25th Anniversary Meeting of the Stanford Center for Reservoir Forecasting, May 9 2012



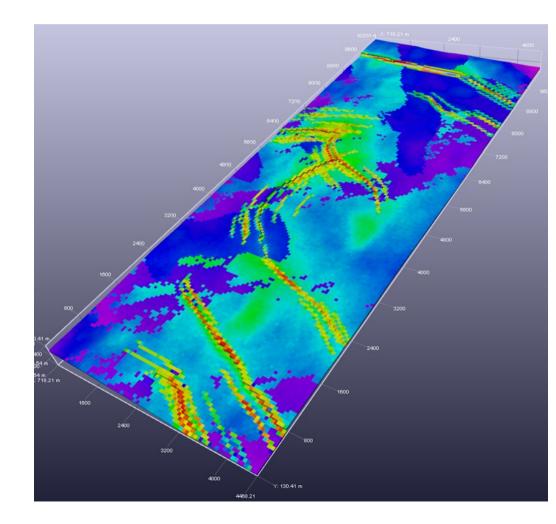
Accounting for Uncertainty in Dual Porosity Descriptions of Fractured Systems

Andre Jung, Jef Caers, Stanford University Darryl Fenwick, Streamsim Technologies

Two Worlds

Models <u>built</u> by Geologists

Models <u>used</u> by Engineers

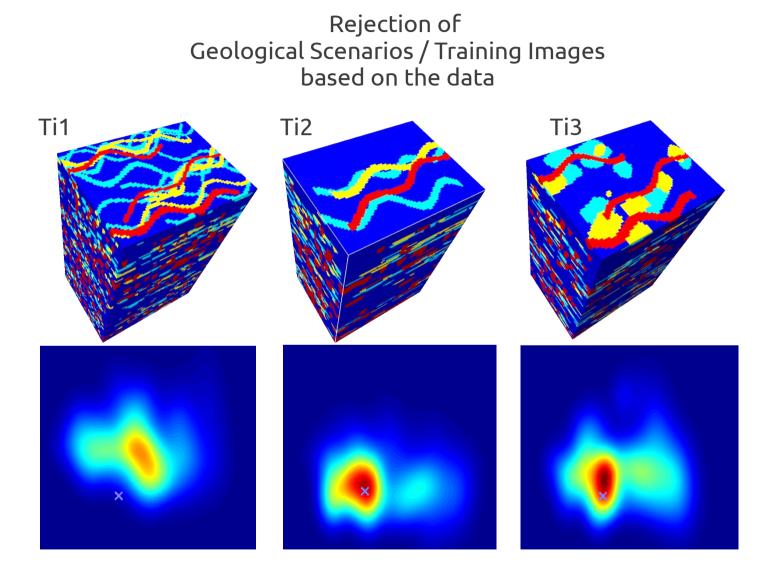


Naturally Fractured Reservoirs

The Purpose:

- Uncertainty Quantification
- History Matching
- The Challenge:
 - Geologically consistent
 - Field Scale
 - Integration with existing Software

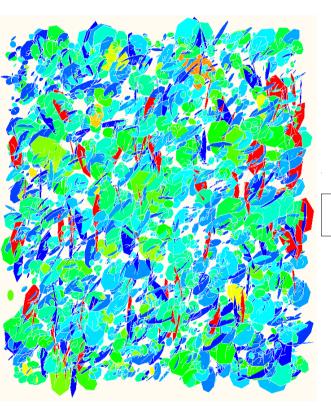
... this morning: Scenario Uncertainty



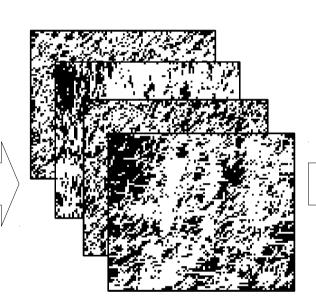
Hyucksoo Park, Céline Scheidt and Jef Caers

Can we build Training Images for Fractures?

