

Deterministic and stochastic modelling for protection zone delineation

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- ***Definitions / General scope***
- ***Deterministic approach***
 - *pure advection*
 - *advection-macrodispersion*
 - *advection-dispersion + multi tracer tests*
 - *challenges*
- ***Stochastic approach***
 - *Background*
 - *Synthetic case*
 - *Stochastic generation of K-fields combined to inverse modelling*
 - *Additional conditioning by geoelectrical resistivity data*
 - *Discussion of results*
- ***Conclusions / Perspectives***

Definitions

Well capture zone : ... the set of points on the groundwater surface from which a tracer particle will reach the well

→ total catchment of the well

Time-related capture zones: ... isochrones (contour lines of equal travel time to the well)

→ only parts of the total catchment

In many regions: **protection zones** at the surface corresponding to particular isochrones in saturated zone

→ a time-related protection

Protection zones ↔ time-related capture zones

➔ *numerical computational methods are used to obtain a delimitation...*

➔ *What are the conceptual choices ?
What are the needed data ?
What are the uncertainties linked to the obtained delineation ?
Deterministic or Stochastic approach ?*

➔ *... in heterogeneous geological formations, all direct and indirect data, respectively hard and soft data, must be used in an optimal way !*

Deterministic approach

Advection + Macrodispersion

- different values of hydraulic conductivity (and possibly also for effective porosity)***
- a macrodispersion term representing statistically the general contaminant behaviour around the advective mean position***
- smaller scale heterogeneities are not introduced in detail***
- 'scale effect' is observed and difficulties to assess upscaled values***

Advection + Dispersion

➔ *heterogeneous conditions for both the groundwater flow model and the transport model*

Example: a methodology proposed to water suppliers in Walloon Region of Belgium (Dassargues, 1994)

- *geology, geomorphology, basic hydrology*
- *geophysical prospecting*
- *piezometers and observation wells*
- *pumping tests in each borehole*
- *multi-tracer tests in pumping conditions*
- *first analytical interpretations*
- *building of a flow-transport model considering heterogeneity in the layers*
- *calibration for flow (on measured piezometric levels)*
- *calibration for transport (on the measured breakthrough curves)*
- *simulations and computations of the travel times for different injection points (including the dispersion)*

... extrapolation of parameters

for each geological unit:

- lithology
- fissuration/
fracturation degree

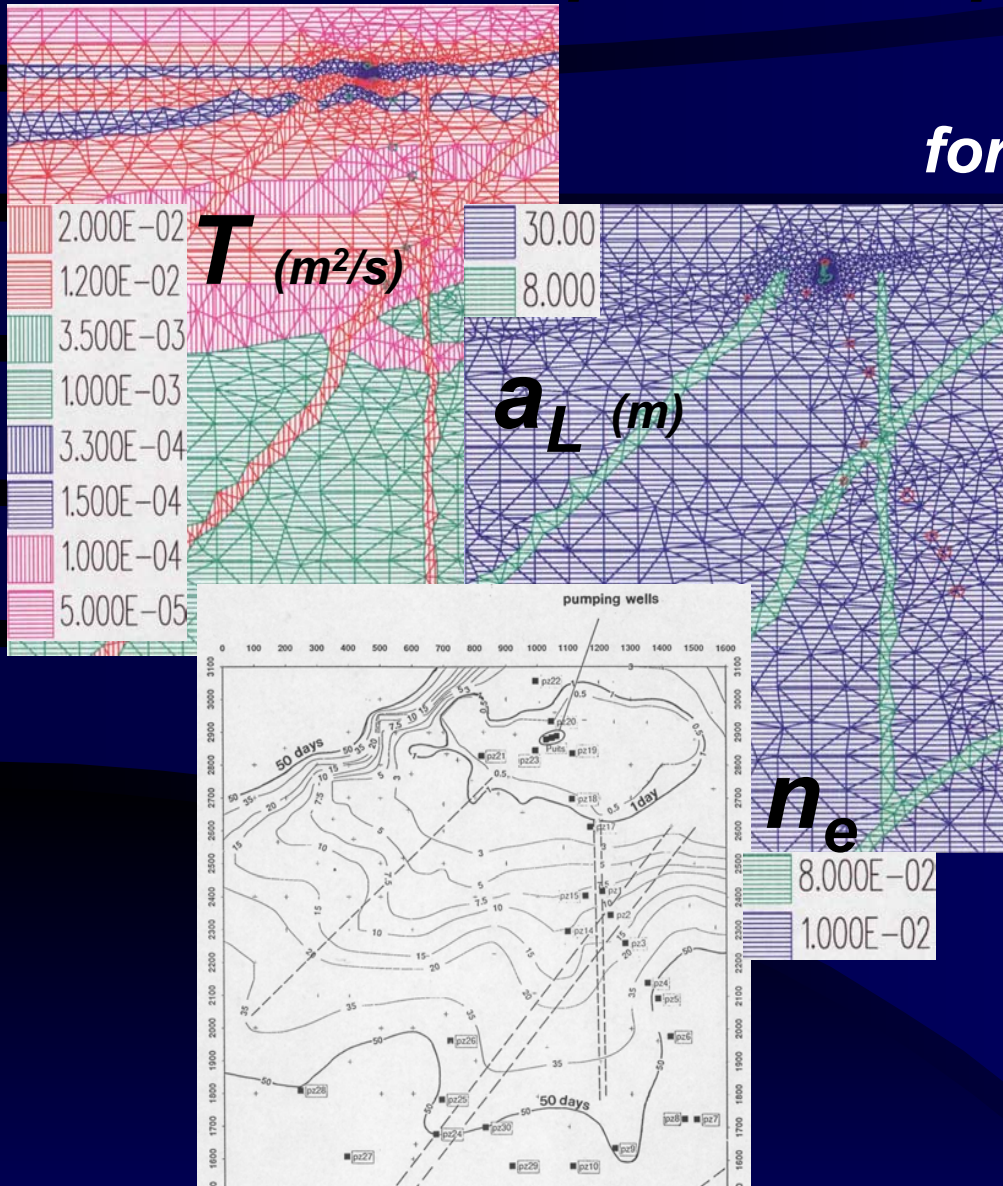


values of - K

- n_e
- a_L, a_T
- ...



purely deterministic
extrapolation based
on 'hard data' and
'soft data'



... challenges



- **tracers behaviours in different geological media (adsorption, ...)**
- **injection control and measurement of the real input function in the aquifer**
- **2D and 3D aspects of the tracer tests and modelling**
- **boundary conditions for groundwater flow and transport models**
- ➔ • **multiple possible calibrations using 'trial and error' calibration or automatic calibration**
- ➔ • **how the role of the geology (soft data) can be combined with a calibration objective function ?**
- ➔ • **upscaling and spatial extrapolation of the parameters taking the spatial variability and specific heterogeneities into account**
- ➔ • **uncertainty of the results...**
 - **immobile water effect, ...**
 - **legal aspects concerning - the 'first arrival' of tracer (contaminant) - the non-saturated zone**

Stochastic approach

General aims

- to obtain a quantification of the uncertainty of results;
- optimising the use of the available data;
- how to combine inverse modelling procedure with integration of soft-data;

Many different approaches !

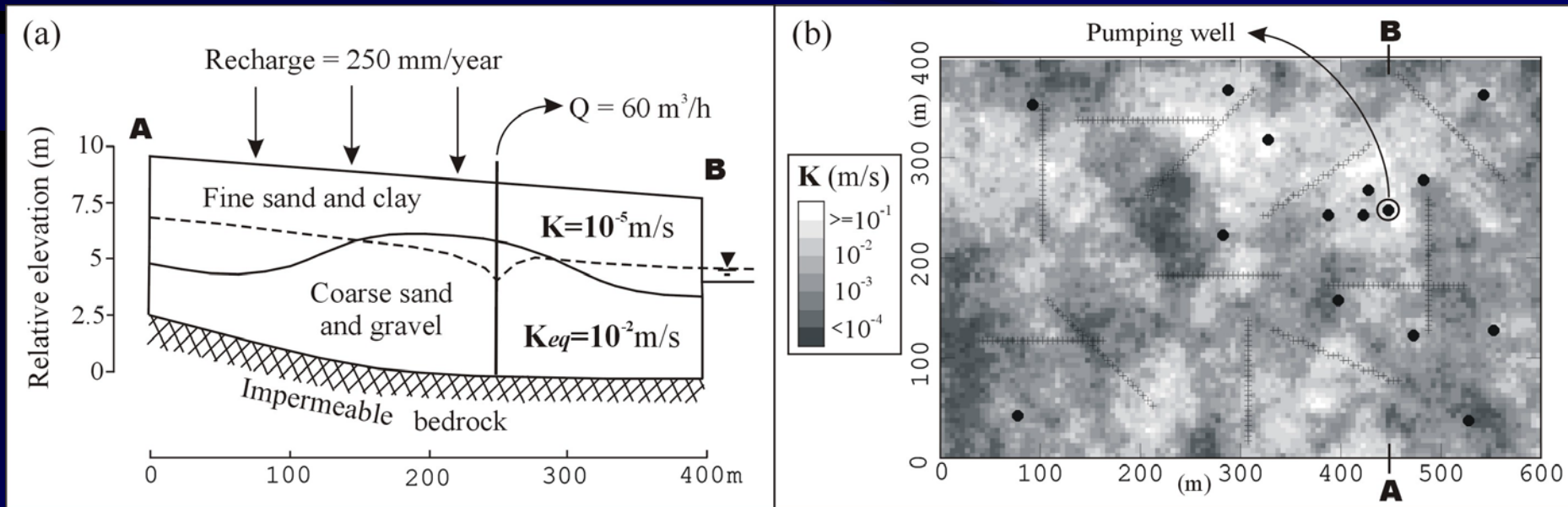
Our example of methodology:

➔ a stochastic approach integrating different sorts of data:

- K -data (hard data) by **conditional** stochastic simulations;
- h data (soft data) by **inverse modelling**;
- ρ data (soft data) by conditional stochastic **co-simulations**

Synthetic example

- hypothetical groundwater flow domain with hydrogeological conditions similar to actual alluvial sites
- domain large enough to avoid boundary effects
- two layers:
 - fine sand and clay layer ($K = 10^{-5}$ m/s)
 - lower coarse sand and gravel layer for which a "true" hydraulic conductivity field representing the "reality" is created using a non-conditional simulation

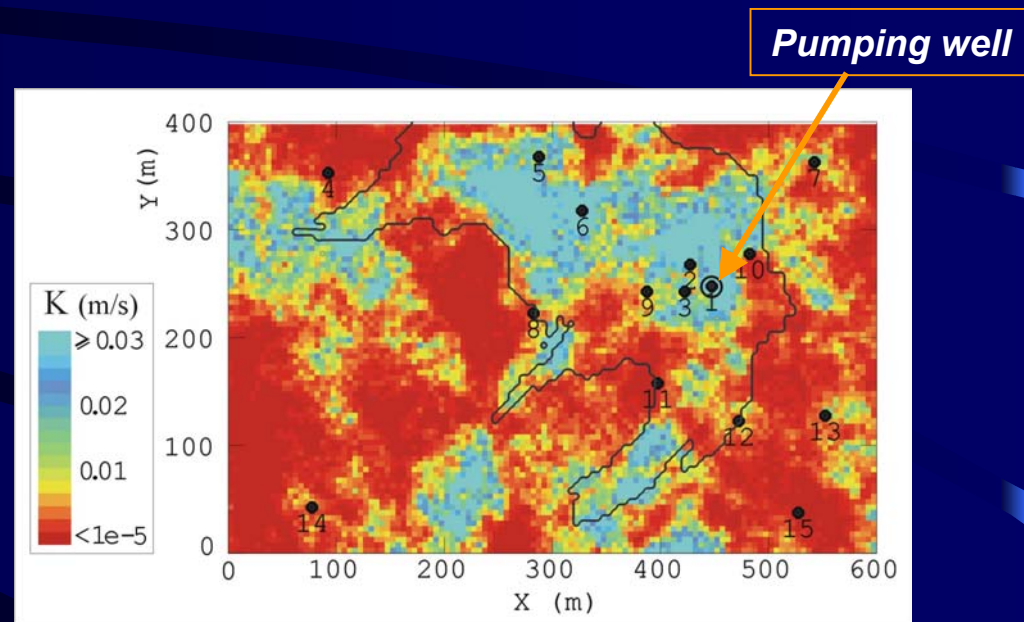


→ using the **Turning Band algorithm** (Mantoglou & Wilson, 1982) with an isotropic, exponential correlation structure of $\log K$ (alluvial sediments of the Meuse River valley downwards to Liège, Belgium)

→ from a grid of 9600 cells: selection 15 K values providing the hard data set

- pumping well ($60 \text{ m}^3/\text{h}$)
- flow simulation providing the synthetic "measured" heads at the 15 virtual piezometers (first set of soft data)
- considering advective transport time to the well,

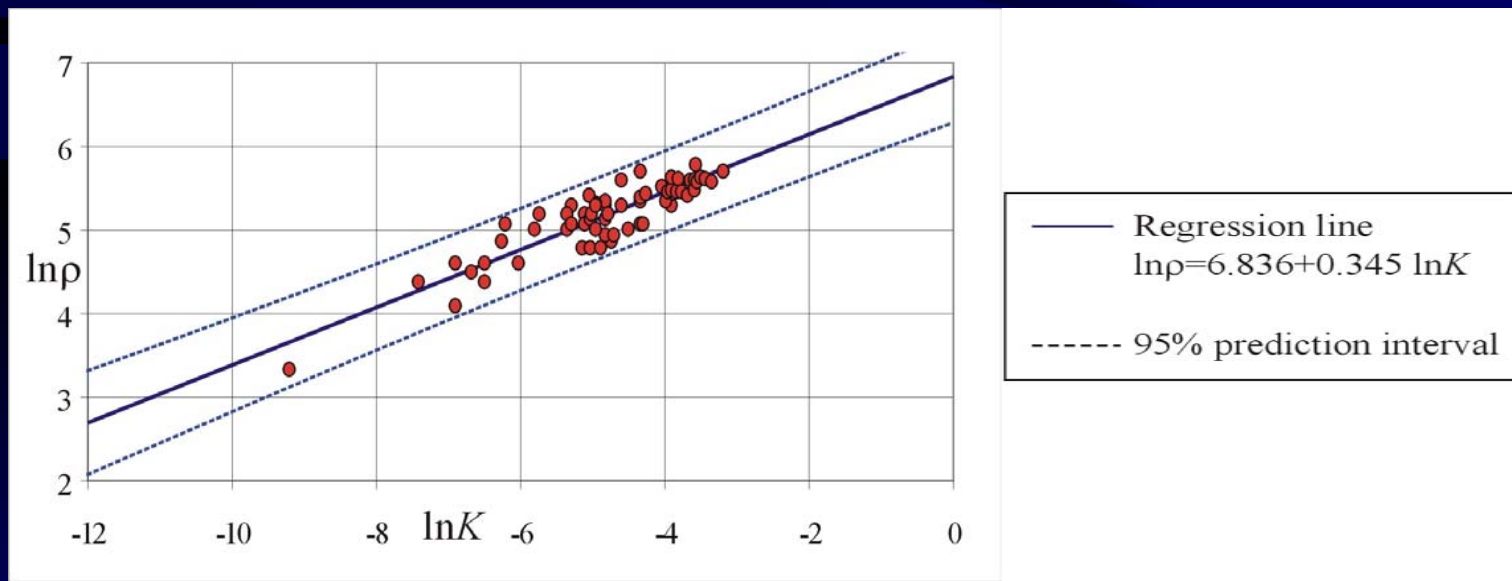
→ the "true" 20-day isochrone (associated with the concerned pumping rate and $n_e = 0.05$) resulting from the "true" hydraulic conductivity field → reference isochrone



