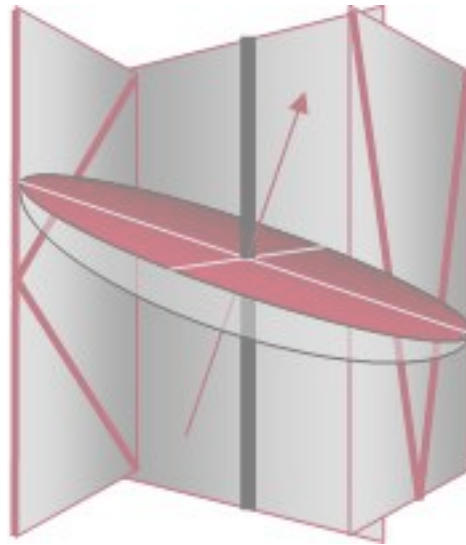


# Calibration of NFR models with interpreted well-test k.h data



**Michel Garcia**

# Calibration with interpreted well-test k.h data

## Intermediate step between

- Reservoir characterization
  - **Static model** conditioned by structural, core and fracture data
- History matching
  - **Dynamic model** conditioned by production data

## Focusing on equivalent permeabilities

- Locally measured permeabilities of the fracture system

## Step specific to NFR models

- Permeability fields **not directly modeled** but derived (calculated) from
  - Fracture densities
  - Fracture geometric and flow properties

# Aim of the calibration

## Relating

- NFR model parameters defined on a fracture-set basis
  - Frature densities
  - Geometric properties: orientation, length, height
  - Flow properties: conductivity

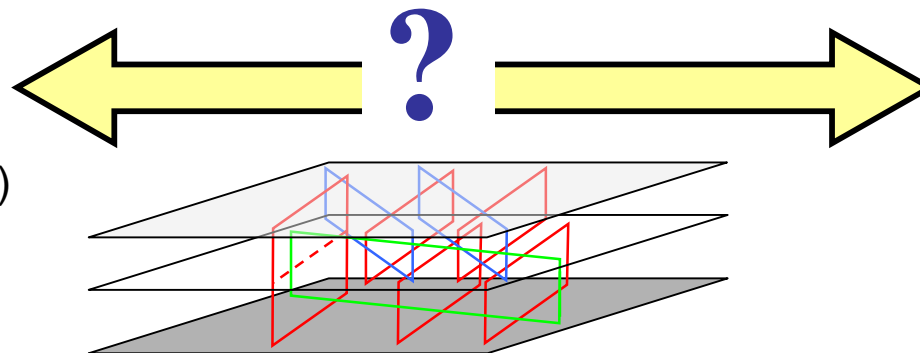
## To

- Locally measured permeabilities of the fracture system
  - Interpreted well-test k.h

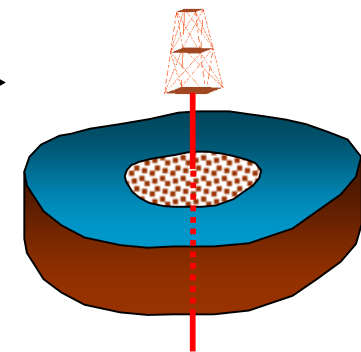
### One Fracture-set

$FD$  = density ( $m^{-1}$ )  
 $L$  = length (m)  
 $H$  = fracture height (m)  
 $ke$  = conductivity (md.m)  
 $n$  = orientation  
...

### Fracture system = all fracture-sets



### Interpreted well-test k.h



# Choice of a conceptual fracture system model

## Discrete fracture network vs. continuous fracture model

- DFN and CFM
  - Same model parameters to characterize the fracture system
  - Equivalent flow properties are to be calculated in gridblocks or at gridblock interfaces as required for reservoir flow simulation
- Particularity of DFN
  - Realization of the fracture system is part of the model  
**↻ the realization should be part of the calibration**
- Particularity of CFM
  - Equivalent flow properties locally derived from global or locally defined fracture parameters (local DFN, analytical solution)
  - If non-bijective relationship between fracture parameters and equivalent permeability tensor  
**↻ need for an additional connectivity-like parameter**

## Evaluation of the (flow) well-test response of NFR models

- Requirements on the calculation method
  - Fast and automatic method for inversion purposes
  - **Grid support** for consistency with the reservoir flow simulator
  - Applicable to permeability **tensor** fields & anisotropic drainage areas
  - Distinction between fracture and matrix permeabilities

## Optimization of model parameters

- Numerous model parameters
  - Several mechanical unit dependent directional fracture-sets
  - Each fracture-set characterized by  $\geq 4$  parameters
- Poorly known fracture parameters
  - Spatial vs. non-spatial parameters
  - Single (effective) value vs. probability distribution
- Connectivity of the fracture system = missing parameter

# Evaluation of interpreted well-test k.h

## Forward problem

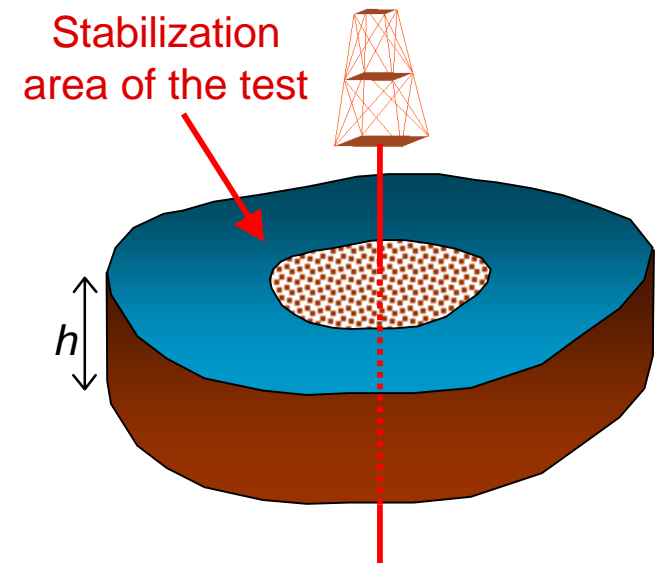
– Interpreted k.h = average k.h corresponding to

