



# Computational Science and Engineering at DuPont

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**Richard W. Nopper, Jr.**  
**Senior Research Associate**  
DuPont Engineering Research & Technology





## Topics

The DuPont Company and our diverse portfolio of science and engineering problems

Flagship problems

underground injection

stratospheric ozone depletion and greenhouse warming

computational chemistry

computational material science

finite-element analysis and CFD

bio-informatics

data processing

Final comments about modeling



## The Foundation of DuPont



### **CORE VALUES**

*Safety and Health*

*Environmental Stewardship*

*Highest Ethical Behavior*

*Respect for People*



## The Vision of DuPont



### **WE ARE A MARKET-DRIVEN SCIENCE COMPANY**

*Our vision is to be the world's  
most dynamic science company,  
creating sustainable solutions essential  
to a better, safer, healthier life  
for people everywhere.*



## The Mission of DuPont



### **SUSTAINABLE GROWTH**

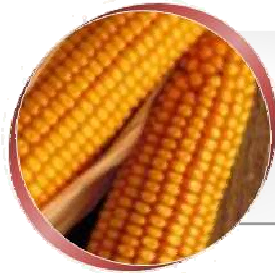
*The creation of shareholder and societal value while we reduce the environmental footprint\* along the value chains in which we operate.*

\* DuPont defines “footprint” as all injuries, illnesses, incidents, waste, emissions, use of water and depletable forms of raw materials and energy.



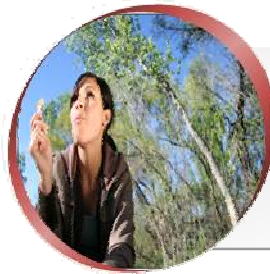
# DuPont Continues to Evolve

DuPont Pro Forma Sales – 2010\*



\$7.8 B

**Agriculture**



\$3.0 B

**Nutrition & Health**



\$3.8 B

**Performance Coatings**



\$2.8 B

**Electronics & Communications**

**\$34.2B\***

\$6.3 B

**Performance Materials**



\$6.3 B

**Performance Chemicals**

\$3.4 B

**Safety & Protection**



\$0.9 B

**Industrial Biosciences**

\* Includes \$0.2B in 'other' sales.  
Total company sales exclude transfers.



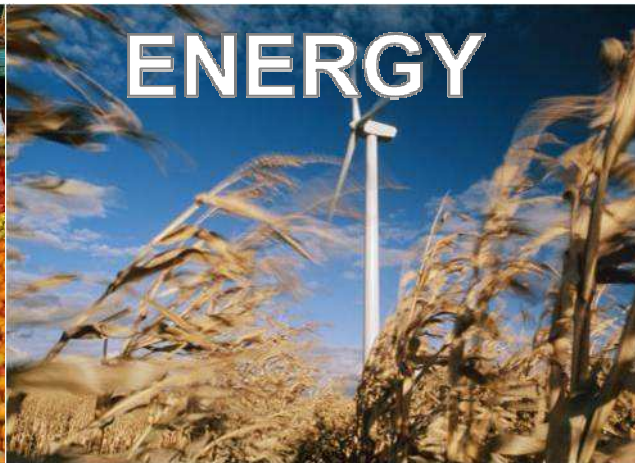


Sometime in 2011, the earth's population will reach 7 billion.  
By 2050, it will be 9 billion.



