# Radon-222 atmospheric transport model: Statistical evaluation and analysis for an optimized framework



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## RADON-222 ATMOSPHERIC TRANSPORT MODEL: STATISTICAL EVALUATION AND ANALYSIS FOR AN OPTIMIZED FRAMEWORK

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### Plan

#### Context

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### Context

- In the event of an incident or accident involving radioactive materials, IRSN provides guidance to public authorities on the technical, public health and medical measures to be taken to protect the population and the environment.
- The monitoring network Téléray in France: +400 of gamma dose rate monitoring stations recording data each ten minutes all year round.
- Several times a year, alarms of this emergency monitoring network are triggered: gamma dose rate peaks due to Radon-222 progeny scavenged by precipitations.
- Although these peaks do not present any risks to the population or environment, it is necessary to determine their origins: natural radioactivity or an accidental release of radioactive materials

Source: EGU2020-14940 Lessons learned on atmospheric radon modelling by statistical model-to-data comparison on gamma dose rate peaks





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### Simulation framework





### **Methods: Model to measurement comparison**





### **Methods: Metrics**

#### Peak-to-peak

**Recall** =  $\frac{TP}{TP+FN}$  tendency to miss events

**Precision** =  $\frac{TP}{TP+FP}$  tendency to produce false alarms

#### Time series

1st Wasserstein distance: minimal cost of transforming one distribution to another:
lower → higher similarity

**Fac 2**: % 
$$\frac{ddd_obs}{2} \leq sim_ddd \leq 2obs_ddd$$

Pearson correlation coefficient: from -1 (total anti-correlation) to 1 (total correlation)



### Methods: case study

#### Comparison over a period of 6 months with more than 15 000 peaks > 10 nSv/h



Fig.1: Total number of peaks by threshold

Two types of precipitation data:

predictions from Météo-France Arpege Model:
Δt=1h and Δx=Δy=0.1°≈10km

 radar observations : Δt=5 minutes and Δx=Δy=1km historical and more precise. Used for reanalysis.

- Two exhalation flux configurations:
- Constant in time : BERAD<sup>1</sup> (France) + Karstens<sup>2</sup> (rest of Europe)
- Varying by month: Karstens<sup>2</sup> (all of Europe)

1: Bureau d'étude et d'expertise du radon – IRSN – based on lelsch et al. 2017 http://dx.doi.org/10.1016/j.jenvrad.2016.05.022 2: Karstens, U., Schwingshackl, C., Schmithüsen, D., and Levin, I.: A processbased <sup>222</sup>radon flux map for Europe and its comparison to long-term observations, Atmos. Chem. Phys., 15, 12845–12865, https://doi.org/10.5194/acp-15-12845-2015, 2015



### **Results: three primary statistical evaluations**



Using radar observations instead of rain predictions yields better results → subsequent evaluations are carried out using radar data

Recall is higher than

false negatives

precision: failures tend to

be false positives more than



#### Setting: Fac2, $\Delta t$ =3h, combined BERAD+Karstens source term



### **Results: statistical evaluation**

The improvement from increasing the time tolerance from 3h to 6h is marginal compared to that obtained considering Fac5 instead of Fac2 between the observed and simulated peaks.

failures are due to simulated values being far apart from observations in amplitude rather than in time

This is confirmed by the distribution of bias between observed and simulated peaks: 0.75 quantile < 0

a consistent tendency to overestimate peaks







Fig.3: Distribution of bias between observed and simulated peaks





**Results: statistical evaluation** 

Wasserstein distance exceptionally high for stations with anomalous observations (missing data, technical malfunction, etc.).

Good correlation with some exceptions.

Low Fac2: bias between observations and simulations is mainly due to the inefficient extraction of the background radiation from the observations.



### **Results: effect of exhalation flux on peaks**



Fig.5: Effect of source term on the distribution of bias (left) and metrics (right) by threshold

#### BERAD+Karstens has a negative bias in mean

- -> its gamma dose rate peaks have higher magnitude
- -> lead to a greater number of false positives

**Karstens** has a better precision (less false positives), worse recall (more false negatives) but globally better accuracy (higher f1\_score).



The choice of an exhalation flux should be guided by which aspect is more critical: missing true events or producing false alarms

### **Effect of station environment**

Weak correlations between a station's metrics and characteristics of its immediate 1km environment: soil occupancy<sup>1</sup>, type<sup>2</sup> and geological age<sup>3</sup>



Fig.6: Pearson correlation coefficient between a station metrics and type of soil (left) and soil occupancy (right)

Same conclusion from Anova tests: p-value >  $\alpha$  =0.05 significance level.

The model's metrics are unaffected by the measurement station immediate environment.



1: CORINE Land Cover

3: BRGM geol. map at 1/1 000 000

2: INRAE

### **Conclusions**

We presented a model capable of successfully simulating around 60% of observed gamma dose rate peaks within a Fac2 in amplitude and 3h time tolerance.

This performance is sensitive to the used exhalation flux and precipitations accuracy, but not to the soil characteristics around the measurement station

Our simulated gamma dose rate peaks are often higher than observed ones, which makes the model prone to produce false positives more than false negatives



### Outlook

Many optimization paths: exhalation flux, wet deposition schemes...etc.

Evaluation on the capacity to predict Radon air concentration

Validation over a longer period of 2 years