- Planar and radial-like flows around the well
- Uniform homogenization of k (or more precisely k.h) within a drainage (stabilization) area around the well
- Instantaneous pressure equilibrium in the matrix and the fracture system
- Average (effective)  $k = (k_{max} k_{min})^{1/2}$   
 $k(u) \approx k_f(u) + k_m(u)$

– Particularities of NFR

CFM

- Upscaling of fracture densities
- Permeability field = tensor field
- Anisotropic drainage area  
( $R_{max} / R_{min}$  up to 10 or more)



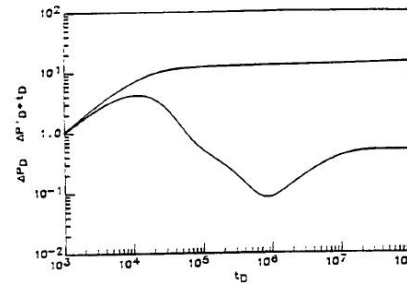
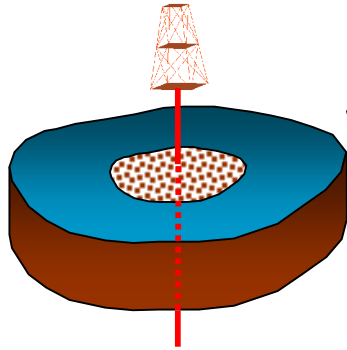
# Evaluation of interpreted well-test k.h (cont.)

## Non-exhaustive list of methods for CFM

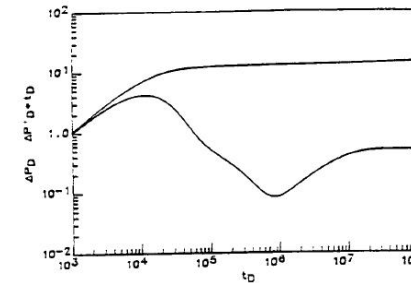
- 1 Numerical simulation of transient (well-test) flow responses
- 2 Power or other types of averaging based on calibrated data
- 3 Annulus ring-based calculation of apparent test-permeability
- 4 Steady-state flow based calculation of around-well average permeability (TOTAL/KIDOVA approach)

# Evaluation of interpreted well-test k.h (cont.)

## Numerical simulation of transient (well-test) flow responses



*Simulated test*



*Experimental test*

### – Possible comparisons

- Automatically interpreted simulated tests vs. interpreted well-test
- Simulated vs. experimental pressure (or pressure variation) curves  
○ relying on raw data instead of a reservoir engineer's interpretation

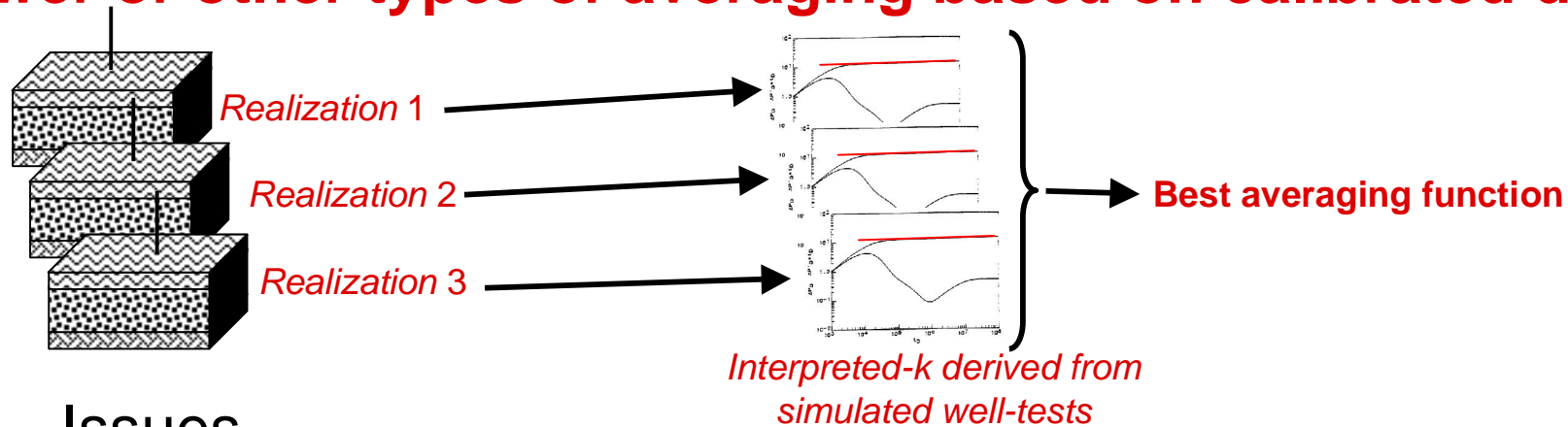
### – Issues

- Complex numerical simulations: local grid refinement, well-model
- Transient flows = additional model parameters (e.g. fracture porosity, compressibility)
- What about fracture/matrix exchanges?



