



Geovariances
Where no one has gone before

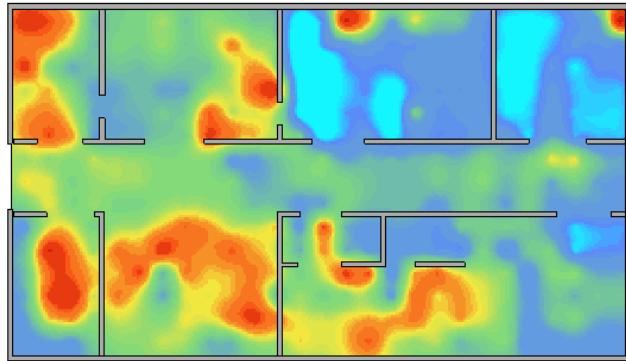
SMART USE OF THE VARIOGRAM TO
EXPLORE SPATIAL DATA, TO BREAK DOWN
VARIANCE CONTRIBUTIONS AND TO
MODEL RADIOLOGICAL CONTAMINATIONS

Yvon Desnoyers

Geostatistician - Nuclear sector
desnoyers@geovariances.com

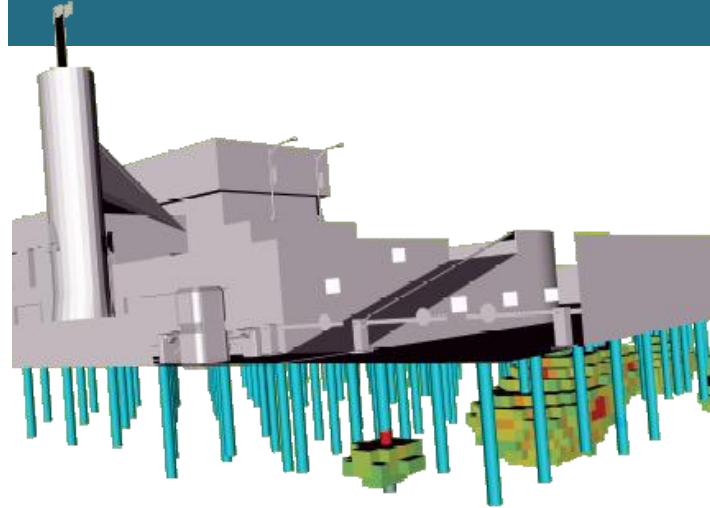
Several characterisation stages

Surface mapping



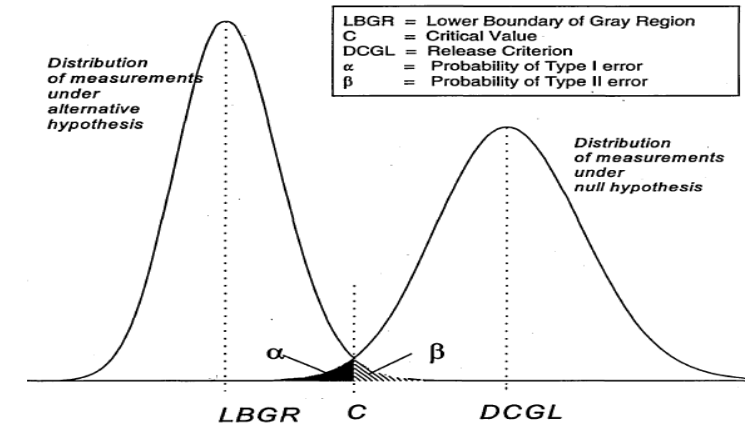
- Hot spot identification
- Zone prioritisation
- Radiation protection
- Doubt removal

Volume categorisation



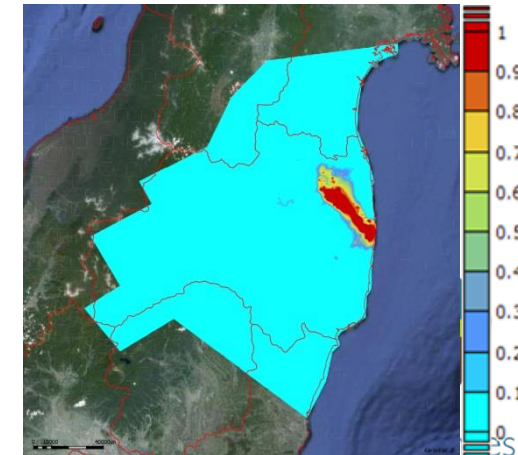
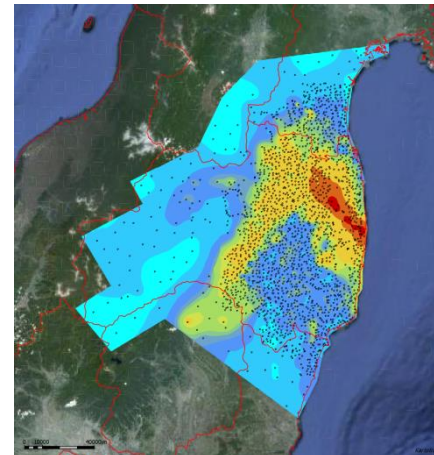
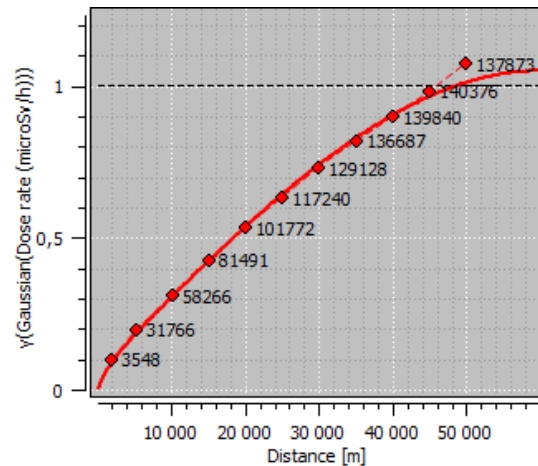
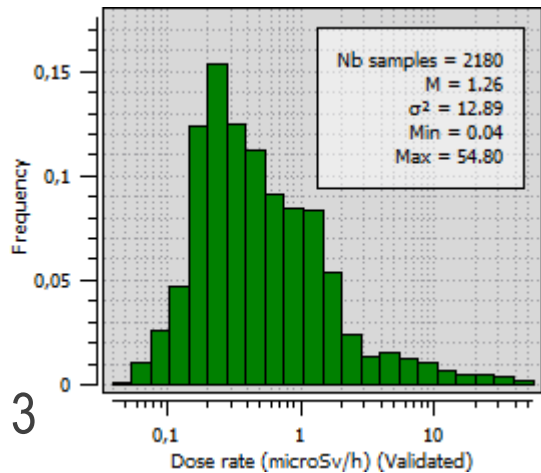
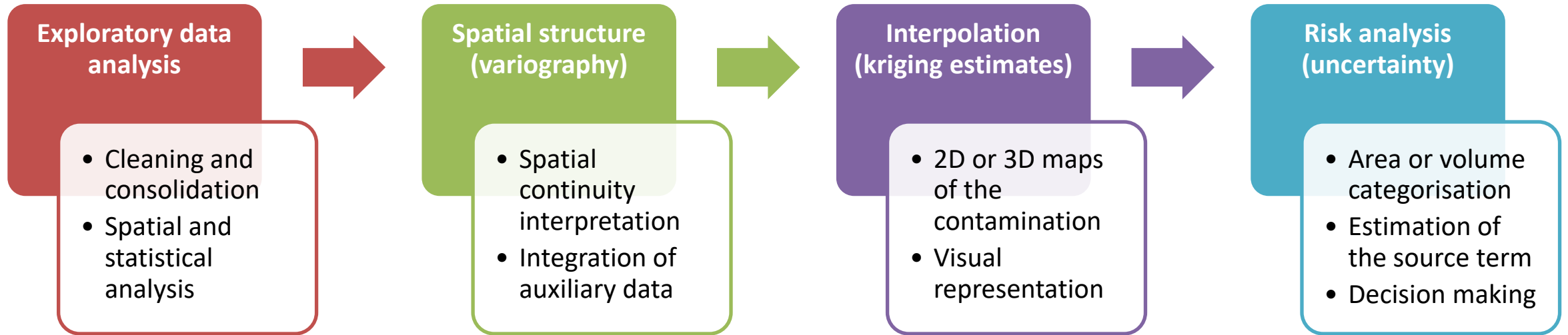
- Operational waste zoning: radiological thresholds, migration profiles
- Source term: scaling factors, average activity

Final control



- Compliance with clearance level
- Residual average activity for impact assessment

To introduce geostatistics



Smart use of the variogram to explore spatial data, to break down variance contributions and to model radiological contaminations



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1 Variogram presentation

2 Application cases

3 Advanced use of the variogram

4 Impact on sampling strategy

5 Conclusions

Classical formula for the statistical variance

$$\sigma^2 = \frac{1}{N} \sum_{i=1}^N (z_i - m)^2 \text{ avec } m = \frac{1}{N} \sum_{i=1}^N z_i$$

Alternative formula for the statistical variance

*So we can take benefit from
the distance between point i and point j*

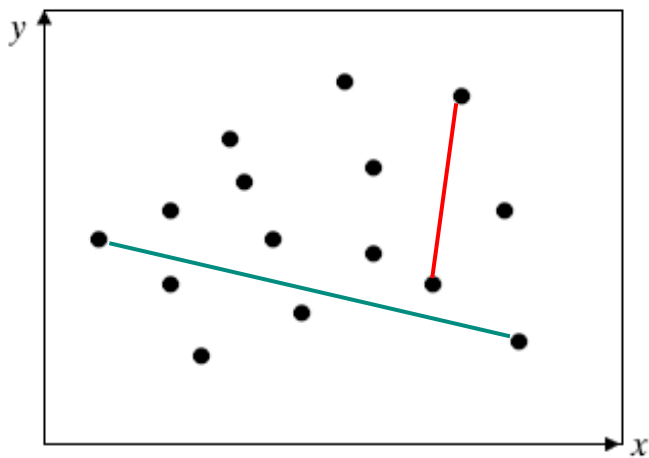
$$\sigma^2 = \frac{1}{N^2} \sum_{i=1}^N \sum_{j=1}^N \frac{1}{2} (z_i - z_j)^2$$

Calculating the mean

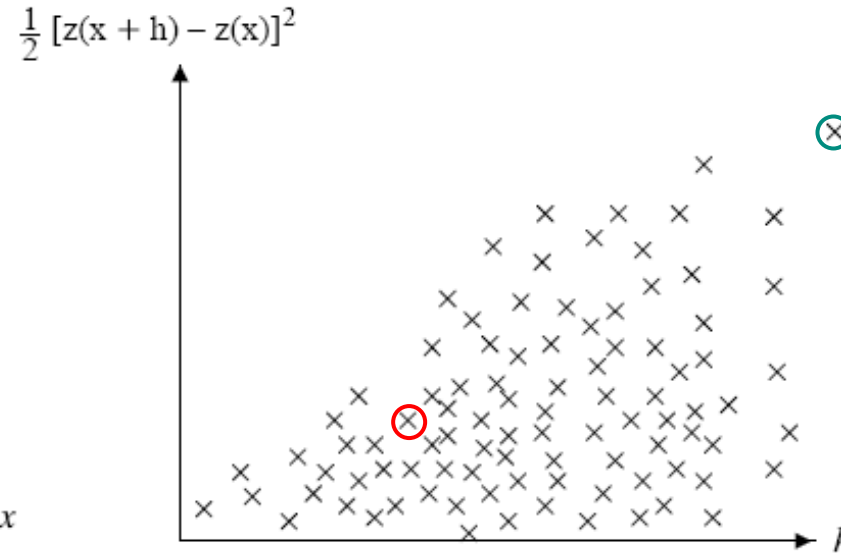
*of variabilities
between each
pair of points*