© National Geographic images

*Feeding The World*



*Reducing Our Dependence On Fossil Fuels*



*Keeping People And The Environment Safe*

**67,000 DuPont employees in more than 90 countries are working to find solutions through are applying our SCIENCE to find solutions to some really BIG challenges...**

**\$1.7 billion DuPont Actual R&D Spend in 2010**



## Integrating our science & technology to find solutions.



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**Engineering, Analytical Science, Scientific Computing, Modeling & Simulation, Toxicology, Environmental Science, Regulatory Science, Application Development**

■ Chemical Sciences   
 ■ Biological Sciences   
 ■ Material Sciences   
 ■ Enabling Technologies



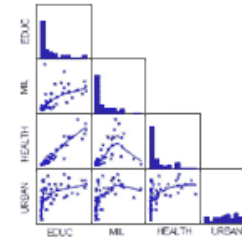
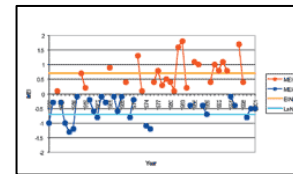
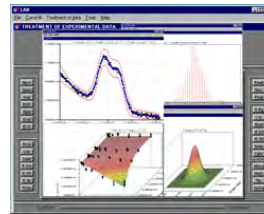


# Modeling is a powerful technology used to:

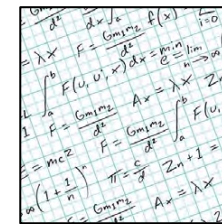
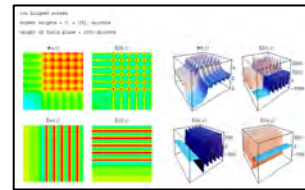
(0) help define the problem



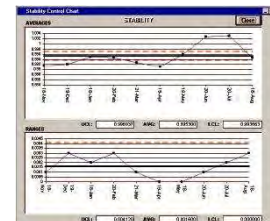
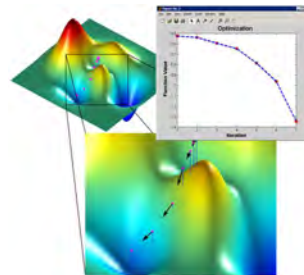
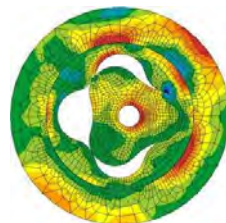
(1) organize & interpret data



(2) capture fundamental understanding  
-- qualitatively & quantitatively



(3) support the design, optimization, & control of products & processes







Over the years, the DuPont HPC Environment has followed developments in hardware -- and hardware costs.

Run a host of applications  
CFD, FEA, Comp Chem, Bioinformatics, DB, etc.

Include storage & backup services

Use multiple platforms (SMP, Clusters, etc)

Run batch jobs through LSF

Provide a cloud computing environment for large jobs







## Overview: underground injection

Modeling is one of the means required by USEPA to demonstrate that operations are protective of human health and the environment.

The demonstration of safe operations may be based on:

**flow and containment:** advective transport, diffusion  
injectate is confined both vertically and horizontally

**chemical fate:** chemical kinetics and transport  
injectate is rendered non-hazardous

There are also structural integrity issues to demonstrate:

