

Overview of Fracture Simulation Methods



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Outline

- Basics (Defining Fractures)
 - Density
 - Length
 - Orientation
 - Aperture & Transmissivity
- Discrete Fracture Models
 - Different approaches for fracture locations
- Pixel-Based
 - Fracnet
 - FCM



How Do Rocks Break?

- Multiple processes and stress fields lead to the final fracture pattern that we can observe

Scanline: measurement of fracture locations, or distances between locations, on a 1-D line perpendicular to the orientation of the fractures

Examine a few end-member fracturing mechanisms to understand spacing distributions



Fracture Measurements

- Intensity, λ , fractures per length ($1/L$ in 1-D)

$$\lambda = \frac{N}{L}$$

– In 2-D, length/area and in 3-D, area/volume

- Spacing, S , length between fractures (L in 1-D)

$$\bar{S} = \frac{1}{N} \sum_{i=1}^N S_i = \frac{L}{N} = \frac{1}{\lambda}$$



Random Breakage

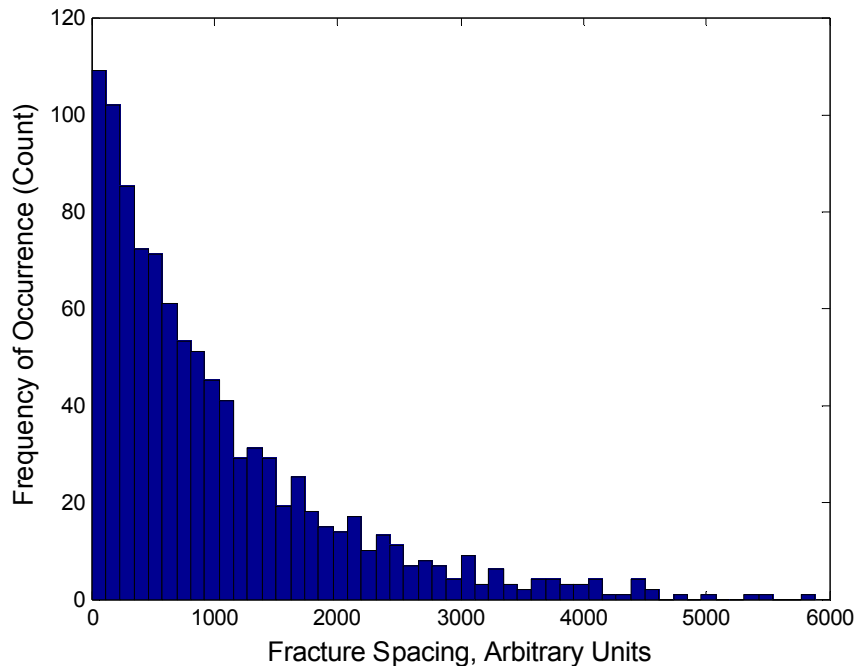
- Fracture locations are random over a distance of rock
- Could occur due to uniform stress applied to a rock with randomly located pre-existing flaws
- Fractures are the result of a Poisson process
 - What does this say about fracture spacing?



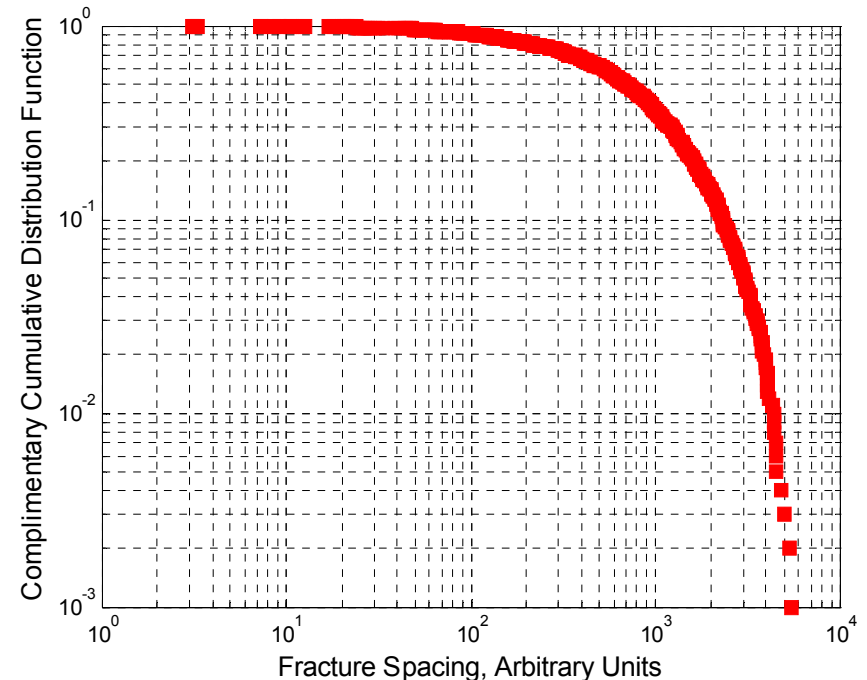
Random Breakage

Random locations of breaks (Poisson process) leads to exponential distribution of spacing between fractures

Histogram

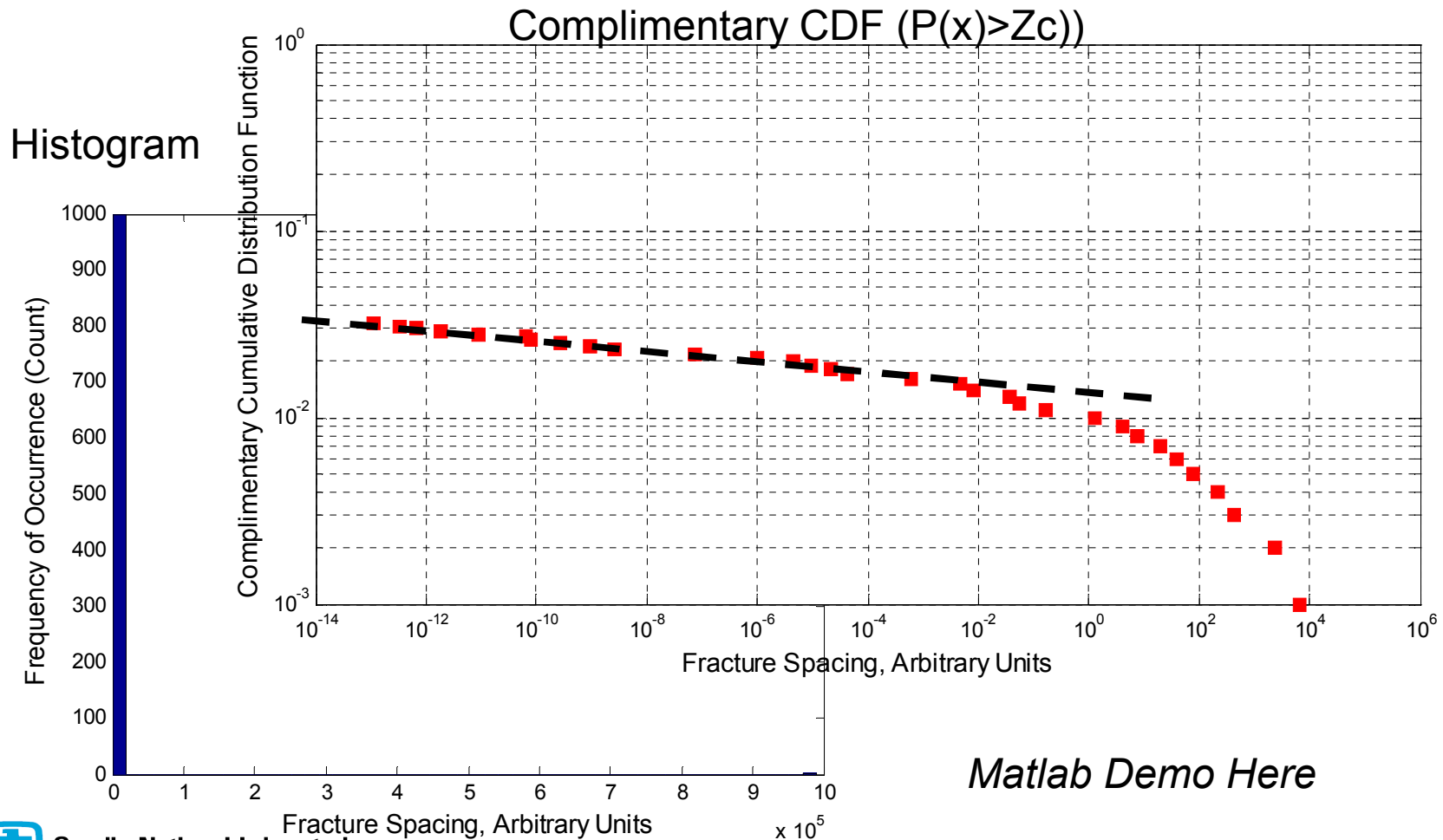


Complimentary CDF ($P(x) > Z_c$)



Non-Random Breakage

Preferential breaking of smallest piece leads to a power-law distribution of fracture spacing. Power-law distribution has a straight line in log-log space



Power-Law Relationship

$$Y = \alpha X^\beta$$

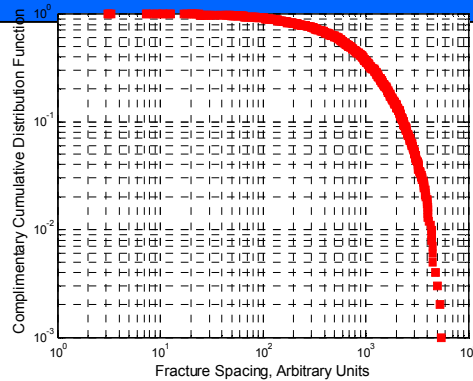
$$\log_{10} Y = \beta_0 + \beta_1 \log_{10} X$$

$$\alpha = 10^{\beta_0} \quad \beta = \beta_1$$

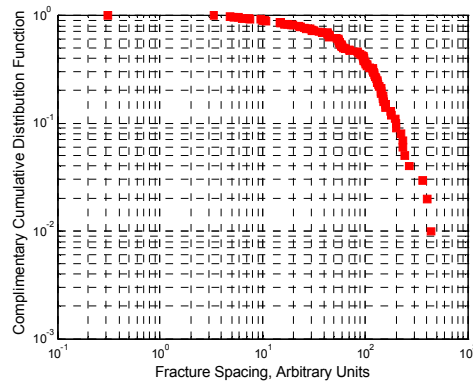
Slope, in log-log space, is the fractal dimension ($\beta = D$)



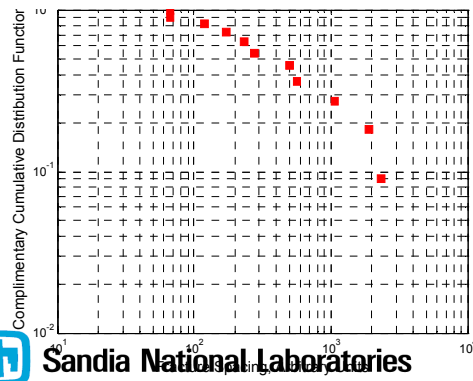
Limited Sampling: Spacing



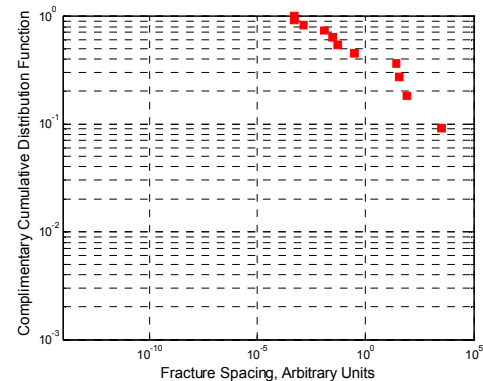
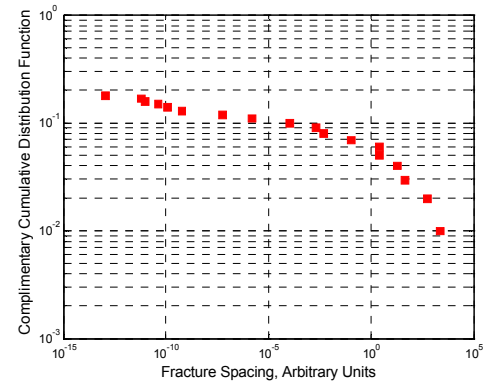
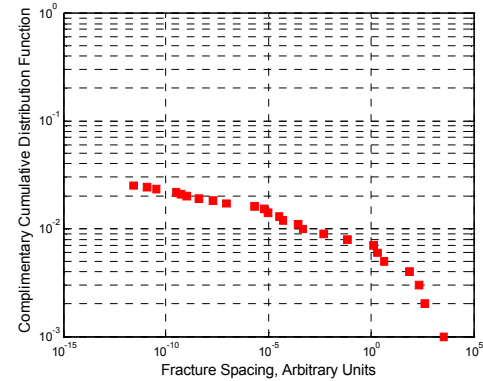
$N = 1000$



$N = 100$



$N = 10$



Fracture Spacing

- We just covered several things
 - Measured fracture spacing (scanline, borehole, etc) will be exponential if fractures are randomly located
 - Spacing will have a power-law distribution if fractures preferentially break the smallest intact piece
 - Inferring statistical distributions from limited data is a risky approach

