

Accepted refereed manuscript of:

Varley A, Tyler A, Dowdall M, Bondar Y & Zabrotski V (2017) An in situ method for the high resolution mapping of ^{137}Cs and estimation of vertical depth penetration in a highly contaminated environment, *Science of the Total Environment*, 605-606, pp. 957-966.

DOI: [10.1016/j.scitotenv.2017.06.067](https://doi.org/10.1016/j.scitotenv.2017.06.067)

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1 **An *In situ* Method for the High Resolution Mapping of ¹³⁷Cs and Estimation of Vertical**
2 **Depth Penetration in a Highly Contaminated Environment.**

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35 **Abstract**

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37 The Chernobyl nuclear power plant meltdown has to date been the single largest release of
38 radioactivity into the environment. As a result, radioactive contamination that poses a
39 significant threat to human health still persists across much of Europe with the highest
40 concentrations associated with Belarus, Ukraine and western Russia. Of the radionuclides
41 still prevalent with these territories ^{137}Cs presents one of the most problematic remediation
42 challenges. Principally, this is due to the localised spatial and vertical heterogeneity of
43 contamination within the soil (~ 10 's of meters), thus making it difficult to accurately
44 characterise through conventional measurement techniques such as static *in situ* gamma-ray
45 spectrometry or soil cores. Here, a practical solution has been explored, which utilises a large
46 number of short-count time spectral measurements made using relatively inexpensive,
47 lightweight, scintillators (sodium iodide and lanthanum bromide). This approach offers the
48 added advantage of being able to estimate activity and burial depth of ^{137}Cs contamination in
49 much higher spatial resolution compared to traditional approaches. During the course of this
50 work, detectors were calibrated using the Monte Carlo simulations and depth distribution was
51 estimated using the peak-to-valley ratio. Activity and depth estimates were then compared to
52 five reference sites characterised using soil cores. Estimates were in good agreement with the
53 reference sites, differences of $\sim 25\%$ and $\sim 50\%$ in total inventory were found for the three
54 higher and two lower activity sites, respectively. It was concluded that slightly longer count
55 times would be required for the lower activity ($< 1 \text{ MBq m}^{-2}$) sites. Modelling and reference
56 site results suggest little advantage would be gained through the use of the substantially more
57 expensive lanthanum bromide detector over the sodium iodide detector. Finally, the potential
58 of the approach was demonstrated by mapping one of the sites and its surrounding area in
59 high spatial resolution.

60

61 **Keywords**

62 **Cs-137, *in situ*, field gamma-ray spectrometry, peak-to-valley, PVT, Chernobyl**

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69 **Highlights**

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71 • Chernobyl-derived ^{137}Cs poses a risk to human health.

72 • Highly heterogeneous contamination within tens of kilometres of Chernobyl particularly
73 difficult to characterise.

74 • Numerous short count time gamma-ray measurements can address spatial resolution issues.

75 • Peak-to-valley method provides improved depth and activity estimates.

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77 **1. Introduction**

78
79 The accident that occurred at the Chernobyl Nuclear Power Plant (ChNPP) in April 1986
80 resulted in the radioactive contamination of wide areas of Europe and, to a greater extent, the
81 territories of Belarus, Ukraine and the western Russian Federation. In areas of south Belarus
82 the deposition exceeded 1500 kBq/m² ¹³⁷Cs, 3 orders of magnitude greater than that which
83 resulted from atmospheric nuclear weapons as of 1986 (UNSCEAR, 2000). As a result of the
84 accident, a total area exceeding 60,000 km² of forested land was contaminated, including over
85 20,000 km² of the Gomel and Mogilev regions of Belarus in addition to areas near Kiev and in
86 the Russian Federations' Bryansk region. The primary contaminants as a result of the accident
87 were ¹³⁷Cs, ⁹⁰Sr and ²³⁹Pu. Highest levels of contamination, in excess of 1.4 MBq/m² of ¹³⁷Cs
88 occur with the southernmost parts of the reserves territory over lands nearest the Chernobyl
89 reactor while localised areas of high contamination can be found distributed throughout the
90 reserve. Sr-90 is present at densities up to 3 MBq/m² while ²⁴¹Am levels have approximately
91 doubled over the past twenty years and now approach 0.1 MBq/m² in places as a result of
92 ingrowth from ²⁴¹Pu (Smith and Beresford, 2002). Actinides such as ²³⁸Pu and ^{239,240}Pu are
93 present at levels up to 0.037 MBq/m² and 0.074 MBq/m² respectively. The Polessie State
94 Radioecology Reserve (PSRER) in southern Belarus is one of the most contaminated regions
95 with associated total inventory estimates of 2.5 x 10⁹ MBq of ¹³⁷Cs and 0.18 x 10⁹ MBq of
96 ⁹⁰Sr (Izrael and Bogdevich, 2009). Contamination is spatially highly heterogeneous over
97 distances of 10's of meters or less while significant contamination is also found in the tree
98 canopy and vegetation and in aquatic systems of the area. A general overview of ¹³⁷Cs levels
99 in the PSRER is presented in Figure 1.

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101Insertion point for Figure 1.....

102
103 ***Figure 1. Spatial distribution of the estimated ¹³⁷Cs inventory (kBq m²) across the Polessie***
104 ***State Radioecology Reserve***

105 Cs-137 is usually considered to be the most important long-term anthropogenic contributor to
106 the dose-levels in the environment (Miller, 2007). Its radiological prominence is due to its
107 relatively long half-life of 32 years, its potential to be remobilised through sediment
108 redistribution as it binds strongly to clays and its affinity to be taken in by living tissue given
109 that it readily displaces biological potassium within the body (Povinec et al., 2003). These

110 factors, amongst others, impart a certain emphasis on the monitoring of levels of this isotope
111 within the environment and such monitoring is especially important in highly contaminated
112 areas such as those of the PSRER where considerable efforts have been expended in
113 establishing the spatial distribution and environmental behaviour of this and other nuclides
114 (see for example Kudelsky, et al., 1996; Knatko et al., 1996; Guermentchuk et al., 1997;
115 Sokolik et al., 2001; Yoschenko et al., 2006).

116
117 Monitoring of ^{137}Cs in soils is conventionally achieved by either of two approaches. The first
118 consists of extracting discrete samples or cores from a site and measuring the activity of either
119 the whole sample or of individually sliced samples representative of different depths using
120 laboratory based gamma-ray spectrometers (Tyler, 2008). Although precise information as to
121 the ^{137}Cs depth penetration can be obtained using this method, the resources it takes to obtain,
122 prepare and count individual samples results in spatial coverage being limited and spatial
123 heterogeneities may not always be captured (Tyler et al., 1996a). Estimates of ^{137}Cs
124 penetration down the soil column are particularly important for two reasons. Firstly,
125 knowledge of depth distribution can contribute greatly to remediation strategies and decisions
126 on subsequent site use. Secondly, depth distribution can influence the environmental
127 behaviour of ^{137}Cs particularly with regards to surface remobilisation.

128
129 The environment of the PSRER constitutes an interesting problem with respect to how high
130 resolution mapping of contaminant ^{137}Cs can be conducted. The territories not covered by
131 forest (basically abandoned agricultural lands) constitute 38.0 % and non-agricultural
132 unforested lands occupy 9.3 %. Road coverage is sparse and the terrain is generally rugged.
133 While topographically flat, the area has local depressions and elevations as well as a variety
134 of watercourses and bodies (both natural and man-made) in addition to extensive seasonally
135 flooded areas. In general, much of the area is inaccessible by vehicle and can be considered as
136 difficult terrain. In light of the above considerations, the development of a man portable
137 gamma-ray spectrometry system capable of accurate high spatial resolution ^{137}Cs mapping of
138 contamination patterns is paramount (Plamboeck et al., 2006). Within the PSRER, due to
139 spatial heterogeneity ^{137}Cs activity the capability to map changes over metres to tens of metres
140 would provide significant improvement in spatial resolution compared to current approaches.

