



Geo-Electrical data fusion by stochastic co-conditioning simulations for delineating groundwater protection zones

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- ... from a practical point of view : one of the classical questions of hydrogeology :

delineation of protections zones around pumping wells
in aquifers ... integrating (as far as possible) all
available data

- "... the future of « dealing with heterogeneity » in hydrogeology depends largely on a conscious decision to better characterize, describe and model the geology of the sites ..."

de Marsily et al., 2005

- "... the central question of whether the stochastic method, which treats aquifer heterogeneity as a random field, is applicable to real aquifers under field conditions, has not been definitively answered"

M.P. Anderson, 1995

A. Dassargues 2006



Outline

- Capture zones delineation in aquifers
 - Co-conditional stochastic method for delineation of time-related capture zone combined with an inverse modelling procedure
 - Application to a virtual (synthetic) study case
 - Application to a real case
- Conclusions/perspectives from a practical point of view

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Groundwater protection zones

- Time-related capture zone (protection zone)

➡ delineation based on the concept of travel time



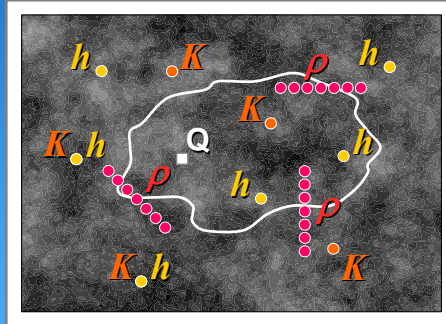
= area around the well from which water is captured within a certain time t

- ➡ based on the reliability of the hydrogeological models
- ➡ depends strongly on our ability to describe the aquifer system properties (K, n_e, \dots)
- ➡ based on our knowledge of the geology (limited by the existing field data)

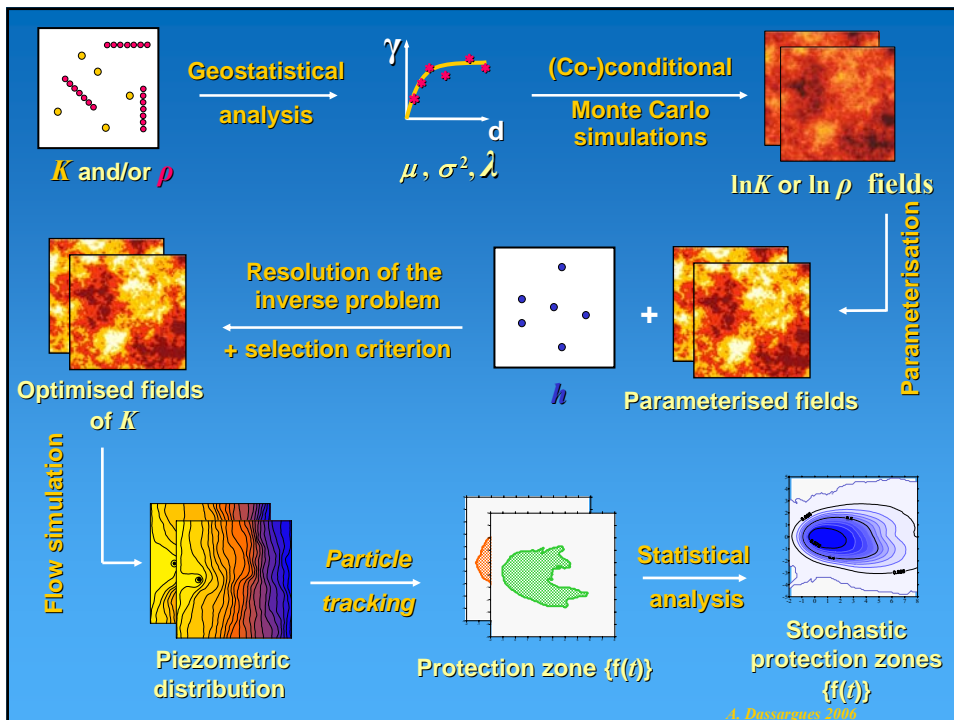
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Objectives

Propose a stochastic method to delineate protection zones, applicable to real study case with fusion of data from geophysical prospecting