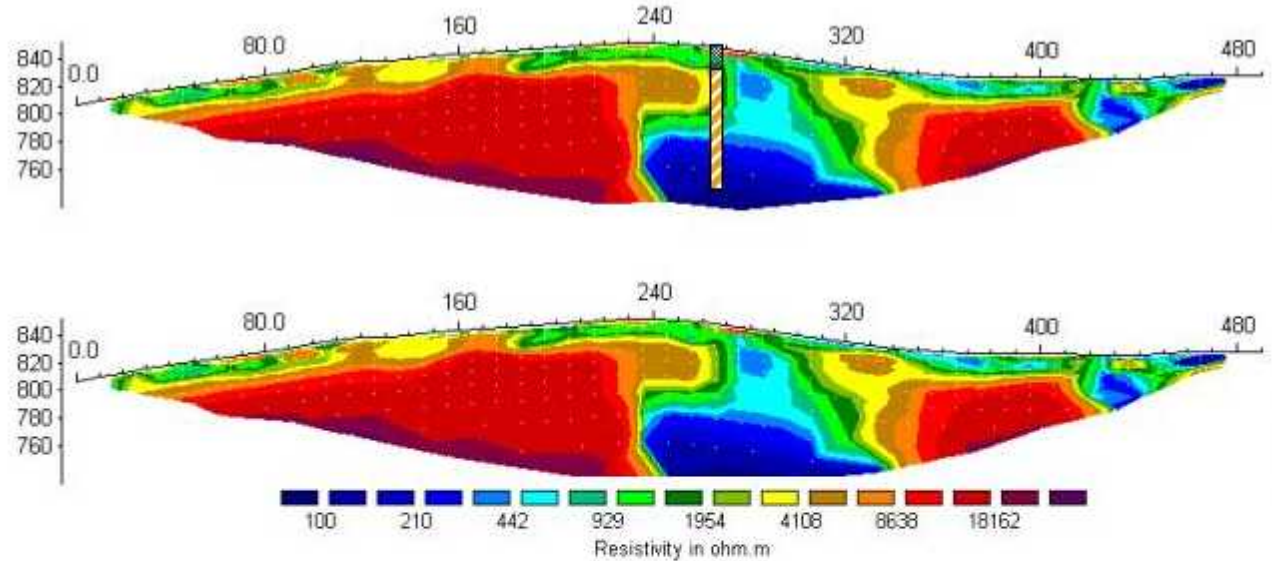


Intensity in 2-D

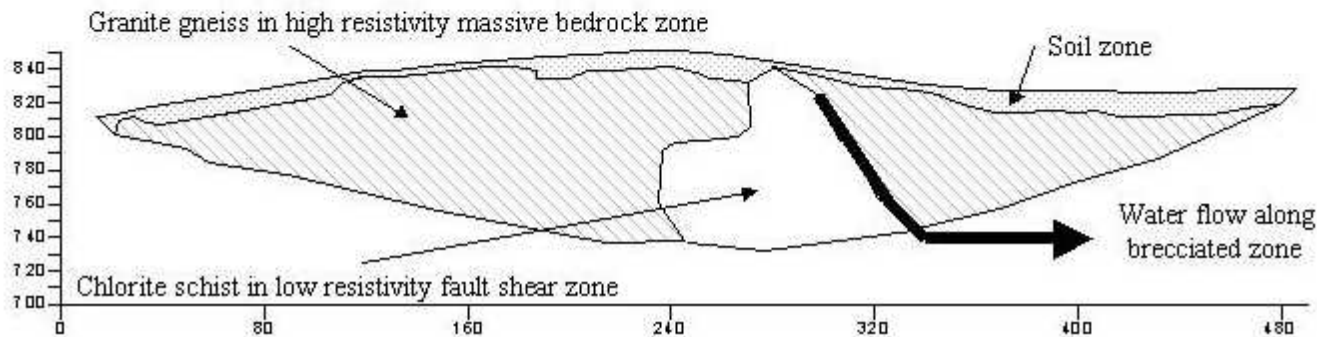
- Homogeneous Poisson Process (HPP) and Non-Homogeneous Poisson Process (NHPP) as models for fracture locations



Intensity Data: Resistivity Profiling



From: Thomas Burbey, Virginia Tech University



Two-dimensional surface resistivity profiles collected using a variety of array techniques combined with borehole geophysical logs revealed anomalous low resistivity areas in crystalline bedrock associated with fault zones.



Length

- Measuring fracture length
 - Have to derive from outcrop data
 - Has anyone here ever seen both ends of a significant fracture?



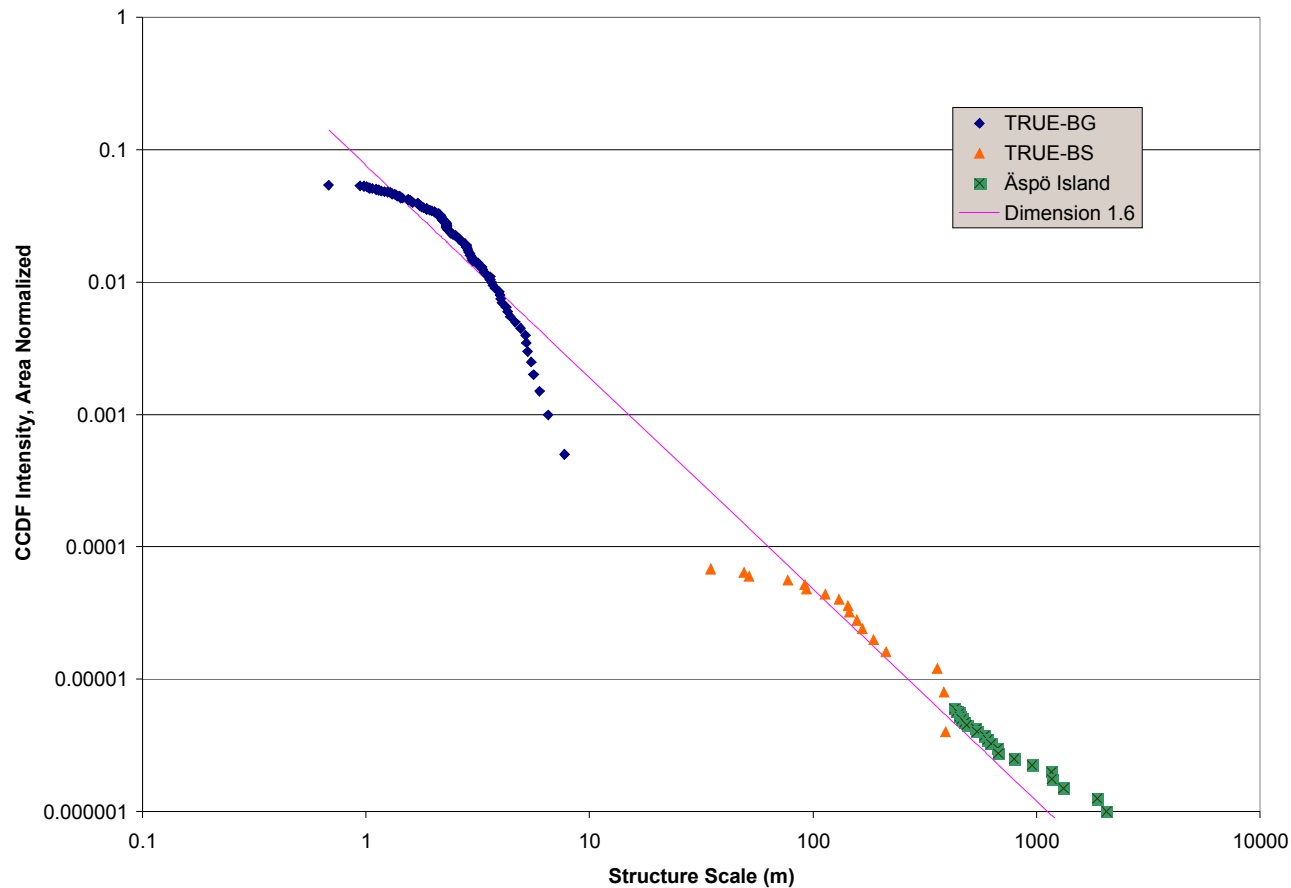
Length Distributions

- Exponential
 - Uniformly random growth of all fractures
- Power-Law
 - Preferential growth of long fractures (growth is proportional to current length)
- Log-Normal
 - Products of uniform random numbers produce log-normal distributions



Length

- Simulated Aspo feature length distribution



Aspo site characterization:

Borehole-Cannister Scale
(0.1-5m)

Block Scale (200m)

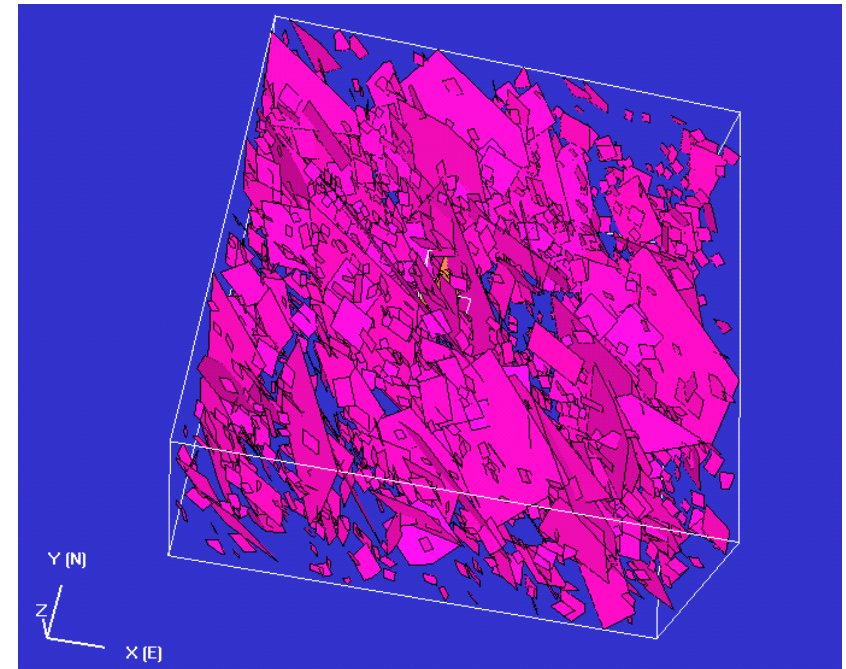
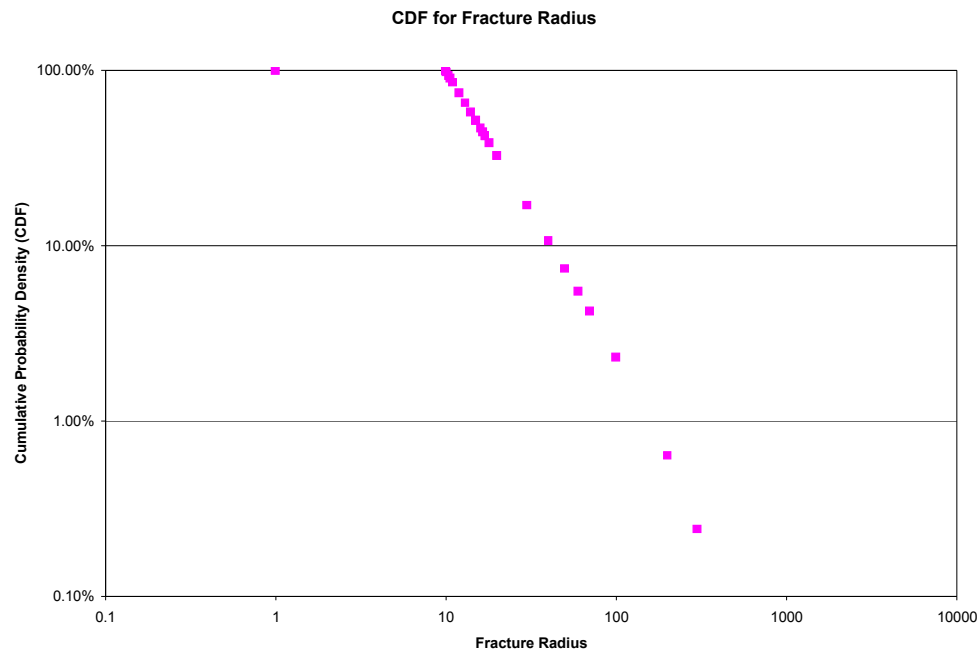
Site Scale (2000m)

Power-Law?

Significant effect of
censored measurements

Length/Size

Discrete feature simulation of rectangular objects with a power-law size distribution



Aperture and Transmissivity

- Cubic Law
 - Flow \propto aperture³

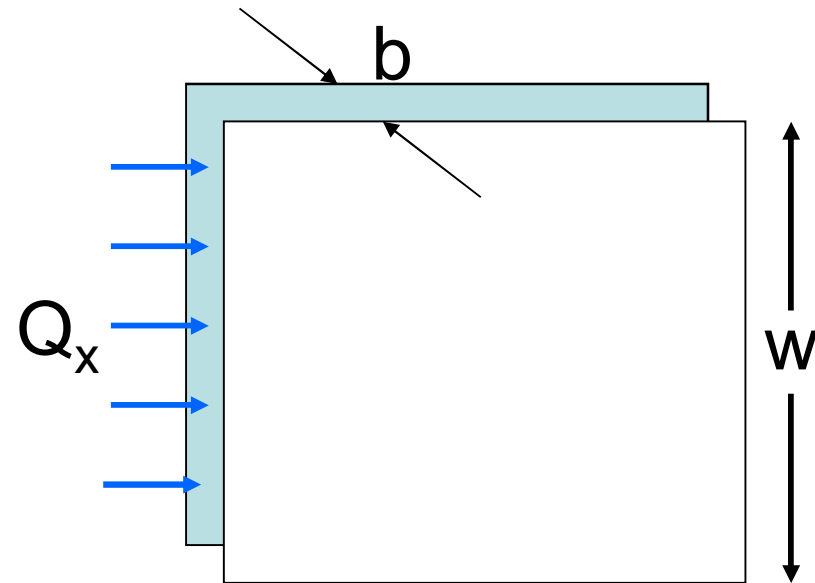
$$Q = \frac{\rho_w g b^2}{12\mu} (bw) \frac{\partial h}{\partial l}$$

$$K = \frac{\rho_w g b^2}{12\mu}$$

- Snow's Equation (multiple fractures)

$$K = \frac{\rho_w g N b^3}{12\mu}$$

$$k = \frac{N b^3}{12}$$



Fracture Measurements

- *“However, in most subsurface cases, there will be insufficient fractures having the size of interest (i.e., large conductive fractures) to derive a statistically significant estimate of spacing”.*

Ortega et al., 2006, AAPG Bulletin



Stochastic Simulation

- Observational limits force us to use a stochastic approach to fracture modeling to capture significant uncertainty

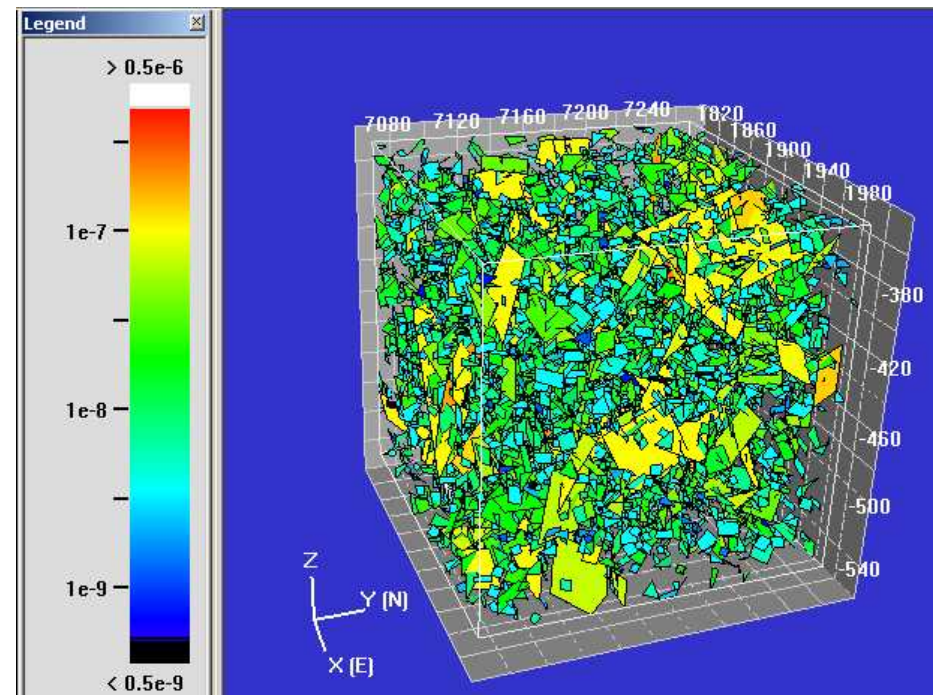
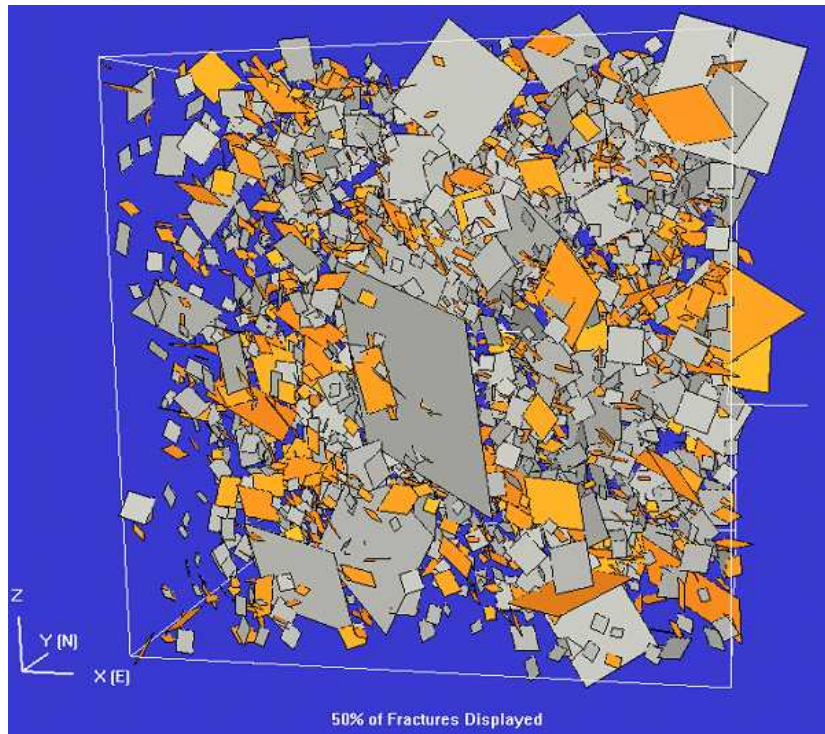


