



## Spatio-temporal variation of radionuclide dispersion from nuclear power plant accidents using FLEXPART ensemble modeling

5 S. Omid Nabavi<sup>1</sup>, Theodoros Christoudias<sup>1</sup>, Yiannis Proestos<sup>1</sup>, Christos Fountoukis<sup>2</sup>, Huda Al-Sulaiti<sup>2</sup>,  
Jos Lelieveld<sup>1,3</sup>

<sup>1</sup>Climate & Atmosphere Research Centre, the Cyprus Institute, Nicosia, 2121, Cyprus (o.nabavi@cyi.ac.cy)

<sup>2</sup>Qatar Environment and Energy Research Institute (QEERI), Hamad Bin Khalifa University, Doha, Qatar

<sup>3</sup>Max Planck Institute for Chemistry, Mainz, 55128, Germany

10 *Correspondence to:* S. Omid Nabavi<sup>1</sup> ([o.nabavi@cyi.ac.cy](mailto:o.nabavi@cyi.ac.cy))

### Abstract.

We estimate the seasonal and diurnal changes in the transport and intensity of radionuclides including Iodine-131 (<sup>131</sup>I) and Cesium-137 (<sup>137</sup>Cs), transported to Qatar from a fictitious accident at the Barakah nuclear power plant (B-NPP) in UAE. For dispersion modeling, we have used the Lagrangian particle/air parcel dispersion model FLEXible PARTicle (FLEXPART) driven by forecast and (re)analysis products, and coupled with the Weather Research and Forecasting model (FLEXPART-WRF). A four-member ensemble of meteorological inputs, including one forecast dataset (CFSv2) and three (re)analysis datasets (native resolution and downscaled FNL and downscaled ERA5), is used to force FLEXPART/FLEXPART-WRF. According to the age spectrum of Lagrangian particles, radionuclides entered southern Qatar about 10 to 20 hours after emission, and almost all emitted particles are transported to and/or deposited in the study area within the 80 hours after the release. A higher number of long-lived particles was found in FNL simulations and when particles are released in the afternoon and spring. The highest levels of simulated <sup>131</sup>I concentrations and <sup>137</sup>Cs deposition were found in FNL simulations in the south/southeast of Qatar. The frequent coincidence of high radionuclide concentrations and deposition with particles released between 5 a.m. and 2 p.m. and in the cold period of the year was attributed to diurnal and seasonal changes in the planetary boundary layer height (PBLH) and synoptic circulations. The difference in input PBLH explains well the inter-member variations of simulated radionuclide concentrations. Simulated concentrations were found with the same level of consistency as reported for real case studies.

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## 1 Introduction

Modeling the spatio-temporal distribution of radioactive materials through chemical transport models (CTMs), especially after major nuclear events, has received widespread attention (Chino et al., 2011, Stohl et al., 2012, Christoudias and Lelieveld, 2013, Evangeliou et al., 2017). Whether explicitly stated or implied, all these studies seek to determine the magnitude and  
35 transport of radionuclides at different spatial scales. Such research activities help to improve the performance of CTMs as the core of preparedness programs for potential nuclear accidents (or releases) related to the increasing number of nuclear facilities (Farid et al., 2017). However, case studies of real accidents of the order of a few days are not suited to examine the impact of seasonal (atmospheric) changes on the radionuclide dispersion. For instance, Long et al. (2019) studied the effect of the East  
40 Asian northeast monsoon on the transport of radionuclides from the Fukushima nuclear power plant accident to the tropical western Pacific and Southeast Asia. They found that these regions received lower airborne radioactivity than other regions in the northern hemisphere due to the late arrivals of radionuclide plumes carried by the monsoon circulations. That is, the dispersion of radionuclides significantly differs under various atmospheric conditions. Maurer et al. (2018) found that the performance of dispersion models is largely dependent on successfully capturing boundary layer processes that vary in seasons and times of the day. In addition, the meteorological inputs of CTMs are produced by atmospheric models that in some cases  
45 exhibit variability when simulating atmospheric conditions over nuclear accident areas (Arnold et al., 2015). This is mainly due to the differences in their initial conditions, spatial and temporal resolutions, mathematical formulation, physical parameterization, etc.

Using an ensemble of meteorological inputs, this study primarily aims to investigate the seasonal and diurnal changes in the transport and surface concentration and deposition magnitude of radionuclides in the event of a potentially possible nuclear  
50 accident. Two radioactive tracers including Iodine-131 ( $^{131}\text{I}$ ) and Cesium-137 ( $^{137}\text{Cs}$ ) are assumed to be released from a fictitious accident at the Barakah nuclear power plant (B-NPP) in the UAE. Both radionuclides are emitted as gas, but due to lower volatility,  $^{137}\text{Cs}$  condenses onto aerosol particles shortly after the release (Christoudias and Lelieveld, 2013). The relatively long half-life of  $^{137}\text{Cs}$  (about 30 years), compared to  $^{131}\text{I}$  (about 8 days), gives significance to the amount of  $^{137}\text{Cs}$  that undergoes surface deposition and to  $^{131}\text{I}$  concentration in the biosphere (Tsuruta et al., 2019, Takagi et al., 2020, Kinase  
55 et al., 2020, Wai et al., 2020). For dispersion modeling, we use the Lagrangian particle model FLEXible PARTicle (FLEXPART) (Stohl et al., 1998, Stohl et al., 2005, Brioude et al., 2013). The Eulerian CTMs model the transport of air pollutants by solving numerically the equation for the conservation of mass (Moussiopoulos, 1997, Zhang and Chen, 2007) which is computationally expensive (Brioude et al., 2013). Furthermore, in an Eulerian model, a particle released from a point source loses its position in the grid-box (numerical diffusion). Lagrangian particle dispersion models (LPDMs) instead have  
60 minimal numerical diffusion as they accurately model the trajectory of each particle individually (Nabi et al., 2015). Furthermore, LPDMs are more computationally efficient for particle dispersion modeling because these models calculate the advection and diffusion only for the location of each particle rather than for the entire model domain. However, LPDMs also have limitations. They suffer from numerical errors when interpolating meteorological fields in space and time. In some cases,



65 due to the lack of accuracy required to simulate the stochastic motion of particles and/or because the vertical velocity is not  
very precisely mass balanced with horizontal winds, particles may not remain well-mixed during the simulation (Brioude et  
al., 2013). Regardless of which type of CTM is being deployed, Girard et al. (2016) showed that uncertainties in meteorological  
fields, namely wind speed and direction and precipitation, and emission rate substantially contribute to errors in dispersion  
modeling. The three common methods for sensitivity analysis of LPDM simulations to meteorological information are: i)  
70 creating perturbations in the horizontal and vertical location of particles, ii) using a single meteorological model with  
perturbations to the initial conditions and/or model physics, and iii) using a suite of different meteorological models (Galmarini  
et al., 2004). By adopting the third approach, we used a four-member ensemble in which the performance of the forecast  
member, forced by 6-hourly data from NCEP climate forecast system version 2 (CFSv2), is compared against the (re)analysis  
members. (Re)analysis-based simulations are expected to be closer to (unavailable in a real-world scenario) actual values than  
forecast-based ones (Leadbetter et al., 2022). Two (re)analysis members are forced by dynamically downscaled meteorological  
75 inputs to study the effect of downscaling on FLEXPART dispersion modelling. We also studied the sensitivity of simulations  
regarding the choice of turbulence schemes under convective conditions. To study pollution risks for local communities, the  
radionuclide simulations were examined in relation to the population densities of the catchment region of interest (Qatar). The  
paper proceeds as follows: Section 2 outlines the LPDM, meteorological inputs, study area, and source term of radionuclides  
used in this study. Results are presented and discussed in Section 3. The study conclusions are given in Section 4.

## 80 **2 Ensemble model configuration**

We have used FLEXPART and FLEXPART driven by the Weather Research and Forecasting (FLEXPART-WRF) model for  
the dispersion modeling of  $^{131}\text{I}$  and  $^{137}\text{Cs}$ , fictitiously released from the B-NPP. The following is a brief description of the  
FLEXPART modeling structure.

### **2.1 FLEXPART and FLEXPART-WRF dispersion modelling**

85 The details of the FLEXPART and FLEXPART-WRF modeling are available in Stohl et al. (2005) and Brioude et al. (2013),  
respectively. This subsection only discusses the principles of FLEXPART modeling that facilitate the presentation of results  
in the next section. FLEXPART, as a LPDM, is developed based on the zero acceleration scheme (Eq. 1). It solves a Langevin  
equation (Eq. 2) for modeling the trajectories of Lagrangian particles. The new location of particles is under the influence of  
large-scale winds, local turbulence (stochastic component), and mesoscale motions.

$$90 \quad X(t + \Delta t) = X(t) + v(X, t)\Delta t \quad (1)$$

$$\frac{dX}{dt} = v[X(t)]$$



where  $t$  is time,  $\Delta t$  is the time increment,  $X$  is the position vector, and  $v = \bar{v} + v_t + v_m$  the wind vector that is composed of the grid-scale wind  $\bar{v}$ , the turbulent wind fluctuations  $v_t$  and the mesoscale wind fluctuations  $v_m$ . FLEXPART also quantifies changes to the mass, or mixing ratio, of advected particles by computing various removal processes (Stohl and Thomson, 1999, Cassiani et al., 2015, Tipka et al., 2020). Turbulent motions  $v_{t_i}$  for wind components  $i$  are parameterized assuming a Markov process based on the Langevin equation (Eq. 2).

$$dv_{t_i} = \alpha_i(x, v_t, t)dt + b_{ij}(x, v_t, t)dW_j \quad (2)$$

Where  $\alpha$  is the drift term,  $b$  the diffusion term, and  $dW_i$  incremental components of a Wiener process with mean zero and variance, which are uncorrelated in time (Stohl et al., 2010). The minimum value of  $\Delta t_i$  is 1 second.  $\Delta t_i$  is used only for the horizontal turbulent wind components of Eq. 2.

$$\Delta t_i = \frac{1}{cti} \min(\Delta \tau_{L\omega}, \frac{h}{2\omega}, \frac{0.5}{\frac{\partial \sigma_{\omega}}{\partial z}}) \quad (3)$$

where  $h$  is the height of the atmospheric boundary layer (ABL). For solving the Langevin equation for the vertical wind component, a shorter time step  $\Delta t_{\omega} = \frac{\Delta t_i}{\text{ifine}}$  is used. Under convective conditions, when turbulence is skewed, larger areas are occupied by downdrafts (rather than updrafts). This may lead to higher surface concentrations (deposition) in areas neighboring the pollution sources (Pisso et al., 2019). To investigate the sensitivity of radionuclide magnitudes to the used turbulence scheme, FLEXPART-WRF members were rerun after replacing the default Gaussian turbulence model (GTM) with the skewed turbulence model (STM) (Luhar et al. (1996) and Cassiani et al. (2013)). The output of either of these schemes is used for the computation of the vertical velocity component of the drift term in Eq. 2. The implementation of STM requires shorter time steps,  $dt$  in Eq. 2, to better resolve turbulence in the convective planetary boundary layer. Hence, we used a tenfold finer time step, as recommended by Pisso et al. (2019), in sensitivity runs. For the computation of  $\sigma_{v_i}$  and  $\Delta \tau_{L_i}$ , FLEXPART uses the parameterization scheme proposed by (Hanna, 1982). In order to take mesoscale motions ( $v_m$ , in Eq. 2) into account, a similar method as Maryon (1998) is followed.

