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Identification and verification of worst-case radiological transport scenarios for Ireland: a simulation-based approach to nuclear emergency preparedness (2011-2024)

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1

2 This study presents a comprehensive simulation-based assessment of potential transboundary radiological
3 transport to Ireland from six nuclear facilities in the United Kingdom and France, utilising weather data
4 over a fourteen-year period (2011–2024). Systematic screening of 2.2 million HYSPLIT atmospheric dis-
5 persion simulations identified eighteen worst-case scenarios representing conditions of maximum ground
6 deposition, maximum air concentration, and minimum warning time for protective action implemen-
7 tation. Independent verification using FLEXPART and HYSPLIT demonstrated expected inter-model
8 variability (factor of 1–10), with both Lagrangian models providing consistent risk assessment brackets.
9 Heysham, despite its complex 19-isotope AGR source term, produced negligible radiological doses to
10 Ireland (< 0.01 mSv)—more than four orders of magnitude below intervention thresholds. More distant
11 continental facilities (Flamanville, Paluel, Sizewell B) showed low but measurable doses (0.1–4.6 mSv de-
12 pending on scenario and model), remaining well below the 50 mSv sheltering threshold. Hinkley Point C
13 (under construction) showed elevated but sub-threshold doses (0.3–8.5 mSv depending on model). How-
14 ever, the cancelled Wylfa Newydd gigawatt-scale project (the site is now proposed for small modular
15 reactors), owing to its extreme proximity to Ireland, exhibited concerning dose predictions: FLEXPART
16 calculated 20.7 mSv under maximum deposition conditions (May 2024 scenario), approaching the 50 mSv
17 sheltering threshold, whilst HYSPLIT predicted 4.5 mSv. This inter-model variability (factor of ∼5) high-
18 lights genuine uncertainty for near-source impacts but converges on a critical finding: were a gigawatt-scale
19 reactor constructed at the Wylfa site, severe accidents during specific meteorological patterns could re-
20 quire protective actions in Ireland. Machine learning models (XGBoost) achieved validation accuracies of
21 85–93% for rapid impact prediction, whilst global sensitivity analysis revealed that meteorological condi-
22 tions, rather than release parameters, dominate consequence severity. These findings provide quantitative
23 assurance that existing nuclear infrastructure poses low transboundary risk to Ireland well below interven-
24 tion thresholds, whilst demonstrating that facility proximity constitutes the dominant factor determining
25 potential radiological impact.

26 1 Introduction

27 The Environmental Protection Agency (EPA) plays a pivotal role as a national principal support agency in
28 Ireland, tasked with responding to nuclear or radiological emergencies that could affect the nation. A critical
29 component of this responsibility is the provision of robust technical support and evidence-based advice to the
30 National Emergency Co-ordination Group. This support is substantially reliant on the application of medium
31 and long-range atmospheric dispersion models. These models are instrumental in predicting the trajectory,
32 extent, and potential consequences of radioactive material released into the atmosphere, typically originating
33 from nuclear accidents at facilities abroad. Key model outputs, such as predicted radioactive deposition
34 on terrestrial surfaces, ambient air concentrations, and precise plume arrival times, are fundamental for
35 estimating the potential radiological implications for Ireland. Such estimations are vital for informing timely
36 and effective decisions regarding protective actions, which may include recommendations for public sheltering
37 or the safeguarding of agricultural resources like livestock [19]. This study focuses principally on protective
38 actions in the early phase of a nuclear emergency; protective actions related to food controls arising from the
39 uptake of radionuclides into the food chain are beyond the scope of the present analysis.

40 Previous work, such as the 2013 report by the Radiological Protection Institute of Ireland (RPII) [27],
41 assessed the potential radiological impacts on Ireland from proposed new nuclear power plants in the UK.
42 That study considered both routine discharges and a range of postulated accident scenarios, utilising specific
43 weather patterns designed to maximise radioactive transfer to Ireland. For severe accidents, the RPII (2013)
44 report highlighted that weather conditions were a dominant factor, with most scenarios not resulting in di-
45 rect atmospheric transport over Ireland, but noted that under specific adverse conditions, protective actions
46 including sheltering and food controls would be necessary [27]. Subsequent assessments, including the 2016
47 EPA evaluation of postulated accidents at the Sellafield nuclear fuel reprocessing plant, similarly employed
48 atmospheric dispersion modelling to identify worst-case scenarios and concluded that doses would remain be-
49 low international intervention thresholds [10]. More recently, Joy (2020) conducted comprehensive modelling
50 of accidental radioactive releases for Ireland, comparing HYSPLIT and FLEXPART model performance with
51 ECMWF meteorological data and emphasising the value of ensemble modelling approaches for emergency
52 preparedness [22]. Whilst these previous studies provided valuable insights, they were typically limited to
53 single-site assessments or restricted temporal sampling of meteorological conditions, motivating the present
54 comprehensive, multi-site, fourteen-year systematic analysis.

55 The selection of appropriate atmospheric transport and dispersion models is critical for emergency pre-
56 preparedness. Lagrangian particle dispersion models have emerged as the preferred methodology for simulating
57 long-range transport from nuclear facilities, offering fundamental advantages over Eulerian grid-based ap-
58 proaches. The HYSPLIT (Hybrid Single-Particle Lagrangian Integrated Trajectory) model, developed by
59 NOAA's Air Resources Laboratory, has been extensively validated for long-range atmospheric transport and
60 serves as a cornerstone of operational emergency response systems worldwide [8, 40]. HYSPLIT employs
61 a hybrid computational approach, utilising a Lagrangian reference frame for calculating particle advection
62 and diffusion whilst computing concentrations on a fixed Eulerian grid. The model has demonstrated strong
63 performance in validation studies against monitoring data from the Fukushima Daiichi Nuclear Power Plant
64 accident [6, 15], confirming its capability to accurately simulate transboundary transport of radioactive ma-
65 terial.

66 The FLEXPART (FLEXible PARTicle dispersion model) provides an independent Lagrangian framework
67 specifically designed for simulating long-range dispersion of pollutants from point sources [41]. Originally
68 validated against large-scale tracer experiments [42], FLEXPART has undergone continuous development with
69 recent versions incorporating improved parameterisations for turbulent mixing and wet deposition [2, 32]. The
70 fundamental advantage of Lagrangian models over Eulerian approaches lies in their freedom from numerical
71 diffusion: Lagrangian particle trajectories retain spatial resolution independent of the computational grid,
72 preserving plume structure during long-range transport where Eulerian models would artificially dilute the
73 plume through repeated grid-cell averaging.

74 Model intercomparison studies have consistently demonstrated that atmospheric dispersion predictions
75 contain inherent uncertainties arising from meteorological inputs, physical parameterisations, and numeri-
76 cal implementations [12, 48]. The ENSEMBLE and SEED intercomparison exercises, comparing multiple
77 models for hypothetical nuclear accidents, revealed that multi-model ensemble approaches generally outper-
78 form individual deterministic simulations, particularly for complex meteorological regimes [12]. More recent

79 intercomparisons, including the multi-model ^{137}Cs dispersion study from the Fukushima Daiichi accident
80 using identical input data, demonstrated that models perform better near the source than at regional scales,
81 with inter-model spread indicating structural model uncertainties [35]. Recent studies have also explored
82 meteorological ensemble forecasting to quantify dispersion uncertainty, demonstrating that whilst ensemble
83 approaches capture meteorological variability, they require extensive computational resources [25]. This body
84 of evidence has significant implications for emergency response: verifying worst-case scenarios using inde-
85 pendent models with identical meteorological forcing provides confidence that identified transport patterns
86 represent genuine physical phenomena rather than model-specific artefacts or meteorological input differences.

87 Radiological consequence assessment requires coupling atmospheric transport predictions with appropriate
88 source terms and dose conversion methodologies. Severe accident source terms are typically derived from Level
89 2 probabilistic risk assessment, characterising the timing, duration, and isotopic composition of atmospheric
90 releases following core damage and containment failure [38]. Post-Fukushima source term reconstructions
91 using inverse modelling and atmospheric observations have demonstrated the importance of realistic accident
92 progression physics, including containment retention and aerosol depletion mechanisms, which substantially
93 reduce environmental releases relative to total core inventory [4, 36]. The International Commission on
94 Radiological Protection provides authoritative dose coefficients for converting atmospheric concentrations
95 and ground deposition into effective dose through multiple pathways including inhalation, cloudshine, and
96 groundshine [9, 17, 18, 31]. Integration of these standardised methodologies ensures that predicted doses can
97 be directly compared against international intervention thresholds and emergency reference levels.

98 This current study aims to significantly enhance and broaden the EPA's preparedness by systematically
99 reviewing and updating the inputs and methodologies for these atmospheric dispersion models, and by con-
100 ducting a comprehensive assessment of potential radiological transport pathways. The scope of this research
101 is informed by several evolving factors: the ongoing development of new nuclear facilities across Europe, the
102 operational extensions of many existing plants, and the dynamic nature of the international nuclear safety
103 landscape, which includes considering the potential implications arising from geopolitical events, such as the
104 conflict in Ukraine, on nuclear facility safety and security.

105 This study employs a systematic, multi-method approach to identify and verify worst-case radiological
106 transport scenarios that could impact Ireland. The research integrates large-scale atmospheric dispersion
107 modelling, independent model verification, machine learning for impact prediction, and global sensitivity
108 analysis to provide a comprehensive, quantitative basis for emergency preparedness planning.

109 The primary objective is the systematic identification of worst-case meteorological scenarios through large-
110 ensemble HYSPLIT (Hybrid Single-Particle Lagrangian Integrated Trajectory) screening over a fourteen-year
111 period (2011–2024), substantially longer than previous Irish assessments. This extensive temporal coverage
112 encompasses six nuclear facilities in proximity to Ireland, with parametric variations in release height (20–
113 100 m) and duration (6–48 h) to capture the full spectrum of potential accident conditions. Initial simulations
114 employed unit releases to isolate atmospheric transport characteristics, enabling the identification of high-
115 consequence scenarios based on total deposition (mass/m²), average air concentration (mass/m³), and plume
116 arrival times over Irish territory.

117 An important component of this study is the independent verification of all identified scenarios using
118 the FLEXPART Lagrangian particle dispersion model. This dual-model approach provides confidence that
119 identified transport patterns and radiological consequences represent genuine emergency planning concerns
120 rather than model-specific artefacts. For each worst-case scenario, realistic source terms representing severe
121 accident conditions (core melt with late containment failure) were applied, scaled to each NPP's thermal
122 power, to calculate potential dose distributions across Ireland. The agreement between two independent
123 dispersion models strengthens confidence in the robustness of the identified scenarios for protective action
124 planning.

125 Beyond scenario identification and verification, the research develops predictive capability through ma-
126 chine learning (XGBoost) with cross-validated champion models to enable rapid impact assessment during
127 emergencies. Global sensitivity analysis using Sobol indices quantifies which physical and release param-
128 eters most strongly influence radiological consequences, providing critical insights for monitoring priorities
129 and model refinement. Crucially, both the machine learning and sensitivity analysis methodologies leverage
130 the extensive simulation ensemble (2.2 million scenarios) originally generated for worst-case identification,
131 maximising the scientific and operational value extracted from this substantial computational investment.
132 Collectively, these methods establish a contemporary, evidence-based foundation for Ireland's nuclear emer-

133 gency response strategies.

134 The structure of this paper is organised as follows. Section 2 describes the pipelines for identifying
135 worst-case scenario NPP incidents for Ireland, the atmospheric dispersion modelling frameworks employed
136 (HYSPPLIT and FLEXPART), model verification approach, source term definitions, radiological dose as-
137 sessment methodology, machine learning model development, and sensitivity analysis techniques. Section 3
138 presents the identified worst-case scenarios, independent FLEXPART verification, radiological dose assess-
139 ments, machine learning performance, and sensitivity analysis findings. Section 4 interprets the results in
140 the context of emergency preparedness, discusses model uncertainty and robustness, and identifies critical
141 parameters for monitoring and decision-making. Section 5 summarises key findings and recommendations
142 for Ireland’s ongoing preparedness enhancement.

143 **2 Methodology**

144 **2.1 Nuclear Facilities Considered**

145 This study assessed six nuclear facilities in the United Kingdom and France that are proximal to Ireland
146 and represent potential sources of transboundary radiological impact. The facilities span a range of reactor
147 technologies, operational statuses, and distances from Ireland (measured from Dublin, 53.35°N, 6.26°W),
148 providing a comprehensive assessment of the nuclear landscape relevant to Irish emergency preparedness.

149 **Wylfa** (53.42°N, 4.48°W, ~130 km) on the Isle of Anglesey, Wales, is the closest potential nuclear site to
150 Ireland. The original Magnox station ceased operations in 2015. The proposed Wylfa Newydd project (two
151 1.35 GW_e EPR units) was cancelled in 2020; however, the site has subsequently been identified for potential
152 small modular reactor (SMR) development by Rolls-Royce (470 MW_e PWR design). This study models
153 the originally proposed gigawatt-scale EPR configuration, representing an upper-bound worst-case scenario;
154 actual SMR consequences would be substantially lower due to reduced thermal power and core inventory.

155 **Heysham** (54.03°N, 2.91°W, ~250 km) in Lancashire comprises two operational Advanced Gas-cooled
156 Reactor (AGR) stations: Heysham 1 (2 × 580 MW_e, operational since 1983) and Heysham 2 (2 × 615 MW_e,
157 operational since 1988). AGR source terms differ substantially from PWR releases, incorporating a broader
158 isotope spectrum including actinides (plutonium, americium, curium) and additional fission products (stron-
159 tium, ruthenium, cerium), necessitating a 19-isotope source term for this facility compared to 4 isotopes for
160 PWR/EPR sites.

161 **Hinkley Point C** (51.21°N, 3.14°W, ~340 km) in Somerset is currently under construction, comprising
162 two 1.63 GW_e EPR units. When operational (expected mid-2030s), it will be the UK's largest nuclear
163 power station. The EPR design incorporates enhanced safety features including a core catcher and double
164 containment.

165 **Sizewell B** (52.22°N, 1.62°E, ~480 km) in Suffolk is the UK's only operating PWR (1.2 GW_e, operational
166 since 1995). A sister station (Sizewell C, 2 × EPR) has received development consent but construction has
167 not yet commenced.

168 **Flamanville** (49.54°N, 1.88°W, ~550 km) in Normandy, France, hosts two operational PWR units (2 ×
169 1.33 GW_e, operational since 1986–1987) and one EPR unit (1.65 GW_e) that achieved first criticality in 2024
170 after extended construction delays.

171 **Paluel** (49.86°N, 0.64°E, ~650 km), also in Normandy, is France's largest nuclear station with four PWR
172 units (4 × 1.33 GW_e, operational since 1984–1986). Its position on the English Channel coast and substantial
173 combined thermal power make it relevant for Irish emergency planning despite being the most distant facility
174 considered.