**induced seismicity** evaluation  
changes in stresses due to fluid injection do not de-stabilize the rock mass

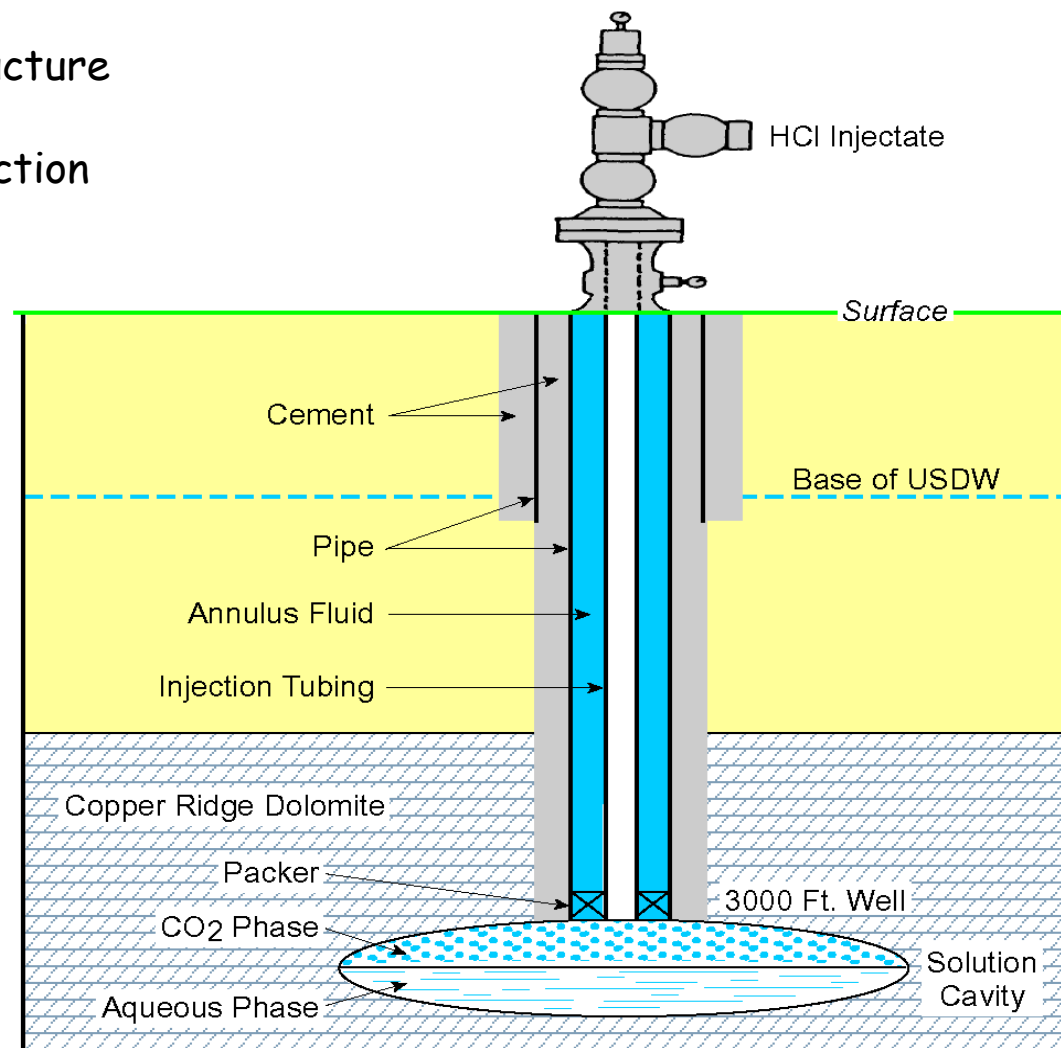
**stability** of nearby rock mass  
presence of well and cavity (if present) do not de-stabilize the rock mass

## Class I injection wells are regulated by USEPA and state and local authorities.

"... not just a hole in the ground"