# Evaluation of interpreted well-test k.h (cont.)

## Power or other types of averaging based on calibrated data



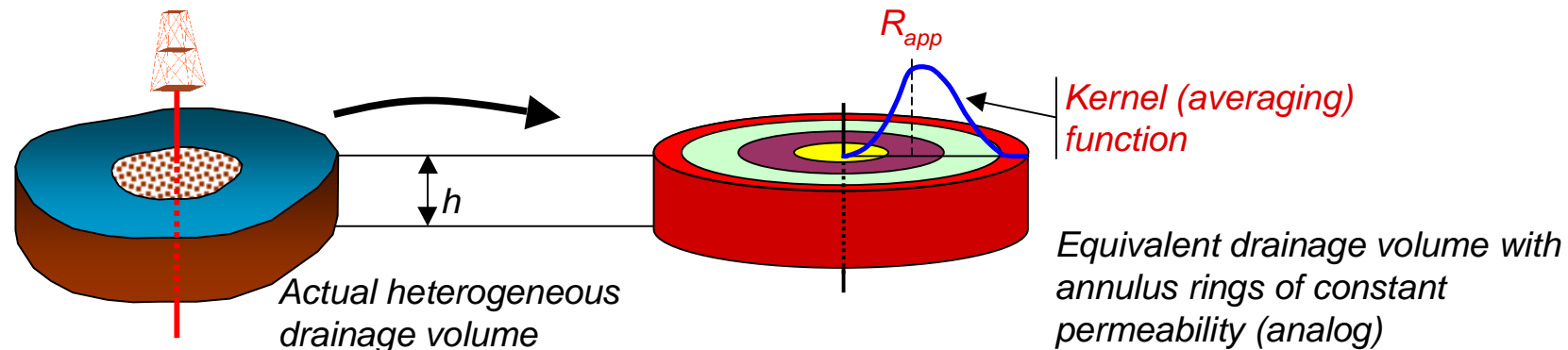
### – Issues

- Applied to scalar permeability fields (not permeability tensor field)
- Defining drainage volumes (generally cylindrical)
- Relying and depending on (inferred) spatial statistics
- Same calibration function applying everywhere

Ref. Power averaging: Deutsch 1992. Multiple point proxy: Srinivasan & Caers 2000, Noetinger 1994

# Evaluation of interpreted well-test k.h (cont.)

## Annulus ring-based calculation of apparent test-permeability



### – Issues

- Applied to scalar permeability fields (not permeability tensor field)
- Defining equi-travel time rings (generally circular)
- Calibration of the weighting function that should apply everywhere

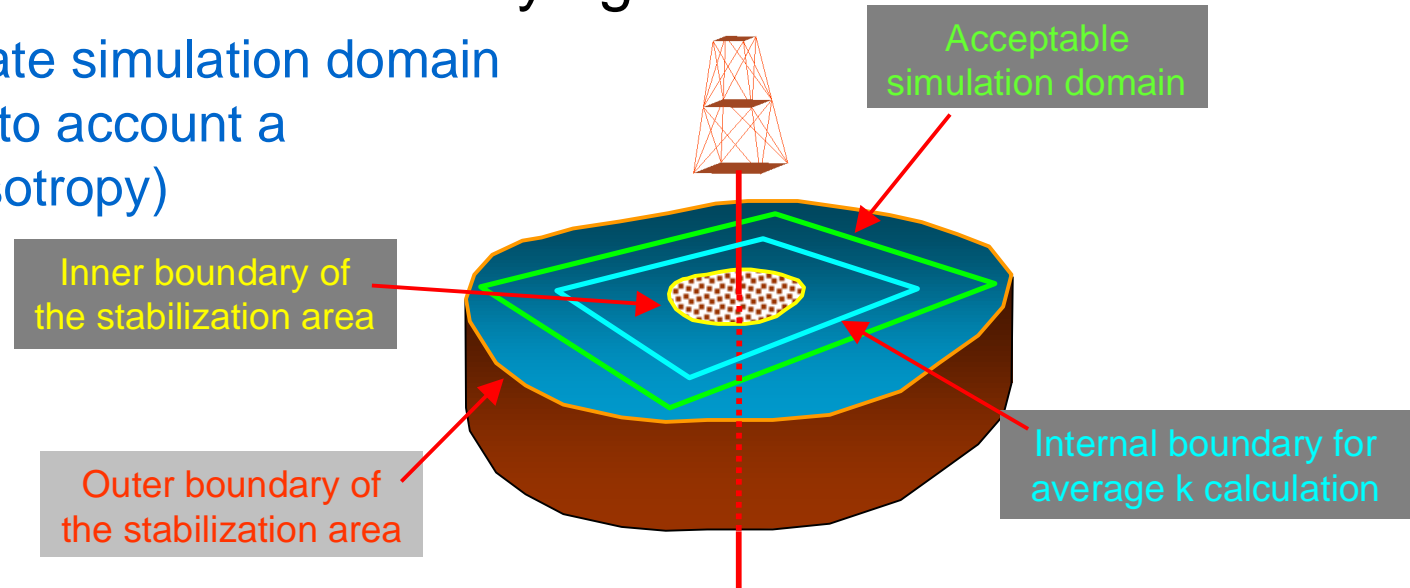
Ref. Gautier & Noetinger 2004, Sagar 1993, Oliver 1989

# Evaluation of interpreted well-test k.h (cont.)

## Effective-gradient based averaging (TOTAL/KIDOVA method)

– Steady-state flow solution relying on:

- Appropriate simulation domain (taking into account a local anisotropy)



- Appropriate boundary conditions (notion of energy)
- Appropriate calculation of average k.h vs. Rd (equivalent drainage radius) around the well (notion of effective gradient)

$$k_{\Gamma} = \frac{\int_{\Gamma} q(\gamma) d\gamma}{\int_{\Gamma} q(\gamma) \|\mathbf{K}(\gamma)^{-1} \mathbf{n}(\gamma)\| d\gamma}$$

– Ref. Garcia, Goult & Gosselin 2007 and 2006

# Evaluation of interpreted well-test k.h (cont.)

## Effective-gradient based averaging (TOTAL/KIDOVA method)

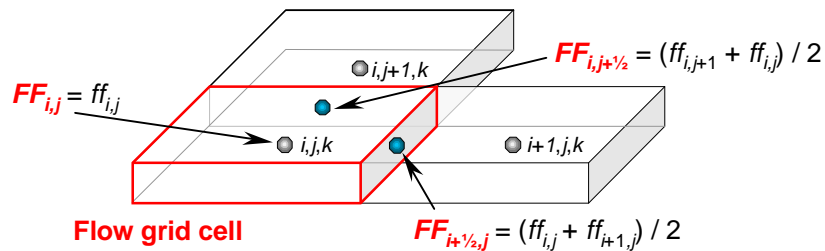
– Flow simulation

- Need for a flow simulator allowing full permeability-tensors

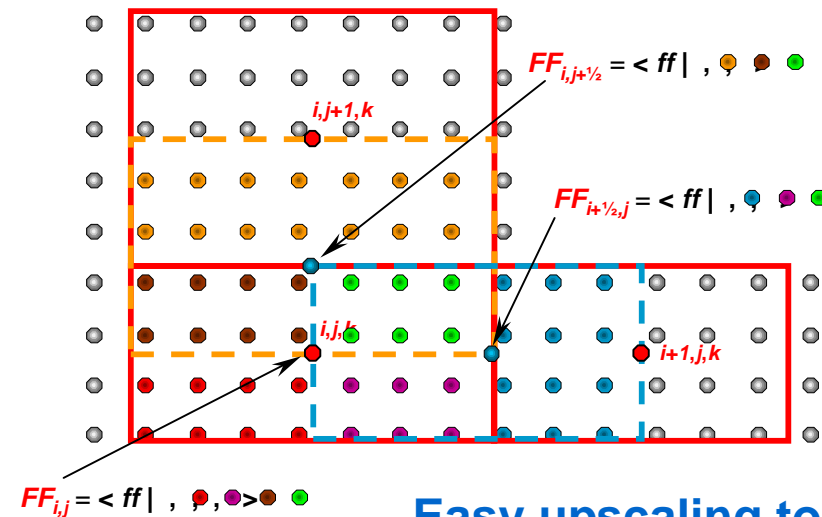
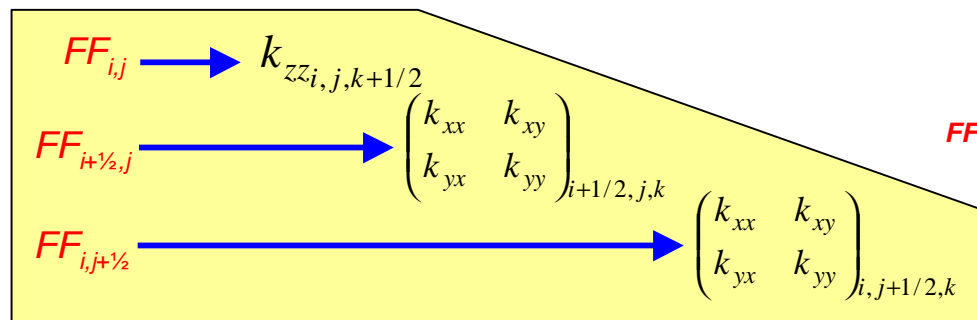
$$\begin{pmatrix} k_{xx} & k_{xy} & 0 \\ k_{yx} & k_{yy} & 0 \\ 0 & 0 & k_{zz} \end{pmatrix}_f + \begin{pmatrix} k_{xx} & k_{xy} & 0 \\ k_{yx} & k_{yy} & 0 \\ 0 & 0 & k_{zz} \end{pmatrix}_m$$

