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Evaluation of ambient dose equivalent rates influenced by vertical and horizontal distribution of radioactive cesium in soil in Fukushima Prefecture

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Abstract

The air dose rate in an environment contaminated with ^{134}Cs and ^{137}Cs depends on the amount, depth profile and horizontal distribution of these contaminants within the ground. This paper introduces and verifies a tool that models these variables and calculates ambient dose equivalent rates at 1 m above the ground. Good correlation is found between predicted dose rates and dose rates measured with survey meters in Fukushima Prefecture in areas contaminated with radiocesium from the Fukushima Dai-ichi Nuclear Power Plant accident. This finding is insensitive to the choice for modelling the activity depth distribution in the ground using activity measurements of collected soil layers, or by using exponential and hyperbolic secant fits to the measurement data. Better predictions are obtained by modelling the horizontal distribution of radioactive cesium across an area if multiple soil samples are available, as opposed to assuming a spatially homogeneous contamination distribution. Reductions seen in air dose rates above flat, undisturbed fields in Fukushima Prefecture are consistent with decrement by radioactive decay and downward migration of cesium into soil. Analysis of remediation strategies for farmland soils confirmed that topsoil removal and interchanging a topsoil layer with a subsoil layer result in similar reductions in the air dose rate. These two strategies are more effective than reverse tillage to invert and mix the topsoil.

Keywords: Fukushima Dai-ichi NPP, dose rate evaluation, cesium profile in soil, soil remediation, PHITS

1. Introduction

Dose reconstruction performed after the accident at the Fukushima Dai-ichi Nuclear Power Plant (FD-NPP) showed that in the main regions affected by the accident, like the evacuated areas, external exposure to radiation from radionuclides deposited on the ground (groundshine) was the most important pathway

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5 contributing to effective doses (WHO, 2012; UNSCEAR, 2014). Since the accident the Japanese Govern-
6 ment has restricted the sale of contaminated foodstuffs, and the short-lived tellurium, iodine and xenon
7 radioisotopes released (^{131m}Te , ^{132}Te , ^{131}I , ^{132}I , ^{133}I , ^{133}Xe) have decayed to completion. Therefore, the
8 main radiological hazard that persists in the environment is exposure to groundshine from radioactive cesium
9 (^{134}Cs and ^{137}Cs).

10 Groundshine after a nuclear accident tends to decrease due to radioactive decay of short-lived isotopes
11 (e.g. ^{132}Te , ^{131}I , ^{132}I and ^{134}Cs) and the penetration of fallout radionuclides into soil (ICRU, 1994). The
12 Japan Atomic Energy Agency (JAEA) and partner organizations have been monitoring the environment
13 in North-East Japan since the accident in March 2011 under contract from the Japanese Government. In
14 particular, the consortium has been measuring radiocesium activity depth distributions within soil and
15 monitoring air dose rates at locations of flat, undisturbed fields (Saito and Onda, 2015).

16 Understanding the relationship between distributions of radioactive cesium within the ground and air
17 dose rates is vital for tracking radiocesium migration, predicting future dose rates, and evaluating remedi-
18 ation strategies for reducing dose rates. Previous authors have published conversion factors between the
19 concentration of radionuclides within the ground and various air dose rate quantities (Beck and de Planque,
20 1968; Beck et al., 1972; Beck, 1980; Saito and Jacob, 1995; Quindos et al., 2004; Saito and Petoussi-Henss,
21 2014; Askri, 2015). These conversion factors assume spatially constant radionuclide inventories and depth
22 distributions. To assist the recovery from the Fukushima disaster, Satoh et al. (2014) developed a calculation
23 system to evaluate air dose rates allowing spatially varying radionuclide inventories, using a method based
24 on summing contributions from radionuclides in different volumes of soil.

25 In this paper we present and verify a tool to calculate ambient dose equivalent rates to high precision
26 for arbitrary depth profiles and horizontal distributions of ^{134}Cs and ^{137}Cs fallout within soil. We describe
27 the workings of the tool and demonstrate the validity of its predictions by comparing against monitoring
28 data of air dose rates in Fukushima Prefecture. The tool is applied for understanding reductions in dose
29 rate seen in North-East Japan in terms of migration of radiocesium within soil, and for evaluating different
30 soil remediation options for contaminated farmlands.

31 2. Methods

32 2.1. Tool to evaluate air dose rates

33 The tool calculates ambient dose equivalent rates at 1 m above the ground, $\dot{H}E * (10)$ ($\mu\text{Sv/h}$) (ICRP,
34 1996), hereafter referred to as air dose rates. The tool consists of conversion factors between ^{134}Cs and ^{137}Cs
35 activity concentrations in different cells and layers of soil, and their contribution to the air dose rate. This
36 method allows the dose rate to be calculated for any radiocesium depth profile within soil and horizontal

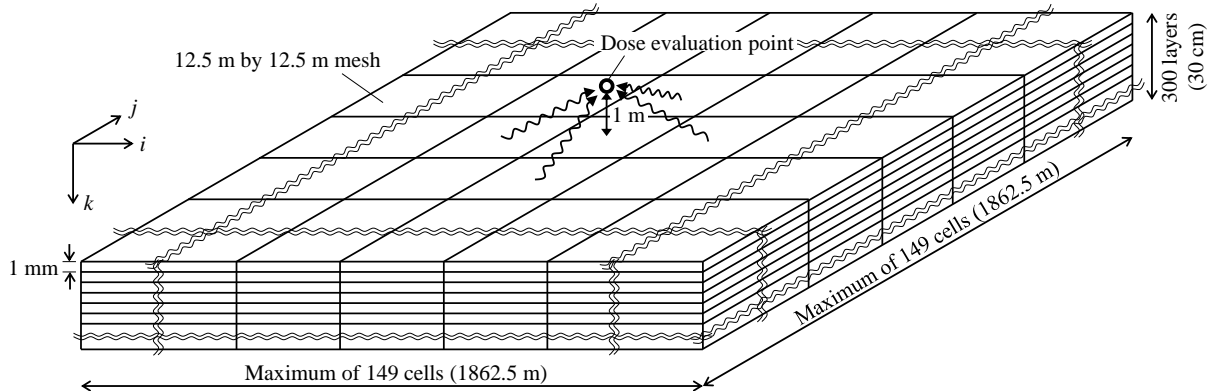


Figure 1: The geometry of the simulations, showing the discretization of soil into small volumes with variable activity concentrations.

37 distribution of the activity, to the limit of the precision of the discretization of the ground into the different
 38 soil volumes.

39 The geometry considered is the infinite half-space (ICRU, 1994) and the land surface is divided into cells
 40 by a 12.5 by 12.5 m mesh (Fig. 1). The tool supports up to 149 by 149 cells horizontally, which equates to
 41 an 1862.5 by 1862.5 m area of land. Up to 300 soil layers, each 1 mm thick, are modelled below each cell on
 42 the mesh. Thus the maximum depth of radiocesium contamination is 30 cm.

43 The half-space geometry is a model for open and uniformly flat land. Therefore any natural or man-made
 44 geographical features that could significantly alter the air dose rate, such as buildings, hilly topography or
 45 dense forests, currently cannot be modelled accurately with the tool. The model also does not consider the
 46 effects of ground roughness on air dose rates. These effects are most significant when modelling planes of
 47 radionuclides on the surface of the ground in the half-space geometry, such as in the period immediately
 48 after fallout deposition. However, they are negligible when modelling radionuclides dispersed within the
 49 ground after this initial weathering period has completed (Jacob et al., 1994), as is the case in this paper.

50 The input data for an air dose rate calculation consist of ^{134}Cs and ^{137}Cs activity concentrations within
 51 each discrete volume of soil. The calculation for the air dose rate performed by the tool is given by

$$\dot{H}E * (10) = \sum_{n,i,j,k} A_{v,n,i,j,k} c_{n,i,j,k}, \quad (1)$$

52 where $A_{v,n,i,j,k}$ (Bq/m^3) is the activity concentration of radiocesium in the soil volume, and $c_{n,i,j,k}$ ($\mu\text{Sv}/\text{h}$
 53 per Bq/m^3) is the activity to dose conversion factor for that soil volume. The index n denotes the ^{134}Cs or
 54 ^{137}Cs isotope, indices i, j denote the cell position of the soil volume on the mesh, and k indexes the depth
 55 of the volume for layers numbered down from the surface. As the calculation in Eq. 1 is a simple sum over
 56 all the soil volumes in the problem, the run-time of the tool on a standard desktop computer is about 10 s.

