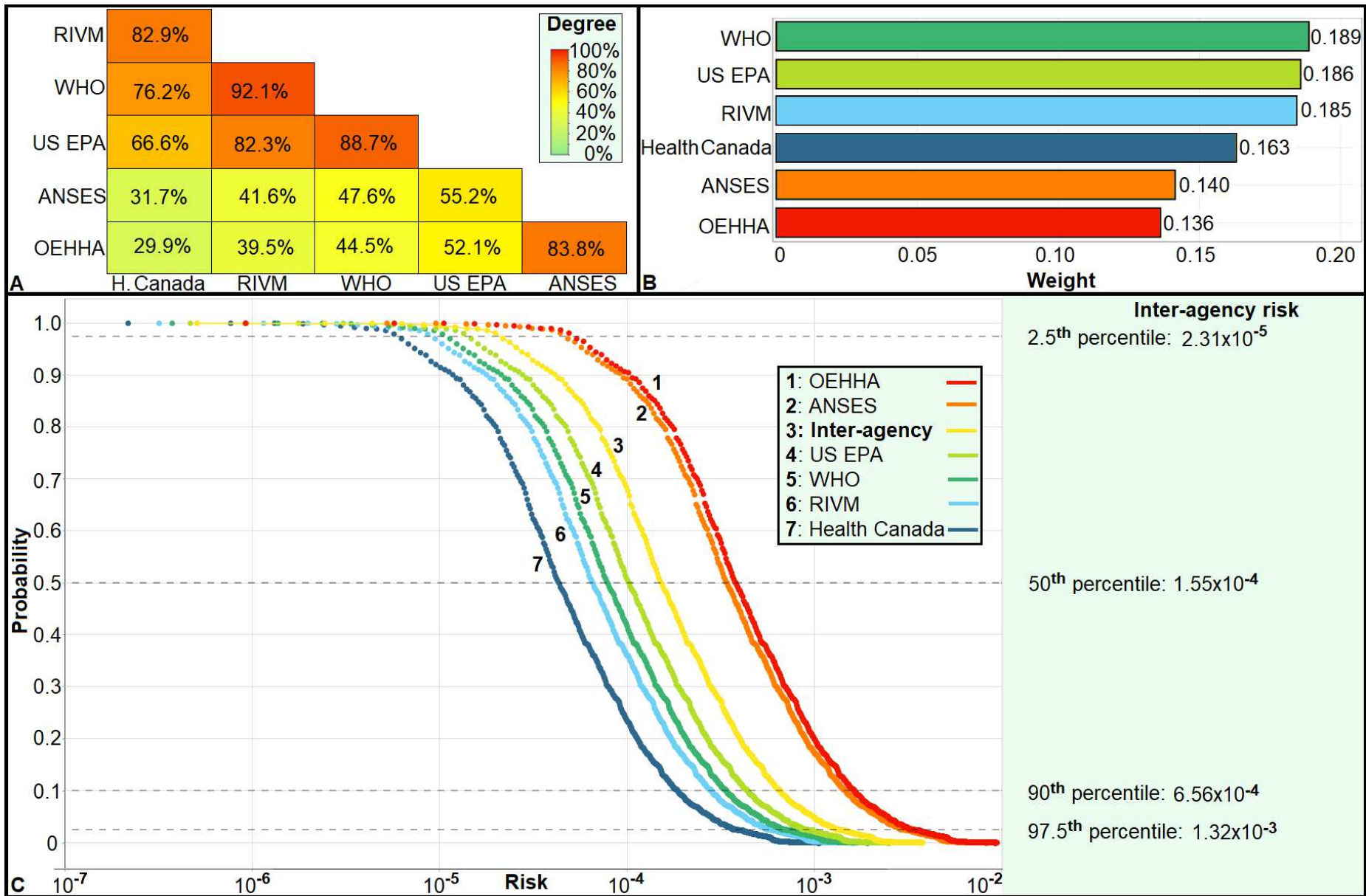
209

210 For the first and third scenarios, the aggregated multi IUR-based risk distribution (AREMRI) followed  
211 a lognormal distribution of parameters  $GM = 1.65 \times 10^{-4}$  and  $GSD = 2.8$  for case 1 and  $GM = 3.82 \times 10^{-5}$   
212 and  $GSD = 2.84$  for case 3. The AREMRI resulted in the third highest risk, with a leukemia risk of  
213  $6.56 \times 10^{-4}$  [IC 95%:  $2.31 \times 10^{-5} - 1.32 \times 10^{-3}$ ] for case 1 and  $1.53 \times 10^{-4}$  [IC 95%:  $5.30 \times 10^{-6} - 3.07 \times 10^{-4}$ ] for  
214 case 3. For both scenarios, these leukemia risks were 2.5 and 2.2 times lower than the cancer risk  
215 estimated with the reference indices (IURs) of OEHHA and ANSES, respectively (Fig. 2C and 4C).  
216 Conversely, the aggregated multi-IUR-based risks were 1.5, 1.9, 2.3 and 3.5 times higher than the  
217 leukemia risk estimated with the reference indices from the US EPA, WHO, RIVM and Health  