- a resistivity data set (second set of soft data) is created based on the observed correlation ($r = 0.9$) existing between electrical resistivity (ρ) and hydraulic conductivity (K) in the alluvial sediments of the Meuse River

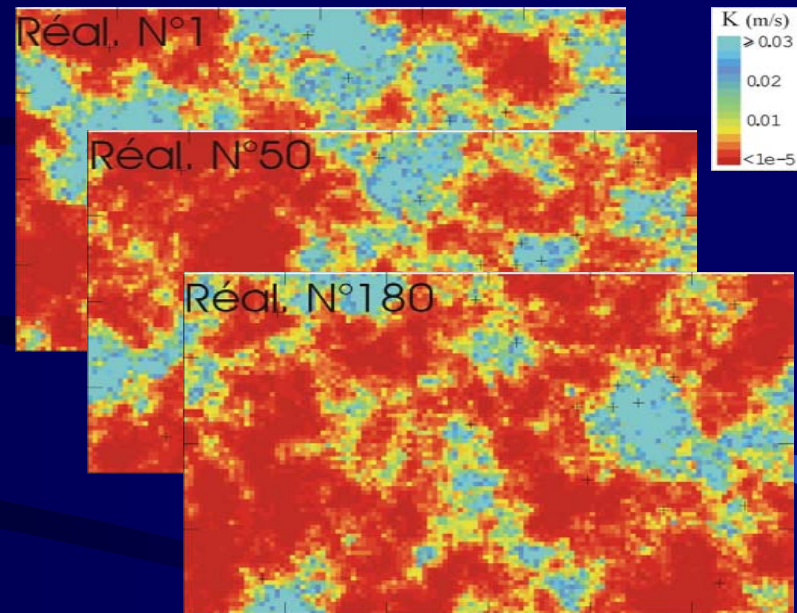
- considering $N(0,1)$ as a random draw within a standard normal distribution and $\hat{\sigma}$ the standard deviation of the regression residual, \Rightarrow 300 resistivity values, distributed on 12 tomographic profiles

$$\ln \rho = 6.836 + 0.345 \ln K + \hat{\sigma} N(0,1)$$



Stochastic conditional simulations

- *four hundred stochastic simulations of equally likely hydraulic conductivity fields*
- *subsequently conditioned on hydraulic conductivity measurements by a kriging technique*