57 The conversion factors for all the soil volumes were calculated using the Particle and Heavy Ion Transport

58 code System (PHITS) (version 2.64 - Sato et al. (2013)). PHITS is a Monte Carlo radiation transport code.
 59 The conversion factors represent the dose rate at 1 m above the middle of the central cell on the mesh per
 60 unit activity concentration within that volume of soil. The density of soil was $\rho_s = 1.6 \text{ g/cm}^3$ and air
 61 was $\rho_a = 0.0012 \text{ g/cm}^3$. The soil and air chemical compositions followed Eckerman and Ryman (1993).
 62 The ^{134}Cs and ^{137}Cs emission spectra were drawn from NuDat2 (2014). Note that the ^{137}Cs energy lines
 63 in NuDat2 (2014) include the contribution from the short-lived daughter product $^{137\text{m}}\text{Ba}$. In each case
 64 the source region was scaled to a vertical line and the detectors transformed to planes to maximize the
 65 computational efficiency of the Monte Carlo simulations (Namito et al., 2012).

66 2.2. Transforming measured activity depth profiles for input into the tool

67 The tool cannot accurately simulate scenarios where there is significant variation in the soil density
 68 horizontally across the simulation region, as a constant soil density ($\rho_s = 1.6 \text{ g/cm}^3$) was employed in the
 69 PHITS simulations. However, other constant soil densities ($\rho_s \neq 1.6 \text{ g/cm}^3$) or soils with varying density as
 70 a function of depth (i.e. $\rho_s(z)$, where z (cm) is the depth in soil from the ground surface) can be simulated.
 71 The solution is to transform the depth coordinate of the source activity depth profiles using the mass
 72 depth (ICRU, 1994). The mass depth, ζ (g/cm^2), is defined as

$$\zeta(z) = \int_0^z \rho_s(z') dz'. \quad (2)$$

73 A contamination depth profile per unit soil mass measured in a field survey, $A_m(\tilde{z})$ (Bq/kg), can be recast
 74 using Eq. 2 into a function of soil mass depth

$$A_m(\zeta) = A_m \left(\int_0^{\tilde{z}} \rho_s(\tilde{z}') d\tilde{z}' \right), \quad (3)$$

75 where $\rho_s(\tilde{z})$ is the density-depth profile of the soil measured from field samples. Here tildes are used to
 76 distinguish the depth coordinate in the field, \tilde{z} (cm), from the depth coordinate z applicable for inputting
 77 data into the tool.

78 The activity profile as a function of mass depth, $A_m(\zeta)$, can be transformed into a function of z applicable
 79 for the tool's constant soil density conversion factors, by reapplying Eq. 2:

$$A_m(z) = A_m(\zeta/1.6). \quad (4)$$

80 Finally, an activity concentration depth profile ($A_v(z)$) for inputting into the tool can be obtained by
 81 multiplying $A_m(z)$ by the constant soil density

$$A_v(z) = 1000 \cdot 1.6 \cdot A_m(z). \quad (5)$$

82 The factor of 1000 ensures $A_v(z)$ has units of Bq/m³.

Table 1: Details of the soil sampling campaigns and sites used for air dose rate predictions.

Soil sampling campaign	Dates	Number of sites used	Number of sites with hyperbolic secant fits
1 st	Dec 12–22, 2011 and Apr 17–19, 2012	41	8
2 nd	Aug 21 to Sep 26, 2012	82	13
3 rd	Nov 26 to Dec 26, 2012	81	28
4 th	Jun 4–27, 2013	80	12
5 th	Oct 28 to Nov 29, 2013	79	23

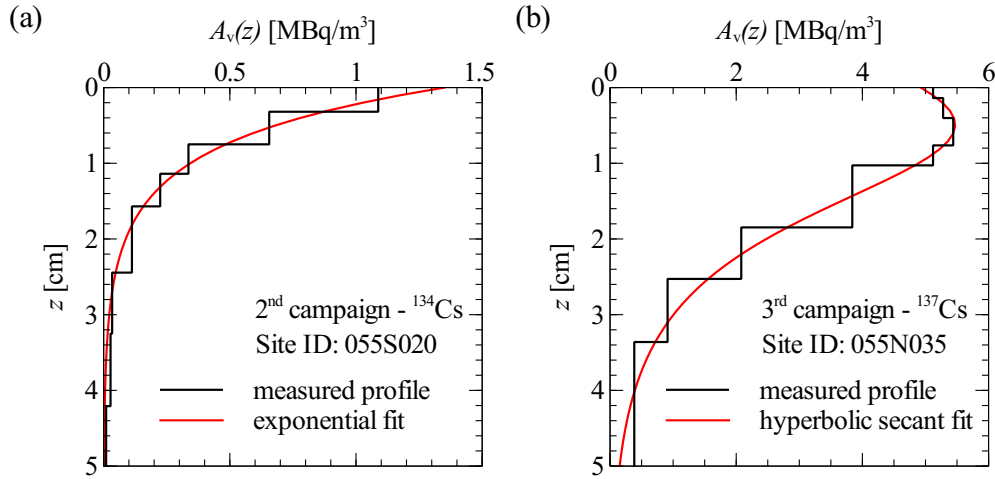


Figure 2: Examples of depth distributions of ^{134}Cs and ^{137}Cs inputted to the tool for dose rate calculations. (a) Exponential depth profile, and (b) hyperbolic secant depth profile. Site identification codes (IDs) follow Matsuda et al. (2015).

83 2.3. Scenarios considered

84 2.3.1. Dose rates above flat, undisturbed fields

85 Field survey teams have measured depth profiles of radioactive cesium in soil at approximately 80 loca-
86 tions near to FDNPP since December 2011 (Matsuda et al., 2015). The samples were taken from sites at
87 wide, flat areas of land and at least 5 m from buildings and trees. Soil samples were collected using a scraper
88 plate. This apparatus was used to remove individual soil layers with thickness between 0.5–3 cm and with
89 increasing depth from the ground surface for radiochemical analysis. Properties analyzed included the in
90 situ density ($\rho_s(\tilde{z})$) and the ^{134}Cs and ^{137}Cs activity per unit mass ($A_m(\tilde{z})$) of each soil layer. All survey
91 data are published online (JAEA, 2015a).

92 We used our calculation tool to predict the air dose rate at each sampling site based on the soil activity
93 measurements. We then compared the results with 1 m air dose rates measured in the field using hand-held

94 survey meters. Data over five soil sampling campaigns were considered in the analysis. The dates of the
 95 campaigns are listed in Table 1. The measured activity depth profiles ($A_m(\tilde{z})$) were scaled into activity
 96 concentration profiles applicable for the tool ($A_v(z)$) using the procedure described in section 2.2. Examples
 97 of the processed depth profiles for two sites are shown in Fig. 2 (black lines).

98 As a scraper plate sample was taken at only one point on the ground per location visited in each soil
 99 sampling campaign, it was assumed that the measured soil activity depth distribution applied homogeneously
 100 across the whole region simulated by the tool. Explicitly, the activity concentrations for each soil layer
 101 inputted to the tool were identical across all cells on the simulation mesh.

102 We also considered a second method for evaluating air dose rates from the soil activity samples, based
 103 on modelling empirical fits to the activity depth profiles. Matsuda et al. (2015) characterized the activity
 104 depth profiles as a function of mass depth ($A_m(\zeta)$) by fitting exponential and hyperbolic secant functions.
 105 The exponential depth distribution is

$$A_m(\zeta) = A_{m,0} \exp(-\zeta/\beta), \quad (6)$$

106 where $A_{m,0}$ (Bq/kg) is the activity per unit soil mass at the ground surface and β (g/cm²) is the relaxation
 107 mass depth that characterizes the degree of fallout penetration into the soil. Figure 2(a) shows a fit of the
 108 exponential function to soil layer activity measurements at one site. The total inventory of contamination
 109 per unit area of land for this distribution is:

$$A_{\text{inv}} = 10\beta A_{m,0}. \quad (7)$$

110 The factor of 10 ensures A_{inv} has units of Bq/m². The exponential distribution is a satisfactory model for
 111 the soil activity depth profile for the first few years after fallout deposition (ICRU, 1994).

112 Matsuda et al. (2015) observed that some of the measured depth profiles display a maximum in the
 113 radiocesium concentration below the ground surface. They proposed fitting a hyperbolic secant function to
 114 these depth profiles, as this function can reproduce a peak in activity concentration below the surface. The
 115 hyperbolic secant function is

$$A_m(\zeta) = A_{m,0} \cosh(\zeta_0/\beta) \operatorname{sech}(-(\zeta - \zeta_0)/\beta). \quad (8)$$

116 Again $A_{m,0}$ (Bq/kg) is the activity per unit soil mass at the ground surface, and β (g/cm²) is a parameter
 117 characterizing the length scale of the distribution. The peak in activity occurs at the mass depth ζ_0 (g/cm²)
 118 below the surface. The hyperbolic secant function converges to an exponential distribution at large mass
 119 depths. Figure 2(b) shows a fit of the hyperbolic secant function to a measured depth distribution. The
 120 total radionuclide inventory per unit area for the hyperbolic secant distribution is

$$A_{\text{inv}} = 20\beta A_{m,0} \cosh(\zeta_0/\beta) [(\pi/4) + \tan^{-1}(\tanh(\zeta_0/(2\beta)))]. \quad (9)$$

121 We followed Matsuda et al. (2015) and fitted the exponential and hyperbolic secant distributions to measured
122 depth profiles. The hyperbolic secant function was used for profiles displaying a peak in activity below the
123 surface, and the exponential function otherwise. Table 1 lists the number of sites in this analysis and the
124 number of fits with the hyperbolic secant function. Examples of the fitted distributions are shown for two
125 sites in Fig. 2 (red lines)

126 We discounted from the analysis any depth profiles showing signs of soil mixing or disturbance (Matsuda
127 et al., 2015). Soil disturbance included land cultivation and decontamination work. We also discounted sites
128 where the air dose rate was not measured with a survey meter at the time of collecting the soil samples.