218 Canada, respectively. The concentration GSDs used for cases 1 & 3 were the same ( $GSD = 2.80$ ),  
219 which explains why the weights and degrees of agreement were similar for both scenarios. The only  
220 difference between case 1 and case 3 came from the estimated leukemia risk, which was 4.3 times  
221 lower for case 3 than for case 1. This difference corresponded to the ratio of their concentration GM  
222 ( $253 \mu\text{g}/\text{m}^3$  for case 1 vs.  $58.4 \mu\text{g}/\text{m}^3$  for case 3). In other words, for a given concentration GSD,  
223 changing the concentration GM will shift the risk estimate toward higher values when the GM is  
224 increased and toward lower values when the GM is decreased, with little to no change at all of the  
225 degrees of agreement or the weighted values.

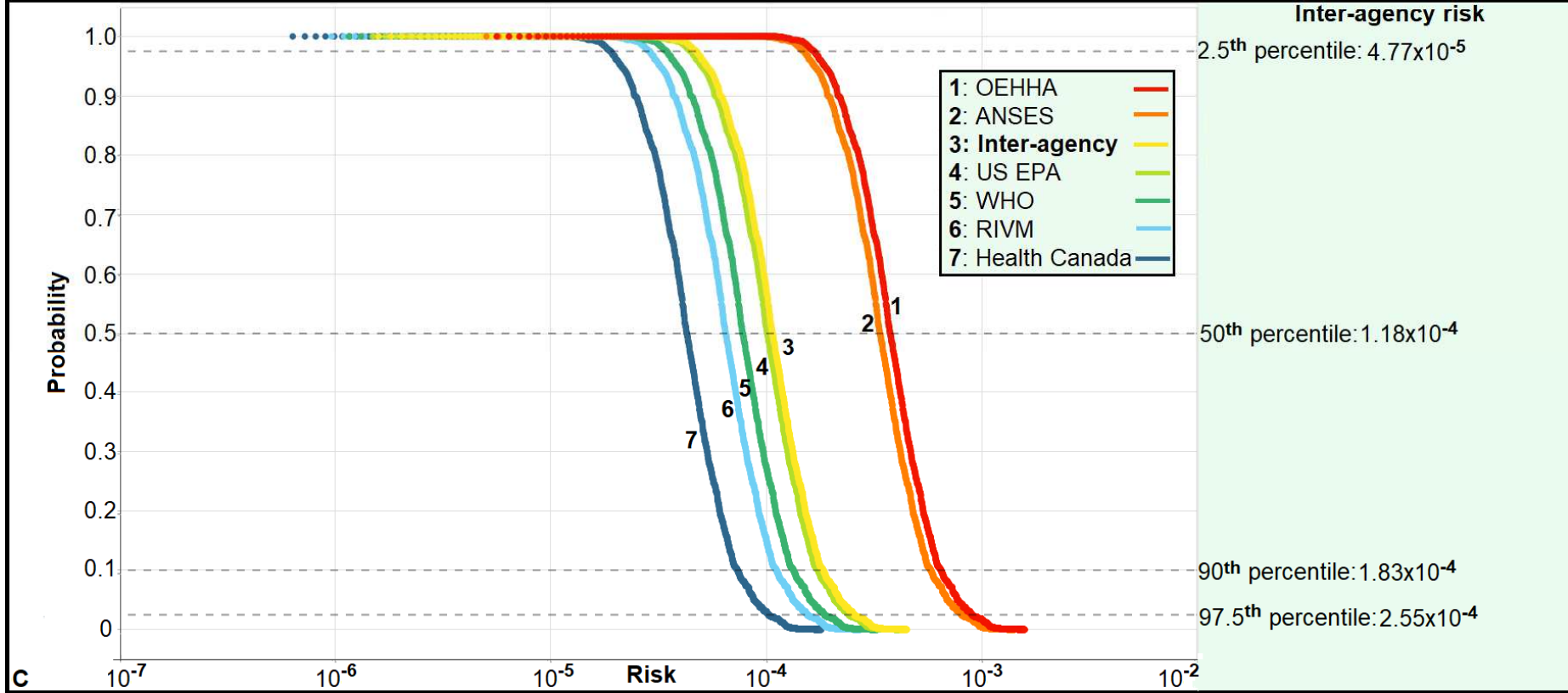
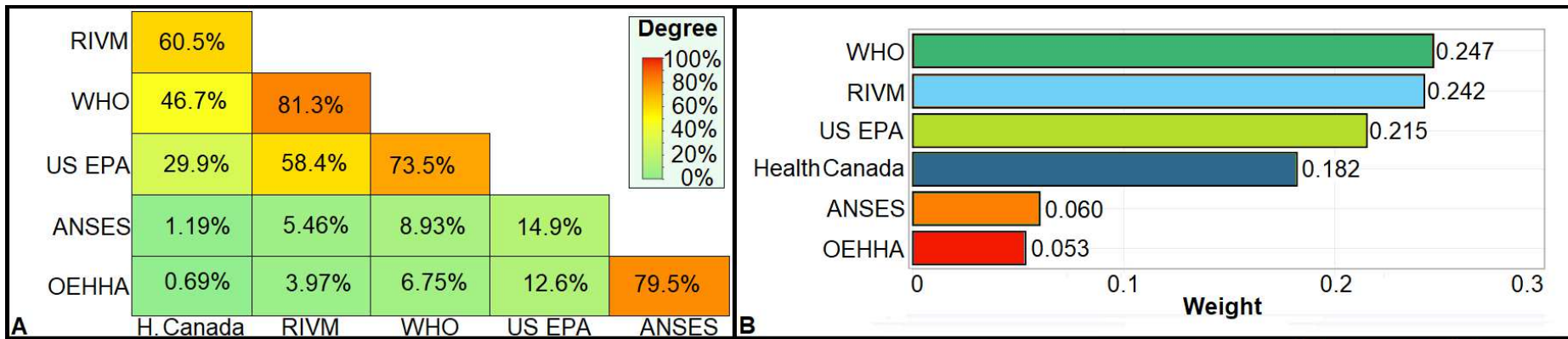
226