– Permeability tensors defined at gridblock interfaces

○ = ff simulation node



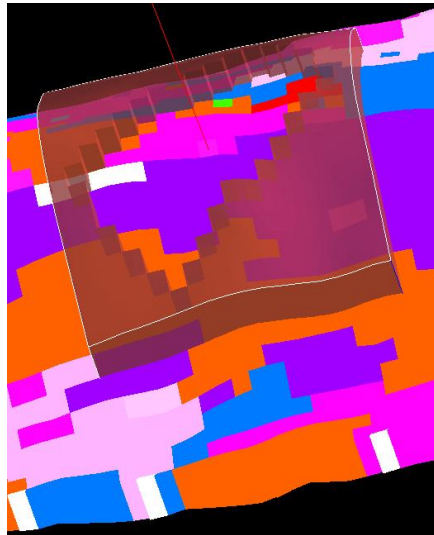
On the FF grid



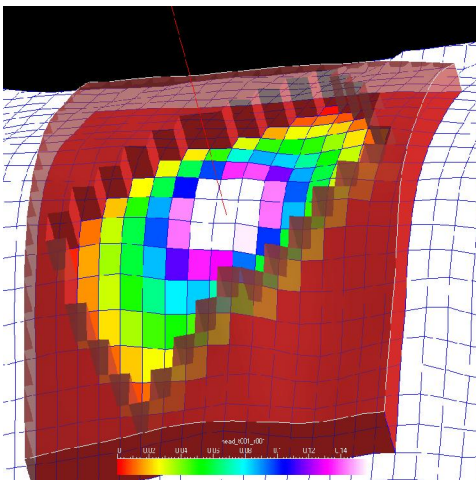
Easy upscaling to a coarser grid

# Evaluation of interpreted well-test k.h (cont.)

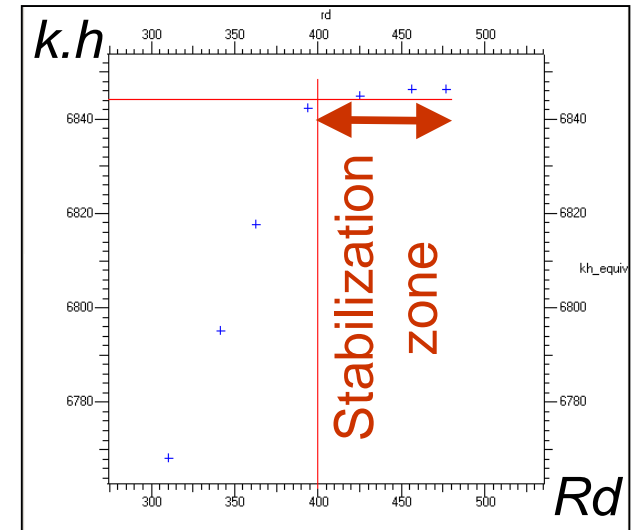
## Illustration of the effective-gradient based averaging method



Appropriate simulation domain:  
based on local  $k$   
anisotropy +  
drainage radius



Flow simulation:  
based on  
appropriate  
boundary conditions  
+ full  $k$ -tensor simulator



Effective-gradient  
based calculation of  
 $k.h$

# Calibration of model parameters

## Position of the problem

- $N_d$  = nb of interpreted well-test (k.h) data  
Well location + interpreted k.h + associated  $R_d$
- $N_p$  = nb of model parameters  
= nb of fracture-sets x nb of fracture parameters  
+ nb structural parameters
- Typically:  $N_p >$  if not  $\gg N_d$ 
  - ⊖ “under determined” problem
  - ⊖ multiple solutions

# Calibration of model parameters (cont.)

## Choice of an uncertainty model for each parameter

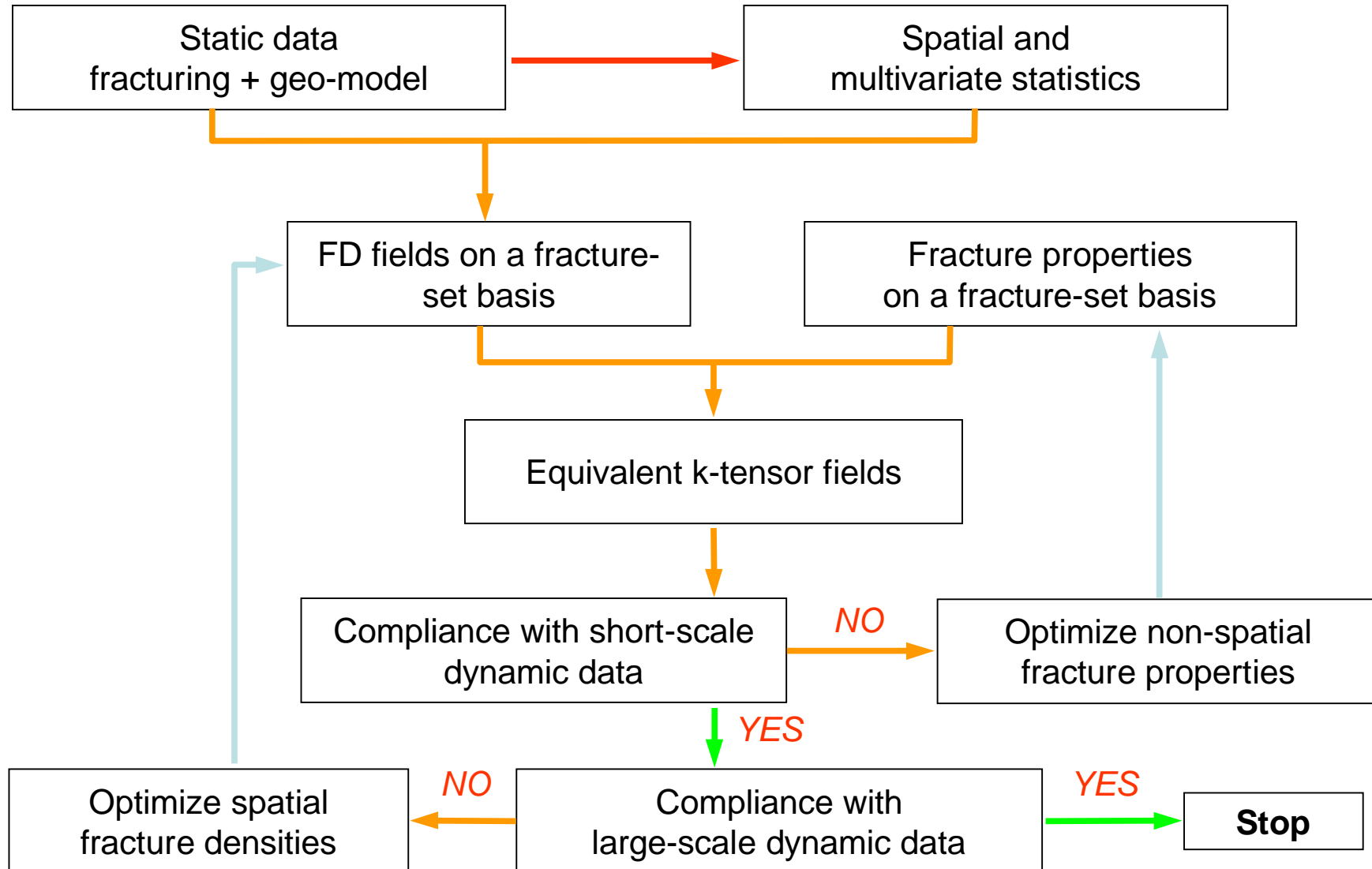
- Single value vs. probability distribution function (pdf)
  - pdf  $\bar{\theta}$  2 or more distribution parameters (central and dispersion statistics) for each (probabilistic) model parameter
- Spatial vs. non-spatial (global)