In dispersion modeling, particle mass reduction mainly occurs through three processes: radioactive decay, dry deposition, and wet deposition. The following exponential equations characterize the radioactive decay (Eq. 4) and wet deposition (Eq. 5).

$$m(t + \Delta t) = m(t) \exp(-\Delta t / \beta) \quad (4)$$

where  $m$  is particle mass, and the time constant  $\beta = \frac{T_{1/2}}{\ln(2)}$  is determined from already calculated the radionuclide half-life  $T_{1/2}$ .

$$m(t + \Delta t) = m(t) \exp(-\Lambda \Delta t) \quad (5)$$

The scavenging coefficient  $\Lambda$  is calculated differently depending on whether the particle in question is in the aerosol or gas phase and whether scavenging takes place inside or below clouds (Stohl et al., 2005).

In this study, we have used FLEXPART 10.4 and a modified version of FLEXPART 9.02 to ingest meteorological simulations from the Weather Research and Forecasting (WRF) (Grell et al., 2005) (Hereafter referred to as FLEXPART and FLEXPART-WRF, respectively). Compared to previous versions, FLEXPART has undergone a significant revision in the calculation of



wet deposition. In recent updates, the dependency of wet deposition on the size of aerosol particles and the type of precipitation has been considered. In addition, FLEXPART only calculates wet deposition for cloudy pixels in which the precipitation rate exceeds  $0.01 \text{ mm h}^{-1}$ . Therefore, the accuracy of cloud pixel detection plays a critical role in the accuracy of location and amount of wet deposition simulated by FLEXPART. In the previous versions (including version 9.02 used in the development of FLEXPART-WRF 3.2), in-cloud grid cells were defined as those with relative humidity above 80% and any grid cells beneath these grid cells as below-cloud ones. Because of recent updates, FLEXPART now uses 3D fields of cloud water mixing ratio and the relative fractions of ice and liquid water (type of precipitation). Therefore, it is expected that part of the difference between FLEXPART and FLEXPART-WRF simulations in the wet period of the year is due to the differences in the schemes used to calculate the scavenging. Pisso et al. (2019) have discussed these and other updates in the recent version of FLEXPART in more detail. Regarding dry deposition, for all particles below two times the reference height ( $h_{ref}$ ), the particle mass reduction is calculated by using Eq. 6:

$$\Delta m(t) = m(t) \left[ 1 - \exp\left(\frac{-v_d(h_{ref})\Delta t}{2h_{ref}}\right) \right] \quad (6)$$

$v_d$  is the dry deposition velocity that is calculated as the ratio of  $v_d(z) = \frac{F_c}{C(z)}$  for the  $h_{ref}$  of 15 m.  $F_c$  and  $C$  are the flux and the concentration of a species at height  $z$  within the constant flux layer. If the necessary information for the parameterization of dry deposition of gases and particles is not available,  $v_d$  can be assumed constant. The concentration of a given gas or aerosol species in a grid cell is equal to the weighted average of total particle mass within the grid cell divided by the grid cell volume (Pisso et al., 2019) as defined in Eq. 7.

$$C(z) = \frac{1}{V} \sum_i^N (m_i f_i) \quad (7)$$

with  $V$  being the grid cell volume,  $m_i$  particle mass,  $N$  the total number of particles, and  $f_i$  the fraction (weight) of the mass of particle  $i$  attributed to the respective grid cell. The amount of dry and wet deposition for the grid cell  $i$  is by default accumulated over the model time dimension unless the age composition of air parcels is required as was the case in this study. A unique feature of FLEXPART/FLEXPART-WRF is the grouping of simulations based on the age of Lagrangian particles. This means that the number of concentrations and deposition in each time step is obtained from the aggregation of the simulations through the added dimension of particle ages. In this study, we used the age spectrum of Lagrangian particles to estimate the transport of radioactive materials from source to receptors. We use the terms transport and Lagrangian particle/air parcel age interchangeably in the rest of the paper. The age of air parcels was examined in an hourly resolution. As a result, a new dimension with a size of 96 (hours) is added to the history output grids that have a horizontal resolution of 10 km in 14 vertical levels from 5 to 5000 m agl. While modeled concentrations (subsection 3.3) are vertically integrated through all model levels in analyses related to air parcel ages (subsection 3.1) and inter-member evaluation, modeled concentrations are averaged between 5 and 100 m agl (model levels 1 to 4) for the study of extreme cases (subsection 3.2).



## 2.2 Meteorological data

155 FLEXPART/FLEXPART-WRF runs in offline mode with grid-scale meteorological fields from numerical weather prediction  
models. In this study, we obtained meteorological inputs from NCEP climate forecast system version 2 (CFSv2) (Saha et al.,  
2011, Saha et al., 2014), the NCEP final analysis (FNL) (NCEP, 2015), and the ECMWF reanalysis 5th generation (ERA5)  
(Hersbach et al., 2020). CFSv2, which entered into the operational mode in March 2011, is a fully coupled climate model  
representing the interactions between the Earth's oceans, land, and atmosphere. CFSv2 can be used to provide 6-hourly forecast  
160 inputs for FLEXPART at the spatial resolution of 0.5 degrees from 2011 onward. FNL provides the three-hourly combination  
of analysis (at 00, 06, 12, and 18) and forecast meteorological fields (at 03, 09, 15, and 21) using the Global Data Assimilation  
System (GDAS). It is available at a spatial resolution of 0.25 degrees starting from 2015. ERA5, the successor to ERA-  
Interim, is the latest generation of ECMWF reanalysis data that covers from January 1, 1950, to nearly the present. They are  
produced at a spatial resolution of about 31 km at hourly time steps. CFSv2 and FNL are used as the inputs of FLEXPART.  
165 ERA5 and FNL analyses were first dynamically downscaled by WRF to create two members ERA5- and FNL-WRF.  
Subsequently, they are fed into FLEXPART-WRF with the new spatial and temporal resolution of 10 km and hourly. To the  
best of our knowledge, this is the first time that ERA5 inputs are used for the implementation of FLEXPART-WRF.  
An overview of the input data and corresponding simulation codes is presented in Table 1. A single simulation code is built  
for each meteorological dataset to be ingested by FLEXPART whereas the FLEXPART-WRF takes inputs via one common  
170 simulation code. Therefore, there are three independent simulation codes (in the second column in Table 1) for the four  
meteorological datasets used in this study.

**Table 1** The summary of meteorological inputs used to run FLEXPART and FLEXPART-WRF. The abbreviated names of meteorological datasets are affixed to their corresponding simulation codes in the second column.

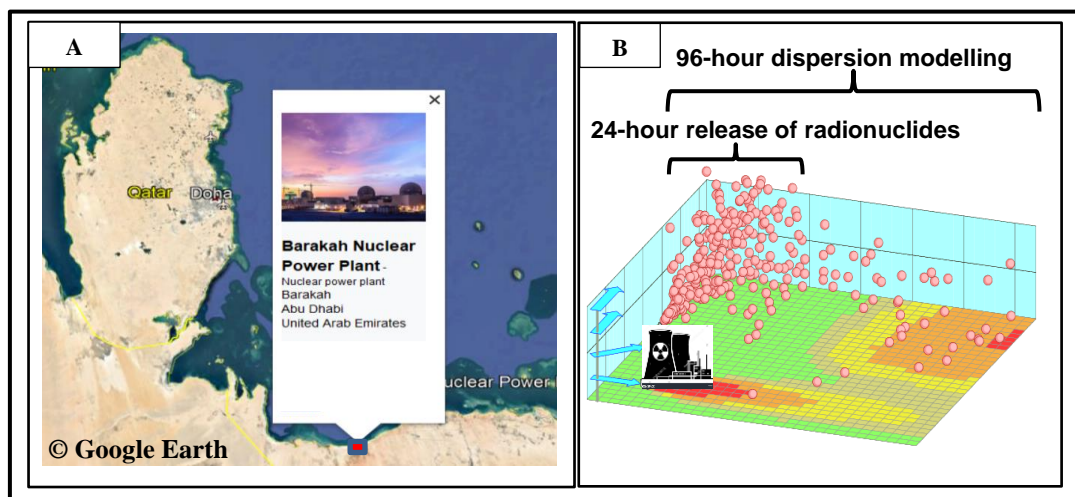
Inputs	Simulation code	Spatial resolution	temporal resolution	time coverage	downscaled	Type of input
CFSv2	FLEXPART_CFSv2	0.5 degrees	6-hourly	2011-present	X	forecast
FNL	FLEXPART_FNL/FLEXPART-WRF	10 km	hourly	2015-present	X/✓	analysis
ERA5	FLEXPART-WRF	10 km	hourly	1950-present	✓	reanalysis

## 2.3 Emission scenario and study area

175 We have simulated a nuclear accident at the B-NPP (Fig. 1-A) during which 22 PBq (6.9 kg) of  $^{137}\text{Cs}$  and 192 PBq (0.042 kg) of  $^{131}\text{I}$  are released during the first 24 hours. These amounts are the upper bound of emissions that Babukhina et al. (2016) estimated for the Fukushima Daiichi nuclear accident in May 2011. A total number of  $10^5$  Lagrangian particles were released



almost uniformly during the emission period (~417 particles per hour). To choose the optimal number of particles, we have followed the literature (Papagiannopoulos et al., 2020, Thompson et al., 2015, Fast and Easter, 2006). The computational time scales linearly increase with the increasing number of particles, while the statistical error of simulations decreases with the square root of the particle density (Pisso et al., 2019). The particles are initially distributed at height levels between 100 and 300 m above the ground level over the emission point. Each simulation starts at the beginning of each day and lasts 96 hours i.e., 4 days (Fig. 1-B). This experiment has been performed for each day of 2019 leading to a total of 1460 simulation days (365 days of the year x 4 forward simulation days). For the diurnal and seasonal stratification of simulations, we always refer to the time when a particle is released (not to when it travels or reaches receptors). The simulation domain is bounded between 17°N-33°N and 40°E-60°E, and for post-processing, the geographical box over Qatar, from 24.25°N to 26.35°N and from 50.65°E to 51.75°E, has been analysed.



190 **Figure 1** A is the study area embracing the B-NPP (red square) and the state of Qatar. The base map and overlaying information are taken from Google Earth. B is the schematic illustration of the LPDM simulation cycle. The original figure is available at <https://www.janicke.de/en/lasat.html> (last access: 30 May 2022).

### 3 FLEXPART/FLEXPART-WRF simulations:

It should be noted that due to high solubility and the relatively long half-life of  $^{137}\text{Cs}$ , further emphasis is placed on the spatio-temporal distribution of  $^{137}\text{Cs}$  deposition. Conversely, the ambient surface-level concentration of  $^{131}\text{I}$  over the study area will be considered since it resides mostly in the gaseous phase and has a short half-life of 8 days.