Key tool in geostatistics

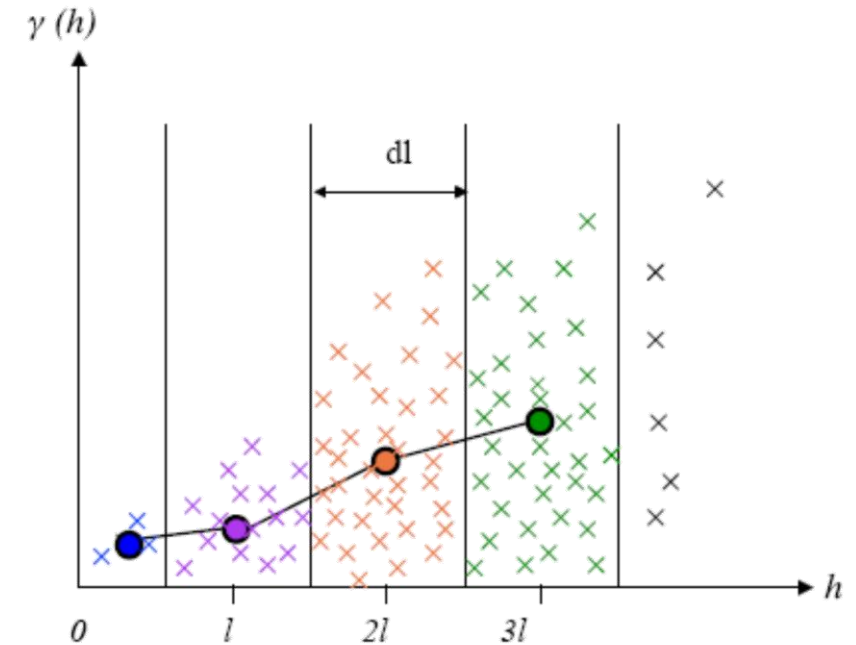
Base map



Variogram cloud

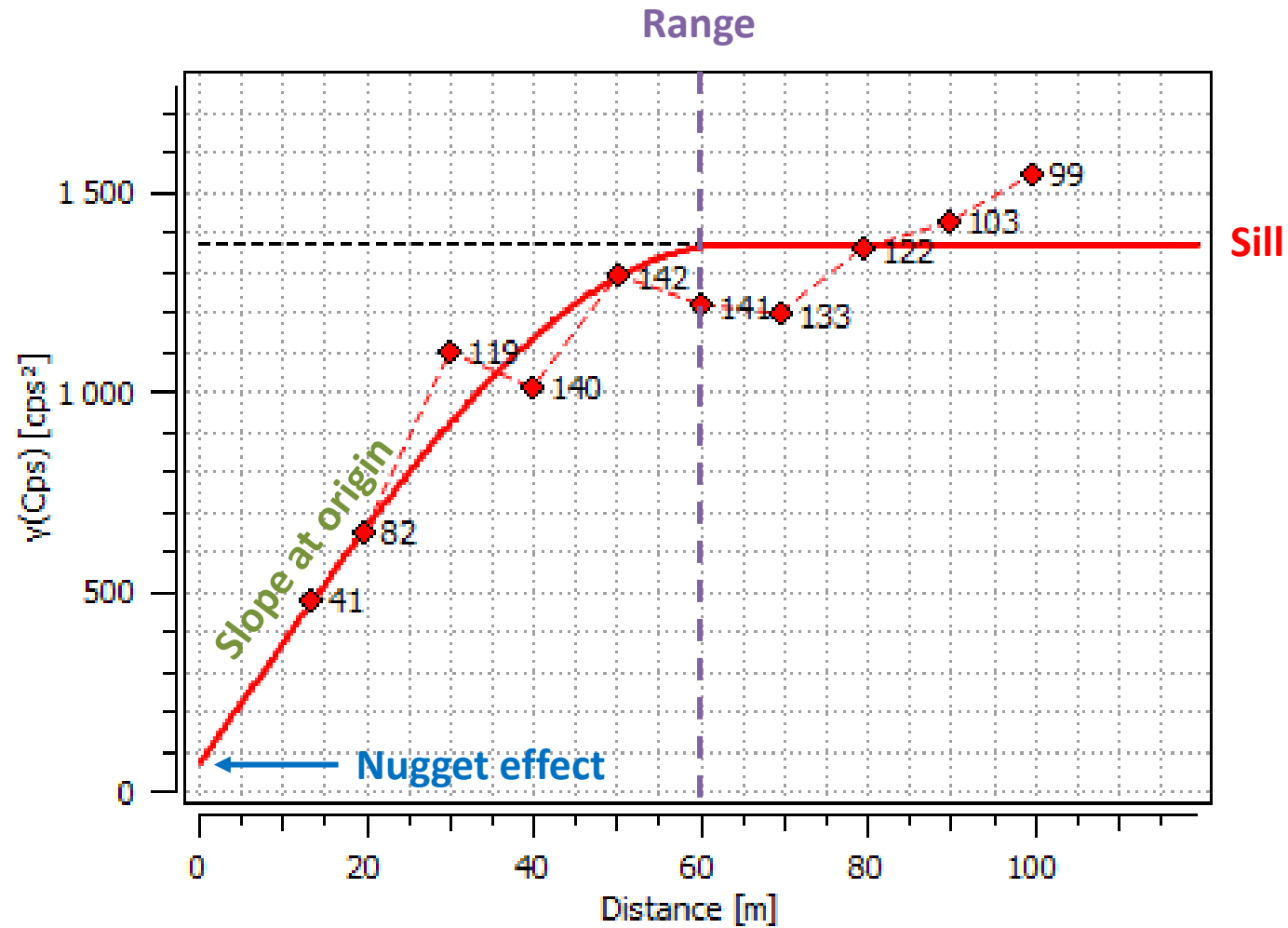


Variogram

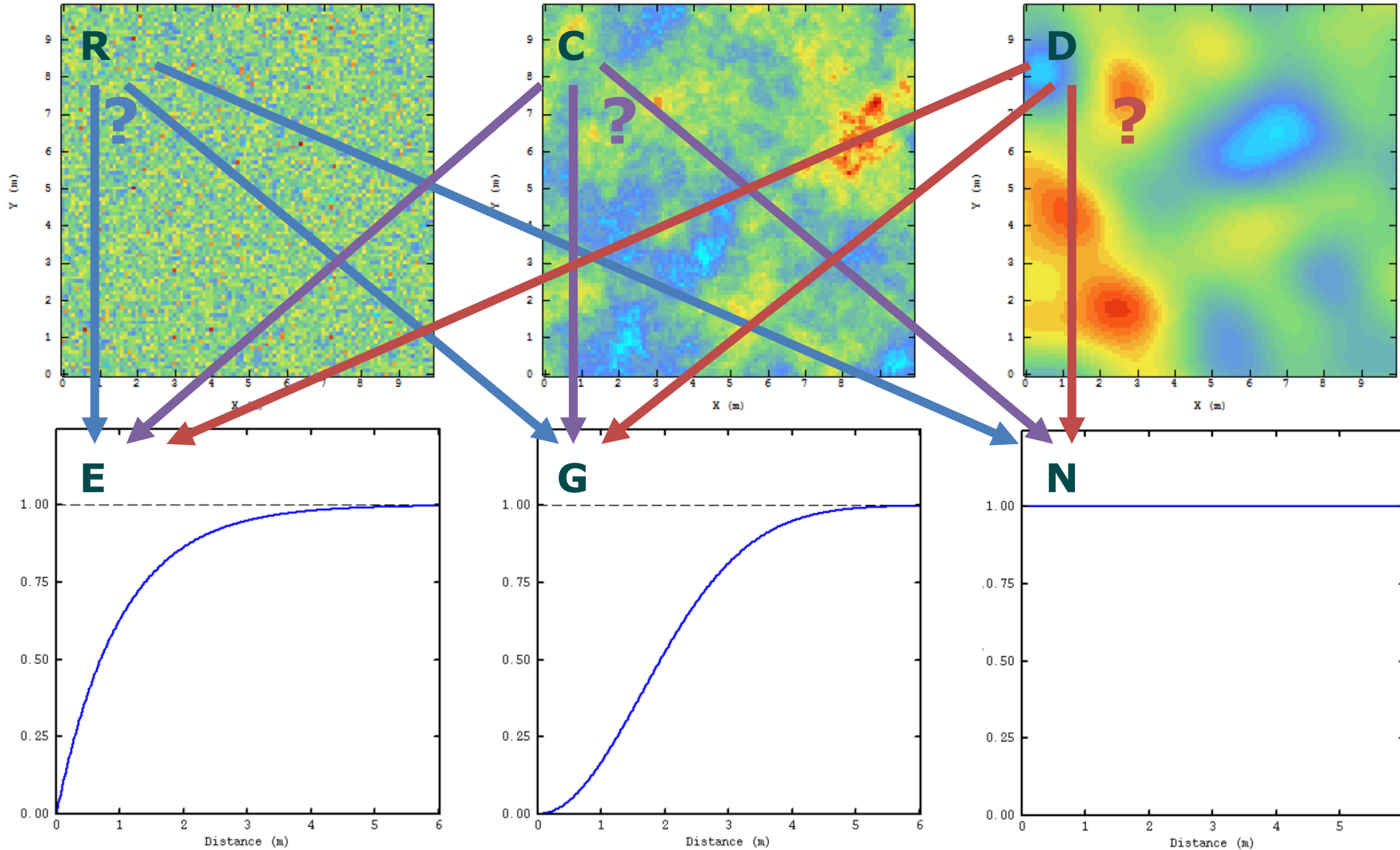


 Spatial variability to be interpreted and fitted

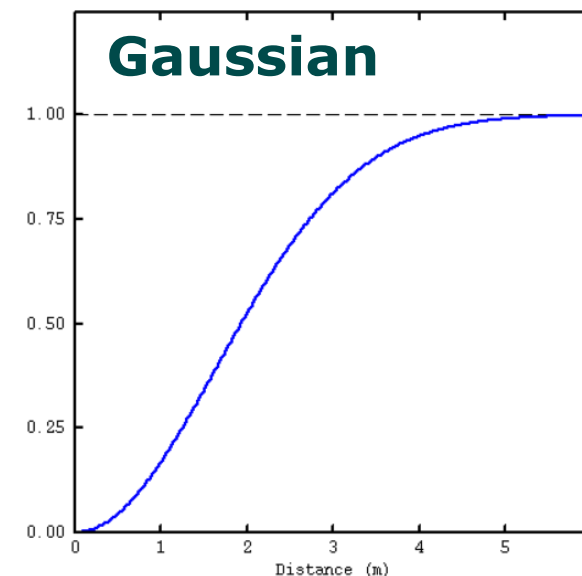
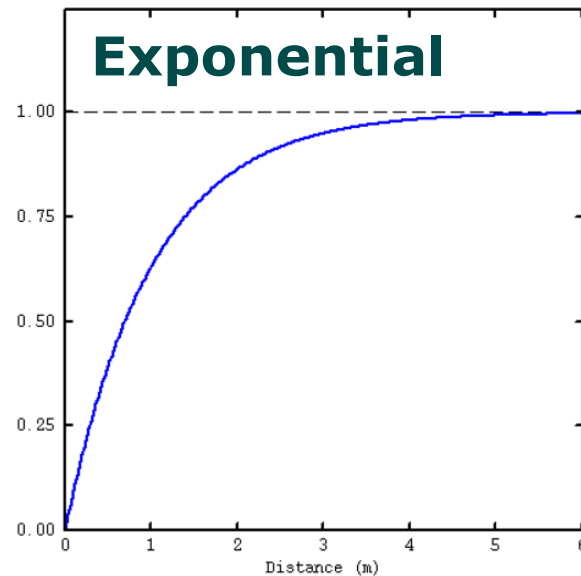
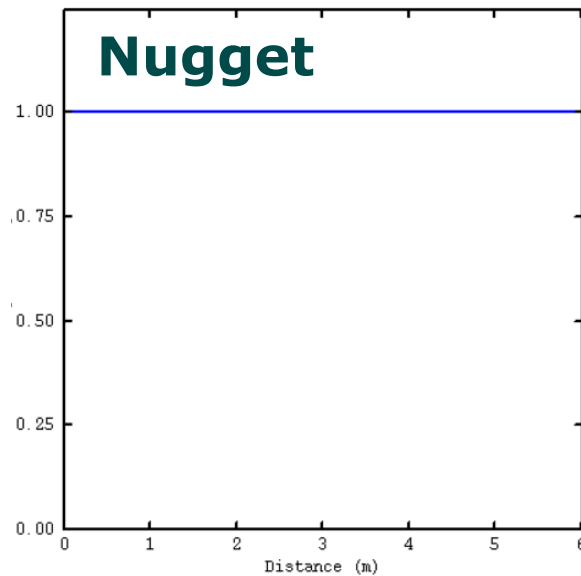
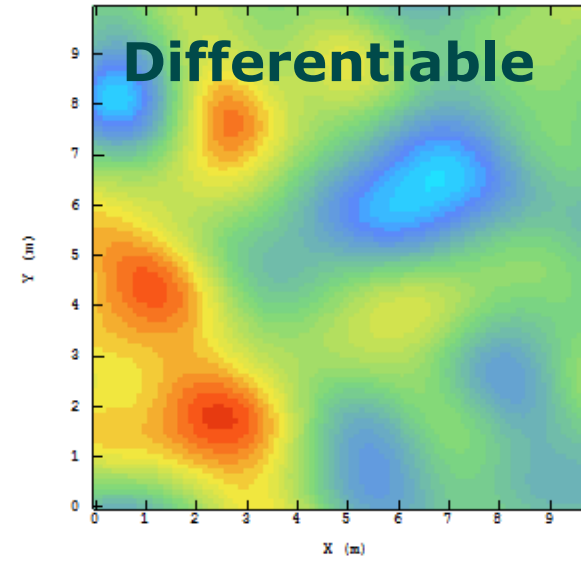
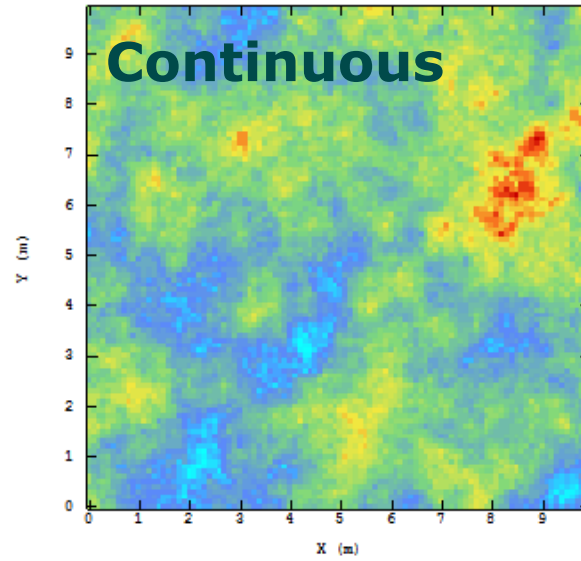
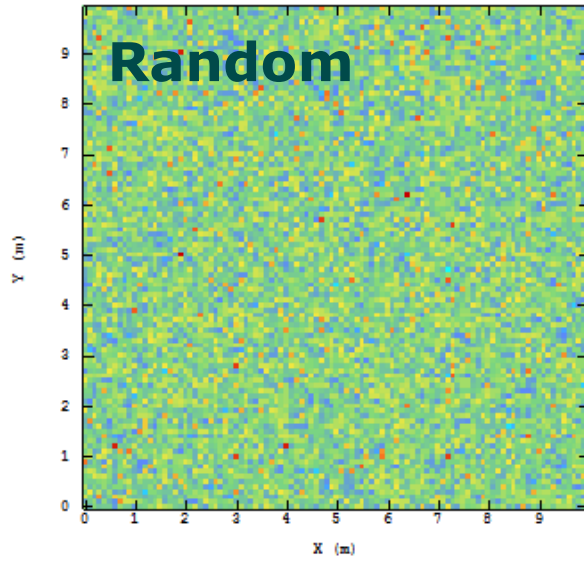
Variogram key points



Three spatial signatures...



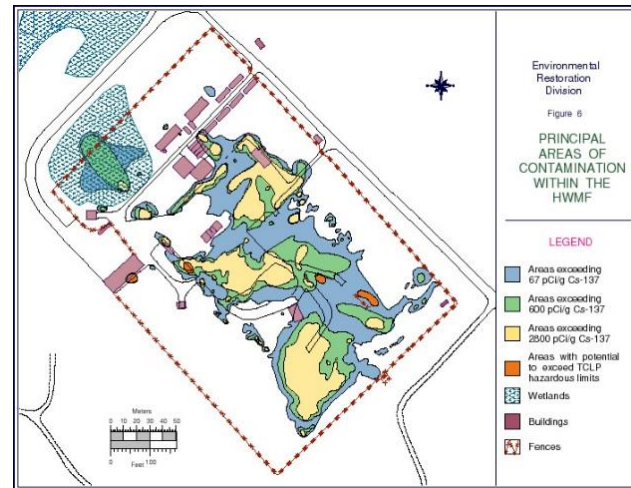
Three spatial signatures...



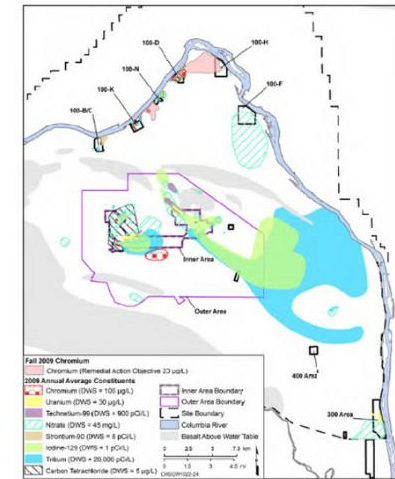
...in the nuclear context (environment)



Waste in a trench



Contaminated soils



Plume in groundwater

More and more continuous behaviour

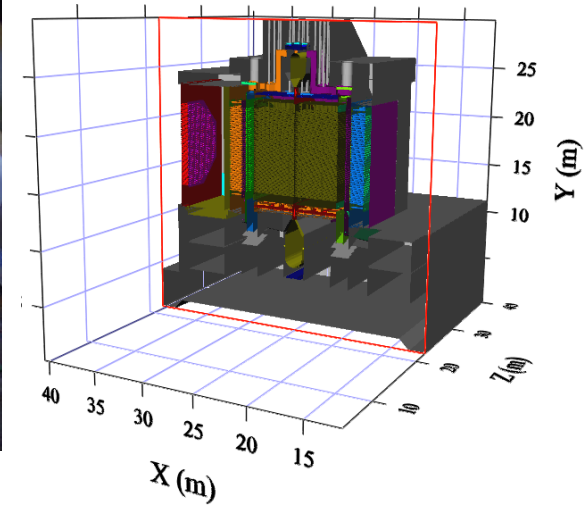
...in the nuclear context (buildings)



Waste packages



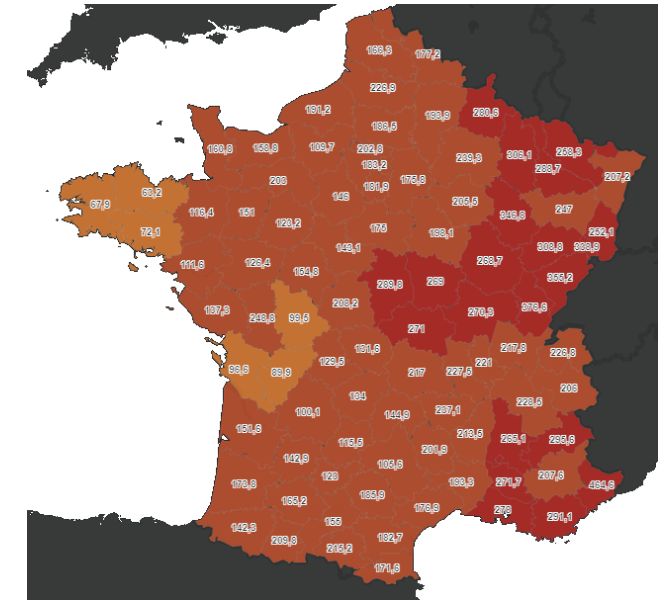
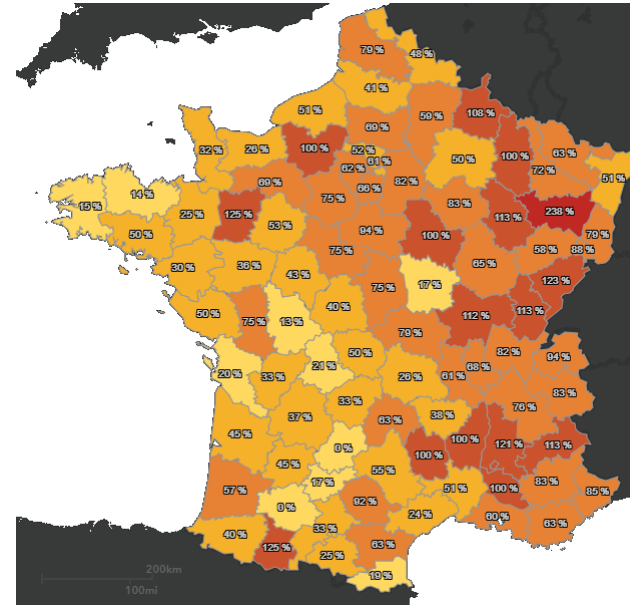
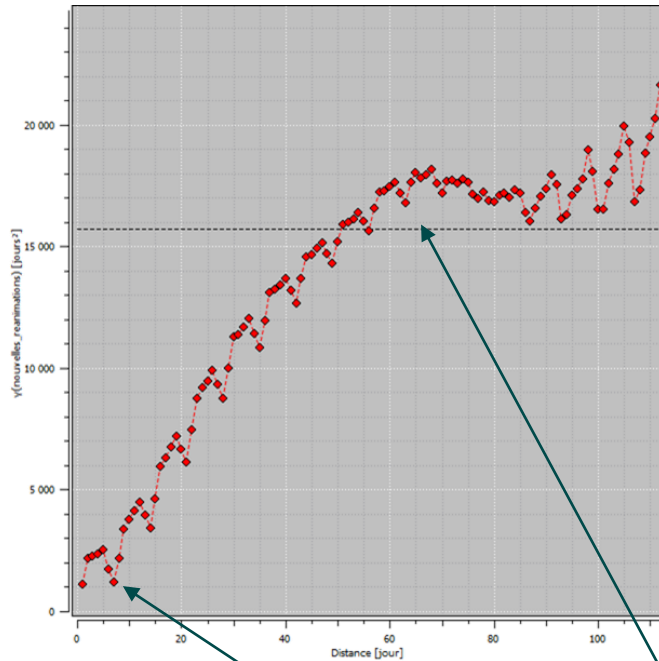
Building structures



Activation

More and more continuous behaviour

...for temporal variations (Covid-19)



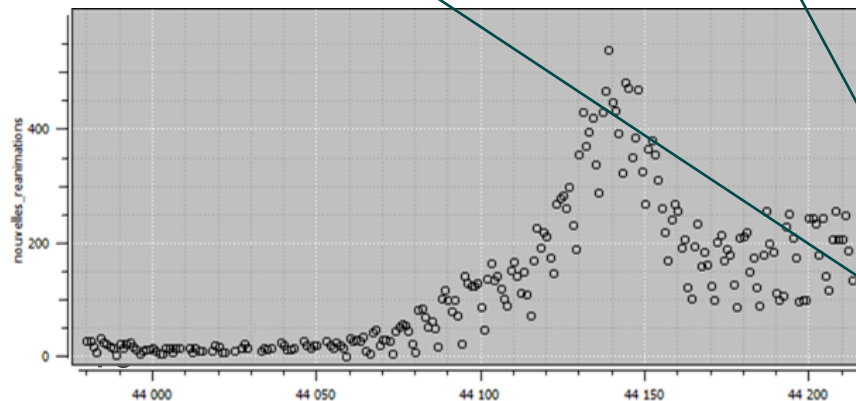
Hospital capacities

Incidence rate

Temporal fluctuations of reanimation entries

• Long-term range (60 days)

• Weekly variations (7 days)



Smart use of the variogram to explore spatial data, to break down variance contributions and to model radiological contaminations



Geovariances
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1 Variogram presentation