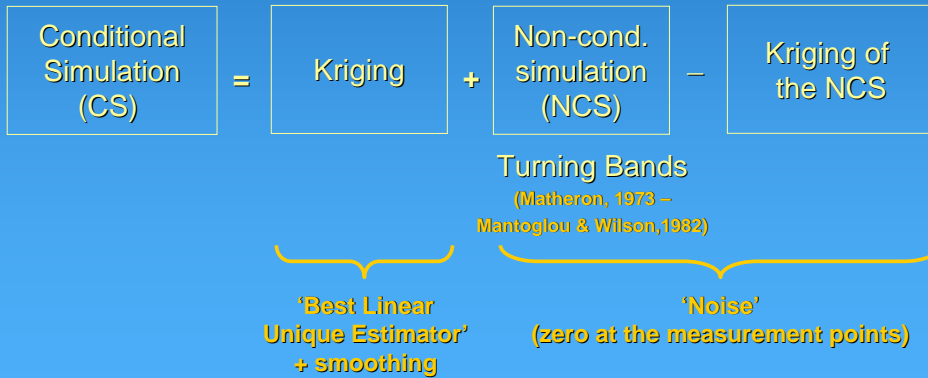


- ✓ quantification of the well capture zone uncertainty
- ✓ reduction of this uncertainty by fusion of direct and indirect measures of K



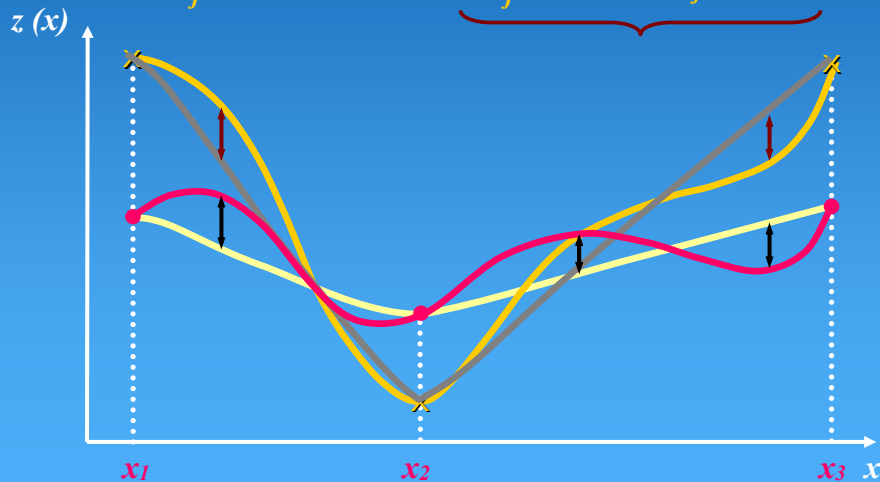
Conditional simulations

$$z_j^{CS}(\underline{x}) = z^*(\underline{x}) + z_j^{NCS}(\underline{x}) - z_j^{NCS*}(\underline{x})$$



Conditional simulations

$$z_j^{CS}(\underline{x}) = z^*(\underline{x}) + z_j^{NCS}(\underline{x}) - z_j^{NCS*}(\underline{x})$$



Co-conditional simulations

$$z_j^{SC}(\underline{x}) = z^*(\underline{x}) + z_j^{SNC}(\underline{x}) - z_j^{SNC*}(\underline{x})$$