General Overview of Methodology



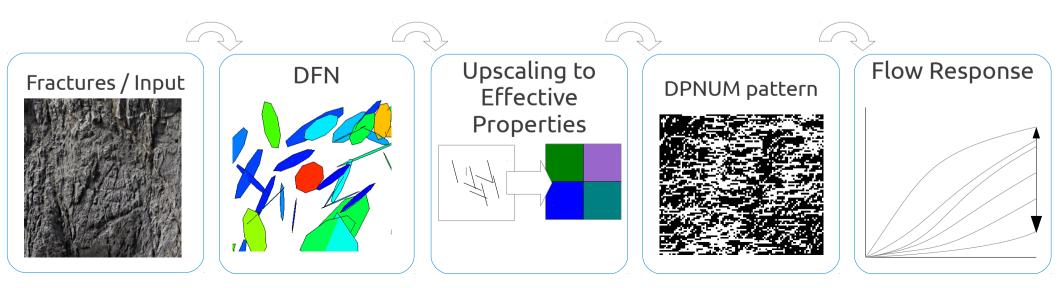
Complex Fracture Models



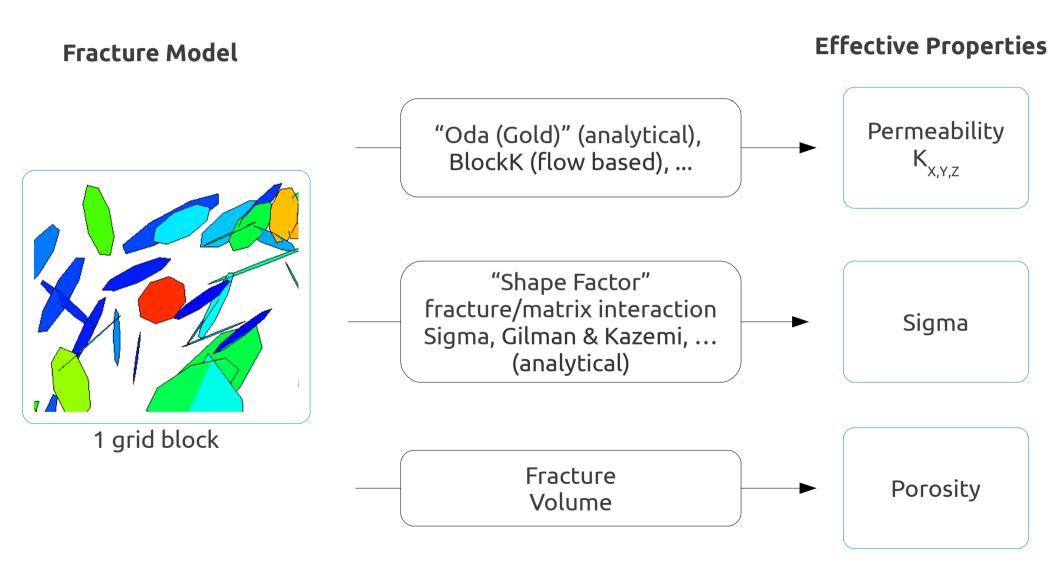
Limited / Manageable Set of flow-based scenarios for Training Images Geostatistics
History Matching
PPM
....

The Idea

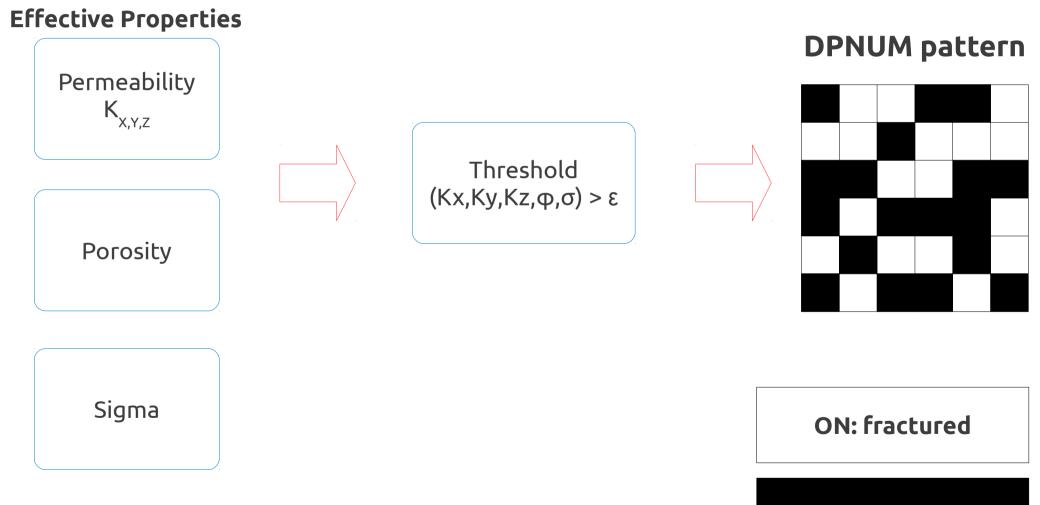
- Generate Discrete Fracture Network (FracMan)
- Upscale to Effective Properties (grid)
- Decide which cells are Dual-Porosity → pattern
- Run Streamline Simulation (3DSL)
- Select flow responses via pattern-based distances