Conditional simulation

=

**Non-cond. simulation
= realisation**

–

Kriging of each realisation

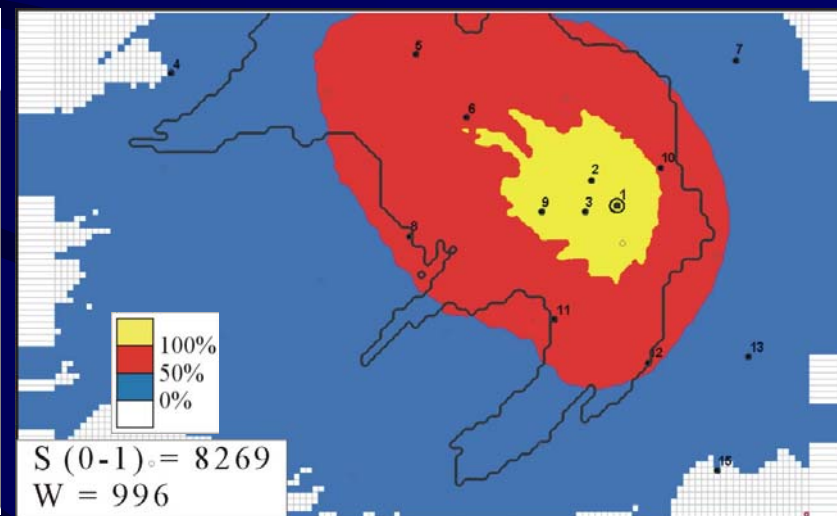
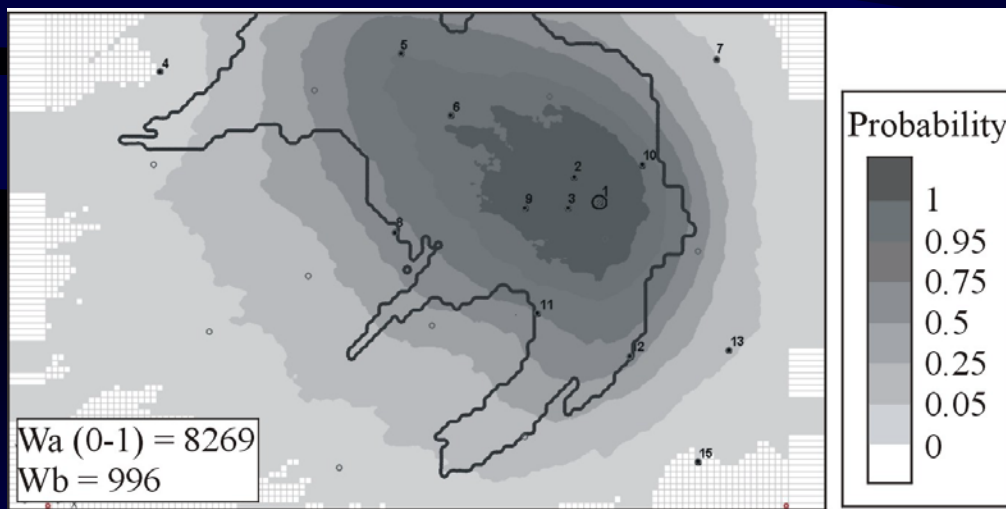
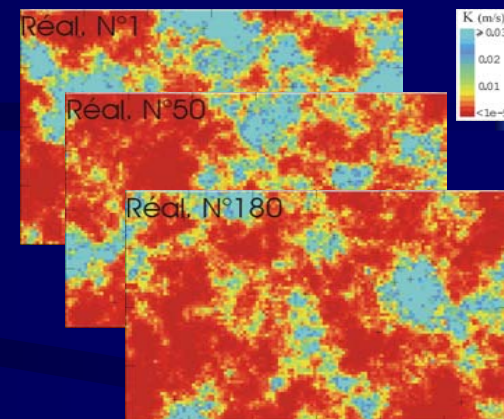
+

Kriging

noise except in the measurement points

**Exact estimator
+ smoothing**

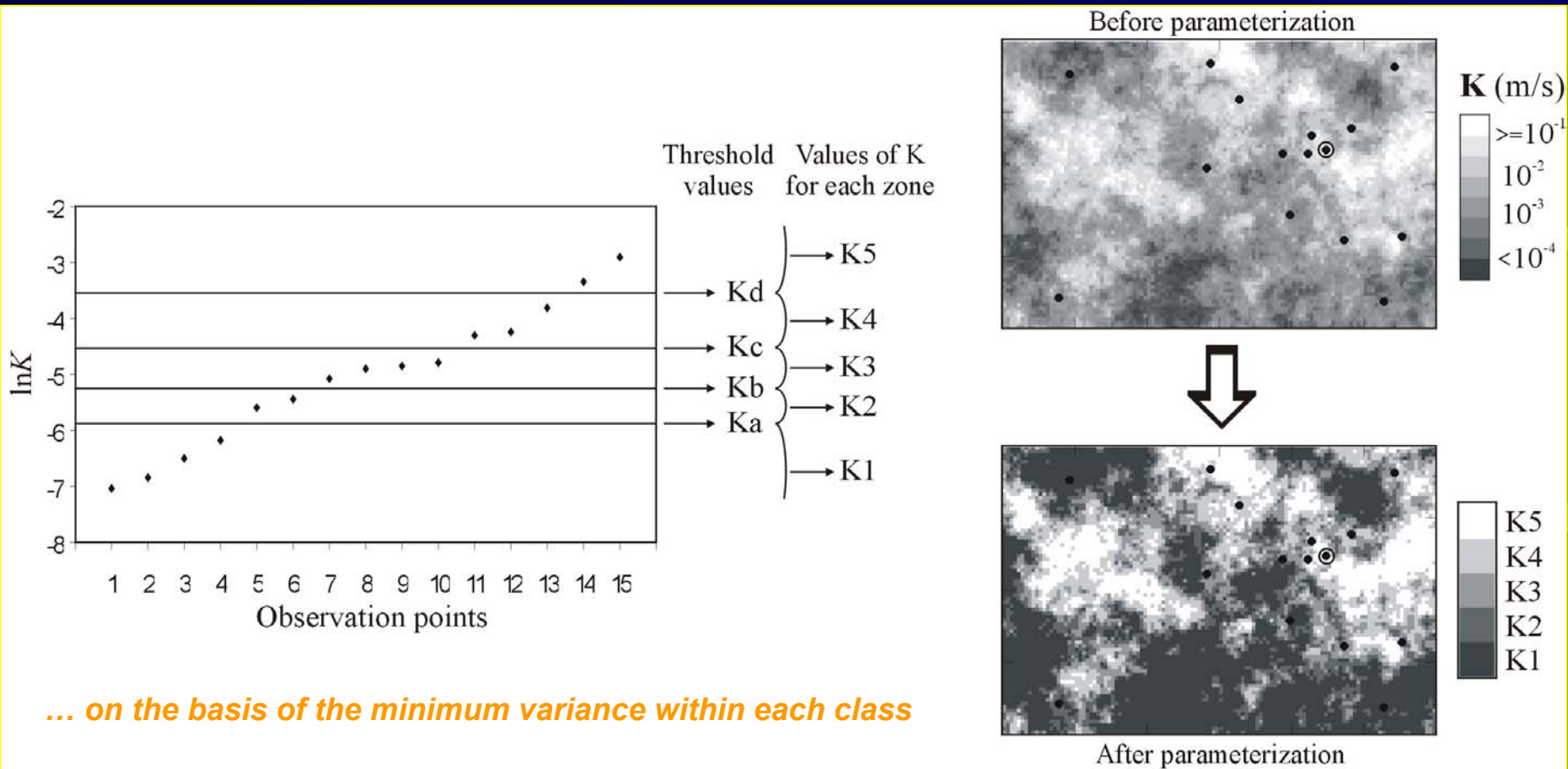
- groundwater flow and a particle tracking process were computed for each realization
- ensemble of obtained capture zones
- treated statistically to infer the capture zone probability distribution (CaPD)
- CaPD gives the spatial distribution of the probability that a conservative tracer particle released at a particular location is captured by the well within a specified time span (van Leeuwen, 2000)