## Usual practice

- Layer thickness ( $T$ ): spatial + deterministic structural model
- Fracture parameters on a (directional) fracture-set basis
  - Fracture density (FD): stochastic spatial distribution (realization)
  - Orientation ( $dir, dip$ ): pdf + possibly spatially varying mean direction
  - Other fracture properties ( $L, ke, H$  or  $Pij$ ): different options
    - ∅ Single (effective) value often sufficient  
Ex. effective length (Masihi & King 2008)
    - ∅ Spatially defined: if correlated to a well-known spatial variable  
Ex.  $ke \propto e^{-\lambda.FD}$  (Pollard & Segall 1987, Pollard & Gross 2000)

# Calibration of model parameters (cont.)

## Two-step approach for non-spatial and spatial parameters





# Calibration of model parameters (cont.)

## Assisted and automatic approach (TOTAL/KIDOVA)

- Sensitivity analysis
  - Characterization of parameter uncertainty
- Inversion/optimization of non-spatial parameters
  - Experimental design (random starting points)
  - Gradient-based method (use of NPSOL)
  - Objective function

$$O(\mathbf{np}) = \left( \frac{1}{\sum_{r=1}^{N_r} 1/O_r^0(\mathbf{np})} \sum_{i=1}^{N_d} w_i \sum_{r=1}^{N_r} \frac{1}{O_r^0(\mathbf{np})} \sum_{i=1}^{N_d} w_i \left| \frac{d_i - d_i^*(\mathbf{np} | \mathbf{sp}^r(\mathbf{u}))}{\sigma_i} \right|^\omega \right)^{1/\omega}$$

- Aim

- ∅ Finding non-spatial parameter values matching at best dynamic data
- ∅ Identifying “best” fracture density realizations
- ∅ Managing local minima or non-unique solutions (multiple starting points)

Weighted FD  
realizations

Model response vs.  
Dynamic data

# Calibration of model parameters (cont.)

## Uncertainty characterization and sensitivity analysis

### – Objectives

- Identifying the most consequential uncertain model parameters
- Fine-tuning their possible range
- Eliminating parameters without or with limited effects on matching

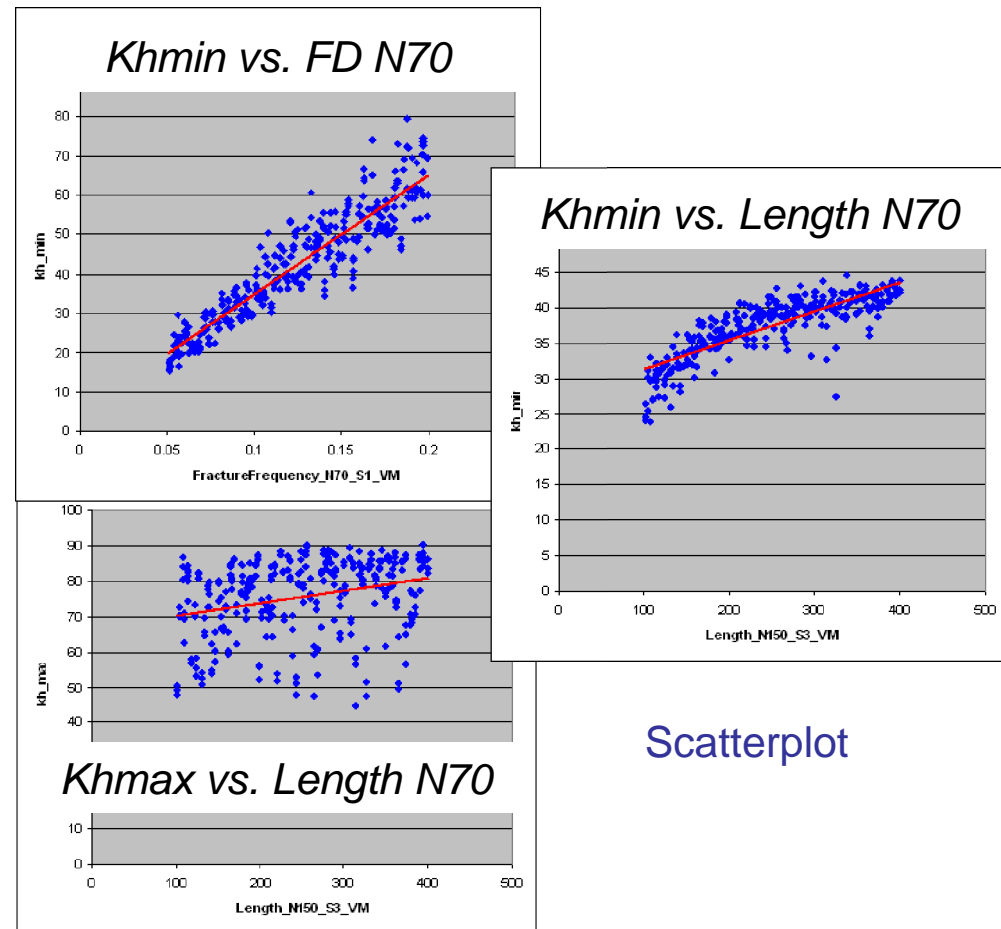
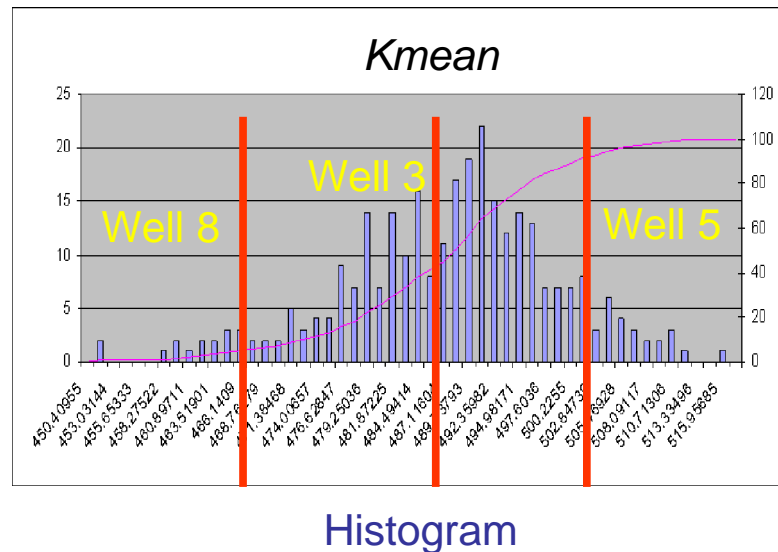
### – Tools