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144 *1.1. In Situ Gamma-ray Spectrometry of ¹³⁷Cs and Depth Estimation.*

145 The fundamentals of gamma-ray spectrometry can be found elsewhere, for example Knoll
146 (2010). The principle of *in situ* gamma-ray spectrometry is the placing of a suitable
147 instrument above the ground and making a measurement over a specified time period. The
148 resultant measurement offers information as to the localised radiation field through the
149 differential measurement of gamma-rays that interact with the active volume of the detector.
150 Critically, the energies of individual gamma-rays are recorded and accumulated into a
151 differential energy spectrum. Estimating the ¹³⁷Cs inventory (Bq m⁻²) directly from the count
152 rate under the characteristic ¹³⁷Cs 662 keV photopeak is not always a straightforward task
153 given that a number of factors can significantly change the count rate. Ground curvature
154 (Laedermann et al., 1998) and the influence of trees and other vegetation (Gering et al., 2002)
155 can introduce a considerable amount of uncertainty. However, the primary source of
156 uncertainty often arises when significant changes in depth distribution are encountered, as
157 considerable variations can alter the count rate by orders of magnitude (Sowa et al., 1989).
158 Besides altering the count rate, changes in depth distribution can also change the detector's
159 Field of View (FoV) and can have an influence on the spatial response of single
160 measurements. The FoV of a detector can be defined as the effective volume of soil being
161 sampled and is often characterised by the lateral distance or radius from the detector over
162 which the sampling area extends to (Tyler et al., 1996a). Problems can arise when
163 contamination is highly associated with the soil surface, following fresh deposition, reducing
164 the detectors ability to resolve small scale localised heterogeneity given that the FoV for a
165 detector at 1 m above the ground is greater than 70 m radius. Moreover, at this scale
166 topographic anomalies become important in defining the FoV. Nevertheless, Chernobyl
167 deposition has aged over 30 years and much of the contamination is understood to have
168 penetrated sufficiently into the soil column (Smith and Beresford, 2002). This process should
169 have led to reduction in the FoV for a detector at 1m above the ground to approximately less
170 than 10 m (Tyler et al., 1996a). A further way to reduce the FoV is to operate detectors at a
171 nominal height of 0.1 m, which should reduce the FoV down to between 2 to 5 m.

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173 Importantly, prior to any inventory estimations the count rate must be corrected for the
174 appropriate depth distribution and the influence of the FoV must also be considered prior to
175 interpolation mapping. The peak-to-valley ratio (PTV) provides a means of performing this
176 task through spectral information alone (Zombori et al., 1992). PTV has been widely used to

177 estimate the burial depth distribution of ^{137}Cs and improve the overall accuracy of inventory
178 estimations (Feng et al., 2012; Gering et al., 1998; Hillmann et al., 1996; Tyler, 2004). Tyler
179 et al (1996b) puts forward a particularly visual description of the physical processes that the
180 PTV method is developed upon. In summary, the PTV is the ratio between the number of full
181 energy and forward scattered photons (Figure 2). Forward scattered photons are essentially
182 photons that have lost a small amount of energy whilst exiting the soil and can be located in
183 the “valley” region of the spectrum between the full energy peak at 662 keV and the Compton
184 edge at 480 keV (Figure 2b). Importantly, the greater the depth at which a source is present
185 (Figure 2a), the higher is the probability of photons undergoing forward scattering and
186 occupying the valley region, leading to an increase in valley height and a decrease in the PTV
187 (Figure 2c).

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189 The PTV method can be applied to any detector of adequate energy resolution (Kastlander
190 and Bargholtz, 2005; Tyler, 2004). At present, two of the most commercially widely viable
191 available scintillators that are eminently suitable for portable field operations and that offer
192 high enough energy resolution to measure ^{137}Cs are sodium iodide (NaI:Tl) and lanthanum
193 bromide (LaBr:Ce) (Guss et al., 2010). Of these two, NaI:Tl has been the workhorse of *in*
194 *situ* and mobile gamma-ray spectrometry for a number a decades as it is relatively cheap,
195 robust and provides acceptable energy resolution (~7% at 662 keV). LaBr:Ce on the other
196 hand is a relatively novel scintillator providing superior energy resolution (2.5% at 662 keV)
197 and higher energy efficiency due to its higher density (5.1 g cm^{-3}) compared to NaI:Tl (3.7 g
198 cm^{-3}). LaBr:Ce does contain an intrinsic background component that can deteriorate the
199 signal to noise ratio especially at low count rates (Nilsson et al., 2014). Another key
200 limitation of LaBr:Ce is that it costs almost an order of magnitude significantly more than
201 NaI:Tl. There have been a relatively large number of studies comparing the performance of
202 NaI:Tl and LaBr:Ce through theoretical or experimental means (Guss et al., 2010; Iltis et al.,
203 2006; Milbrath et al., 2007). Nonetheless, limited comparisons have been undertaken in the
204 field (Duval and Arnold, 2013; Nilsson, 2010) and furthermore only a subset of these have
205 pertained to the measurement of ^{137}Cs (Nilsson et al., 2014).

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207Insertion Point for Figure 2.....

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209 **Figure 2. The relationship between changes in depth distribution (β) and the peak-to-valley**
210 **ratio observed in a $71 \times 71 \text{ mm}$ sodium iodide detector. A) Normalised depth distributions**

211 *for the same inventory ; B) Peak normalised spectra for different β values demonstrating*
212 *the relative increase in valley counts with increasing burial; C) A straight-forward*
213 *regression between calibrated peak-to-valley ratio and β value.*

214 The overarching aim of this investigation is to compare NaI:Tl and LaBr:Ce detector systems
215 for the purposes of characterising spatial ^{137}Cs contamination including its vertical depth
216 distribution at high spatial resolution using one second acquisition times within the PSRER.

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220 2. Methodology

221 2.1. Field Sites.

222 The selected sites within the PSRER were topographically flat over areas of 50 m in diameter,
223 the centre of each site being marked with a square area of the size 100 m². Each site was
224 specifically chosen to be as homogeneous as possible. As a general indicator of the
225 conditions of the site, the gamma dose rate was recorded with an AT6130 Radiation Monitor
226 (Scientific and Production Enterprise ATOMTEX, Belarus) at each corner and in the centre at
227 a height of 1 m and soil samples were taken at the same points. The location and ambient dose
228 rates over the five sites were as follows: Site 1, a soddy podzolic soil at N 51°33'07,9" E
229 029°55'26,1" (2.21 to 2.50 µSv/h); Site 2, a sandy soil at N 51°33'17,1" E 029°55'12,1", (0.29
230 to 0.35 µSv/h); Site 3, an organic peat soil at N 51°32'54,7" E 029°55'52,2" (1.55 to 1.73
231 µSv/h); Site 4, a regularly flooded, mineral floodplain soil at N 51°31'45,2" E 029°56'07,5"
232 (0.52 to 0.89 µSv/h); Site 5, a mineral, ploughed soil at N 51°47'11,8" E 030°01'16,8" (0.25
233 to 0.39 µSv/h). A standard cylindrical corer of 4 cm in diameter and 20 cm extent was used to
234 take cores, each divided into the following increments: 0-3 cm, 3-6 cm, 6-10 cm, 10-15 cm,
235 15-20 cm. All layers of the same depth from the five cores were bulked and homogenized for
236 gamma-spectrometry. The preparation of soil samples for the radioanalytical measurements
237 included air drying and homogenization by mixing and sieving (2 mm aperture). After
238 weighing, a subsample was transferred to the analytical geometry for counting. A HPGe
239 (Canberra, Be5030, carbon fibre window, 50% rel. eff.) detector was used for ¹³⁷Cs
240 determination. Count times varied from 6000 to 80000 sec. Statistical uncertainties for the
241 photo peak of ¹³⁷Cs were less than 5 %. Details as to ¹³⁷Cs contamination levels at the five test
242 sites are provided in Table 1. For further site information and detailed photographs of the sites
243 refer to Dowdall et al., (2015).

244

245 2.2. Gamma-ray Spectrometry System

246 A 71 (Ø) × 71 mm NaI:Tl and 71 (Ø) × 71 mm LaBr:Ce detectors (both made by Saint
247 Gobain) were connected to individual Ortec digiBases recording a 1024 channel spectral
248 every second using Ortec's Maestro software (ORTEC, 2005). Accompanying each spectrum,
249 GPS coordinates were recorded using an SX Blue II differential GPS, with a resolution of 0.6
250 m. The software used to combine spectral measurements and spatial coordinates was built

251 within the JAVA framework and was able to provide real-time feedback on count rates in
252 specified windows. To avoid significant spectral drift occurring, Maestro's gain stabilisation
253 software was operated on the 662 keV peak. In order to capture local inhomogeneities, the
254 detectors were to the operators best abilities held at a constant height of 0.1 m from the
255 ground at all times to narrow the field of view. This was expected to deviate by 0.1 m either
256 way. The shape of the spectrum was assumed not to change significantly therefore, only the
257 count rate was altered according to the changing field of view and calculated through a
258 deterministic model; this was built in as an uncertainty following Monte Carlo simulations. A
259 walking speed of approximately 0.5 m s⁻¹ and transect spacing of 0.5 to 1 m was maintained
260 throughout the surveys. Approximately 10 m outside of the test site area were measured in
261 addition to the areas inside the test sites.