Discrete Fracture Models

- Statistically-based approach to simulating objects (fractures) that represent observational data base
- Typically objects have a very large (length to aperture) aspect ratio



DFM Examples: Aspo



Dershowitz et al., 2002

Background fractures coloured by transmissivity (log scale) in 200m cube



DFM: Drawbacks

- DFM's are completely observationally-based (Statistical models)
 - Do not account for genesis of fractures.
 - DFM can work well if observational database is complete
 - Impossible to determine degree of completeness
 - Observations of length and shape are censored
 - My Experience: Difficult to use in stochastic flow simulation



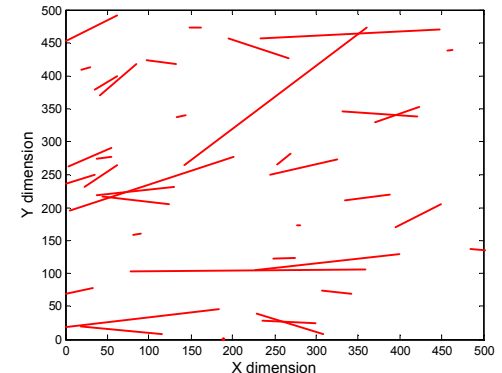
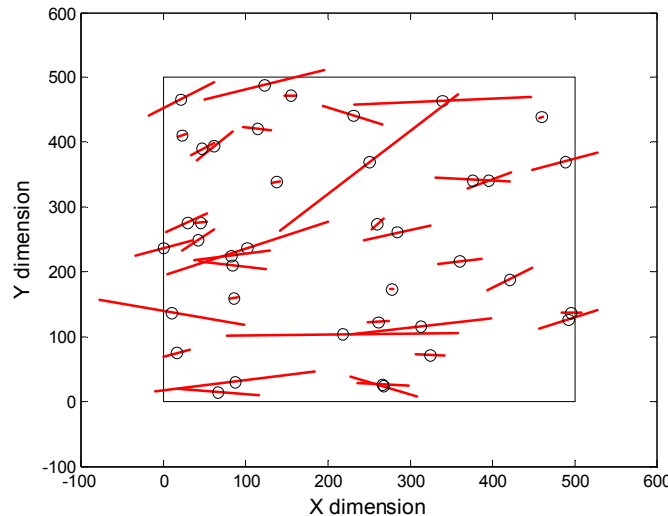
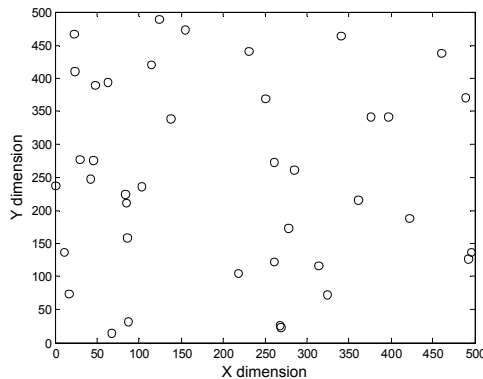
Cell (Pixel)-Based

- Similar to DFN's in that stochastic simulation is used to place objects (fractures)
- Main difference is that fractures are placed on a grid
 - Fracture simulation will not be mesh independent (minimum support is mesh size)
 - May be easier to incorporate rule-based growth and termination (FracNet)
 - Amenable to dual-permeability and stochastic simulation approaches



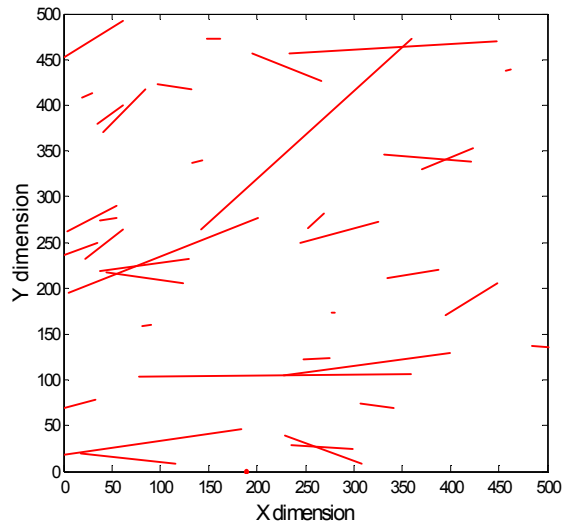
Cell Based: Simulation Example

40 Fracture centers located randomly in domain (left). Fracture lengths and orientations drawn from exponential (80) and normal (10,20) distributions (center). Fractures then trimmed to domain boundaries (right)



Fracture Network

Connectedness of fracture network dictates the amount of flow across the domain. Image on left shows locations of 40 individual fractures. Image on right shows 35 distinct clusters of connected fractures.

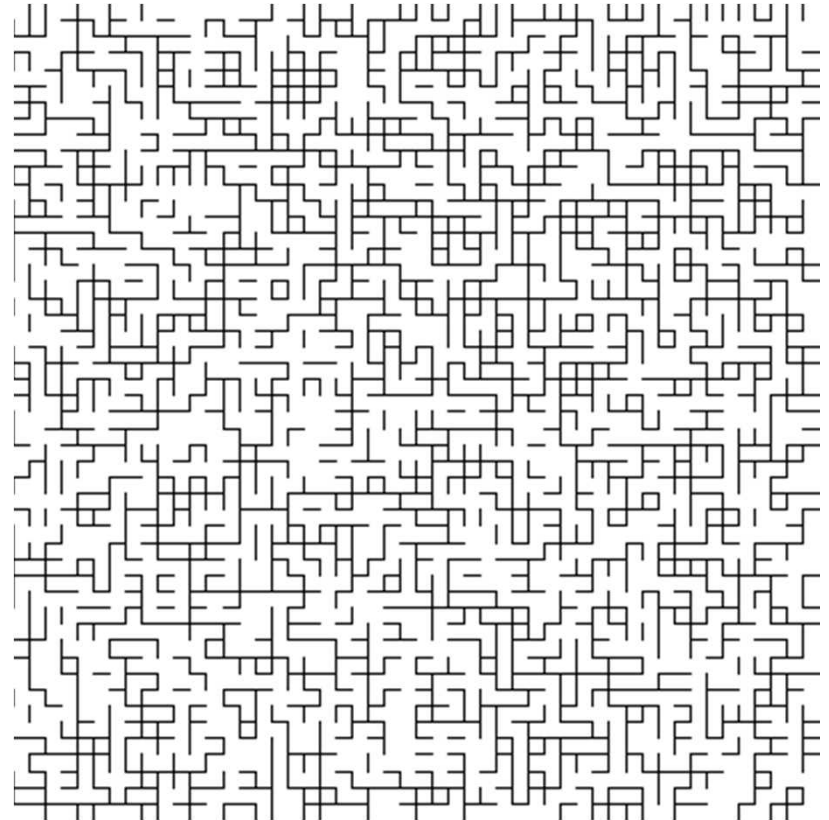


Percolation: Background

Development in statistical physics

Bond percolation on a square lattice where $P(\text{connect})$ is 0.51 for any given edge

Percolation threshold is 0.50 for 2-D and approximately 0.249 For 3-D (square lattice with $z = 4$ and 6, respectively)

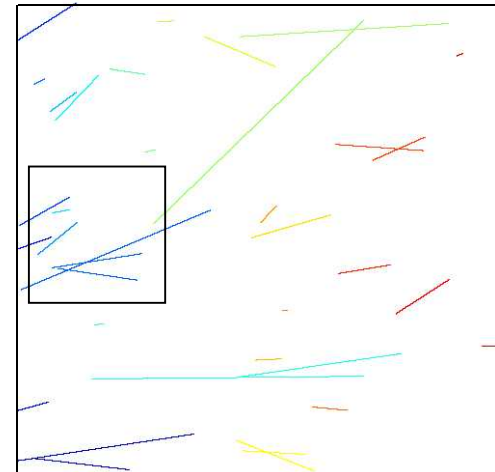


See: Sahimi, M., 1995, *Flow and Transport in Porous Media and Fractured Rock*, VCH, New York, 482 pp.



Percolation

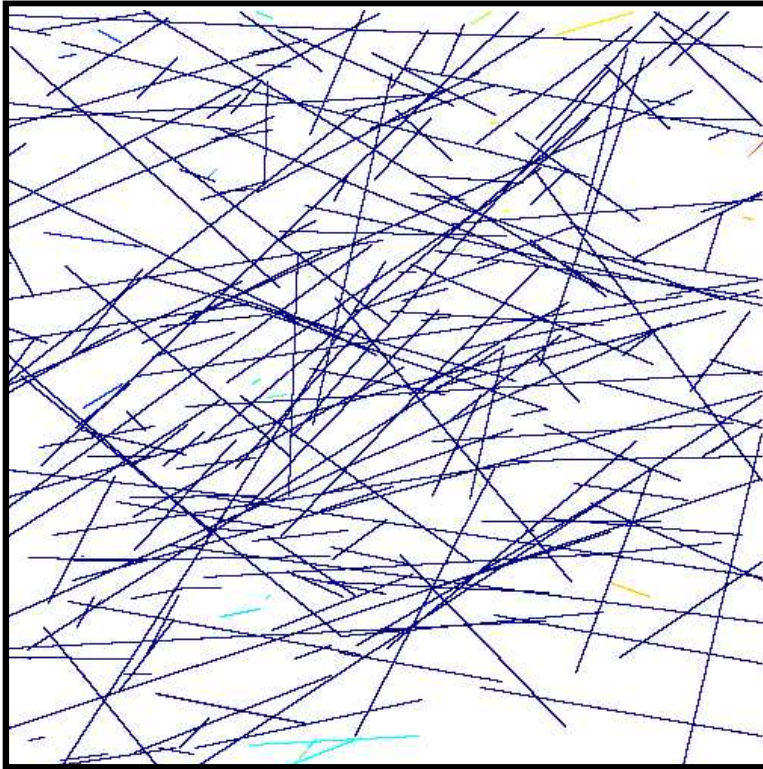
- Critical point at which the fracture network goes from impermeable to permeable
 - Network becomes connected across a volume
- Percolation only has definition within a *specified area/volume*