#### 3.1 Diurnal and seasonal analysis of radionuclide transport

In order to investigate the diurnal and seasonal variations in the transport of radionuclides, we examined the simulated age of the Lagrangian particles entering the study area by the season and time of day in which they were released from B-NPP. The frequency density plots of air parcel ages for  $^{131}\text{I}$  and  $^{137}\text{Cs}$  are shown in Figures 2 and S1, respectively. The significant



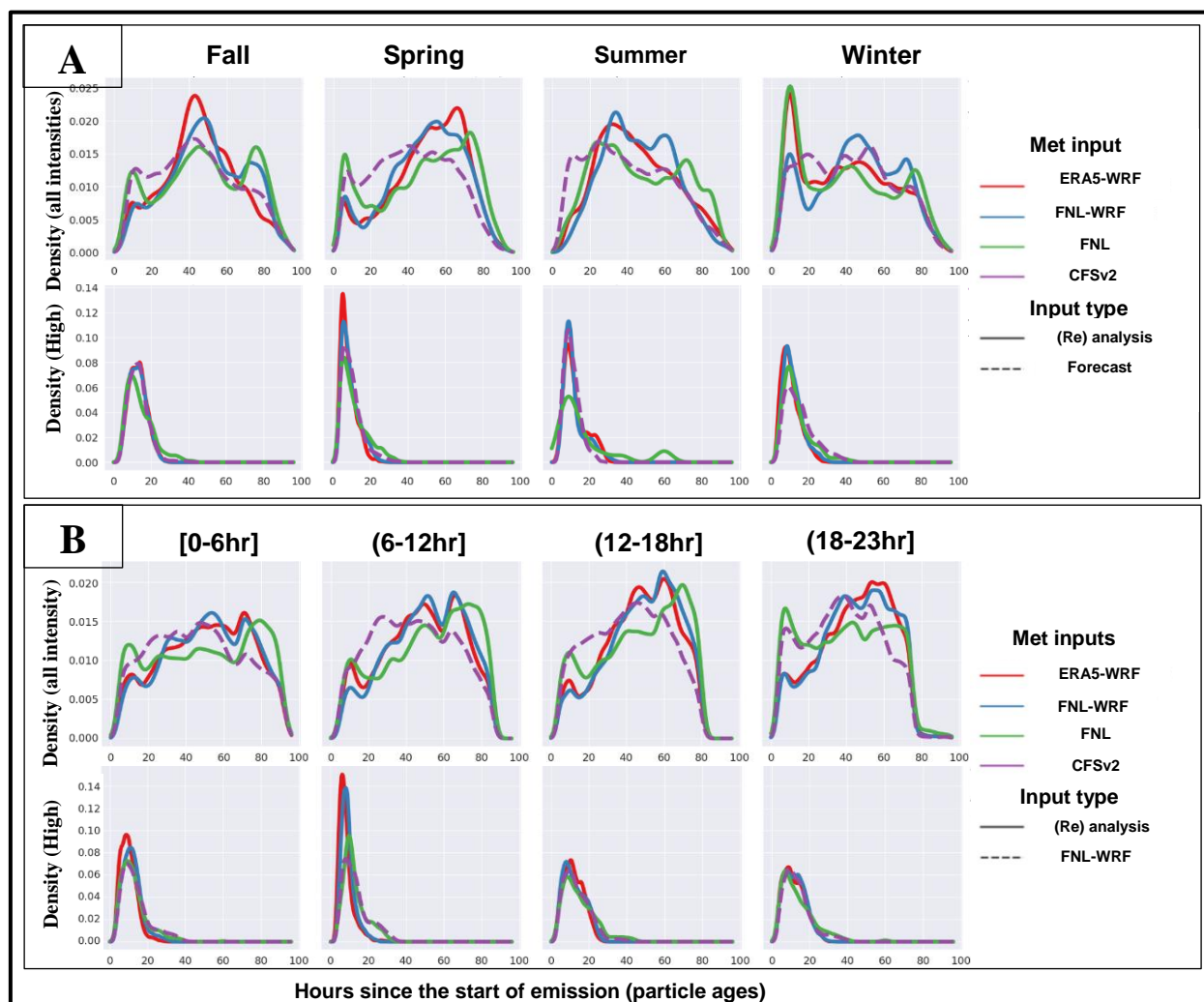
200 similarity of age distributions of  $^{131}\text{I}$  and  $^{137}\text{Cs}$  indicates that differences in the transport characteristics of these radionuclides  
such as the wet and dry deposition rate and radioactive decay could not be so great as to cause the abundance of cases where  
 $^{131}\text{I}$  and  $^{137}\text{Cs}$  particles are not present in a common grid. The close dispersion of  $^{131}\text{I}$  and  $^{137}\text{Cs}$  concentrations (at least base  
concentrations) and, consequently, the age of corresponding air parcels can be attributed to the lack of heavy precipitation in  
the region. This can be seen in S2 where a positively skewed distribution is found for the ratio of  $^{137}\text{Cs}$  wet deposition to total  
205  $^{137}\text{Cs}$  deposition. Therefore, the following discussion related to the ages of particles simulating  $^{131}\text{I}$  concentrations is also valid  
for  $^{137}\text{Cs}$  concentrations (shown in the supplement). All ensemble members in all seasons simulated an abrupt increase (or a  
peak) in the number of air parcels with lifetimes of less than 20 hours (top row in Fig. 2-A). This indicates that, regardless of  
the time of year, radionuclide clouds started crossing the borders of the study area within a few hours after the emission. The  
relatively lower spatial resolution of CFSv2 caused a smooth distribution of its simulated air parcel ages that is close to the  
210 average of other distributions (less variation in radionuclide transport). The age distribution produced by FNL-WRF was found  
to be more similar to the one produced by ERA5-WRF than by FNL. This seems to be due to the use of meteorological inputs  
with the same spatio-temporal resolutions and a common simulation code and, consequently, similar modeling schemes for  
the two former members. Although the base model used for the production of FNL, the Global Forecast System (GFS), is also  
the atmospheric component of CFSv2, FNL age distributions look closer to those from ERA5- and FNL-WRF. In addition to  
215 having finer spatial resolution than CFSv2, FNL assimilates observations like ERA5. Compared to other members, air parcel  
ages are distributed in a wider range in all seasons in FNL (note the location of the first and last peaks). This could be because:  
FNL simulated 1- a decrease in transport speed to the same receptors and/or 2- a higher frequency of air parcels reaching  
further receptors from the source, and/or 3- an increase in the residence time of air parcels over the same receptors. While the  
two former cases are examined in the next section, the study of air parcel residence time requires the calculation of source–  
220 receptor relationship (Seibert and Frank, 2004). The residence time of Lagrangian particles is outputted when FLEXPART and  
FLEXPART-WRF are executed in the backward mode (Pisso et al., 2019), which is beyond the scope of this study.  
To understand how the transport may change with the levels of radionuclide concentrations, beside the previous categorization,  
air parcel ages were separated into three levels representing low (the top row in S3-A), moderate (the bottom row in S3-A),  
and high (the bottom row in Fig. 2-A) concentrations of  $^{131}\text{I}$  (see S1 and S4 for  $^{137}\text{Cs}$ ). These categories, for each member, are  
225 determined based on column (mass) densities, computed from the vertical integration of concentrations. The low, moderate,  
and high values are respectively lower than the 33<sup>rd</sup> percentile, between the 33<sup>rd</sup> and 66<sup>th</sup> percentiles, and greater than the 66<sup>th</sup>  
percentile. The strong positively skewed age distributions of particles that carried moderate (in the supplement) and high  
concentrations of both radionuclides in all members and all seasons indicate that they reach the study area boundaries within  
a few hours (less than 20 hours) after the accident. These results are to be expected as the southeast of the study area is very  
230 close to the source point (B-NPP). It is also worth mentioning that all members of the ensemble model (including the forecast  
member) have a greater agreement in air parcel ages for the moderate and high levels of  $^{131}\text{I}$  and  $^{137}\text{Cs}$  than those simulated for  
low concentrations. This is in line with the principles of FLEXPART dispersion modeling. According to Pisso et al. (2019),





the error rate of simulations decreases with the square root of the particle density. As a result, all four members simulated the age of Lagrangian particles corresponding to moderate and high concentrations more closely and, most likely, more accurately  
235 near the source. For low intensities of both radionuclides, however, the age distributions are almost the same as that seen in all intensities.

The peak of newly arriving air parcels in all intensities occurs earlier in spring than in other seasons. There is also a delayed peak of long-lived air parcels in low and all intensity column densities in the spring. Therefore, it can be concluded that regional atmospheric circulations led to the transport of radionuclides with different intensities to the further parts of the study area in  
240 this season than in other seasons. Figure 2-B (S1-B) shows the categorization of age distributions based on the release time of particles carrying  $^{131}\text{I}$  ( $^{137}\text{Cs}$ ) in the first 24 hours of each simulation period. This new categorization reveals that the number of long-lived air parcels inducing concentrations increased in all members when particles are released between 12 and 6 p. m. (top row in Fig. 2-B and S1-B). This temporal pattern may be due to the coincidence of particle release and the development of the planetary boundary layer in the afternoon that caused the transport of radionuclides to longer distances (represented by  
245 longer Lagrangian particle ages). Compared to ERA5-WRF and FNL-WRF, CFSv2 and FNL have simulated a larger number of shorter-aged air parcels after sunset (between 6 and 11 p.m.). It is because the spatio-temporal resolution of inputs used in these two members is too coarse to resolve the gradual reduction of boundary layer height resulting in fewer particles being transported to remote areas. Like seasonal distribution, the diurnal variations of air parcel ages for high concentrations are very similar in all members (bottom row in Fig. 2-B and S1-B). They all simulated a higher number of shorter-lived parcels when  
250 they are released in the first half of the day. Although a firm conclusion would require further studies at very-high-resolution of land-sea circulations in the region, onshore winds passing the emission point from the south may cause the abundance of shorter-lived parcels in southern Qatar.



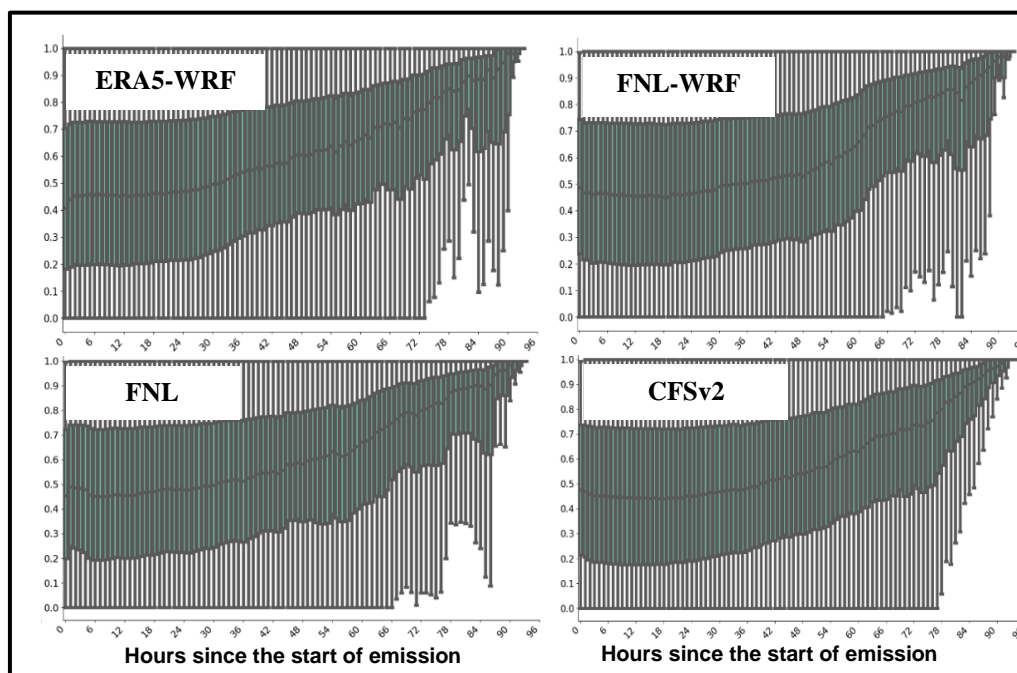
255 **Figure 2 A:** the smooth density estimates of air parcel ages corresponding to all intensities of  $^{131}\text{I}$  column densities (top row) and of those above the 66<sup>th</sup> percentile (bottom row). **B:** the same as A, but for four times of the day. The curves represent simulations from each member of the ensemble. y- and x-axes show density values and air parcel ages, respectively.