262

263 *2.3. Monte Carlo Simulations*

264 The individual instruments were calibrated using Monte Carlo Simulations (MCS). MCS
265 were preferred over analytical calibration given that the relationship between PTV and
266 changes in ¹³⁷Cs depth distribution could be modelled with greater precision compared to a
267 limited number of soil core data (Likar et al., 2000). MCS are widely used to estimate
268 detector responses as radiation transport is probabilistic by nature and very difficult to resolve
269 use deterministic approaches (Stromswold, 1995). Hence, by tracking individual particles,
270 through repeated random sampling of well-defined probability distributions a complex
271 problem can be addressed (Maučec et al., 2009).

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285 Table 1. Inventories, bulk soil density, soil characteristics and depth profiles of ¹³⁷Cs at each
286 of the five sites.

Soil characteristics	Layer	¹³⁷ Cs Bq/kg	% of Total ¹³⁷ Cs Inventory
Soddy podzolic	Site 1		
	0-3 cm	41450+/-8290	66.1+/-13.2
	3-6 cm	15350+/-3070	28.8+/-5.8
	6-10 cm	1360+/-270	3.6+/-0.7
	0-15 cm	314+/-72	1.1+/-0.3
	15-20 cm	92+/-27	0.3+/-0.1
Bulk density / ¹³⁷ Cs Inventory			1.62 g/cm ³ / 2414 kBq/m ²
Sandy	Site 2		
	0-3 cm	876+/-175	24.0+/-4.8
	3-6 cm	814+/-163	19.4+/-3.9
	6-10 cm	693+/-139	22.6+/-4.5
	0-15 cm	813+/-163	33.4+/-6.7
	15-20 cm	16+/-8	0.6+/-0.3
Bulk density / ¹³⁷ Cs Inventory			1.77 g/cm ³ / 215 kBq/m ²
Organic peat	Site 3		
	0-3 cm	33380+/-6675	38.7+/-7.7
	3-6 cm	16000+/-3200	30.1+/-6.0
	6-10 cm	5910+/-1180	17.3+/-3.5
	0-15 cm	2740+/-550	10.0+/-2.0
	15-20 cm	947+/-189	3.9+/-0.8
Bulk density / ¹³⁷ Cs Inventory			1.15 g/cm ³ / 1672 kBq/m ²
Mineral floodplain	Site 4		
	0-3 cm	41810+/-8360	56.6+/-11.3
	3-6 cm	8665+/-1733	22.2+/-4.4
	6-10 cm	3076+/-615	13.4+/-2.7
	0-15 cm	940+/-189	5.9+/-1.2
	15-20 cm	292+/-65	1.8+/-0.4
Bulk density / ¹³⁷ Cs Inventory			1.23 g/cm ³ / 1178 kBq/m ²
Ploughed mineral	Site 5		
	0-3 cm	2114+/-423	17.8+/-3.6
	3-6 cm	2147+/-435	15.3+/-3.1
	6-10 cm	2164+/-433	20.6+/-4.1
	0-15 cm	2235+/-447	24.9+/-5.0
	15-20 cm	1814+/-362	21.5+/-4.3
Bulk density / ¹³⁷ Cs Inventory			1.55 g/cm ³ / 644 kBq/m ²

288 The software package Monte Carlo N-Particle 5 (MCNP5) was used to obtain spectral
289 responses to recover PTV (Briesmeister, 1993). In individual MCS, a basic geometry
290 encompassing the detector, soil column and air filled with constituent materials were
291 described through input cards (Figure 3).

292

293*Insertion Point for Figure 3*.....

294

295 **Figure 3. MCNP setup and method used to derive PTV from an exponential distributed**
296 **source.**

297

298 Descriptions of the starting source were also given to MCNP5, alongside F8 tally instructions
299 that captured the differential energy spectrum within the simulated detector's active volume.
300 Given that the PTV is generated using energies considerably higher than those significantly
301 affected by the photoelectric effect and backscatter (>350 keV), Beck et al. (1972) a standard
302 soil composition was used. All nuclide decay data was acquired from the National Nuclear
303 Data Centre (2013).

304

305 Validated detector geometries utilised in previous studies (Varley et al., 2015a) were used to
306 derive spectral responses for 71 (Ø) × 71 mm NaI:Tl and LaBr:Ce detectors. Varley et al.,
307 (2015a) reported good agreement between MCS and experimental benchmark results
308 collected from doped-concrete calibration pads (1 m²). The code making up the detector
309 geometries was modelled on manufactures specification provided by Saint Gobain along with
310 receipt of detectors. Detector models encompassed the active volume, its outer casing, a
311 simplified photomultiplier tube and the PVC plastic piping used to house the detection unit
312 (Figure 3). The simulated contamination was an unbiased source of 0.662 MeV photons and
313 the geometry was truncated just above the detector where beyond this point all particles were
314 killed to maximise computer time. No specific reduction variance techniques were used such
315 as Russian roulette.

316

317 To optimise the geometry of the simulation, adjoint calculations were utilised to establish the
318 necessary limits or effective FoV of the detector to the source (Varley et al., 2015b). A
319 conservative extent for the surface layer (0-0.01 m) was adopted (25 m) and successively
320 deeper layers were established in the same manner. A Gaussian function was fitted to the

321 adjoint results that included 99.9 % of the total photons. Importantly, MNCP5 does not take
322 into consideration pile-up effects in the electronics of the scintillator for instance when chance
323 coincidence is encountered (Lindstrom and Fleming, 1995). This phenomenon within
324 environmental spectra can significantly alter the shape of the spectrum at high count rates
325 leading to a smearing of the spectrum and importantly removing counts from the full energy
326 peak to a higher energy channel when combined with lower energy photons (Mowlavi and
327 Hadizadeh Yazdi, 2011). Pile-up can be modelled using a Monte Carlo regime if the spectrum
328 is accurate enough across all energies. However, it was accepted that the lower energy part of
329 the spectrum could not be modelled accurately due to uncertainties associated with the
330 photoelectric effect which is driven by the chemical composition of the soil. Therefore no
331 attempt was made to model this occurrence through Monte Carlo methods. Yet, full energy
332 photon losses in the field were not expected to exceed 0.5 % even at the highest count rates
333 observed. This was established by monitoring the smearing of photons to higher energies.

334

335 To achieve spectral responses to a variety of depth distributions and inventories, MCS were
336 simulated at discrete 10 mm depth intervals down to a depth of 70 g cm⁻² (Figure 3). Beyond
337 70 g cm⁻² it was found that contributions to PTV were infinitesimal even for large β values.
338 To form the spectral response for a given value of β and inventory, spectral contributions
339 from each 10 mm slice were weighted in accordance with the appropriate exponential
340 function value (see Eqn [1]). This approach not only permitted the PTV to be defined at a
341 higher depth resolution but also improved counting statistics for later regression models. The
342 reason for this is that the maximum number of particles in MCNP5 (2×10^9) could be run in
343 each simulation effectively increasing source density without repeated random number
344 sampling (Briesmeister, 1993).

345

346 Environmental spectra obtained from the test sites contained contributions from the natural
347 radioelements: ⁴⁰K and the ²³⁸U and ²³²Th series. Although spectral inputs from the natural
348 nuclides were small in comparison to ¹³⁷Cs, contributions to the valley region were found to
349 influence the PTV particularly at low ¹³⁷Cs photon fluxes. Therefore, straightforward uniform
350 distributions for the natural radioelements were included as separate MCS (Thummerer and
351 Jacob, 1998).