a highly engineered structure

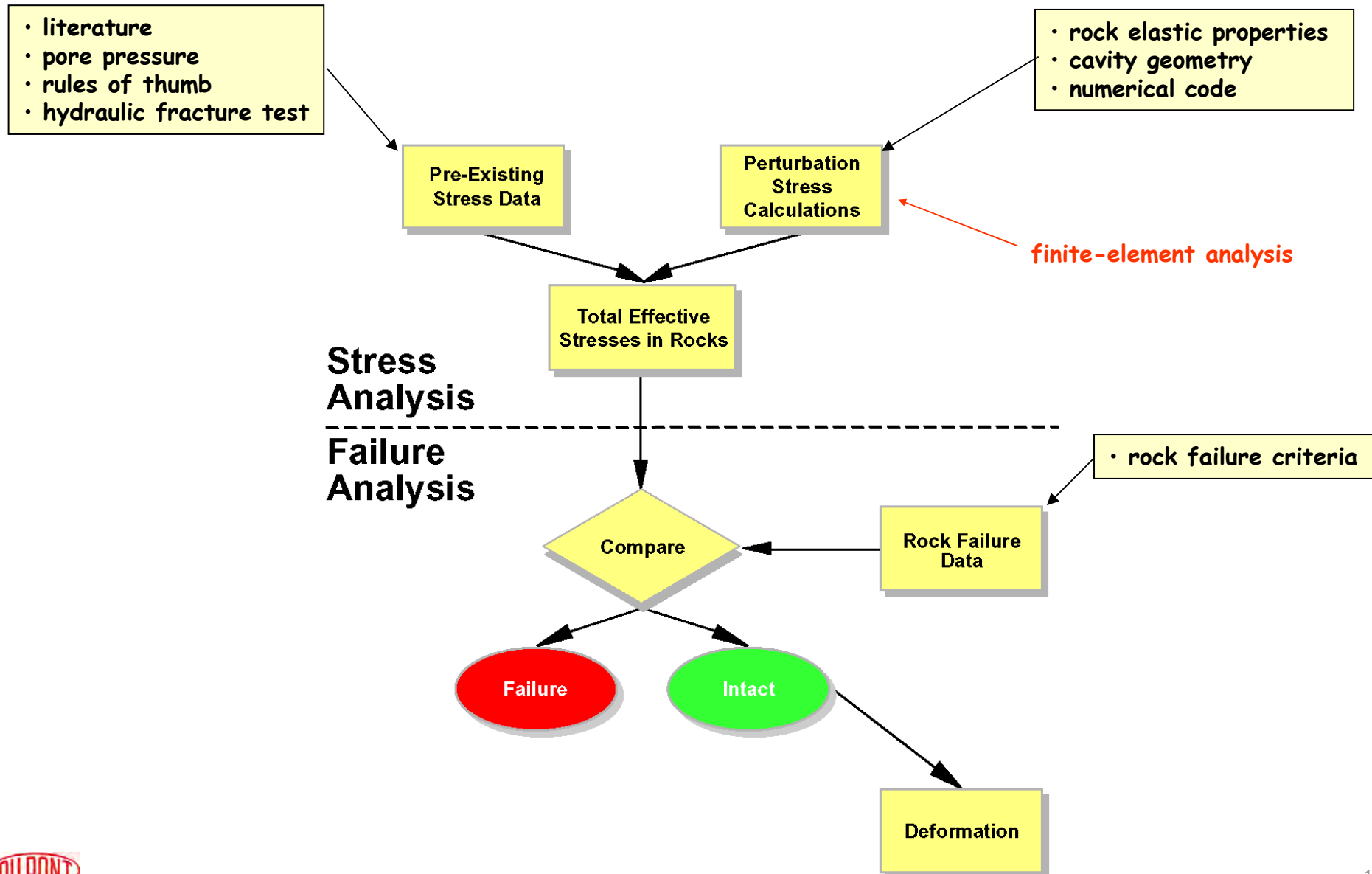
multiple layers of protection



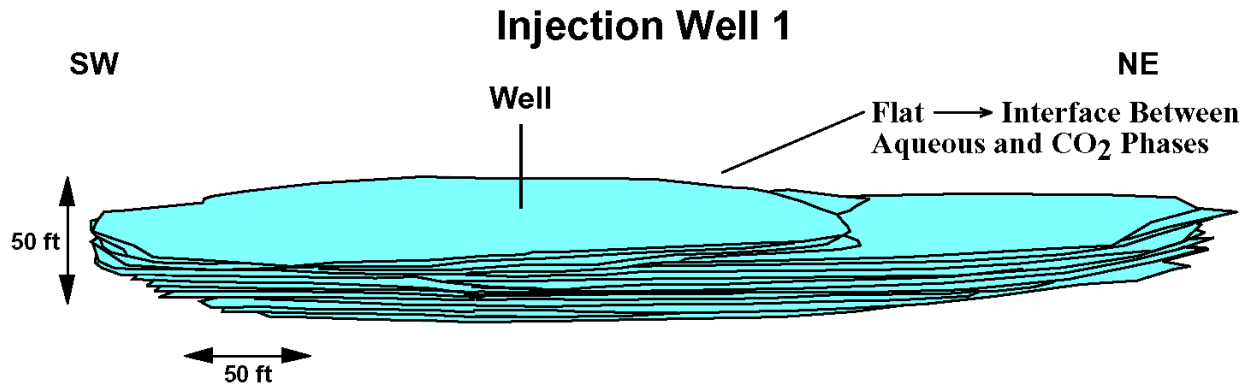
**Case study. Acidic fluid injected into a carbonate rock creates a solution cavity.**