Co-conditional
simulation


= Cokriging

+

Non cond.
simulation
NCS




-

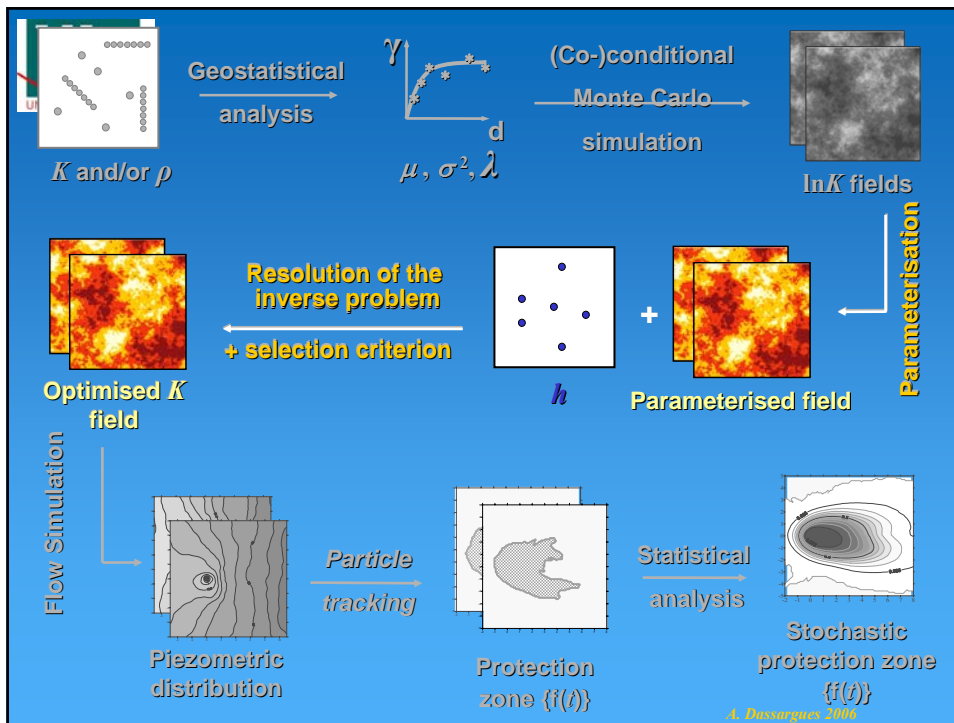
Estimation
by cokriging
of the NCS

... by this cokriging (fusion of secondary data),
 a better characterisation of the spatial variability is obtained

(Co-)conditional simulations

Data Availability

- K  conditional simulations of $\ln K$
- K and ρ  co-conditional simulations of $\ln K$
- ρ  conditional simulations of $\ln \rho$



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Conditioning on measured h through an inverse modelling procedure

→ ... requires parameterisation (zonation in 'facies')
 → How to perform that without losing the geostatistical properties of the K fields ?

Actual variation range for $\ln K$

$-\infty \longleftarrow \longrightarrow +\infty$

↓ Zonation

$-\infty \quad S1 \quad S2 \quad S3 \quad S4 \quad +\infty$
 $\underbrace{\hspace{1cm}} \quad \underbrace{\hspace{1cm}} \quad \underbrace{\hspace{1cm}} \quad \underbrace{\hspace{1cm}} \quad \underbrace{\hspace{1cm}}$
 $C1 \quad C2 \quad C3 \quad C4 \quad C5$
 avec
 $K_{C1} < K_{C2} < K_{C3} < K_{C4} < K_{C5}$

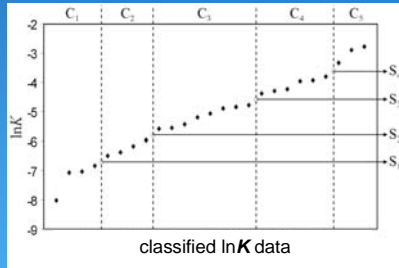
→ thresholds values for dividing in 'facies' of uniform value ($K_i, i = 1, \dots, 5$)

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Which threshold values (S_i) ?

- Principle : minimum variance within each 'facies'

→ minimising function f



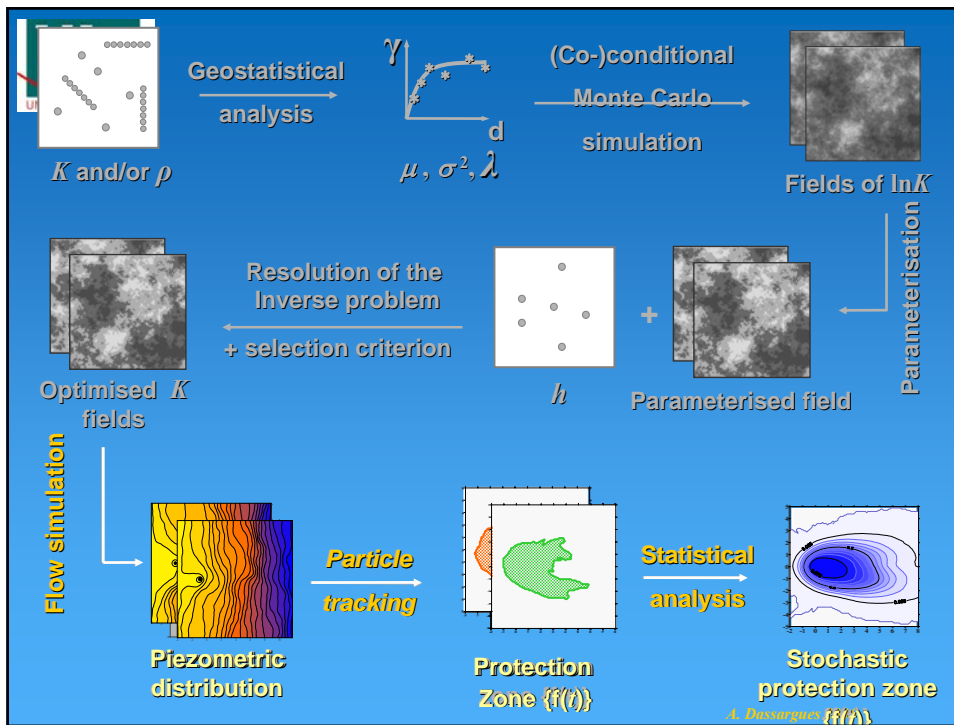
$$f = \sum_{i=1}^{N_c} \sum_{j=1}^{N_{di}} (\ln K_{ij} - \overline{\ln K}_i)^2$$

$$\overline{\ln K}_i = \frac{1}{N_{di}} \sum_{j=1}^{N_{di}} \ln K_{ij} \quad , i = 1, N_c$$