227 For the second scenario, the aggregated multi-IUR-based risk distribution (AREMRI) followed a  
228 lognormal distribution of parameters  $GM = 1.08 \times 10^{-4}$  and  $GSD = 1.52$ . The AREMRI resulted in the  
229 third highest risk, with a leukemia risk of  $1.83 \times 10^{-4}$  [IC 95%:  $4.77 \times 10^{-5} - 2.55 \times 10^{-4}$ ], which was 3.5  
230 and 3.1 times lower than the cancer risk estimated with the reference indices (IURs) of OEHHA and  
231 ANSES, respectively (Fig. 3C). Conversely, the aggregated multi-IUR-based risk was 1.1, 1.4, 1.6 and  
232 2.5 times higher than the leukemia risk estimated with the reference indices from the US EPA, WHO,  
233 RIVM and Health Canada, respectively. For this scenario, the GM concentration was the same as that  
234 in the first scenario ( $GM = 253 \mu\text{g}/\text{m}^3$ ), but the GSD concentration was lower (1.03 for case 2 vs. 2.80  
235 for case 1). A lower GSD concentration for case 2 resulted in lower degrees of agreement (Fig. 3A)  
236 and better risk estimation (narrower 95% confidence interval) than for case 1. Indeed, there was a 5.3

237 factor between the lower and upper bounds of the 95% confidence interval of the risk distribution for  
238 case 2 compared to a factor of 57 for case 1.