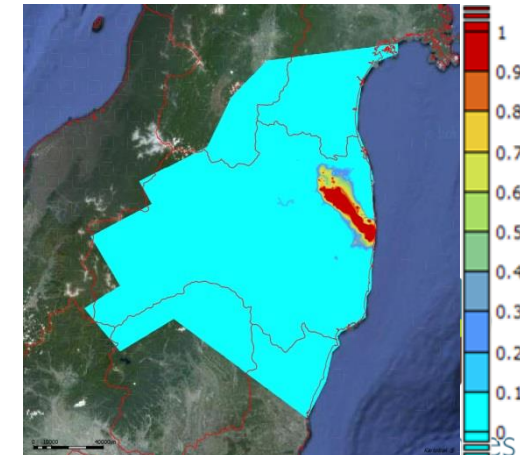
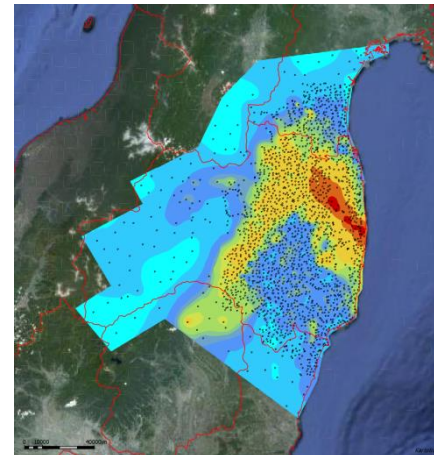
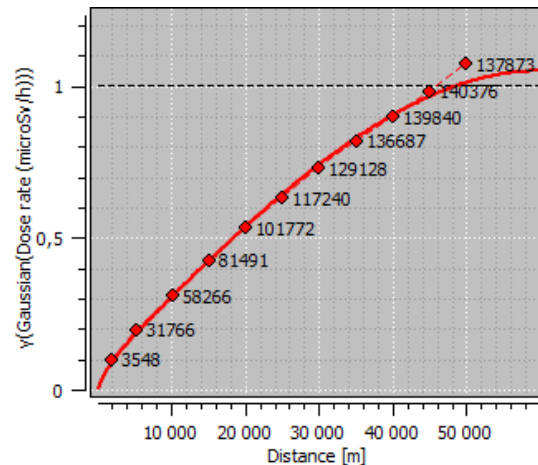
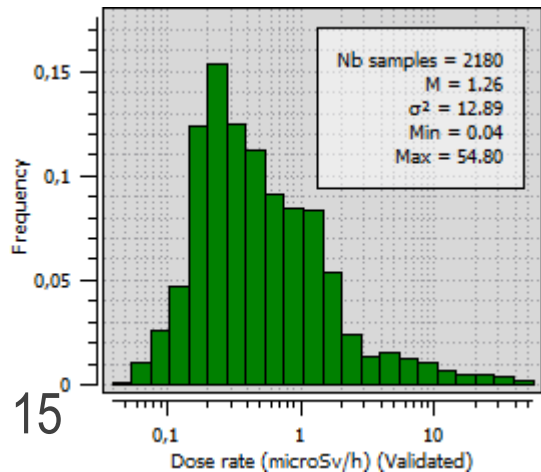
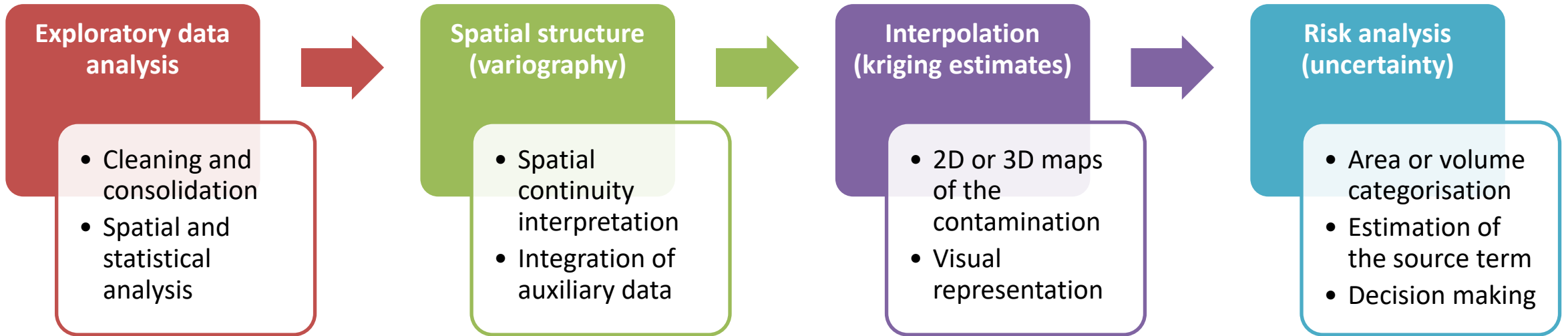
2 Application cases

3 Advanced use of the variogram

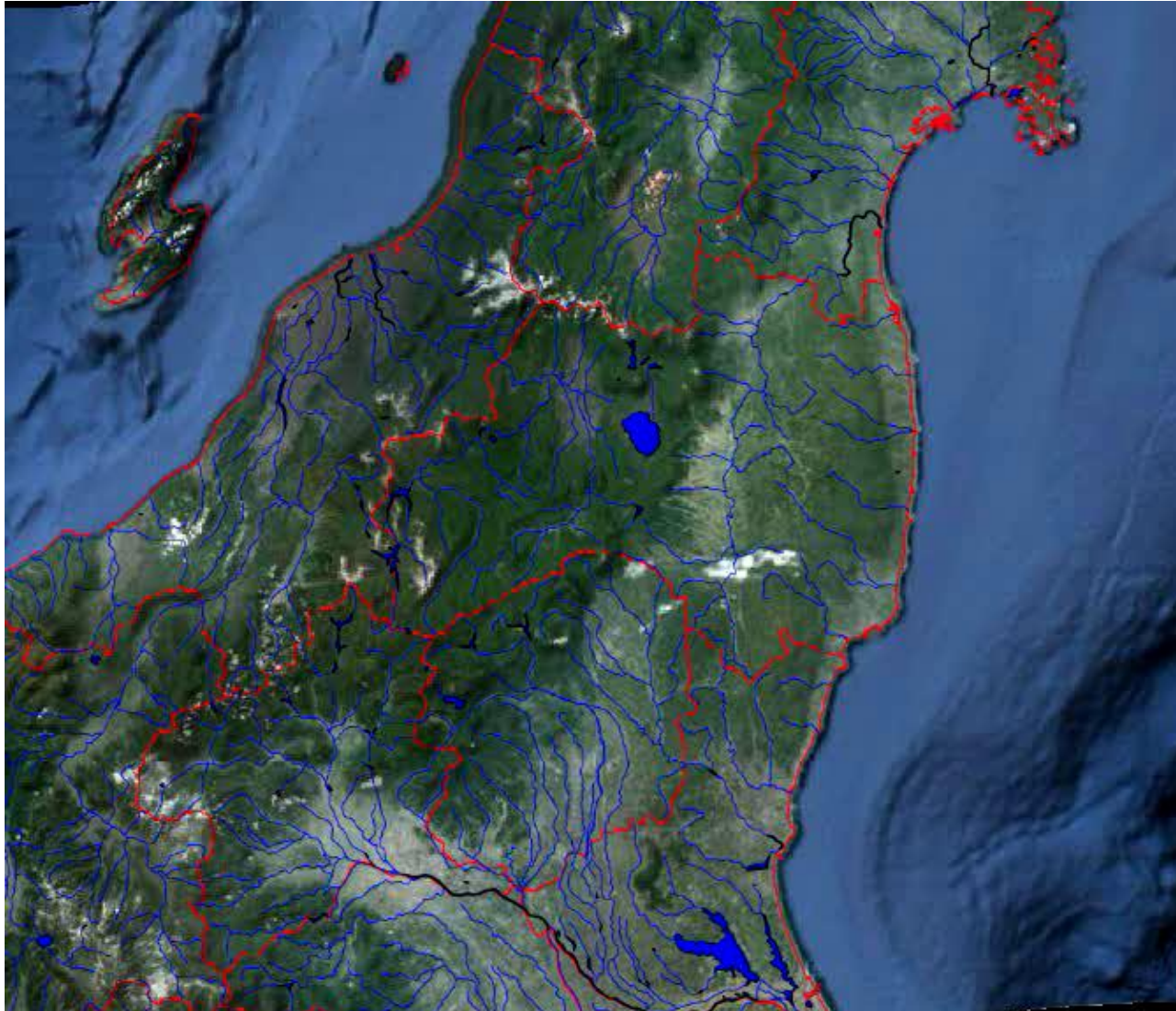
4 Impact on sampling strategy

5 Conclusions

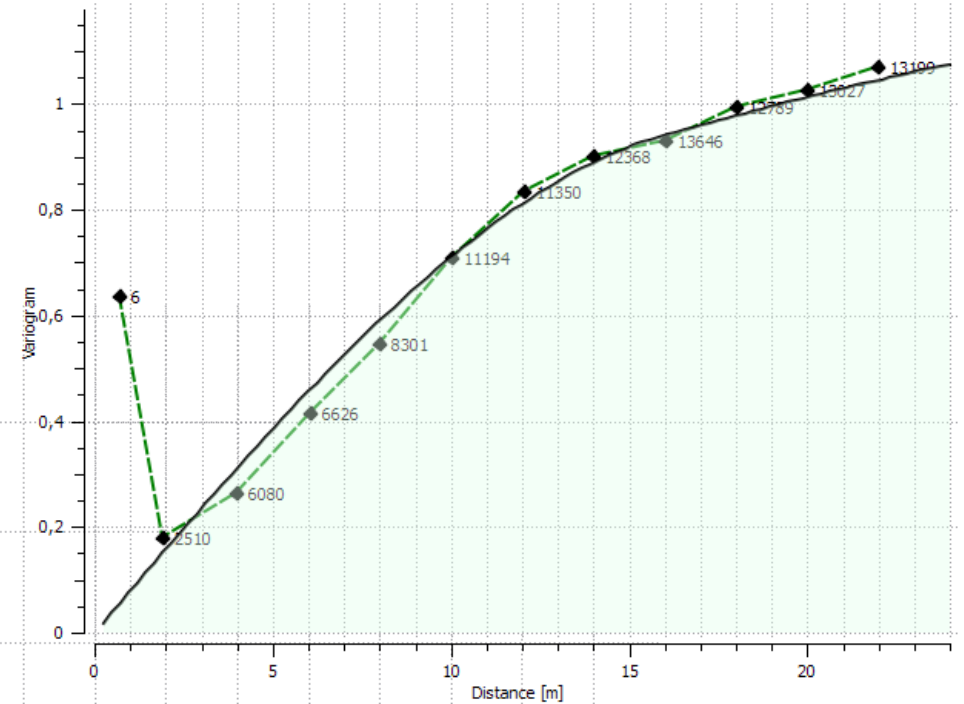
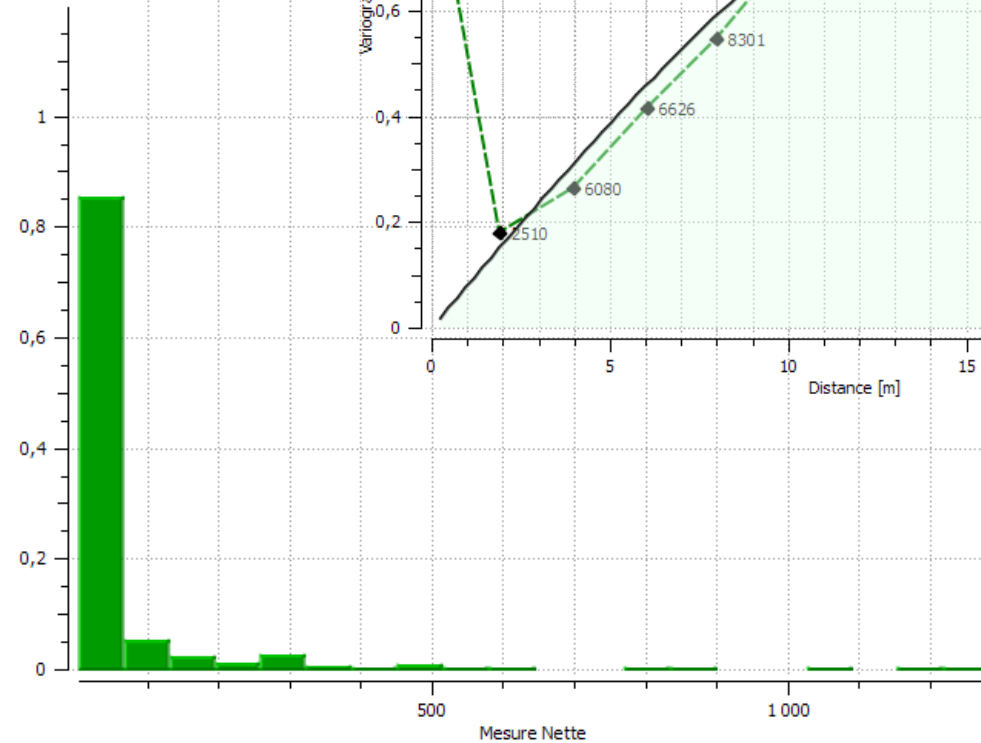
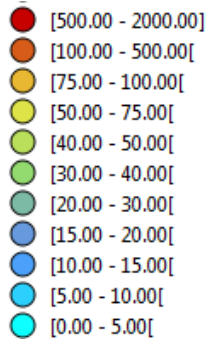
Geostatistics workflow on Fukushima (large scale)



Application case: Fukushima



Application case: RM2 facility



Application case: RM2 facility

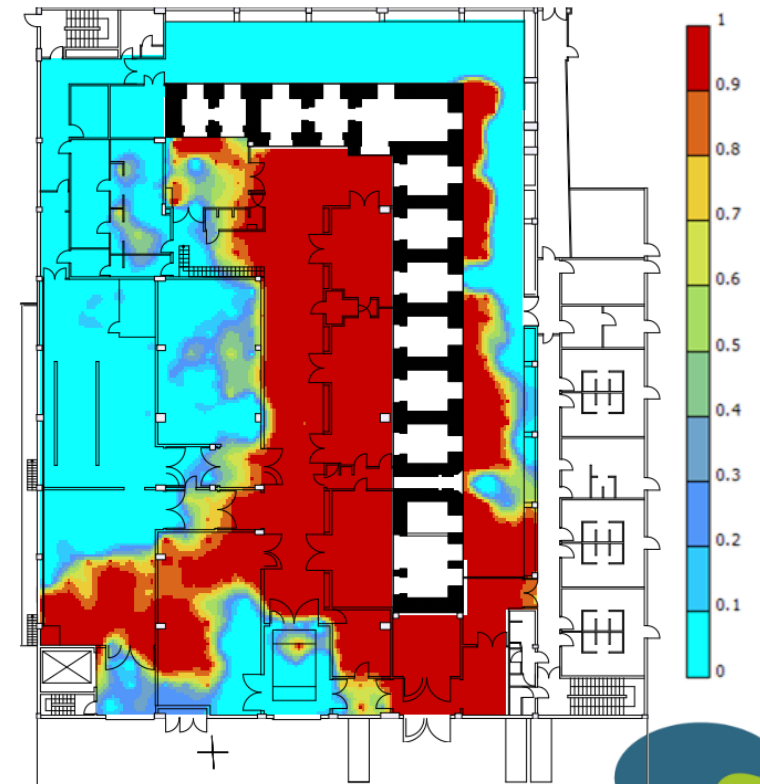
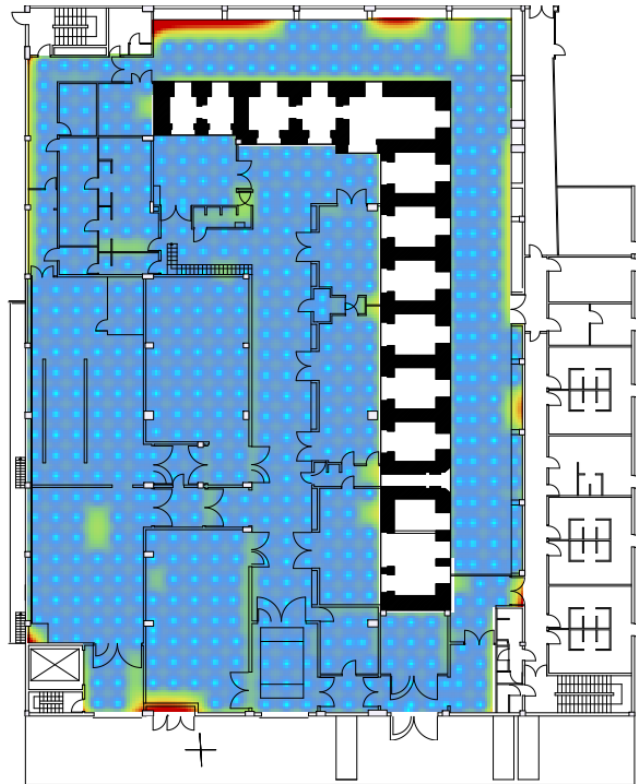
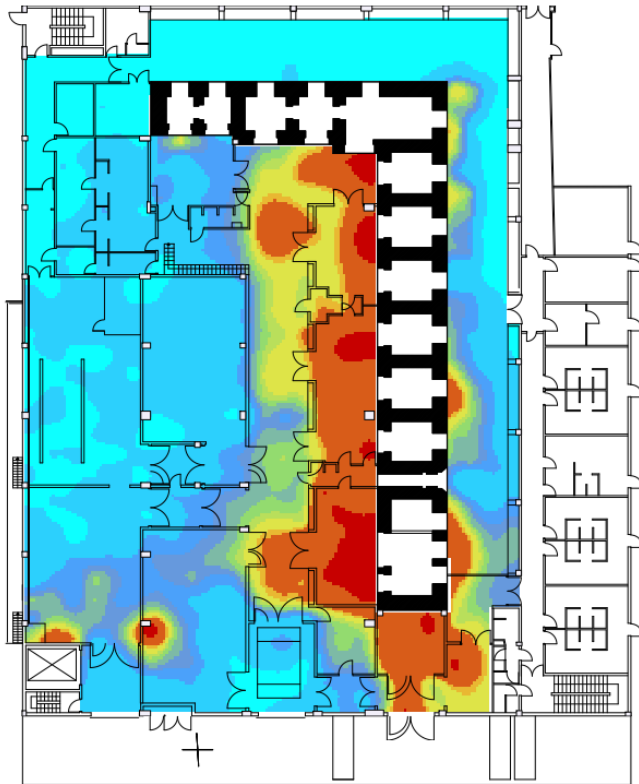
Interpolation

+

Uncertainty

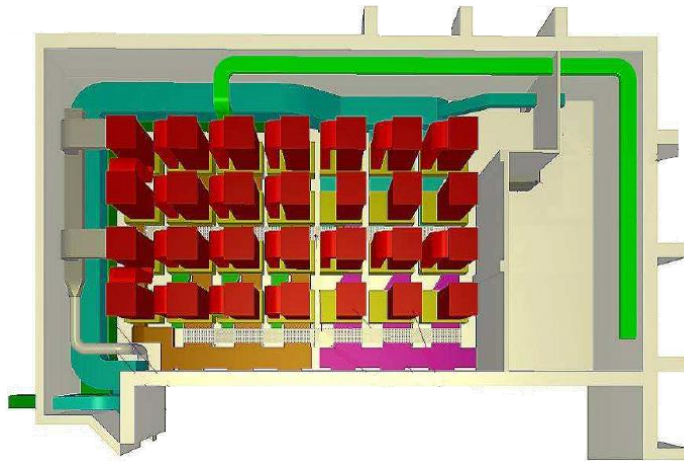
=

Classification



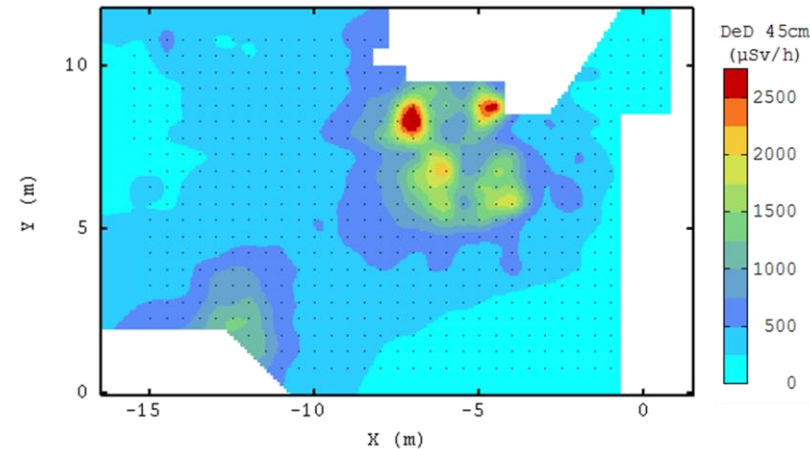
Filter room MAR200 (CEA Marcoule)