- then the value of K_{Cj} = average value / facies
- serve as initial value in (not adjoining) zones of the facies
- optimised by the inverse procedure

After the inverse procedure: selection criterion (order criterion)

- after optimisation of the K values in each 'facies', we add the selection criterion that the initial respective order must be respected : $K_{C_i} < K_{C_{(i+1)}}$
or relaxing:
→ only one permutation
- eliminating realisations not respecting this criterion as « unrealistic fields for a geological point of view »



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Statistical analysis

Capture zone probability distribution

$$[CAP(\underline{x}, t)] = \frac{1}{n} \sum_{k=1}^n (I(\underline{x}, t))_k$$

= probability that a tracer particle released at a particular location is captured by the well within a specified time

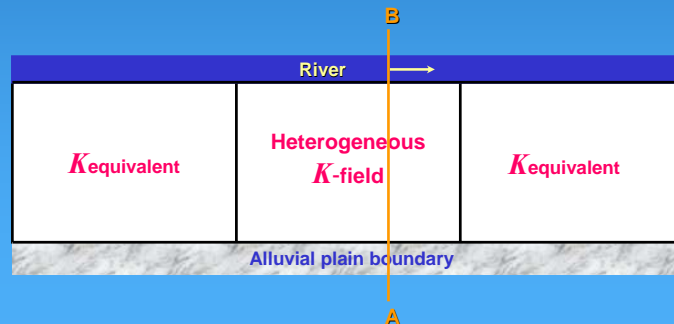
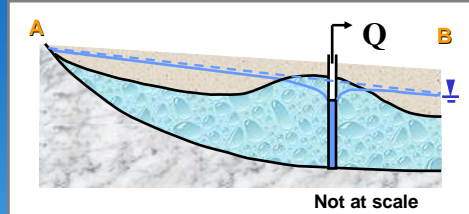
The isolines $\Gamma(t)$ connect the locations in the CAP with the same probability $p(CAP(x, t)) = i$

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Synthetic study case

Groundwater model
representing
« a reference situation »

very similar to alluvial aquifer
conditions

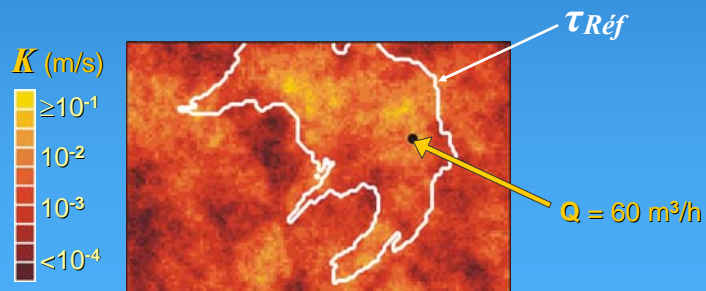


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Construction of a reference capture zone

One non-conditional K -field generation: reference medium

- ➔ Reference protection zone $f(t)$
(by direct groundwater flow simulation and particle tracking)
- ➔ Create sets of 'measured data' K , h and ρ

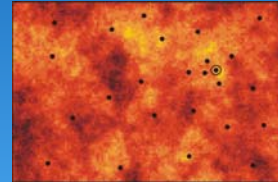


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Creating realistic 'measured data sets'

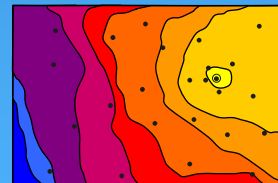
- Creating a (K) hydraulic conductivity data set

→ in few 'measurement points': value from the K field



- Creating a (h) piezometric heads data set

→ in few 'measurement points': value from the reference flow field



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Creating realistic 'measured data sets' (following)

- Creating a (ρ) geoelectrical resistivity data set

Data measured and collected in many sites of the alluvial sediments of the River Meuse

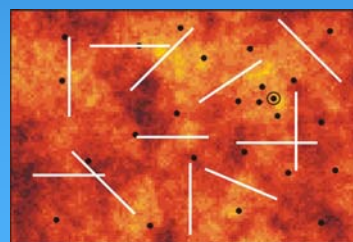
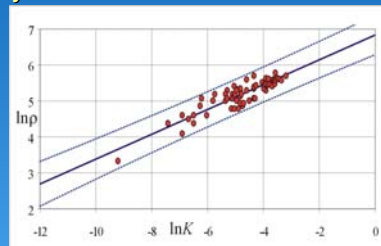
→ correlation between $\ln K$ and $\ln \rho$