129 Modelling both the measured and the empirical fits for the soil depth profiles yielded two predictions
130 for the air dose rate at each site. The air dose rate was calculated as the sum of contributions from ^{134}Cs
131 and ^{137}Cs , and an additional $0.05\ \mu\text{Sv/h}$ contribution representing the background dose rate from natural
132 radionuclides (Mikami et al., 2015b).

133 2.3.2. Evolution of air dose rates

134 In addition to the soil sampling campaigns, JAEA and partners have been measuring air dose rates with
135 hand-held survey meters at thousands of locations across Fukushima Prefecture, including locations with
136 flat, undisturbed fields (Saito et al., 2015; Mikami et al., 2015b). The monitoring results show that dose rates
137 at these locations decreased faster than expected by just the physical decay of ^{134}Cs and ^{137}Cs (Saito and
138 Onda, 2015). Mikami et al. (2015a) demonstrated that, for the period between March 2012 and December
139 2012, relatively little migration of the ^{137}Cs inventory away from these fields occurred. Mikami et al.
140 (2015b) explained the decrease in dose rates, beyond what could be expected by radioactive decay alone, by
141 the downward migration of radioactive cesium into the soil.

142 The relaxation mass depth, β , characterizes the penetration of fallout into soil for the exponential
143 distribution (Eq. 6). Conversion coefficients published for various values of β can be used to evaluate the
144 1 m ambient dose equivalent rate given the radionuclide inventory per unit area of soil (Saito and Petoussi-
145 Henss, 2014). In contrast, two parameters characterize the penetration of the radionuclides within soil for
146 the hyperbolic secant distribution - a relaxation mass depth β and a mass depth ζ_0 for the peak in activity
147 concentration below the surface (Eq. 8).

148 To allow direct comparison between exponential and hyperbolic secant depth profiles, Matsuda et al.
149 (2015) proposed an effective relaxation mass depth parameter, $\beta_{\text{eff}}^{\dot{K}}$ (g/cm^2), for the hyperbolic secant dis-
150 tribution. The effective relaxation mass depth is defined as the value β of an exponential depth distribution
151 yielding the same air kerma rate at 1 m as the hyperbolic secant distribution (\dot{K} - $\mu\text{Gy/h}$), given an identical
152 inventory of fallout radionuclides in both distributions (i.e. A_{inv} is equal for both distributions).

153 In this study we used our calculation tool to calculate effective relaxation mass depths for the hyperbolic
154 secant fits over the five soil sampling campaigns. The effective relaxation mass depths were calculated

Table 2: Data for $\beta_{\text{eff}}^{\dot{H}E*(10)}$ over the five soil sampling campaigns.

Soil sampling campaign	Number of sites used	Number of sites with hyperbolic secant fits	Mean $\beta_{\text{eff}}^{\dot{H}E*(10)}$ (g/cm ²)	Median $\beta_{\text{eff}}^{\dot{H}E*(10)}$ (g/cm ²)	Min $\beta_{\text{eff}}^{\dot{H}E*(10)}$ (g/cm ²)	Max $\beta_{\text{eff}}^{\dot{H}E*(10)}$ (g/cm ²)
1 st	83	12	1.13	0.93	0.24	5.95
2 nd	82	13	1.41	1.00	0.11	8.72
3 rd	81	28	1.56	1.23	0.43	10.36
4 th	80	12	1.64	1.36	0.29	7.73
5 th	79	23	2.17	1.85	0.38	6.41

155 by matching $\dot{H}E*(10)$ from each hyperbolic secant distribution to an exponential distribution of equal
156 inventory, i.e. $\beta_{\text{eff}}^{\dot{H}E*(10)}$ (g/cm²). Note that the definition of effective relaxation mass depth means that
157 $\beta_{\text{eff}}^{\dot{H}E*(10)}$ for an exponential distribution is equal to the relaxation mass depth (β) of the distribution.

158 We calculated arithmetic mean, median, minimum and maximum $\beta_{\text{eff}}^{\dot{H}E*(10)}$ values for each of the five
159 soil sampling campaigns (Table 2). More sites from the first soil sampling campaign could be used in this
160 analysis than for the dose rate predictions (c.f. Table 2 with Table 1), as it was not necessary to have a field
161 survey measurement of the air dose rate in order to calculate $\beta_{\text{eff}}^{\dot{H}E*(10)}$.

162 We considered the decrement of the components of the air dose rate attributable to radioactive cesium
163 fallout, i.e. $\dot{H}E*(10)$ measurements minus a 0.05 $\mu\text{Sv/h}$ contribution from natural background radiation,
164 over the first four air dose rate surveys (JAEA, 2015a). The dates of the air dose rate surveys and the mean
165 air dose rates at flat, undisturbed fields are listed in Table 3.

166 We modelled the decrement in dose rates due to radioactive decay and cesium migration deeper within
167 soil. First, we matched the mean $\beta_{\text{eff}}^{\dot{H}E*(10)}$ values from the soil sampling campaigns to the periods of the
168 air dose rate surveys (Table 3). We then modelled exponential distributions with β parameters equal to the
169 mean $\beta_{\text{eff}}^{\dot{H}E*(10)}$ values with the tool. The inventories supplied were decay corrected to dates at the middle
170 of each air dose rate survey period. The decay corrections assumed an activity ratio of released ¹³⁴Cs and
171 ¹³⁷Cs from FDNPP of 1.00 on March 11, 2011 (UNSCEAR, 2014). The results show little sensitivity to
172 plausible alternatives (in the range 0.90–1.08) for this initial activity ratio. The calculated dose rates were
173 then normalized to June 21, 2011, the date at the middle of the first air dose rate survey (Table 3), for
174 comparison with the measured dose rates. As no scraper plate soil samples were available for the period
175 of the first air dose rate survey (June 4 to July 8, 2011), a $\beta_{\text{eff}}^{\dot{H}E*(10)}$ value of 1.00 g/cm² was assumed as
176 applicable for this period (Mikami et al., 2015b). The sensitivity of the results upon this assumption is
177 checked in the results section.

Table 3: Details of the air dose rate surveys and results for the models of dose rate reductions. ^aAssumed value. See text for details.

Air dose rate survey		Soil sampling campaign	Mean $\beta_{\text{eff}}^{\dot{H}E^*(10)}$ (g/cm ²)	Measurements		Models	
Campaign	Dates			$\dot{H}^*(10) - 0.05$ (μSv/h)	Relative change	Decay only	Decay & migration
1 st campaign	Jun 4 to Jul 8, 2011	-	1.00 ^a	1.25	1.0	1.0	1.0
2 nd campaign	Dec 13, 2011 to May 29, 2012	1 st	1.13	1.01	0.81	0.84	0.82
1 st part of 3 rd campaign	Aug 14 to Sep 7, 2012	2 nd	1.41	0.84	0.67	0.76	0.70
2 nd part of 3 rd campaign	Nov 5 to Dec 7, 2012	3 rd	1.56	0.78	0.62	0.72	0.65
1 st part of 4 th campaign	Jun 3 to Jul 4, 2013	4 th	1.64	0.64	0.51	0.64	0.57
2 nd part of 4 th campaign	Oct 28 to Dec 2, 2013	5 th	2.27	0.55	0.44	0.59	0.48

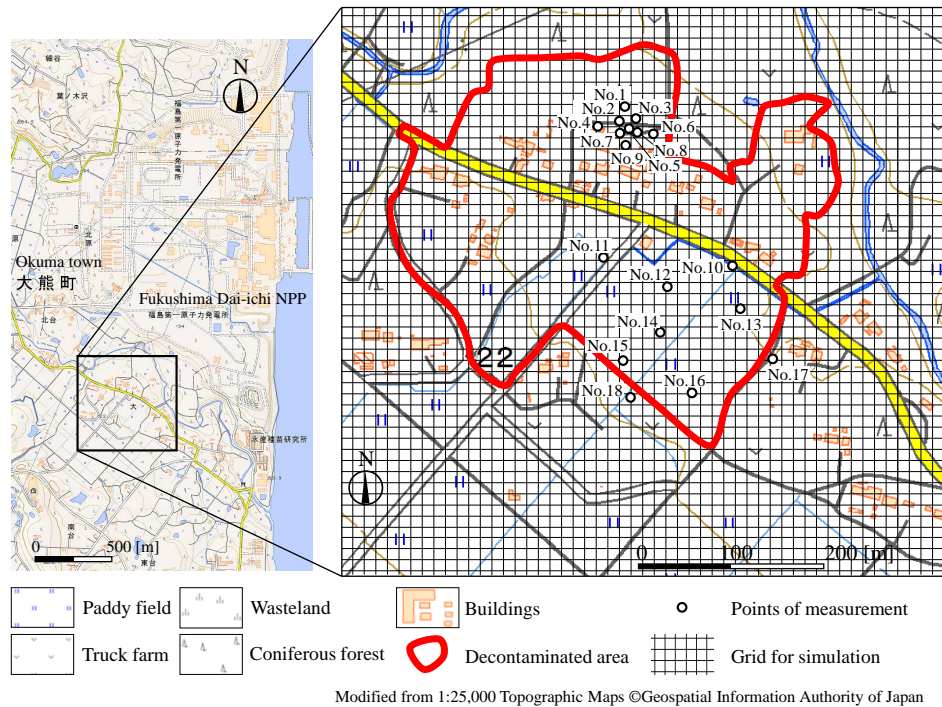


Figure 3: The decontamination boundary and soil sampling locations at Ottozawa.