Wa : extent of the uncertainty zone for which the probability P of capture is $0 < P < 1$
Wb : difference between reference isochrone and isoline 50% (probability of capture)

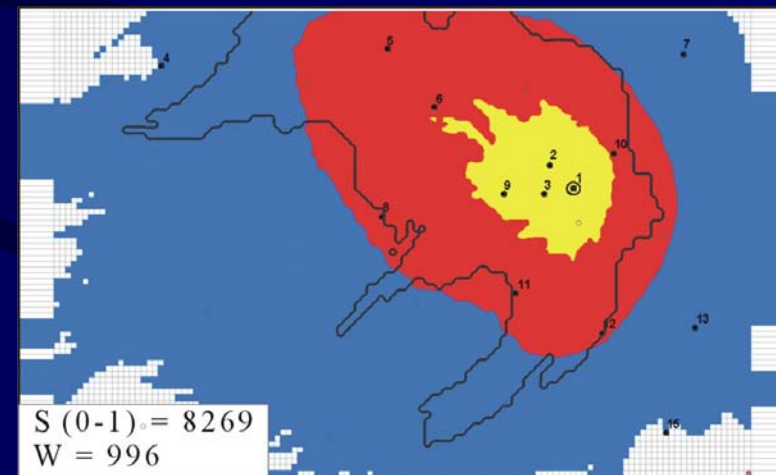
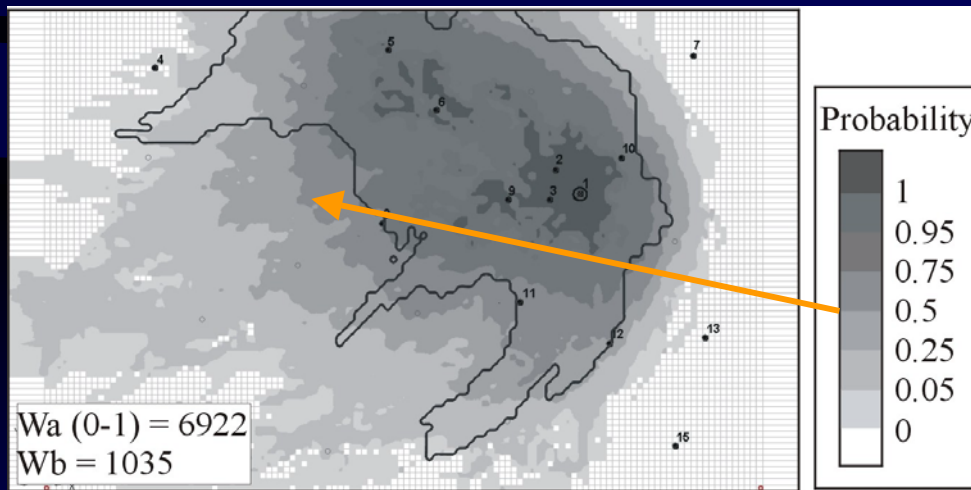
Additional conditioning by head measurements (first set of soft data)