→ artificial generation of a ρ data set from the reference K field

considering $N(0,1)$ as a random draw within a standard normal distribution and σ the standard deviation of the regression residual,

$$\ln \rho_i = b_0 + b_1 \ln K_i + \sigma \cdot N_i(0,1)$$

300 resistivity values, distributed on 12 tomographic profiles



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Application to the synthetic study case

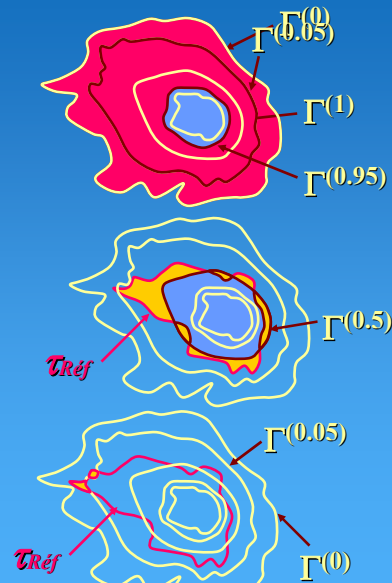
	0 K	15 K	25 K
0 ρ	[CAP(x,t)]	[CAP(x,t) K_{15}, h_{15}] [CAP(x,t) K_{15}, h_{25}]	[CAP(x,t) K_{25}, h_{15}] [CAP(x,t) K_{25}, h_{25}]
150 ρ	[CAP(x,t) h_{15}, ρ_{150}] [CAP(x,t) h_{25}, ρ_{150}]	[CAP(x,t) $K_{15}, h_{15}, \rho_{150}$]	[CAP(x,t) $K_{25}, h_{25}, \rho_{150}$]
300 ρ	[CAP(x,t) h_{15}, ρ_{300}] [CAP(x,t) h_{25}, ρ_{300}]	[CAP(x,t) $K_{15}, h_{15}, \rho_{300}$]	[CAP(x,t) $K_{25}, h_{25}, \rho_{300}$]

$t = 1, 5, 10, 20$ days

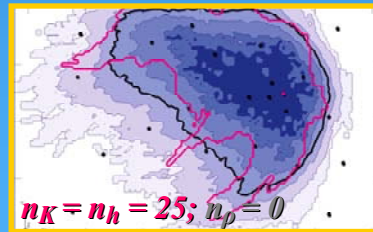
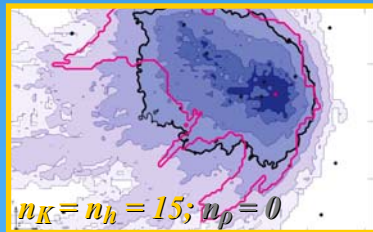
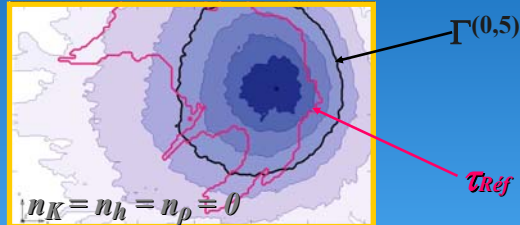
Performance quantification

(after van Leeuwen & al., 2000)

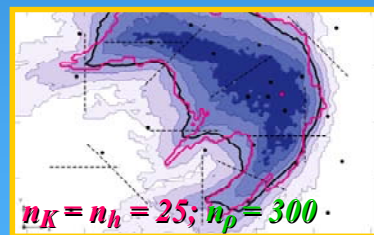
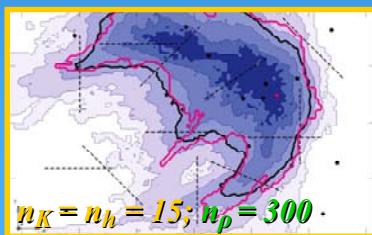
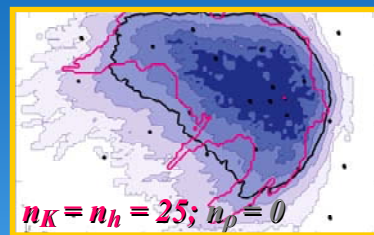
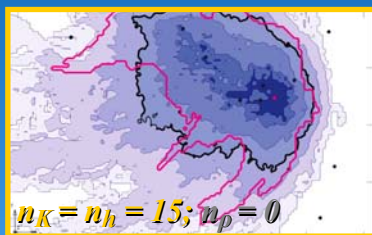
- w_{a1} and w_{a2} = a measure of the surface between $\Gamma^{(0)}$ and $\Gamma^{(1)}$; and between $\Gamma^{(0.05)}$ and $\Gamma^{(0.95)}$
- w_b = deviation of $\Gamma^{(0.5)}$ with regards to $\mathcal{T}^{Réf}$
- w_{s1} and w_{s2} = how far $\mathcal{T}^{Réf}$ is well included in the zone $\Gamma^{(0)}$ and $\Gamma^{(0,05)}$