- Latin hyper-cube Monte Carlo sampling
- Uni and multivariate statistical analyses of results

# Calibration of model parameters (cont.)

## Uncertainty characterization and sensitivity analysis

– Typical results

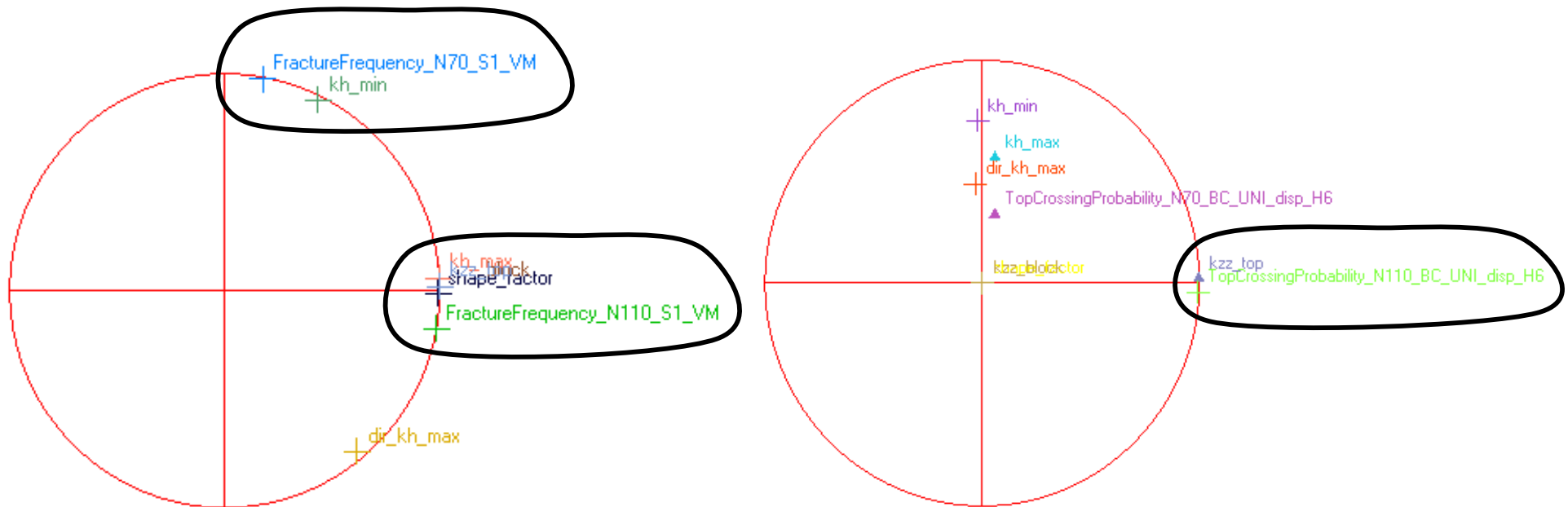


– Ref. M.E. Luna Orosco Garcia, MSc. Reservoir Geosciences and Engineering, IFP School, 2007

# Calibration of model parameters (cont.)

## Uncertainty characterization and sensitivity analysis

– Typical results



– Ref. M.E. Luna Orosco Garcia, MSc. Reservoir Geosciences and Engineering, IFP School, 2007

# Calibration of model parameters (cont.)

## Automatic calibration of non-spatial parameters

The image shows two overlapping screenshots of a software interface. The top screenshot displays the 'Dynamic Data to Be Matched' section, which includes a table for selecting realizations and a table for selecting well-test data for calibration. The bottom screenshot displays the 'Matching Method' section, where the user can choose between non-spatial and spatial optimization methods.