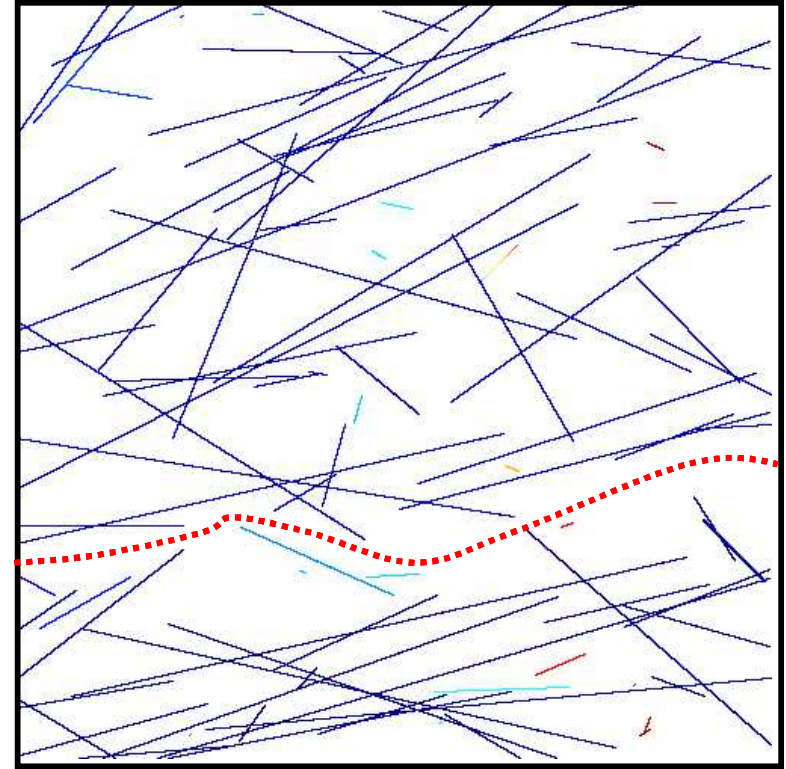
Percolation and Fractures

Fractures drawn from same length and orientation distributions

240 Fractures

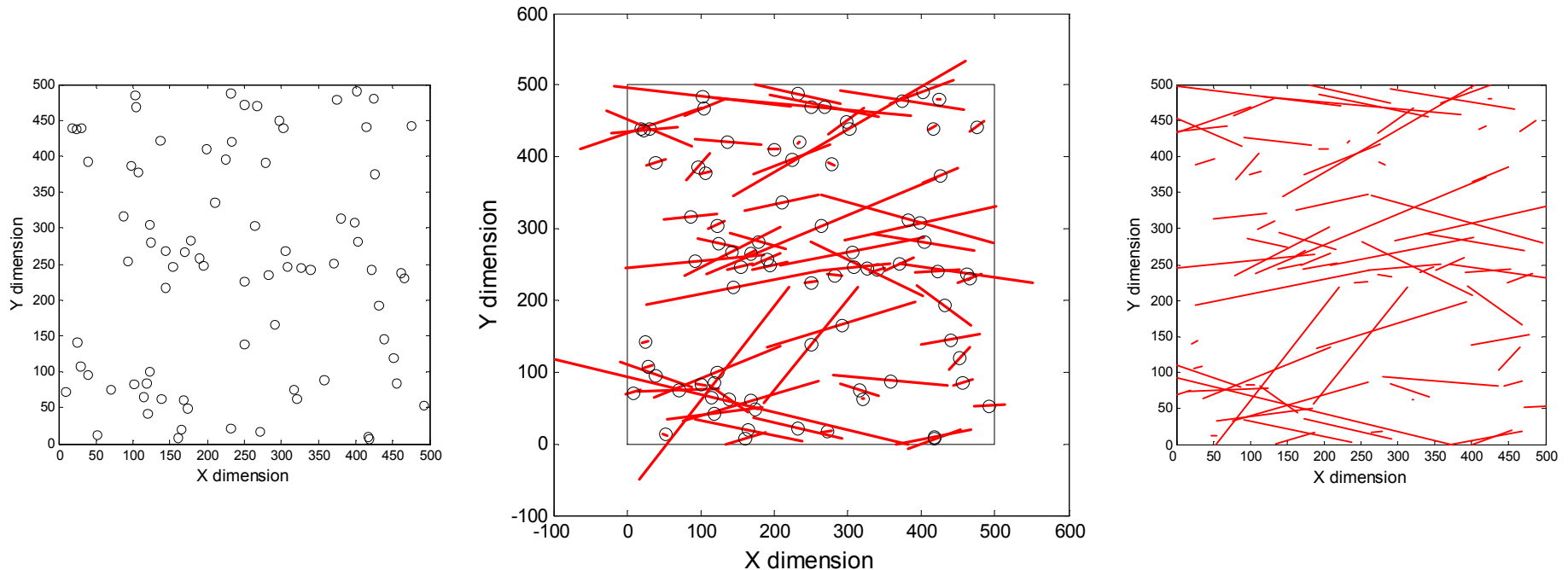


100 Fractures

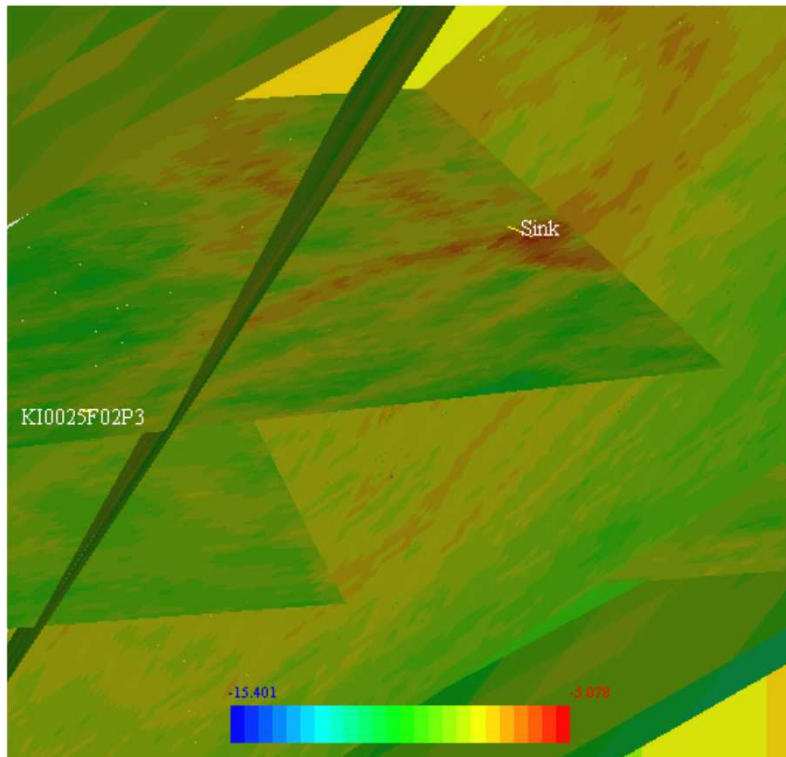


Cell Based: Simulation Example

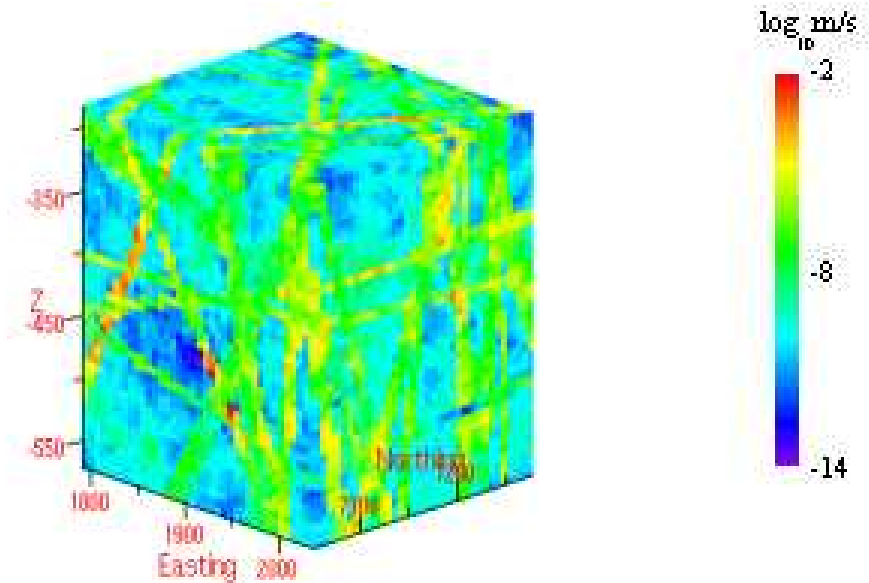
80 Fracture centers located randomly in domain (left). Fracture lengths and orientations drawn from exponential (120) and normal (10,20) distributions (center). Fractures then trimmed to domain boundaries (right)



Cell-Based Model Examples



Stochastic transmissivity field on
Structure 20 (Holton, 2001)

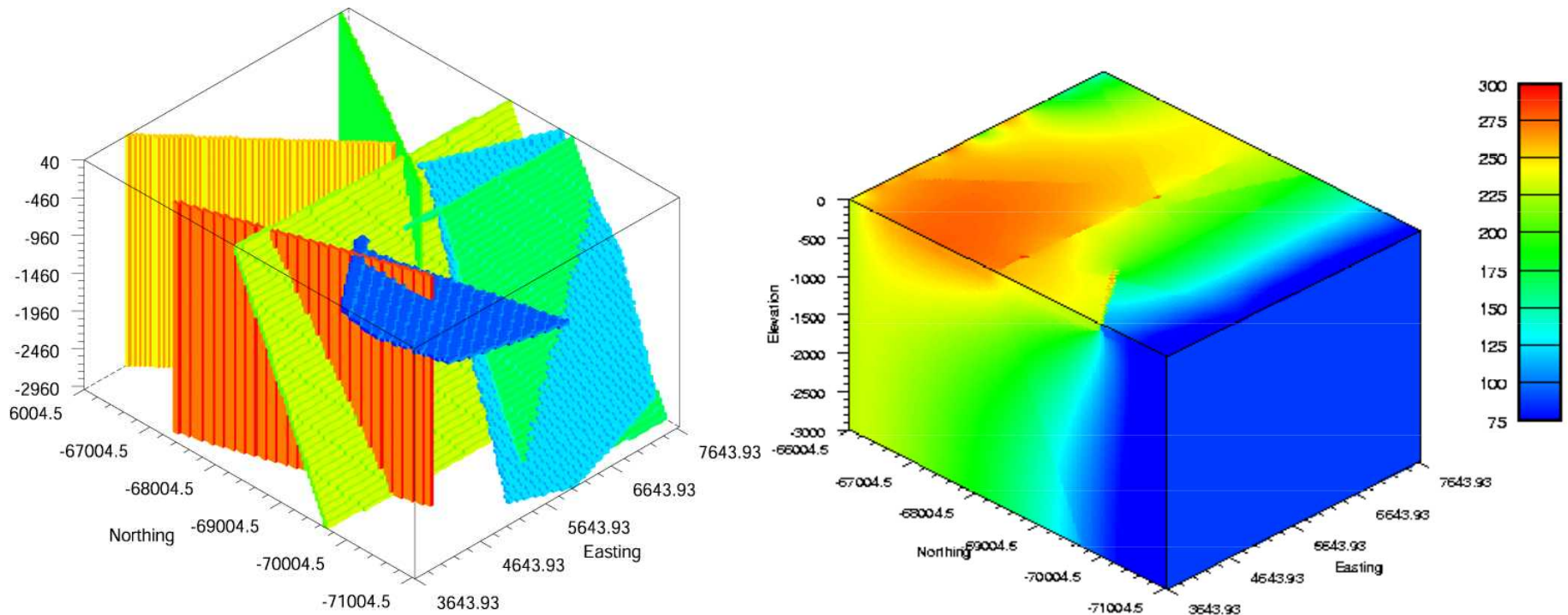


*Stochastic field of hydraulic conductivity
on structures (Gómez-Hernández et al., in
prep.)*



Cell Based: JNC Shobasama Site

Model of seven existing (confirmed faults)

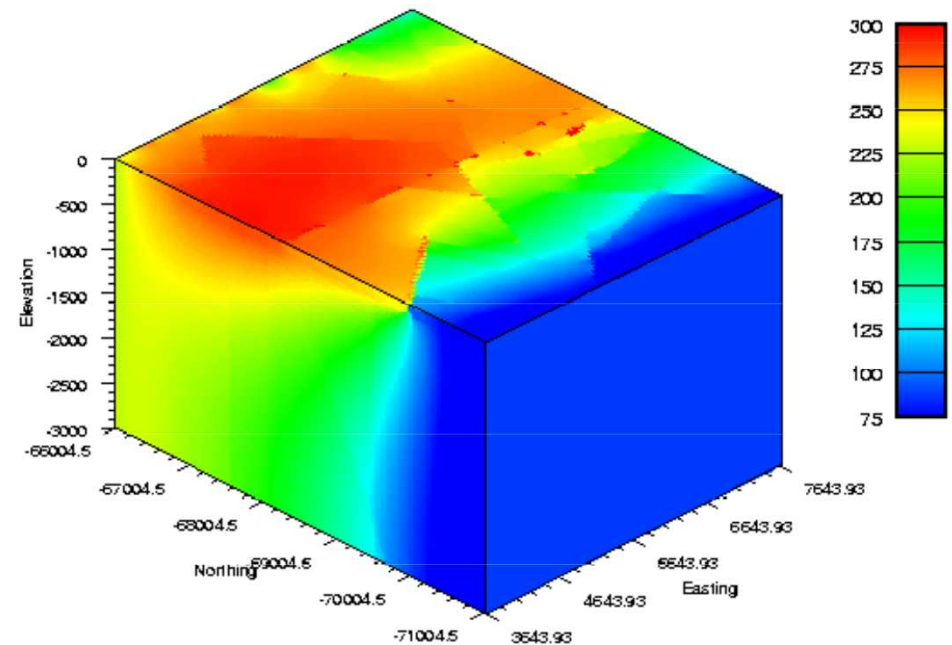
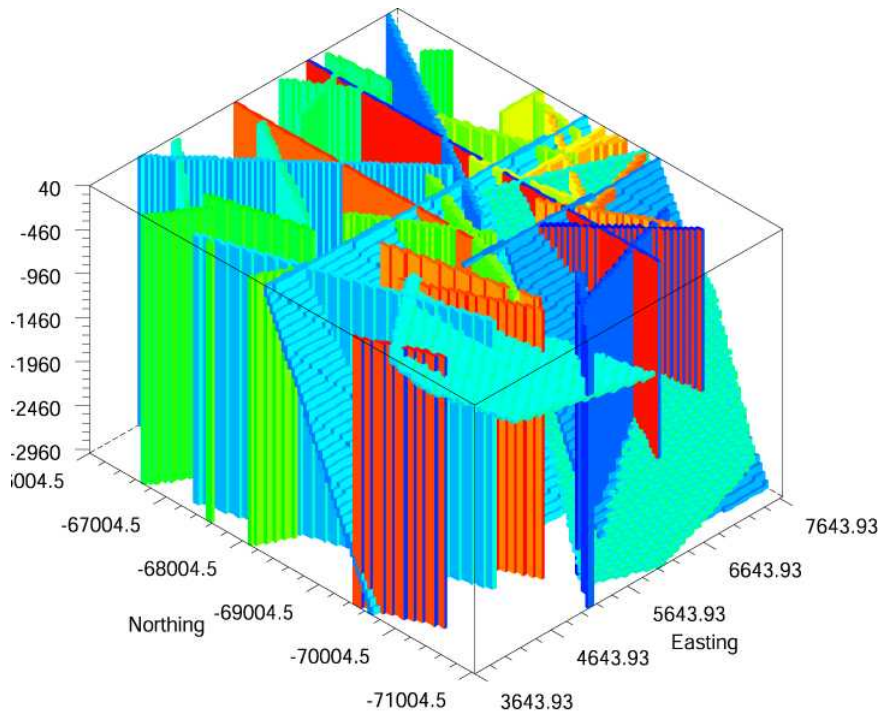


937,500 cells, 40m cube gridblocks



Cell Based: JNC Shobosama Site

Model of existing (confirmed faults) + unconfirmed faults from lineament analysis (23 faults total)



McKenna, S.A., M. Eliassi, K. Inaba and H. Saegusa, 2001, Steady-state groundwater flow modeling of the MIU site area, Groundwater Flow in Discrete Fractures Symposium, Japanese Geotechnical Society, Tokyo, September 10-11, 14 pp.

Fractured Continuum Model

- Problem Statement:
 - How to honor observations made on various discrete fractures in continuum models of fracture permeability?
- What we care about:
 - Flow characteristics of the fracture network
- What we measure:
 - Characteristics of individual fractures

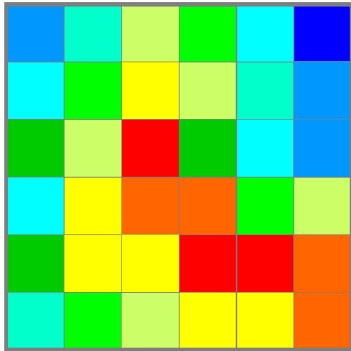
McKenna, S.A. and P.C. Reeves, 2006, Chapter 14: Fractured Continuum Approach to Stochastic Permeability Modeling, in: Coburn, Yarus, and Chambers, eds., *Stochastic Modeling and Geostatistics: Principles, Methods, and Case Studies, Volume II: AAPG Computer Applications in Geology 5*, p. 173–186.