352

353

354

355 2.4. Vertical Distribution of ¹³⁷Cs

356 In the first decade or so following Chernobyl, many early studies that used field gamma-ray
357 spectrometry to measure ¹³⁷Cs at undisturbed sites tended to adopt an exponential function,
358 whereby the activity of contamination was assumed to decrease exponentially with increasing
359 depth as the heavily concentrated ¹³⁷Cs source on the surface gradually penetrated into the soil
360 column (Hillmann et al., 1996). A number of studies have confirmed this in Belarus and
361 PSRER (Askbrant et al., 1996; Ivanov et al., 1997) However, more recent observations from
362 other areas of Europe have shown the exponential function to break down as the source
363 further penetrates into the soil column and fresh relatively uncontaminated soil buries
364 contamination to form a subsurface maximum (Almgren and Isaksson, 2006; Bernhardsson et
365 al., 2015). However, the most recent study conducted in the PSRER suggests that the
366 exponential function is perhaps still the most appropriate (Dowdall et al., 2017). There reason
367 are three reasons that might explain this outcome. Firstly, the PSRER in comparison to other
368 parts of Europe, where many of the aforementioned studies were conducted, receives
369 relatively low annual rainfall possibly limiting penetration. Secondly, the primary mode of
370 contamination deposition at the time of the accident was dry deposition rather than wet
371 deposition reducing initial infiltration (Drozdovitch et al., 2013). Finally, within a 30 km
372 radius of the ChNPP, radiocaesium was thought to be highly associated with fuel fragments
373 significantly affecting its physiochemical form and ultimately its mobility (Smith and
374 Beresford, 2002).

375

376 These facts combined with the knowledge that the PSRER has not been disturbed to any great
377 extent in over 30 years, therefore justify an exponential decay function to be used in the
378 modelling of the ¹³⁷Cs depth distribution in this environment (ICRU, 1994). Wherein the
379 activity (A_m) (Bq kg⁻¹) of the soil changes with mass per unit area (ζ) (g cm⁻²) (Eqn [1]).

380

$$381 \quad A_m(\zeta) = A_{m,0} \cdot e^{(-\zeta/\beta)} \quad [1]$$

382

383 Notice, mass per unit area is preferred over physical depth (cm) as it takes into account
384 density (ρ), which generates a better approximation of shielding and thus permits inventory
385 (Bq m⁻²) to be calculated more accurately (Eqn [1]). The relaxation mass per unit area (β)
386 relates the rate at which the surface activity (A_0) decreases with mass depth. Resultantly, β

387 values approaching 0 will define surface contamination whilst uniform contamination will
388 approach ∞ .

389

390 *2.5. Peak and Valley Calculation*

391 The first step in calculation of the PTV was identification of a peak within the correct area of
392 a spectrum obtained in the field. The gain stabilisation software, in the vast majority of cases,
393 centred the 662 keV peak in the correct channel, however small amounts of spectral drift
394 occurred during the survey. To ensure the peak (NaI:Tl = 586.8-737.2 keV; LaBr:Ce = 618.3-
395 705.7 keV), valley (NaI:Tl = 496.5-583.8 keV; LaBr:Ce = 498.7-615.3 keV) and background
396 (NaI:Tl = 740.2-755.3 ; LaBr:Ce = 708.7-726.8 keV) region of interests remained the same
397 between spectra, a Gaussian function was fitted to the peak, the maximum of which could
398 then act as a more precise energy calibration. Thereafter, the background under the peak was
399 estimating using a straightforward linear model fitted to five channels either side of the base
400 of the peak. The background was then subtracted and a second Gaussian function was fitted
401 and integrated to calculate the area. The valley height was calculated by subtracting the area
402 to the high energy side of the peak from the valley region (Tyler, 2004). This routine was
403 written using the base features of the R statistical package (R Core Development Team,
404 2016).

405

406 *2.6. Data Generation*

407 For each detector a calibration dataset was generated, comprised of 1000 spectral responses,
408 derived through randomly sampling input β and inventory parameters and applying them to
409 the MCS dataset. The calibration dataset provided spectra with relatively low uncertainty to
410 facilitate fitting of statistical models. An independent dataset (of 2000 spectral responses) was
411 also formed. This dataset, known as the cross-validation dataset, provided a means of
412 assessing performance through Root Mean Square Error (RMSE) for β and inventory
413 estimates. The same modelling constraints applied to the calibration dataset were used to form
414 the cross-validation dataset, but with the addition of the natural radionuclides (^{40}K , ^{238}U and
415 ^{232}Th series) and Poisson noise generated from the one second counts. A similar
416 bootstrapping subroutine of MCS has proven valuable in previous studies to sample both
417 background and source populations when *a priori* knowledge is not available (Varley et al.,
418 2015b).

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Model parameters (Table 2) were chosen to be representative of the variation that could take place within the PSRER environment. Simulated inventories were markedly higher than those at the five test sites, but it was deemed necessary to develop an approach that could be used for higher deposition densities. Additionally, a maximum β value of 100 was chosen as it was discovered through preliminary investigations that larger values contributed negligible counts into the valley regions and would be almost impossible to differentiate. In any case, β values above 100 make very little difference to final inventory estimations and will have negligible contribution if dose rate calculations are performed (ICRU, 1994).

Table 2. Parameters used to construct a synthetic dataset.

Parameter	Minimum	Mean	Median	Maximum
^{137}Cs inventory (kBq m^{-2})	7.2	2479.3	282	86962.2
β (g cm^{-2})	0.13	18.3	7.6	100.0
^{40}K inventory (kBq m^{-2})	28.2	226.7	222.6	565.2
^{238}U inventory (kBq m^{-2})	7.1	17.5	16.7	42.39
^{232}Th inventory (kBq m^{-2})	7.1	19.3	18.4	42.4

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2.7. Calibration Curves

433 To establish appropriate calibration curves to estimate β and inventory for each detector, a
434 linear model was fitted to the respective calibration datasets through a conventional least-
435 squares fitting regime (Crawley, 2012). The relationship between explanatory and response
436 variables was found to be log-linear whereby both response (β) and explanatory (PTV)
437 variables were logarithmic. Formulation of the depth model was performed by regressing
438 PTV against β , where β_0 is the point at which PTV is equal to zero and β_s is the slope factor
439 applied to PTV (Eqn [1])

440
441
442

$$\log(\beta) = \beta_0 + \beta_s \cdot \log(\text{PTV}) \quad [2]$$

443 The inventory model was estimated by calculating the calibration coefficient C , by dividing
444 the inventory (A) by the number of counts in the full energy peak (N_p) (Eqn [3]). C could then
445 be regressed against the PTV to generate the intercept (C_0) and the slope factor (C_s) (Eqn [4]).

446

$$447 \quad C = A/N_p \quad [3]$$

448

$$449 \quad \log(C) = C_0 + C_s \cdot (PTV) \quad [4]$$

450

451 Applied to field estimates the PTV can be used to derive C , which need only be multiplied by
452 N_p to estimate inventory (Tyler et al., 1996b).

453

454 *2.8. Detector-core Comparisons*

455 To develop a firm basis for statistical comparison with the soil core results, all spectra
456 obtained inside the sites were isolated using a georeferenced polygon, run through the
457 calibrated models and then aggregated to form a median and interquartile ranges. On average
458 approximately 600 spectra were collected at each site. Median and interquartile ranges were
459 preferred over the mean and standard deviation as estimates were found to be highly skewed.
460 Core inventory estimates were the sum of the total activity of the core multiplied by the
461 reciprocal of the total core diameter (m^2). Uncertainties were principally based on counting
462 uncertainty. Derivation of β for core data was performed by using a least-squares fitting
463 regime on the function described in Eqn [1]. Uncertainties were derived by Monte Carlo
464 simulation.

465

466 **3. Results and Discussion**

467 *3.1. Statistical Modelling of Monte Carlo Simulations*

468 Statistically significant relationships (R^2 values greater than 0.93; $p < 0.05$) were found for all
 469 models derived from the calibration datasets (Table 3). Model parameter values are similar
 470 between detectors, suggesting that the detectors respond in a similar manner to an extended
 471 environmental ^{137}Cs source.

472

473 Table 3. Parameter and R^2 values for depth and inventory models for lanthanum bromide and
 474 sodium iodide detectors.

Detector	B_0	Depth model			Inventory model			
		B_s	R^2	$RMSE$	C_0	C_s	R^2	$RMSE$
LaBr:Ce	1579.71	0.97	0.93	19.32	10.23	0.98	0.98	407.21
NaI:Tl	1423.46	0.97	0.94	22.04	11.92	0.98	0.99	483.85

475

476

477 Based on RMSE values on validation datasets in which Poisson noise and natural background
 478 were introduced, LaBr:Ce (407.21 kBq m^{-2} and 19.32 g cm^{-2}) marginally outperformed NaI:Tl
 479 (483.85 kBq m^{-2} and 22.04 g cm^{-2}) (Table 3). LaBr:Ce would at this stage appear to offer
 480 increased performance. Nonetheless, to investigate the response of each detector further the
 481 average uncertainty made on inventory and β estimates, as a function inventory, has been
 482 presented (Figure 4).