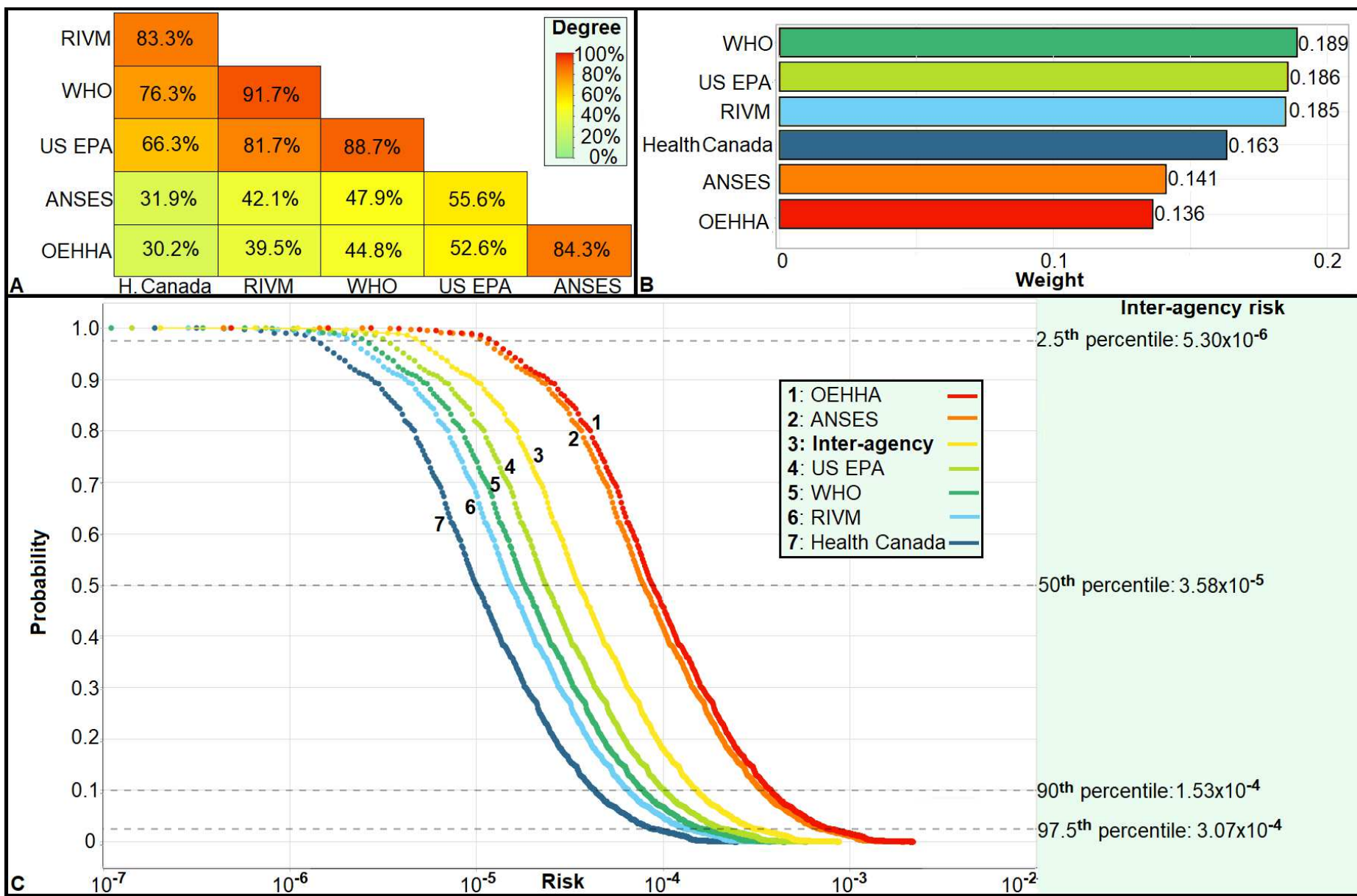


240 **Fig. 2:** Side-by-side comparison of single IUR-based risk distributions for the construction of the aggregated multi-IUR-based risk distribution – case 1  
241 A: Matrix of degrees of agreement between a single IUR-based risk distribution, B: Weights attributed to the IUR-based risk distribution for the aggregated  
242 multi-IUR-based risk distribution construction, C: IUR-based risk distributions.  
243 ANSES: French Agency for Food, Environmental and Occupational Health & Safety, H. Canada: Health Canada, OEHHA: California Office of  
244 Environmental Health Hazard Assessment, RIVM: National Dutch Institute for Public Health and the Environment, US EPA: US Environmental Protection  
245 Agency, WHO: World Health Organization. Each agency name refers to the risk distribution generated using its benzene IUR and the exposure scenario  
246 considered. The term “Inter-agency” refers here to the aggregated risk estimate based on multiple reference indices (AREMRI).