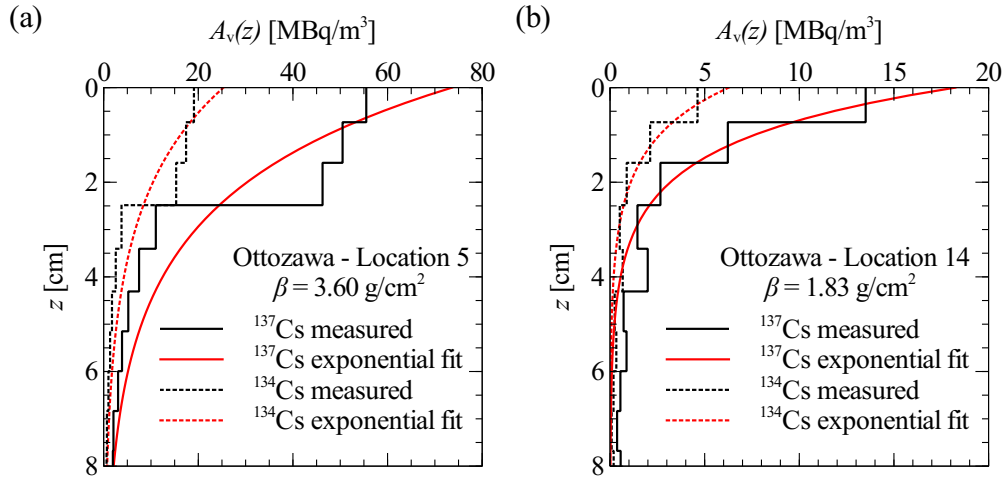


Figure 4: Measured soil activity depth distributions and exponential fits for the Ottozawa area. (a) Location 5, and (b) location 14.

178 2.3.3. Spatial variability in soil activity levels

179 The calculations with the tool in sections 2.3.1 and 2.3.2 assumed spatially uniform radiocesium distri-
 180 butions, as only one soil sample was available at each location. We considered the effect of spatial variability
 181 in the radiocesium distribution on evaluating dose rates by studying the Ottozawa area. The area lies within
 182 2 km of FDNPP and soil samples were taken at multiple locations across the paddy fields and scrubland in
 183 the area. Figure 3 shows a map of the area and the soil sampling locations.

184 This area was remediated between November 2011 and May 2012 as part of a decontamination pilot
 185 project coordinated by JAEA and is now subject to long-term environmental monitoring (JAEA, 2015b).
 186 Remediation consisted of removing the top 5 cm of topsoil from paddy fields and areas around residential
 187 buildings, and cleaning road and building surfaces. The air dose rates ranged from 22–263 μ Sv/h before
 188 decontamination, and dropped to between 4–110 μ Sv/h afterwards. Decontamination of this area was studied
 189 numerically by Hashimoto et al. (2014).

190 The soil samples and air dose rates were taken on July 24, 2014 at 18 locations across the area. Soil
 191 samples were collected by inserting a cylindrical plastic cup (U-8 type, 58 mm internal height, 50 mm internal
 192 diameter) into the topsoil and collecting the soil contents into plastic bags (Onda et al., 2015). One sample
 193 was taken at locations 1–16, while five samples were collected for locations 17 and 18. Locations 17 and 18
 194 lie outside the decontaminated area.

195 The ¹³⁴Cs and ¹³⁷Cs depth distributions at locations 5 and 14 were determined by using a scraper plate
 196 to remove 1 cm thick soil layers down to a maximum depth of 10 cm. The activity per unit soil mass in each
 197 layer was measured using a high resolution gamma spectrometer. Unfortunately due to oversight we did
 198 not measure the in situ densities at the time of collecting the soil samples. Therefore we had to make an

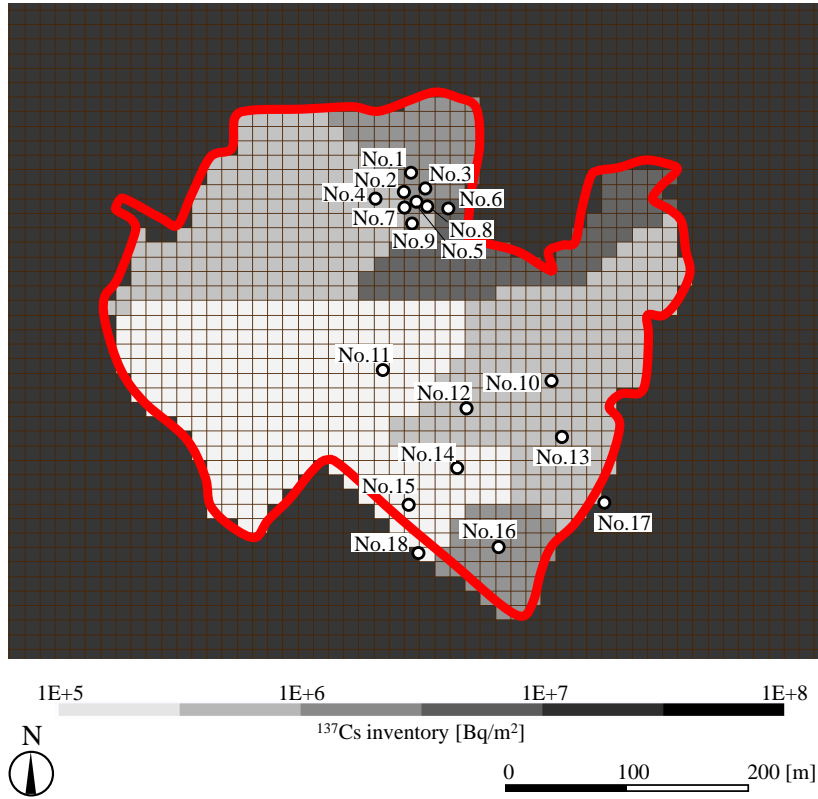


Figure 5: ^{137}Cs inventories assigned to cells across the Ottozawa area for simulation of air dose rates at locations 1–18.

199 assumption for the layer densities. We chose densities equal to the mean densities of the soil layers collected
 200 over the five scraper plate sampling campaigns described in section 2.3.1. The measured activity depth
 201 profiles at Ottozawa were exponential to a reasonable approximation (Fig. 4).

202 Scraper plate analyses of the activity depth distributions were not performed at the other locations
 203 (locations 1–18, excluding 5 and 14). An exponential depth distribution was assigned to these locations
 204 based on the β value applicable at the nearest of locations 5 and 14 to the site. Locations 1–9 were thus
 205 assigned an exponential depth distribution with $\beta = 3.60 \text{ g/cm}^2$, and locations 10–18 an exponential depth
 206 distribution with $\beta = 1.83 \text{ g/cm}^2$. The total inventory per unit area, A_{inv} , was inferred by correcting
 207 the cylindrical cup activity measurement for radioactivity at depths greater than 58mm as given by the
 208 exponential distribution. The inventory for locations 5, 14, 17 and 18, where multiple soil samples were
 209 taken, was taken to be the mean over the various samples.

210 Two strategies were used to predict the air dose rate at locations 1–18. The first strategy assumed that
 211 the radiocesium distribution was spatially homogeneous. The inventory and depth distribution for that
 212 location was applied uniformly across the simulation region.