**Is the cavity stable?**

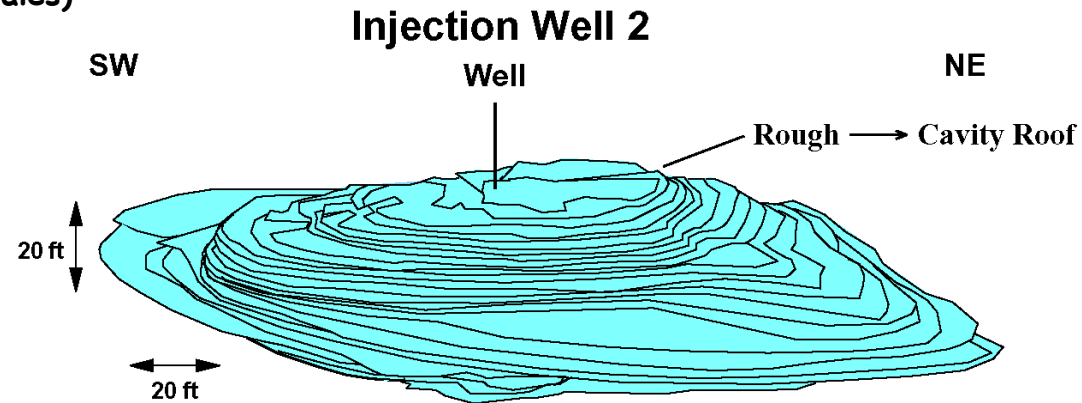
The structural integrity analysis is composed of a stress analysis, followed by a failure analysis.



Geometry of cavities as measured by Sonar Caliper Survey and confirmed by borehole televiewer.



(note different spatial scales)



*Question: Could these cavities fail?*



The stress calculation is straightforward once the basic data are in hand.

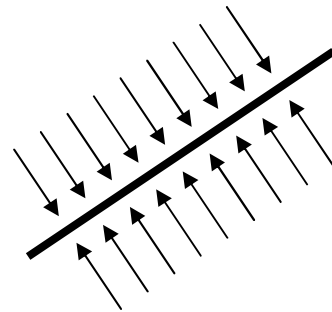
The test for failure is straightforward, although tedious.

Specify a set of potential failure planes

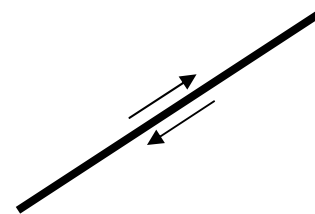
- at various locations near the cavity
- at various orientations