Upscaling Fracture Model (DFN) to Flow Model

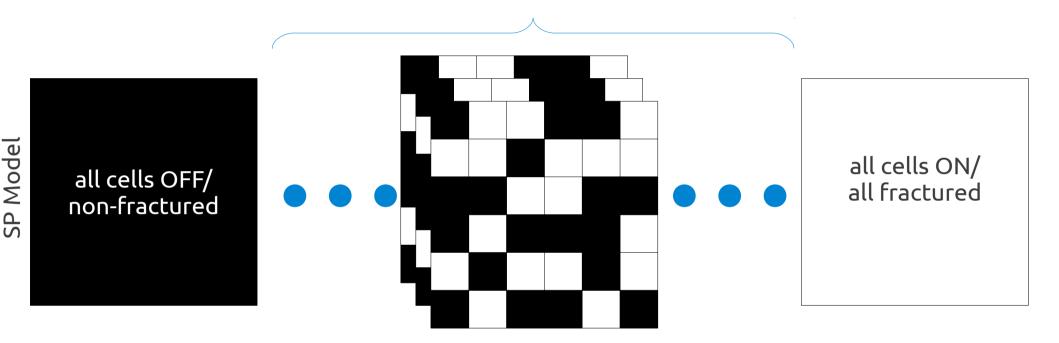


DP model: ON/OFF



OFF: non-fractured

DPNUM pattern

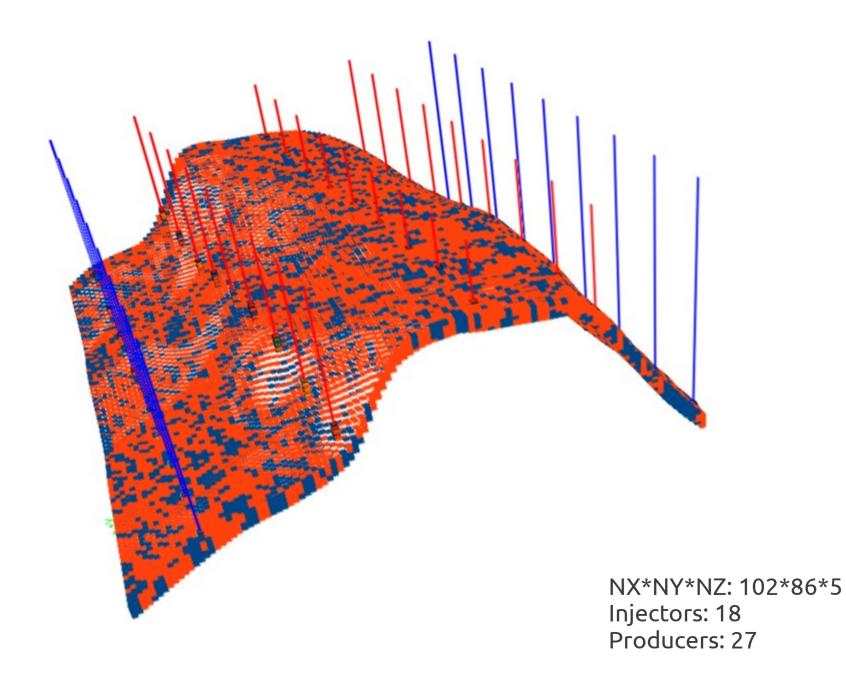


ON: fractured

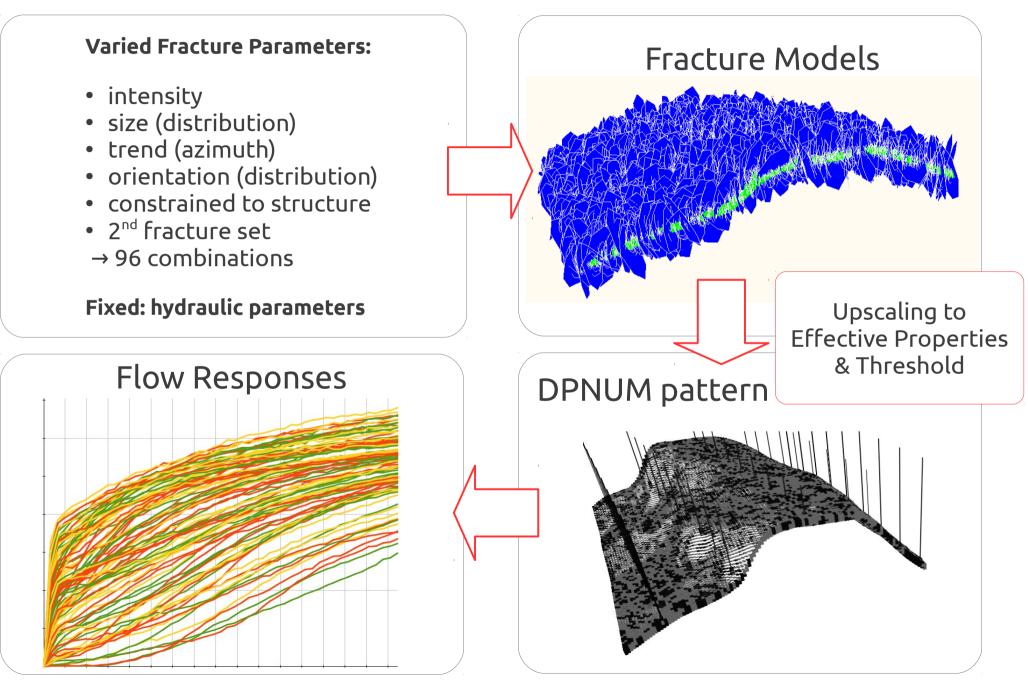
OFF: non-fractured