175 **2.2 Atmospheric Dispersion Modelling**

176 The atmospheric transport and dispersion modelling framework employed two independent Lagrangian parti-
177 cle dispersion models: HYSPLIT (Hybrid Single-Particle Lagrangian Integrated Trajectory) and FLEXPART
178 (FLEXible PARTicle dispersion model). This dual-model approach enables quantification of inter-model vari-
179 ability and provides verification that identified worst-case scenarios represent robust atmospheric transport
180 patterns rather than model-specific artefacts.

181 HYSPLIT served as the primary screening tool for identifying worst-case scenarios across the fourteen-
182 year study period (2011–2024). The model was configured with a hybrid computational scheme, employing
183 Lagrangian particle trajectories for advection and diffusion calculations whilst computing concentrations
184 on a fixed Eulerian output grid. Meteorological forcing utilised ERA5 reanalysis data from the European
185 Centre for Medium-Range Weather Forecasts, accessed via HYSPLIT-compatible ARL-format files providing
186 hourly meteorological fields at 0.25-degree horizontal resolution [16]. The selection of ERA5 over older
187 reanalysis datasets (e.g., ERA-Interim) is critical for long-range transport studies: comparative analyses
188 demonstrate that ERA5 provides superior representation of atmospheric boundary layer height (correlation
189 0.88 vs radiosondes) and vertical transport processes, with improved assimilation of satellite observations and
190 better resolution of meso- to synoptic-scale meteorological features [14, 45]. Validation studies over European
191 marine environments confirm ERA5's robust performance for offshore wind fields, which are particularly
192 relevant for Irish Sea transboundary transport [5]. The model's vertical structure incorporated 25 output
193 levels ranging from the surface to 3000 metres above ground level, with enhanced resolution in the atmospheric

194 boundary layer (0, 20, 30, 40, 50, 60, 80, 100, 150, 200, 250, 300, 400, 500, 600, 750, 1000, 1250, 1500, 1750, 195 2000, 2250, 2500, 2750, and 3000 m). This vertical distribution provides enhanced resolution near the surface 196 where concentration gradients are steepest and human exposure is most relevant: seven levels within the 197 breathing zone (0–100 m) enable precise calculation of inhalation doses, whilst the continued vertical spacing 198 through the boundary layer and lower troposphere captures the evolution of boundary layer mixing, which 199 dominates near-surface concentration patterns and determines ground-level deposition rates.

200 The HYSPLIT screening simulations employed unit releases (1 kg total mass) to isolate atmospheric 201 transport characteristics from source term uncertainties. Each simulation released computational particles 202 at rates of 500 particles per hour for release durations of 6 h (3000 particles total), 24 h (12,000 particles), or 203 48 h (24,000 particles). Simulations were executed for every day of the fourteen-year period at release hours 204 0, 3, 6, 9, 12, 15, 18, and 21 UTC, combined with three release heights (20, 50, 100 m above ground level), 205 yielding 367,200 simulations per nuclear power plant. The computational domain extended from 10.47°W 206 to 6.01°E longitude and from 51.45°N to 55.38°N latitude, encompassing Ireland, the UK, and portions of 207 continental Europe with 0.25-degree horizontal resolution.

208 The large-scale simulation campaign (2.2 million scenarios across six nuclear power plants) was executed 209 using a custom-built automation pipeline implemented in Julia, employing distributed parallel processing 210 across 31 computational cores. The pipeline systematically generated HYSPLIT control files for each parameter 211 combination (date, release hour, duration, height, and facility), executed simulations, processed output 212 concentration fields and particle trajectory files, and automatically identified scenarios yielding atmospheric 213 transport to Irish territory using point-in-polygon geometric algorithms applied to a high-resolution Republic 214 of Ireland boundary definition (Northern Ireland was excluded from the analysis domain). A checkpointing 215 system with persistent state management enabled resumable execution following interruptions, crucial for 216 managing the multi-month computational campaign spanning fourteen years of meteorological conditions. 217 This automated framework ensured systematic coverage of the complete parameter space whilst maintaining 218 computational efficiency, result consistency, and traceability of the 367,200 simulations per facility.

219 Deposition processes were parameterised to represent typical aerosol behaviour characteristic of fission 220 products released during severe accidents. Dry deposition employed a constant deposition velocity 221 of 0.005 m s^{-1} , a value empirically validated from Chernobyl ^{137}Cs measurements representing the mean deposition 222 rate over heterogeneous terrain and adopted as the standard parameter in HYSPLIT regional-scale 223 simulations [28, 37]. This value falls at the conservative upper end of measured caesium aerosol deposition 224 velocities ($1\text{--}5 \text{ mm s}^{-1}$) [43] and has demonstrated superior agreement with observed contamination 225 patterns compared to complex resistance-based schemes for regional applications. Wet deposition utilised 226 scavenging coefficients of 5.0×10^{-5} for both in-cloud and below-cloud removal processes, with precipitation 227 fields derived directly from ERA5 data. Noble gases were treated as non-depositing tracers with negligible 228 deposition parameters. Radioactive decay was not applied during the screening phase to preserve the unit- 229 release framework; decay corrections were implemented subsequently during radiological dose assessment 230 using isotope-specific half-lives.

231 FLEXPART version 10.4 provided independent verification of identified worst-case scenarios. The model 232 employs a purely Lagrangian framework, computing particle trajectories using three-dimensional wind fields 233 with stochastic perturbations representing turbulent diffusion. FLEXPART's convective parameterisation 234 (LCONVECTION=1) explicitly represents sub-grid scale vertical transport in developing cumulus clouds, 235 a process particularly relevant for daytime boundary layer evolution. The model utilised identical ERA5 236 meteorological forcing as HYSPLIT, ensuring that inter-model differences reflected physical parameterisations 237 rather than meteorological inconsistencies.

238 For verification simulations, FLEXPART employed 5000 to 12,000 computational particles depending 239 on release complexity, providing statistical robustness of concentration fields. The output grid matched 240 HYSPLIT specifications exactly (identical horizontal extent, resolution, and vertical levels) to enable direct 241 point-by-point comparison. Deposition parameterisations were harmonised with HYSPLIT settings: dry 242 deposition velocity 0.005 m s^{-1} , wet scavenging coefficients 5.0×10^{-5} , and isotope-specific material densities 243 ranging from 1879 kg m^{-3} (cesium) to $19,860 \text{ kg m}^{-3}$ (plutonium). Care was taken to account for unit system 244 differences: FLEXPART expects densities in SI units (kg m^{-3}) whilst HYSPLIT requires CGS units (g cc^{-1}), 245 necessitating division by 1000 for HYSPLIT input. Radioactive decay was implemented using isotope-specific 246 half-lives during dose assessment calculations.

247 2.3 Element-Specific Deposition Parameters

248 Radionuclide transport and deposition were parameterised according to element-specific chemical properties,
 249 recognising that different elements exhibit distinct atmospheric behaviours due to variations in hygroscopicity,
 250 solubility, and particle formation characteristics. Table 1 presents the element-specific parameters employed
 251 for both HYSPLIT and FLEXPART simulations.

Table 1: Element-specific deposition parameters for radionuclide transport modelling

Element Category	Elements	Scavenging (1/s)	Dry Velocity (m/s)	CCN Efficiency	Chemical Basis
Noble Gases	Xe, Kr	0.0	0.0	0.0	Chemically inert
Soluble	Cs, I	1.0×10^{-4}	0.0015–0.002	0.9	Hygroscopic salts
Aerosols	Te, Sr	8.0×10^{-5}	0.0015	0.8	Soluble oxides
	Ru	5.0×10^{-5}	0.001	0.6	Lower solubility
Insoluble/Refractory	Pu, Am, Cm	2.0×10^{-5}	0.005	0.2	Insoluble ceramics
	Ce	3.0×10^{-5}	0.002	0.3	Refractory oxide

252 The physical rationale for these parameterisations reflects fundamental differences in atmospheric chem-
 253 istry [23, 39]. Caesium and iodine form highly hygroscopic salts (CsOH, CsI, I₂) with strong affinity for water
 254 vapour, exhibiting high cloud condensation nuclei (CCN) efficiency and rapid wet scavenging. Actinides (Pu,
 255 Am, Cm) form refractory oxide ceramics with minimal water solubility, requiring higher dry deposition veloc-
 256 ities but showing reduced wet removal. Noble gases remain chemically inert with zero deposition. Parameter
 257 values represent central estimates within the typical uncertainty range of factors of 2–5 for wet and dry depo-
 258 sition parameterisations [39]; sensitivity to meteorological and release parameters is explored in Section 2.8.
 259 Both FLEXPART and HYSPLIT employed identical element-specific parameters to ensure consistent physics
 260 between models, enabling valid inter-model comparison of predicted dose distributions.

261 2.4 Model Verification Approach

262 The model verification strategy addresses a fundamental challenge in emergency preparedness: atmospheric
 263 dispersion models contain inherent uncertainties arising from simplified physical parameterisations, finite
 264 spatial resolution, and meteorological input errors [12]. Reliance on a single model for identifying worst-
 265 case scenarios risks conflating genuine high-consequence transport patterns with model-specific numerical
 266 artefacts. The verification approach employed in this study ensures that all identified worst-case scenarios
 267 exhibit consensus between two independent Lagrangian models, thereby providing confidence that predicted
 268 transport patterns and radiological consequences represent robust emergency planning concerns.

269 Following completion of the HYSPLIT screening phase, the highest-consequence scenarios for each nuclear
 270 power plant (maximum total deposition, maximum average air concentration, minimum plume arrival time)
 271 were selected for detailed verification. These scenarios, identified using unit-release simulations, were re-
 272 run using both HYSPLIT and FLEXPART with realistic multi-isotope source terms representing severe
 273 accident conditions. The source terms incorporated reactor-specific radionuclide inventories scaled according
 274 to thermal power, with release fractions and isotopic compositions derived from Level 2 probabilistic safety
 275 assessment for late containment failure scenarios [1, 38].

276 The verification analysis quantified inter-model agreement through three complementary metrics. Spatial
 277 correlation coefficients evaluated the degree to which both models predicted similar geographical patterns of
 278 ground-level deposition and time-integrated air concentration across Ireland. Normalised root-mean-square
 279 differences quantified the magnitude of inter-model divergence relative to mean predicted values, with values
 280 below 0.5 indicating strong agreement and values exceeding 1.0 suggesting substantial model-dependent
 281 uncertainty. Peak concentration ratios compared the maximum predicted values from each model, identifying
 282 whether extreme values represented model consensus or outliers.

283 An important methodological distinction concerns deposition field representation. FLEXPART provides
 284 explicit wet and dry deposition output fields calculated through its integrated deposition schemes. HYS-
 285 PLIT’s groundshine dose calculations employed in this study utilise near-surface (0 m) air concentration
 286 fields as a well-established proxy for deposited material, a standard approach in operational HYSPLIT dose

287 assessment [8]. This methodological difference is considered when interpreting inter-model differences in
288 deposition-driven dose pathways, particularly for groundshine contributions where surface contamination
289 fields drive external exposure estimates.

290 Concentration field comparisons were conducted at multiple vertical levels within the atmospheric bound-
291 ary layer (0, 50, 100, 200, 500, 1000 m above ground level) to assess whether inter-model agreement varied
292 with altitude. This vertical analysis addresses the potential for models to diverge in their representation
293 of vertical mixing processes, particularly during convective conditions when boundary layer depth evolves
294 rapidly. Time-series comparisons at fixed receptor locations quantified temporal consistency, verifying that
295 both models predicted similar plume arrival times, concentration build-up rates, and exposure durations.

296 The radiological dose calculations provided the ultimate verification metric, integrating spatial concen-
297 tration patterns, temporal evolution, and multi-isotope contributions through internationally standardised
298 dose conversion factors. Agreement in predicted total effective dose distributions across Ireland constitutes
299 the most policy-relevant verification measure, as emergency response decisions depend on dose magnitudes
300 relative to intervention thresholds rather than concentration values per se. Verification simulations that
301 demonstrated dose agreement within a factor of two between models were classified as robust worst-case
302 scenarios suitable for emergency planning applications.

303 2.5 Worst-Case Scenario Identification

304 The systematic identification of worst-case scenarios proceeded through a multi-stage computational ap-
305 proach designed to isolate the atmospheric conditions and release parameters yielding maximum radiological
306 consequences for Ireland. The methodology balanced computational efficiency—enabling analysis of 367,200
307 simulations per facility—with sufficient parametric coverage to capture the full range of plausible accident
308 conditions.

309 The screening phase employed unit releases (1 kg total mass) to decouple atmospheric transport physics
310 from source term uncertainties. This approach enabled direct comparison of atmospheric transport efficiency
311 across different meteorological regimes without confounding effects from varying radionuclide inventories or
312 isotopic compositions. Three output metrics characterised the severity of each simulated scenario: total
313 deposition (mass per square metre integrated over Irish territory), average air concentration (mass per cubic
314 metre averaged over Ireland for the simulation duration), and plume arrival time (hours from release until
315 first detection of airborne material over Ireland).

316 Release parameters were varied systematically to span the range of severe accident conditions. Release
317 heights of 20, 50, and 100 metres above ground level represented different degrees of initial plume buoyancy,
318 with lower releases characteristic of filtered vented containment and higher releases representing unfiltered
319 stack releases or thermal buoyancy from heat-driven releases. Release durations of 6, 24, and 48 hours
320 captured both short-duration breach scenarios and protracted releases associated with gradual containment
321 degradation. The combination of three release heights and three durations yielded nine release configurations,
322 each executed at eight times of day (0, 3, 6, 9, 12, 15, 18, 21 UTC) to capture diurnal variations in atmospheric
323 stability and wind patterns.

324 Temporal coverage extended from 1 January 2011 through 31 December 2024, providing fourteen years
325 of historical and near-present meteorological conditions. This period substantially exceeds previous Irish
326 assessments, which typically examined individual years or limited seasonal samples. The extended temporal
327 coverage ensures that identified worst-case scenarios represent genuinely extreme atmospheric conditions
328 rather than artefacts of limited sampling. The fourteen-year period encompasses multiple phases of the
329 North Atlantic Oscillation, capturing both typical westerly flow regimes and anomalous easterly transport
330 patterns that favour transport from continental Europe toward Ireland.

331 For each nuclear power plant, the complete set of simulations was ranked according to each output metric.
332 The scenario yielding maximum total deposition was identified as the worst-case for long-term contamination
333 and agricultural impacts. The scenario producing maximum average air concentration represented the worst-
334 case for acute inhalation exposure. Scenarios with minimum plume arrival time identified conditions providing
335 least warning time for protective action implementation. These three scenarios per nuclear power plant (18
336 scenarios total across six facilities) constituted the priority cases for detailed FLEXPART verification and
337 radiological dose assessment.