Calculate normal and shear stresses for each

Observe whether failure envelope is exceeded



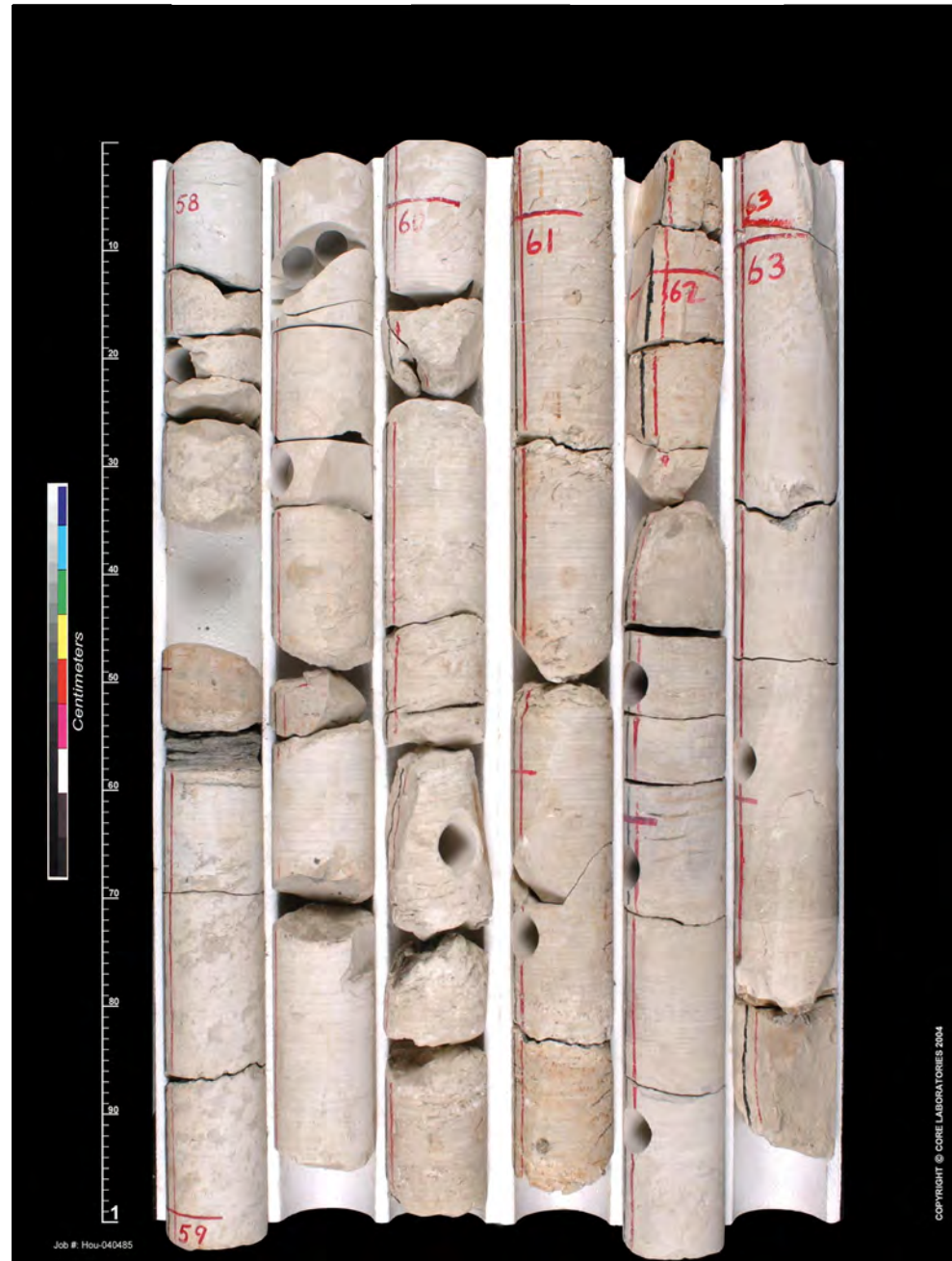
normal stress



shear stress

Rock mechanical data  
come from selected core  
samples.