213 The second strategy was to model the spatial heterogeneity in soil activity levels, as revealed by the

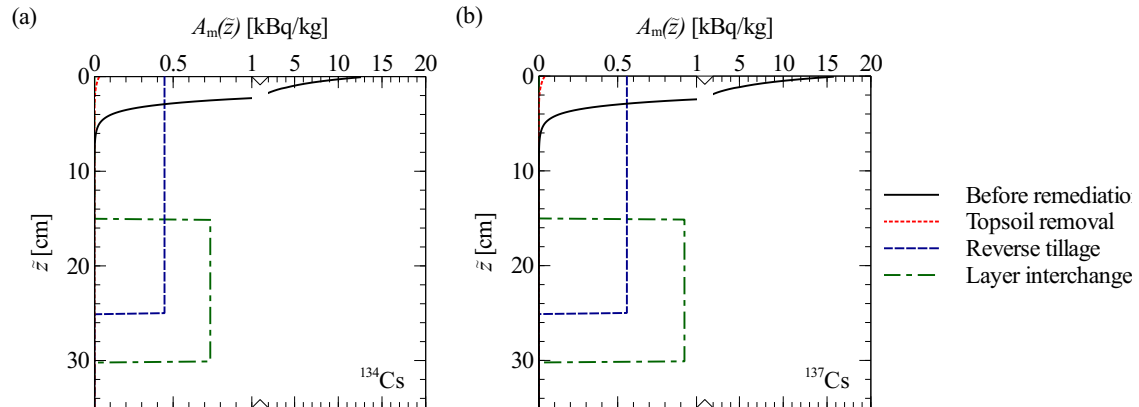


Figure 6: Activity depth distributions of ^{134}Cs and ^{137}Cs for three farmland soil remediation methods, shown for the in situ depth coordinate \bar{z} . Note break in horizontal axes at 1 kBq/kg to show full distribution of activity with depth.

214 soil samples at the other locations. A 12.5 by 12.5 m mesh was overlaid onto a map of the area (Fig. 3).
 215 Cells containing a soil sampling location were assigned the inventory and relaxation mass depth for that
 216 sample. We adopted a simple interpolation method to assign inventories and relaxation mass depths to
 217 the other cells on the mesh. The inventories and β values were set equal to the values applicable at the
 218 nearest cell hosting a sampling location. Cells equidistant from more than one sampling location were
 219 assigned inventories randomly from one of the equidistant locations. Because of a large disparity between
 220 soil activity levels inside and outside the bounds of the remediated area, locations outside the remediated
 221 area were assigned the inventory of the closest of either location 17 or 18. It would also be possible to
 222 employ other interpolation techniques to assign inventories to cells without soil samples, for example based
 223 on inverse distance weighting techniques or Kriging (IAEA, 2003).

224 The assigned inventories for all cells across the area are depicted in Fig. 5. The mesh size simulated in
 225 the tool was 149 by 149 cells for both dose rate prediction methods.

226 2.3.4. Evaluation of farmland soil remediation methods

227 To evaluate different methods for remediating farmland soils, we used the tool to calculate air dose rates
 228 after remediation by topsoil removal, reverse tillage, or topsoil-subsoil layer interchange. Figure 6 shows a
 229 typical exponential depth distribution for ^{134}Cs and ^{137}Cs within undisturbed farmland soil in Fukushima
 230 Prefecture (solid black lines). The ^{134}Cs to ^{137}Cs activity ratio is applicable on December 01, 2011. This date
 231 falls within a pilot project on decontamination techniques, and allows comparison of dose rate predictions
 232 from the tool against environmental measurements from the decontamination project (JAEA, 2015b).

233 The relaxation mass depth of the exponential profile in Fig. 6 is $\beta = 1.13 \text{ g/cm}^2$. This follows the result
 234 from the first soil depth distribution sampling campaign (Table 2). The air dose rate under these ^{134}Cs and

235 ^{137}Cs inventories and depth profiles is $1.25\ \mu\text{Sv/h}$ before remediation, including a $0.05\ \mu\text{Sv/h}$ contribution
236 from natural background radiation.

237 The different remediation methods alter the activity depth distributions of the farmland soil. Figure
238 6 shows idealized activity depth distributions after topsoil removal, reverse tillage, or topsoil-subsoil layer
239 interchange. We used the tool to evaluate the air dose rate after completion of each of these remediation
240 options.

241 Topsoil removal involves mechanically stripping the top 5 cm of the soil, and disposing the excavated soil
242 as radioactive waste. The activity profile for the remaining soil is, to a first approximation, the exponential
243 distribution for depths greater than 5 cm prior to decontamination (dotted red lines in Fig. 6).

244 Reverse tillage employs a tractor pulled plough to invert the topsoil. The ploughing creates small ridges
245 and furrows on the land surface, which flatten off as the soil weathers and relaxes. We approximated the
246 soil as being homogeneously mixed after this process. Ploughing down to a depth of 25 cm thus results in a
247 constant radioactivity profile initially with depth, followed by the exponential distribution at depths below
248 25 cm (dashed blue lines in Fig. 6).

249 In Topsoil-subsoil layer interchange a layer of topsoil is switched with a layer of subsoil. Typically a
250 topsoil layer down to 15 cm is excavated with a digger and this soil is placed aside on a plastic sheet. The
251 next 15 cm of subsoil is then excavated and stored temporarily on adjacent ground. The pit that has been
252 created is refilled by first adding a 15 cm layer of the original topsoil, and then levelling to the ground surface
253 with the excavated subsoil layer. This strategy can be approximated as creating two homogenized layers of
254 activity concentration below the ground surface. The top layer, down to 15 cm depth, contains the activity
255 originally between the depths of 15 cm and 30 cm. The subsequent 15 cm thick layer below contains the
256 activity that was originally in the top 15 cm of soil (green dash-dot lines in Fig. 6).

257 To model these remediation scenarios, we considered remediation of a $37.5\ \text{m}$ by $37.5\ \text{m}$ ($1.4\ \text{km}^2$) area
258 of land, equivalent to a 3 by 3 square of cells on the simulation mesh (Fig. 7). The simulation models
259 consisted of remediated depth distributions within these cells, while the depth distributions outside the area
260 remained unchanged. Reductions in the dose rates were calculated for the center and near to the corners
261 of the remediated square of land. All dose rate evaluations included a $0.05\ \mu\text{Sv/h}$ contribution from natural
262 background radiation.

263 3. Results and discussion

264 3.1. Dose rates above flat, undisturbed fields

265 The predictions for air dose rates above flat, undisturbed fields made using the measured activity depth
266 profiles compare well with the dose rates measured at the sampling sites, as shown by Fig. 8(a). The
267 correlation holds over the range of dose rates covered by the dataset ($0.09\text{--}5.3\ \mu\text{Sv/h}$). The predicted dose

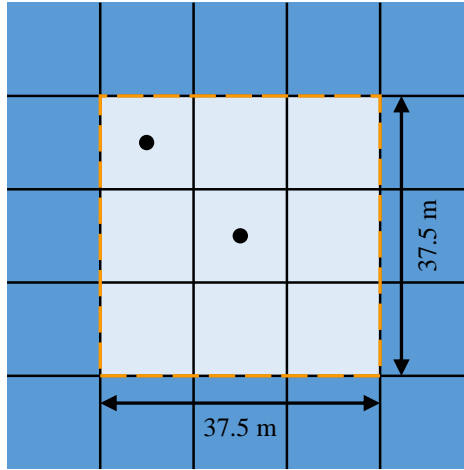


Figure 7: Setup of farmland soil remediation simulations: light blue area within orange dashed line is remediated land. Land in the dark blue area outside is not remediated. Dose rates before and after remediation were calculated for the locations marked by black spots.

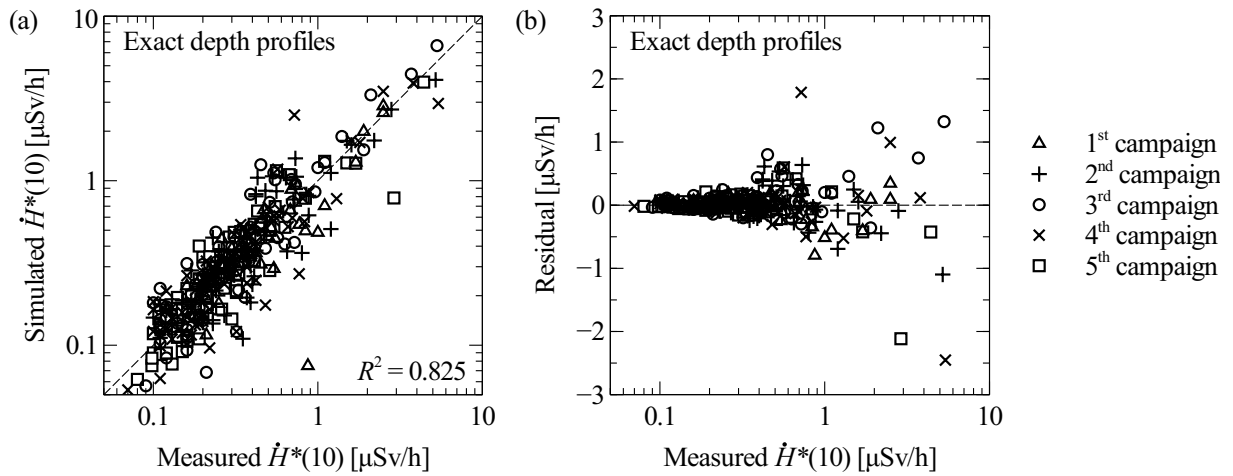


Figure 8: Correlation between measured air dose rates from the soil sampling campaigns and simulation predictions using measured soil depth profiles as inputs. (a) Measurement-prediction correlation. The dotted line indicates $y = x$. (b) Scatter plot of residual errors in the predictions. The dotted line is $y = 0$.