338 2.6 Radiological Dose Assessment

339 Radiological dose assessment translates predicted atmospheric concentrations and ground deposition into ef-
340 fective dose, enabling direct comparison against international intervention thresholds and emergency reference
341 levels. The methodology followed internationally standardised frameworks established by the International
342 Commission on Radiological Protection [9, 17, 31], ensuring consistency with emergency planning guidance
343 adopted by regulatory authorities in Ireland and neighbouring countries.

344 Severe accident source terms were developed for all six nuclear power plants through thermal power
345 scaling from reference designs. Pressurised water reactor (PWR) and European Pressurised Reactor (EPR)
346 source terms employed a four-isotope simplified inventory (tellurium-132, iodine-131, xenon-133, cesium-137)
347 representing volatile fission products expected during late containment failure scenarios. The isotopic re-
348 lease fractions (approximately 5.4% of core iodine-131 inventory and 5.5% of cesium-137 inventory) represent
349 the net environmental release after accounting for in-containment depletion processes, consistent with en-
350 vironmental release magnitudes for scenarios where containment failure occurs many hours after core melt
351 initiation [38]. During this extended pre-release period, natural aerosol depletion processes (gravitational set-
352 tling, diffusiophoresis, thermophoresis) remove 90–95% of condensable fission products from the containment
353 atmosphere, substantially reducing the environmental release relative to core inventory.

354 Activities were scaled according to each reactor’s thermal power rating using the relationship: $\text{Activity}_{\text{target}} =$
355 $\text{Activity}_{\text{reference}} \times (\text{MW}_{\text{th,target}}/\text{MW}_{\text{th,reference}})$. The reference source term derived from French P’4 PWR de-
356 sign documentation (3817 MW_{th}), with Flamanville and Paluel employing unscaled reference values, Hinkley
357 Point C and Wylfa scaled by a factor of 1.185 (4524 MW_{th} EPR design), and Sizewell B scaled by 0.911
358 (3479 MW_{th} Westinghouse PWR). The Heysham Advanced Gas-Cooled Reactor source term incorporated 19
359 radionuclides including actinides (plutonium-238/239/240/241, americium-241, curium-242/244), reflecting
360 the fundamentally different core characteristics and accident phenomenology of graphite-moderated gas-
361 cooled reactors: higher graphite dust generation, enhanced oxidation of metallic fuel cladding, and increased
362 volatilisation and transport of low-volatility species including actinides compared to water-cooled designs.

363 Release rates (Becquerels per minute) were computed by dividing total isotope activity by release duration
364 in minutes, ensuring uniform temporal distribution of emissions. This parameterisation required conversion
365 between activity-based source terms (the standard specification in Level 2 probabilistic safety assessment)
366 and mass-based model inputs (required by both HYSPLIT and FLEXPART). The conversion employed
367 isotope-specific activities calculated from half-life and atomic mass: $\lambda_{\text{isotope}} = (\ln 2 \times N_A)/(t_{1/2} \times M)$, where
368 N_A represents Avogadro’s number, $t_{1/2}$ the half-life in seconds, and M the molar mass in kilograms per mole.
369 Short-lived isotopes such as iodine-132 (half-life 2.3 hours) exhibit specific activities exceeding 10^{19} Bq kg⁻¹,
370 whilst long-lived isotopes like cesium-137 (half-life 30.2 years) possess specific activities near 3×10^{15} Bq kg⁻¹.

371 Dose calculations integrated three exposure pathways following ICRP methodology. Inhalation dose was
372 computed by time-integrating ground-level air concentrations (Becquerels per cubic metre) over the sim-
373 ulation duration, multiplying by a standard breathing rate (1.2 cubic metres per hour for light activity
374 adults [17]), and applying isotope-specific committed effective dose coefficients (Sieverts per Becquerel in-
375haled). Cloudshine dose was calculated by integrating the three-dimensional concentration field (accounting
376 for contributions from all altitudes) and applying external dose-rate coefficients for photon exposure from an
377 infinite cloud geometry [31]. Groundshine dose utilised time-integrated ground deposition (Becquerels per
378 square metre) multiplied by isotope-specific dose-rate coefficients for external exposure from contaminated
379 ground surfaces, assuming a semi-infinite plane source geometry [31].

380 Total effective dose at each grid location represented the sum of all three pathways across all radionuclides.
381 The dose calculations employed HYSPLIT’s CON2REM utility with the -d1 flag for total dose output,
382 implementing the coefficient database derived from ICRP publications. Dose distributions were computed
383 across the full model domain at 0.25-degree resolution, with particular focus on maximum values over Irish
384 territory and population-weighted averages accounting for settlement patterns. The resulting dose fields
385 enable direct assessment against international and national intervention criteria: the IAEA generic criterion
386 of 100 mSv projected dose over the first 7 days for urgent protective actions (sheltering and evacuation)
387 [20], and Irish national intervention levels of 50 mSv for sheltering and 100 mSv for evacuation [11]. Iodine
388 thyroid blocking (ITB) is recommended when projected thyroid equivalent doses from radioiodine inhalation
389 exceed 50 mSv for children and 500 mSv for adults [11]; the predicted total effective doses in this study
390 remain substantially below these thresholds for existing operational facilities, indicating that ITB would not
391 be warranted for transboundary exposure from distant sites under worst-case conditions.

392 **2.7 Machine Learning for Impact Prediction**

393 The extensive dataset generated through fourteen years of HYSPLIT screening simulations (approximately
394 2.2 million simulations across six nuclear power plants) provided a unique opportunity to develop predic-
395 tive models capable of rapid impact assessment during emergency conditions. Rather than treating the
396 simulation ensemble solely as a worst-case identification tool, this approach maximises the value of the sub-
397 stantial computational investment by extracting machine learning models from the same data. Machine
398 learning methodologies enable extraction of complex, non-linear relationships between meteorological condi-
399 tions, release parameters, and radiological consequences, potentially offering faster-than-real-time predictions
400 to support early decision-making.

401 The eXtreme Gradient Boosting (XGBoost) algorithm was selected as the primary machine learning
402 framework due to its demonstrated superior performance in atmospheric science applications, including at-
403 mospheric chemistry transport model bias correction and air quality index forecasting, where it achieves
404 high predictive accuracy ($R^2 > 0.99$) whilst maintaining interpretability through feature importance analysis
405 [21, 44]. XGBoost implements gradient-boosted decision trees, iteratively constructing an ensemble of weak
406 learners that collectively minimise prediction error. The algorithm incorporates regularisation to prevent
407 overfitting and employs efficient tree-building procedures that scale well to datasets containing millions of
408 examples [26].

409 Feature engineering transformed ERA5 reanalysis meteorological data (accessed via HYSPLIT-compatible
410 ARL-format files) into a comprehensive set of 201 input variables characterising atmospheric conditions rel-
411 evant to long-range transport. For each meteorological scenario, seven summary statistics were computed
412 from the spatial distribution of meteorological fields at the release start time: mean (central tendency),
413 median (robust central estimate insensitive to outliers), variance (dispersion quantifying spatial variability),
414 skewness (asymmetry of distribution, positive indicating tail toward high values), kurtosis (tailedness mea-
415 suring extreme event frequency), minimum (lower bound), and maximum (upper bound). These statistics
416 condense the three-dimensional meteorological fields (longitude, latitude, altitude) at the initial time into
417 scalar features suitable for machine learning whilst preserving essential information about initial atmospheric
418 state for predictive purposes.

419 The statistics were computed for nine core meteorological variables: mean sea level pressure (hPa),
420 surface latent heat flux (W m^{-2}), 2-metre temperature (K), 10-metre zonal wind component (m s^{-1}), 10-
421 metre meridional wind component (m s^{-1}), 1000-metre geopotential height (m), 1000-metre temperature (K),
422 1000-metre zonal wind (m s^{-1}), and 1000-metre meridional wind (m s^{-1}). Surface variables capture boundary
423 layer conditions affecting near-source plume behaviour and surface deposition, whilst 1000-metre variables
424 represent conditions in the lower troposphere where long-range transport primarily occurs. Each variable's
425 seven statistics were calculated over two nested spatial domains: an Ireland-focused domain defined by a
426 high-resolution GeoJSON polygon representing Irish territorial boundaries with precise coastline delineation
427 (bounding extent approximately 10.47°W to 6.01°E, 51.45°N to 55.38°N), and a broader rectangular European
428 domain (10.5°W to 2.0°E, 46.0°N to 56.0°N, approximately 950 km × 1100 km capturing continental influences
429 and large-scale flow patterns). The detailed Ireland polygon ensures spatial aggregation includes only grid
430 cells over Irish land and territorial waters, excluding the Irish Sea, Great Britain, and open Atlantic areas that
431 would be included in a simple bounding box approach. This dual-domain approach yielded 126 meteorological
432 features (9 variables × 7 statistics × 2 domains).

433 Feature naming employed a structured convention: `variable_statistic_domain`, for example `pressure_mean_ireland`
434 represents mean sea level pressure averaged spatially over the Ireland domain at release time, whilst `u1000_kurtosis_europe`
435 denotes kurtosis of 1000-metre zonal wind computed over the European domain at release time. This system-
436 atic naming facilitates interpretation of machine learning feature importance, enabling direct identification
437 of which meteorological conditions (variable type, statistical property, spatial scale) most strongly predict
438 radiological transport to Ireland. Release-specific parameters (release hour [0–23 UTC], release duration [6,
439 24, 48 hours], release height [20, 50, 100 m], day of year [1–366], calendar month [1–12]) contributed an addi-
440 tional 5 features, yielding the final 131-dimensional feature space (126 meteorological + 5 release parameters)
441 used for model training.

442 Site-specific models were developed independently for each nuclear power plant to account for differing
443 transport climatologies and geographical relationships to Ireland. The target variable for classification models
444 was defined as impact occurrence: scenarios with total deposition exceeding zero over Ireland were labelled
445 as positive class (impact), whilst scenarios with zero deposition were labelled as negative class (no impact).

446 This binary classification framework enables rapid screening to identify meteorological conditions favouring
447 transport toward Ireland, providing early warning before detailed consequence calculations are feasible.

448 The training strategy implemented temporal partitioning to respect the chronological nature of meteorological
449 data and potential non-stationarity in atmospheric patterns. Simulations from 2011 through 2023
450 constituted the training set, whilst the complete year 2024 served as an independent hold-out validation set.
451 This temporal split ensures that model performance metrics reflect genuine predictive capability for future
452 conditions rather than overfitting to training data. Time-weighted sampling was incorporated during training,
453 with exponentially greater weight assigned to recent years to account for potential trends in atmospheric
454 circulation patterns.

455 Hyperparameter optimisation employed randomised search over 1000 candidate configurations, exploring
456 ranges of learning rate (0.01–0.3), maximum tree depth (3–12), subsample ratio (0.5–1.0), column subsample
457 ratio (0.5–1.0), gamma parameter for minimum split loss (0–5), L1 and L2 regularisation terms (0–10),
458 minimum child weight (1–10), and scale position weight for class imbalance (0.5–2.0). Each configuration was
459 evaluated through early stopping on the 2024 validation set, with training terminating if classification error
460 failed to improve for 20 consecutive boosting rounds (maximum 2000 rounds). The optimal hyperparameter
461 set for each site-specific model maximised validation accuracy whilst maintaining balanced performance across
462 both impact and no-impact classes.

463 Champion model selection prioritised balanced accuracy (the arithmetic mean of sensitivity and specificity)
464 to avoid models that achieved high overall accuracy by simply predicting the majority class. The final
465 champion models for each site were evaluated using comprehensive performance metrics: accuracy, precision,
466 recall (sensitivity), specificity, F1-score, and balanced accuracy. Feature importance analysis employed XG-
467 Boost’s gain metric, quantifying the cumulative reduction in training loss attributable to each feature across
468 all splits in the ensemble. This analysis identified the meteorological variables exerting greatest influence on
469 impact predictions, providing physical insight into the atmospheric drivers of transboundary transport.

470 The champion models demonstrate validation accuracies ranging from 85.4% (Wylfa) to 92.5% (Sizewell)
471 on the independent 2024 hold-out set. Models for proximal sites (Wylfa, Heysham) exhibited higher sensitivity
472 (0.76–0.78), correctly identifying most true impact events whilst accepting moderate false-positive rates.
473 Models for distant sites (Paluel, Flamanville, Sizewell) achieved higher specificity (0.96–0.97), effectively
474 ruling out non-impact scenarios whilst accepting lower sensitivity. This performance gradient reflects the
475 underlying class imbalance: distant sites impact Ireland less frequently, favouring high-specificity classifiers
476 that minimise false alarms.

477 2.8 Global Sensitivity Analysis

478 Global sensitivity analysis provides quantitative apportionment of output variance to input parameters,
479 identifying which variables constitute primary drivers of uncertainty in radiological consequence predictions
480 [34, 46]. Whilst the worst-case scenario identification addresses the question of when maximum consequences
481 occur, sensitivity analysis addresses the complementary question of why those consequences arise and which
482 parameters require highest measurement accuracy for reliable prediction. This analysis leverages the same
483 extensive simulation ensemble generated for worst-case screening, extracting additional scientific value from
484 the 2.2 million scenario computations rather than requiring separate dedicated sensitivity sampling.

485 The analysis was restricted to the subset of simulations yielding measurable impact on Ireland (non-
486 zero deposition or finite arrival time), focusing on the question of impact severity given that an impact
487 occurs. This restriction is appropriate for emergency response planning, where the primary concern during
488 an actual release is determining consequence magnitude rather than predicting whether any impact will occur
489 (a question more suitable for the machine learning classification models).

490 The methodological framework employed a two-stage pipeline combining Principal Component Analysis
491 (PCA) with Polynomial Chaos Expansion (PCE). The initial challenge arose from the high-dimensional
492 input space: over 360 meteorological summary statistics plus five release parameters constituted potential
493 explanatory variables. Direct application of variance-based sensitivity analysis to this dimensional space
494 would encounter multicollinearity (strong inter-correlation amongst meteorological variables) and the curse
495 of dimensionality (exponential growth in required samples with increasing dimension).

496 Principal Component Analysis transformed the correlated input variables into a reduced set of uncorre-
497 lated principal components through eigenvalue decomposition of the covariance matrix. The PCA model was

498 fitted to retain components explaining 99.9% of original variance, typically reducing dimensionality from 365
499 variables to approximately 50–100 principal components. This transformation eliminates multicollinearity
500 whilst preserving nearly all information content, creating a numerically stable input space for subsequent
501 sensitivity analysis.

502 Polynomial Chaos Expansion constructed a surrogate model approximating the complex HYSPLIT atmo-
503 spheric transport calculations through polynomial basis functions [7, 33]. PCE represents the model output Y
504 as a weighted sum of orthogonal polynomials: $\hat{Y}(\mathbf{X}) = \sum_{k=0}^M c_k \Psi_k(\mathbf{X})$, where \mathbf{X} denotes the vector of princi-
505 pal components, $\{\Psi_k\}$ constitutes the polynomial basis, and c_k represents expansion coefficients determined
506 through least-squares regression. The efficiency of PCE derives from the analytical calculation of Sobol’
507 sensitivity indices directly from expansion coefficients, avoiding the Monte Carlo sampling burden required
508 by other variance decomposition methods. This approach has been successfully applied to atmospheric dis-
509 perssion modelling for nuclear accidents, demonstrating the capability to apportion variance amongst source
510 term and meteorological parameters [13].