248 **Fig. 3:** Side-by-side comparison of single IUR-based risk distributions for the construction of the aggregated multi-IUR-based risk distribution – case 2  
249 A: Matrix of degrees of agreement between a single IUR-based risk distribution, B: Weights attributed to the IUR-based risk distribution for the aggregated  
250 multi-IUR-based risk distribution construction, C: IUR-based risk distributions.  
251 ANSES: French Agency for Food, Environmental and Occupational Health & Safety, H. Canada: Health Canada, OEHHA: California Office of  
252 Environmental Health Hazard Assessment, RIVM: National Dutch Institute for Public Health and the Environment, US EPA: US Environmental Protection  
253 Agency, WHO: World Health Organization. Each agency name refers to the risk distribution generated using its benzene IUR and the exposure scenario  
254 considered. The term “Inter-agency” refers here to the aggregated risk estimate based on multiple reference indices (AREMRI).





256 **Fig. 4:** Side-by-side comparison of single IUR-based risk distributions for the construction of the aggregated multi-IUR-based risk distribution – case 3  
257 A: Matrix of degrees of agreement between a single IUR-based risk distribution, B: Weights attributed to the IUR-based risk distribution for the aggregated  
258 multi-IUR-based risk distribution construction, C: IUR-based risk distributions.  
259 ANSES: French Agency for Food, Environmental and Occupational Health & Safety, H. Canada: Health Canada, OEHHA: California Office of  
260 Environmental Health Hazard Assessment, RIVM: National Dutch Institute for Public Health and the Environment, US EPA: US Environmental Protection  
261 Agency, WHO: World Health Organization. Each agency name refers to the risk distribution generated using its benzene IUR and the exposure scenario  
262 considered. The term “Inter-agency” refers here to the aggregated risk estimate based on multiple reference indices (AREMRI).

#### 263 **4. Discussion**

264 Performing a risk assessment requires the use of toxicity reference indices. When dealing with several  
265 reference indices, risk assessment can be challenging, as there is no objective way to choose one  
266 reference index over the others. By definition, all indices available from sanitary/regulatory agencies  
267 are legitimate, but in practice, we tend to use an index from a source or agency to which we  
268 subjectively attach a certain importance. In this paper, we propose a novel approach consisting of  
269 constructing and using an aggregated multi-reference index-based risk estimate constructed only on  
270 mathematical objectivity, assuming that the reference indices considered are all relevant and suitable  
271 for the risk assessment to be conducted. An R code allowing for the estimation of risks (AREMRI) by  
272 inputting multiple reference indices is provided in the Supplemental Material.

273 The construction of the aggregated risk distribution was based on the weighted linear aggregation  
274 (arithmetic mean) of the probability risk distribution generated using a given reference index and using  
275 the degree of agreement among the reference index-based risk distributions for the assignment of  
276 weight. Other approaches could be used, such as stochastic dominance, which is a form of stochastic  
277 ordering that would favor one reference index-based risk distribution over the others based on a  
278 specific criterion. For example, using stochastic dominance, it would be possible to favor only the risk  
279 distribution resulting from the source or agency with the highest reference index, that is to say the  
280 most protective, or with the lowest reference index (weight of 1 while a weight of 0 for all the other  
281 reference index-based risk distribution) (Verteramo Chiu et al. 2020). That kind of approach is,  
282 however, different from our goal, which was not to judge or provide arguments for choosing a  
283 reference index over the others but rather to objectively construct an aggregated risk indicator that  
284 takes into account several reference indices from different sources (e.g., sanitary agencies), as all  
285 reference indices are legitimate and carry part of the truth.

286 The proposed approach was illustrated using three practical cases where the exposure concentration  
287 distributions were assumed to follow lognormal distributions. Our approach is not limited to  
288 lognormal distributions and can work with any type of distribution. Two additional examples are  
289 available in the Supplemental Material (Fig. S1 and S2), with exposure concentrations following  
290 normal and Poisson distributions, respectively.