483

484Insertion Point for Figure 4.....

485

486 **Figure 4. A) Percentage error associated with inventory estimates as a function of**
 487 **inventory for: lanthanum bromide (purely based on peak counts), lanthanum bromide**
 488 **(peak-to-valley corrected) and sodium iodide (peak-to-valley corrected) B) Percentage error**
 489 **associated with β estimates as a function of inventory for lanthanum bromide (peak-to-**
 490 **valley corrected) and sodium iodide (peak-to-valley corrected).**

491
492 Firstly, the advantage of using PTV corrected data can be realised when compared to
493 uncorrected data derived from peak counts (Figure 4A). At inventories greater than 100 kBq
494 m⁻², there is clear divergence in the percentage error on inventory estimates between corrected
495 and uncorrected LaBr:Ce results, which becomes significantly larger as inventory increases.
496 This occurrence can be explained through counting statistics. For example, at low peak count
497 rates (typically of the order of a few hundred) the uncertainty associated with the calculation
498 of PTV is large enough that it does not aid in the estimation of inventory; the highest source
499 of uncertainty in this instance will be associated with the valley height calculation. However,
500 as count rates rise with increasing contamination, and the valley calculation becomes more
501 reliable, the PTV correction significantly decreases the percentage error associated with
502 inventory estimation.

503
504 Reinforcing the RMSE values (Table 3), corrected NaI:Tl and LaBr:Ce would appear to
505 produce similar percentage error across the inventory range (Figure 4A). Yet, at inventories
506 less than ~1500 kBq m⁻², NaI:Tl exhibits an advantage of ca. 5% lower percentage error than
507 LaBr:Ce. Above ~1500 kBq m⁻², LaBr:Ce would appear to improve on NaI:Tl by
508 approximately 1%. This fact is associated with the intrinsic background contained with the
509 LaBr:Ce active volume. Therefore, at lower count rates, the signal to noise ratio is lower than
510 that of NaI:Tl, nonetheless, when count rates exceed a roughly thousand or more the
511 improved energy resolution deems it superior. A similar aspect is observed when analysing
512 percentage error associated with β estimates as a function of inventory (Figure 4B). Note,
513 there is a similar crossover at approximately ~1000 kBq m⁻², but in this case the advantage
514 above this value is clearer for the LaBr:Ce detector (~10%). Interestingly, percentage error
515 on β values are seen to plateau for both detectors at densities around 4000 kBq m⁻². This
516 possibly represents the absolute limit of detection where very little improvement on depth
517 estimate can be made using the devices, this is due to the sensitivity of the ratio at greater
518 values of β .

519
520 *3.2. Application to the Reference Sites*

521 Core results confirmed that there was a large variation in both the depth distribution and total
522 inventory across the five test sites (Figure 5). General patterns included a clear surface
523 association of contamination at the high activity organic-rich sites (1, 3 and 4) and sites with

524 significant vertical penetration: sites 2 (sandy) and 5 (ploughed). Applying the models
525 developed from MCS to data collected over the test sites yielded encouraging results.
526 Generally, there was good agreement between core and detector inventory estimates
527 particularly for sites 1, 2 and 4 with less than 25 % difference between core and detector
528 estimates (Figure 6A).

529

530*Insertion Point for Figure 5*.....

531

532 ***Figure 5. Inventory depth distributions derived from soil cores extracted from the five***
533 ***GAMFAC sites. Uncertainties are one standard deviation and based on counting statistics.***

534

535 Results are very similar for both detectors for all of the sites confirming earlier percentage
536 error investigations (see Figure 4A). The detectors struggled equally to characterise the
537 inventory at sites 2 and 5 where differences of ca. 53 % and 45 % were encountered (see
538 Figure 6A). It became clear that this method may be limited at sites where the level of
539 contamination is relatively low and at greater depths when operating with one second count
540 times. To lower the uncertainty, it might be appropriate to lengthen the count time or,
541 alternatively, conduct slower traverses and increasing the spatial density of spectra, which
542 could, retrospectively, be summed to improve statistics. It must also be noted that remote
543 measurements tend to only sample the top 20 cm or so of soil, much of the contamination at
544 these sites is below this range and most possibly beyond the detection capabilities of these
545 instruments (see Figure 4A). In addition, it may be argued that these sites could be described
546 as having subsurface maxima ($\sim 20 \text{ g cm}^{-2}$), which the exponential model would have
547 struggled to accurately characterise (see Figure 5). Alternate more flexible models, for
548 example the one described by Hillman et al. (1996) or (Almgren and Isaksson, 2006) might be
549 more appropriate in this case. This could have been more representative of the PTV values
550 found in the field and reduced the uncertainty associated with final inventory estimates.

551

552*Insertion Point for Figure 6*.....

553

554 ***Figure 6. Median and interquartile range values for inventory (A) and β (B) estimates for***
555 ***the five GAMFAC sites for lanthanum bromide and sodium iodide compared to core***
556 ***results.***

557 An example of how complex the spatial distribution of ^{137}Cs in the PSRER can be is
558 provided, whereby inventory and β values obtained using NaI:Tl from site 2 and the area
559 surrounding have been smoothed using universal kriging (Graphical abstract). First of all,
560 notice the reference site chosen on the sand dune (red broken lines) was relatively stable with
561 little variation in inventory and depth confirming it a good choice of reference site to apply
562 static *in situ* systems. However, outside of the isolated site large fluctuations in inventory and
563 depth were witnessed. For instance two relatively large hotspots (5-15 meters in diameter) to
564 the north of the site were estimated to possess more than six times the inventory ($\sim 1500 \text{ kBq}$
565 m^{-2}) of the reference site. Due to the geometry and distance away from the reference site
566 these sources would nonetheless have little impact on final calibration measurements taken at
567 the site. Together with greater inventory, β values associated with these hotspot were
568 estimated to be greater than the surrounding area suggesting contamination was deeper in the
569 soil column. This finding could suggest accumulation and penetration of contamination
570 particularly in areas of the topographic depression in the foreground, which was lower in
571 elevation than the sand dune and could potentially accumulate ^{137}Cs through erosion
572 processes. Further investigative studies involving coring and chemical analysis would have to
573 be performed to confirm this hypothesis.

574

575 ***Graphical abstract. Estimated inventory and β obtained for $71 \times 71 \text{ mm}$ sodium iodide***
576 ***detector for site 2 (red box) and surrounding area. Notice the reference site (red box) is***
577 ***significantly more stable than surrounding areas in terms of depth and activity.***

578 **4. Conclusions**

579 The Chernobyl Nuclear power plant disaster has been the single largest nuclear reactor
580 meltdown in our history. Subsequently, highly radioactive material can be found in large
581 areas of Ukraine, Belarus and Russia presenting a considerable remediation challenge that
582 could take decades or even centuries to overcome. An important step in this process is
583 develop methods to characterise the activity and burial depth of ^{137}Cs in high spatial
584 resolution. To this end, a highly practical method has been described and validated against
585 reference sites within the Polessie State Radioecological Reserve in southern Belarus. It was
586 shown that a conventional 71×71 mm sodium iodide scintillation detector could exhibit
587 equal performance to a significantly more expensive lanthanum bromide detector of the same
588 size and power requirements. This outcome can be attributed to the intrinsic contamination
589 associated with lanthanum bromide. Depth and activity maps of ^{137}Cs contamination have
590 demonstrated the advantage of the technique applied to the sodium iodide detector. Further
591 application of this approach could significantly support future remediation activities
592 ultimately lowering the risk to human health.

593

594 **5. Acknowledgements**

595 This work was performed as part of the GAMFAC intercomparison exercise funded by the
596 Nordic Nuclear Safety Research. The authors would like to acknowledge the staff of the
597 Polessie State Radiation Ecology Reserve of Belarus for the organisation of the GAMFAC
598 project and processing of cores. Additonal gratitude must be served to Ms. Elena Almås for
599 her invaluable translation.