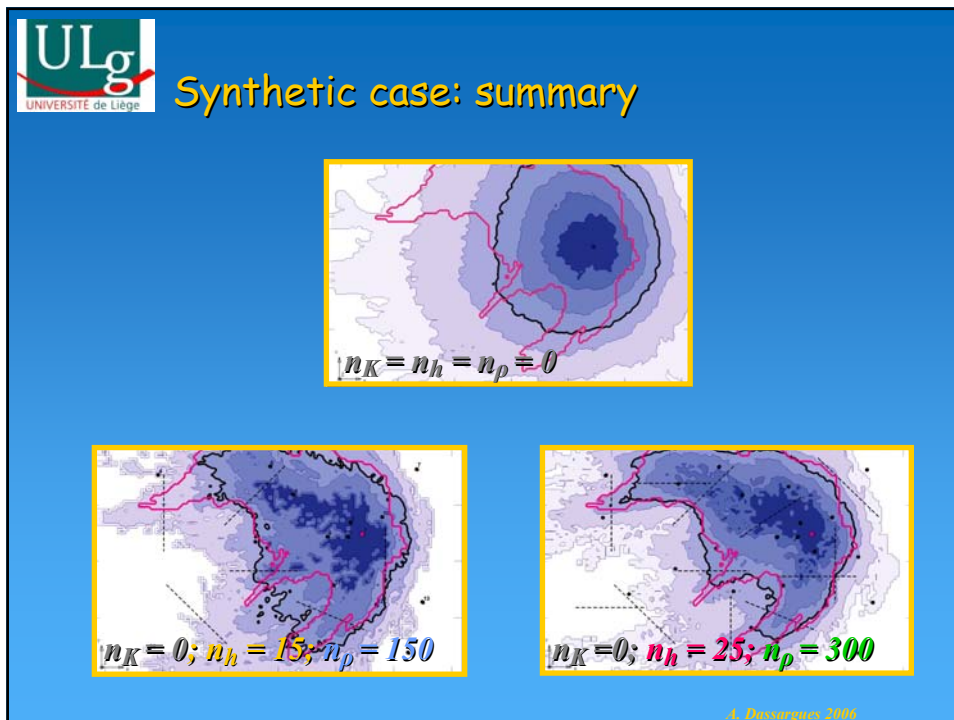
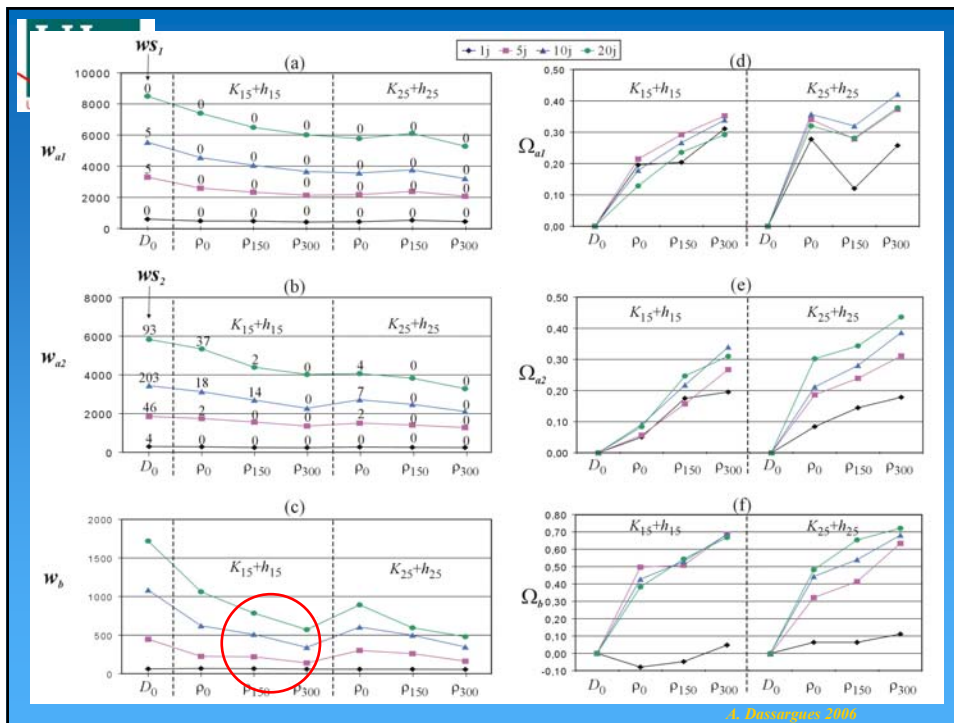