Historical

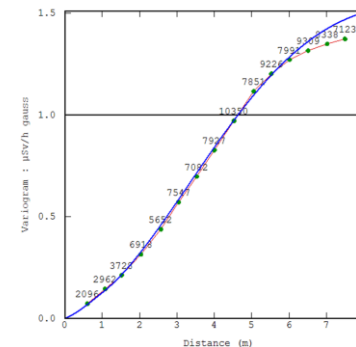


170 m², concrete slab
30 cm thickness
Two major events of
contamination

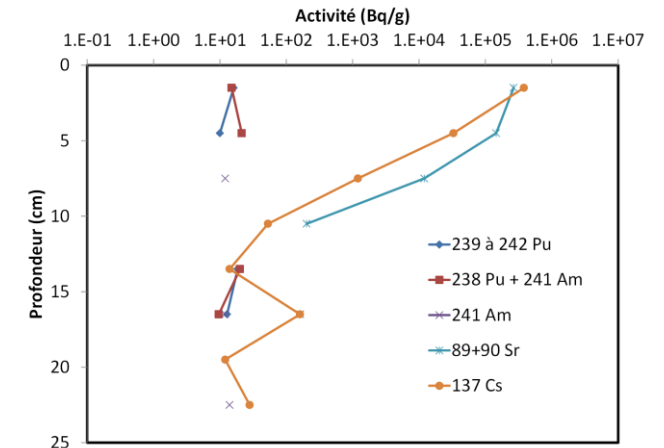
Dose rate



545 measures,
remotely
(50 cm mesh)



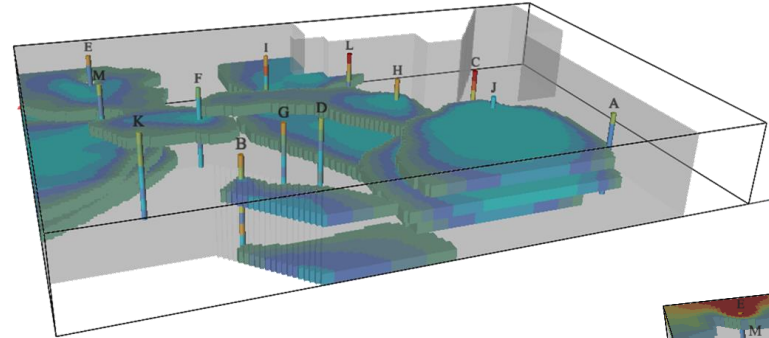
Samples



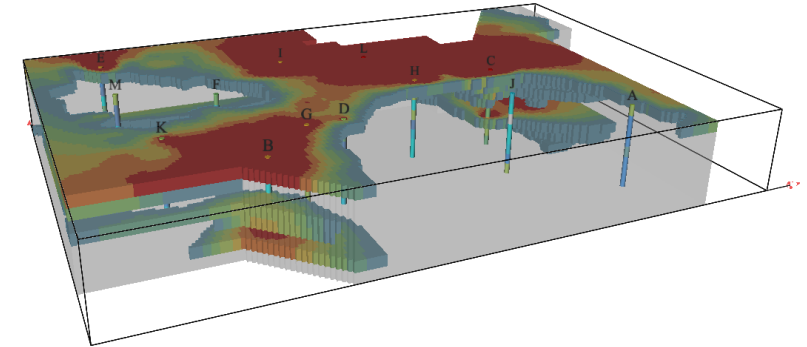
13 points in 3 phases
Migration profiles
Correlation between
nuclides

Filter room MAR200 (CEA Marcoule)

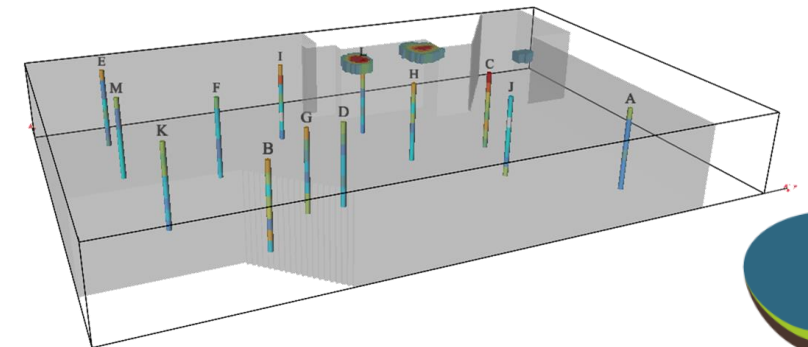
Waste categorisation



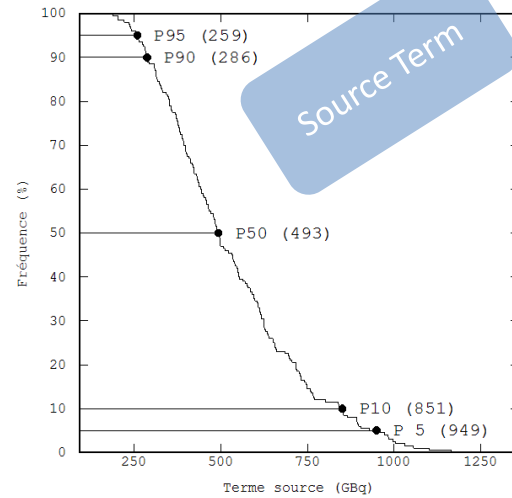
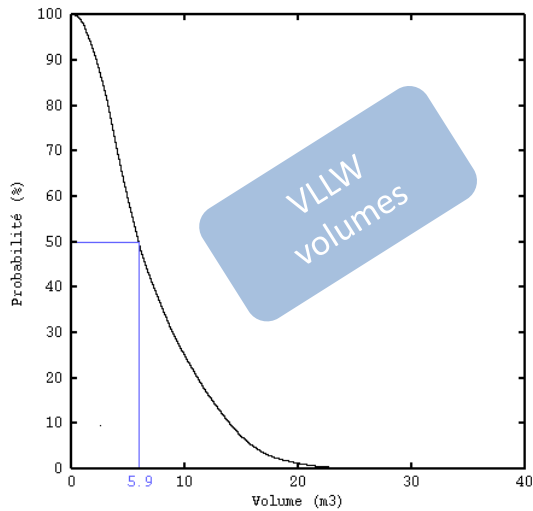
VLLW/LLW



LLW/ILW

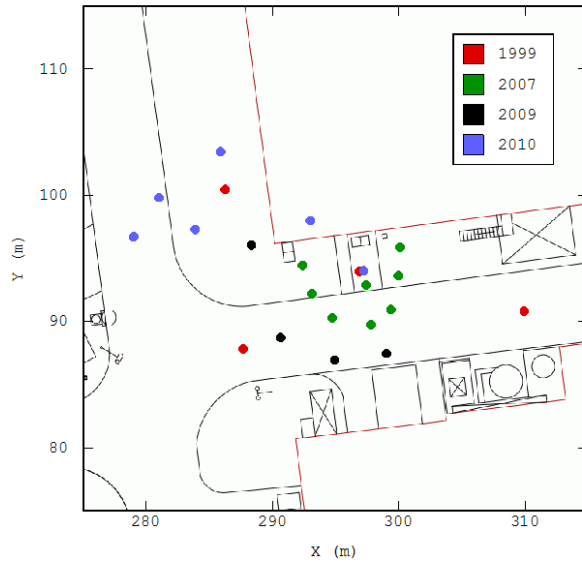


ILW/HLW



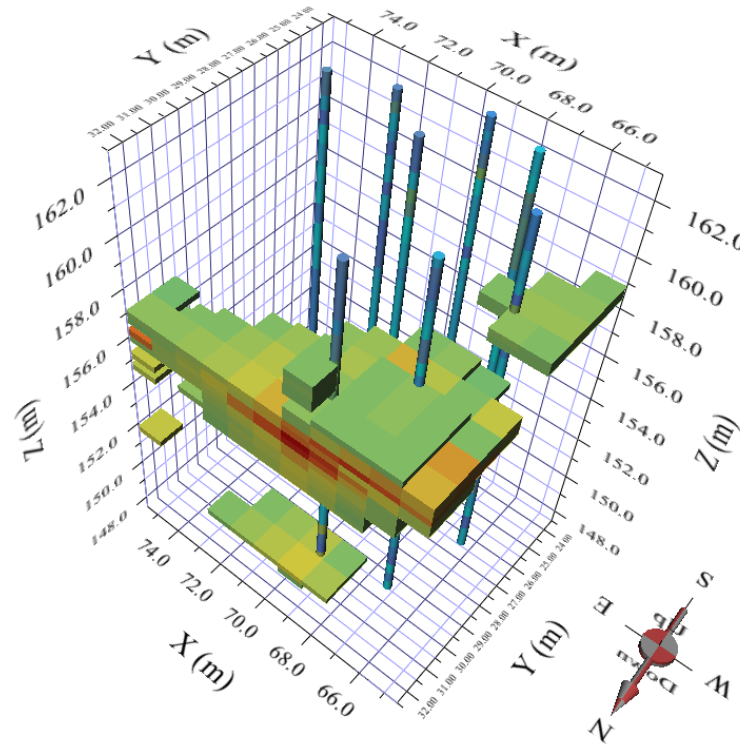
Former moat (CEA Fontenay-aux-Roses)

Historical



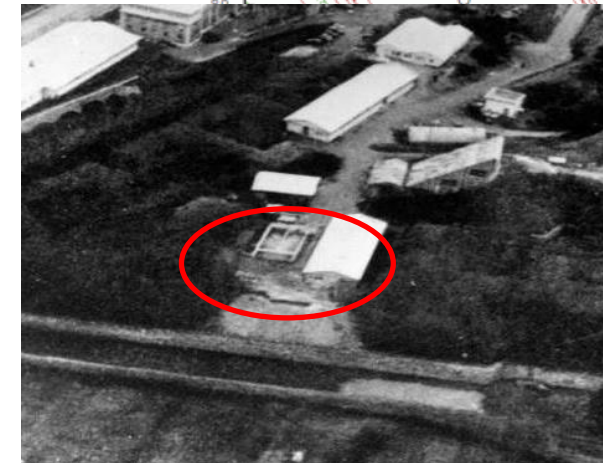
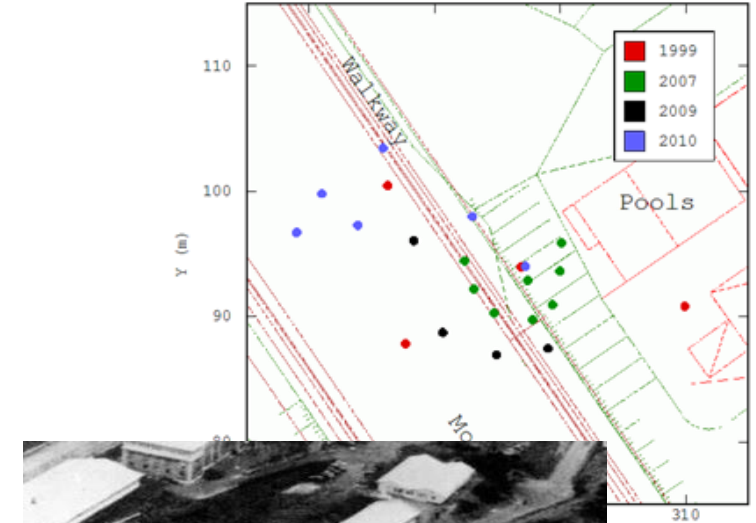
Presence of Cs-137
at 3-4 m depth in 1999

Investigations



Volumes with
existing data in 2007

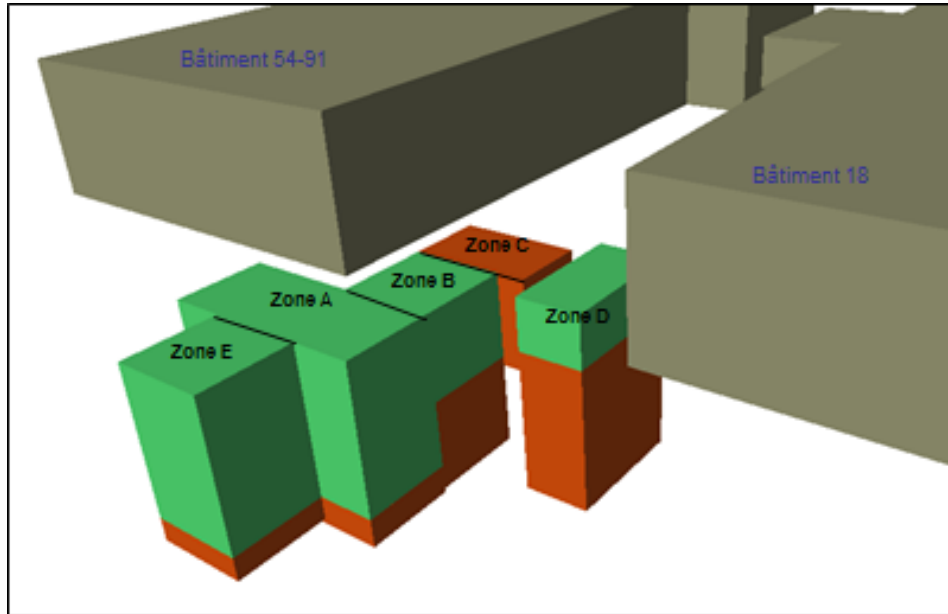
Historical (bis)



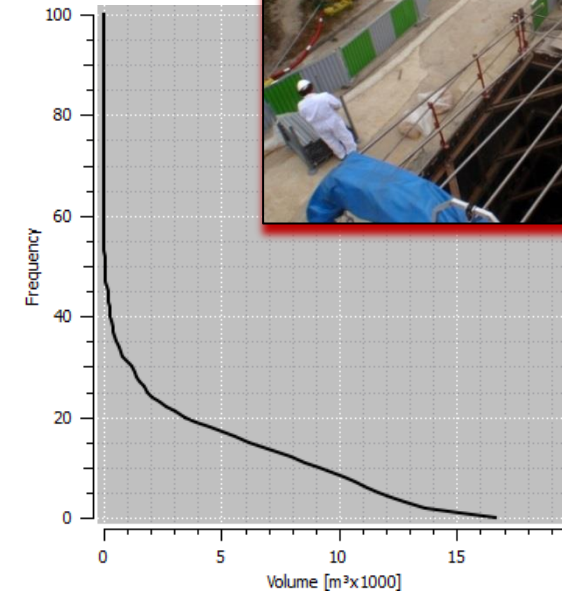
Former moat (CEA Fontenay-aux-Roses)



Former moat (CEA Fontenay-aux-Roses)



- 2000 m³ of conventional waste
- 2000 m³ of Very-Low-Level waste



Smart use of the variogram to explore spatial data, to break down variance contributions and to model radiological contaminations



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1 Variogram presentation

2 Application cases

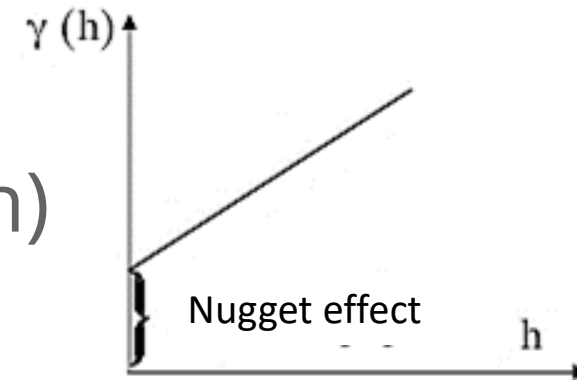
3 Advanced use of the variogram

4 Impact on sampling strategy

5 Conclusions

Precisions on nugget effect

Random contribution (no spatialisation)



Possible origins of nugget effect

Phenomena

- Huge heterogeneity at small scale
- Perturbation of the environment during the sampling

Metrology

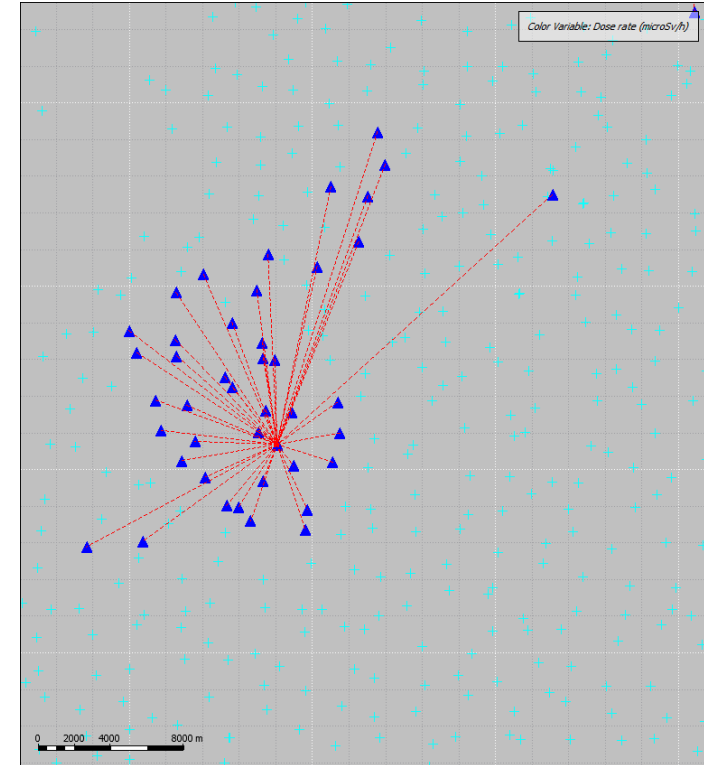
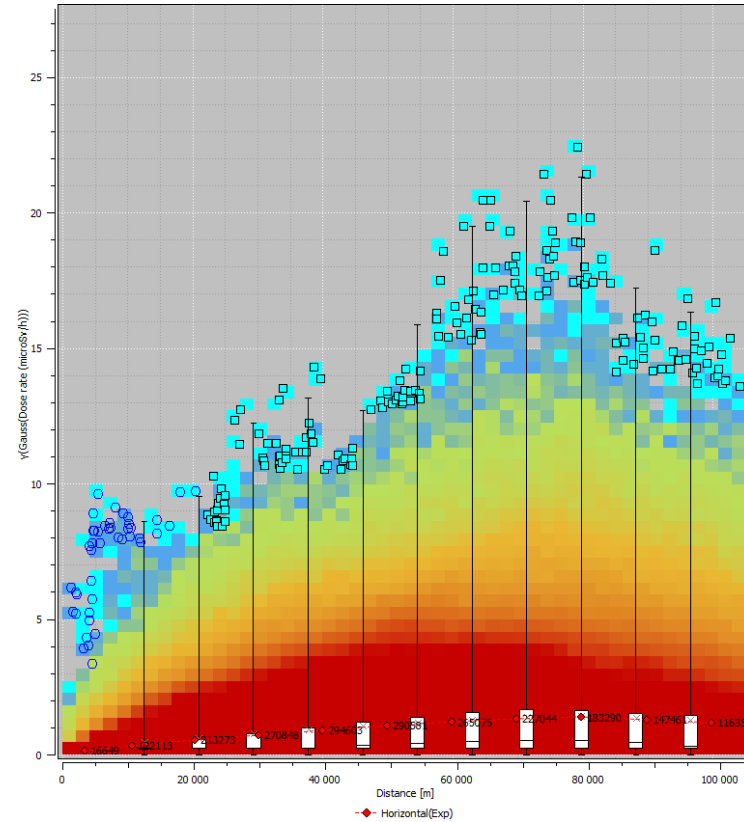
- Measurement uncertainty or laboratory uncertainty
- Successive measurements are not identical (temporal fluctuation for dose rate for instance)

Location

- Errors and uncertainties of positioning (GPS)
- Height variation, matrix...