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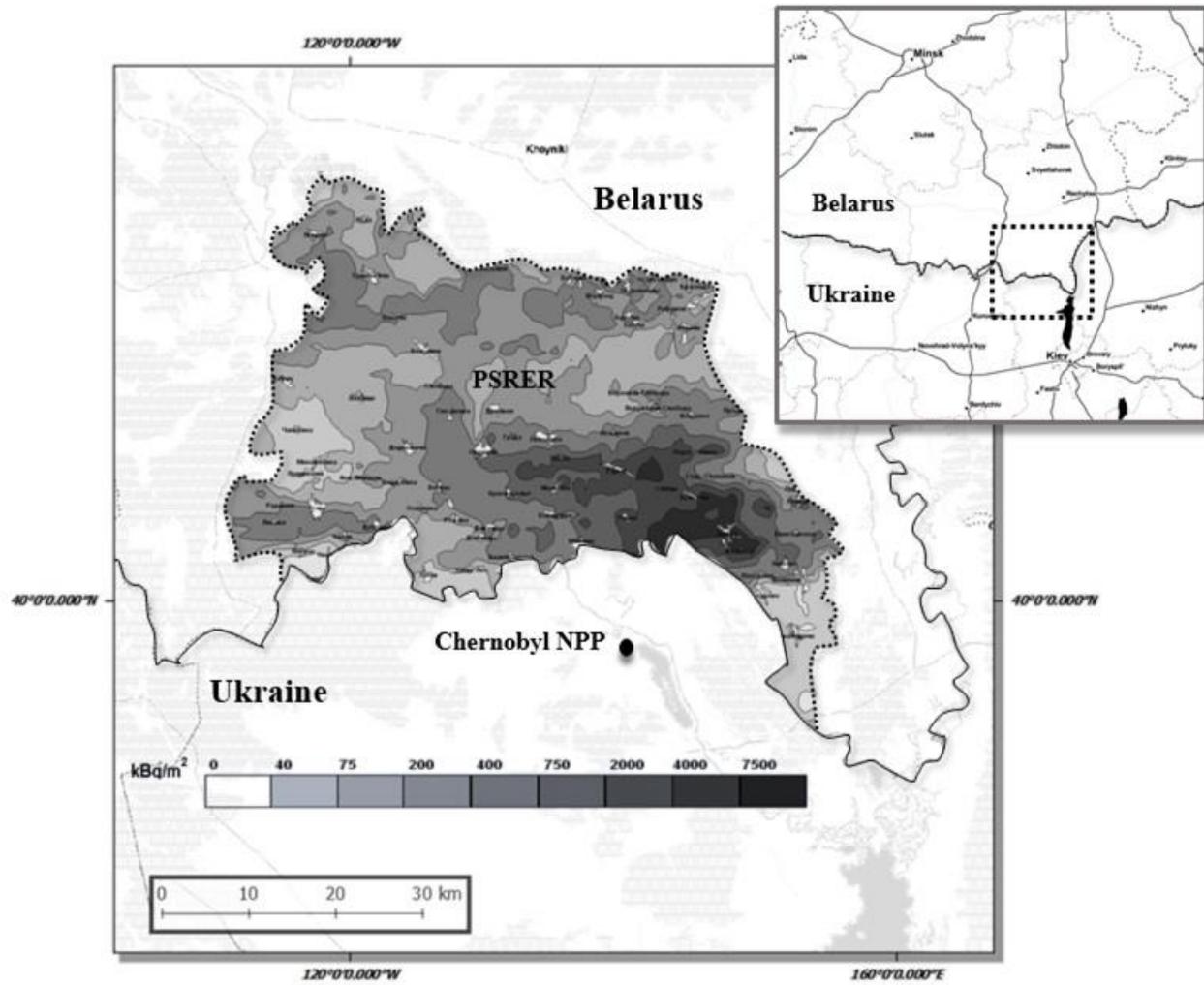
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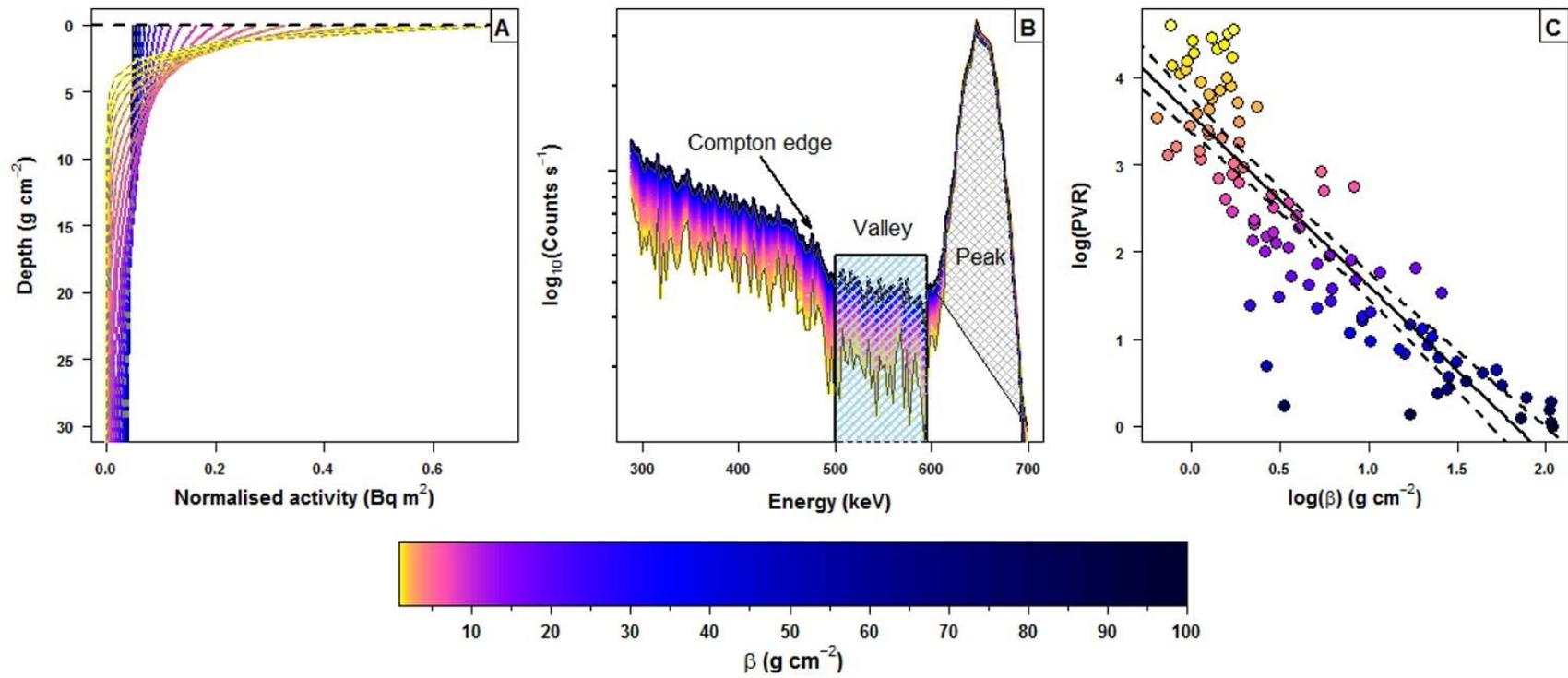
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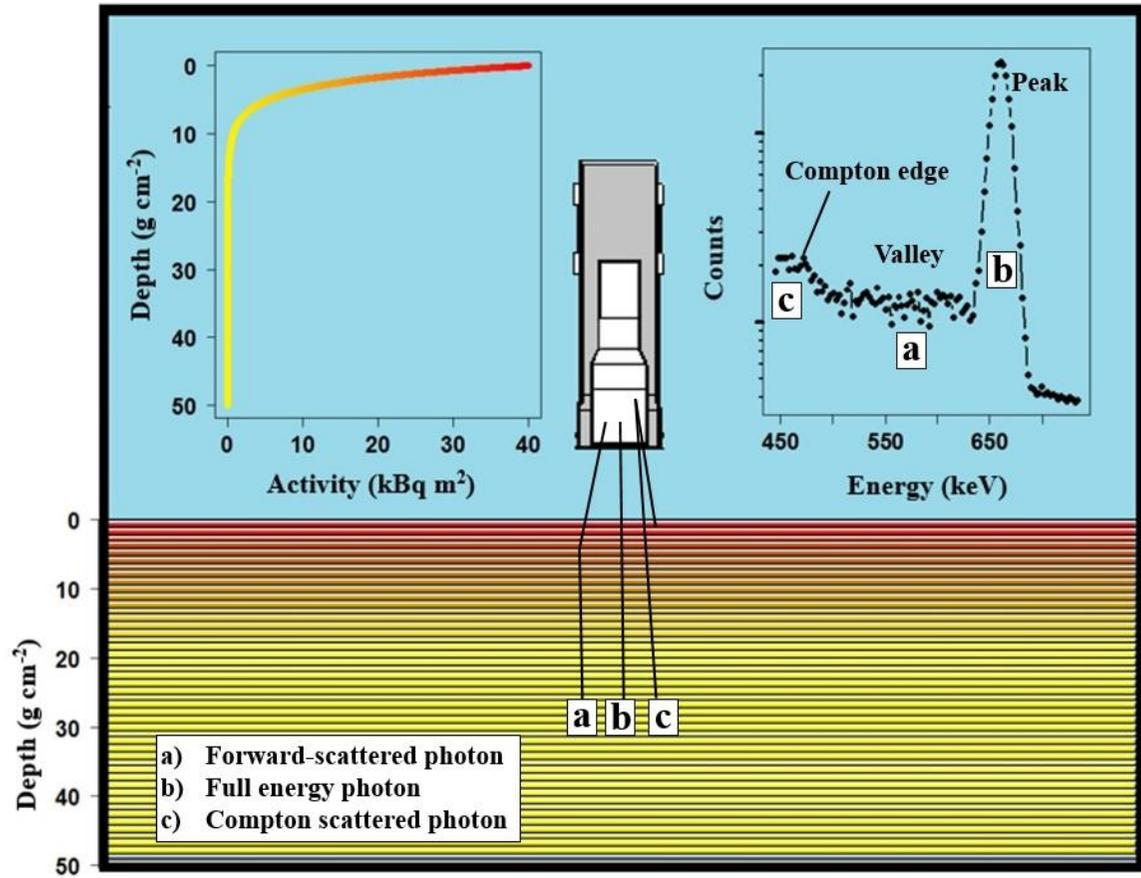