Due to the relatively long half-life of  $^{137}\text{Cs}$ , its deposition rate in the affected areas is of great importance. Figure 3 shows the relationship between the age compositions of air parcels and the amount of  $^{137}\text{Cs}$  deposition normalized to the maximum of each simulation period at each grid cell in winter when both dry and wet deposition occur in the study area. We found very similar results in other seasons (Fig. S5). As expected, the values of the  $^{137}\text{Cs}$  deposition increase cumulatively with the time after the accident. All members simulated a similar pattern (median) in which around eighty percent of the deposition (the median of normalized deposition > 0.8) happens within 80 hours after the assumed accident. However, there is considerable variability in the spatial pattern of deposition. Given that the figure shows the amount of deposition across the whole study area, the decrease in the levels of accumulated deposition at the end of the simulation period pertains to the areas that are far

260



265 from the source. However, in all parts of the region, the deposition is close to 100% approaching the end of the 96-hour simulation period. This can also be seen in Figure S6 where most of the particles inducing all intensities of  $^{137}\text{Cs}$  deposition peak around 20 hours and a quite small number of air parcels are aged above 80 hours.



270 **Figure 3** the deposition of  $^{137}\text{Cs}$  normalized to the maximum amount of each simulation period at each grid cell in winter. the x-axis shows the age of Lagrangian particles and the y-axis is the normalized deposition. Error bars show the range of normalized deposition within the study area. To prepare this plot, the deposition values for each time step are accumulated (through age spectra and time dimensions) and, then, are normalized to the maximum value of each four-day run at each grid cell.

### 3.2 Spatio-temporal distribution of radionuclides

275 In this section, we examine the seasonal and diurnal changes in the concentrations and deposition of radionuclides in the study area. To perform analysis related to radionuclide concentrations, the average of the simulations in the lowest four layers of the model between 5 to 100 meters has been used. Due to the high radioactivity of  $^{131}\text{I}$  and the solubility and relatively long half-life of  $^{137}\text{Cs}$ , in this part, we focus on  $^{131}\text{I}$  concentrations and  $^{137}\text{Cs}$  deposition. Figures 4 and 5 show the seasonal median of the maximum  $^{131}\text{I}$  concentrations ( $^{131}\text{I}_{\text{conc\_seas\_max}}$ ) and of the maximum (total)  $^{137}\text{Cs}$  deposition ( $^{137}\text{Cs}_{\text{depos\_seas\_max}}$ ) within the study area. The maximum values are calculated through each four-day simulation period and, then, medians are found within  
280 the respective season. To identify the highest possible level of pollution at each point, regardless of its frequency, local maxima are calculated only from non-zero intensities. Using conversion factors from Spiegelberg-Planer (2013),  $^{131}\text{I}_{\text{conc\_seas\_max}}$  (in a unit of  $\text{Bq m}^{-3}$ ) are converted to the maximum hourly doses from inhalation (in a unit of  $\mu\text{Sv}$ ). Results show that CFSv2 and FNL simulated the lowest and highest levels of inhalation doses in all seasons than the other members (note varying color



scales between members in both Figures 4 and 5). While CFSv2 simulations are consistently less than 200  $\mu\text{Sv}$ , FNL and two  
285 dynamically downscaled members simulated doses up to more than 1000 and 600  $\mu\text{Sv}$ . In other words, the forecast member  
of the ensemble has underpredicted  $^{131}\text{I}$  doses by a factor of about 3 to 5 compared to (re)analysis members in severely polluted  
areas. Considering that FNL and CFSv2 are executed using the same simulation code, the remarkable difference between the  
resulting  $^{131}\text{I}_{\text{conc\_seas\_max}}$  can be attributed to the difference in meteorological inputs. Figure S7 shows the seasonal median of  
planetary boundary layer height (PBLH) estimated based on CFSv2 (top row) and FNL (bottom row) meteorological inputs.  
290 CFSv2-based PBLH is found to be 3 to 5 times higher over emission point in UAE and, with less intensity, within the study  
area than FNL-based PBLH. The obvious consequence of elevated PBLH is the dilution of pollutants near the surface. In  
addition, the poorer spatial resolution of CFSv2 inputs caused the faulty separation of land and sea boundary layer process  
along the coastlines of Qatar. This also may lead to the suboptimal modeling of particle dispersion across the study area,  
especially along the coastlines. The lower (higher) PBLH in the cold (warm) period of the year coincides with escalated  
295 (downgraded)  $^{131}\text{I}_{\text{conc\_seas\_max}}$  by a factor of up to 3 to 5 in the cold period (warm period) in the south/southeast of Qatar. In  
other words, southern Qatar is the first area to be affected by dense  $^{131}\text{I}$  clouds in the event of a nuclear accident, especially in  
the cold period of the year. FNL simulated a dense  $^{131}\text{I}$  cloud in the center of Qatar in the summer that is due to an exceptional  
transport of radionuclides to Qatar in the summer of 2019, discussed later. This unusual pattern in summer, and even prevalent  
spatial patterns in other seasons, may change with the increasing length of the study period. Inhalation doses simulated by  
300 three reanalysis members are relatively high compared to equivalent doses modeled by Christoudias and Lelieveld (2013) for  
areas adjacent to the Fukushima accident. They reported the cumulative inhalation of 100 and 500  $\mu\text{Sv}$  doses from a total of  
 $^{131}\text{I}$ ,  $^{134}\text{Cs}$ , and  $^{137}\text{Cs}$  concentrations in around three months (between March 11 and May 31, 2011). Compared to the outputs  
of FNL and FNL-WRF, the significant similarity of FNL-WRF and ERA5-WRF simulations suggests the downscaling of  
inputs and the application of the same (different) simulation codes can profoundly affect FLEXPART modelling. For example,  
305 the simulations of FNL in southern Qatar in autumn are more than twice that of WRF-FNL. The examination of the particle  
release time (shown by contours) shows that  $^{131}\text{I}_{\text{conc\_seas\_max}}$  in most parts of the study area is caused by particles released  
between 5 a.m. and 12 p.m. As mentioned before, it seems that the intensification of the thermal gradient between the land and  
sea at this time of day and the strengthened onshore winds increase the transfer of freshly released air parcels carrying  
concentrated  $^{131}\text{I}$  clouds inland to the south of the study area. Figure 5 shows  $^{137}\text{Cs}_{\text{depos\_seas\_max}}$  simulated by four ensemble  
310 members. According to the definition by the International Atomic Energy Agency (IAEA, 2009) any area covered with  
radioactive substances which emit beta particles and gamma rays, like  $^{131}\text{I}$  and  $^{137}\text{C}$ , in quantities above 40  $\text{kBqm}^{-2}$  is  
considered to be contaminated. Accordingly, all members have simulated high levels of contamination for a significant portion  
of the study area in the cold period of the year. For example, FNL simulations of  $^{137}\text{Cs}_{\text{depos\_seas\_max}}$  in winter and autumn are  
above 40  $\text{kBqm}^{-2}$  in entire southern Qatar. Contrary to  $^{131}\text{I}_{\text{conc\_seas\_max}}$ , CFSv2 yielded comparable and even higher  
315  $^{137}\text{Cs}_{\text{depos\_seas\_max}}$  to other members in winter when the entire Qatar is subject to  $^{137}\text{Cs}_{\text{depos\_seas\_max}}$  of more than 100  $\text{kBqm}^{-2}$ .  
However, in other seasons, CFSv2 simulations fall below 40  $\text{kBqm}^{-2}$  in most areas. In the simulations of ERA5- and FNL-



WRF,  $^{137}\text{Cs}_{\text{depos\_seas\_max}}$  levels above  $40 \text{ kBq m}^{-2}$  cover almost the entire Qatar in winter whereas they are confined to the southern half in autumn. The  $^{137}\text{Cs}_{\text{depos\_seas\_max}}$  magnitudes in the warm period of the year (except in the simulations of FNL in summer) are mostly either close to or below the threshold. The higher  $^{137}\text{Cs}_{\text{depos\_seas\_max}}$  in the cold period of the year, especially  
320 in winter, can be largely attributed to the seasonal increase in the transport of  $^{137}\text{Cs}$  by northward winds, discussed later, and to the increased contribution of wet deposition. According to Figure S2, the wet deposition of  $^{137}\text{Cs}$  in CFSv2 simulations is greater than that of other members, which can well explain the higher CFSv2  $^{137}\text{Cs}_{\text{depos\_seas\_max}}$  in winter. The above results indicate that if a nuclear accident happens in the cold period of the year, especially in winter, the magnitudes of extreme  $^{131}\text{I}$  concentrations and total  $^{137}\text{Cs}$  deposition in the south of Qatar can be up to 3 times stronger than in the warm seasons. In Figure  
325 5,  $^{137}\text{Cs}_{\text{depos\_seas\_max}}$  mainly coincides with the contours of particle release time between 10 a.m. and 2 p.m. when the boundary layer is expanding. The analysis of age spectra in subsection 3.1 disclosed the greater frequency of long-lived air parcels in FNL simulations. To further examine these results, the spatial distribution of air parcel ages overlying the full-year median of the maximum  $^{131}\text{I}$  concentrations and of the maximum (total)  $^{137}\text{Cs}$  deposition is shown in Figures 6-A and -B, respectively. As expected, the age of Lagrangian particles is decreasing southward with the decreasing distance from the source. In Figure  
330 6-A, we see that all ensemble members simulated the age contours of less than 20 hours along with high levels of  $^{131}\text{I}$  at the south-eastern corner of the study area. The longer-lived particles (above 37 to 40 hours) were found in the north of Qatar. The observation of longer-lived air parcels and expanded dense  $^{131}\text{I}$  clouds in the center of Qatar in FNL simulations indicate slower air parcel transport and the greater expansion of radionuclides to areas further from the source. Therefore, these two transport characteristics seem to cause the abundance of long-lived particles in FNL simulations compared to other members. The  
335 residence time of Lagrangian particles, as the third factor, is not investigated in the present study because it is not the standard output of forward FLEXPART runs. According to Figure 6-B, FNL simulated the fastest end to the total  $^{137}\text{Cs}$  deposition in southern Qatar (about 40 hours after the accident) whereas air parcel ages corresponding to the deposition completion in the north of Qatar are above about 60 hours in all members. In other words, FNL produced the biggest difference (around 20 hours) between the particle ages simultaneous with the completion of deposition in the north and south of Qatar compared to  
340 that of other members (around 10 hours).