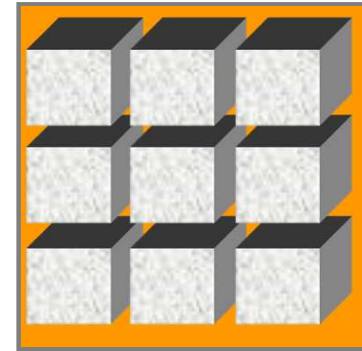
MODELING APPROACHES

Dual Porosity / Permeability Systems

Equivalent Continuum Model

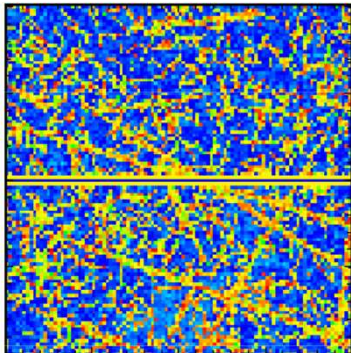


Dual Permeability (K) Model

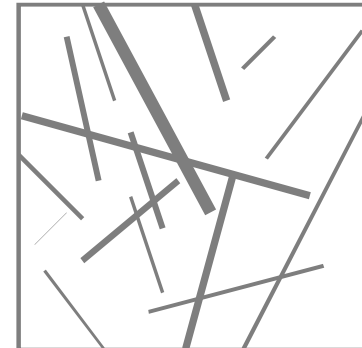


No Fractures

Fractured Continuum Model



Discrete Fracture Network



No Matrix

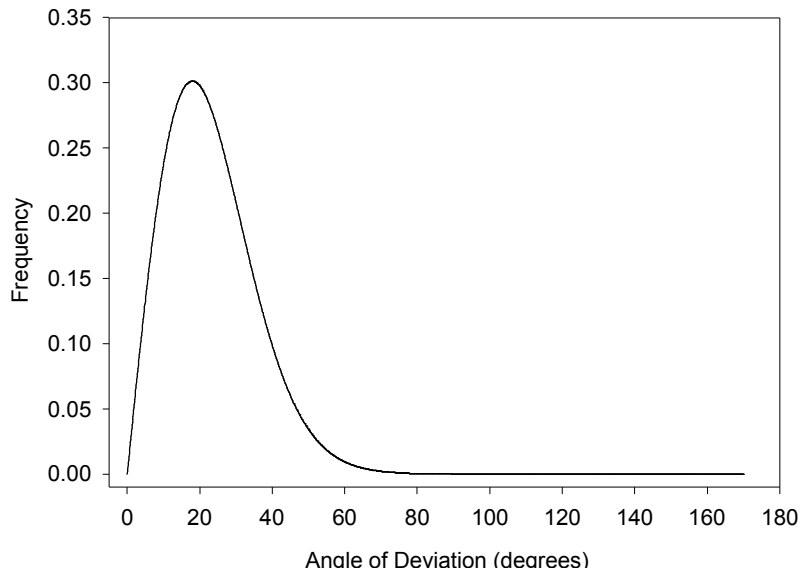
Honoring Discrete Fracture Observations

- At the gridblock scale, FCM is an effective permeability value derived from knowledge of discrete fracture network
- For this study, observations were made on discrete fractures to characterize:
 - Radius: Truncated Power-Law
 - Frequency: Poisson
 - Orientation: Fisher (approximated by Triangular)
 - Transmissivity: Log-normal
 - Aperture: Deterministic relation with Transmissivity

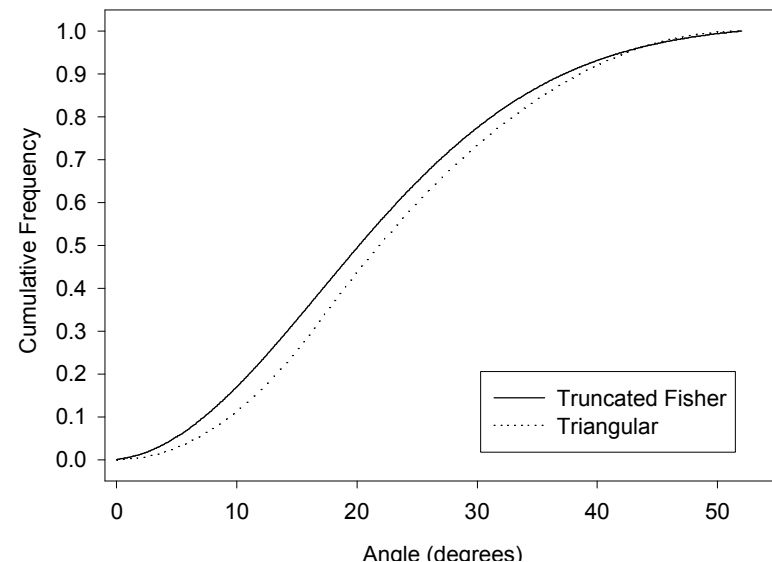


Feature Orientation

Fisher distribution of fracture orientation deviations about the mean orientation

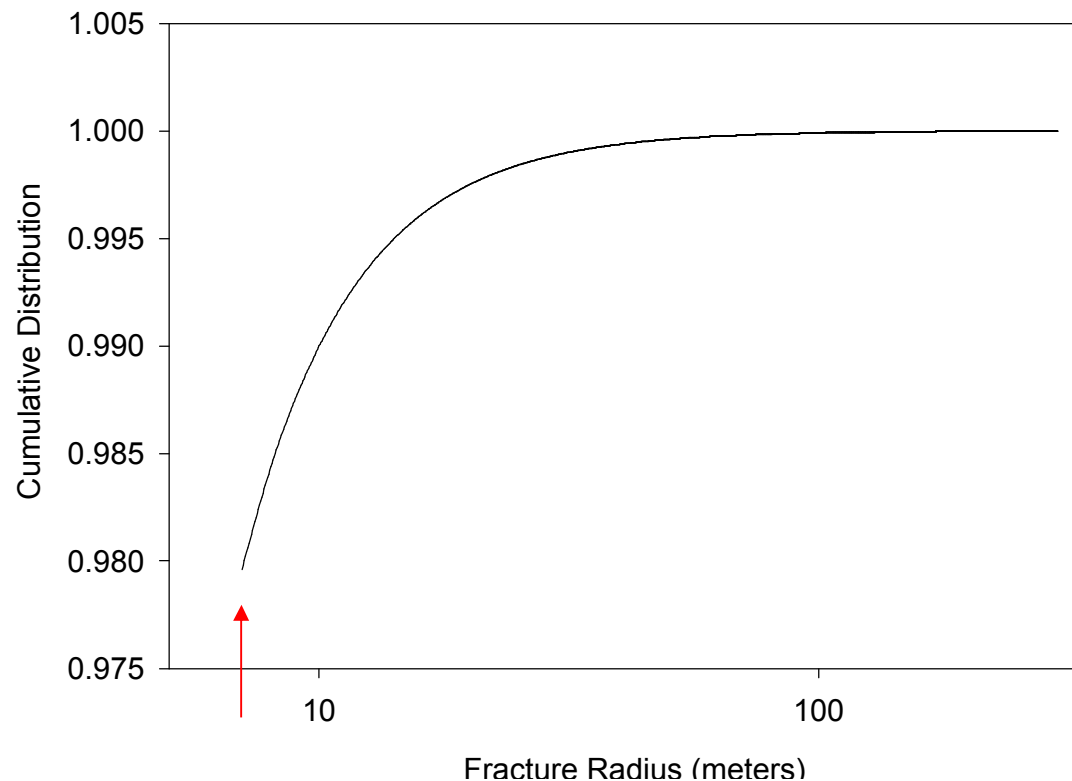


Fisher distribution approximated by triangular distribution



Fracture Radius

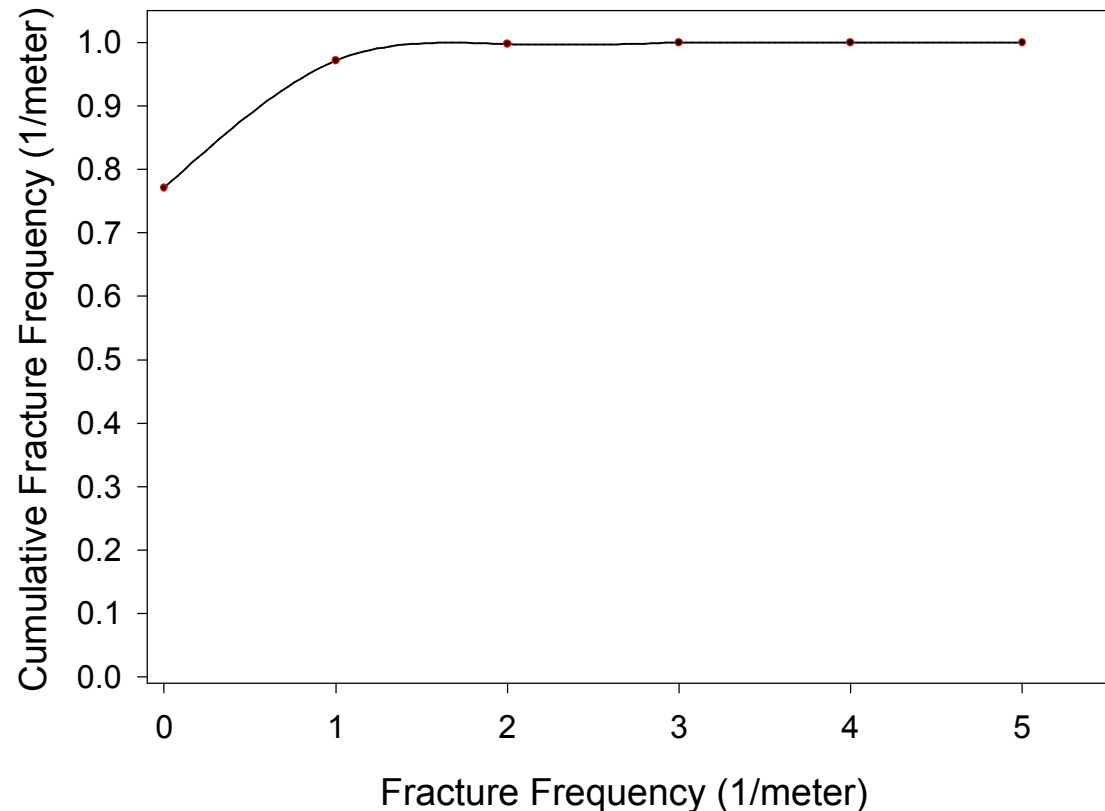
Fracture radius specified with a truncated power-law
(minimum length = 7.0 meters)



Fracture Frequency

Poisson distribution of fracture frequency (1/m)

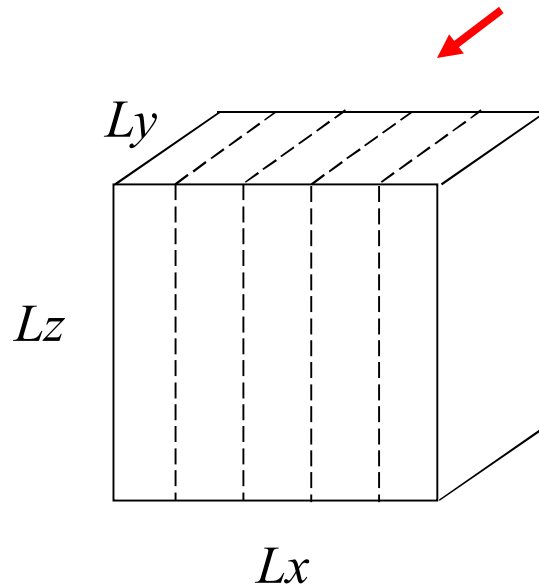
Poisson distribution modified to define frequency per grid-block. Note that Poisson distribution is only defined for integer values.



Effective Gridblock Permeability

Local Model

Define gridblock permeability in terms of geometric information on fracture network, gridblock dimensions and effective fracture conductance



$$C_i = \frac{gb^3 \rho_w}{12\mu L}$$

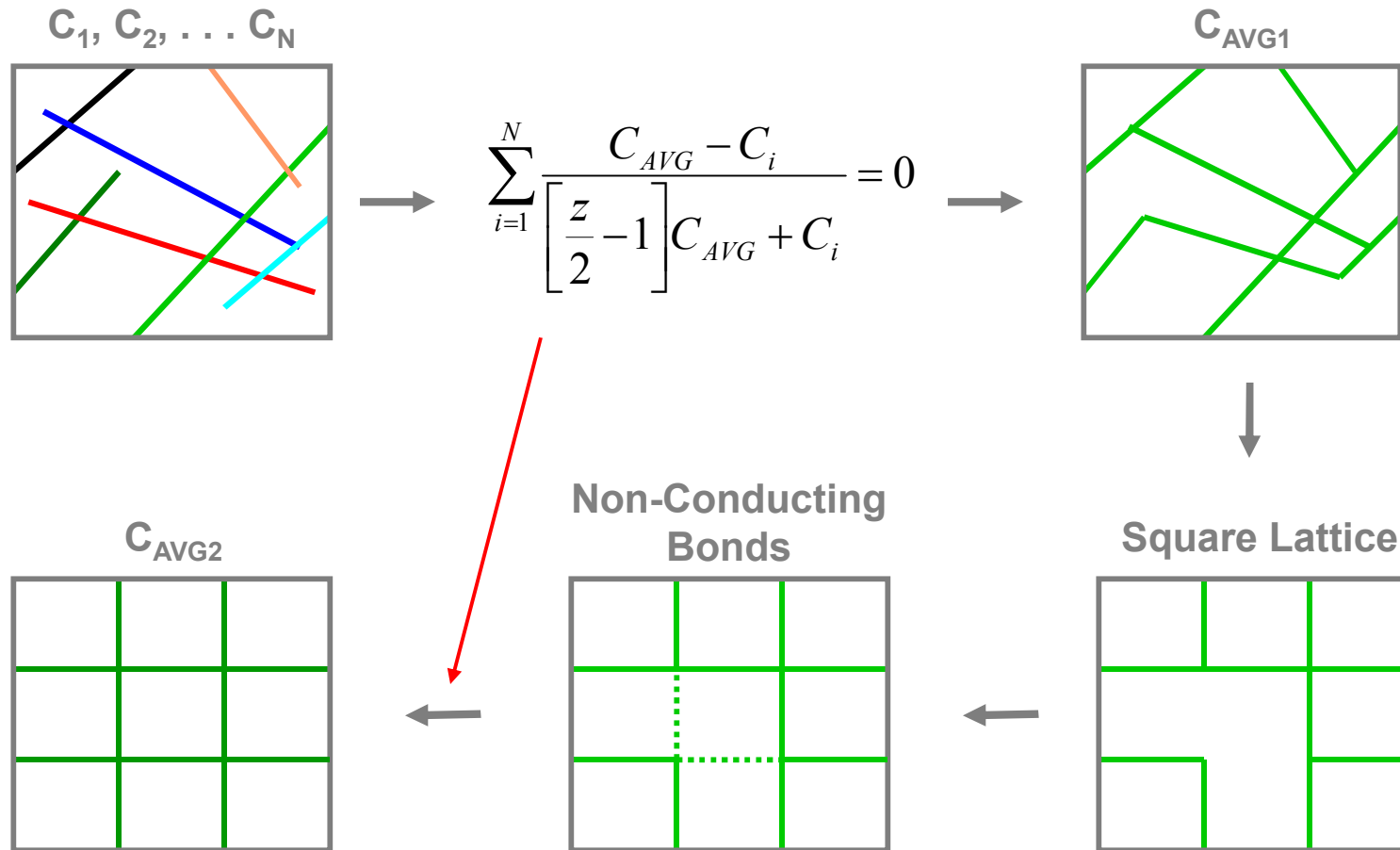
$$Q_y = N_x C^* \Delta H$$

$$Q_y = K_y Lz Lx \Delta H / Ly$$

$$K_y = \frac{Ly N_x C^*}{Lx Lz}$$

Effective Medium Approximation

Local Model



Spatial Simulation

Domain Model

- FCM uses combination of geostatistical and object-based simulation to populate flow model domain
- Components of Effective Medium Approximation (frequency and coordination number) are considered to be realizations of a spatial random function and are modeled with geostatistical simulation.
- Proportion of “conductive/non-conductive” cells, based on Poisson distribution, modeled with object-based simulation
- Three simulations combined to produce final permeability model