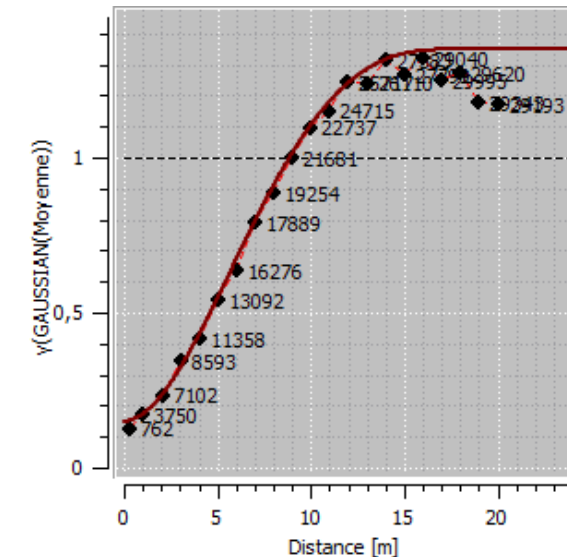
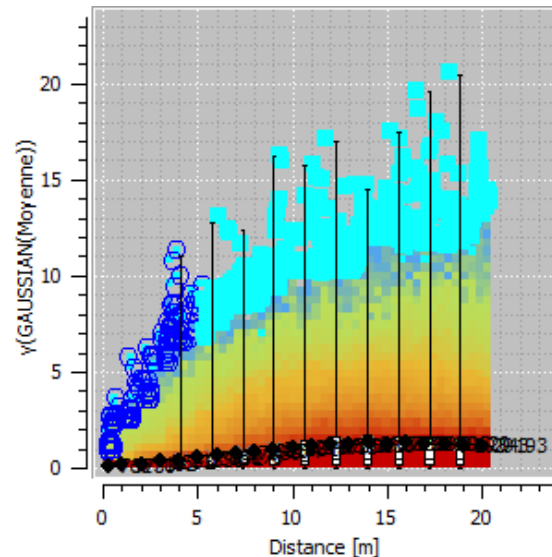
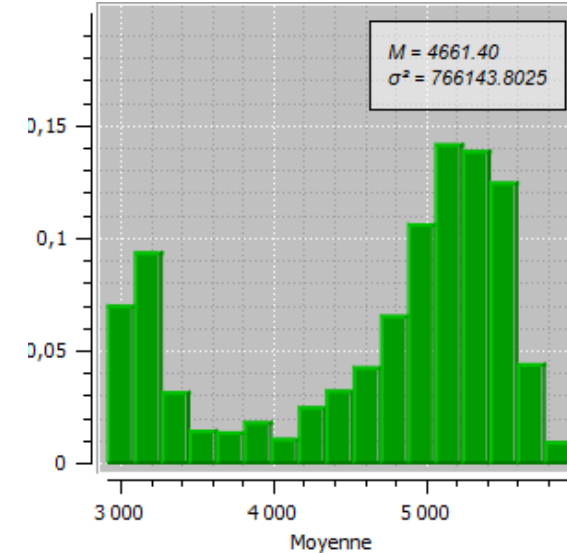
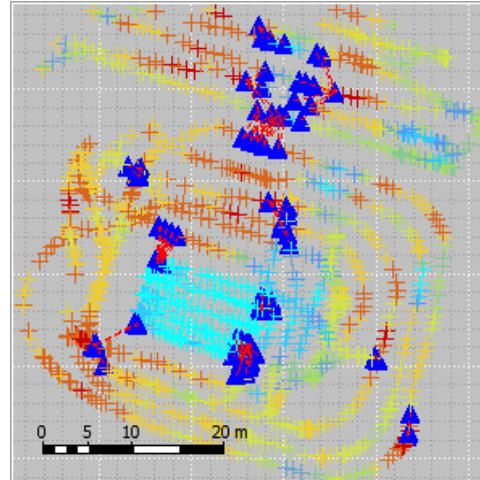
Identification of spatial outliers

Spatially close pairs of points but with high variability on the variogram cloud

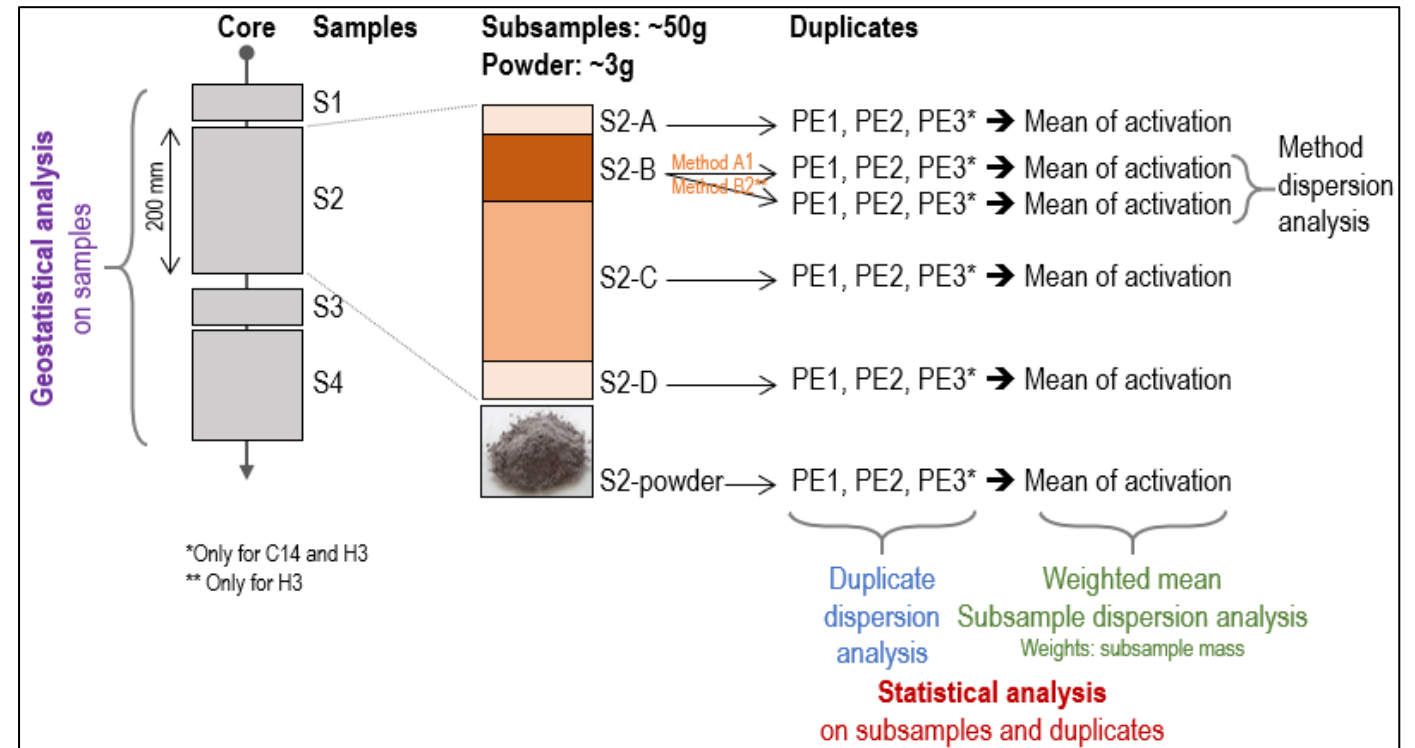
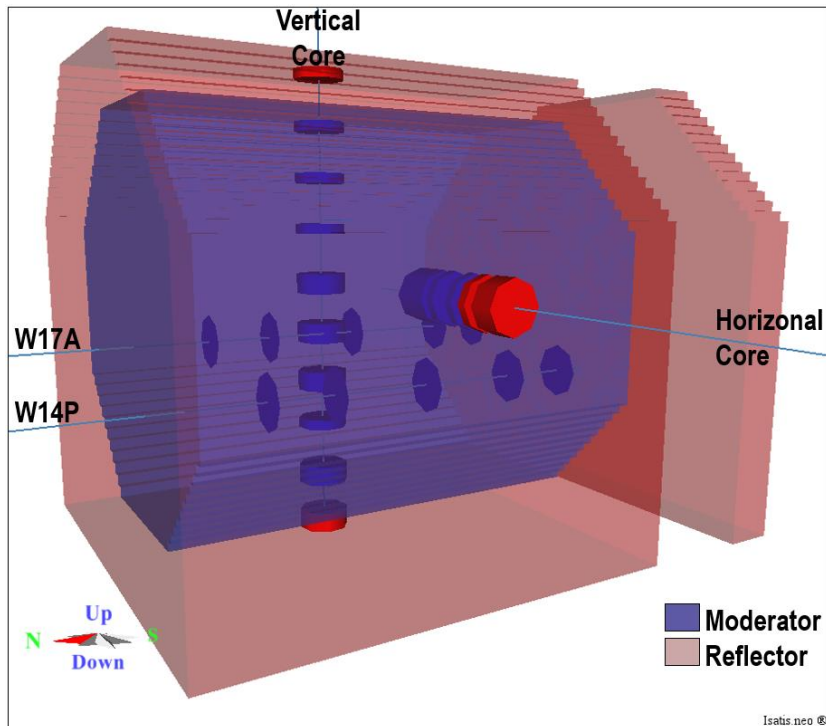


Back to exploratory data analysis

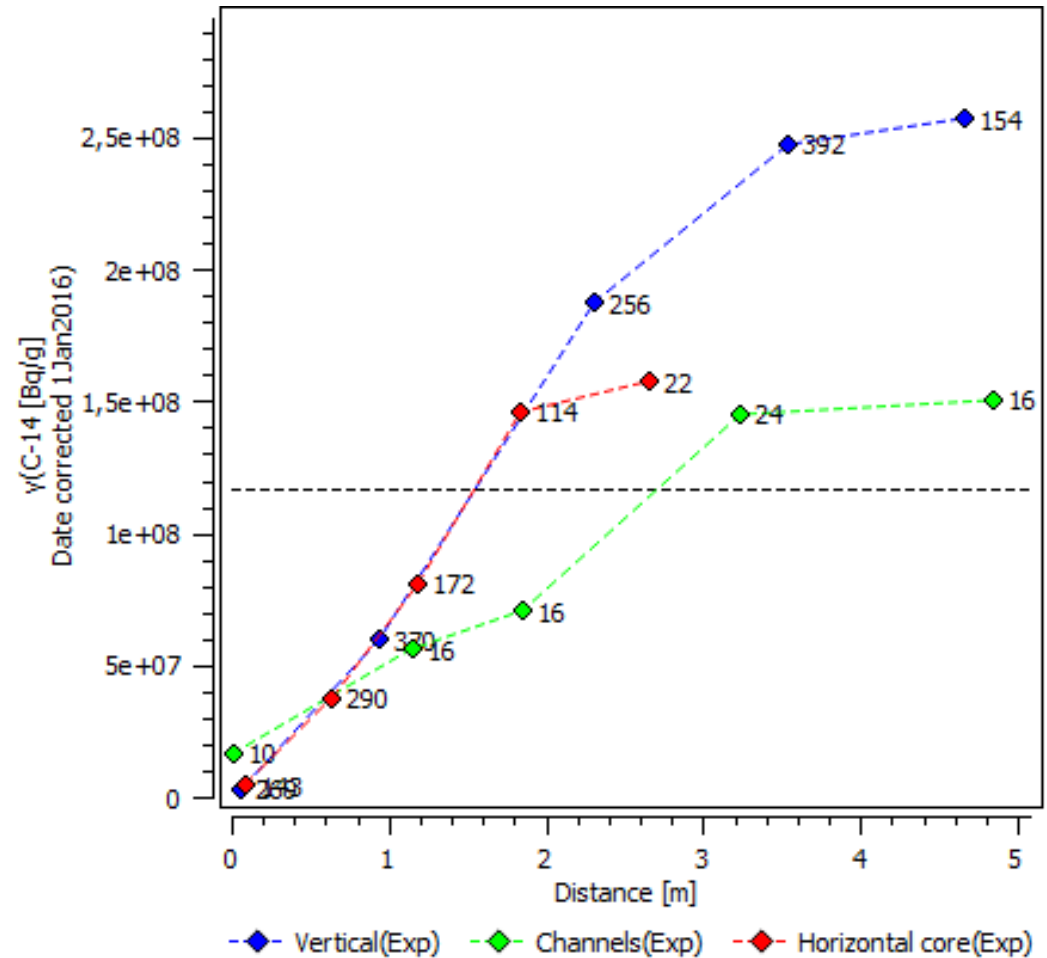
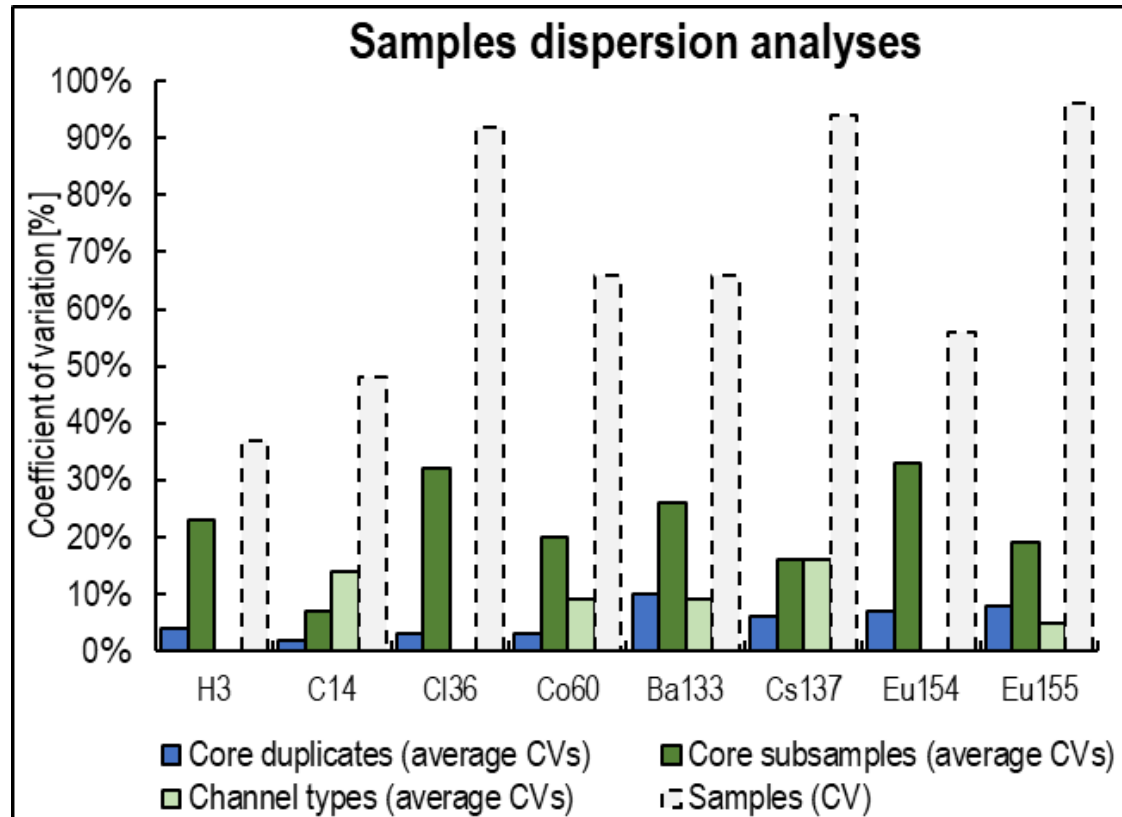
Soil composition and GPS issues



Variance decomposition (CEA G2 dataset)



Variance decomposition (CEA G2 dataset)



Variance decomposition (Eurachem dataset)

100 Samples (p45)

Table A2.1: Measured lead concentrations at each target on the sampling grid (mg kg^{-1}), shown by the actual coordinates used in the regular sampling grid (spacing 30 m) [56]. They show a high degree of variability between-locations of roughly a factor of 10. The variability within 10 of these locations selected at random (i.e. A4, B7, C1, D9, E8, F7, G7, H5, I9 and J5) was used for the estimation of uncertainty from sampling (Table A2.2). This within-target variation is substantial (e.g. a factor of 2) but substantially less than the between-target variability.