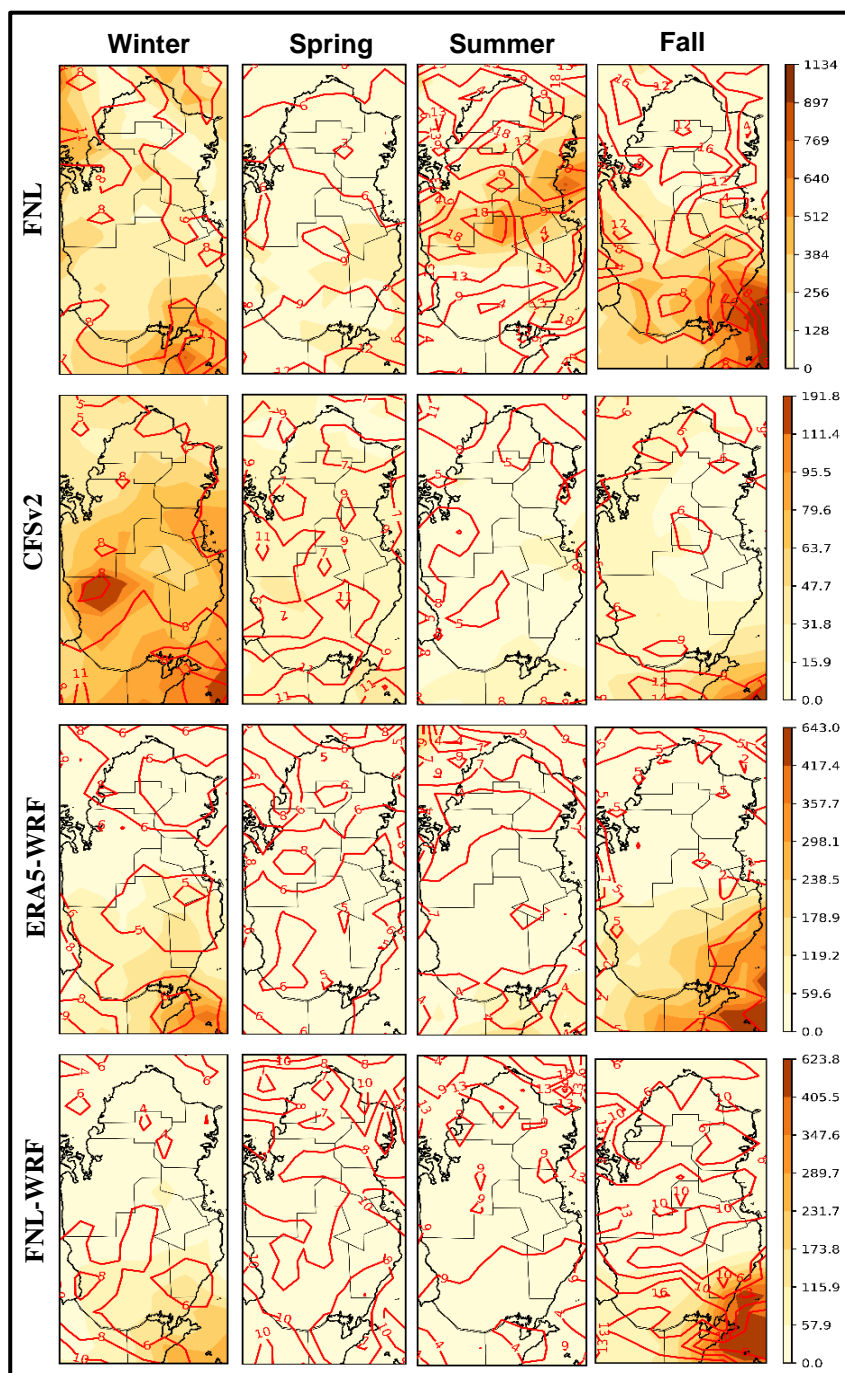
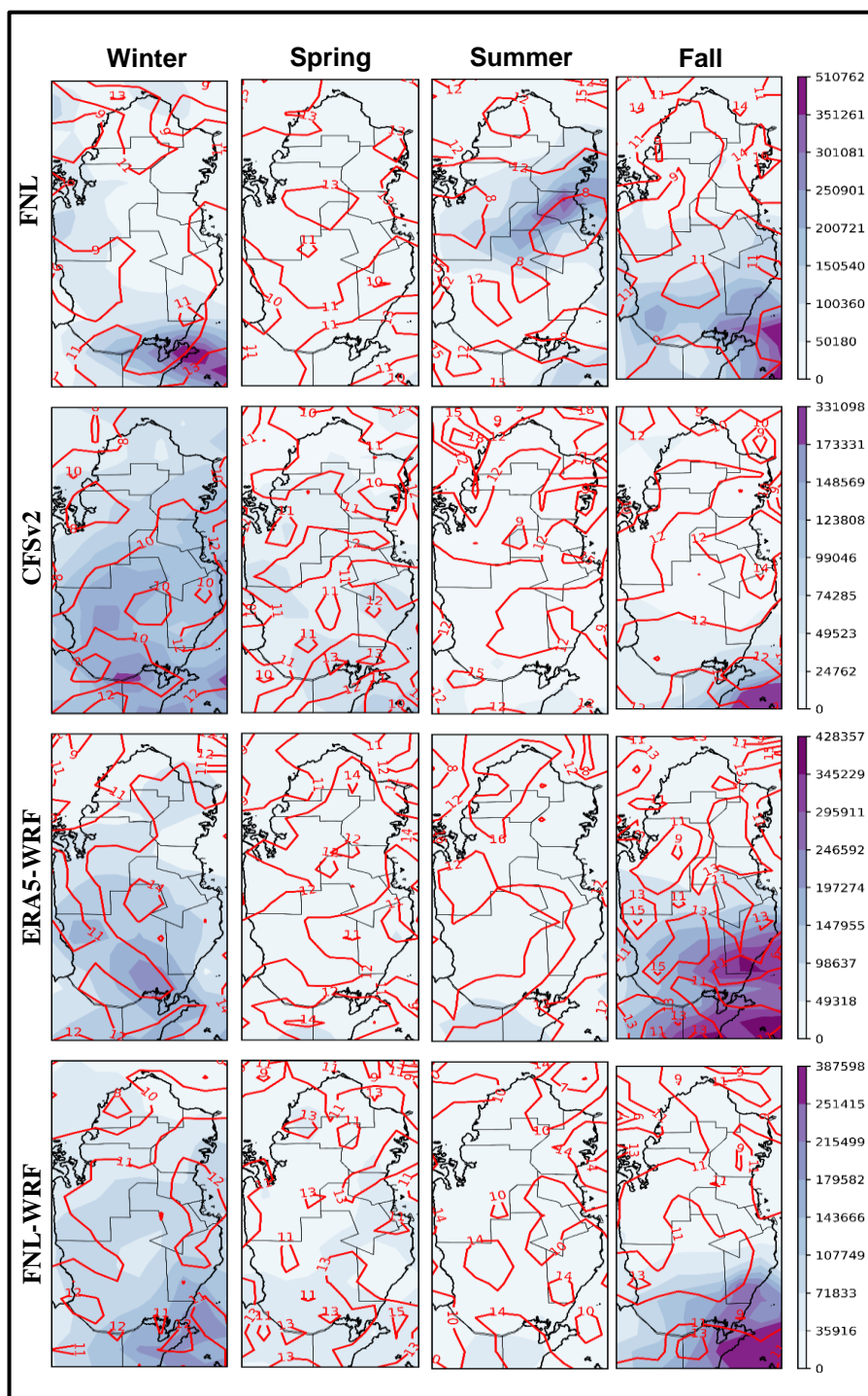


Figure 4 Color scale shows the seasonal median of maximum  $^{131}\text{I}$  concentrations ( $^{131}\text{I}^{\text{conc\_seas\_max}}$ ), converted to maximum hourly doses from inhalations ( $\mu\text{Sv}$ ). Contour lines are the seasonal median of release time of Lagrangian particles simultaneous with maximum  $^{131}\text{I}$  concentrations.

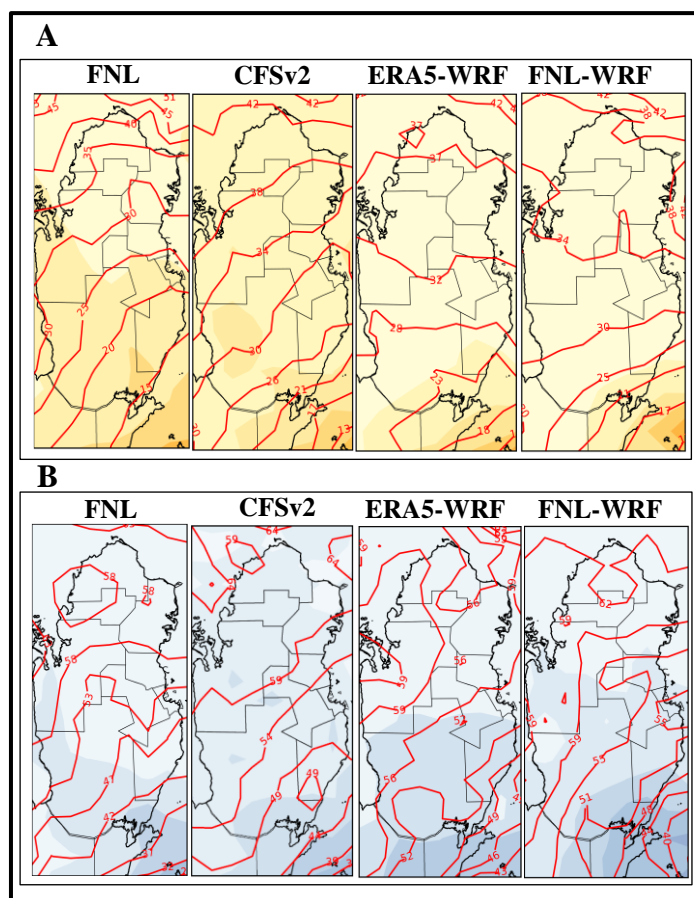


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Figure 5 Same as 4, but for the seasonal median of  $^{137}\text{Cs}$  deposition ( $^{137}\text{Cs}_{\text{depos\_seas\_max}}$ ) in the unit of  $\text{Bq m}^{-2}$ .



Because of the high importance of the population being exposed to air pollution radionuclides,  $^{131}\text{I}^{\text{conc\_seas\_max}}$ , converted to the maximum hourly doses from inhalation, and  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$  are analyzed against the population density of Qatar (Fig. 7). The spatial distribution of the population shows that the desert areas of southern and southeastern Qatar, as the domain entry of radionuclides in the region, host a small number of people or are nearly uninhabited (Fig. 7-A). Figures 7- B and C show that the extremely high levels of inhalation doses (higher than  $200\ \mu\text{Sv}$ ) and  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$  (higher than  $100\ \text{kBqm}^{-2}$ ) occurred mostly in areas with a population density of fewer than five people per arc-second. The populated areas (with a density of more than 15 people per arc-second) have coincided with inhalation doses and  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$  less than above thresholds. The highest inhalation doses are often found in FNL simulations, in autumn and winter, whereas the lowest doses are simulated by CFSv2. Compared to  $^{131}\text{I}^{\text{conc\_seas\_max}}$  inhalation doses, there are no pronounced seasonal and inter-member differences in  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$  at different population levels.