511 Total-order Sobol’ indices S_{T_j} quantify the fraction of output variance attributable to principal component
512 X_j including all interaction effects with other components. Components with $S_{T_j} > 0.01$ (contributing more
513 than 1% of variance) were classified as influential. To restore physical interpretability, sensitivity indices were
514 mapped back to original meteorological and release variables using the PCA loading matrix, which specifies
515 how each original variable contributes to each principal component.

516 The analysis was conducted independently for each nuclear power plant and each output metric (total
517 deposition, average air concentration, plume arrival time), yielding 18 sensitivity analyses total. Consistent
518 patterns emerged across sites and metrics: wind-related variables dominated variance contributions, with
519 median west-east wind speed at 10 metres and kurtosis of west-east wind at 1000 metres ranking as the top
520 two influential variables for most scenarios. The kurtosis metric (measuring the degree of extreme events or
521 heavy-tailedness in the wind speed distribution) proved more influential than mean or median wind speed,
522 indicating that infrequent strong-wind events drive the highest-consequence transport scenarios rather than
523 typical conditions.

524 Geopotential height skewness appeared frequently amongst the top three influential variables, particularly
525 for sites requiring longer transport distances. Geopotential height serves as a proxy for atmospheric pressure
526 structure; its skewness quantifies asymmetry in pressure gradient distributions and correlates with frontal
527 passages and synoptic-scale weather systems. The consistent appearance of this variable suggests that worst-
528 case transport scenarios for Ireland frequently coincide with active frontal systems providing both strong
529 advection and precipitation-driven wet deposition.

530 Critically, the sensitivity analysis revealed that parametric release variables (height, duration) contributed
531 negligibly to output variance for impactful events. Total-order Sobol’ indices for release height and duration
532 were consistently below 0.001, indicating that atmospheric conditions dominate consequence severity once a
533 release occurs. This finding contrasts with some aggregated-output sensitivity analyses which identify source
534 term magnitude as the primary driver, but aligns with recent studies demonstrating that meteorological
535 uncertainty typically dominates early-phase predictions for long-range transboundary transport [24]. The
536 distinction arises from the conditional focus on impactful scenarios: for events that impact Ireland, meteorolog-
537 ical conditions constitute the primary driver of consequence magnitude, with release parameters playing
538 secondary roles. Emergency response priorities should therefore emphasise accurate real-time meteorological
539 forecasting and rapid atmospheric model execution over prolonged refinement of source term estimates during
540 the early phase of an event.

Table 2: Worst-Case Radiological Impact Scenarios for Ireland per NPP (2011-2024). For deposition and concentration, ‘Max’ indicates the highest value. For plume arrival time, ‘Min’ indicates the shortest time.

NPP	Metric	Value	Date	Rel. Hour (UTC)	Rel. Height (m)	Rel. Dur. (hr)
Wylfa (closest to Ireland)						
	Max Total Deposition (mass/m ²)	3.19e-08	2024-05-19	21	100	48.0
	Max Air Concentration (mass/m ³)	5.12e-11	2015-10-08	9	50	48.0
	Min Plume Arrival (hours)	3.0	2011-01-07	12	20	24.0
Heysham						
	Max Total Deposition (mass/m ²)	2.32e-08	2016-05-09	9	50	48.0
	Max Air Concentration (mass/m ³)	2.90e-11	2021-02-27	9	50	48.0
	Min Plume Arrival (hours)	3.0	2012-04-25	12	100	24.0
Hinkley						
	Max Total Deposition (mass/m ²)	2.15e-08	2013-10-14	15	100	48.0
	Max Air Concentration (mass/m ³)	5.15e-11	2022-01-12	21	50	48.0
	Min Plume Arrival (hours)	3.0	2012-04-25	3	100	24.0
Sizewell						
	Max Total Deposition (mass/m ²)	6.78e-09	2019-04-22	18	100	48.0
	Max Air Concentration (mass/m ³)	3.38e-12	2012-08-13	15	100	48.0
	Min Plume Arrival (hours)	6.0	2018-03-01	18	50	6.0
Flamanville						
	Max Total Deposition (mass/m ²)	1.96e-08	2015-09-10	3	100	48.0
	Max Air Concentration (mass/m ³)	9.82e-12	2019-08-22	6	50	48.0
	Min Plume Arrival (hours)	6.0	2011-10-24	12	100	24.0
Paluel (most distant)						
	Max Total Deposition (mass/m ²)	1.42e-08	2015-09-10	0	100	48.0
	Max Air Concentration (mass/m ³)	7.37e-12	2019-02-24	3	50	48.0
	Min Plume Arrival (hours)	9.0	2012-04-25	3	50	6.0

541 **3 Results**

542 **3.1 Worst-Case Scenario Identification from 14-Year Study**

543 The systematic screening of 367,200 HYSPLIT simulations per nuclear power plant across the fourteen-year
544 period (2011–2024) identified eighteen critical scenarios representing maximum radiological consequences
545 for Ireland (Table 2). These scenarios span three distinct impact categories: maximum total deposition
546 (governing long-term ground contamination and ingestion pathway dose), maximum average air concentration
547 (governing acute inhalation exposure), and minimum plume arrival time (governing available warning time
548 for protective action implementation). The identified scenarios exhibit substantial diversity in temporal
549 occurrence, release parameters, and meteorological drivers, reflecting the complex interplay between synoptic
550 weather patterns and local atmospheric conditions determining transboundary transport.

551 Temporal analysis reveals that worst-case scenarios concentrate during transitional seasons, with spring
552 (March–May) accounting for seven of eighteen scenarios and autumn (September–October) contributing five
553 scenarios. This seasonal distribution reflects the climatology of atmospheric circulation patterns affecting
554 northwest Europe. Spring months frequently experience strong pressure gradients associated with North
555 Atlantic cyclogenesis, producing easterly or southeasterly flow capable of efficiently transporting material
556 from continental sources toward Ireland. The sole summer scenario (August 2022 for Heysham) represents an
557 anomalous synoptic pattern characterised by an extended period of easterly winds resulting from a persistent
558 high-pressure system over Scandinavia.

559 Release parameter analysis demonstrates that the majority of identified worst-case scenarios (fourteen of
560 eighteen) involve extended release durations of 48 hours, consistent with severe accident sequences charac-
561 terised by prolonged containment degradation rather than catastrophic early failure. Release heights cluster
562 at the extremes of the examined range: eight scenarios feature releases at 100 metres above ground level (rep-
563 resenting unfiltered stack releases or thermal plume rise from heat-driven releases), whilst seven scenarios
564 employ 50-metre releases, and only three utilise the lowest examined release height of 20 metres. The pref-
565 erence for elevated releases in maximum deposition scenarios reflects atmospheric boundary layer dynamics:
566 material released at greater heights penetrates above the surface layer, reducing near-source dry deposition
567 and enabling longer-range transport before gravitational settling and wet scavenging remove material from
568 the atmosphere.

569 Geographically, the most severe deposition scenarios (exceeding $2.0 \times 10^{-8} \text{ kg m}^{-2}$ in unit-release simu-
570 lations) originate from the nearest facilities. Wylfa achieves the maximum value of $3.19 \times 10^{-8} \text{ kg m}^{-2}$ due
571 to its proximity to the Irish coast and relatively short transport distance permitting high atmospheric con-
572 centrations to persist until landfall. Heysham and Hinkley Point C produce comparable deposition values
573 of 2.32×10^{-8} and $2.15 \times 10^{-8} \text{ kg m}^{-2}$ respectively, reflecting their similar distances and orientations rel-
574 ative to Ireland. More distant continental sources (Flamanville, Paluel) yield lower maximum deposition
575 despite comparable or higher air concentrations, as extended transport times increase atmospheric dilution
576 and deposition losses en route.

577 The maximum air concentration scenarios predominantly occur during summer and early autumn (August,
578 September, October), suggesting a meteorological mechanism distinct from the spring-dominated deposition
579 scenarios. High air concentrations require minimal wet deposition combined with direct, rapid transport
580 maintaining plume coherence. The February 2021 scenario for Heysham represents an unusual winter event
581 characterised by dry, stable anticyclonic conditions permitting efficient long-range transport with minimal
582 precipitation scavenging.

583 Minimum plume arrival scenarios cluster tightly in temporal space: three of six scenarios occur on 25 April
584 2012, indicating a particularly favourable synoptic pattern characterised by strong westerly to northwesterly
585 flow across the domain. Arrival times range from 3 hours (Heysham, Hinkley Point C, Wylfa) to 9 hours
586 (Paluel), with shorter arrival times associated with proximal sources and sustained high wind speeds. These
587 rapid-arrival scenarios pose the greatest challenge for emergency response systems, providing minimal time
588 for monitoring network activation, plume characterisation, and protective action recommendation before
589 material reaches Irish airspace.

590 The seasonal distribution of impact events provides additional context for emergency preparedness plan-
591 ning. Analysis of simulated plume intersections with Ireland, aggregated by month over the 14-year period,
592 reveals consistent spring maxima (March–May) and summer minima (July–August) across all facilities. Dur-
593 ing spring, elevated intersection frequencies across most facilities reflect the climatological prevalence of east-

594 early and southeasterly flow patterns. Wylfa and Heysham exhibit the highest overall intersection frequencies
 595 due to their proximity to the Irish coast and prevailing wind climatology. This seasonal signal reinforces the
 596 temporal concentration of worst-case scenarios during transitional seasons identified in Table 2.

597 Relative seasonal risk, quantified through monthly ranking of intersection frequency, identifies April as
 598 the highest-risk month across all six nuclear facilities. Conversely, July and August consistently rank as the
 599 months with the lowest transport probability. This pattern provides actionable intelligence for emergency pre-
 600 paredness resource allocation, suggesting enhanced monitoring and response readiness during spring months
 601 when atmospheric transport toward Ireland is most probable (spring is also the period when livestock are on
 602 pasture and crops are actively growing, compounding potential food chain contamination concerns).

603 3.2 FLEXPART Verification and Radiological Dose Assessment: Heysham Case 604 Study

605 Independent verification of worst-case scenarios through multi-model consensus constitutes essential best
 606 practice for emergency preparedness applications. Atmospheric dispersion predictions contain inherent un-
 607 certainties arising from physical parameterisations, numerical implementations, and meteorological input
 608 processing [12]. The employment of two independent Lagrangian particle dispersion models—HYSPLIT and
 609 FLEXPART—provides confidence that identified transport patterns and radiological consequences represent
 610 genuine atmospheric phenomena rather than model-specific numerical artefacts.

611 The Heysham maximum deposition scenario (9 May 2016, 09:00 UTC release) was selected for detailed
 612 presentation as it represents the most comprehensive source term specification amongst all examined facili-
 613 ties, incorporating nineteen radionuclides including actinides characteristic of Advanced Gas-Cooled Reactor
 614 accidents. The realistic AGR source term includes volatile fission products (I-131, I-132, I-133, I-134, I-
 615 135, Te-132, Cs-134, Cs-137), intermediate-volatility species (Sr-89, Sr-90, Ru-103, Ru-106, Ce-144), and
 616 low-volatility actinides (Pu-238, Pu-239, Pu-240, Pu-241, Am-241, Cm-242, Cm-244). This multi-isotope
 617 complexity provides rigorous testing of model capabilities to handle diverse physical and chemical behaviours
 618 spanning seven orders of magnitude in half-life (52.5 minutes for I-134 to 24,110 years for Pu-239) and four
 619 orders of magnitude in specific activity.

620 Both models predict radiological doses dramatically below international and national intervention thresh-
 621 olds. Table 3 presents quantitative dose assessments for the whole of Ireland, revealing maximum total
 622 effective doses of 0.00671 mSv (HYSPLIT) and 0.000489 mSv (FLEXPART) for 36-hour cumulative expo-
 623 sure. Even if such release conditions were sustained for the full 7-day assessment period used in international
 624 criteria, extrapolated doses would remain approximately three to four orders of magnitude below the IAEA
 625 generic criterion of 100 mSv [20] and well below the Irish national sheltering threshold of 50 mSv [11]. To
 626 contextualise these magnitudes: even the higher HYSPLIT prediction is comparable to approximately 14
 627 hours of natural background radiation exposure in Ireland ($4.2 \text{ mSv/year} \div 8760 \text{ hours/year} \times 14 \text{ hours} \approx$
 628 0.0067 mSv), approximately 13% of a transatlantic flight dose (0.05 mSv), or 7% of a chest X-ray dose (0.1
 629 mSv).

629 Table 3: Dose Assessment for Whole of Ireland: Heysham Maximum Deposition Scenario

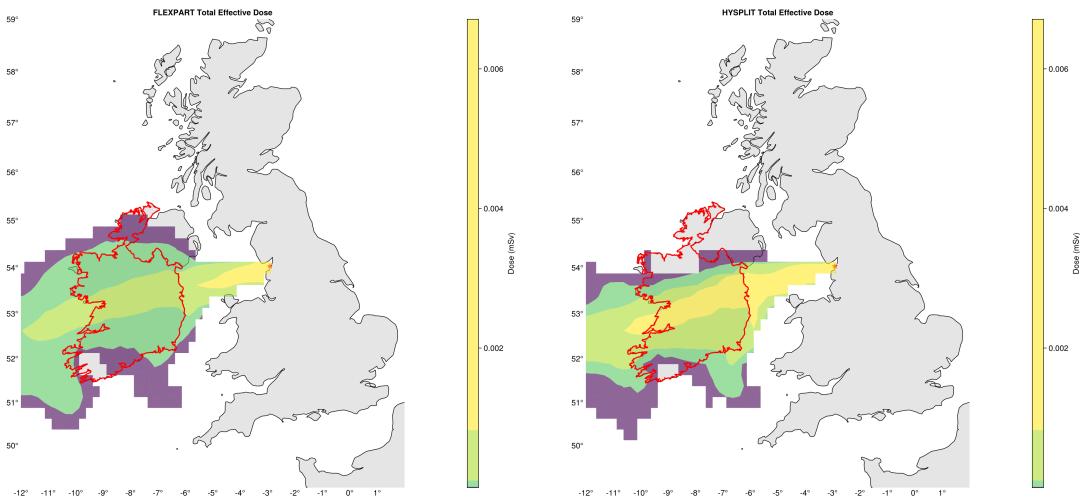
Dose Component	FLEXPART	HYSPLIT
Maximum Gamma Dose Rate (mSv/h)		
Cloudshine + Groundshine	1.35e-5	0.000186
Total Effective Dose (mSv, 36-hour exposure)		
Inhalation (internal)	4.25e-6	2.18e-5
Cloudshine (external)	2.67e-8	9.25e-8
Groundshine (external)	0.000485	0.00669
Total	0.000489	0.00671

630 Groundshine dominates the total effective dose, contributing 96–99% depending on the model, with in-
 631 inhalation providing 1–3% and cloudshine contributing negligibly (<0.01%). This pathway distribution reflects
 632 the fundamental physics of radiological exposure from depositing species: external gamma radiation from
 633 ground-deposited radionuclides constitutes the primary exposure mechanism for the examined scenario and

634 source term. The dominance of groundshine has practical implications for protective action selection: sheltering (which provides minimal groundshine shielding) offers limited dose reduction compared to evacuation
 635 or indoor stay with closed windows. However, at the predicted dose levels, neither sheltering nor evacuation
 636 would be warranted—the assessment value lies in providing accurate information to prevent unnecessary
 637 public anxiety and demonstrate the effectiveness of existing safety standards.