291 The approach consisting of constructing and using an aggregated multi-reference index-based risk  
292 estimate was illustrated for one kind of stochastic reference value (inhalation cancer unit risk). The  
293 presented approach could be used with other types of toxicological reference values, such as oral slope  
294 factors, drinking water unit risks or threshold/deterministic values. For some toxicological reference  
295 values, the distribution of the underlying toxicity data is available, and the reference value itself can  
296 sometimes correspond to a certain point on that distribution (e.g., for BMDL values). In that particular  
297 case, these reference values could potentially be set as distributions instead of single values using the  
298 distribution of the underlying toxicity data when available. While the presented approach can still be  
299 used with reference value distributions, it should be noted that the choice of the reference value  
300 distribution (e.g., gamma, lognormal, Poisson, Weibull) and the associated uncertainties (confidence  
301 intervals) will both have an important impact on the resulting risk estimate. In particular, in the  
302 presence of large confidence intervals (high data dispersion), the degrees of agreement (overlaps  
303 between distributions of risks) are so high that all of the reference indices contribute almost equally to  
304 the aggregate distribution and the notion of multi-reference indices becomes less relevant. This issue is  
305 particularly true with deterministic values (threshold values), for which the choice of the derivation  
306 method and, in particular, the choice and use of large safety factors may dominate the differences  
307 between the toxicological reference values. One possible solution to address this issue would be to  
308 replace traditional noncancer reference doses (threshold values) with probabilistic estimates (Chiu et  
309 al. 2018).

310 Regardless of the practical case considered, IUR-based risk distributions from OEHHA and ANSES  
311 had a high degree of agreement with each other but a medium to low degree of agreement with other  
312 IUR-based risk distributions. The six benzene reference values used in this illustration were derived  
313 for the same purpose (lifetime benzene exposure), all based on cohort studies of benzene-exposed  
314 workers, but using different key studies and extrapolation/derivation methods among agencies, which  
315 may explain the observed differences (ANSES 2019; Health Canada 2010; OEHHA 2019; RIVM  
316 2001; US EPA 2018; WHO 2000). OEHHA and ANSES used the same extrapolation method, a  
317 relatively risk procedure, but chose two different key studies, Rinsky et al. (1981) for OEHHA and  
318 Richardson (2008) for ANSES. RIVM chose the same key study as OEHHA but used a nonthreshold

319 extrapolation method. The WHO used an average relative risk model with Crump (1994) as a key  
320 study. The US EPA applied a low-dose linearity utilizing the maximum likelihood estimates method  
321 based on six key studies (Crump and Allen 1984; Crump 1994; Paustenbach et al. 1993; Rinsky et al.  
322 1981, 1987; US EPA 1998). Health Canada applied a linear quadratic model of the exposure response  
323 relationship based on four key studies (Bond et al. 1986; Rinsky et al. 1987; Wong 1987a,b).  
324 Differences in policies and/or expert judgments could also have played a role in the observed  
325 differences (Beck et al. 2016).

326 The aggregated multi-reference index-based distribution (the degrees of agreement and weights)  
327 depends on the number of reference indices included in the aggregation as well as the concentration  
328 distribution parameters. The narrower the concentration distributions (lower GSD), the lower the  
329 degrees of agreement and the better the risk distribution estimation (narrower confidence interval). If  
330 the GSD concentration tends to 1, as in the second scenario, the AREMRI tends to the mean risk value  
331 and also tends to approach and equal a single reference index-based risk distribution (e.g., the risk  
332 distribution using the reference value from the US EPA for case 2).

333 Regarding the reference indices, we found, for our practical cases, six agencies that set/proposed an  
334 IUR for benzene and the risk of leukemia. There are most likely other agencies and studies around the  
335 world which we are not aware of that may also have proposed an IUR for benzene and the risk of  
336 leukemia for lifetime exposure. Integrating these potential other IURs into the aggregated multi-  
337 reference index-based risk distribution construction would change the leukemia risk obtained with the  
338 aggregated risk indicator. However, the approach presented in this paper does not remove the  
339 requirement of a prior screening of reference values to be included in the integrated approach. This  
340 screening should always be made to determine how reference indices were derived, which data they  
341 are based on, and their intended use (e.g., assessment of occupational exposure, exposure via  
342 foodstuffs and unintentional exposure). Indeed, it is paramount that the risk assessor has some  
343 knowledge regarding the reference values used in an assessment and understands the way the values  
344 were derived, their associated uncertainties and their relevance and validity for the risk assessment  
345 being carried out. If the approach presented in this paper is used to interpret the results of a risk

346 assessment, then only reference values derived for the same purpose(s) should be included in the  
347 analysis.

348

## 349 **5. Conclusions**

350 The three practical cases highlighted the usefulness of comparing risk estimates generated using  
351 multiple reference indices from several sources (e.g., regulatory agencies) when assessing the human  
352 health risk associated with exposure to environmental pollutants.