For each realisation,
computation of the 20-day
capture zone \rightarrow CaPD

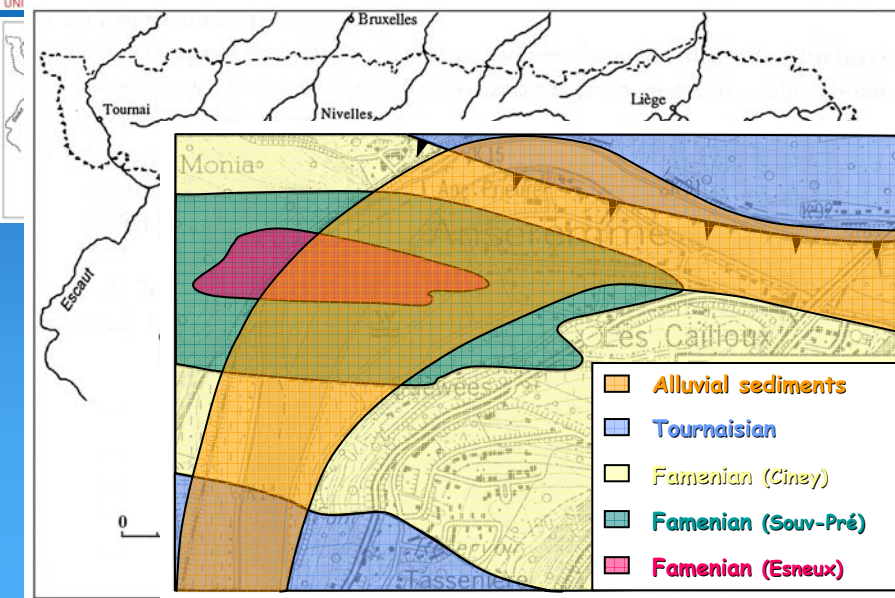


... strong effect of the co-conditioning



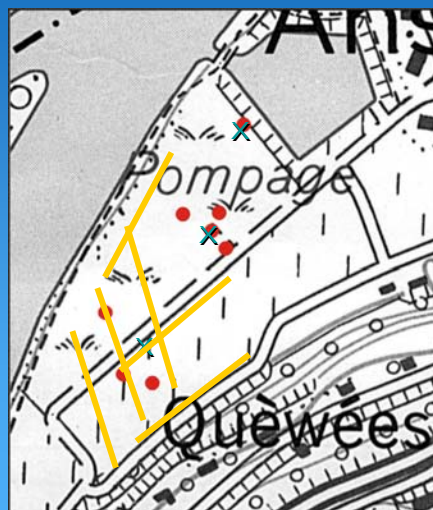


Application to a real case (Anseremme near Dinant, Belgium)



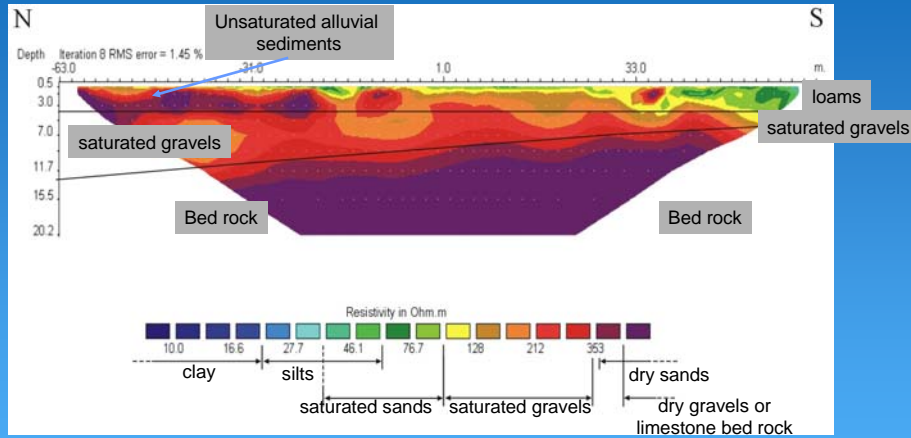
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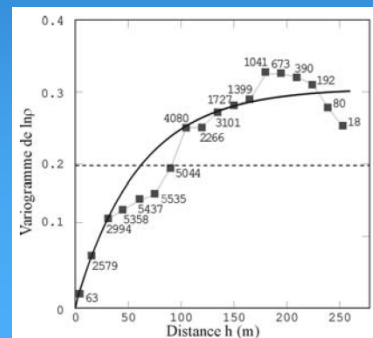
Equivalent value of ρ obtained in the gravel layer (alluvial sediment)