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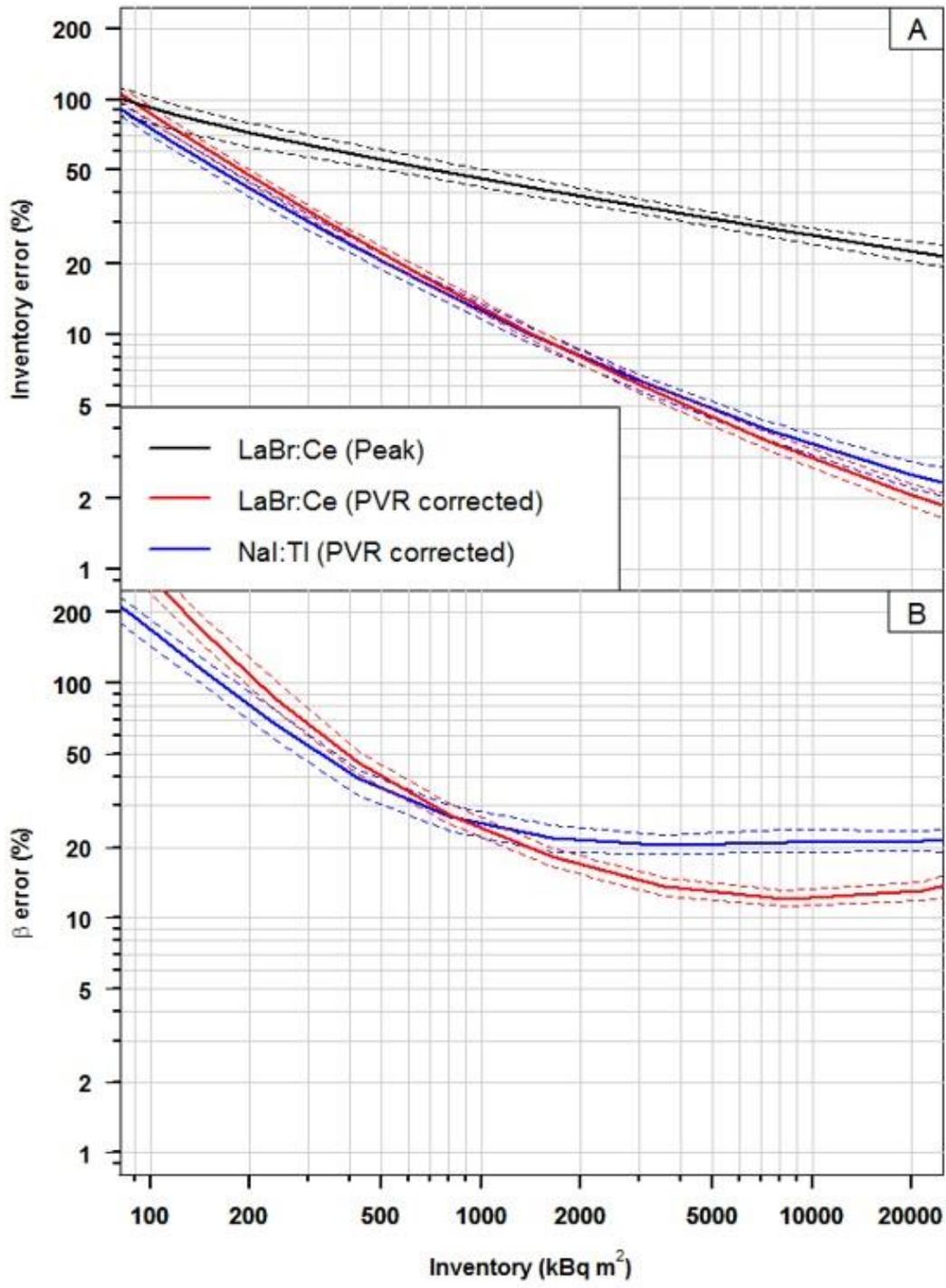
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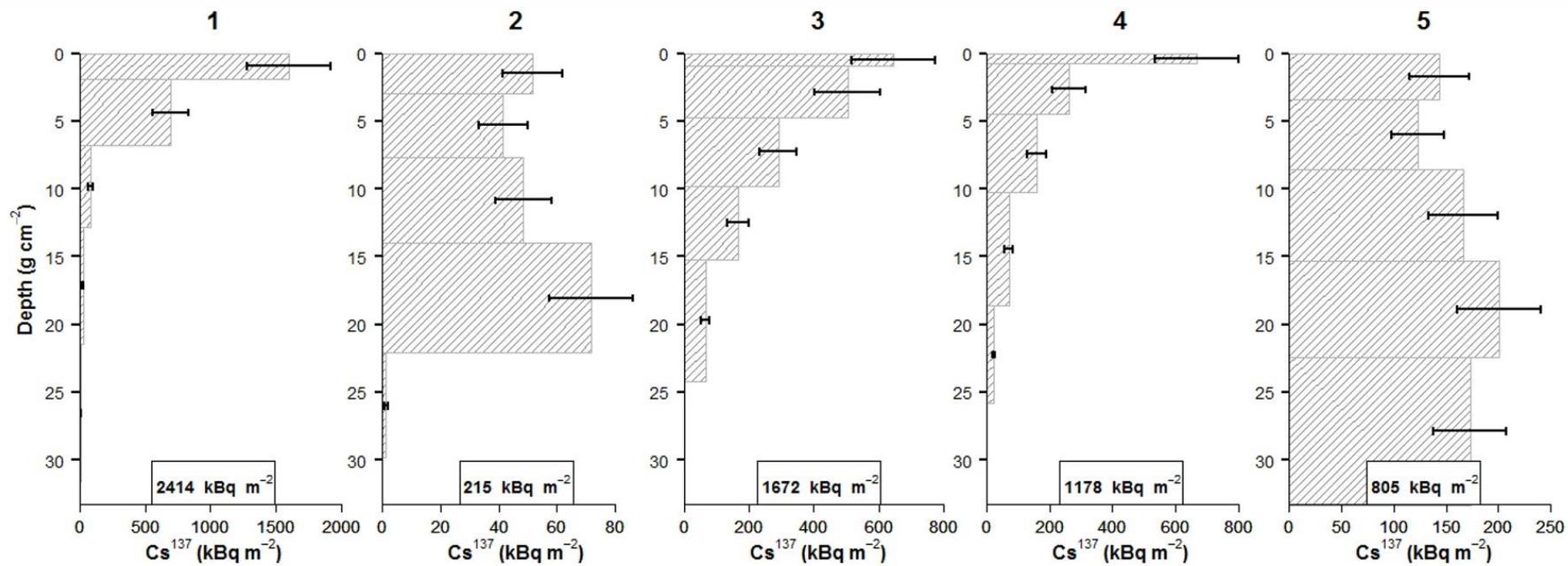
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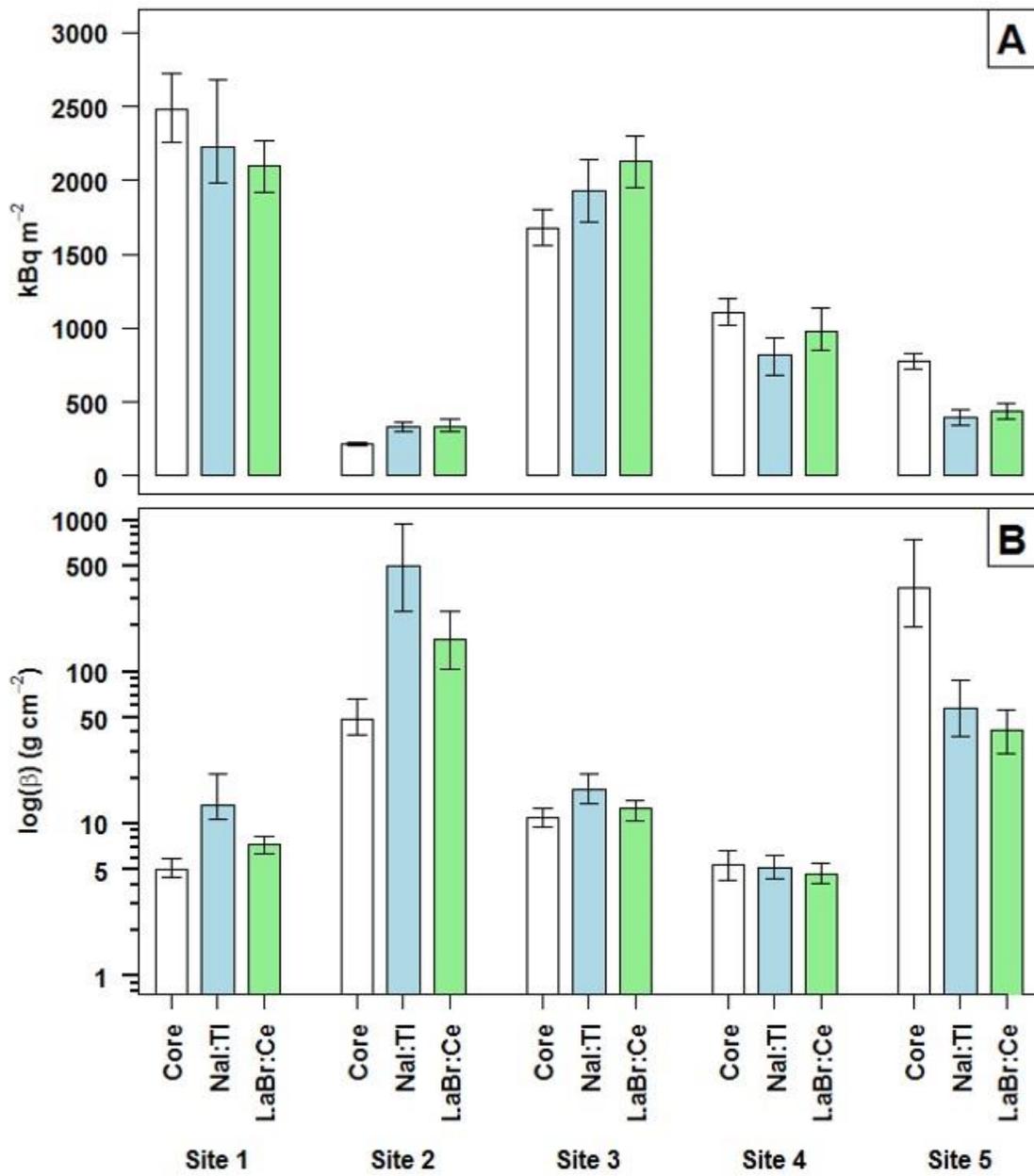
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842 *Figure 4*