GEOSTATISTICAL MODEL

“Fractured Continuum”

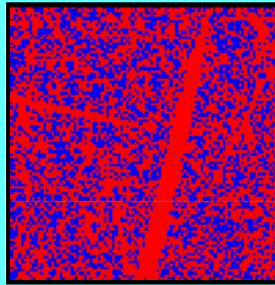
Fractured Continuum Model



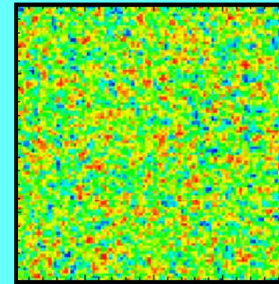
Fracture Statistics

- *Shape*
- *Radius*
- *Orientation*
- *Transmissivity*
- *Aperture*
- *Spatial Frequency*

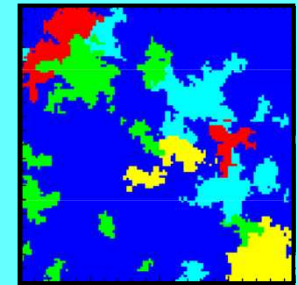
“Conductive Domains”
Boolean



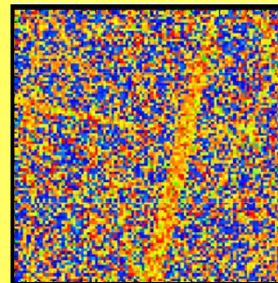
Coordination Number
Multigaussian



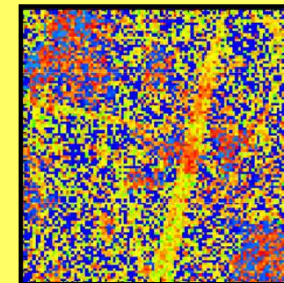
Number of Fractures



Permeability



Porosity



FRACTURED CONTINUUM MODEL

Dual Porosity / Permeability Systems

Advantages

- Matrix is Not Ignored → *Unlike DFNs*
- Fractures are Not Abstracted to a Network of 1-D Pipes → *Unlike DFNs*
- Spatial Correlation Reflects Underlying Spatial Structure of Fractures → *Unlike ECMs / DFNs*
- Statistics Underlying Fracture Geometry Are Directly Utilized → *Unlike ECMs / DFNs*
- Influence of Fracture Geometry on Flow and Transport Can be Explicitly Studied → *Unlike ECMs / DFNs*



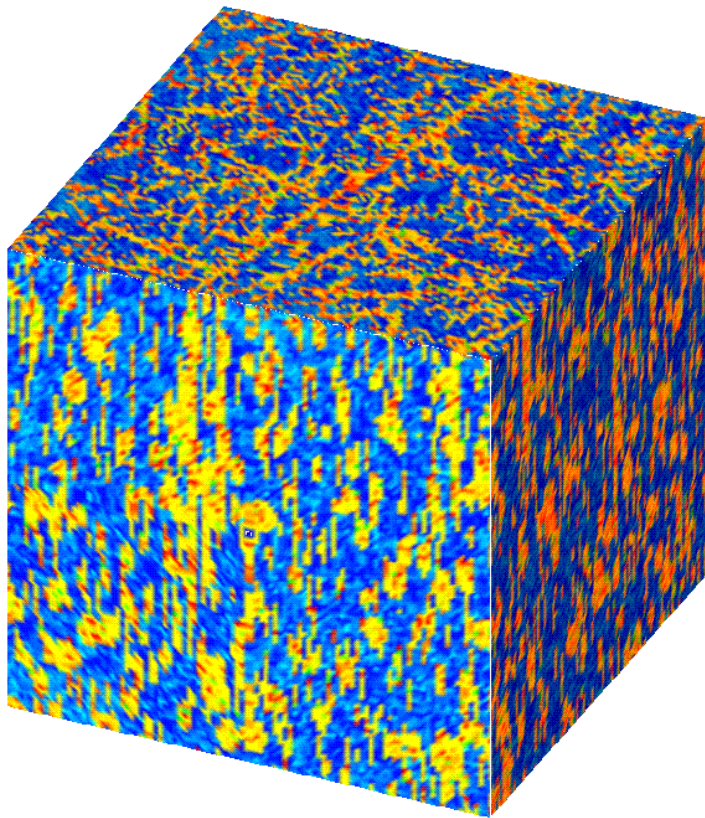
EXAMPLE REALIZATION

Stage 1 R44/50

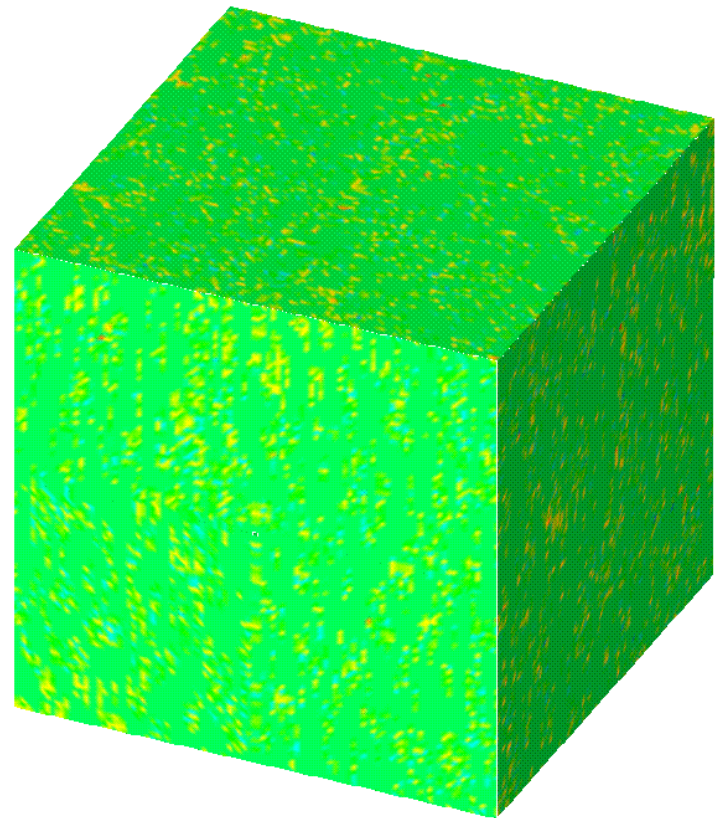
POR-SALSA: H12 FLOW COMPARISON

Intrinsic Permeability

Porosity



$\text{Log}_{10} (k) [\text{meters}^2]$

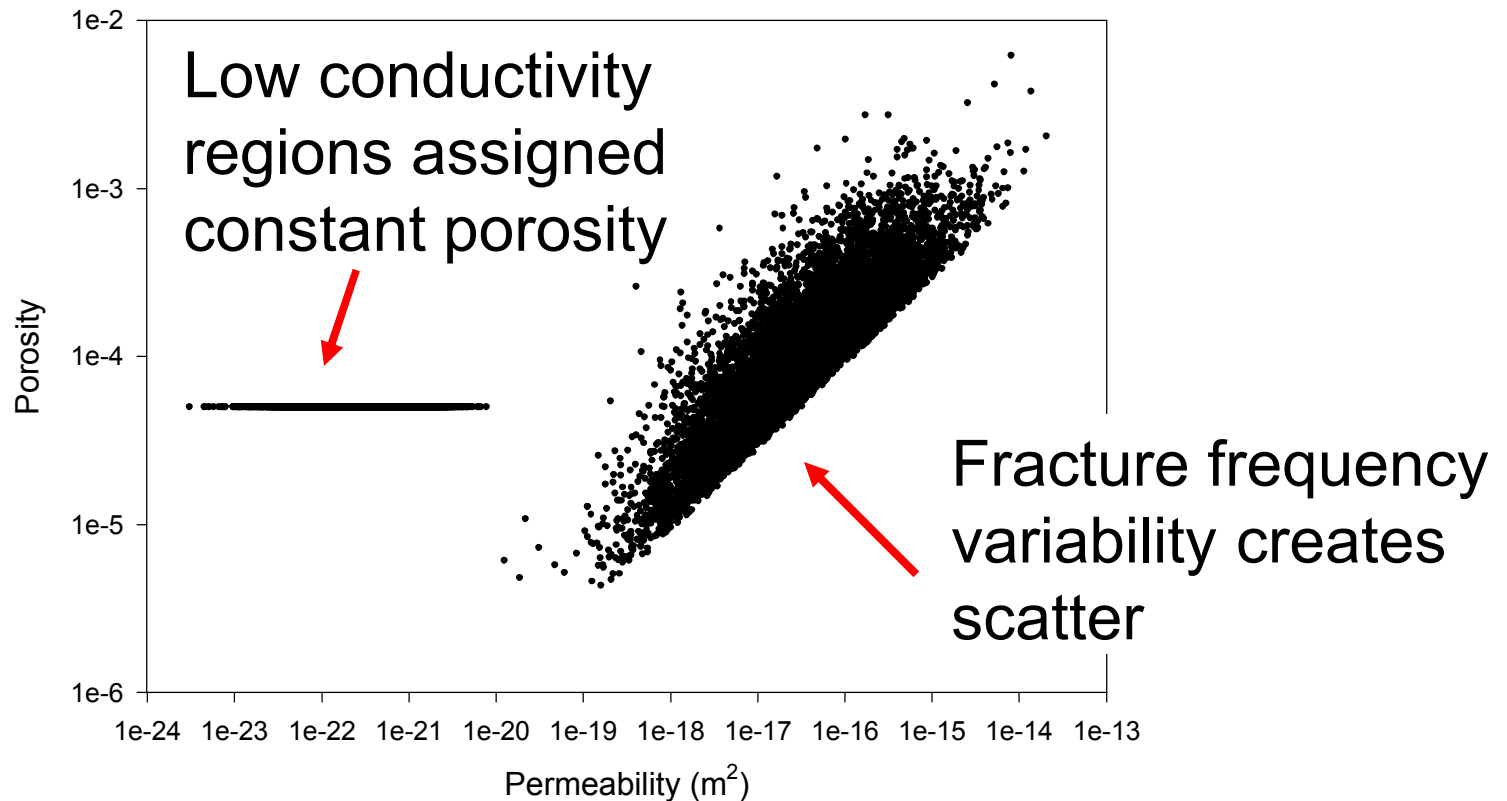


$\text{Log}_{10} (\phi)$



Permeability-Porosity Relationship

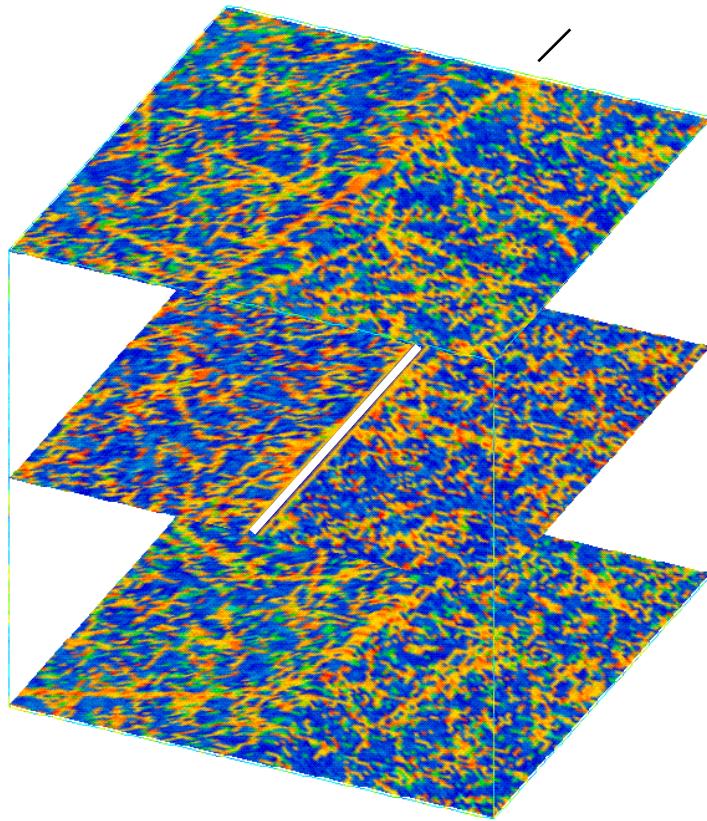
Stage 1 result, single realization



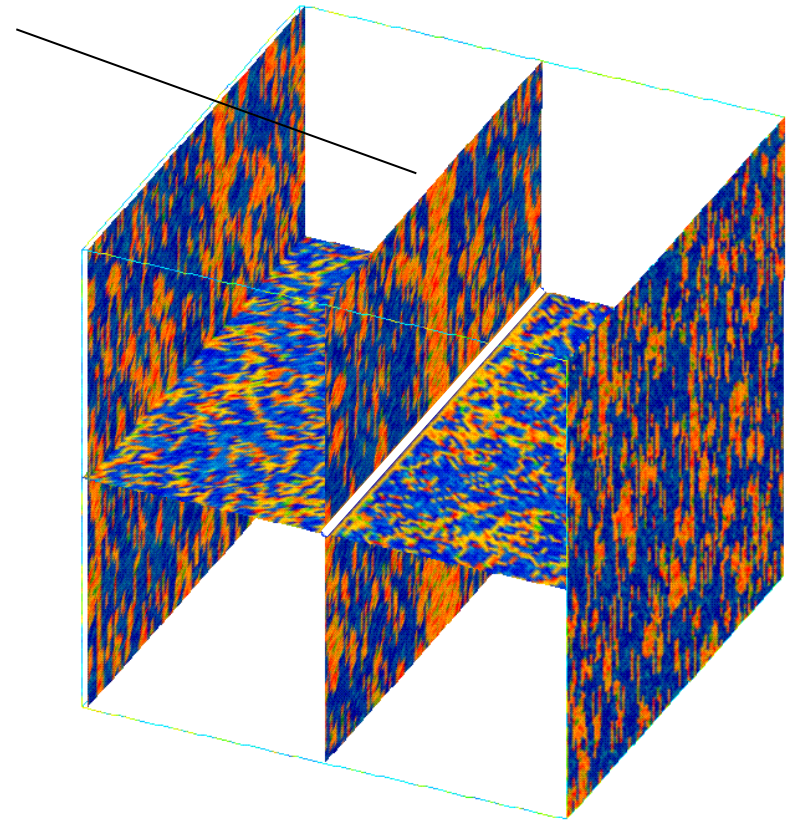
EXAMPLE REALIZATION

Stage 1 R44/50

POR-SALSA: H12 FLOW COMPARISON
Intrinsic Permeability Transects



$\text{Log}_{10}(k)$ [meters²]