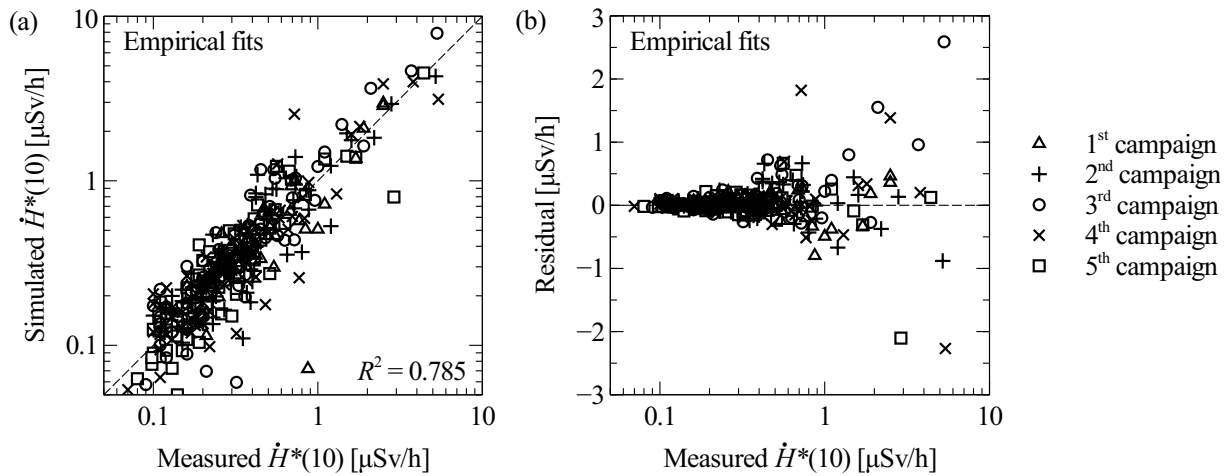


Figure 9: As per Fig. 8, except showing dose rate predictions made the exponential and hyperbolic secant fits to the measured soil activity depth distributions.

268 rates are always within a factor of three of the true dose rate, with one exception. At one site a dose rate
 269 of $0.87 \mu\text{Sv/h}$ was observed, but the tool predicted $0.075 \mu\text{Sv/h}$.

270 The residual differences between the predictions and the measured dose rates are shown in Fig. 8(b). A
 271 positive residual indicates an over-estimation by the tool, and a negative residual, an under-estimate. There
 272 is no tendency for the tool to either over-estimate or under-predict dose rates across the range of dose rates
 273 measured in the surveys.

274 Tyler et al. (1996) noted previously that individual soil samples can be poor representations of the mean
 275 soil activity across a wide area. The mean free path in air of the primary gamma rays emitted by $^{134/137}\text{Cs}$
 276 decay is around 100 m. Satoh et al. (2014) showed that radioactivity within 500 m contributes significantly
 277 to an air dose rate. Thus, the total volume of soil contributing to the dose rate, down to a depth of 8 cm,
 278 is $62\,800 \text{ m}^3$. As the volume of soil collected down to the same depth with a 15 cm by 30 cm scraper plate
 279 is 0.0036 m^3 , the sample represents only $6 \cdot 10^{-8}$ parts of the total soil volume contributing to the air dose
 280 rate.

281 Highly variable ^{134}Cs and ^{137}Cs activity concentrations are often found between different soil samples
 282 taken at the same location. Saito et al. (2015) confirmed this was the case for soil samples taken in Fukushima
 283 Prefecture. The variations are caused by heterogeneity in the fallout deposition, and by scrubbing and
 284 concentration of fallout nuclides by local earth surface processes. Therefore, a large sampling uncertainty
 285 for the inventory of the total soil volume contributing to the air dose rate should be expected if only a single
 286 soil sample is available. We ascribe the sampling uncertainty from the scraper plate measurement as the
 287 main source of error in the predictions for the air dose rate shown in Fig. 8(a).

288 We next considered the quality of the dose rate predictions obtained by modelling the empirical fits to

289 the measured depth profiles (Fig. 9(a)). The coefficient of determination obtained in this case is slightly
290 lower than the models employing the measured depth profiles directly ($R^2 = 0.785$ versus 0.825). The
291 slight difference in R^2 values is caused by the predictions for the high dose rate locations being slightly less
292 accurate from the models employing the empirical fitting functions. The residual errors for the predictions
293 at these high dose rate locations dominate the squared residuals sum in the calculation of R^2 , and hence
294 the resulting R^2 value.

295 The residuals for the predictions obtained by modelling the empirical fits are shown in Fig. 9(b). Exclud-
296 ing the high dose rate locations, the amount of scatter in the residuals is comparable to Fig. 9(a). Another
297 way to quantify the accuracy of the predictions is to consider the mean absolute percentage error. This
298 statistic is less susceptible to being skewed by the squared residuals for the predictions at the high dose
299 rate locations than R^2 . The mean absolute percentage error of the predictions made using the exact depth
300 profiles is 29%. This compares with a mean absolute percentage error of 30% for the predictions obtained
301 by modelling the fitted activity depth profile functions.

302 The results thus indicate that no significant error is introduced by modelling the empirical fits to the
303 activity depth profiles instead of the measured step-wise profiles. This conclusion necessarily depends on the
304 details of the soil sampling procedure. Matsuda et al. (2015) measured the activity within 0.5 cm layers of
305 topsoil, followed by 1 and 3 cm thick layers at deeper depths. If coarser soil layer thicknesses are employed,
306 modelling the empirical fits may yield more accurate predictions than modelling the measured depth profiles,
307 as it is plausible for the empirical fits to offer a better representation of the true activity profile in the soil.

308 3.2. Evolution of air dose rates

309 The distributions of $\beta_{\text{eff}}^{\dot{H}E^{*(10)}}$ values obtained from the exponential and hyperbolic secant fits to the
310 depth profiles are shown in Fig. 10(a) for the five soil sampling campaigns. Both the mean and median
311 values of $\beta_{\text{eff}}^{\dot{H}E^{*(10)}}$ increase over time (Table 2), indicating that the radiocesium is migrating deeper into the
312 soil.

313 The component of the mean air dose rate attributable to radiocesium at the flat, undisturbed fields is
314 plotted for the air survey campaigns with solid diamonds in Fig. 10(b). The data are plotted relative to
315 June 2011, the date of the first air dose rate survey (Table 3).

316 The solid line in Fig. 10(b) represents the decrement in dose rates that would be expected on the basis of
317 radioactive decay of ^{134}Cs and ^{137}Cs , and without migration of the radiocesium at the sites. The measured
318 dose rates decrease faster than expected by just radioactive decay.

319 Mikami et al. (2015b) explained the additional reduction in dose rates between June 2011 and December
320 2012 by migration of the cesium fallout deeper into the soil pack. This trend continued through 2013, as
321 shown by the results of our decay and migration calculations (open circles, Fig. 10(b)). The decay and

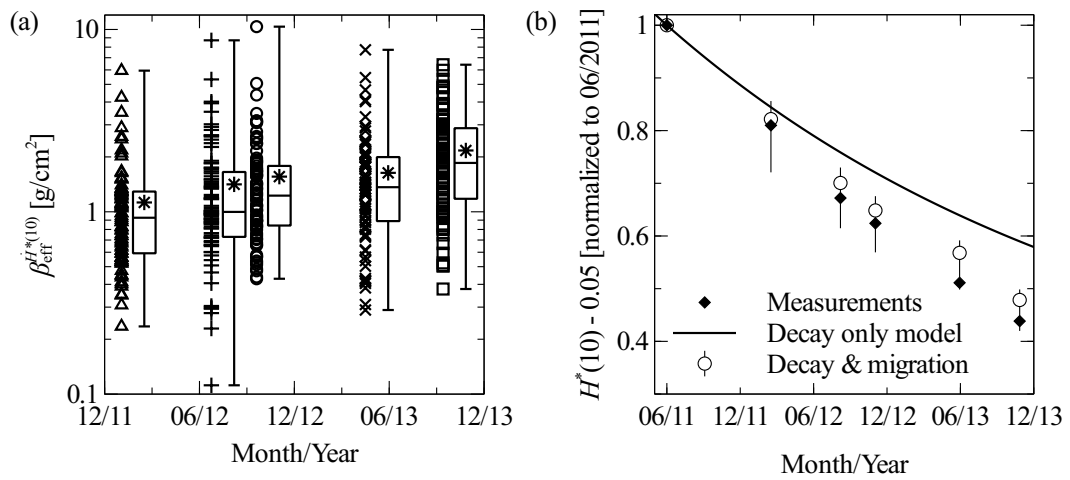


Figure 10: (a) Box and whisker plot showing distribution of $\beta_{\text{eff}}^{i*(10)}$ values over the five soil sampling campaigns. The whiskers show the maxima and minima of the distributions. The boxes show the range in between the 25th and 75th percentiles of distributions. The mean values are indicated by asterisks. The full distributions are plotted with symbols, offset to the left of each box and whiskers. (b) Measurements and modelling results for the reduction in air dose rate component attributable to radioactive cesium at locations of flat, undisturbed fields. The measurements (solid diamonds) show the mean air dose rate attributable to radiocesium from the air survey campaigns, normalized to the value at the first air dose rate survey (Table 3). The vertical bars on the data for the decay and migration model (circles) indicate results when varying $\beta_{\text{eff}}^{i*(10)}$ between 0.5–2.0 g/cm² at the time of the first air dose rate survey (June 2011).