$$\frac{e_{tot}}{\rho} = \sum_{i=1}^n \frac{e_i}{\rho_i}$$

A. Desbordes 2004

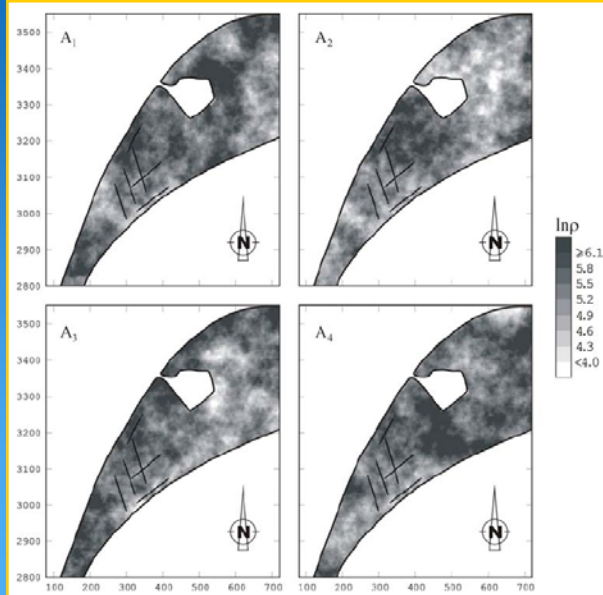
Application to a real case (Anseremme near Dinant, Belgium)

- conditioning on $\ln \rho$ (293 data points)
- inverse modelling on h (8 data points)
- verifying the coherence of the obtained K fields with regards to the 3 data points of measured K



A. Desbordes 2004

Application to a real case (Anseremme near Dinant, Belgium)

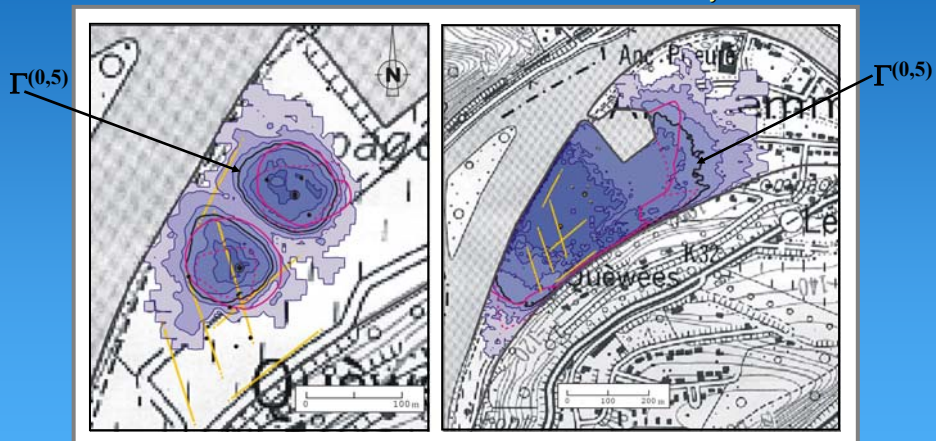


A. Desbordes 2004

Application to a real case (Anseremme near Dinant, Belgium)

$t = 24$ hours

$t = 50$ days



— \mathcal{T} of a previous 3D deterministic model
 \mathcal{T} of a previous 2D deterministic model

A. Desbordes 2004

Lessons

- stochastic co-conditional approaches bring improvements
- it does not spare us the acquisition of measured data
- selection of 'best' locations for geophysical measurements can be crucial
- if piezometric heads are used for inverse modelling
 - ➔ parameterisation / zonation in 'facies'
 - ➔ thresholds values / selection criterion
 - ➔ statistics on remaining realisations
 - ➔ if selection criterion not too bad
geostatistical structure seems to be preserved
- the 'facies' zones can be disjoint ... possible links to be found with more 'genetic' based geological analysis

Conclusions

- advantages of fusing available geological / geophysical / hydrogeological data
- optimizing soft data measurements is required ...
- assumption that geology generates 'sediment facies'
- genetic/genesis models modelling rock formation processes should help more and more ?
- when zones corresponding to 'facies' are defined, ... how to deal with a calibration ?
 - * keeping the geometry of the facies: changes in parameters values and find criteria for selection
 - * adjustment of the geometry: 'gradual deformation' (Hu et al., 2001)
- future combination with multiple point geostatistical techniques ?

- FNRS Belgium
- SWDE
- DGRNE

Thanks !

Effect of the zonation in facies and selection on the spatial structure ?