**Dynamic Data to Be Matched**

Select one or more realizations

	Realization
1	<input type="checkbox"/> No 1
2	<input checked="" type="checkbox"/> No 2
3	<input type="checkbox"/> No 3

**Choice of FD realizations and dynamic data**

Select k.h test(s) for calibration

	Well	Kh	Rd	Bottom layer	Top layer	Error(%)
1	<input checked="" type="checkbox"/> B3	1500	1000	EclGrid_H10	EclGrid_H5	5
2	<input checked="" type="checkbox"/> B3	4300	500	EclGrid_H26	EclGrid_H0	5

**Matching Method**

Non-spatial parameters

- EST
- Gradient-based optimization method

Spatial parameters

- Geostatistical inversion

**Choice of optimization method(s)**

# Calibration of model parameters (cont.)

## Automatic calibration of non-spatial parameters

- Grid – 200\*200\*6 cells
- Four zones with FD of 0.3, 1.5, 3 & 7 m<sup>-1</sup>
- Two fracture sets N70 & N150
  - N70:  $L=10$  m,  $ke = 400$  mD.m
  - N150:  $L=15$  m,  $ke = 250$  mD.m

Matching results

k,h test: 4

Well: Vert\_04 - Kh: 74620 - Rd: 800 - Bottom: SquareMechUnitGrid\_H3 - Top: SquareMechUnitGrid\_H6

	Nb iterations	Objective function	End status	K,h real 1
Run 1	6	0.412819	6	73541.1
Run 2	4	0.328113	6	74789.3
Run 3	4	0.601585	6	72297.4

Best result Run: 2 Realization: 1

Show output files

Apply calibrated model parameters

Back Next Report Help

Fracture set	Parameter	Run 1	Run 2	Run 3	TRUE
N70	$L$	170.1	348.0	871.8	<b>400</b>
N70	$ke$	19.5	11.8	11.5	<b>10</b>
N150	$L$	125.6	909.1	136.2	<b>250</b>
N150	$ke$	13.8	22.3	28.7	<b>15</b>

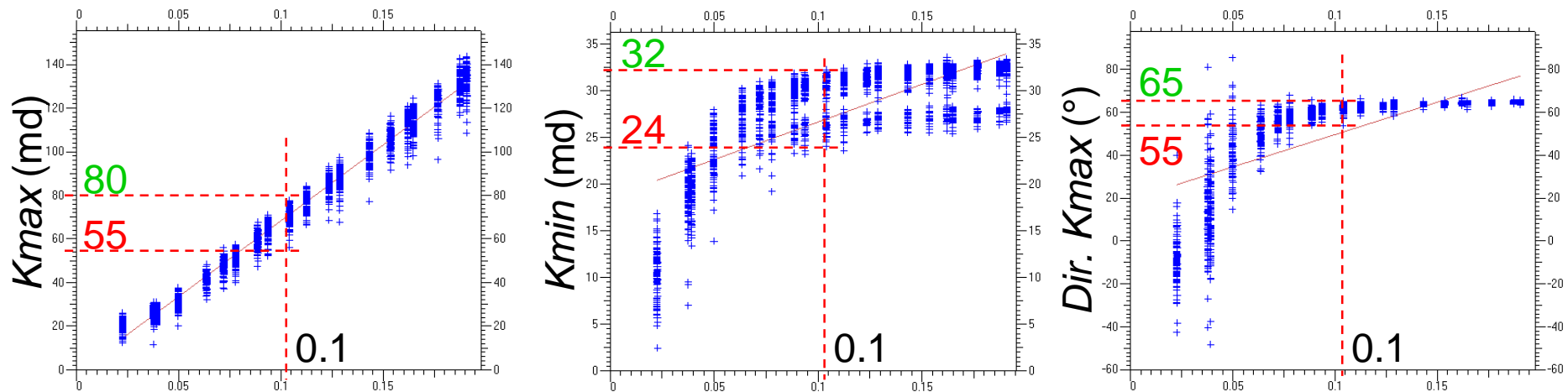
– Ref. Egor Mikhaylenko, MSc. Reservoir Geosciences and Engineering, IFP School, 2008

# Calibration of model parameters (cont.)

## About the need for a connectivity parameter

- Calculation uncertainty between fracture-set parameters and equivalent permeability tensor

Non-bijective relation = sensitivity to some DFN realizations



$FD-N70$  (m<sup>-1</sup>) = only non-constant parameter

- Connectivity parameter = additional (locally or globally) defined model parameter to discard calculation uncertainty

***Acknowledgements:***



**TOTAL**

***for its support and permission to publish results***



- **Evaluation of interpreted k.h test data on NFR models**
  - Easy forward problem, complex evaluation task
- **Automated vs. assisted model parameter calibration**
  - Practice of sensitivity analysis and optimization
- **Which support for calibration: DFN vs. CFM?**
  - Consistency of model calibration with history matching
- **Multiresolution of NFR model parameters**
  - Trade-off between model complexity and available data

# References

## Conditioning to well-test data

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## Effective properties

- Masihi, M. and King, P.R. (2008). Connectivity Prediction in Fractured Reservoirs With Variable Fracture Size: Analysis and Validation. SPE J.13 (1): 88-98. SPE-100229-PA