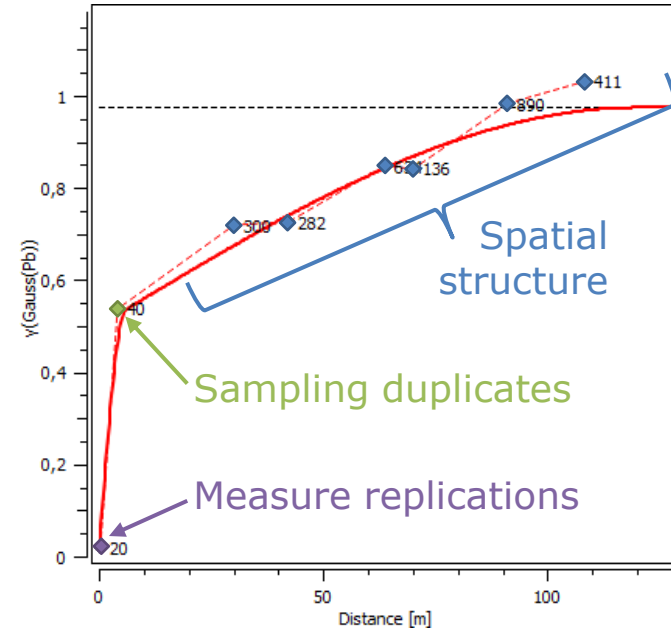
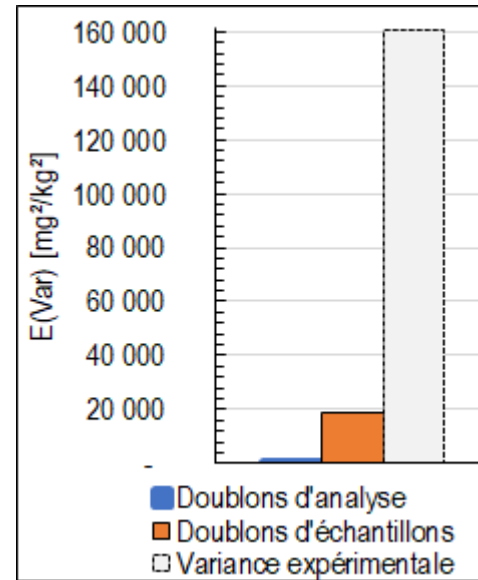
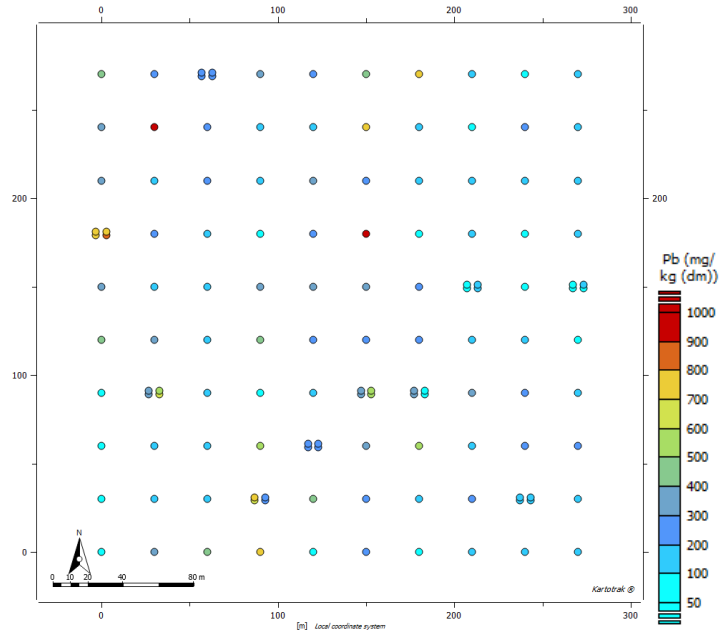
Row	A	B	C	D	E	F	G	H	I	J
1	474	287	250	338	212	458	713	125	77	168
2	378	3590	260	152	197	711	165	69	206	126
3	327	197	240	159	327	264	105	137	131	102
4	787	207	197	87	254	1840	78	102	71	107
5	395	165	188	344	314	302	284	89	87	83
6	453	371	155	462	258	245	237	173	152	83
7	72	470	194	82.5	162	441	199	326	290	164
8	71	101	108	521	218	327	540	132	258	246
9	72	188	104	463	482	228	135	285	181	146
10	89	366	495	779	60	206	56	135	137	149

2x2x10 Replicates (p46)

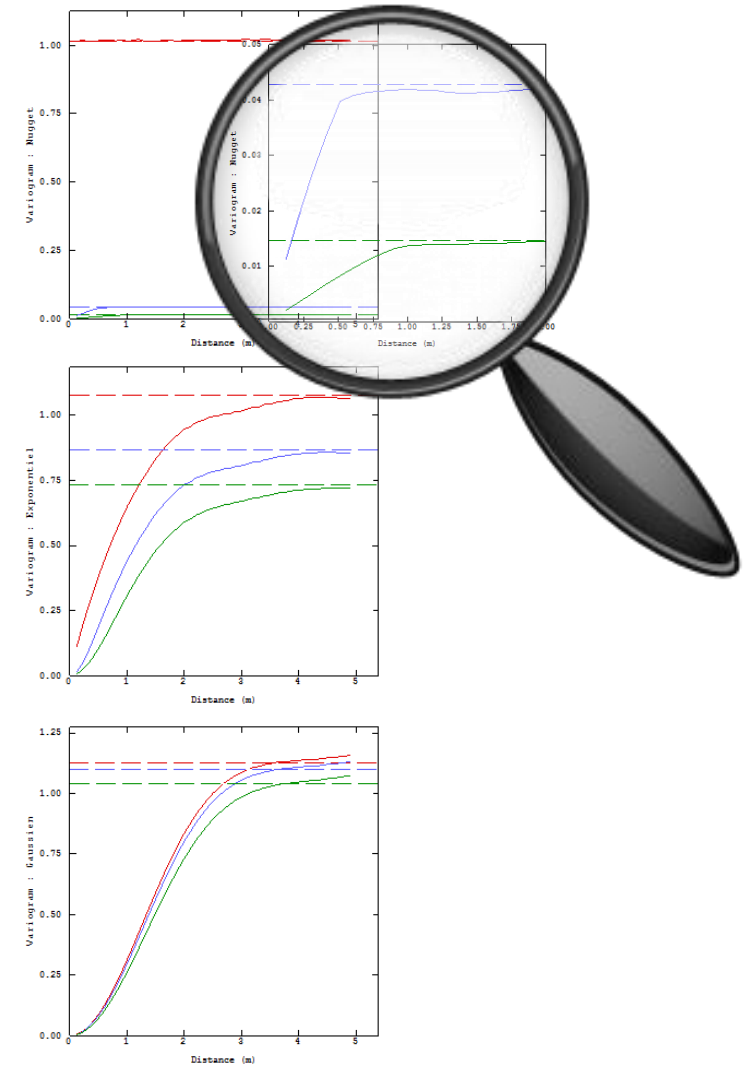
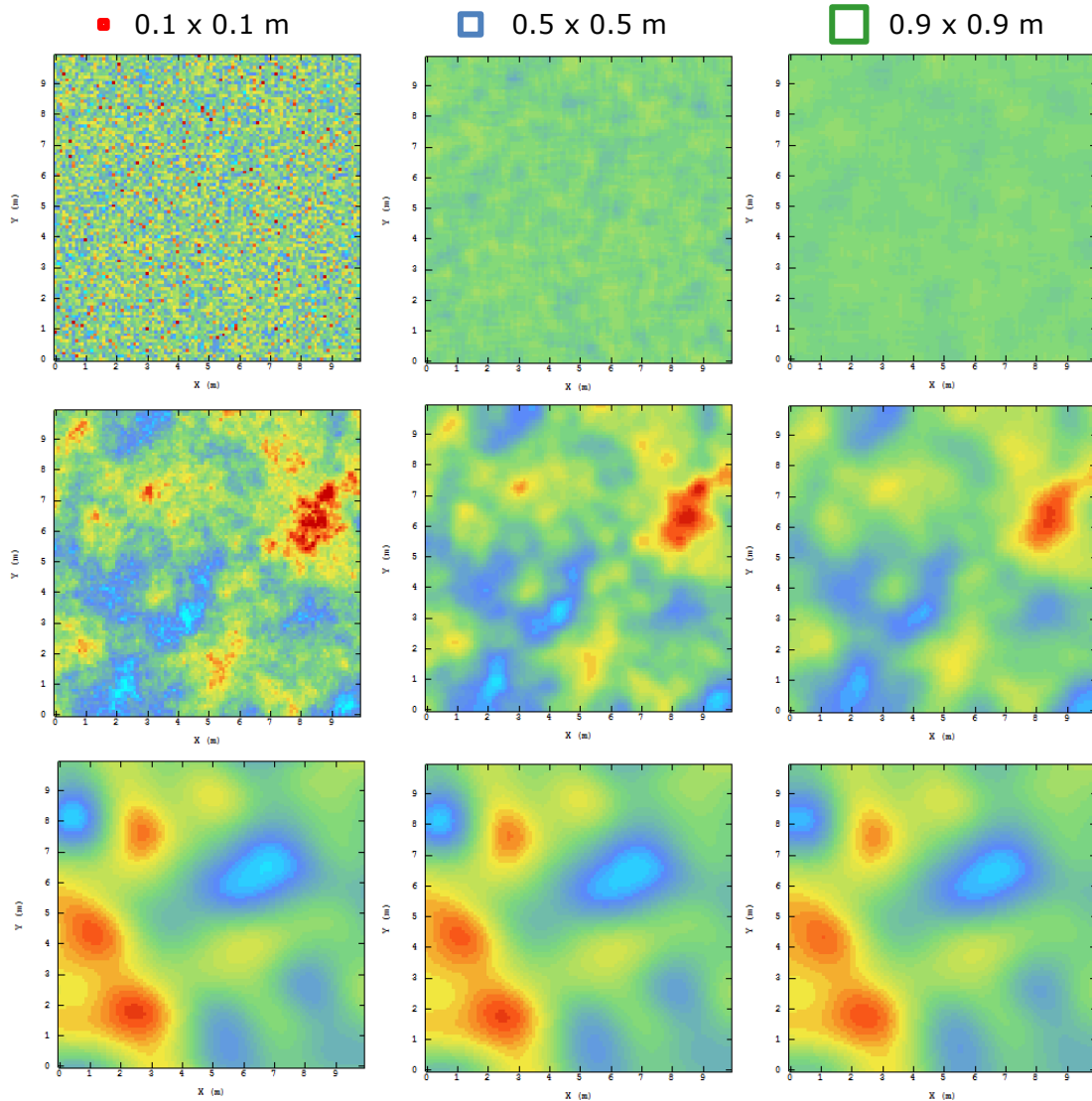
Table A2.2: Measurements of the concentration (mg kg^{-1}) of a lead on 10 duplicated samples from the total of 100 targets in a survey of contaminated land (Table A2.1) [56]. The duplicate samples are labelled S1 and S2. Likewise, duplicate analyses are labelled A1 and A2. Hence, D9S1A2 (value 702 mg kg^{-1}) is analysis 2, from sample 1 from sampling target D9. Values shown are rounded for clarity, and used for subsequent calculations, but generally un-rounded values are preferable for these calculations.

Sample target	S1A1	S1A2	S2A1	S2A2
A4	787	769	811	780
B7	338	327	651	563
C1	289	297	211	204
D9	662	702	238	246
E8	229	215	208	218
F7	346	374	525	520
G7	324	321	77	73
H5	56	61	116	120
I9	189	189	176	168
J5	61	61	91	119

Variance decomposition (Eurachem dataset)

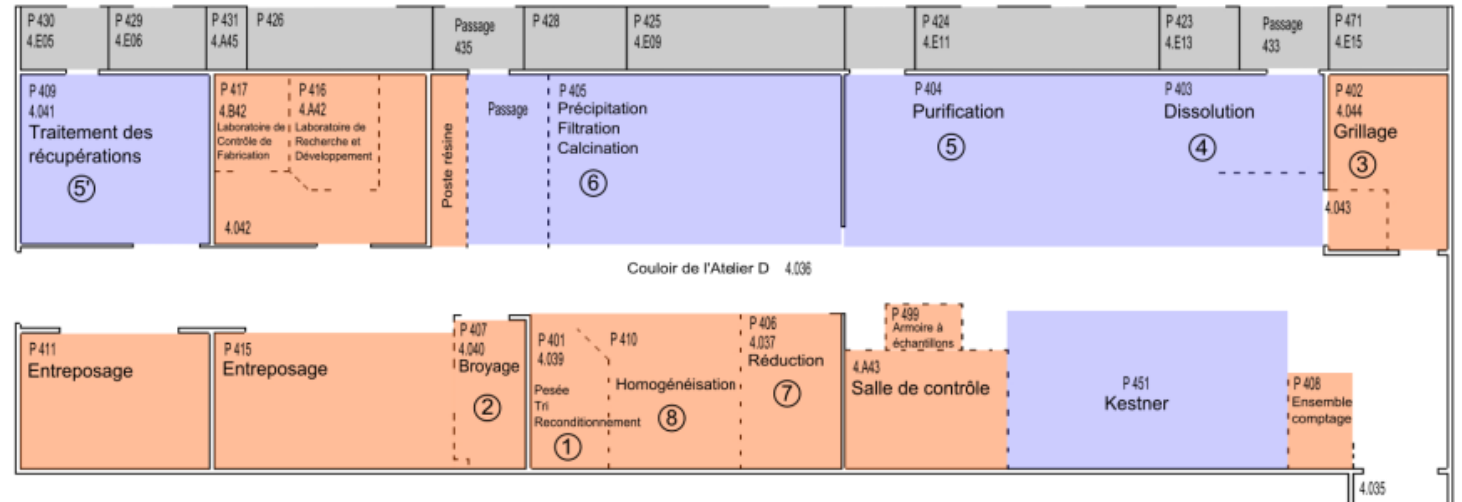


Support effect and spatial structure



Multivariate geostatistics: Case presentation

Workshop D of ATUE facility (CEA Cadarache)
 Historical and functional analysis



Production of metallic uranium, then uranium oxides

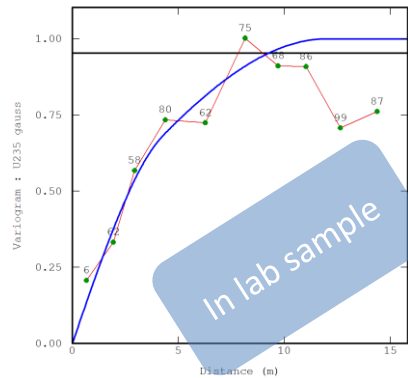
Two physical states of uranium : **dry way** / **wet way**

800 m²

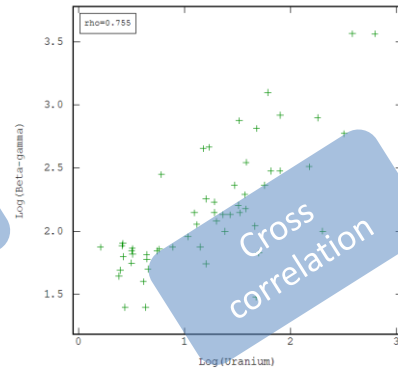
Multivariate geostatistics: Variography analysis

Coupling between different radiological investigations

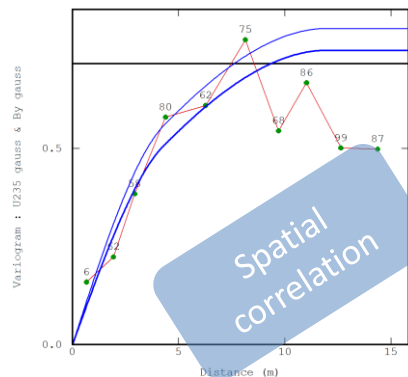
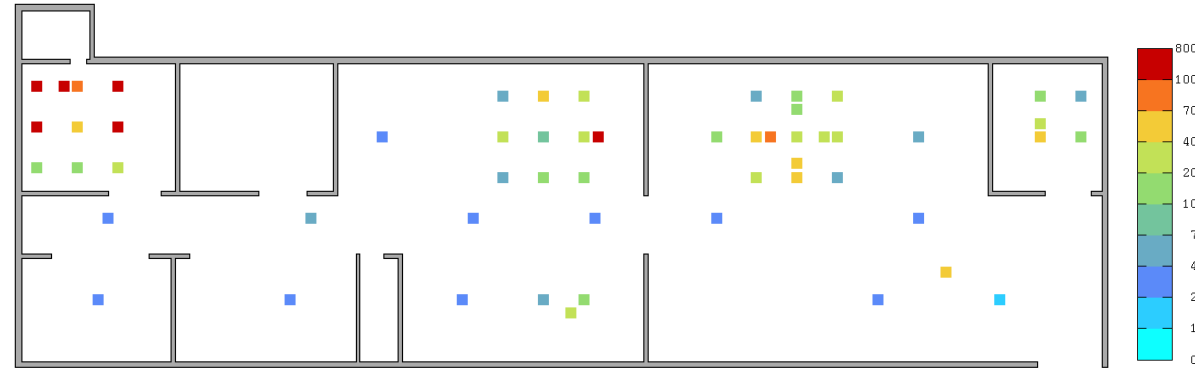
Cross-variogram:
$$\gamma_{z_1 z_2}(h) = \frac{1}{2} E\{[z_1(x+h) - z_1(x)][z_2(x+h) - z_2(x)]\}$$



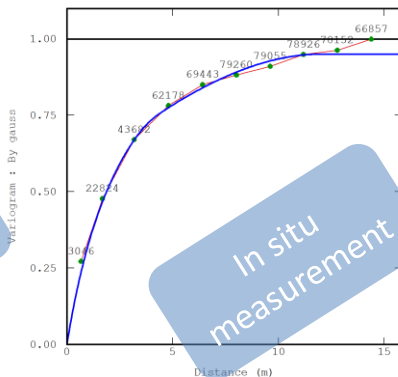
In lab sample



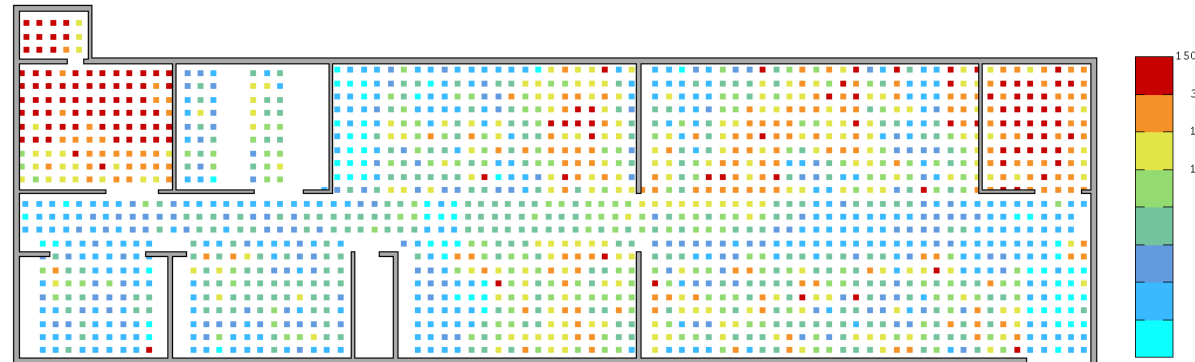
Cross correlation



Spatial correlation



In situ measurement

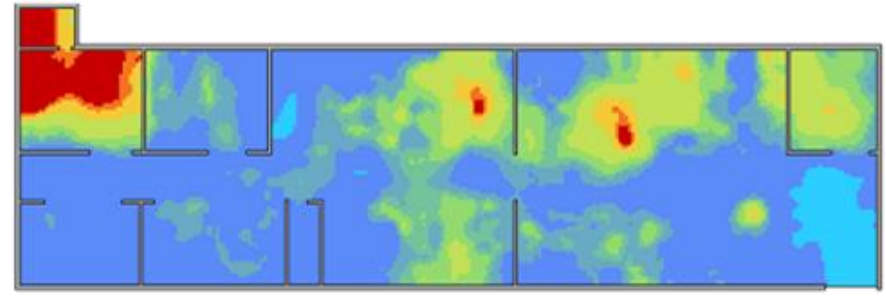
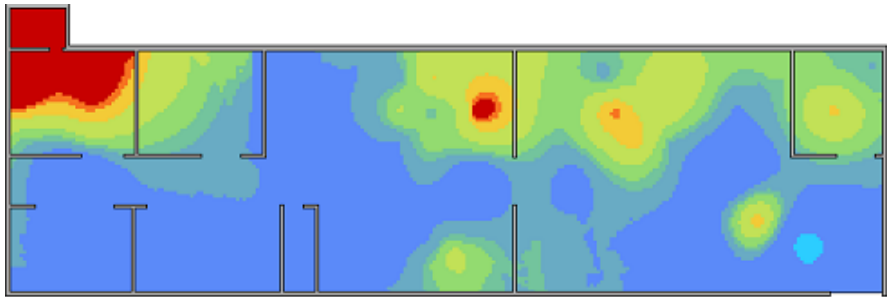


Multivariate geostatistics: Mapping comparison

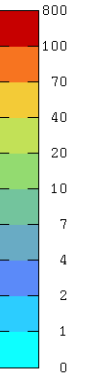
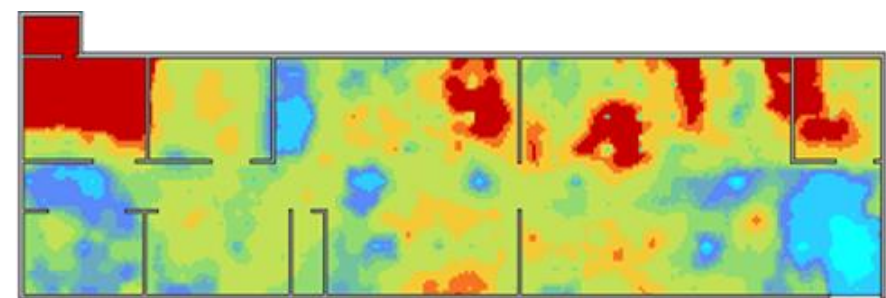
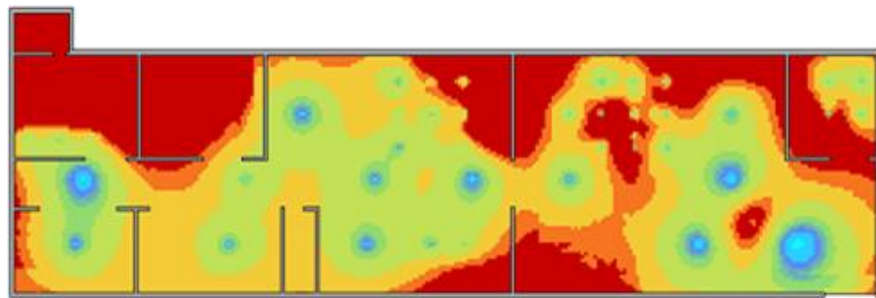
0 In lab data

+ In situ data

z^*



[..]