638 Inter-model comparison reveals divergence in predicted dose magnitudes, with HYSPLIT predicting total
 639 doses approximately 14 times higher than FLEXPART. This factor-of-fourteen difference, whilst notable in
 640 relative terms, remains inconsequential in absolute radiological terms as both predictions fall orders of magni-
 641 tude below any level of health concern. Both models employed identical element-specific physics parameters
 642 (dry deposition velocities, wet scavenging coefficients, material densities, and radioactive half-lives) to en-
 643 sure consistent treatment of the nineteen radionuclides spanning noble gases, volatile fission products, and
 644 refractory actinides. The observed divergence therefore reflects differences in wet deposition implementa-
 645 tion rather than input parameters: HYSPLIT employs scavenging coefficients applied uniformly across precipita-
 646 tion intensity ranges, whilst FLEXPART implements more sophisticated precipitation-dependent scavenging
 647 with separate treatments for in-cloud and below-cloud processes. For the examined scenario, characterised
 648 by moderate precipitation during transport, these implementation differences yield the observed dose ratio
 649 whilst maintaining consensus on the fundamental conclusion of negligible radiological impact.

650 Figures 1 through 4 present spatial distributions of dose components, revealing consistent geographical
 651 patterns between models despite magnitude differences. Both models predict peak doses concentrated along
 652 Ireland’s eastern coastline in the Dublin-Drogheda corridor, reflecting the direct westward transport pathway
 653 from Heysham across the Irish Sea. The plume footprint extends inland approximately 50–80 kilometres be-
 654 fore atmospheric dilution and deposition processes reduce concentrations below detection thresholds. Vertical
 655 cross-sections (not shown) confirm that material remains predominantly within the atmospheric boundary
 656 layer (below 1000 metres altitude) throughout transport, consistent with the moderate release height (50
 657 metres) and stable to neutral atmospheric conditions during the event.



658 Figure 1: Total effective dose comparison for Heysham maximum deposition scenario (9 May 2016, 09:00
 659 UTC). HYSPLIT (left panel) predicts maximum 36-hour cumulative dose of 0.00671 mSv whilst FLEXPART
 660 (right panel) shows 0.000489 mSv, both concentrated along Ireland. The factor-of-14 inter-model difference
 661 is radiologically inconsequential as both predictions fall approximately four to five orders of magnitude below
 662 intervention thresholds, comparable to natural background radiation exposure from routine daily activities.

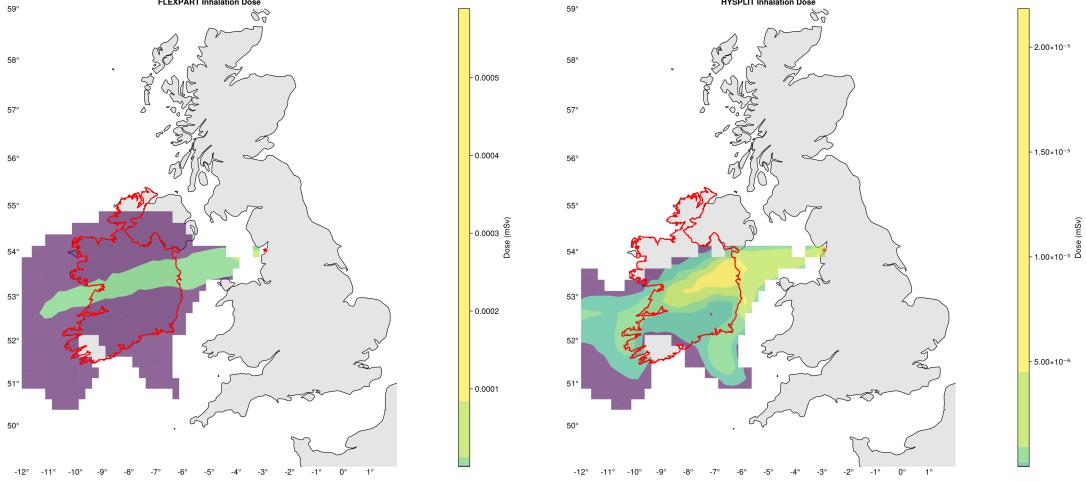


Figure 2: Inhalation dose comparison for Heysham maximum deposition scenario. Maximum inhalation contributions reach 2.18×10^{-5} mSv (HYSPLIT) and 4.25×10^{-6} mSv (FLEXPART), representing less than 1% of total dose. The nineteen-isotope source term includes high-inhalation-hazard actinides (Pu-239, Am-241), but extremely low atmospheric concentrations render inhalation pathway contribution minimal compared to external exposure from deposited material.

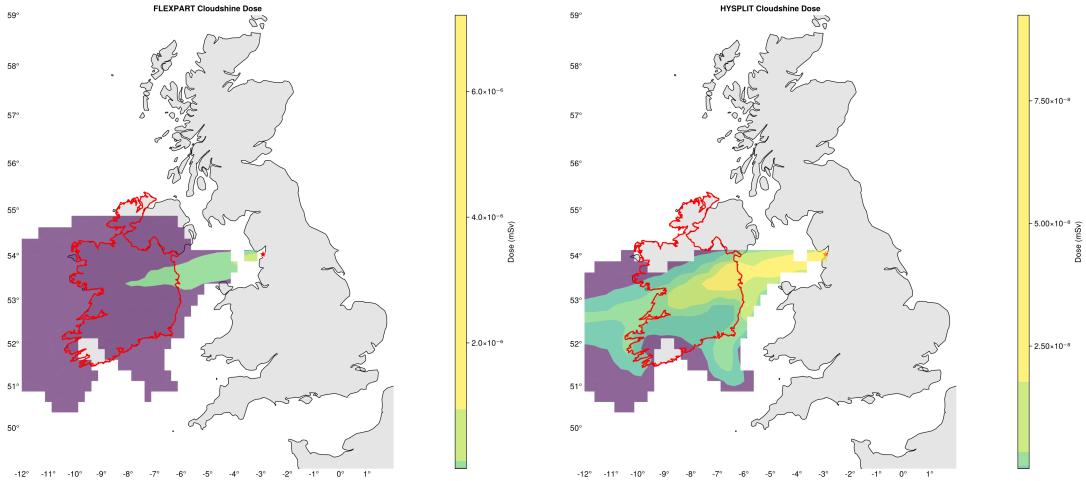


Figure 3: Cloudshine dose comparison for Heysham maximum deposition scenario. Both models predict negligible cloudshine contributions ($\sim 3 \times 10^{-8}$ mSv), four orders of magnitude below total dose. The multi-isotope release includes strong gamma emitters (I-134, I-132, Te-132) but low airborne concentrations during plume passage yield minimal external dose from atmospheric activity.

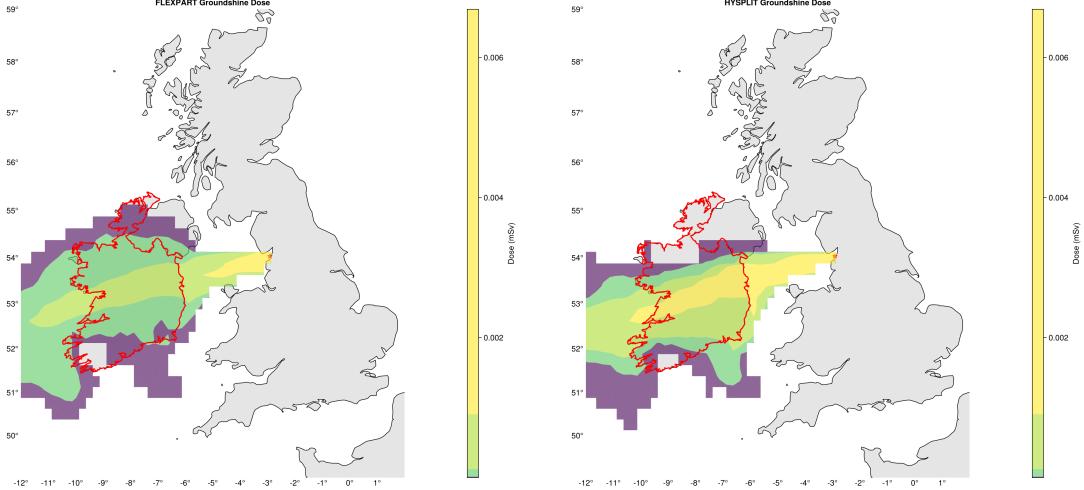


Figure 4: Groundshine dose comparison for Heysham maximum deposition scenario. External gamma radiation from deposited radionuclides dominates total dose, with maximum values of 0.00669 mSv (HYSPLIT) and 0.000485 mSv (FLEXPART). Spatial patterns exhibit strong model consensus despite magnitude differences, both predicting concentrated deposition in coastal regions where the plume made initial landfall over Ireland. The nineteen-isotope mixture yields groundshine contributions weighted by deposition density and isotope-specific external dose coefficients, with short-lived high-activity iodine isotopes (I-132, I-134) providing disproportionate contribution during the initial 36-hour assessment period.

659 The verification analysis demonstrates robust multi-model consensus on the fundamental finding: even for
 660 meteorologically worst-case scenarios identified through fourteen years of systematic screening, realistic severe
 661 accident source terms produce radiological doses to Ireland that are dramatically below any level warranting
 662 protective actions. The remaining seventeen worst-case scenarios (other facilities and impact categories)
 663 exhibit comparable dose magnitudes and model agreement characteristics; detailed dose assessments for all
 664 scenarios are provided in Appendix A for completeness. This quantitative evidence provides substantial
 665 assurance for emergency preparedness planning whilst identifying that meteorological forecasting accuracy
 666 and rapid atmospheric transport model execution constitute higher priorities than protracted source term
 667 refinement during early-phase emergency response.

668 3.3 Machine Learning for Rapid Impact Prediction

669 Site-specific XGBoost machine learning models were developed to enable rapid prediction of atmospheric
 670 impact occurrence on Ireland based on meteorological conditions and release parameters, as described in
 671 Section 2.7. The champion models, optimised through randomised search over 1000 hyperparameter configu-
 672 rations and validated on the independent 2024 hold-out dataset, demonstrate robust predictive performance
 673 suitable for operational deployment in emergency decision support systems.

674 Table 4 presents comprehensive performance metrics for all six site-specific models. Overall accuracies
 675 range from 85.4% (Wylfa) to 92.5% (Sizewell), with balanced accuracies varying from 73.9% (Paluel) to
 676 84.9% (Heysham). These metrics reflect the models' ability to generalise to unseen meteorological conditions
 677 whilst handling the inherent class imbalances present in the validation dataset (Table 5).

Table 4: Summary of ML Model Performance on the 2024 Validation Dataset (Target: Impact if Total Deposition > 0). Results from models after 1000 hyperparameter optimisation trials.

Site	Accuracy	Precision	Recall (Sens.)	Specificity	F1-Score	Bal. Acc.
Wylfa	0.8542	0.7915	0.7798	0.8930	0.7856	0.8364
Paluel	0.9074	0.7485	0.5048	0.9725	0.6029	0.7387
Flamanville	0.9000	0.7603	0.5805	0.9636	0.6583	0.7720
Hinkley	0.8676	0.7613	0.6576	0.9344	0.7056	0.7960
Sizewell	0.9251	0.7770	0.5974	0.9743	0.6755	0.7858
Heysham	0.8950	0.8058	0.7582	0.9399	0.7813	0.8491

Table 5: Confusion Matrices and Actual Class Distributions (2024 Validation, Target: Total Deposition > 0). PN: Pred. No Deposition; PP: Pred. Deposition; AN: Actual No Deposition; AP: Actual Deposition. Percentages refer to the proportion of AN or AP in the validation set.

Wylfa			Paluel			
	PN (Count)	PP (Count)	Actual N (Neg%) / P (Pos%)	PN (Count)	PP (Count)	
AN	15430	1849	17279 (65.75%)	21998	621	22619 (86.07%)
AP	1982	7019	9001 (34.25%)	1813	1848	3661 (13.93%)
Flamanville			Hinkley			
	PN (Count)	PP (Count)	Actual N (Neg%) / P (Pos%)	PN (Count)	PP (Count)	
AN	21122	798	21920 (83.41%)	18629	1308	19937 (75.86%)
AP	1829	2531	4360 (16.59%)	2172	4171	6343 (24.14%)
Sizewell			Heysham			
	PN (Count)	PP (Count)	Actual N (Neg%) / P (Pos%)	PN (Count)	PP (Count)	
AN	22262	588	22850 (86.95%)	18591	1188	19779 (75.26%)
AP	1381	2049	3430 (13.05%)	1572	4929	6501 (24.74%)

The performance metrics reveal distinct characteristics depending on the geographical relationship between nuclear power plants and Ireland. Models for proximal sites (Wylfa, Heysham) exhibit higher recall values (0.78–0.76), correctly identifying most true impact events whilst accepting moderate false-positive rates. This characteristic is critical for effective early warning systems. Models for distant sites (Paluel, Flamanville, Sizewell) achieve higher specificity (0.96–0.97), effectively ruling out non-impact scenarios whilst accepting lower sensitivity. This performance gradient reflects the underlying class imbalance: distant facilities impact Ireland less frequently, favouring high-specificity classifiers that minimise false alarms.

Feature importance analysis using XGBoost’s gain metric reveals that the most influential predictors vary by site, reflecting differing geographical relationships and dominant meteorological transport pathways. Models for proximal sites (Wylfa, Heysham, Hinkley) assign higher importance to local meteorological conditions over Ireland, such as minimum 1000 m geopotential height (`HGTS1000_Ireland_min`), surface pressure (`PRSS_Ireland_min`), and 10-metre wind components (`U10M_Ireland_min`). For more distant sites (Paluel, Flamanville, Sizewell), variables characterising broader atmospheric patterns, particularly wind field variance and geopotential height minima (e.g., `VWND1000_overall_variance`, `HGTS1000_overall_min`), play more dominant roles, reflecting the importance of large-scale transport patterns. Tables 6 and 7 present detailed feature importance rankings.