Illustration 1

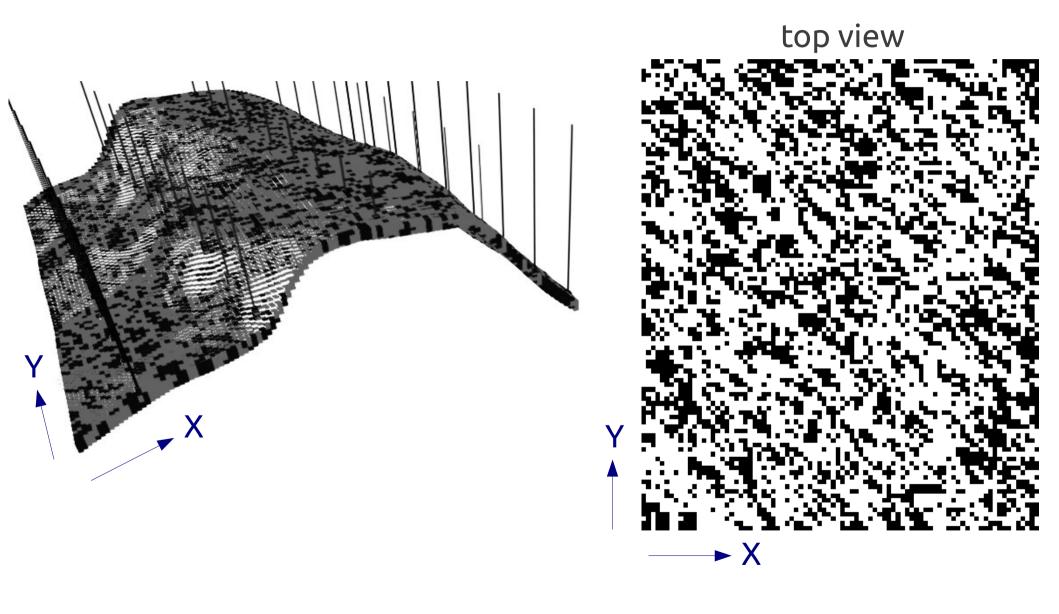
Streamline Simulation Model



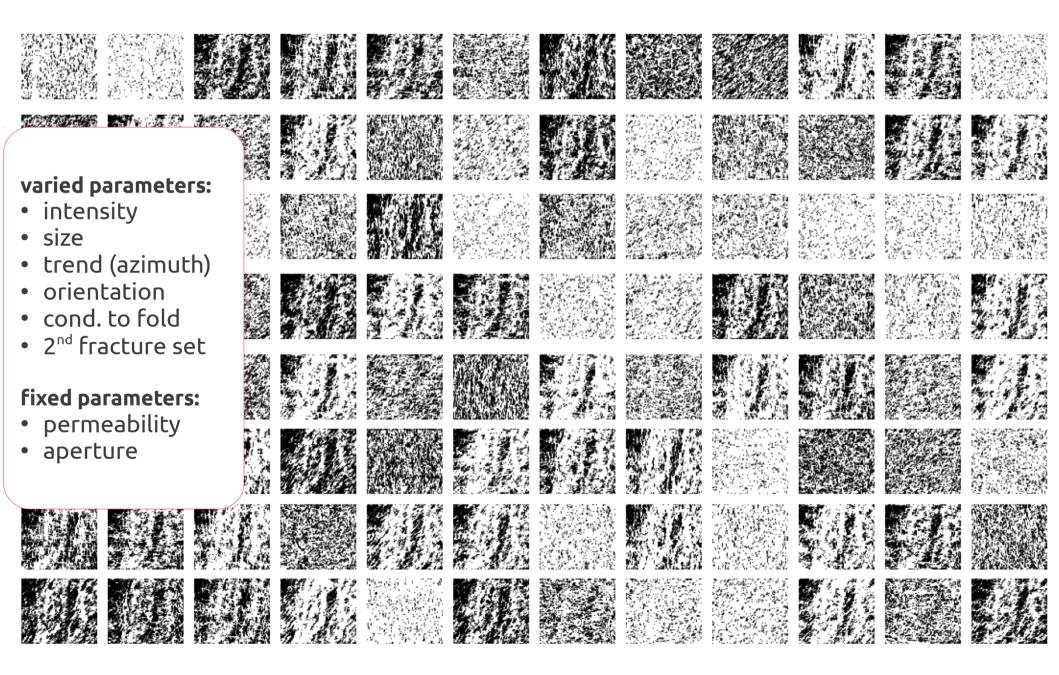
Building the Illustration Case



pattern from different perspectives



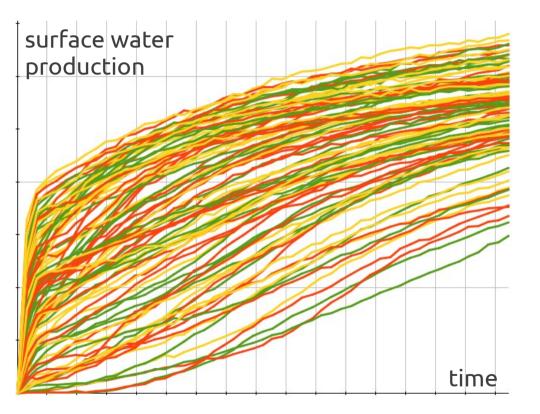