353 Therefore, the proposed approach hinges on its ability to provide useful information when conducting  
354 risk assessment in the presence of several reference indices from various sources. These results are  
355 essential for risk managers, practitioners, and decision-makers by providing guidance and by  
356 supporting direct comparisons between studies. Indeed, our approach can provide a more complete,  
357 consistent and transparent risk assessment of pollutants that pose a threat to human health and support  
358 better-informed risk management decisions by giving some insight into the uncertainties associated  
359 with the choice of the reference index in the risk assessment. Our approach could also serve to  
360 integrate risk assessments, such as by obtaining an integrated view of risk assessments when trying to  
361 prioritize among different cases for remedial action.

362 To facilitate the use of our approach, the matrix of degrees of agreement, side-by-side comparison of  
363 reference index-based risk distributions, the choice of the reference values to consider and the  
364 construction of the aggregated multi-reference index-based risk distribution will be implemented in the  
365 *InCaRisk* app for inhalation cancer risk assessment. In addition to making information more readily  
366 understood and retained in a quicker time, the visual aids and representations provided by *InCaRisk*  
367 could be useful for improving risk communication and promoting transparency among studies when  
368 conducting risk assessments (Beck et al. 2016).

369 Regardless of the type of exposure (e.g., acute, subchronic, or chronic), absorption pathways, exposure  
370 setting (environmental or occupational), stages of the life cycle of a chemical (e.g., manufacture, use,  
371 disposal, consumer products or waste) and stages of the life of an individual (e.g., adulthood), the  
372 approach presented in this paper can be used. However, this approach can only be used with reference  
373 indices intended for the same purpose (e.g., the same absorption pathway, same type of exposure and

374 same exposure setting), and some prior screening of reference indices has to be done to ensure that the  
375 aggregated approach does not look at a mixture of reference values that should not be integrated.  
376 Finally, the proposed approach has applicability in other fields and studies with similar problematics,  
377 as long as there is some constraint on the choice of reference values to be included.

378

#### 379 *Author contributions*

380 **Pascal Petit:** Conceptualization, Methodology, Software, Formal analysis, Investigation, Data  
381 Curation, Writing - Original Draft, Visualization. **Dominique J. Bicot:** Conceptualization,  
382 Methodology, Validation, Writing - Review & Editing.

383

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464 **Tables**

465

466 **Table 1:** Summary of the lognormal exposure concentration distribution parameters used for three  
 467 practical cases

Reference index source	Reference index	Case 1	Case 2	Case 3
	IUR (m <sup>3</sup> /μg)	GM; GSD	GM; GSD	GM; GSD
Using the reference index from ANSES	2.6x10 <sup>-5</sup>	253 μg/m <sup>3</sup> ; 2.80	253 μg/m <sup>3</sup> ; 1.03	58.4 μg/m <sup>3</sup> ; 2.80
Using the reference index from Health Canada	3.3x10 <sup>-6</sup>	253 μg/m <sup>3</sup> ; 2.80	253 μg/m <sup>3</sup> ; 1.03	58.4 μg/m <sup>3</sup> ; 2.80
Using the reference index from OEHHA	2.9x10 <sup>-5</sup>	253 μg/m <sup>3</sup> ; 2.80	253 μg/m <sup>3</sup> ; 1.03	58.4 μg/m <sup>3</sup> ; 2.80
Using the reference index from RIVM	5.0x10 <sup>-6</sup>	253 μg/m <sup>3</sup> ; 2.80	253 μg/m <sup>3</sup> ; 1.03	58.4 μg/m <sup>3</sup> ; 2.80
Using the reference index from US EPA	7.8x10 <sup>-6</sup>	253 μg/m <sup>3</sup> ; 2.80	253 μg/m <sup>3</sup> ; 1.03	58.4 μg/m <sup>3</sup> ; 2.80
Using the reference index from WHO	6.0x10 <sup>-6</sup>	253 μg/m <sup>3</sup> ; 2.80	253 μg/m <sup>3</sup> ; 1.03	58.4 μg/m <sup>3</sup> ; 2.80

468 *Note:* GM: geometric mean; GSD: geometric standard deviation, IUR: inhalation unit risk, ANSES:  
 469 French Agency for Food, Environmental and Occupational Health & Safety, H. Canada: Health  
 470 Canada, OEHHA: California Office of Environmental Health Hazard Assessment, RIVM: National  
 471 Dutch Institute for Public Health and the Environment, US EPA: US Environmental Protection  
 472 Agency, WHO: World Health Organization.