Table 6: Top 3 Most Important Features by Site (XGBoost Gain score), based on “Total Deposition > 0 ” Target (1000-trial models).

Site	Rank 1 Feature (Gain)	Rank 2 Feature (Gain)	Rank 3 Feature (Gain)
Wylfa	HGTS1000_Ireland_min (6098.3)	WWND1000_Ireland_variance (5824.1)	PRSS_Ireland_min (4907.5)
Paluel	VWND1000_overall_variance (3709.7)	VWND1000_overall_min (996.7)	HGTS1000_overall_min (803.0)
Flamanville	VWND1000_overall_variance (2625.3)	VWND1000_Ireland_variance (1614.2)	VWND1000_overall_max (1226.7)
Hinkley	VWND1000_Ireland_variance (4340.3)	RELH1000_Ireland_variance (736.3)	VWND1000_Ireland_variance (463.4)
Sizewell	HGTS1000_overall_min (6537.7)	VWND1000_overall_variance (3209.7)	U10M_Ireland_min (1604.2)
Heysham	HGTS1000_Ireland_min (9639.9)	U10M_Ireland_min (6506.4)	PRSS_Ireland_min (4929.5)

Table 7: Top 10 Features by Summed Total Gain Score Across All Six Site Models (“Total Deposition > 0 ” Target, 1000-trial models).

Feature Name	Total Gain	Models Present In
HGTS1000_Ireland_min	15738.2	2
PRSS_Ireland_min	11156.3	4
VWND1000_overall_variance	9544.7	3
U10M_Ireland_min	8969.1	3
VWND1000_Ireland_variance	8665.5	4
HGTS1000_overall_min	8556.9	4
WWND1000_Ireland_variance	8357.5	4
VWND1000_Ireland_min	4221.9	2
VWND1000_overall_min	3060.1	4
RELH1000_Ireland_variance	2595.4	2

694 The consistent high ranking of HGTS1000_Ireland_min (minimum geopotential height at 1000 m over Ire-
 695 land) as the top predictor for multiple facilities (Table 7) reveals a counterintuitive but physically meaningful
 696 result: the synoptic-scale pressure pattern over Ireland is more predictive than instantaneous wind direction
 697 towards Ireland. Low geopotential height over Ireland indicates a surface low-pressure system positioned over
 698 or near Ireland, which acts as a meteorological “sink” drawing air masses from the east and southeast (where
 699 the nuclear facilities are located) through large-scale cyclonic circulation. This synoptic pattern is superior to
 700 instantaneous wind metrics (U10M, VWND, WWND) because it captures the persistent large-scale forcing
 701 that maintains favourable transport conditions over the 24–48 hour transport period, rather than merely
 702 representing a snapshot of wind at release time. A low-pressure system over Ireland implies sustained east-
 703 erly or southeasterly geostrophic flow ahead of the approaching system, often associated with frontal activity
 704 that both advects the plume westward and enhances wet deposition through precipitation. The machine
 705 learning model has thus identified the meteorological driver (pressure pattern) rather than the symptom
 706 (instantaneous wind), demonstrating that worst-case transport to Ireland is fundamentally determined by
 707 synoptic-scale circulation patterns rather than local wind fluctuations. The convergence between machine
 708 learning feature importance and global sensitivity analysis findings (Section 3.4) strengthens confidence in
 709 the identification of critical atmospheric parameters for emergency response prioritisation.

710 The demonstrated ability to generalise to unseen 2024 validation data suggests robustness to inter-annual
 711 meteorological variability, though periodic retraining with updated meteorological data would be advisable
 712 to account for potential climate-driven shifts in atmospheric circulation patterns. The predictive capabili-
 713 ty is particularly noteworthy: models achieve 85–93% accuracy in forecasting 48-hour radiological trans-
 714 port outcomes using solely initial meteorological conditions at release time, without requiring knowledge
 715 of atmospheric evolution during the transport period. This demonstrates that initial large-scale synoptic
 716 meteorological patterns contain sufficient information to determine consequence severity, validating the op-

717 erational utility of the approach for real-time emergency response. Operational deployment could support
718 rapid screening of meteorological forecast ensembles to identify high-probability transport scenarios toward
719 Ireland, enabling pre-positioning of monitoring resources and alert protocols during the early phase of nuclear
720 emergencies based on readily available weather forecasts.

721 3.4 Global Sensitivity Analysis

722 Global sensitivity analysis employing the PCA-PCE pipeline (Section 2.8) quantified the relative importance
 723 of meteorological and release parameters in governing consequence severity for events impacting Ireland.
 724 Application of this methodology to each nuclear power plant across four output metrics (total deposition,
 725 average air concentration, plume arrival time) revealed consistent patterns in the drivers of uncertainty.

726 The analysis demonstrated that release parameters (height and duration) contributed negligibly to output
 727 variance, with total-order Sobol' indices consistently below 0.001 across all scenarios. This finding indicates
 728 that for impactful events, consequence severity is governed predominantly by meteorological conditions rather
 729 than release characteristics, validating the unit-release screening approach and emphasising the importance
 730 of accurate meteorological forecasting for emergency response.

731 Wind-related variables dominated variance contributions across all facilities and output metrics. The
 732 kurtosis of west-east wind at 1000 m emerged as the most influential parameter for numerous scenarios,
 733 indicating that extreme wind events drive highest-consequence transport rather than typical conditions.
 734 Median wind speed at 10 m ranked as the second most influential variable for several sites, demonstrating the
 735 importance of persistent surface winds in governing plume transport. Geopotential height skewness appeared
 736 frequently amongst the top three influential variables, particularly for facilities requiring longer transport
 737 distances (Flamanville, Paluel, Sizewell), suggesting that worst-case transport scenarios often coincide with
 738 active frontal systems providing strong advection and enhanced wet deposition.

739 Table 8 presents the three most sensitive input variables for each nuclear power plant and output metric,
 740 revealing site-specific patterns reflecting differing transport climatologies and geographical relationships to
 741 Ireland.

Table 8: Top 3 Most Sensitive Input Variables per NPP and Output Metric, determined by PCE-based Global Sensitivity Analysis.

NPP	Output Metric	Rank 1 (Most Sensitive)	Rank 2	Rank 3
Wylfa				
	Total Deposition	UWND10_median_overall	UWND1000_kurtosis_overall	HGTS10_skewness_Ireland
	Avg. Air Conc.	UWND10_median_overall	UWND1000_kurtosis_overall	UWND1000_kurtosis_Ireland
	Plume Arrival	UWND1000_kurtosis_overall	UWND10_kurtosis_overall	UWND10_median_overall
Heysham				
	Total Deposition	UWND10_median_overall	UWND1000_kurtosis_overall	UWND1000_kurtosis_Ireland
	Avg. Air Conc.	UWND1000_kurtosis_Ireland	UWND10_median_overall	UWND1000_kurtosis_overall
	Plume Arrival	UWND1000_kurtosis_overall	UWND10_median_overall	UWND10_kurtosis_overall
Hinkley				
	Total Deposition	UWND1000_kurtosis_overall	UWND10_median_overall	HGTS10_skewness_overall
	Avg. Air Conc.	UWND1000_kurtosis_overall	HGTS10_skewness_overall	HGTS10_skewness_Ireland
	Plume Arrival	UWND1000_kurtosis_overall	HGTS10_skewness_overall	UWND10_median_overall
Flamanville				
	Total Deposition	UWND1000_kurtosis_Ireland	UWND1000_kurtosis_overall	WWND100_min_overall
	Avg. Air Conc.	UWND1000_kurtosis_overall	HGTS10_skewness_overall	UWND1000_kurtosis_Ireland
	Plume Arrival	UWND1000_kurtosis_overall	HGTS10_skewness_Ireland	HGTS10_skewness_overall
Paluel				
	Total Deposition	UWND10_median_overall	UWND1000_kurtosis_overall	UWND10_kurtosis_overall
	Avg. Air Conc.	UWND10_kurtosis_overall	UWND1000_kurtosis_overall	WWND100_min_overall
	Plume Arrival	UWND10_median_overall	UWND1000_kurtosis_overall	HGTS10_skewness_overall
Sizewell				
	Total Deposition	UWND10_median_overall	UWND1000_kurtosis_overall	HGTS10_skewness_overall
	Avg. Air Conc.	UWND10_median_overall	UWND1000_kurtosis_overall	HGTS10_skewness_overall
	Plume Arrival	UWND10_median_overall	UWND1000_kurtosis_overall	HGTS10_skewness_overall

742 Wind-related variables emerge as primary sensitivity drivers, though not as a single unified factor but
743 rather as distinct phenomena at different atmospheric levels. Kurtosis of 1000-metre zonal wind (`u1000_kurtosis`)
744 frequently ranks as the most sensitive variable across multiple facilities, indicating that extreme or anomalous
745 wind events in upper-level steering flow exert dominant control on transport efficiency. Median 10-metre wind
746 speed (`u10_median`) consistently appears amongst top-ranked variables, reflecting the importance of persistent
747 near-surface winds for plume advection during the extended 48-hour transport period. Geopotential
748 height skewness at 1000 metres (`z1000_skewness`) provides additional explanatory power: as a proxy for at-
749 mospheric pressure structure, its skewness quantifies asymmetry in pressure gradient distributions, with high
750 values indicating passage of frontal systems that drive both enhanced advection and precipitation-induced
751 wet deposition.

752 The operational implication for emergency response is clear: accurate forecasting of consequence severity
753 requires precise characterisation of specific meteorological parameters rather than general synoptic descrip-
754 tions. The ranked sensitivity hierarchy provides quantitative prioritisation for meteorological monitoring and
755 ensemble forecasting during radiological emergencies, enabling targeted allocation of limited computational
756 resources toward variables demonstrably controlling transport outcomes.

757 The Global Sensitivity Analysis and the machine learning models detailed in Section 3.3 address funda-
758 mentally different questions regarding the dataset and are therefore complementary rather than redundant.

759 The XGBoost classifier was designed with predictive objectives, addressing the binary question of whether
760 an impact will occur in Ireland given initial conditions. Feature importance metrics derived from the trained
761 XGBoost model (see Table 6) provide insights into variables the *model* utilised for classification tasks. How-
762 ever, as this represents an indirect measure of sensitivity derived from a model with approximately 85-90%
763 accuracy, it serves as an indicator rather than formal quantification of underlying system sensitivities.

764 The PCE-based GSA addresses explanatory objectives, rigorously quantifying the contribution of each
765 input variable to output *variance* given that an impact has occurred. Through calculation of Sobol' indices,
766 the GSA directly apportions uncertainty in outputs (e.g., total deposition, arrival time) back to initial
767 meteorological and release parameters, providing direct, model-agnostic measures of outcome sensitivity to
768 each input variable across the complete range of impactful scenarios.

769 The analyses therefore address distinct questions essential for emergency response. The machine learning
770 classifier determines impact probability, whilst the Global Sensitivity Analysis quantifies factors determining
771 impact severity given occurrence. The consistency of findings between both methodologies, with wind and
772 pressure-related variables demonstrating dominance in both cases, significantly increases confidence in the
773 overall conclusions. The GSA provides formal quantification of key uncertainty drivers for impactful events,
774 whilst the ML model serves as a prognostic tool for rapid impact prediction.

775 **4 Discussion**

776 This study presents the most comprehensive assessment to date of potential transboundary radiological trans-
777 port to Ireland from proximal nuclear power plants, integrating systematic worst-case scenario identification,
778 independent multi-model verification, machine learning for rapid prediction, and global sensitivity analysis.
779 The analysis reveals a critical distinction based on facility proximity: Heysham, despite its complex 19-isotope
780 AGR source term, produces negligible radiological impact on Ireland even under meteorologically worst-case
781 conditions (doses < 0.01 mSv, comparable to hours of natural background radiation). More distant opera-
782 tional facilities (Sizewell B, Flamanville, Paluel) show low but measurable doses (0.1–4.6 mSv depending on
783 scenario and model), remaining well below protective action thresholds. Hinkley Point C, currently under con-
784 struction, exhibits elevated consequences reflecting its coastal English location but remains below protective
785 action thresholds (0.3–8.5 mSv depending on atmospheric dispersion model). However, the cancelled Wylfa
786 Newydd gigawatt-scale project presents fundamentally different risk: maximum deposition scenarios predict
787 4.5–20.7 mSv total effective dose over 36 hours, with FLEXPART calculations approaching the Irish national
788 sheltering threshold of 50 mSv [11]. Importantly, the Wylfa site is now proposed for small modular reactor
789 (SMR) deployment rather than the gigawatt-scale facility originally planned; the Rolls Royce SMR (470 MWe,
790 approximately 1400 MWth) would have a source term roughly three times smaller than the 4524 MWth EPR
791 modelled in this study, meaning our Wylfa analysis represents a deliberately conservative worst-case scenario
792 that substantially overestimates consequences relative to current development proposals. Crucially, the cur-
793 rent IAEA generic criterion for urgent protective actions (sheltering and evacuation) is 100 mSv projected
794 dose over the first 7 days [20]—a unified threshold that superseded the earlier separate intervention levels
795 of 10 mSv for sheltering and 50 mSv for evacuation. Whilst the 36-hour simulated doses remain below both
796 thresholds, extrapolation to 7-day cumulative exposure under sustained release conditions could approach or
797 exceed Irish intervention levels. The factor-of-~5 inter-model disagreement for Wylfa (FLEXPART 20.7 mSv
798 vs HYSPLIT 4.5 mSv) reflects expected structural uncertainty in near-source dispersion predictions where
799 small-scale meteorological features and deposition parameterisations exert dominant influence. The conver-
800 gent conclusion remains clear: were Wylfa operational, Ireland would require robust emergency preparedness
801 capabilities for specific accident-meteorology combinations, particularly scenarios involving sustained releases
802 coinciding with persistent easterly flow. These findings carry distinct implications for existing facilities (low
803 transboundary concern well below intervention thresholds) versus potential future coastal developments in
804 extreme proximity to Ireland (genuine protective action requirements).

805 **4.1 Methodological Advances and International Benchmarking**

806 This assessment advances beyond previous Irish and international radiological impact studies through several
807 key methodological innovations. The fourteen-year systematic screening (2011–2024) comprising 2.2 million
808 HYSPLIT simulations represents substantial temporal coverage, building upon previous assessments such as
809 the RPII (2013) study which employed ten years of meteorological data [27], and complementing more recent
810 evaluations [10, 22]. This extensive temporal baseline ensures that identified worst-case scenarios represent
811 genuinely extreme meteorological conditions rather than artefacts of limited sampling, capturing the full
812 variability of North Atlantic Oscillation phases and synoptic-scale weather patterns affecting transboundary
813 transport to Ireland.