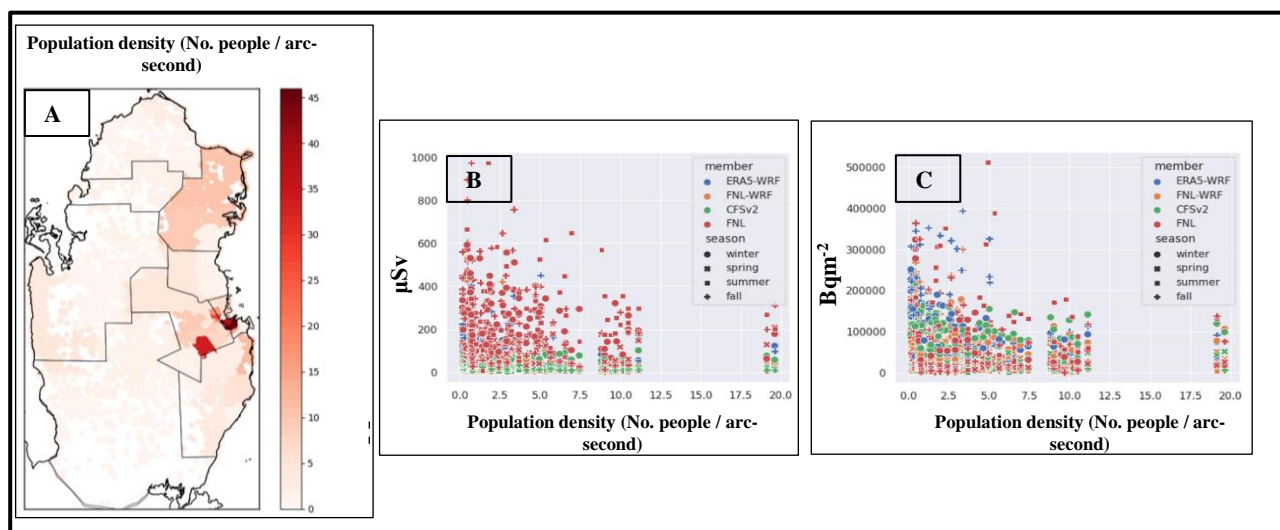


360 **Figure 6 A: the full-year median of  $^{131}\text{I}^{\text{conc\_seas\_max}}$ , converted to maximum hourly doses from inhalations ( $\mu\text{Sv}$ ), and B: the full-year median of  $^{137}\text{Cs}$  deposition ( $^{137}\text{Cs}^{\text{depos\_seas\_max}}$ ) in the unit of  $\text{Bqm}^{-2}$ . Contour lines are the full-year median of air parcel ages simultaneous with maximum  $^{131}\text{I}$  concentrations and  $^{137}\text{Cs}$  deposition. The colorbar of each subplot, representing each member, in the top (bottom) row is the same as in Fig. 4 (Fig. 5), omitted here for space.**





365



**Figure 7** the subplot A shows the gridded number of the population of Qatar at a resolution of one arc-second in 2020. The relationship between  $^{131}\text{I}^{\text{conc\_seas\_max}}$ , converted to maximum hourly doses from inhalations ( $\mu\text{Sv}$ ), and  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$  and Qatar's population densities are shown in B and C, respectively. Markers' shapes and colors represent seasons and ensemble members.

370 In addition to the magnitude of the radioactive materials transported, the temporal distribution of extreme events in the affected area is of importance in preparedness programs. Figure 8-A shows the frequency of occurrences (FoO) of  $^{131}\text{I}$  column densities above the 66<sup>th</sup> percentile. In all members, more than half of the events in the east, with a high population density, and north of Qatar occurred in winter. Other parts of Qatar also see a large FoO (greater than 30% in FNL-WRF to 50% in FNL) in winter. While the springtime FoO of high  $^{131}\text{I}$  varies between 15 and 30%, the lowest FoO is observed in the summer when less than

375 15% of extreme events have reached receptors in Qatar. The seasonal FoO of  $^{137}\text{Cs}$  column densities (S8) also shows almost identical results. The panel at the top in Figure 8-B shows a case in which ERA5-WRF simulated the north-westward movement of  $^{131}\text{I}$  concentrations ( $\text{Bq}/\text{m}^3$ ) on January 14 at noon. We observe a similar pattern in a large number of events during which high levels of radionuclides are transported to Qatar. This synoptic pattern seems to be due to the juxtaposition of low and high-pressure cells located in the west and east of the region. The resulting pressure gradient has caused the

380 formation of strong south/southeast winds between two cyclonic and anti-cyclonic cells that move the dense  $^{131}\text{I}$  clouds to the study area. This pattern occurs mainly in the late winter-early spring period simultaneous with the southward movement of westerlies and the eastward movement of the Saudi Arabian subtropical high pressure (De Vries et al., 2016). The panel at the bottom in Figure 8-B shows the summertime average of ERA5-WRF  $^{131}\text{I}$  concentrations ( $\text{Bq}/\text{m}^3$ ) and of near-surface atmospheric circulation. The seasonal pattern found here illustrates well the main reason for the scarcity of extreme events

385 observed in summer in the study area. In this pattern, northwest-southeast winds, known as Shamal winds (Yu et al., 2016),



caused simulated radionuclides to move away from the study area. Therefore, one may assume that the occurrence of a nuclear accident and the resulting release of radionuclides in the cold period of the year pose a higher contamination risk to the population in Qatar.

Concerning the sensitivity analysis of FLEXPART-WRF simulations to the turbulence scheme of choice, the use of STM  
390 overall reduces the column densities aggregated over each 96-hour simulation period (96-hour integrated column densities).  
As shown in Figure 9-A, the median of simulated column densities of  $^{131}\text{I}$  is more than doubled in GTM-based simulations in  
the fall. There is no comparable change in other seasons that is reflected in a minute increase in the full-year distribution of  
GTM-based simulations. In contrast, we see a tangible increase in extremely high levels of STM-based  $^{137}\text{Cs}$  deposition  
simulations in almost all seasons (Fig. 9-B). The upper quartile of  $^{137}\text{Cs}$  deposition with STM in winter is around 25% higher  
395 than those simulations with GTM in the same period. This is largely found across the year leading to an annual increase of  
around 20% in the upper quartiles of STM-based simulations of  $^{137}\text{Cs}$  deposition. The implementation of STM in FNL-WRF  
has had a similar impact on radionuclide simulations (shown in S9). The increase in high  $^{137}\text{Cs}$  deposition and the decline in  
the moderate levels of  $^{131}\text{I}$  column densities (mainly in fall) after the application of STM, in both sensitivity runs, can be  
400 interpreted as the increase of deposition in areas closer to the emission point and the equivalent decline in the transport of  
radionuclides to remote areas, respectively.

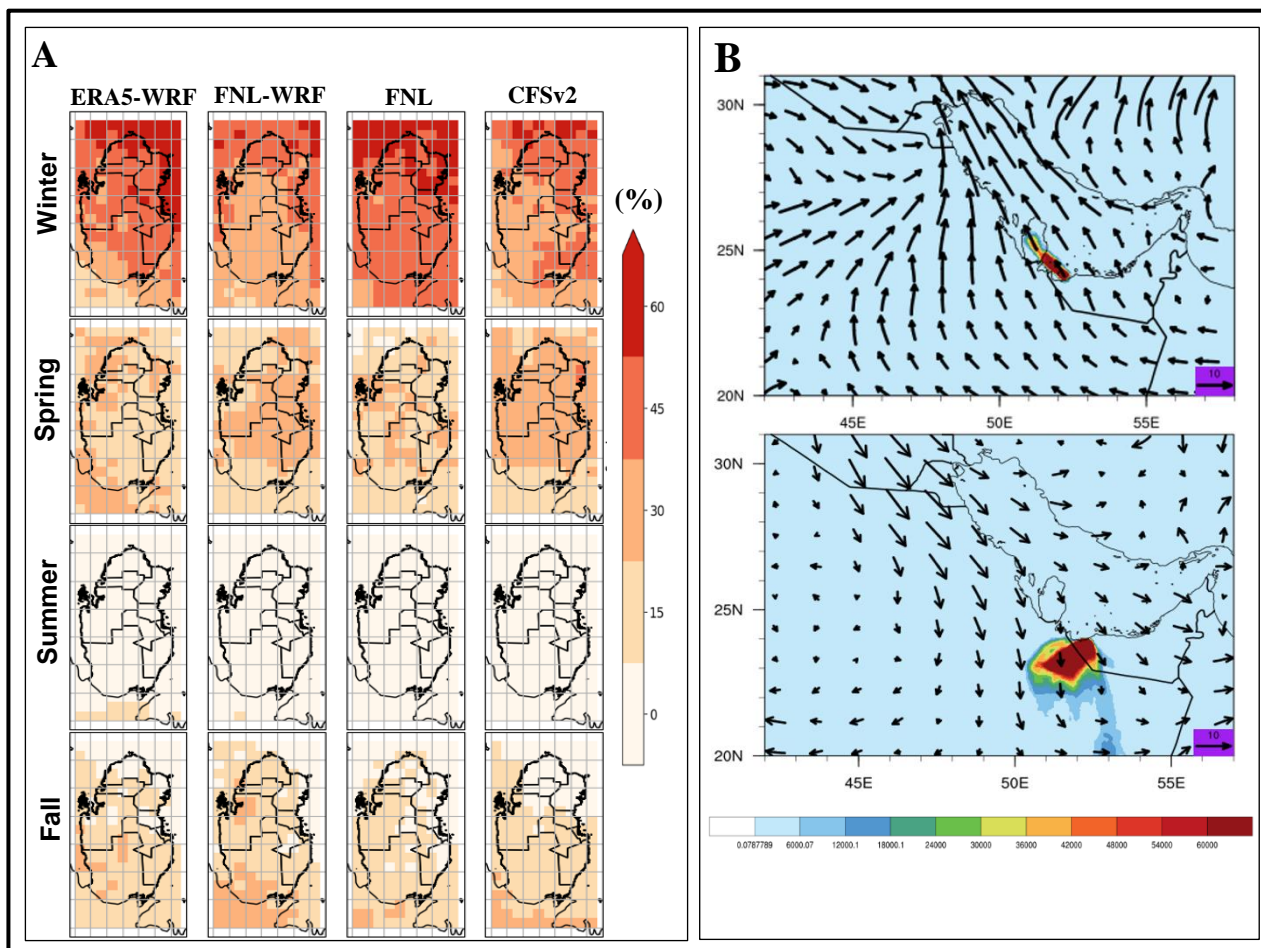
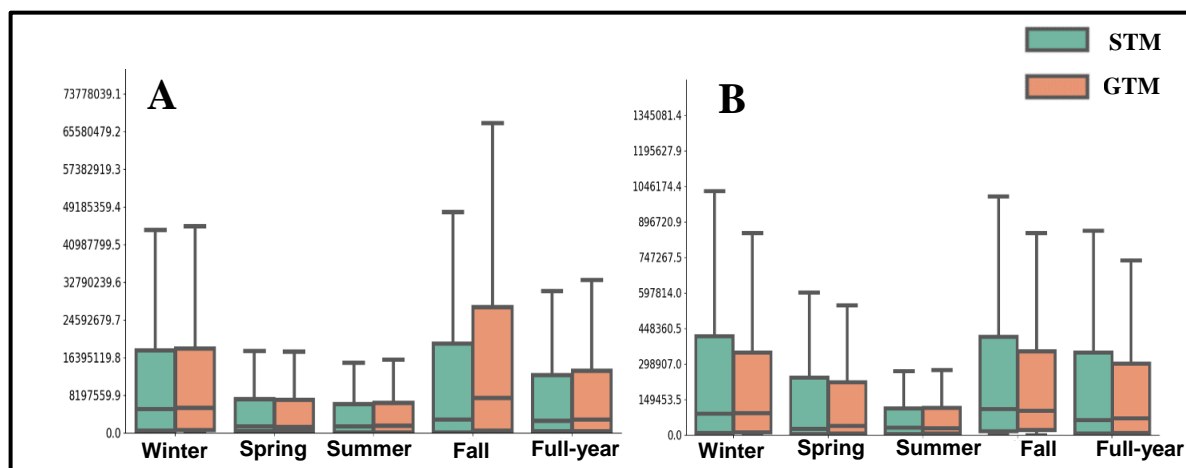


Figure 8 A: The frequency of occurrence (%) of  $^{131}\text{I}$  column densities above the respective 66<sup>th</sup> percentile. B: ERA5-WRF  $^{131}\text{I}$  column density ( $\text{Bq m}^{-3}$ ) on January 14, 2019, at noon (top) and the summertime average of  $^{131}\text{I}$  column density (bottom). Overlying vectors show wind streams at 150 m agl. To represent atmospheric circulations during the release of radionuclides, wind data at 12 p.m. on the first day of the four-day simulation period was used.