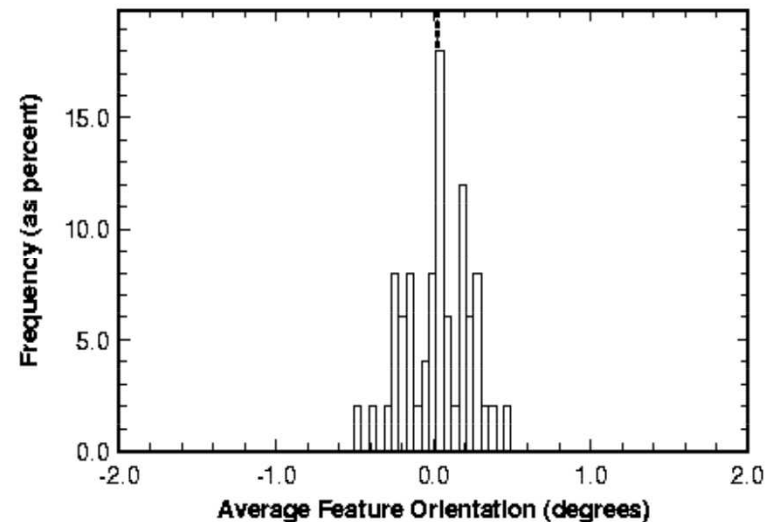
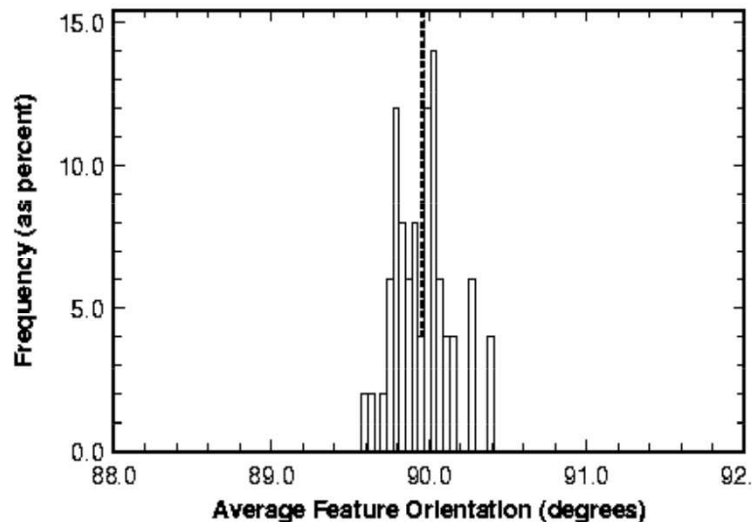


$\text{Log}_{10}(k)$ [meters²]



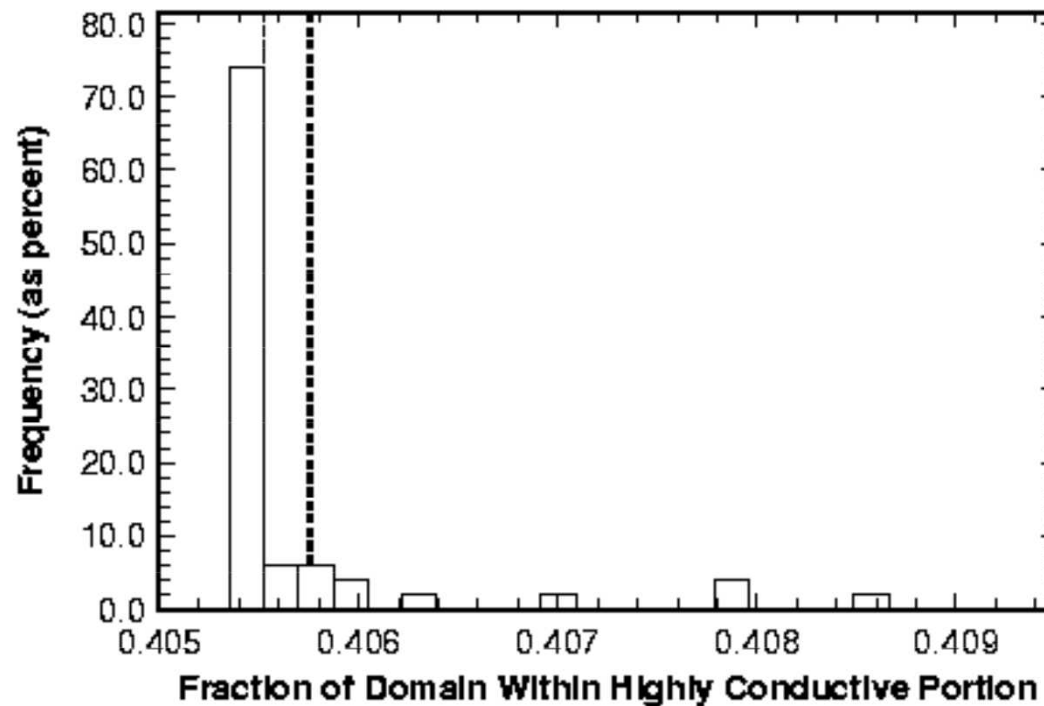
Model Validation (orientation)

Mean orientation of both feature sets calculated for 50 realizations and compared to target values of 0.0 and 90.0 degrees



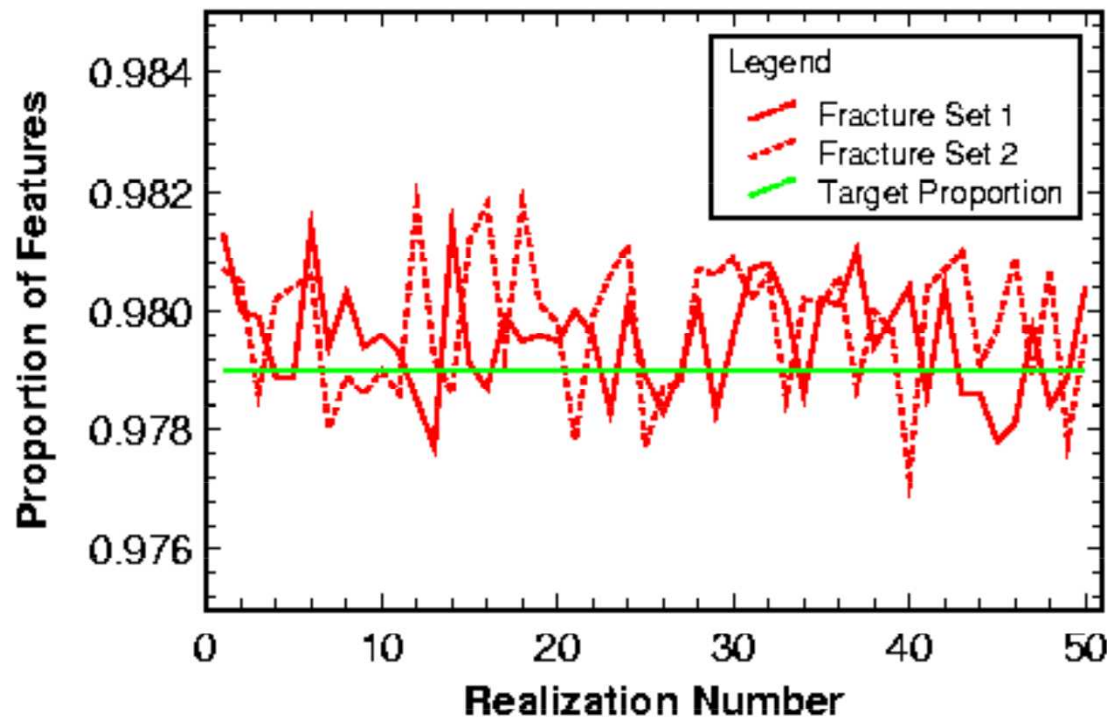
Model Validation (Frequency Minimum)

Median proportion of conductive gridblocks across 50 realizations equals target proportion of 0.4055



Model Validation (Minimum Fracture Radius)

97.9 percent of all features have radius greater than 7 meters

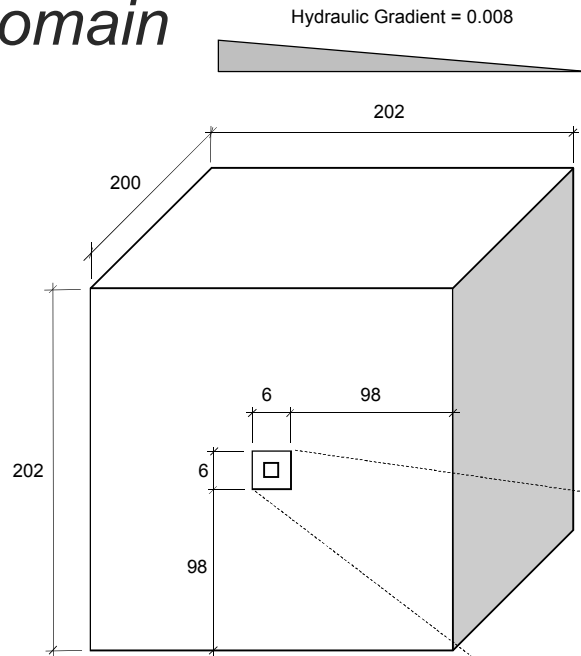


Proportion of features with radii greater than 7 meters is shown against the target proportion for both feature sets

Flow Modeling (MESH DESCRIPTION)

POR-SALSA: H12 FLOW COMPARISON

Domain



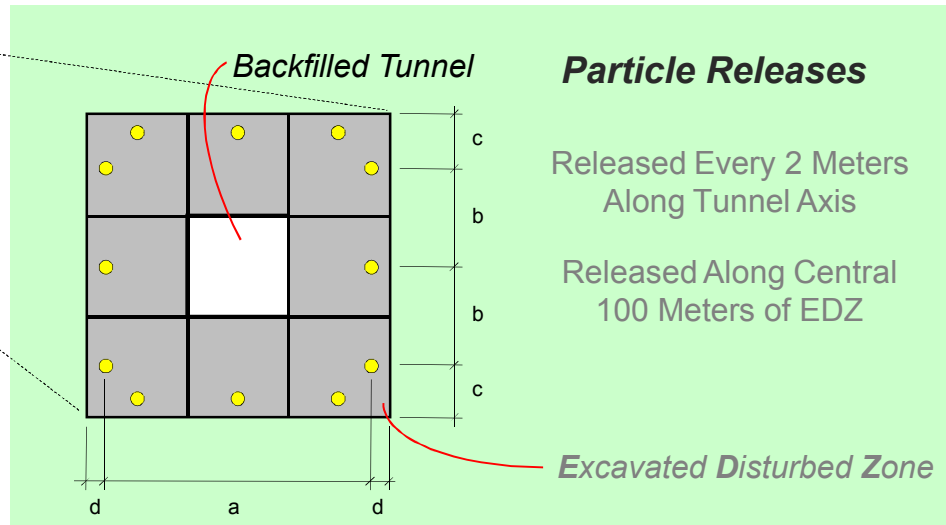
Dimensions (meters)

$a = 5.50$ $b = 2.00$
 $c = 1.00$ $d = 0.25$

100 x 101 x 101 Elements
Regular Hexahedral Elements
(2 x 2 x 2 Meter / 8 Nodes)

1,020,000 Elements
1,050,804 Nodes

1 Primary Unknown (ρ_w)
4 Secondary Unknowns (H, v_x, v_y, v_z)



Flow and Transport Model PERFORMANCE

POR-SALSA: H12 FLOW COMPARISON

Flow Solution

- *1,050,804 Primary Unknowns*
- *4,203,216 Secondary Unknowns*

Per Realization

15-25 Minutes

Particle Tracking

- *600 Particles*
- *Advective Travel Times / Distances*
- *Cumulative F-Quotients*