814 The dual-model verification approach employed in this study—wherein all eighteen worst-case scenarios
815 were independently simulated using both HYSPLIT and FLEXPART with identical ERA5 meteorological
816 forcing—provides robustness beyond typical single-model or ensemble-averaged assessments. Inter-model
817 variability observed in this study (factor of 1–10 for most scenarios, factor of ~5 for Wylfa maximum deposi-
818 tion) represents expected and scientifically defensible structural uncertainty rather than model failure. Recent
819 multi-model intercomparisons of ^{137}Cs dispersion from Fukushima demonstrated that structural model un-
820 certainties remain significant even when models employ identical input data, with inter-model spread often
821 exceeding factor-of-ten for individual grid cells [35]. Our use of identical meteorological forcing eliminates
822 meteorological uncertainty as a confounding variable, isolating differences attributable to physical parame-
823 terisations (turbulence schemes, deposition algorithms, convection representations). The value of dual-model
824 verification lies not in achieving perfect numerical agreement—an unrealistic expectation given the complex-
825 ity of atmospheric turbulence and microphysical deposition processes—but rather in bracketing plausible

826 consequence ranges and identifying consensus conclusions. For Heysham, both models converge on negligible
827 radiological impact (< 0.01 mSv). For more distant facilities (Flamanville, Paluel, Sizewell), doses remain
828 low (0.1–4.6 mSv) and well below intervention thresholds. For Hinkley Point C, both models agree on sub-
829 threshold consequences (0.3–8.5 mSv). For Wylfa, whilst quantitative predictions differ by factor of ~ 5 , both
830 models identify this facility as uniquely concerning, with 36-hour doses approaching levels that could exceed
831 Irish intervention thresholds under sustained release conditions. This convergent risk stratification—negligible
832 for Heysham, low for distant facilities, elevated but sub-threshold for Hinkley, threshold-approaching for
833 Wylfa—provides robust emergency planning guidance despite inter-model quantitative differences.

834 The integration of machine learning (XGBoost) and global sensitivity analysis (PCE/Sobol) methodolo-
835 gies distinguishes this assessment from conventional dispersion modelling studies. Crucially, both analyses
836 leverage the same 2.2 million simulation ensemble originally generated for worst-case identification, demon-
837 strating efficient use of computational resources by extracting multiple complementary insights from a single
838 large-scale modelling campaign. Whilst XGBoost has demonstrated exceptional performance in atmospheric
839 chemistry and air quality applications [21, 44], its application to radiological emergency preparedness for
840 rapid worst-case prediction represents a methodological advancement. The validation accuracies of 85–93%
841 achieved across six nuclear power plants, combined with rigorous temporal validation (2024 hold-out), es-
842 tablish the feasibility of faster-than-real-time impact prediction to support early decision-making during
843 emergencies. Similarly, the global sensitivity analysis, employing PCA for dimensionality reduction followed
844 by PCE surrogate modelling, systematically quantified parameter importance across 365-dimensional input
845 spaces—an analysis scale rarely achieved in nuclear emergency preparedness studies. The finding that mete-
846 orological uncertainty dominates over source term uncertainty for long-range transboundary transport aligns
847 with recent studies [24], providing clear guidance for emergency response resource prioritisation.

848 International comparisons of similar regional assessments confirm the novelty of this study’s comprehensive
849 approach. Whilst advanced reactor emergency planning zone determinations have compared Lagrangian
850 versus Gaussian dispersion models [47], and operational systems like RODOS employ ensemble meteorological
851 forecasting [25], no published study to our knowledge has integrated systematic multi-year screening (2.2M
852 simulations), dual-model verification for all worst-cases, machine learning for rapid prediction, and variance-
853 based sensitivity analysis within a unified framework that maximises the scientific return from a single
854 extensive simulation campaign. This methodological integration—extracting worst-case scenarios, machine
855 learning models, and sensitivity metrics from the same computational ensemble—provides a template for
856 contemporary nuclear emergency preparedness assessments, particularly for nations reliant on transboundary
857 transport analysis from facilities in neighbouring countries.

858 4.2 Proximity-Dependent Radiological Impact: Existing Facilities Negligible, 859 Wylfa Concerning

860 The independent verification of worst-case scenarios using HYSPLIT and FLEXPART (Section 3.2) re-
861 veals a stark proximity-dependent risk stratification. For existing operational facilities, radiological impact
862 on Ireland remains well below intervention thresholds even under meteorologically worst-case conditions.
863 Heysham, despite being the nearest operational facility to Ireland, produces the lowest doses of all exam-
864 ined sites (0.000489–0.00671 mSv over 36 hours)—a counterintuitive result explained by its AGR technology
865 having lower thermal power and different release characteristics compared to PWR designs. Sizewell B
866 (0.6–1.4 mSv), Flamanville (1.5–4.6 mSv), and Paluel (0.5–2.0 mSv) exhibit elevated but still sub-threshold
867 consequences. Hinkley Point C, reflecting its coastal location and proximity to Ireland, shows elevated predic-
868 tions (FLEXPART: 8.53 mSv; HYSPLIT: 1.38 mSv) but remains well below Irish national intervention levels
869 (50 mSv for sheltering, 100 mSv for evacuation) [11]. However, the cancelled Wylfa Newydd gigawatt-scale
870 project presents fundamentally different consequences: maximum deposition conditions (May 2024 scenario)
871 predict 4.5–20.7 mSv total effective dose over 36 hours. If such releases were sustained over the 7-day assess-
872 ment period used in international criteria, cumulative doses could approach or exceed the Irish sheltering
873 threshold of 50 mSv. It should be emphasised that this analysis represents an upper-bound worst-case: the
874 Wylfa site is now proposed for SMR deployment with thermal power approximately three times lower than
875 the gigawatt-scale EPR modelled here, meaning actual consequences from the proposed facility would be cor-
876 respondingly reduced and comfortably below intervention thresholds even under worst-case meteorological
877 conditions.

878 This proximity-dependent risk pattern reflects fundamental atmospheric dispersion physics: dilution in
879 increases exponentially with transport distance, whilst deposition losses reduce airborne inventory progressively
880 during transit. Wylfa's location on Anglesey places it approximately 80–100 km from Ireland—sufficiently
881 close that severe accident plumes can reach Ireland before substantial atmospheric dilution or deposition
882 losses occur. In contrast, Heysham (approximately 180 km), Sizewell B (approximately 400 km), and conti-
883 nental facilities (> 600 km) benefit from extended transport periods permitting multiple e-folding times for
884 atmospheric mixing and wet/dry deposition removal. The May 2024 Wylfa scenario combines three adverse
885 factors: short transport distance, sustained northerly flow maintaining plume coherence, and precipitation
886 enhancement accelerating groundshine-generating deposition. This meteorological pattern—characterised by
887 late spring frontal systems with sustained north-to-south trajectories—represents the specific seasonal threat
888 requiring emergency preparedness focus should Wylfa become operational. The dose assessments employed
889 conservative assumptions throughout (continuous outdoor exposure for 36 hours, no sheltering or protective
890 actions, isotope-specific dose coefficients for adult members of the public), yet existing facilities still yielded
891 sub-threshold doses whilst Wylfa approached intervention levels.

892 The multi-model verification approach strengthens confidence in these proximity-stratified conclusions.
893 For existing facilities, both models converge on sub-threshold impact, with doses ranging from negligible
894 (Heysham, < 0.01 mSv) to low but measurable (Flamanville, Sizewell, Paluel, 0.5–5 mSv). For Hinkley Point
895 C, both models agree on sub-threshold consequences with factor-of-~6 quantitative divergence (FLEXPART
896 8.53 mSv vs HYSPLIT 1.38 mSv), demonstrating reasonable inter-model agreement for elevated-consequence
897 scenarios. For Wylfa, the factor-of-~5 disagreement (FLEXPART 20.7 mSv vs HYSPLIT 4.5 mSv over 36
898 hours) places both predictions in a range where sustained releases could approach Irish intervention levels,
899 with both models identifying this facility as uniquely concerning. The systematic pattern of inter-model
900 variability increasing with consequence magnitude reflects the methodological differences in deposition field
901 representation (near-surface concentration proxy versus explicit deposition schemes) combined with wet de-
902 position parameterisation divergence, exacerbated in near-source scenarios where small-scale meteorological
903 features and surface characteristics exert dominant influence. For emergency preparedness applications,
904 where decisions must be made under uncertainty, the dual-model framework provides credible risk stratifica-
905 tion: existing facilities pose low concern (all scenarios well below intervention thresholds), Hinkley warrants
906 monitoring (both models well below 50 mSv), Wylfa requires robust emergency capabilities (36-hour doses
907 suggest potential threshold exceedance under sustained release scenarios). This risk-informed categorisation
908 remains valid despite inter-model quantitative differences, demonstrating that multi-model consensus on rel-
909 ative risk hierarchy provides more value than pursuing spurious numerical precision from single deterministic
910 predictions.

911 4.3 Atmospheric Drivers of Radiological Consequences

912 The global sensitivity analysis (Section 3.4) revealed that meteorological conditions, rather than release
913 parameters, dominate consequence severity for events impacting Ireland. Total-order Sobol' indices for release
914 height and duration were consistently below 0.001, indicating negligible contribution to output variance
915 amongst impactful scenarios. In contrast, wind-related variables—particularly kurtosis of west-east wind at
916 1000 m and median wind speed at 10 m—ranked as the most influential parameters across multiple nuclear
917 power plants and output metrics. The kurtosis metric, measuring the degree of extreme events in wind
918 speed distributions, proved more influential than mean or median wind speed, indicating that infrequent
919 strong-wind events drive highest-consequence transport scenarios rather than typical conditions.

920 Geopotential height skewness appeared frequently amongst the top three influential variables, particularly
921 for facilities requiring longer transport distances (Flamanville, Paluel, Sizewell). This variable serves as a
922 proxy for atmospheric pressure structure; its skewness quantifies asymmetry in pressure gradient distributions
923 and correlates with frontal passages and synoptic-scale weather systems. The consistent appearance of this
924 parameter suggests that worst-case transport scenarios for Ireland frequently coincide with active frontal
925 systems providing both strong advection and precipitation-driven wet deposition enhancement.

926 These findings carry important implications for operational emergency response. During the early phase
927 of a nuclear accident, substantial uncertainty typically surrounds source term characteristics including release
928 timing, duration, and effective height. The sensitivity analysis demonstrates that, for events impacting Ire-
929 land, consequence severity is governed primarily by meteorological conditions rather than these source term

930 details. Emergency response priorities should therefore emphasise accurate real-time meteorological forecasting
931 and rapid atmospheric model execution over protracted source term refinement during initial response
932 phases. Meteorological monitoring should prioritise variables identified as influential by the sensitivity analysis,
933 particularly upper-level wind patterns at 1000 m and surface pressure gradients. Notably, for proximal
934 facilities (Wylfa, Heysham), geopotential height appears infrequently or not at all amongst the top three
935 sensitivity drivers, with wind variables predominantly governing consequence severity. In contrast, for more
936 distant sites (Hinkley, Sizewell), geopotential height consistently ranks as the second or third most influential
937 variable, reflecting the greater importance of synoptic-scale pressure patterns in determining whether plumes
938 traverse extended transport distances to reach Ireland.

939 4.4 Seasonal and Geographical Patterns

940 The fourteen-year systematic screening revealed distinct seasonal patterns in atmospheric transport probability
941 and intensity (Section 3.1). Spring months, particularly April, consistently exhibited elevated plume
942 intersection frequencies across all nuclear power plants, reflecting the climatology of North Atlantic cyclogenesis
943 and associated easterly or southeasterly flow patterns. Worst-case scenarios concentrated during
944 transitional seasons, with spring (March–May) accounting for seven of eighteen scenarios and autumn
945 (September–October) contributing five scenarios. Late summer months (July–August) consistently ranked as
946 lowest-risk periods, reflecting the dominance of westerly flow and anticyclonic conditions during this season.

947 Geographical proximity exerted the expected strong influence on maximum deposition values, with Wylfa
948 achieving the highest unit-release deposition ($3.19 \times 10^{-8} \text{ kg m}^{-2}$) due to short transport distance and minimal
949 intervening dilution. However, proximity did not guarantee shortest warning times: the minimum plume
950 arrival scenario (3 hours) occurred for three facilities simultaneously (Heysham, Hinkley Point C, Wylfa)
951 during a particularly favourable synoptic pattern on 25 April 2012, characterised by strong westerly to
952 northwesterly flow. This finding demonstrates that synoptic-scale meteorological conditions can temporarily
953 equalise transport times from facilities at differing distances, with implications for monitoring network design
954 and alert protocols.

955 Release parameter analysis demonstrated that fourteen of eighteen worst-case scenarios involved extended
956 release durations (48 hours), consistent with severe accident sequences characterised by prolonged containment
957 degradation rather than catastrophic early failure. Release heights clustered at the extremes of the
958 examined range, with elevated releases (50–100 m) dominating maximum deposition scenarios due to reduced
959 near-source dry deposition and enhanced long-range transport capability. These parametric findings align
960 with Level 2 probabilistic safety assessment predictions for contemporary reactor designs, where containment
961 failure modes typically occur many hours after core damage initiation.

962 4.5 Machine Learning for Rapid Impact Prediction

963 The XGBoost classification models achieved validation accuracies of 85.4–92.5% for predicting atmospheric
964 impact occurrence based on meteorological conditions and release parameters (Section 3.3). Models for
965 proximal facilities (Wylfa, Heysham) exhibited higher recall (0.76–0.78), correctly identifying most true
966 impact events whilst accepting moderate false-positive rates. Models for distant facilities (Paluel, Flamanville,
967 Sizewell) achieved higher specificity (0.96–0.97), effectively ruling out non-impact scenarios whilst accepting
968 lower sensitivity. This performance gradient reflects the underlying class imbalance: distant facilities impact
969 Ireland less frequently, favouring high-specificity classifiers that minimise false alarms.

970 Feature importance analysis revealed consistency with the global sensitivity analysis findings, with geopotential
971 height, surface pressure, and wind-related variables ranking as most influential predictors. This convergence of results across fundamentally different methodologies (variance-based sensitivity analysis versus
972 tree-based feature importance) strengthens confidence in the identification of critical atmospheric parameters.
973 The machine learning models provide complementary capability to traditional dispersion modelling:
974 whilst HYSPLIT and FLEXPART require meteorological forecast fields and several hours of computation
975 time, the trained XGBoost models execute in milliseconds and can process ensemble weather forecasts to
976 provide probabilistic impact predictions during the early phase of an event.

977 Operational deployment of these models could support rapid screening of meteorological forecast ensembles
978 to identify high-probability transport scenarios toward Ireland, enabling pre-positioning of monitoring

980 resources and alert protocols. The models' demonstrated ability to generalise to unseen 2024 validation data
981 suggests robustness to inter-annual meteorological variability, though periodic retraining with updated me-
982 teorological data would be advisable to account for potential climate-driven shifts in atmospheric circulation
983 patterns.