322 migration model results are reasonably consistent with the measurements, although they tend to under-
323 estimate the reduction in dose rates by up to 10 %.

324 A source of uncertainty in the decay and migration model is the choice for mean $\beta_{\text{eff}}^{\dot{H}E^{*(10)}}$ for the first
325 air dose rate survey campaign (June 4 to July 8, 2011 - Table 3). The soil sampling campaigns by Matsuda
326 et al. (2015) commenced in December 2011, so cannot provide measurements to derive a mean $\beta_{\text{eff}}^{\dot{H}E^{*(10)}}$
327 value applicable to this period. The circles in Fig. 10(b) represent the assumption that $\beta_{\text{eff}}^{\dot{H}E^{*(10)}} = 1.0 \text{ g/cm}^2$
328 in June 2011. ICRU (1994) cites β values for atmospheric radionuclide fallout in the range 0.1–4 g/cm² for
329 up to one year after fallout deposition. These results are based on measurements for cesium radioisotopes
330 from Chernobyl fallout in Europe and Western Russia.

331 Takahashi et al. (2015) measured depth distributions at two grassland sites and three abandoned agricul-
332 tural fields in Fukushima Prefecture between June 21–28, 2011. They found that the exponential distribution
333 was a good fit for the measured depth profiles, with β values in the range 0.60–3.08 g/cm². However, they
334 noted that the site giving the highest relaxation mass depth (3.08 g/cm²) was pasture land where the soil
335 had been disturbed by cattle grazing. Excluding this site from their dataset yields a mean β value from four
336 sites of 1.20 g/cm².

337 To determine the sensitivity of our decay and migration model on the choice for the mean $\beta_{\text{eff}}^{\dot{H}E^{*(10)}}$
338 value for the first air dose rate survey, we considered the effect of varying this parameter in the range 0.50–
339 1.20 g/cm². This is a range of values that we consider credible for the period between June 4 and July 8,
340 2011, based on the previous literature cited and the mean $\beta_{\text{eff}}^{\dot{H}E^{*(10)}}$ value of 1.13 g/cm² derived from the
341 first soil sampling campaign in December 2011 (Table 2). The effect of varying the initial value of $\beta_{\text{eff}}^{\dot{H}E^{*(10)}}$
342 in this range is shown by vertical bars around circle markers in Fig. 10(b). The ranges indicated by these
343 bars include the measurements, but do not permit the decay only explanation for the reduction in dose
344 rates. The sensitivity analysis is thus consistent with the conclusion that migration of cesium deeper into
345 the ground was the main cause behind the additional decrement in dose rates.

346 There are two other factors that could plausibly explain the underestimation of the true dose rate
347 reduction by the model for decay and migration deeper into soil (Fig. 10(b)). Although Mikami et al. (2015a)
348 suggested that little migration of the radiocesium inventory in the horizontal direction had occurred, within
349 uncertainties their data are consistent with a possible small amount of horizontal migration (on the order
350 of 5–10 % of the inventory).

351 Another factor is as follows. Although the sites featuring in the air dose rate surveys were chosen to be
352 flat, open spaces (Mikami et al., 2015b), certain sites may include urban areas or areas with roads and paved
353 surfaces at the periphery. The wide field of view of environmental radioactivity means that radiocesium
354 within these areas contributes to the air dose rate. As the radiocesium within these areas has a shorter
355 ecological half-life than areas of grassland or agricultural areas (Kinase et al., 2014), i.e. the radiocesium is
356 more easily washed away, this could contribute to the underestimation of the dose rate reduction by the

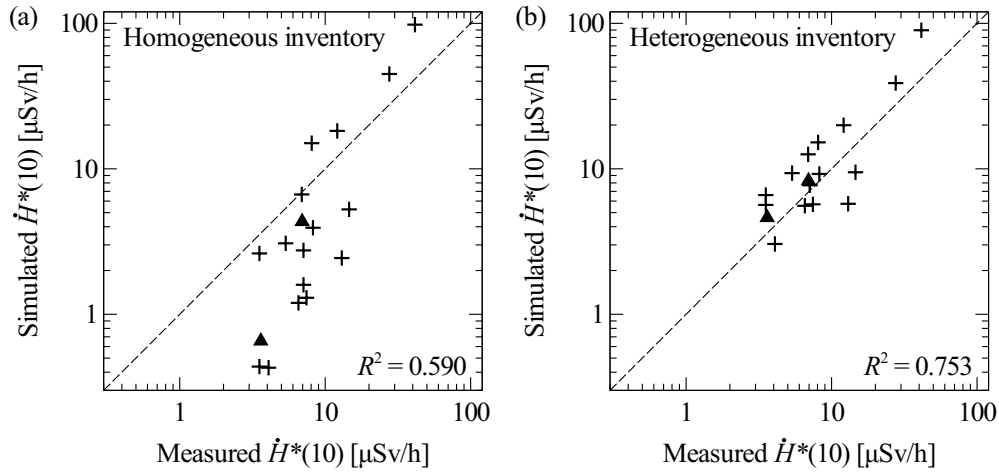


Figure 11: Correlation between measured air dose rates and predictions from soil activity levels at Ottozawa. (a) Assumed a spatially homogeneous ^{134}Cs and ^{137}Cs inventory. (b) Spatially varying inventory informed by all the soil sampling locations. Triangles indicate locations 5 and 14, where scraper plate samples yielded the depth distribution.

357 models in Fig. 10(b).

358 3.3. Effect of spatial variability in soil activity levels

359 The Ottozawa area was used to study the effect of spatial variations in the radiocesium distribution on
 360 air dose rates. The range of dose rates measured at Ottozawa in July 2014 varied between 3.5–41.4 $\mu\text{Sv/h}$
 361 (Table 4). This is a higher range of values than measured in the five soil sampling campaigns (section 3.1),
 362 as Ottozawa lies closer to FDNPP than the sites in the five soil sampling campaigns and is more highly
 363 contaminated with fallout from the accident.

364 Fig. 11 shows two sets of predictions for the air dose rates from soil activity measurements, plotted
 365 against the dose rates measured in the field. The predictions shown in Fig. 11(a) did not account for the
 366 spatial variations in soil activity levels. Figure 11(b) shows predictions from the models incorporating the
 367 measured spatial variations in the cesium inventory.

368 It is clear that modelling the spatial variations in the contamination distribution yields better predictions
 369 for the air dose rate. Therefore modelling the spatial variation is the better strategy if multiple soil samples
 370 across an area are available to include in the dose rate analysis.

371 The coefficient of determination is higher for the predictions modelling the spatial distribution ($R^2 =$
 372 0.753) than for the predictions assuming a homogeneous radiocesium distribution ($R^2 = 0.590$). The origin
 373 of the difference in the R^2 values is traceable to the model accounting for the spatial distribution yielding a
 374 better prediction for the highest dose rate site, location 17, with a measured dose rate of 41.4 $\mu\text{Sv/h}$, than
 375 the model assuming a homogeneous cesium distribution.

Table 4: Results of soil sampling and air dose rate predictions for Ottozawa area on July 24, 2014. **Bold** indicates soil samples taken with scraper plate apparatus. Other samples collected with U-8 cup. *Italic* indicates values of β inferred from depth distributions at locations 5 or 14.