96 patterns through parameter variation

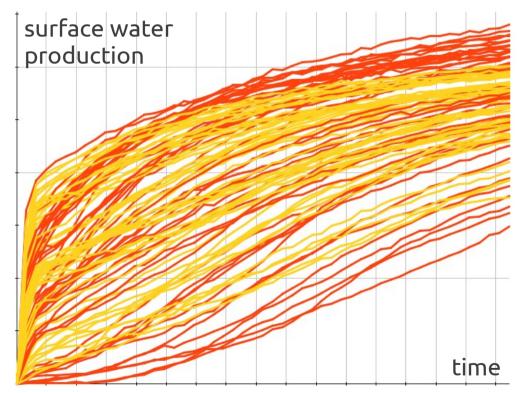


96 Field Responses: sensitivity to DFN input parameters

Trend (Azimuth) of Fractures

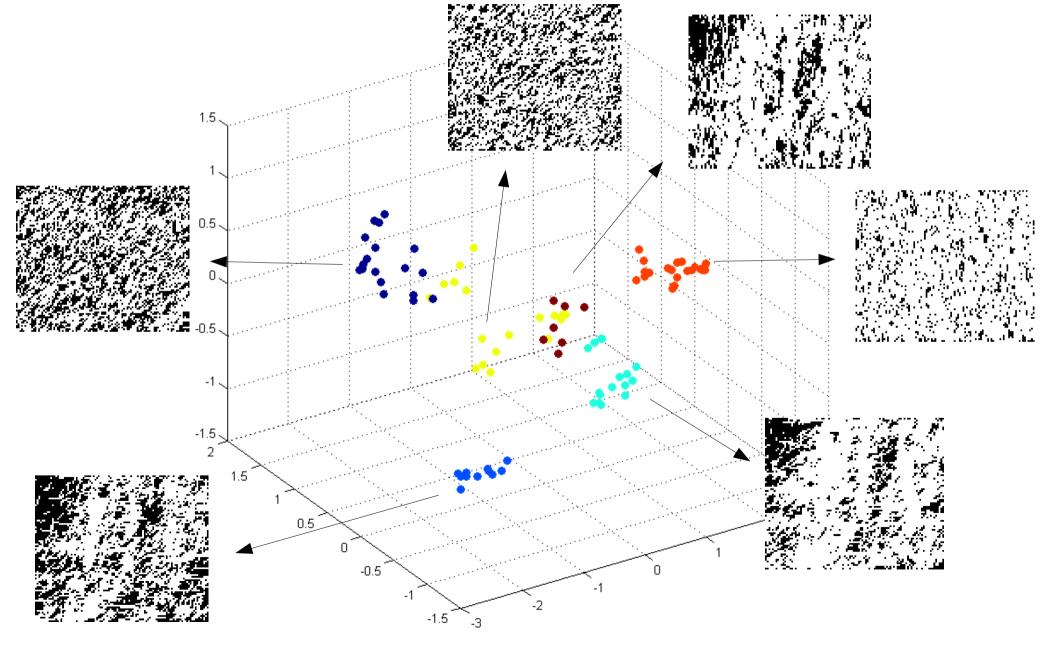
Intensity constraint to structure: yes/no