~10 Seconds

50 Realizations Overnight

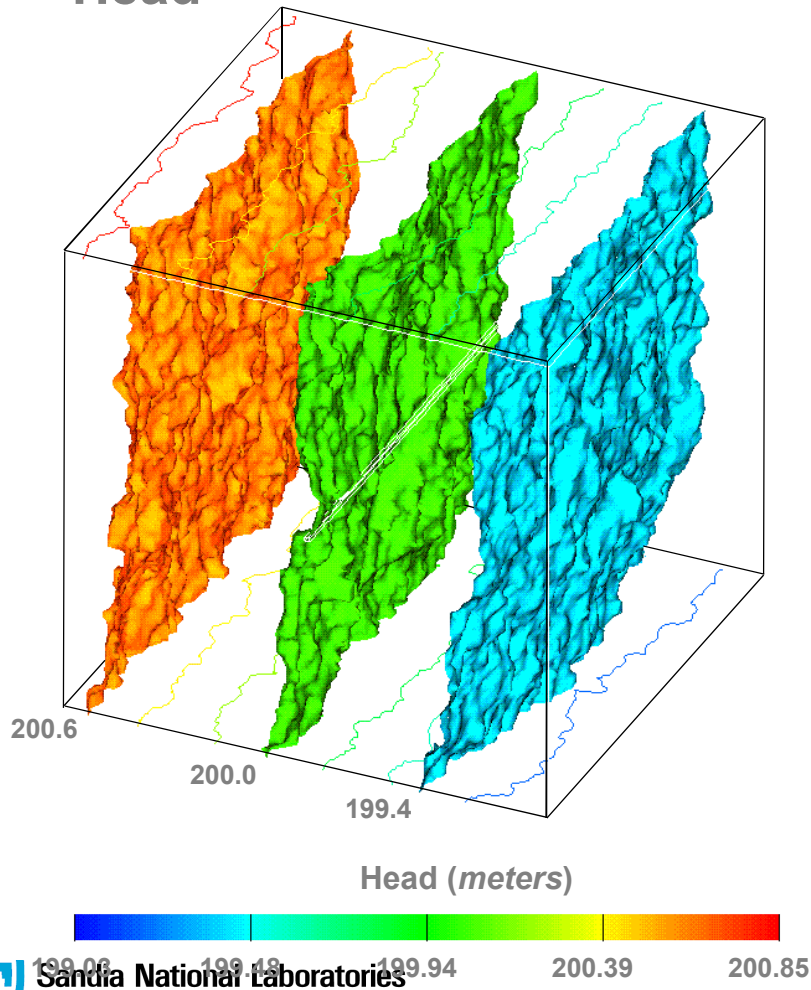


FLOW CALCULATIONS

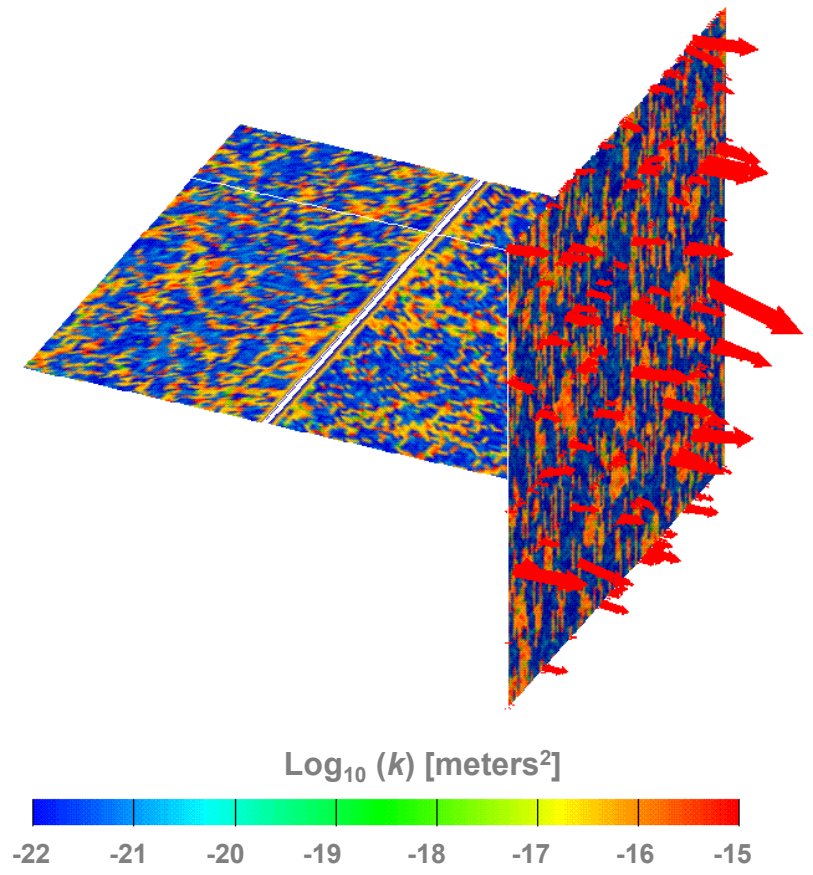
Stage 1 R44/50

POR-SALSA: H12 FLOW COMPARISON

Head



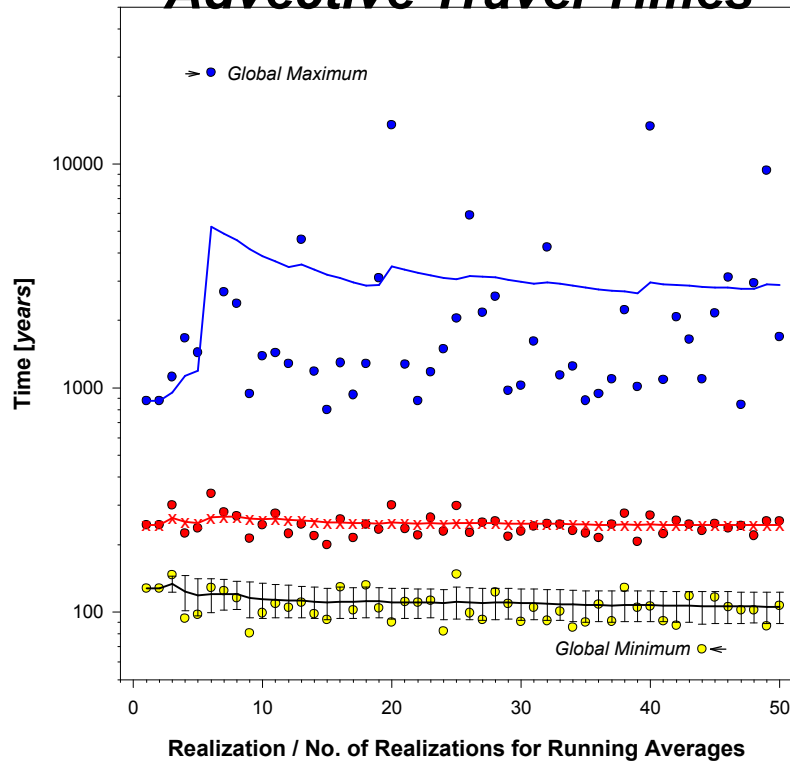
Outflow Fluxes



PARTICLE TRACKING

Stage 1: 50 Realizations

POR-SALSA: H12 FLOW COMPARISON *Advective Travel Times*



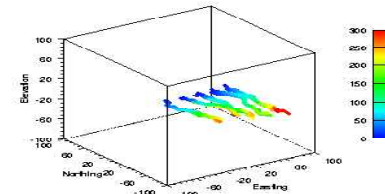
Model: TRACKWAY

(Alex H. Treadway, Dept. 6849)

600 Particles Per Realization

< 10 Seconds

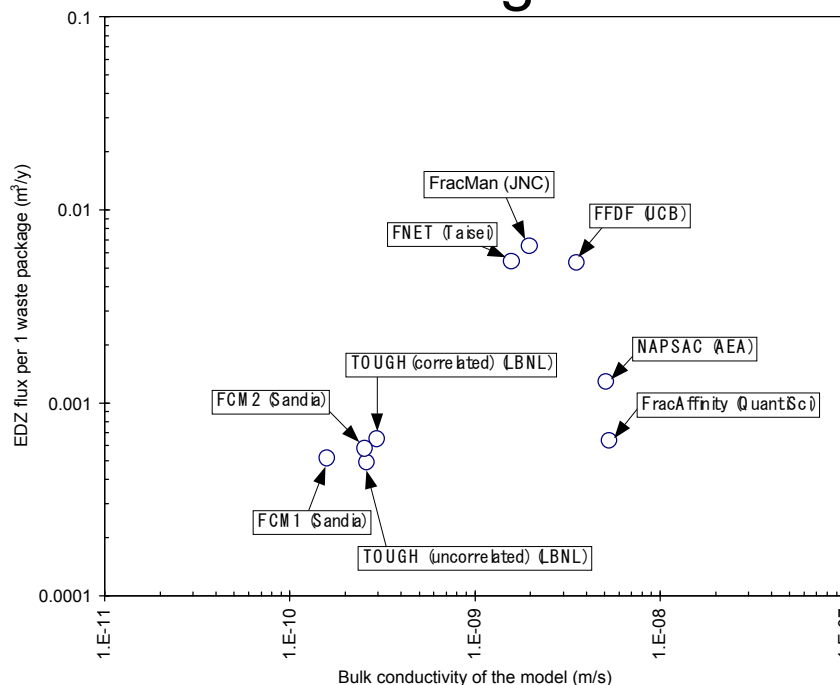
Ensemble Mean Minimum
 105.6 ± 15.6 Years



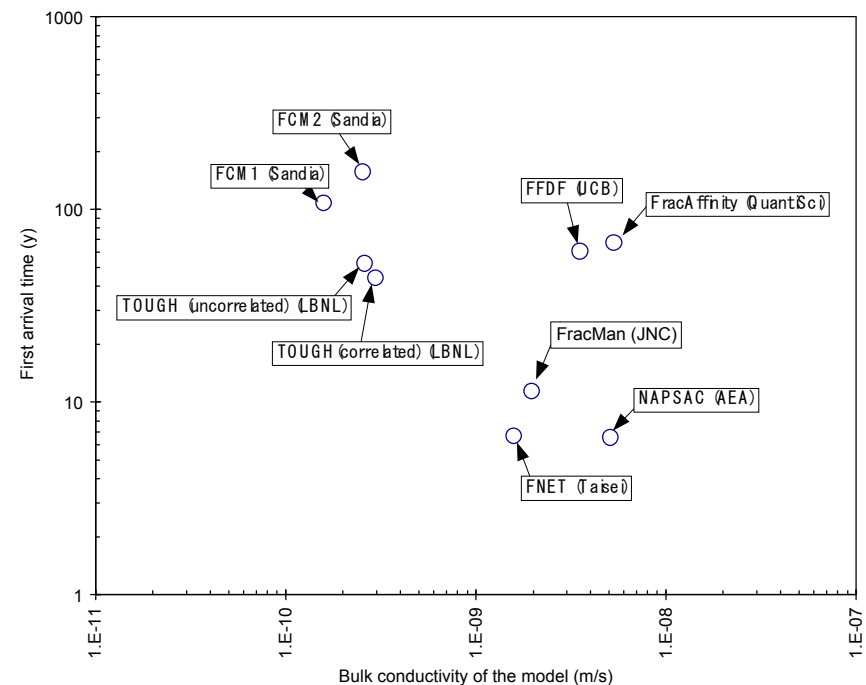
S

Comparing Conceptual Models

Bulk Conductivity vs. Flux through EDZ



Bulk Conductivity vs. First Arrival Time



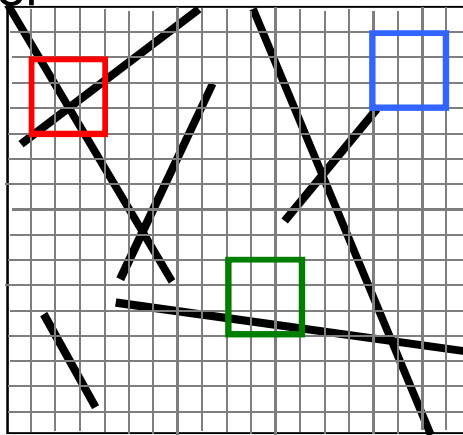
After Sawada, DRAFT, 1999



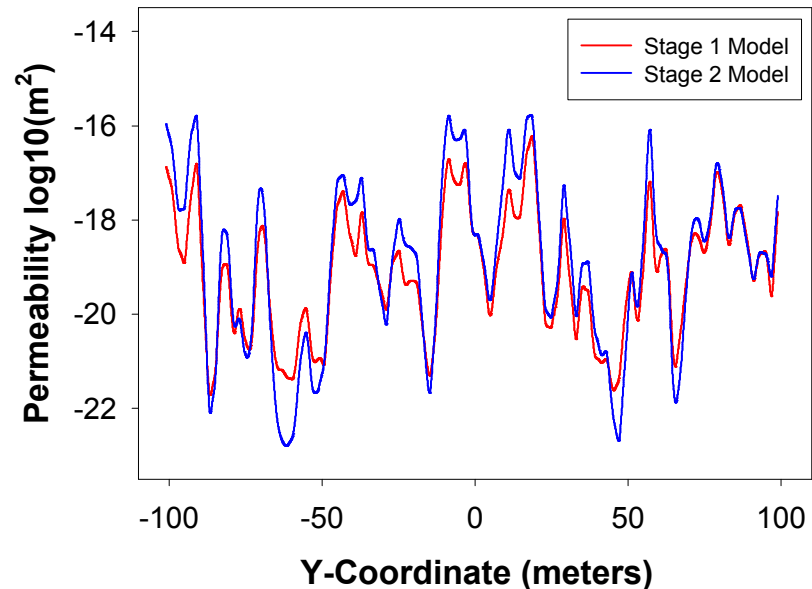
Variation on Conceptual Model

In Stage 1 frequency & coordination number values were drawn independently from location of conductive features

In Stage2 proportion of conductive features in moving template is used to draw from frequency and coordination number



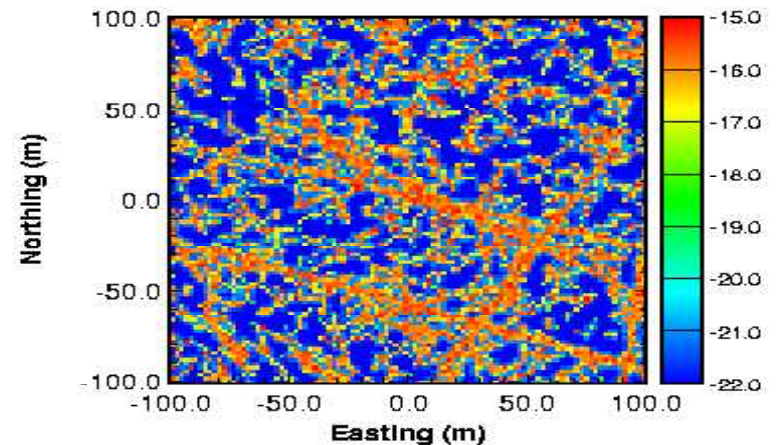
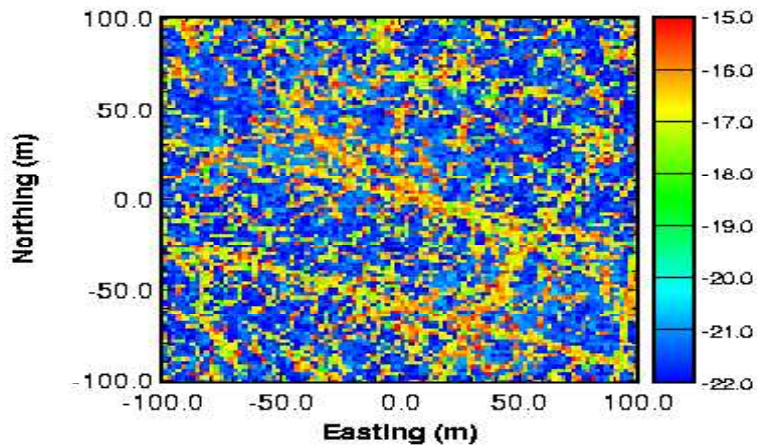
Results: greater difference between conductive and non-conductive zones



Comparing Permeability Models

Stage 1

Stage 2



Permeability in m^2



PROJECT OVERVIEW

POR-SALSA: H12 FLOW COMPARISON

Flow Simulator: **POR-SALSA**

Mesh

100 x 101 x 101 Elements
Regular Hexahedral Elements
(2 x 2 x 2 Meter / 8 Nodes)

1,020,000 Elements

1,050,804 Nodes

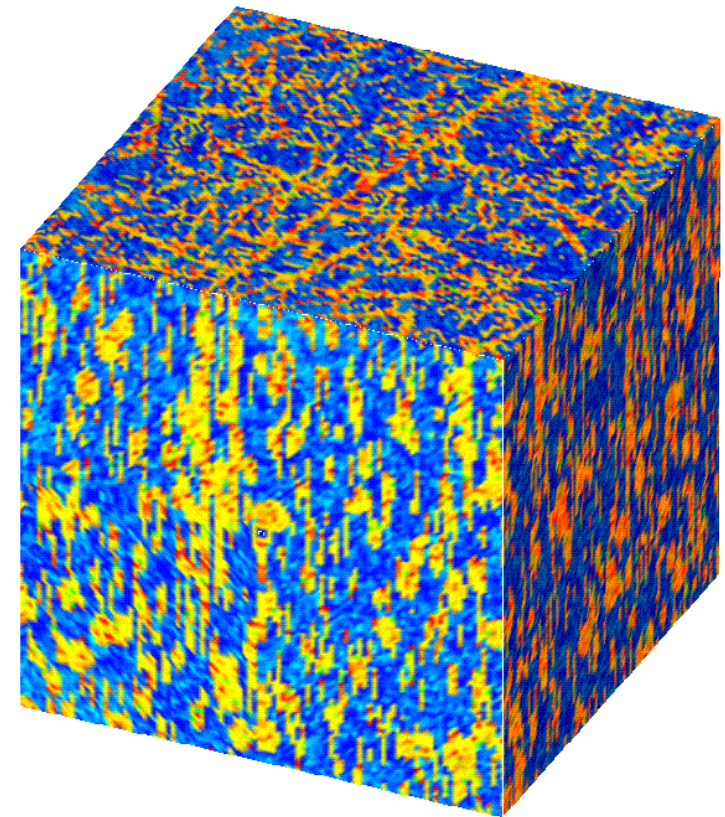
Flow Problem

Steady-State, Saturated
1 Primary Unknown (ρ_w)
4 Secondary Unknowns (H, v_x, v_y, v_z)

Performance

50 Realizations
20 Processors
17 Hours Total Time

Geostatistical Representation:
Fractured Continuum Model



$\text{Log}_{10} (k) [\text{meters}^2]$



Summary

- FCM represents new approach to modeling fracture permeability (EMT + Spatial Simulation)
 - Bridge between DFN and ECM models
 - Exploits capability of MPP flow and transport capabilities
- Compared to other approaches, results indicate:
 - FCM Bulk hydraulic conductivity similar to ECM's (about 1 order of magnitude lower than DFN models)
 - FCM produces longest times to first arrival (100+) years (DFN model first arrivals are roughly 5-15 years)