Smart use of the variogram to explore spatial data, to break down variance contributions and to model radiological contaminations



Geovariances
Where no one has gone before

- 1 Variogram presentation
- 2 Application cases
- 3 Advanced use of the variogram
- 4 Impact on sampling strategy**
- 5 Conclusions

Investigation optimisation

Initial mesh determination



Geometrical context

Historical analysis



Feedback on spatial structures



Target size

Adding extra measurements



Uncertainty reduction



Manual approach



Automatic algorithm

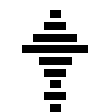
Positioning of destructive samples



Correlation with surface maps

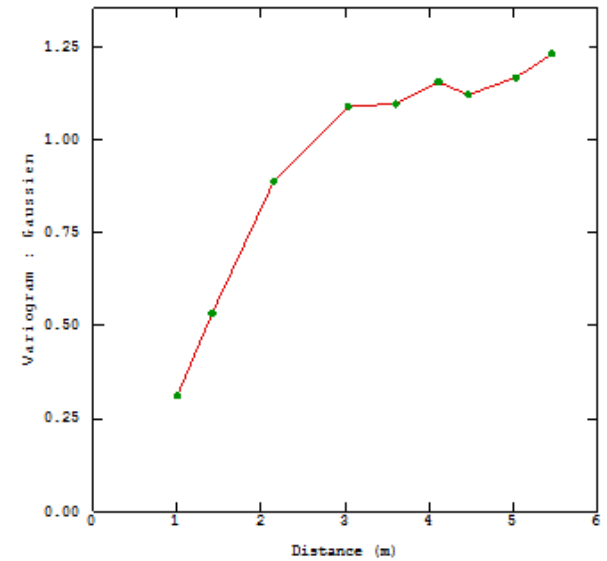
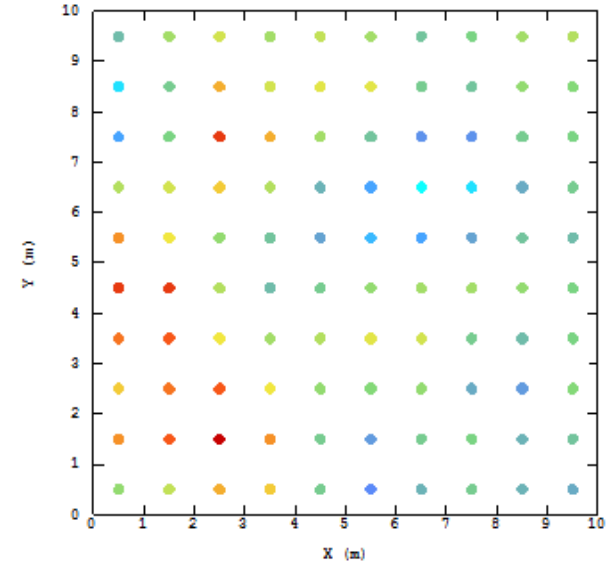
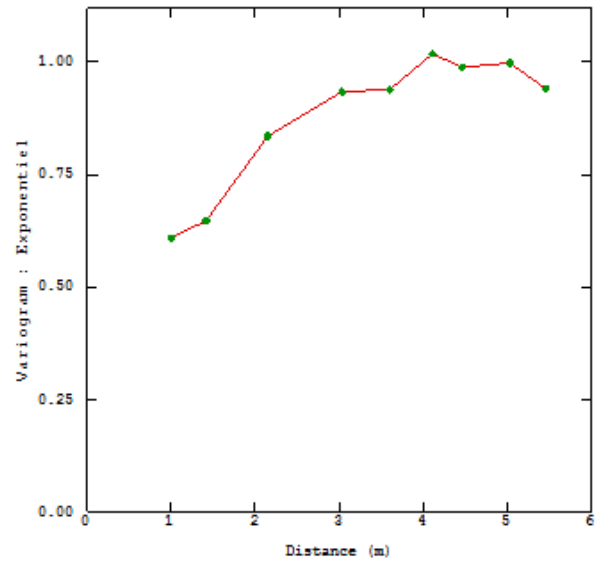
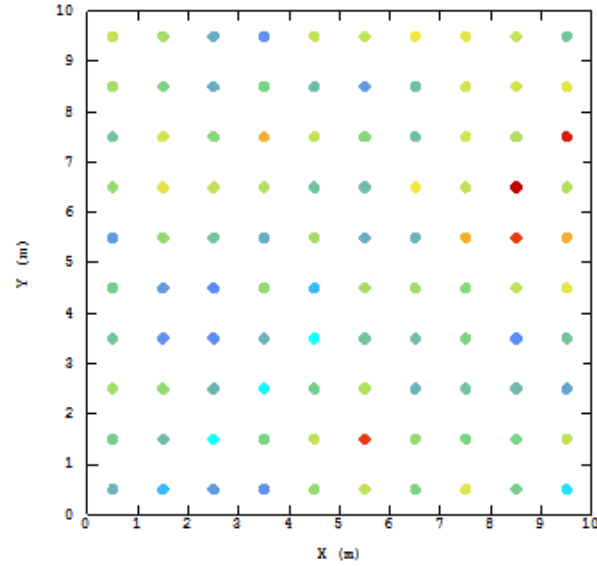
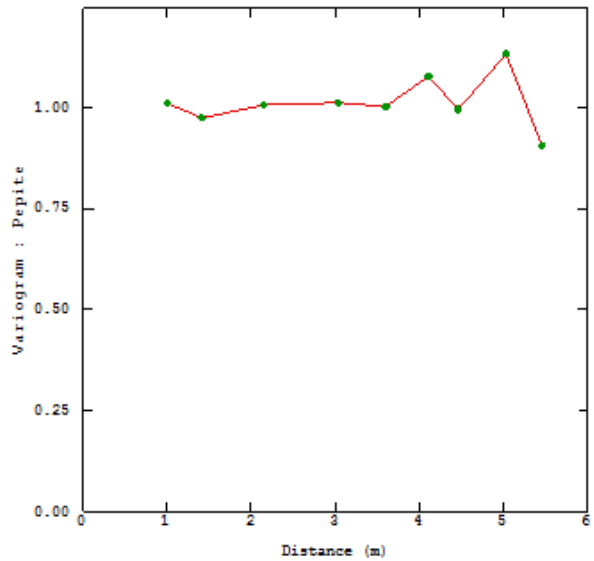
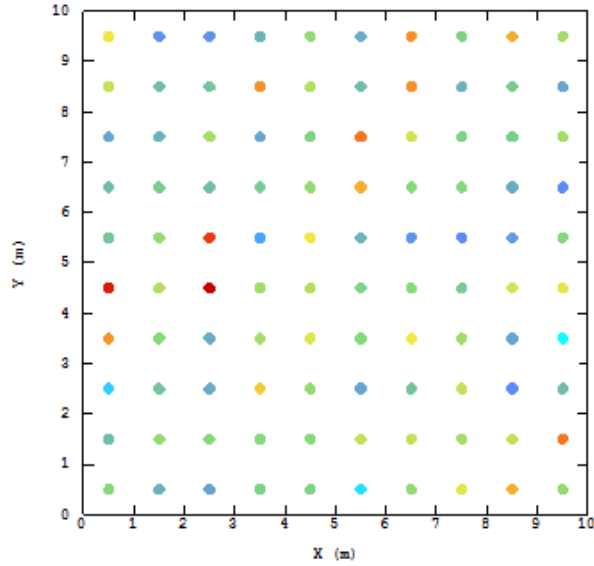


Specific vertical recommendations

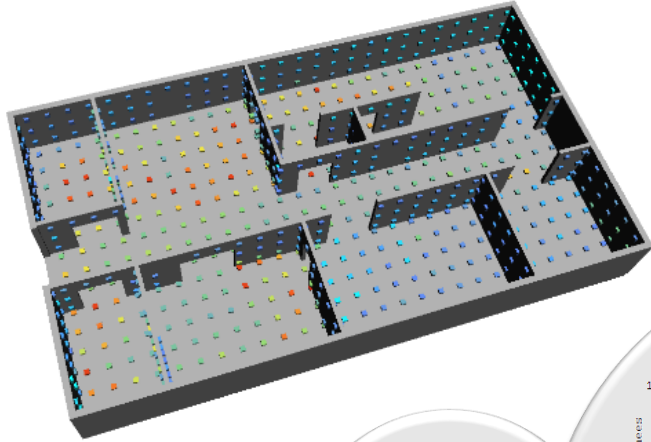


Feedback on migration profiles

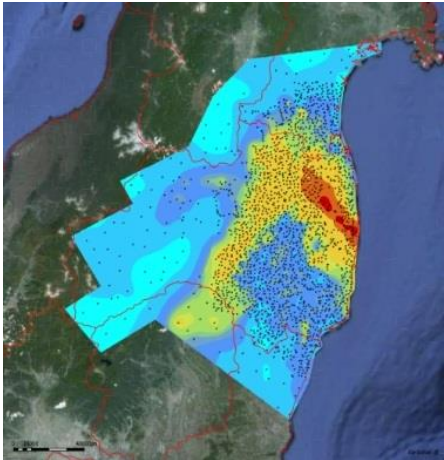
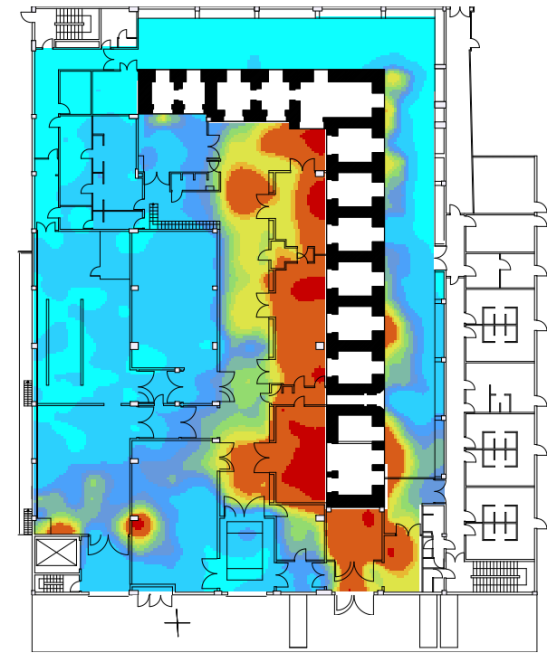
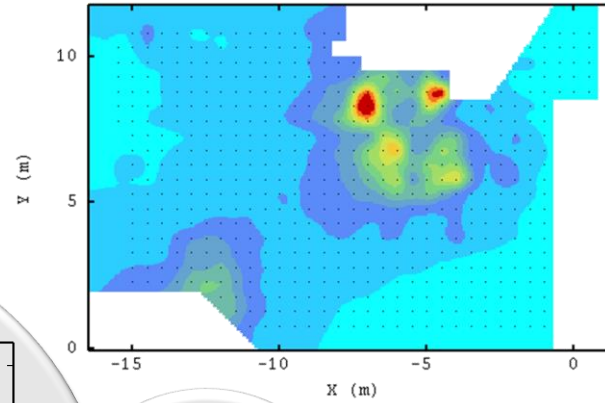
Experimental variograms with 1 m mesh



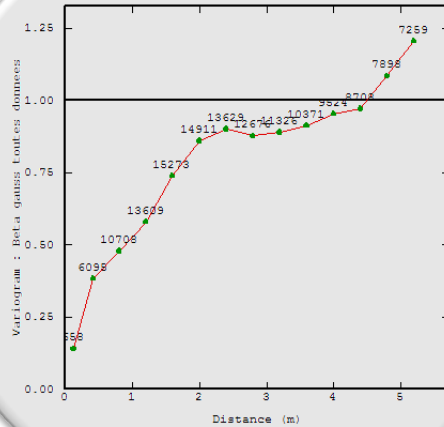
Geostatistics point of view



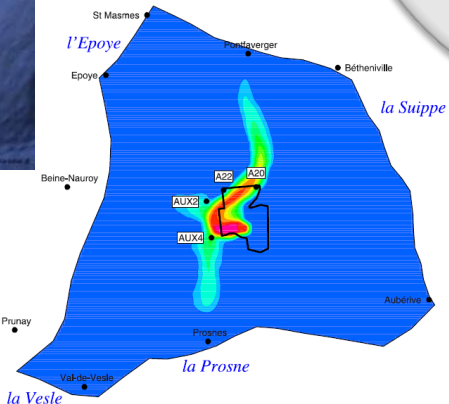
Concrete



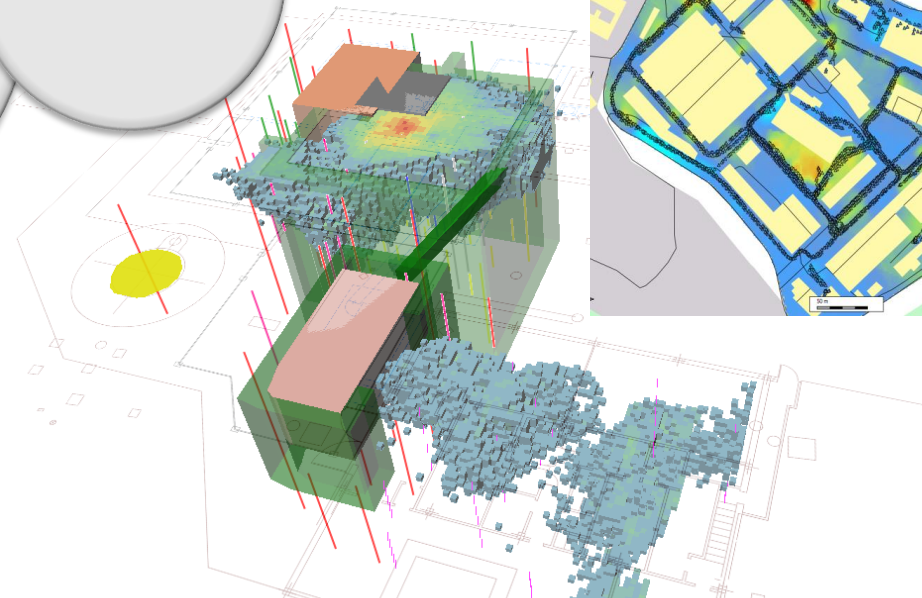
Post-accidental



Soils

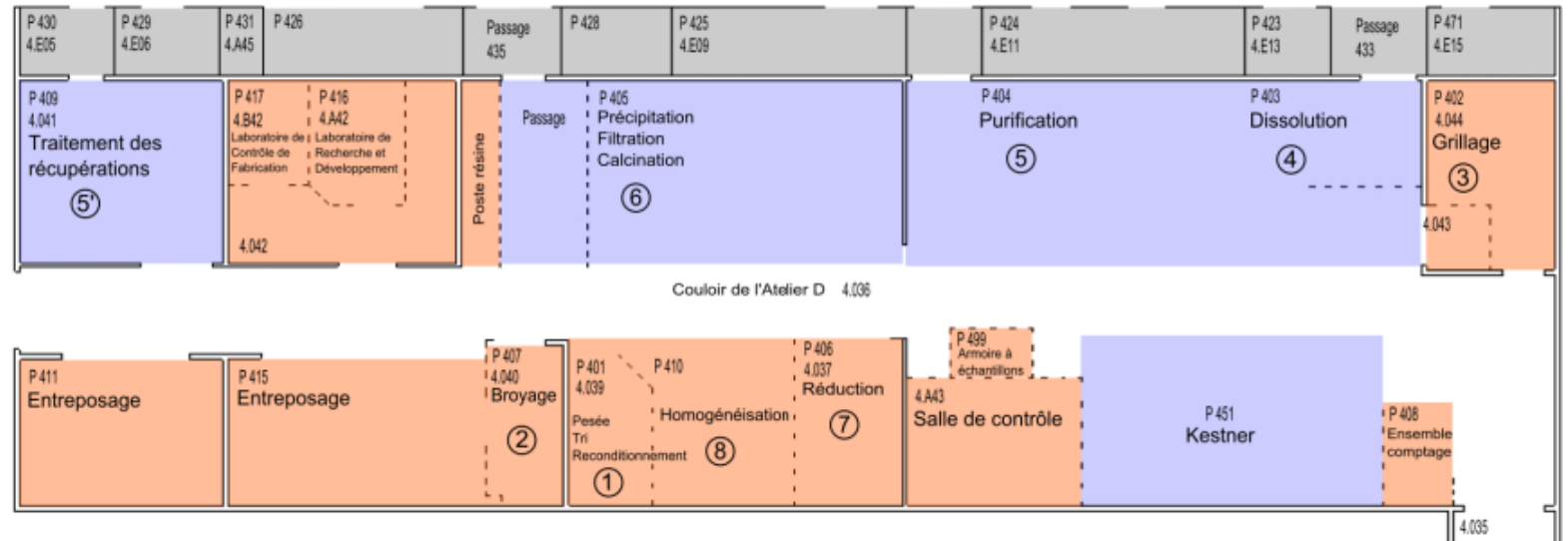


Ground water



Narrative arc: Workshop D (ATUE – CEA Marcoule)