- parameterization of each K-field
- thresholds values for dividing in five zones of uniform value ($K_i, i = 1, \dots, 5$)



for each parameterized K-field:

- a groundwater flow calibration on head measurements using PEST
(Doherty, 1994)
- rejecting obtained (calibrated) realizations that did not respect $(K_i < K_{i+1})$, considering them as geologically erroneous
- for each remaining realization, computation of the 20-day capture zone \rightarrow CaPD for the ensemble of possible capture zones

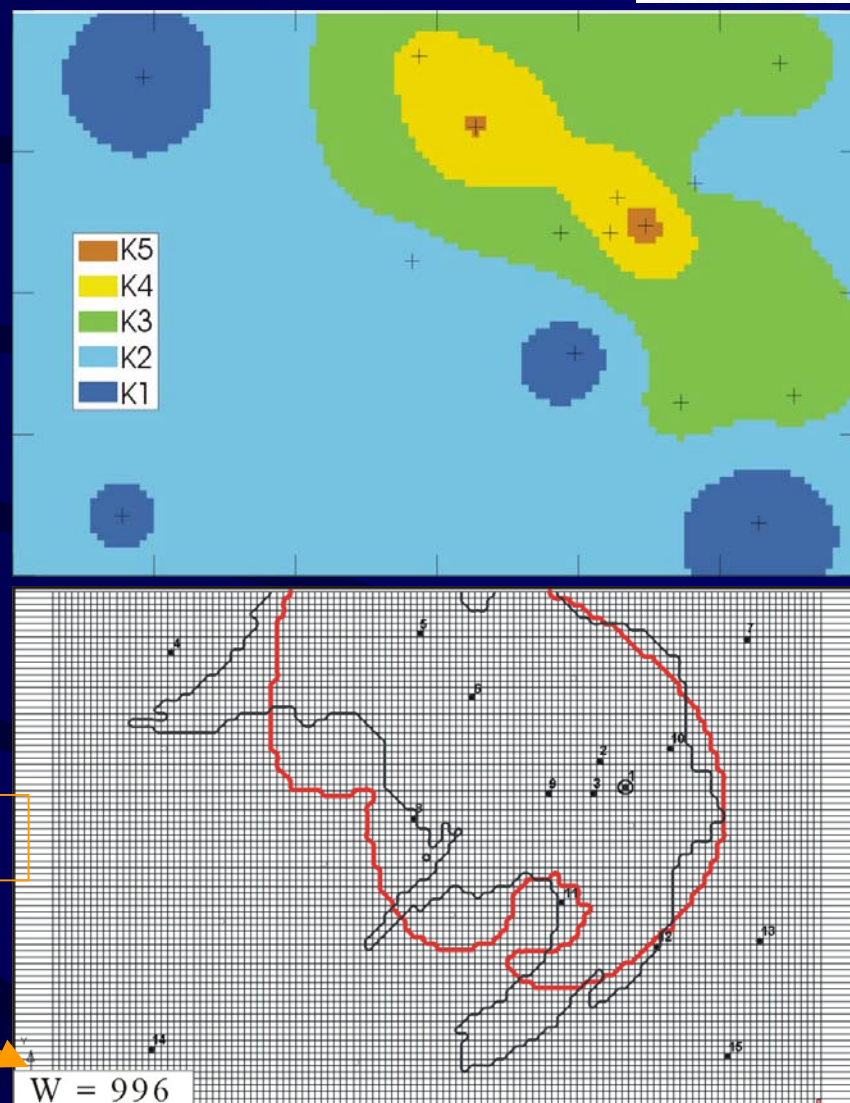


... reducing W_a (uncertainty zone for which the probability P of capture is $0 < P < 1$)
 ... not changing W_b (difference between reference isochrone and isoline 50%)

... if we were deterministic

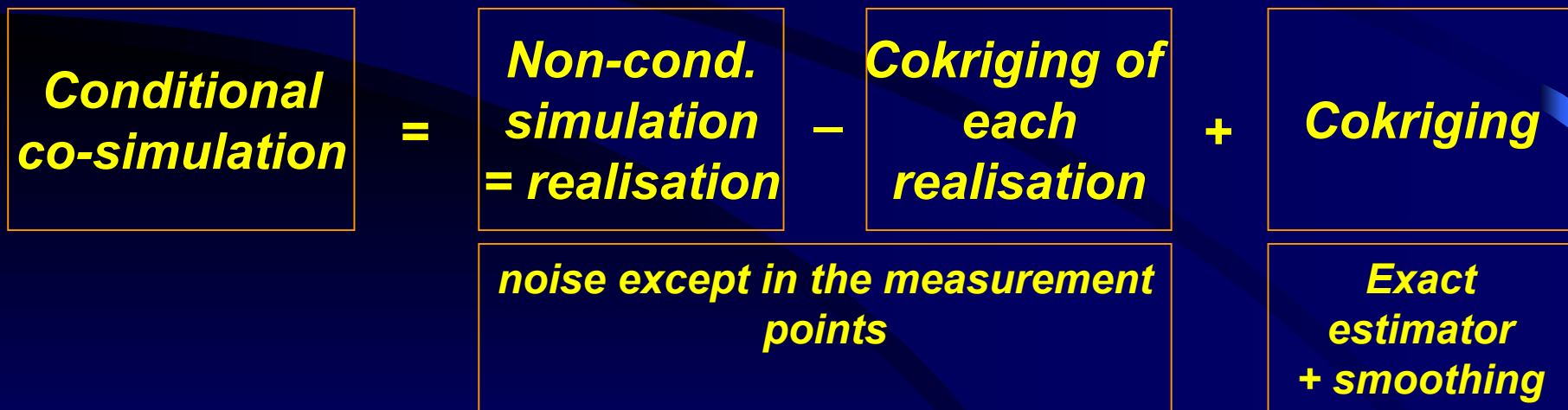
- collecting the 15 measured K values and piezometric heads
- definitions of zones (on the basis of kriging and the threshold method)
- calibration of the groundwater flow model on the 15 measured h
- advective transport simulation
- 20 days isochrone line