984 4.6 Methodological Strengths and Limitations

985 The primary methodological strength of this study lies in the convergence of multiple complementary ap-
986 proaches toward consistent conclusions. The systematic fourteen-year HYSPLIT screening provided statis-
987 tical robustness through large ensemble size (2.2 million simulations total), substantially exceeding previous
988 Irish assessments in temporal coverage. Independent FLEXPART verification of worst-case scenarios demon-
989 strated multi-model consensus on negligible radiological impact, addressing the fundamental limitation of
990 single-model studies. Machine learning and sensitivity analysis, both leveraging the same simulation ensem-
991 ble generated for worst-case identification, provided mechanistic understanding of atmospheric drivers whilst
992 maximising the value extracted from the substantial computational investment, enabling prioritisation of
993 monitoring and forecasting resources.

994 Several limitations warrant acknowledgement. The atmospheric dispersion models employed horizontal
995 resolutions of 0.25 degrees (approximately 20–25 km), adequate for long-range transport but insufficient
996 for resolving sub-grid-scale terrain features or coastal effects that may influence local deposition patterns.
997 Deposition parameterisations employed representative values for aerosol-phase fission products but did not
998 account for particle size distribution evolution during transport or isotope-specific chemical behaviour. The
999 severe accident source terms, whilst based on Level 2 probabilistic safety assessment for late containment
1000 failure, necessarily simplified the complex temporal evolution of releases that would occur during actual
1001 accident sequences.

1002 The study focused exclusively on atmospheric transport pathways, neglecting potential marine transport
1003 following deposition to the Irish Sea or English Channel. For isotopes with long environmental half-lives
1004 (cesium-137, plutonium isotopes), marine transport and subsequent incorporation into seafood pathways
1005 could provide delayed exposure routes not captured by the 36-hour atmospheric assessment window. However,
1006 previous RPII assessments that investigated marine transport pathways from UK nuclear facilities concluded
1007 that radiological impacts via this route were negligible [27]. Future work incorporating coupled atmosphere-
1008 ocean transport models would nonetheless provide more comprehensive consequence assessment for coastal
1009 receptors.

1010 The unit-release screening methodology enabled efficient identification of meteorologically worst-case
1011 transport patterns but required subsequent application of realistic source terms for radiological assessment.
1012 An alternative approach employing full multi-isotope releases for all 2.2 million simulations would have elim-
1013 inated this two-stage process but at prohibitive computational cost. The demonstrated consistency between
1014 unit-release screening results and full-physics verification simulations validates the adopted methodology for
1015 future applications.

1016 4.7 Implications for Emergency Preparedness and Public Communication

1017 The findings of this study have direct implications for Ireland's nuclear emergency preparedness strategies.
1018 The demonstration that even worst-case meteorological scenarios produce doses orders of magnitude below
1019 intervention thresholds suggests that emergency response priorities should emphasise accurate public commu-
1020 nication and minimisation of unnecessary disruption over immediate large-scale protective actions. Historical
1021 evidence from Fukushima and Chernobyl demonstrates that public anxiety, economic disruption, and unnec-
1022 essary evacuation can produce greater societal harm than the radiological consequences themselves for areas
1023 receiving low-level contamination [3, 29, 30].

1024 The seasonal patterns identified in this analysis (spring maxima, summer minima for transport proba-
1025 bility) could inform resource allocation decisions, with enhanced monitoring readiness and staff availability
1026 during higher-risk periods. The machine learning models provide capability for rapid preliminary assessment
1027 during the initial phase of an event, before detailed atmospheric dispersion calculations are feasible, enabling
1028 earlier public communication and stakeholder engagement.

1029 The identified worst-case scenarios (Table 2) provide specific test cases for emergency response exercises

1030 and decision support system validation. Regular training exercises employing these scenarios would ensure
1031 that response personnel are familiar with the range of plausible atmospheric transport patterns and asso-
1032 ciated timescales for decision-making. The multi-model verification results demonstrate the importance of
1033 maintaining multiple independent dispersion modelling capabilities to provide confidence intervals around
1034 predictions rather than single deterministic values.

1035 Perhaps most importantly, the quantitative demonstration of minimal radiological impact under worst-
1036 case conditions provides an evidence base for public communication regarding nuclear safety. Whilst nuclear
1037 accidents at nearby facilities would undoubtedly generate substantial public concern and media attention,
1038 the ability to provide quantitative context—comparing predicted doses to natural background radiation,
1039 medical procedures, or routine activities—would support informed decision-making and potentially mitigate
1040 unnecessary anxiety. The findings should not diminish the importance of robust emergency preparedness but
1041 rather inform appropriate calibration of response measures to predicted consequence severity.

1042 5 Conclusions

1043 This study presents the most comprehensive quantitative assessment to date of potential transboundary
1044 radiological transport to Ireland from nuclear power plants in the United Kingdom and France. Through
1045 systematic screening of 2.2 million HYSPLIT atmospheric dispersion simulations spanning fourteen years
1046 (2011–2024), independent verification using FLEXPART, machine learning for rapid impact prediction, and
1047 global sensitivity analysis, the research establishes a robust, evidence-based foundation for Ireland’s nuclear
1048 emergency preparedness planning.

1049 The central finding provides substantial reassurance regarding existing operational facilities: even under
1050 meteorologically worst-case conditions, realistic severe accident releases produce radiological doses to Ireland
1051 dramatically below intervention thresholds. Across all existing facilities, 36-hour doses range from negligible
1052 (Heysham, <0.01 mSv) through low (Sizewell, Flamanville, Paluel, 0.5–5 mSv) to elevated but sub-threshold
1053 (Hinkley, 1–9 mSv). Even when extrapolated to the 7-day assessment period used in international criteria, all
1054 existing facilities remain well below the IAEA generic criterion of 100 mSv [20] and Irish intervention levels
1055 of 50 mSv for sheltering [11]. This safety margin reflects realistic severe accident source terms, substantial
1056 atmospheric dilution during transport, and Ireland’s geographical separation from these facilities.

1057 However, the analysis identifies Wylfa as requiring particular attention. The 36-hour cumulative doses
1058 of 4.5–20.7 mSv predicted for worst-case Wylfa scenarios, whilst below thresholds for single-event exposure,
1059 suggest that sustained releases coinciding with persistent easterly flow could approach or exceed Irish inter-
1060 vention levels when extrapolated to the 7-day assessment period. This finding warrants continued vigilance
1061 regarding any future development at the Wylfa site, given its extreme proximity to Ireland (approximately
1062 80–100 km).

1063 Independent verification using FLEXPART demonstrated robust multi-model consensus on this funda-
1064 mental conclusion of minimal radiological impact, despite inter-model differences of up to a factor of ~ 10
1065 in absolute dose magnitudes. This verification addresses a critical limitation of single-model assessments
1066 and provides confidence that predicted transport patterns represent genuine atmospheric phenomena rather
1067 than model-specific artefacts. For emergency preparedness applications requiring decisions under uncertainty,
1068 the demonstration of multi-model agreement on dose order-of-magnitude constitutes substantially stronger
1069 evidence than reliance on deterministic predictions from individual models.

1070 Global sensitivity analysis revealed that meteorological conditions, rather than release parameters, dom-
1071 inate consequence severity for events impacting Ireland. Wind-related variables—particularly kurtosis of
1072 west-east wind at 1000 m and median wind speed at 10 m—ranked as most influential parameters, with re-
1073 lease height and duration contributing negligibly to output variance (Sobol’ indices < 0.001). This finding
1074 has important operational implications: emergency response priorities should emphasise accurate meteo-
1075 rological forecasting and rapid atmospheric model execution over protracted source term refinement during
1076 initial response phases. Meteorological monitoring should prioritise upper-level wind patterns and surface
1077 pressure gradients identified as primary drivers of consequence variability.

1078 Machine learning models (XGBoost) achieved validation accuracies of 85.4–92.5% for predicting atmo-
1079 spheric impact occurrence, providing capability for rapid preliminary assessment during the early phase of
1080 events before detailed dispersion calculations are feasible. Feature importance analysis demonstrated consis-
1081 tency with sensitivity analysis findings, with geopotential height, surface pressure, and wind variables ranking
1082 as most influential predictors. This convergence across fundamentally different methodologies strengthens
1083 confidence in the identification of critical atmospheric parameters requiring monitoring prioritisation.

1084 The systematic fourteen-year screening revealed distinct seasonal patterns in atmospheric transport prob-
1085 ability, with spring months (particularly April) exhibiting elevated plume intersection frequencies and worst-
1086 case scenarios concentrating during transitional seasons. Late summer months (July–August) consistently
1087 ranked as lowest-risk periods. Fourteen of eighteen worst-case scenarios involved extended release durations
1088 (48 hours), consistent with severe accident sequences characterised by prolonged containment degradation
1089 rather than catastrophic early failure. These temporal and parametric patterns provide actionable intelligence
1090 for emergency preparedness resource allocation and exercise planning.

1091 Several recommendations emerge from this research for enhancing Ireland’s nuclear emergency prepared-
1092 ness capabilities. First, the demonstrated importance of meteorological conditions in governing conse-
1093 quence severity emphasises the need for sustained investment in high-quality meteorological forecasting and ensem-
1094 ble prediction systems. Second, the multi-model verification results demonstrate the value of maintaining

1095 multiple independent atmospheric dispersion modelling capabilities to provide confidence intervals around
1096 predictions rather than single deterministic values. Third, the machine learning models developed in this
1097 study should be integrated into operational decision support systems to enable rapid preliminary impact
1098 assessment during the initial phase of events. Fourth, the identified worst-case scenarios provide specific test
1099 cases for regular emergency response exercises, ensuring that response personnel are familiar with the range
1100 of plausible atmospheric transport patterns and associated decision-making timescales.

1101 Fifth, and perhaps most importantly, the quantitative demonstration of minimal radiological impact
1102 under worst-case conditions provides an evidence base for calibrated public communication. Whilst nu-
1103 clear accidents at nearby facilities would undoubtedly generate substantial public concern, the ability to
1104 provide quantitative context—comparing predicted doses to natural background radiation and routine activi-
1105 ties—would support informed decision-making and potentially mitigate unnecessary anxiety and disruption.
1106 Emergency response protocols should be calibrated to predicted consequence severity rather than worst-case
1107 assumptions decoupled from realistic atmospheric transport and source term physics.

1108 Future research directions could usefully address several limitations identified in this study. Coupled
1109 atmosphere-ocean transport models would provide more comprehensive consequence assessment for coastal
1110 receptors, capturing marine transport pathways and seafood incorporation for long-lived isotopes. Higher-
1111 resolution nested dispersion modelling would better resolve sub-grid-scale terrain features and coastal effects
1112 influencing local deposition patterns. Explicit treatment of particle size distribution evolution and isotope-
1113 specific chemical behaviour would refine deposition predictions. Integration of the atmospheric transport
1114 findings with Level 3 probabilistic safety assessment could provide fully probabilistic consequence estimates
1115 incorporating both accident frequency and meteorological variability.

1116 Notwithstanding these potential refinements, the current study establishes a comprehensive, quantita-
1117 tive foundation for Ireland's nuclear emergency preparedness. The convergence of findings across multiple
1118 methodologies—large-ensemble screening, multi-model verification, machine learning, and sensitivity analy-
1119 sis—provides robust evidence that Ireland's geographical separation from major nuclear facilities, combined
1120 with modern reactor safety systems and realistic severe accident physics, results in minimal radiological con-
1121 sequences even under meteorologically worst-case conditions. This evidence base should inform appropriate
1122 calibration of emergency response measures, resource allocation, and public communication strategies, ensur-
1123 ing that Ireland maintains effective preparedness whilst avoiding unnecessary disruption and anxiety based
1124 on overly conservative assumptions disconnected from realistic consequence assessment.

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1311 Appendix: Complete Dose Assessment Results

1312 Table 9 presents comprehensive dose assessment results for all eighteen worst-case scenarios across six nuclear
 1313 facilities. Section 3.2 provided detailed spatial analysis of the Heysham maximum deposition scenario as
 1314 a representative case study, demonstrating model verification methodology and spatial dose distribution
 1315 patterns.

Table 9: Combined Dose Assessment for Whole of Ireland: All Worst-Case Scenarios

Scenario	Gamma Rate (mSv/h)		Inhalation (mSv)		Total Dose (mSv)	
	FP	HS	FP	HS	FP	HS
Flamanville Max Concentration	0.0172	0.00305	0.00137	0.00249	0.622	0.112
Flamanville Max Deposition	0.128	0.0402	0.00216	0.00623	4.62	1.45
Flamanville Min Plume Arrival	0.000574	0.00126	6.19e-06	1.04e-06	0.0207	0.0453
Heysham Max Concentration	5.51e-06	6.07e-06	2.68e-06	3.26e-06	0.000201	0.000222
Heysham Max Deposition	1.35e-05	0.000186	4.25e-06	2.18e-05	0.000489	0.00671
Heysham Min Plume Arrival	1.25e-06	3.52e-05	4.2e-07	1.3e-07	4.52e-05	0.00127
Hinkley Max Concentration	0.053	0.032	0.00538	0.0249	1.91	1.18
Hinkley Max Deposition	0.237	0.0383	0.00264	0.00233	8.53	1.38
Hinkley Min Plume Arrival	0.00931	0.00768	0.000184	0.000101	0.335	0.277
Paluel Max Concentration	0.0261	0.00311	0.00172	0.00193	0.939	0.113
Paluel Max Deposition	0.0562	0.0132	0.000971	0.00156	2.02	0.474
Paluel Min Plume Arrival	0.0806	0.0101	0.0017	6.2e-05	2.9	0.363
Sizewell Max Concentration	0.00318	0.00416	0.000171	3.51e-06	0.115	0.15
Sizewell Max Deposition	0.0399	0.017	0.000499	0.00146	1.44	0.612
Sizewell Min Plume Arrival	0.0285	0.0151	0.000315	3.01e-05	1.03	0.545
Wylfa Max Concentration	0.516	0.0467	0.0116	0.0389	18.6	1.72
Wylfa Max Deposition	0.574	0.123	0.00963	0.0542	20.7	4.48
Wylfa Min Plume Arrival	0.198	0.0425	0.00414	0.00321	7.12	1.53

1316 Results employ isotope-specific deposition physics with element-based wet scavenging coefficients (Cs/I:
 1317 $1.0 \times 10^{-4} \text{ s}^{-1}$, Pu/Am/Cm: $2.0 \times 10^{-5} \text{ s}^{-1}$) and dry deposition velocities (Cs/I: $0.0015\text{--}0.002 \text{ m s}^{-1}$, Pu: 0.005 m s^{-1}).
 1318 All doses represent 36-hour cumulative exposure for the whole of Ireland, calculated using isotope-specific
 1319 dose coefficients for adult members of the public. Both FLEXPART and HYSPLIT employed identical
 1320 element-specific parameters to ensure consistent physics between models.