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844 *Figure 5*



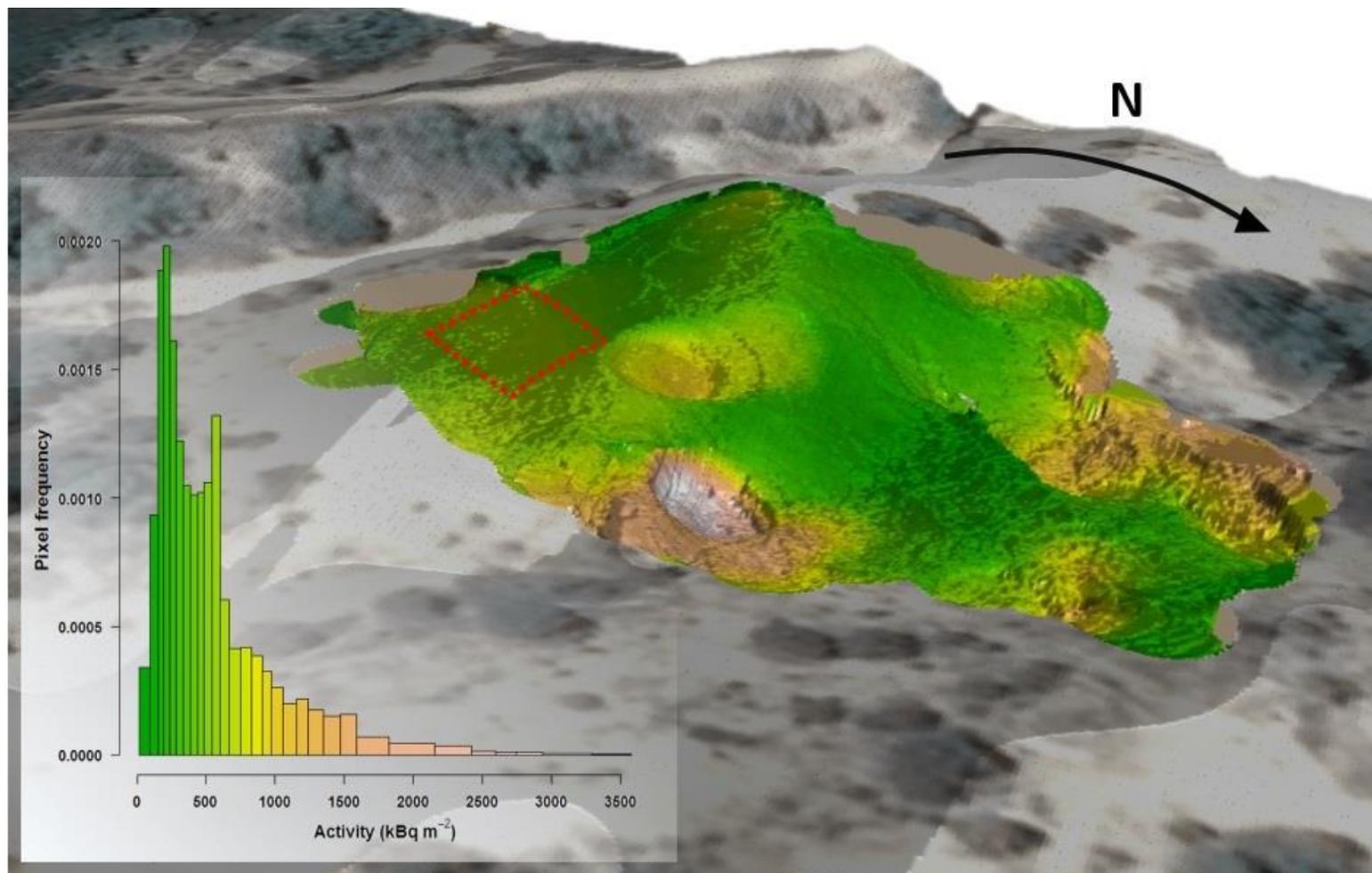
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847 *Figure 6*

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Graphical abstract