difference (in 5m x 5m cells)



Additional conditioning by geoelectrical resistivity data (second set of soft data)

- *integration of the geophysical data set by conditioning each stochastic simulation on both hydraulic conductivity measurements and resistivity values*
- *cokriging technique, providing stochastic conditional "co-simulations"*
- *four hundred K-fields conditioned on K values (hard data) and on ρ values (soft data)*

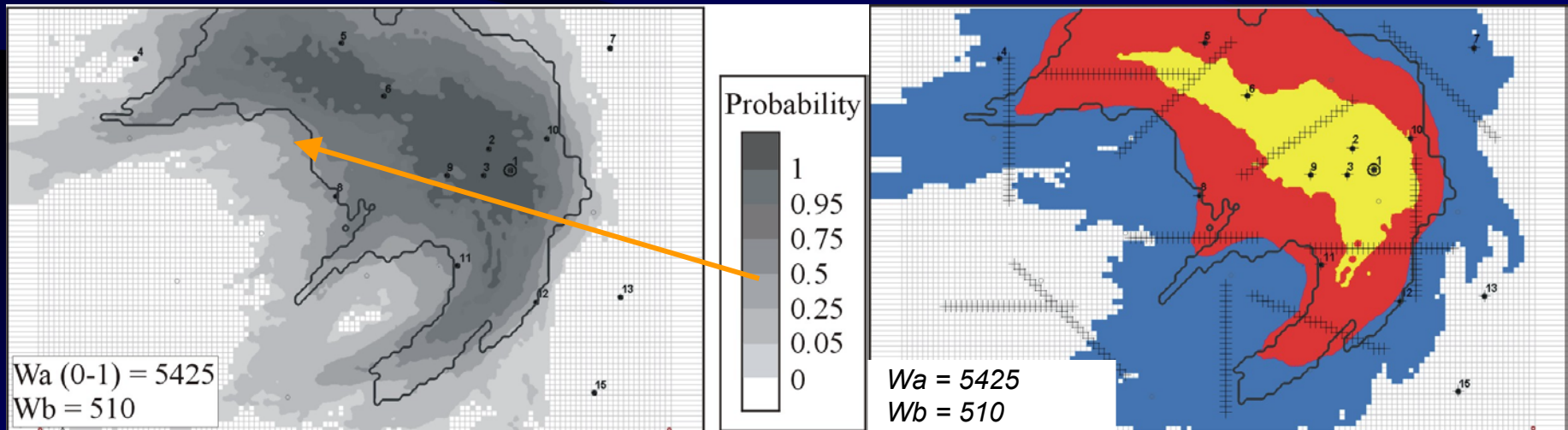


... as previously:

- parameterization of each K-field
- thresholds values for dividing in five zones of uniform value (K_i , $i = 1, \dots, 5$)

... as previously, for each obtained K-field:

- inverse modelling using PEST (conditionning on h)
- rejecting realizations that did not respect ($K_i < K_{i+1}$)
- for each remaining realization, computation of the 20-day capture zone \rightarrow CaPD for the ensemble of possible capture zones



Work being done ...



Sensitivity analysis

- ***number of hard data (K)***
- ***number of first soft data (measured h)***
- ***number of second soft data (ρ)***
- ***threshold method used for parameterization and values for dividing in zones***
- ***relaxation of the rejection criterion***

Application to a practical study case

- ***what are the practical difficulties ?***
 - CPU time ? local minima in inverse modelling ?***
- ***number of K values and measured h : very limited***

Conclusions / Perspectives



- *stochastic approaches bring improvements*
- *it does not spare us the acquisition of measured data*
- *selection of 'best' locations for geophysical measurements*
- *all other issues concerning modelling and tracer tests interpretation remain*
- *further conditioning on tracer travel times, and on other soft data*
- *do we include dispersion ? to which extent ?*
- *are these methodologies applicable at another scale ?*
- *are these methodologies applicable in fissured media ?*