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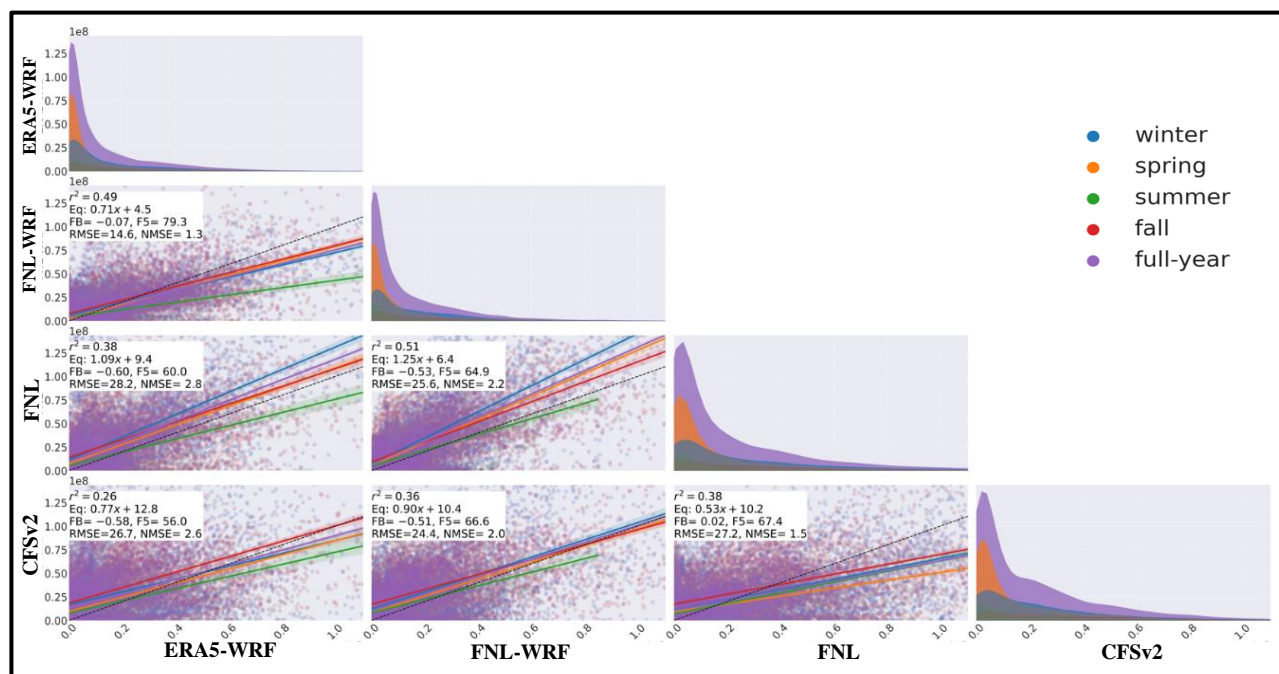
**Figure 9** Panels A and B show the 96-hour integrated simulations of column densities of  $^{131}\text{I}$  ( $\text{Bq}/\text{m}^2$ ) and total deposition of  $^{137}\text{Cs}$  ( $\text{Bq}/\text{m}^2$ ) from ERA5-WRF with STM (green, sensitivity run) and GTM (brown, control run), respectively. The same outputs for FNL-WRF are shown in S9.

### 410 3.3 Inter-comparison of FLEXPART and FLEXPART-WRF runs (effect of downscaling)

Aside from the fact that radionuclides released from a nuclear accident are rarely monitored (Evangelidou et al., 2017), since we are dealing with a hypothetical accident in this study, the simulations from the three (re)analysis members are assumed to be a better approximation to actual conditions than that of CFSv2 (Leadbetter et al., 2022). The evaluation statistical metrics used here are recommended by Maurer et al. (2018) who sought the best transport model for the simulation of Xe-133 dispersion using measurements from the six International Monitoring System (IMS) stations. The R-squared obtained between the 96-hour integrated simulations of  $^{131}\text{I}$  column densities from CFSv2 and other members (the bottom row in Figure 10) indicates that the CFSv2 simulations are well associated with those from (re)analysis members. The Pearson correlation coefficients between CFSv2 and (re)analysis simulations, equivalent to the square root of the coefficient of determination ( $r^2$ ), range between 0.5 and 0.6. Maurer et al. (2018) found that the involved models produced an average correlation coefficient of 0.51 between Xe-133 simulations and observations. The highest correlation of CFSv2 was found with FNL followed by FNL-WRF. The linear regression equations computed between simulations from CFSv2 and other members all have less than unity coefficients indicating that CFSv2 underpredicted  $^{131}\text{I}$  column densities, especially compared to FNL. The seasonal regression suggests that, mainly due to elevated PBLH (Fig. S7), the simulations of CFSv2 are significantly lower than that of FNL in spring and that of FLEXPART-WRF members in summer. The fractional bias (FB), in the range -2 and 2, is the bias of simulation means normalized by the sum of the two means and multiplied by 2. The fraction within a factor of 5 (F5) is the fraction of simulations that is at most a factor larger (5) or smaller (0.2) than the reference values. While CFSv2 simulations show (on average) trivial FB compared to FNL simulations, they are positively biased (-0.51 and -0.58) compared to the simulations from FNL- and ERA5-WRF (the negative sign of FB is due to the placement of the (re)analysis members in the



x-axis). In terms of FB5, 67.4% and 66.6% of CFSv2 simulations are respectively within the factor of 5 of FNL and FNL-  
430 WRF simulations while it is reduced to 56% in comparison against ERA5-WRF simulations. Taking into account the average  
FB of 0.33 and FB5 of 32% between simulations and observations reported by Maurer et al. (2018), it can be concluded that  
the association between CFSv2 and (re)analysis simulations falls within the expected range in radionuclide dispersion  
modeling. Unlike previous metrics, the RMSE between CFSv2 and FNL is higher than what was found between the former  
and other members. It is mainly because of the abundance of extremely high values in FNL simulations. CFSv2 simulations  
435 produced the lowest and highest NMSE, the normalized RMSE which is less sensitive to extreme values, against FNL (1.5)  
and ERA5-WRF (2.8) simulations, respectively. Among the (re)analysis members, the lowest correlation (0.61) is found  
between ERA5-WRF and FNL. Feeding downscaled FNL inputs into FLEXPART-WRF (FNL-WRF) increased the correlation  
of ERA5-WRF and FNL-WRF to 0.7. The highest correlation between all members was obtained between FNL and FNL-  
WRF (0.71) indicating that the downscaling of inputs in FNL-WRF did not have much effect on the association of their  
440 simulations. However, considering other metrics (FB=-0.07, FB5=79.3%, RMSE=75440, and NMSE=1.2), the closest  
simulations are produced by FNL-WRF and ERA5-WRF (rather than by FNL and FNL-WRF). This suggests that the  
downscaling of similar (FNL) and different (FNL and ERA-5) meteorological datasets increased and decreased the absolute  
differences between resulting simulations, respectively. The main diagonal of the evaluation matrix shows the relative  
distribution of simulations. All distributions here depicted the higher frequency of low  $^{131}\text{I}$  column densities in spring than in  
445 other seasons (as also seen in Figure 2-A). The correlations computed between  $^{137}\text{Cs}$  column densities in all members are very  
similar to those found for  $^{131}\text{I}$  column densities with the exception that the correlation between the CFSv2 and ERA5-WRF  
decreased from 0.7 to 0.64 (Fig. S10). CFSv2 caused higher underpredictions compared to FNL- and ERA5-WRF in  $^{137}\text{Cs}$   
column density simulations, leading to the decrease of the respective linear coefficients from 0.9 and 0.77 to 0.64 and 0.53.  
While the RMSEs have increased tangibly, FB and NMSE decreased up to zero and 1.2, respectively, and FB5 increased up  
450 to 83%. This indicates that  $^{137}\text{Cs}$  column density simulations from CFSv2 are less biased against (re)analysis members than  
 $^{131}\text{I}$  simulations.



455 **Figure 10** The inter-comparison of the 96-hour integrated simulations of  $^{131}\text{I}$  column densities ( $\text{Bq}/\text{m}^2$ ) from FLEXPART and FLEXPART-WRF runs. The reported model metrics show the performance of members for all simulations over 2019 (full-year) at each grid point. The seasonal regression lines are shown by colors.

The comparison of simulated  $^{137}\text{Cs}$  deposition shows that ensemble members (both forecast and (re)analysis members) have simulated differently (Fig. S11). For example, CFSv2 yielded correlations of less than 0.22 compared with the three (re)analysis members. The linear coefficients between CFSv2 and (re)analysis members are less than 0.1, which indicates much stronger underpredictions in CFSv2 deposition simulations than its column density simulations. The other statistics not only emphasize the inconsistencies between CFSv2 and (re)analysis members but between (re)analysis members. Although there exists relatively higher consistency between the simulations of  $^{131}\text{I}$  deposition (Fig. S12) than  $^{137}\text{Cs}$  deposition, they are still far less correlated than the simulated column densities of both studied radioactive substances. Gudiksen et al. (1988) found errors in precipitation and the deposition formulation as two main sources of uncertainties in deposition modeling. Having compared meteorological fields from ERA-40 and ERA-Interim to force FLEXPART, Evangelidou et al. (2017) also found very large differences (around 55%) in the posterior source term of  $^{137}\text{Cs}$ . It is attributed to differences in the precipitation fields of these two meteorological datasets that induced substantial differences in  $^{137}\text{Cs}$  deposition (and subsequently in the estimated  $^{137}\text{Cs}$  emissions). We also expect that part of the difference between  $^{137}\text{Cs}$  deposition can be attributed to the recent updates in FLEXPART scavenging scheme, compared to the one used in FLEXPART-WRF.