Let's try an approach based on the patterns

Patterns clustered by Modified Hausdorff Distance



96 Field Responses by pattern

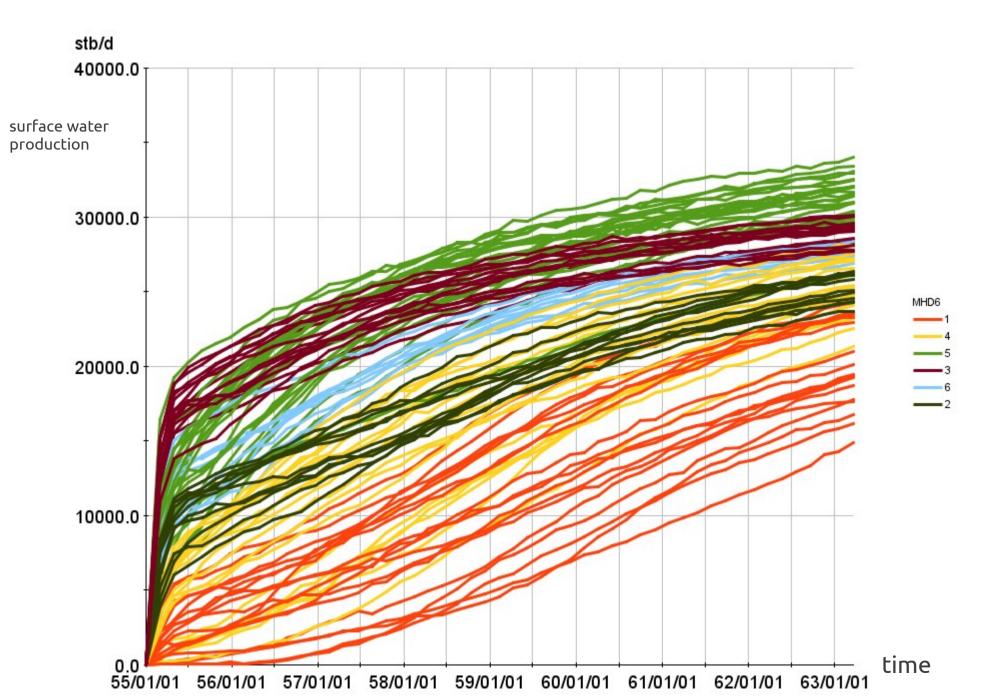


Illustration 2

Spatial Uncertainty

Spatial Uncertainty

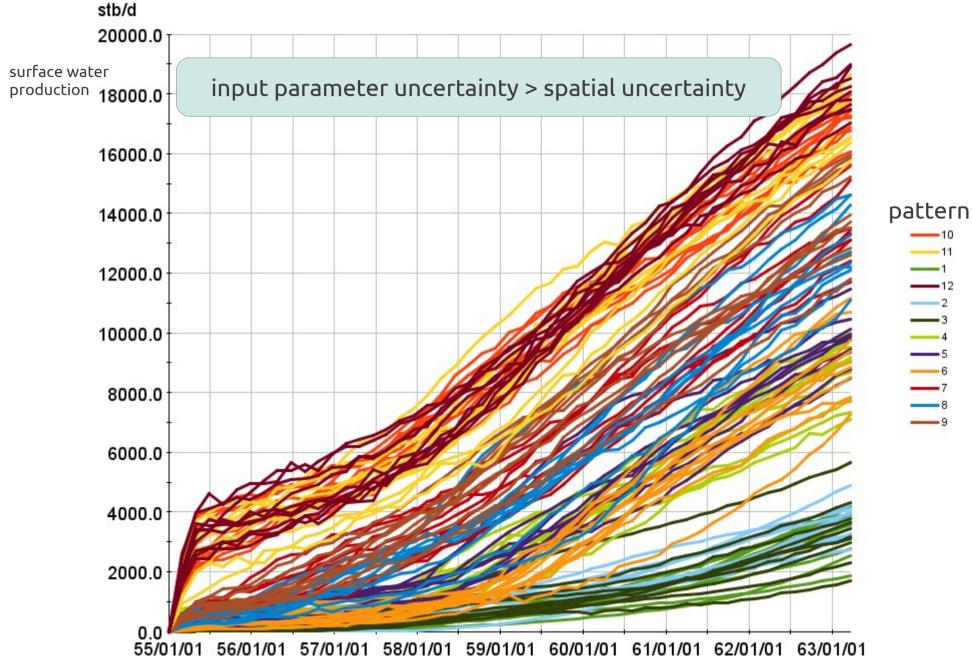
- Pick 4 DFN parameter sets (distinct patterns)
- Apply 3 variations (trend of fractures)
- Run 10 Realizations of DFN per parameter set
 - → 120 Flow Responses
- Group Flow Responses by parameter set

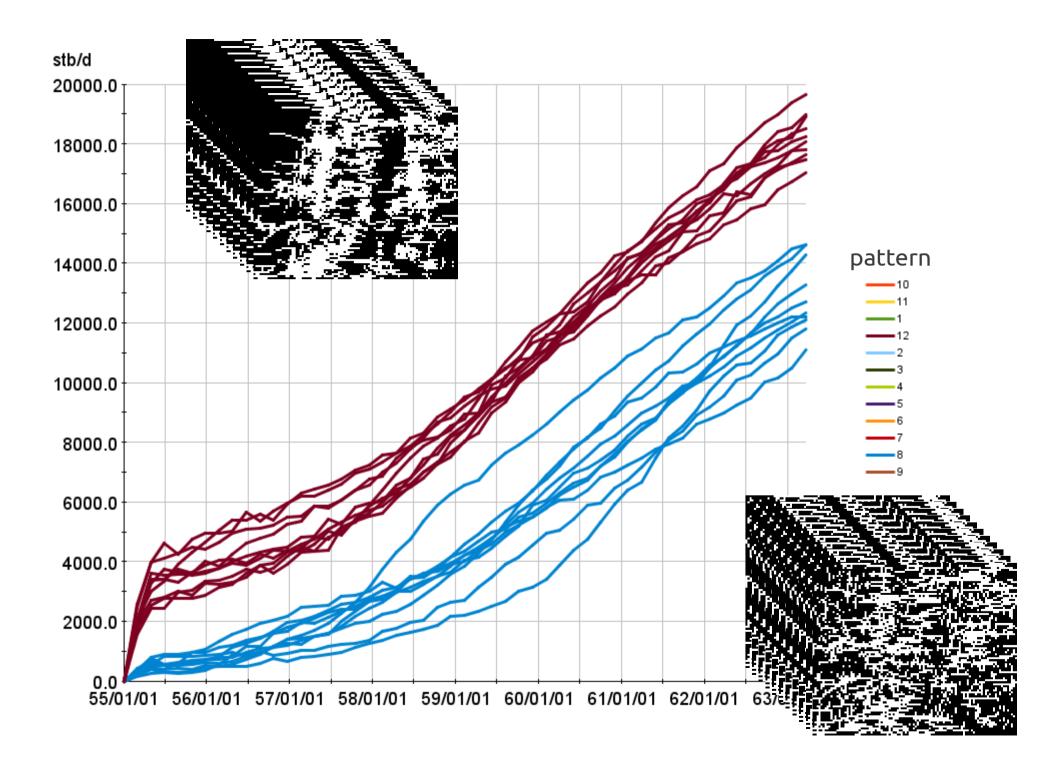
Spatial uncertainty of patterns

A. C. orientation 12 patterns orientation conditioned to fold orientation 2nd fracture set orientation conditioned to fold and 2nd fracture set