Historical and functional analysis



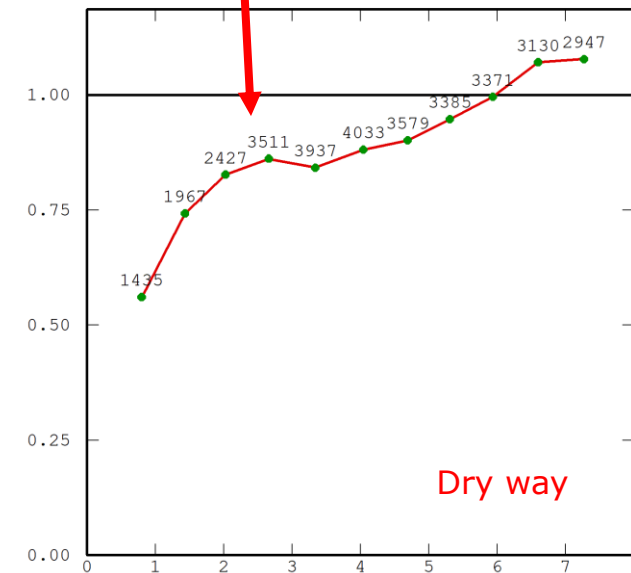
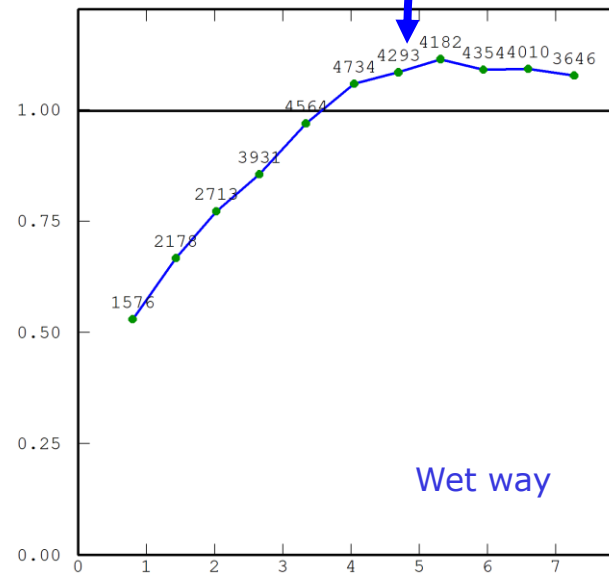
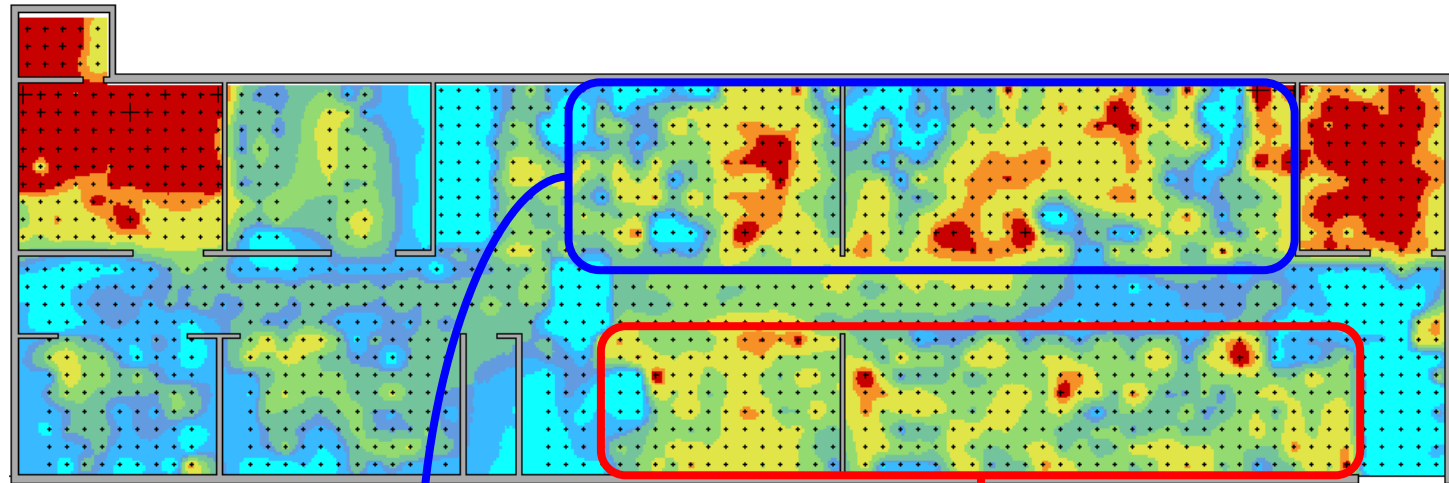
Production of metallic uranium, then uranium oxides
Two physical states of uranium : **dry way** / **wet way**
≈800 m² (54 m by 15 m)



Historical context

Different processes

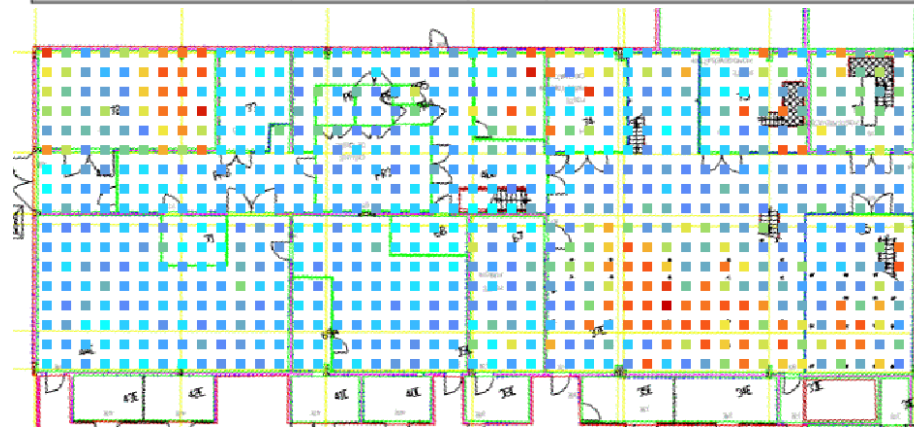
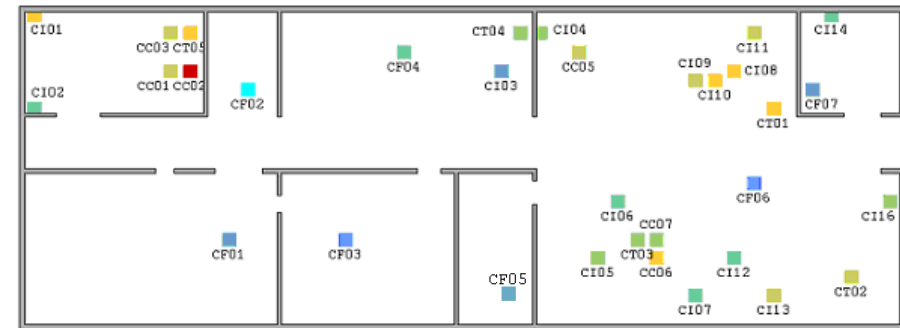
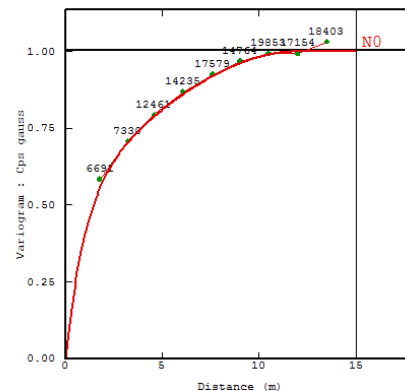
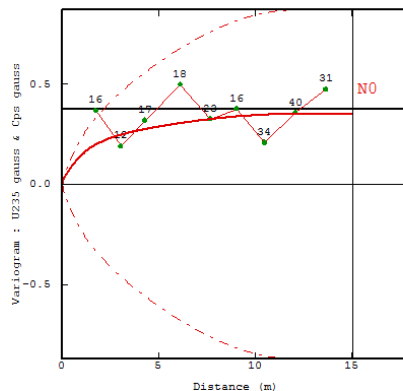
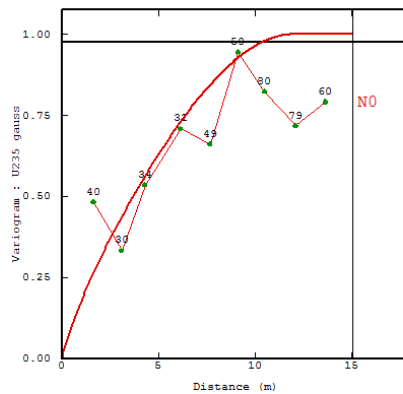
Consequences on the sampling optimisation



Positioning of destructive samples

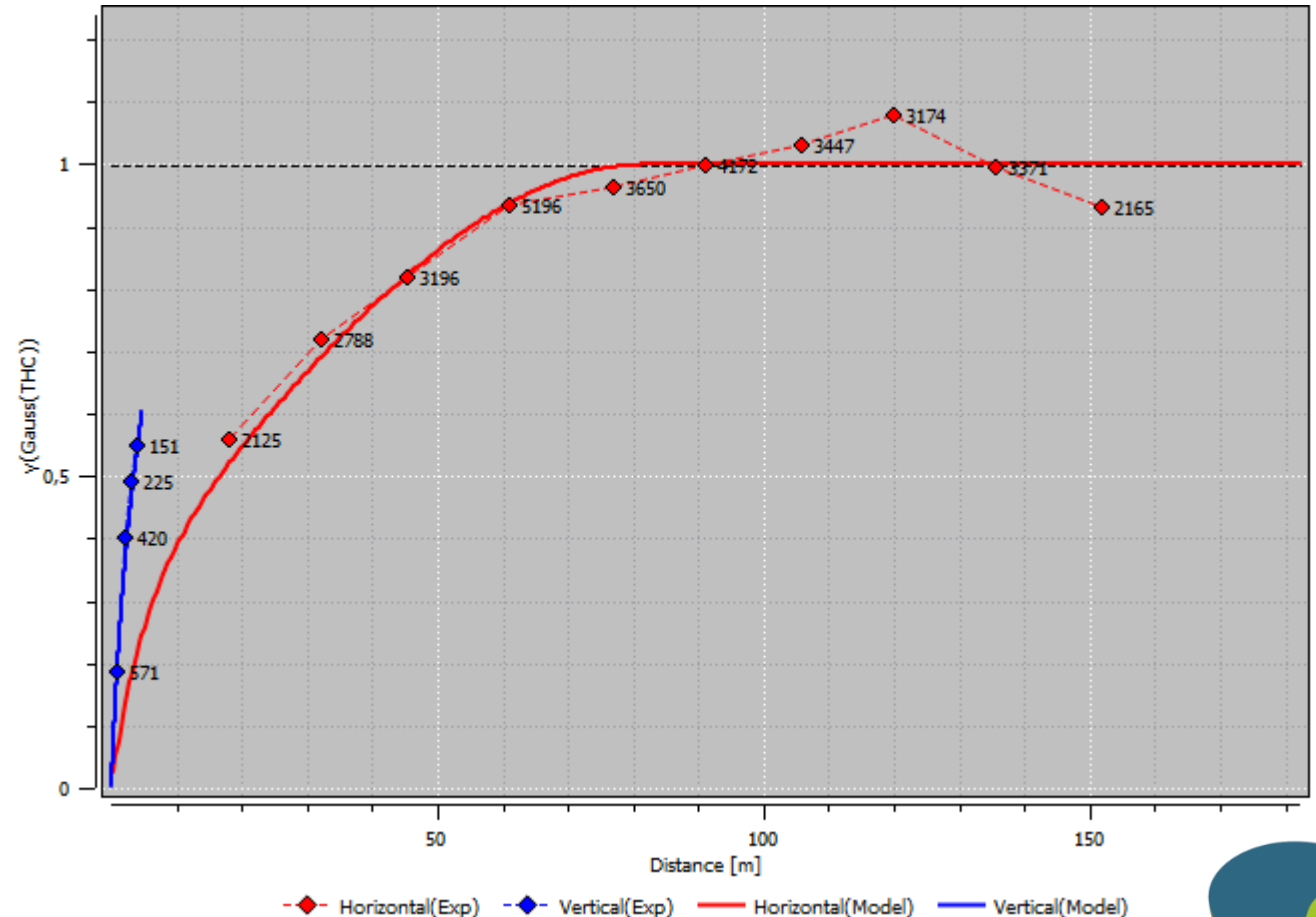
Prior knowledge (incidents...)

Multivariate with in situ mapping



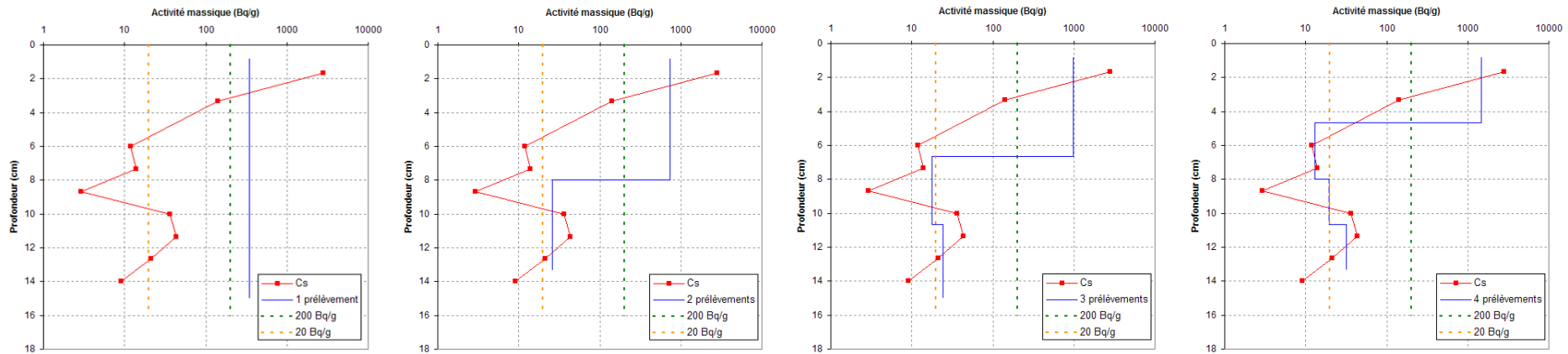
3D variogram and anisotropy

Generally, the vertical variability (along-borehole) is higher than the horizontal one (between-boreholes)



Additional comments

Migration profiles and vertical resolution

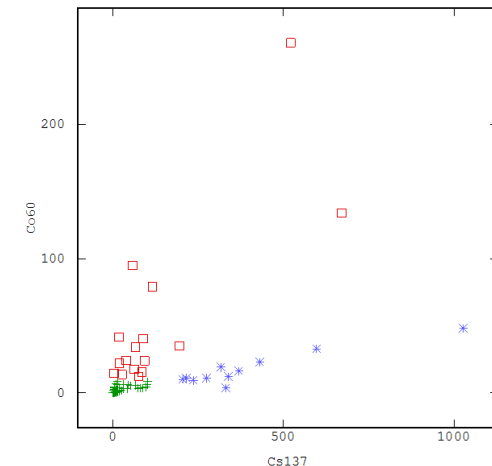


In lab analyses

- Dose | Gamma | Beta | Alpha

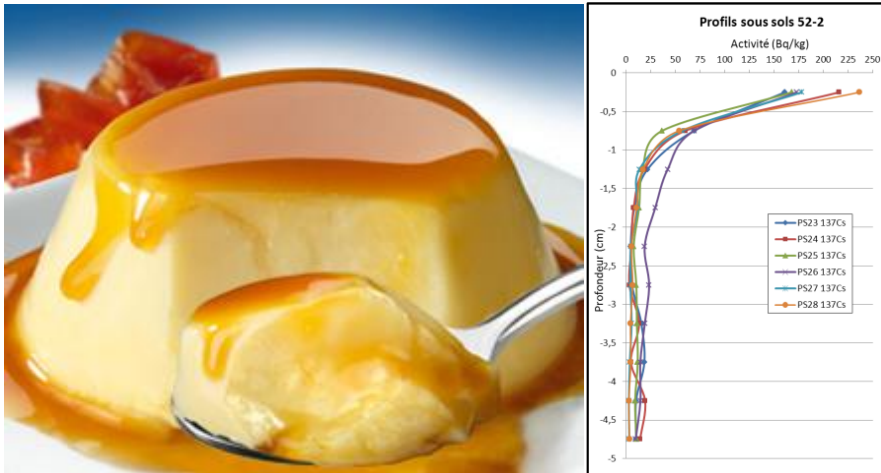


Ratio variation
(both laterally and vertically)



Examples of vertical profiles

« Classical » shape



First mm (even cm)
in concrete

A few tens of cm
in soils

Several peaks



Lithology impact

Difference of concrete quality

Older contamination...

Smart use of the variogram to explore spatial data, to break down variance contributions and to model radiological contaminations



Geovariances
Where no one has gone before

- 1 What is the variogram?
- 2 Application cases
- 3 Advanced use of the variogram
- 4 Impact on sampling strategy
- 5 Conclusions

Variogram, the smart tool

- To explore spatial data
 - Data errors, coordinates uncertainty
 - Quantification of the spatial correlation
- To break down variance contributions
 - Phenomenon, metrology, sampling issues
 - As regards spatial correlation
- To model radiological contaminations
 - 2D and 3D maps, along with estimation uncertainty
 - Taking spatial correlation into account

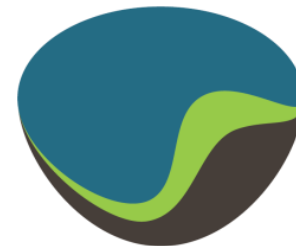


THANK YOU

Smart use of the variogram
to explore spatial data,
to break down variance contributions
and to model radiological contaminations

Yvon Desnoyers
Geostatistician
Nuclear sector

desnoyers@geovariances.com
www.geovariances.com



Geovariances
Where no one has gone before