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## Conclusion

470 In this study, we examined the dispersion of radionuclides, including  $^{131}\text{I}$  and  $^{137}\text{Cs}$ , in the event of a nuclear accident at  
different seasons and times of the day. We have used the Lagrangian particle/air parcel dispersion model FLEXible PARTicle  
(FLEXPART) and FLEXPART coupled with the Weather Research and Forecasting model (FLEXPART-WRF). The studied  
radionuclides are assumed to be released within the first 24 hours of a four-day simulation period from an accident at the  
Barakah nuclear power plant (B-NPP) in the UAE. FLEXPART and FLEXPART-WRF were run iteratively 365 times  
475 (simulating each day of 2019) to produce hourly simulations through 96 hours after the accident at the spatial resolution of 10  
km in 14 vertical levels from 5 to 5000 m agl. The source term was scaled to the maximum estimates of the radioactive  
materials from the Fukushima accident in 2011 (22 PBq (6.9 kg) of  $^{137}\text{Cs}$  and 192 PBq (0.042 kg) of  $^{131}\text{I}$ ). In the course of  
examining the study question, we paid special attention to the differences in the simulations when using different  
meteorological inputs. We quantified meteorological uncertainties by producing an ensemble model with three (re)analysis  
480 members including Final Analysis (FNL) at native resolution, FNL and the ECMWF reanalysis 5th generation (ERA5)  
downscaled by WRF (FNL- and ERA5-WRF), and one forecast member which is forced by the NCEP climate forecast system  
version 2 (CFSv2). This ensemble provided the basis for comparing the output of the forecast member (CFSv2) against three  
other (re)analysis members. The simulations of FNL and CFSv2 were compared against that of FNL-WRF and ERA5-WRF  
to find the effect of downscaling and using different model simulation codes (FLEXPART vs. FLEXPART-WRF) on  
485 FLEXPART dispersion modeling. The simulations of all four members were also examined concerning the population density  
of Qatar to study pollution risks for local human communities in case of a nuclear accident in the region. We also studied the  
sensitivity of ERA5- and FNL-WRF simulations to the turbulence scheme used under convective conditions. The results of  
this study can be summarized in three main points:

1- **The transport of radionuclides:** to analyze the time interval between emission and reception of radionuclides, we  
490 used the age composition of radionuclide plumes. The results show that the transport characteristics of  $^{131}\text{I}$  and  $^{137}\text{Cs}$   
did not significantly affect their movement from B-NPP to Qatar, the study area, which was attributed to the lack of  
heavy precipitation in the study area. Regardless of the time of release, the analysis of air parcel age spectra indicates  
that dense radionuclide clouds enter the south study area about 10 to 20 hours after the emission. A significant portion  
of emitted  $^{131}\text{I}$  is transported to the furthest parts of the study area up to 80 hours after the accident. All members also  
495 simulated that a big portion of  $^{137}\text{Cs}$  deposition occurs within 75 to 80 hours after the emission. The age distributions  
of Lagrangian particles in spring follow a bimodal distribution indicating the distant transport of radionuclides in this  
season. Results also show that the number of long-lived air parcels was higher when released between 12 p.m. and 6  
p.m. This may indicate the impact of the planetary boundary layer height on the range of affected areas. Compared to  
other members, the relatively lower spatial resolution of meteorological inputs caused the distribution of air parcel  
ages from CFSv2 to represent an average of other members' age distributions. In addition, the lower temporal  
500 resolution of CFSv2 inputs and, consequently, abrupt changes in the diurnal variation of boundary layer height caused



the infrequency of long-lived particles after sunset in the resulting simulations. The age distributions produced by FNL-WRF and ERA5-WRF were found to be more similar, compared to the one from FNL, due to the downscaling of inputs and the use of a common simulation code. The larger contribution of long-lived particles in FNL simulations was distinctively found in all seasons. We attributed it to the slower and/or more distant transport of air parcels from the emission point in FNL simulations, compared to other members.

2- **The distribution of extremely high concentrations and deposition of radionuclides:** we investigated the seasonal median of the maximum of  $^{131}\text{I}$  concentrations ( $^{131}\text{I}^{\text{conc\_seas\_max}}$ ), which is converted to the maximum hourly doses from inhalation, and of the total  $^{137}\text{Cs}$  deposition ( $^{137}\text{Cs}^{\text{depos\_seas\_max}}$ ) across the study area. As expected, all members simulated the highest (lowest)  $^{131}\text{I}^{\text{conc\_seas\_max}}$  and  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$  over the south (north) of Qatar, being the closest (furthest) point of the study area to the emission point. The inter-seasonal comparison of simulations shows that all ensemble members have simulated inhalation doses up to three to five times stronger in the cold period of the year in the south/southeast of Qatar than in the warm period. The examination of the air parcel release time resulting in  $^{131}\text{I}^{\text{conc\_seas\_max}}$  shows that they are mainly released between 5 a.m. and 12 p.m. It seems that the intensification of the thermal gradient between the land and sea and the resulting onshore winds increased the transport of high  $^{131}\text{I}$  concentrations (inhalation doses), through freshly released air parcels, to the study area at this time of day. Among the ensemble members, the highest levels of inhalation doses were found in FNL simulations (up to 1000  $\mu\text{Sv}$ ) and then in FNL- and ERA5 simulations (up to 600  $\mu\text{Sv}$ ) whereas CFSv2 underpredicted inhalation doses by a factor of about 3 to 5 compared to these (re)analysis members. Considering that FNL and CFSv2 are executed using the same simulation code, the remarkable difference between the resulting  $^{131}\text{I}^{\text{conc\_seas\_max}}$  can be attributed to the difference in meteorological inputs. We found that the 3 to 5-fold elevated planetary boundary layer height caused the stronger dilution of radionuclides at the surface in CFSv2 simulations. In addition, the poorer spatial resolution of CFSv2 inputs than of FNL inputs caused the faulty separation of the boundary layer process along the coastlines in the former. The significant differences in  $^{131}\text{I}^{\text{conc\_seas\_max}}$  simulated by FNL and FNL-WRF show how the use of different model simulation codes (schemes) and downscaling of meteorological inputs may affect FLEXPART modeling and, consequently, the decisions made based on its simulations for real accidents. As was the case with  $^{131}\text{I}^{\text{conc\_seas\_max}}$ ,  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$  in the cold period of the year, especially in winter, in the south of the study area can be up to three times higher than in the warm period. The highest spread of contaminated surfaces with  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$  greater than 40 kBqm<sup>-2</sup> is found in FNL simulations.  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$  mostly occurred when particles are released between 10 a.m. and 2 p.m. This can be attributed to the development of the boundary layer and the increased transport of radionuclides from the source to conterminous areas at this time of the day. The examination of the frequency with which  $^{131}\text{I}$  and  $^{137}\text{Cs}$  concentrations above the 66<sup>th</sup> percentiles are transported to the populated areas of eastern Qatar shows that, in all members, more than 50% of the extreme cases occurred in winter and between 15 and 30% in spring. The above findings show that any nuclear accident at B-NPP in the winter will be more likely to be accompanied by





535 the transfer of high radionuclide concentrations and deposition within the study area. This pronounced intra-annual  
distribution could be attributed to a seasonal atmospheric pattern in which south/southeast winds transport the dense  
radionuclide clouds to the study area. The collocation of population density and simulations shows that the populated  
areas (with more than 15 people per arc-second) coincided with moderate to low levels of  $^{131}\text{I}^{\text{conc\_seas\_max}}$  (inhalation  
doses) and  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$  of less than 200  $\mu\text{Sv}$  and 100  $\text{kBq}\cdot\text{m}^{-2}$ , respectively. Uninhabited areas in southern Qatar  
540 have received the highest levels of  $^{131}\text{I}^{\text{conc\_seas\_max}}$  and  $^{137}\text{Cs}^{\text{depos\_seas\_max}}$ . In this study, we also investigated the effect of  
the turbulence scheme selected under convective conditions on radionuclide dispersion. The implementation of the  
skewed turbulence model (STM), instead of the Gaussian one (GTM), in ERA5-WRF and FNL-WRF decreased the  
median of simulated  $^{131}\text{I}$  concentrations and increased the high levels of simulated  $^{137}\text{Cs}$  deposition. According to  
Pisso et al. (2019), it can be interpreted as the decline in radionuclide transport to areas away from the source due to  
545 the intensified deposition in the areas around the source under skewed turbulence conditions.

3- **The inter-comparison of ensemble members:** the outputs of CFSv2 were compared with that of three (re)analysis  
members. In general, CFSv2 simulations of  $^{137}\text{Cs}$  and  $^{131}\text{I}$  column density are most highly correlated with FNL,  
producing a Pearson correlation coefficient of around 0.6, and then with FNL-WRF. This is because all three members  
have a similar meteorological base model. The evaluation metrics including fraction bias ( $\text{FB}=0.02$ ), the fraction of  
550 simulations within the factor of 5 ( $\text{FB5}=67.4$ ), and normalized RMSE ( $\text{NMSE}=1.5$ ) also show that CFSv2 produced  
the lowest absolute differences with FNL. Less than unity linear regression coefficients between CFSv2 and other  
members, particularly between CFSv2 and FNL in spring, show that CFSv2 caused considerable underpredictions in  
 $^{131}\text{I}$  column density simulations, mainly due to the elevated input PBLH. Regarding  $^{137}\text{Cs}$  column density simulations,  
CFSv2 produced stronger underpredictions, but smaller absolute differences against (re)analysis members. CFSv2  
555 produced the lowest absolute difference with FNL-WRF in  $^{137}\text{Cs}$  column density simulations ( $\text{FB}=-0.02$ ,  $\text{FB5}=83.7$ ,  
and  $\text{NMSE}=1.2$ ). Considering the performance statistics between simulations from CFSv2 and reanalysis members  
and those computed between Xe-133 observations and simulations (Maurer et al., 2018) indicates that the uncertainty  
of  $^{131}\text{I}$  and  $^{137}\text{Cs}$  column density simulations by CFSv2 fall within the expected uncertainty range. Considering all  
members, the lowest absolute difference between  $^{131}\text{I}$  and  $^{37}\text{Cs}$  column density simulations is found between FNL-  
560 WRF and ERA5-WRF yielding  $\text{FB}=-0.07$  and  $-0.05$ ,  $\text{FB5}=79.3\%$  and  $83.6\%$ ,  $\text{RMSE}=14.9 \times 10^8$  and  $2946.3 \times 10^7$ ,  
and  $\text{NMSE}=1.3$  and  $1.1$ . This underlines that the use of different downscaled inputs but the same simulation code in  
these two members increased the similarity between resulting simulations to the extent that surpasses the association  
between simulations from FNL and FNL-WRF with identical meteorological inputs. The deposition simulations from  
all members show a larger inconsistency for both  $^{131}\text{I}$  and  $^{137}\text{Cs}$ . As already stated in the literature (Girard et al., 2016,  
565 Gudiksen et al., 1988, Evangeliou et al., 2017), this can be partially attributed to uncertainties in input precipitation  
fields and the recent updates in the wet deposition scheme of FLEXPART.



**Data availability.** The FLEXPART and FLEXPART-WRF simulations are available upon request. The open-source codes for the FLEXPART 10.4 and FLEXPART-WRF 3.3.2 can be downloaded from <https://www.flexpart.eu/downloads> (last access: 570 27 May 2022). Qatar's high-resolution population density datasets are freely available at <https://data.humdata.org/dataset/qatar-high-resolution-population-density-maps-demographic-estimates> (last access: 27 May 2022).

**Author contributions.** SON performed the WRF, FLEXPART, and FLEXPART-WRF simulations and led the integration of results and writing. SON and TC designed the experiments. All the co-authors have read the paper and provided professional 575 comments.

**Competing interests.** The authors declare that they have no conflict of interest.

#### Acknowledgments.

The CyI High-Performance Computing Facility and Qatar Environment and Energy Research Institute (QEERI) High-Performance Computer provided computational resources supporting this work. The authors thank the FLEXPART and 580 FLEXPART-WRF developers for providing the transport model source codes.

#### Financial support.

This research has received funding from the Qatar Environment & Energy Research Institute (QEERI), an entity of the Hamad Bin Khalifa University, wholly owned by the Qatar Foundation for Science, Education, and Community Development.

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