473 **Figure captions**

474

475 **Fig. 1:** Scheme of the general approach to construct an aggregated risk distribution based on multiple  
476 reference indices

477  $\omega_i$ : contributing weight of agency “*i*” with  $0 \leq \omega_i \leq 1$  such that  $\sum_{i=1}^n \omega_i = 1$ ;  $R_a(r)$ : aggregated risk  
478 distribution function based on multiple reference indices;  $R_i(r)$ : risk distribution function of agency  
479 “*i*”.

480

481 **Fig. 2:** Side-by-side comparison of single IUR-based risk distributions for the construction of the  
482 aggregated multi-IUR-based risk distribution – case 1

483 A: Matrix of degrees of agreement between a single IUR-based risk distribution, B: Weights attributed  
484 to the IUR-based risk distribution for the aggregated multi-IUR-based risk distribution construction,  
485 C: IUR-based risk distributions.

486 ANSES: French Agency for Food, Environmental and Occupational Health & Safety, H. Canada:  
487 Health Canada, OEHHA: California Office of Environmental Health Hazard Assessment, RIVM:  
488 National Dutch Institute for Public Health and the Environment, US EPA: US Environmental  
489 Protection Agency, WHO: World Health Organization. Each agency name refers to the risk  
490 distribution generated using its benzene IUR and the exposure scenario considered. The term “Inter-  
491 agency” refers here to the aggregated risk estimate based on multiple reference indices (AREMRI).

492

493 **Fig. 3:** Side-by-side comparison of single IUR-based risk distributions for the construction of the  
494 aggregated multi-IUR-based risk distribution – case 2

495 A: Matrix of degrees of agreement between a single IUR-based risk distribution, B: Weights attributed  
496 to the IUR-based risk distribution for the aggregated multi-IUR-based risk distribution construction,  
497 C: IUR-based risk distributions.

498 ANSES: French Agency for Food, Environmental and Occupational Health & Safety, H. Canada:  
499 Health Canada, OEHHA: California Office of Environmental Health Hazard Assessment, RIVM:  
500 National Dutch Institute for Public Health and the Environment, US EPA: US Environmental

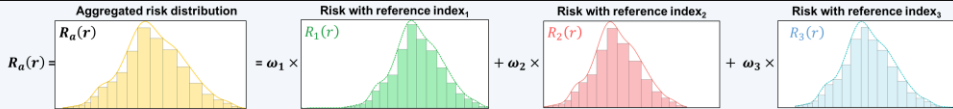
501 Protection Agency, WHO: World Health Organization. Each agency name refers to the risk  
502 distribution generated using its benzene IUR and the exposure scenario considered. The term “Inter-  
503 agency” refers here to the aggregated risk estimate based on multiple reference indices (AREMRI).

504

505 **Fig. 4:** Side-by-side comparison of single IUR-based risk distributions for the construction of the  
506 aggregated multi-IUR-based risk distribution – case 3

507 A: Matrix of degrees of agreement between a single IUR-based risk distribution, B: Weights attributed  
508 to the IUR-based risk distribution for the aggregated multi-IUR-based risk distribution construction,  
509 C: IUR-based risk distributions.

510 ANSES: French Agency for Food, Environmental and Occupational Health & Safety, H. Canada:  
511 Health Canada, OEHHA: California Office of Environmental Health Hazard Assessment, RIVM:  
512 National Dutch Institute for Public Health and the Environment, US EPA: US Environmental  
513 Protection Agency, WHO: World Health Organization. Each agency name refers to the risk  
514 distribution generated using its benzene IUR and the exposure scenario considered. The term “Inter-  
515 agency” refers here to the aggregated risk estimate based on multiple reference indices (AREMRI).



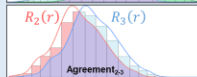
Where:

$$\left. \begin{matrix} \omega_1 \\ \omega_2 \\ \omega_3 \end{matrix} \right\} = \text{function}[\text{agreement}]$$

**Agreement** $[R_1(r), R_2(r)]$   
Risk with reference index<sub>1</sub> – Risk with reference index<sub>2</sub>

**Agreement** $[R_1(r), R_3(r)]$   
Risk with reference index<sub>1</sub> – Risk with reference index<sub>3</sub>

**Agreement** $[R_2(r), R_3(r)]$   
Risk with reference index<sub>2</sub> – Risk with reference index<sub>3</sub>



$R_1(r)$	100%		
$R_2(r)$	30%	100%	
$R_3(r)$	45%	76%	100%
	$R_1(r)$	$R_2(r)$	$R_3(r)$

**Matrix of agreement**