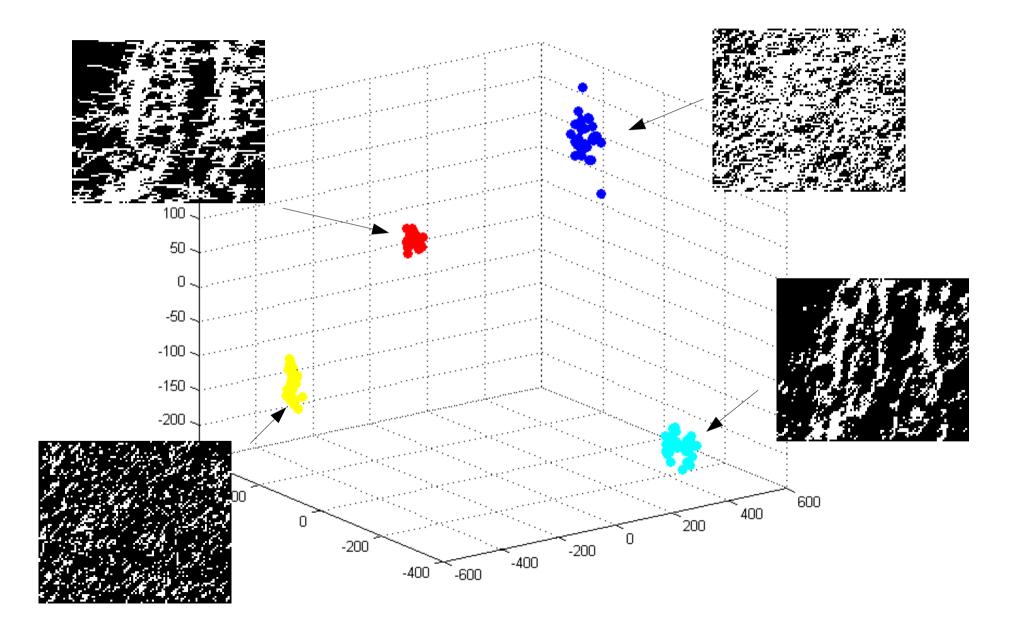
10 realizations

Field response by parameter set / pattern

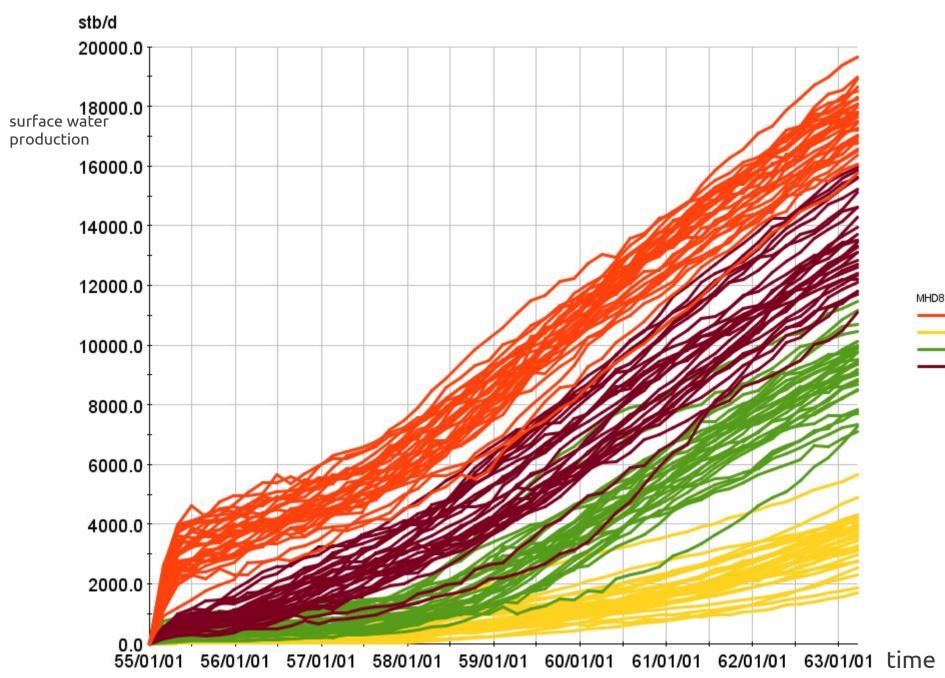




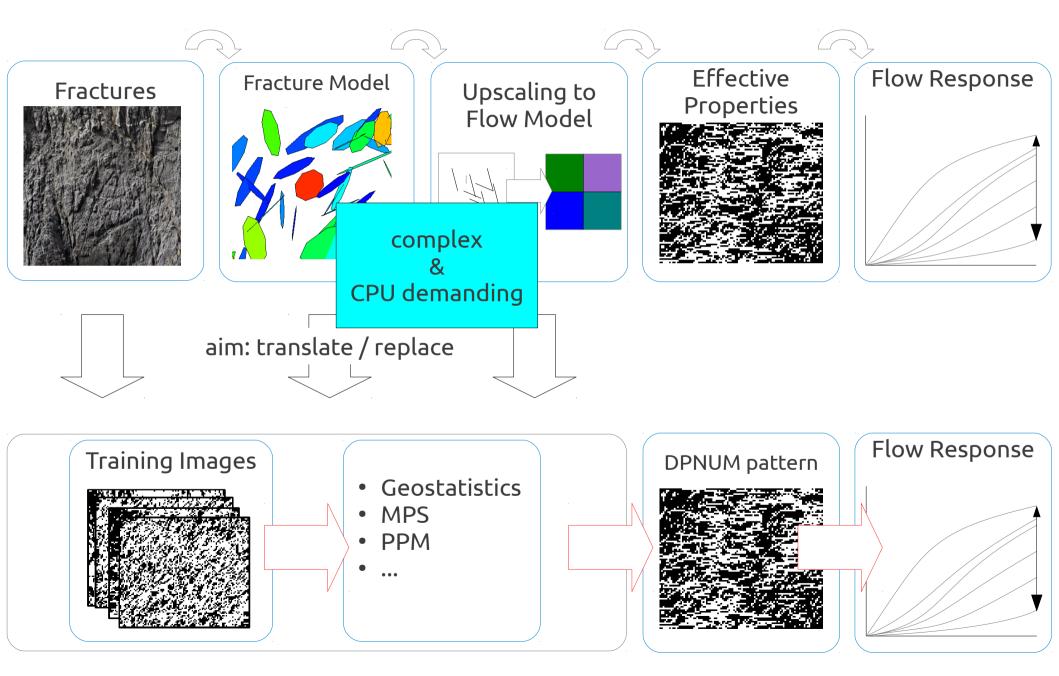
Patterns clustered by Modified Hausdorff Distance



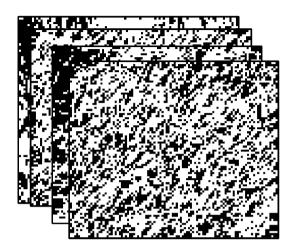
Field response by pattern



establishing a new workflow

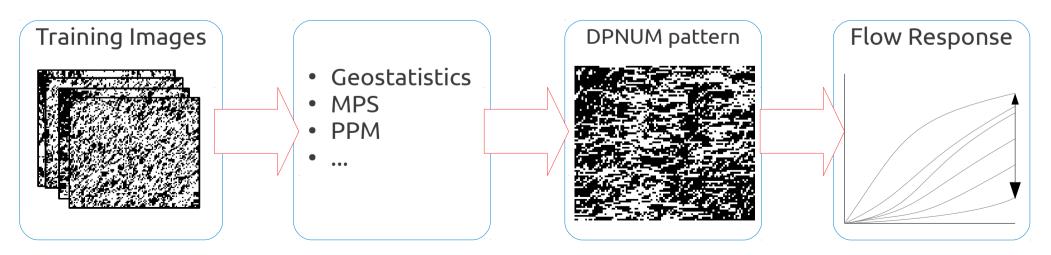


establishing a new workflow



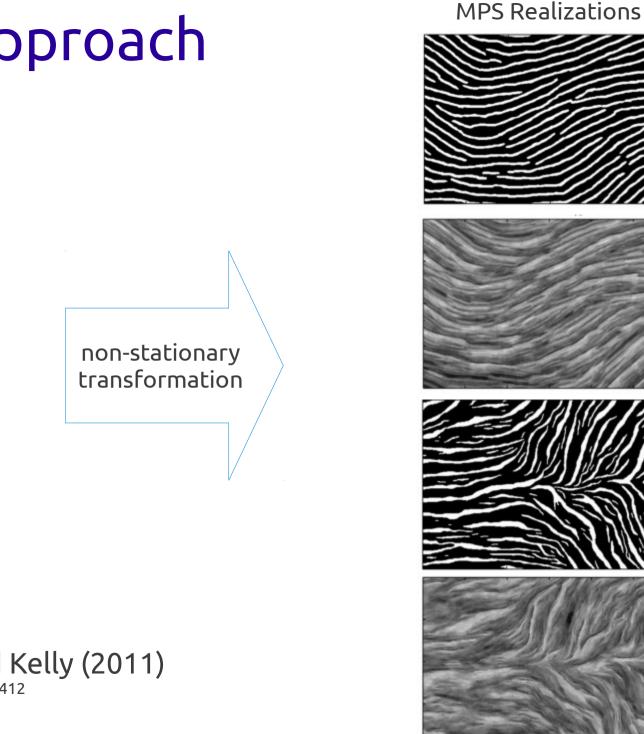
training images for fractures

- bypass complexity of DFN modeling & upscaling
- by using fracture patterns as training images
- conserve geologic realism via training images
- simulate fracture patterns directly
- less CPU demanding



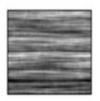
reasonably capture uncertainty





possible approach

Training Images





Mariethoz and Kelly (2011)

DOI: 10.1029/2011WR010412

Conclusions

- Translation of DFN Models to Grid/Pattern Domain
- Training Images for Fractures
- Preservation of Geological Realism
- Easy Integration with existing Software

Acknowledgments

FracMan (Golder Associates)

- Neal Josephson
- Aaron Fox
- Glori Lee

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