Location	Inventory (MBq/m ²)		β (g/cm ²)	Measured $\dot{H}^*(10)$ (μ Sv/h)	Prediction based on assumption for Cs distribution (μ Sv/h)	
	¹³⁴ Cs	¹³⁷ Cs			Homogeneous	Heterogeneous
1	0.643	1.89	<i>3.60</i>	14.6	5.3	9.5
2	0.154	0.453	<i>3.60</i>	7.4	1.3	5.7
3	0.472	1.43	<i>3.60</i>	8.2	3.9	9.2
4	0.294	0.868	<i>3.60</i>	13.0	2.4	5.7
5	0.567	1.67	3.60	6.9	4.3	8.3
	0.483	1.47				
6	2.24	6.61	<i>3.60</i>	12.1	18.2	19.9
7	0.141	0.419	<i>3.60</i>	6.6	1.2	5.6
8	0.189	0.566	<i>3.60</i>	7.1	1.6	7.8
9	1.82	5.52	<i>3.60</i>	8.1	15.0	15.2
10	0.265	0.811	<i>1.83</i>	7.1	2.7	7.7
11	0.0342	0.122	<i>1.83</i>	4.1	0.4	3.0
12	0.258	0.758	<i>1.83</i>	3.5	2.6	5.6
13	0.305	0.888	<i>1.83</i>	5.4	3.1	9.3
14	0.0714	0.210	1.83	3.6	0.7	4.6
	0.0506	0.145				
15	0.0391	0.114	<i>1.83</i>	3.5	0.4	6.6
16	0.665	1.95	<i>1.83</i>	6.9	6.7	12.6
17	8.78	26.1	<i>1.83</i>	41.4	98.0	89.7
	7.88	23.7				
	12.4	37.4				
	9.95	29.3				
	9.68	29.0				
18	9.37	27.8	<i>1.83</i>	27.6	44.9	38.8
	2.17	6.34				
	3.95	11.8				
	4.42	13.3				
	2.41	7.22				

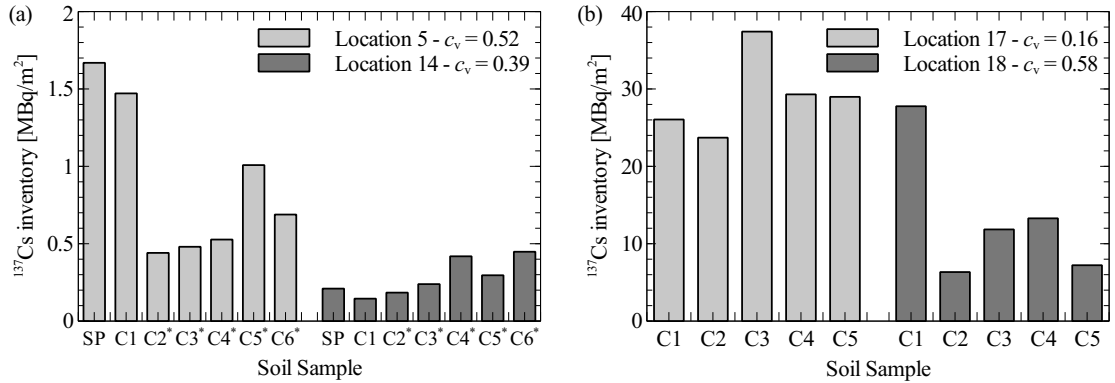


Figure 12: Variation in ^{137}Cs inventory between soil samples at Ottozawa locations 5, 14, 17 and 18. SP denotes a scraper plate sample, and C1, C2, etc. denote the U-8 cup samples. Asterisks denote soil samples taken on October 10, 2014 and inventory decay corrected to July 24, 2014. c_v denotes coefficient of variation between the soil sample inventories at each location. c_v is the ratio of the sample standard deviation to the mean.

376 The mean absolute percentage error for the predictions taking into account the spatial heterogeneity of
 377 the activity is 47%. This result is higher than the $\approx 30\%$ mean absolute percentage error for the predictions
 378 for dose rates above flat, undisturbed fields (section 3.1). This difference is also observable by comparing
 379 the quality of the correlation in Figs. 8(a) and 9(a) with Fig. 11(b).

380 There are a number of distinctions between the modelling at Ottozawa and the flat, undisturbed fields
 381 that contribute to a higher uncertainty for the predictions at Ottozawa. There is a higher degree of mea-
 382 surement uncertainty for many of the soil samples at Ottozawa than for the sites visited in the five soil
 383 sampling campaigns, as the U-8 cup samples collect smaller volumes of soil than the scraper plate. This
 384 can be shown by examining the inventories from the multiple soil samples taken at locations 5, 14, 17 and
 385 18 (Fig. 12). There is large variation between the inventories between the samples at each location. The
 386 highest variation is seen for location 18, where the largest inventory is four times greater than the smallest
 387 inventory.

388 The mean coefficient of variation for the four Ottozawa locations is $c_v = 0.41$. This is larger than the
 389 mean $c_v = 0.36$ observed by Saito et al. (2015) for locations within a 100 km radius of FDNPP, which are
 390 similar to the sites visited in the soil sampling campaigns. Mishra et al. (2015) independently reported a
 391 coefficient of variation of 0.27 between four samples at another site similar to those visited in the Matsuda
 392 et al. (2015) soil sampling campaigns.

393 Another factor contributing to the uncertainty for the predictions at Ottozawa include the fact that
 394 scraper plate samples were only taken at locations 5 and 14. The depth distribution at other locations
 395 had to be inferred from these two measurements. It is notable that some of the best dose rate predictions
 396 obtained for Ottozawa were at locations 5 and 14 (triangles - Fig. 11(b)).

Table 5: The percentage reduction in the air dose rate after remediation of farmland soils by three different methods. The simulation input data (depth profiles, activity levels, etc.) were applicable on December 01, 2011. Full remediation means that all 149 by 149 cells on the simulation mesh were modelled as remediated land.

Remediation method	Observed results (JAEA, 2015b)	Simulation results		
		Centre of 37.5 by 37.5 m remediated area	Corner of 37.5 by 37.5 m remediated area	Full remediation
Topsoil removal	40–70 %	73 %	65 %	96 %
Reverse tillage	30–60 %	54 %	46 %	71 %
Topsoil-subsoil layer interchange	≈65 %	68 %	60 %	90 %

3.4. Evaluation of farmland soil remediation methods

We used the tool to evaluate the effectiveness of three methods for remediating farmland soils for decreasing air dose rates (Table 5). We calculated the reduction in air dose rate at the center and the corner of a 37.5 by 37.5 m square area of remediated land, and compared with field results from a decontamination pilot project in Fukushima Prefecture (JAEA, 2015b). Also shown in Table 5 are theoretical limits for the reduction in dose rates, calculated assuming remediation of all the land surface.

The performance of the topsoil removal and layer interchange methods of remediation are similar. Both methods yield ≈65 % reduction in air dose rates for the square area of remediated land. These methods are more effective than reverse tillage, where the calculations indicated a ≈50 % reduction in the air dose rate.

Experience from the decontamination pilot project (JAEA, 2015b) suggests a range of dose rate reductions for topsoil removal and reverse tillage. A number of factors affect the percentage reduction in dose rates after land remediation, including the size of the area remediated, the homogeneity of the remediation actions, and the magnitude of the dose rate before remediation relative to the natural background dose rate. The remediation parameters, e.g. the thickness of topsoil removed, or the depth of ploughing when performing reverse tillage, may also have varied slightly. However, the general correspondence in Table 5 between the predictions from the tool and observed results is encouraging.

One advantage of the layer interchange method over topsoil removal is that it does not create waste radioactive soil for disposal. However, the fact that the contaminated soil remains at the site after remediation, albeit below the ground surface, is tempered by the possible availability of the radioactive contaminants for uptake by crops or vegetation in future. This point may affect the viability of farming these lands after remediation if the crops or livestock produced approach food safety limits for radioactive cesium content.

418 **4. Conclusions**

419 The simulation predictions for dose rates at flat, undisturbed fields from soil activity depth profiles showed
420 good correlation with measurements. Little error was introduced by modelling exponential and hyperbolic
421 secant fits to measured activity depth profiles. This conclusion necessarily depends on the experimental
422 parameters for measuring activity depth distributions. Soil layers at least as fine as collected by Matsuda
423 et al. (2015) are recommended if the data are to be used to evaluate air dose rates. Simulations of the
424 Ottozawa area demonstrated that modelling spatial variations in contamination levels improves the quality
425 of dose rate predictions. This approach is recommended if multiple soil activity samples across an area are
426 available.

427 The main uncertainty in air dose rate predictions derived from soil samples is due to the sampling uncer-
428 tainty for the true soil inventory distribution based on the limited volume samples. In situ or mobile gamma
429 spectroscopy surveys offer a more comprehensive route to assess environmental radiocesium distributions,
430 as they are subject to much lower sampling uncertainty (ICRU, 1994). The results from these surveys could
431 be used to inform inputs for dose rate modelling and improve prediction quality.

432 Simulations for the decrement in air dose rates seen at undisturbed, flat fields in Fukushima Prefecture
433 for the first 20 months following the Fukushima Dai-ichi accident were consistent with the hypothesis that
434 radiocesium decay and deeper migration in soil are the main responsible factors. Simulations of three
435 farmland soil remediation methods for reducing air dose rates demonstrated that topsoil removal and layer
436 interchange strategies have similar levels of effectiveness, and both methods are more effective than reverse
437 tillage.

438 Techniques for modelling air dose rates from soil activity concentrations, such as described in this paper,
439 would be effective for evaluating air dose rates in future and for planning land remediation works.

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