*Core showing plugs cut out  
for rock mechanics and  
permeability testing.*

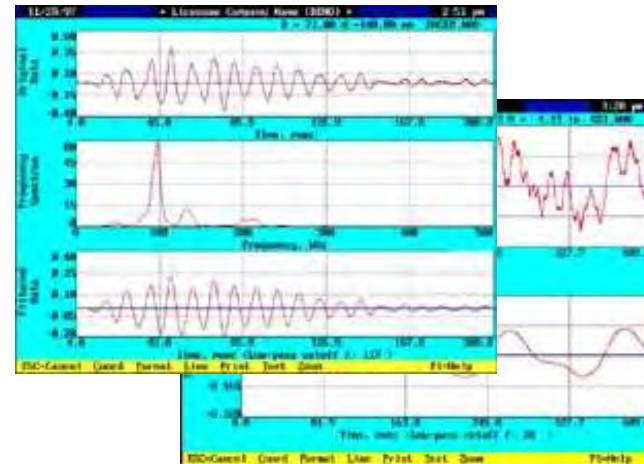


## Rock elastic properties were measured both statically and ultrasonically.

static values from "rock squeezing"  
for the structural integrity analysis



ultrasonic values from wave propagation studies  
for correlation with well logs



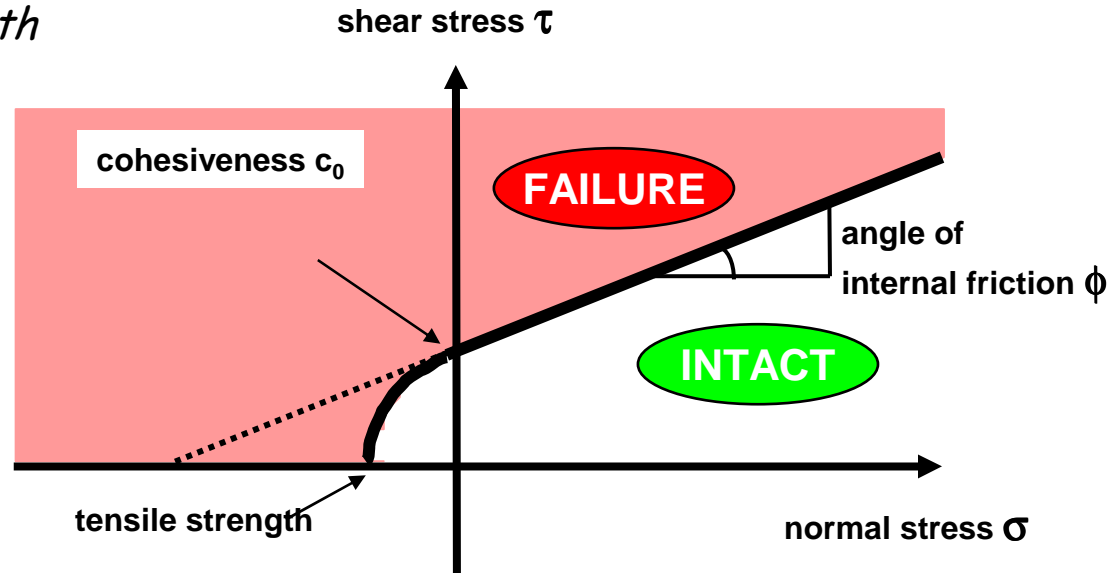
Isotropic elastic materials are characterized by  
density  $\rho$ , Young's modulus  $E$ , & Poisson's ratio  $\nu$

$$E = \text{stress} / \text{strain}$$

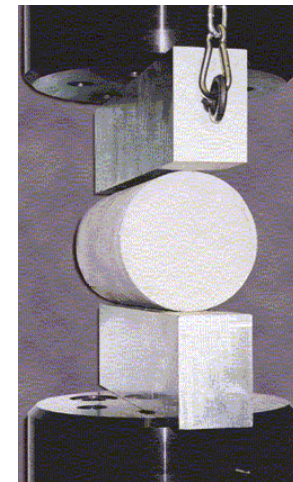
$$\nu = \frac{\text{lateral strain}}{\text{axial strain}}$$

Static "rock squeezing" determined the failure envelope for each cored interval.

*cohesiveness (related to unconfined compressive strength)*  
*angle of internal friction*  
*tensile strength*



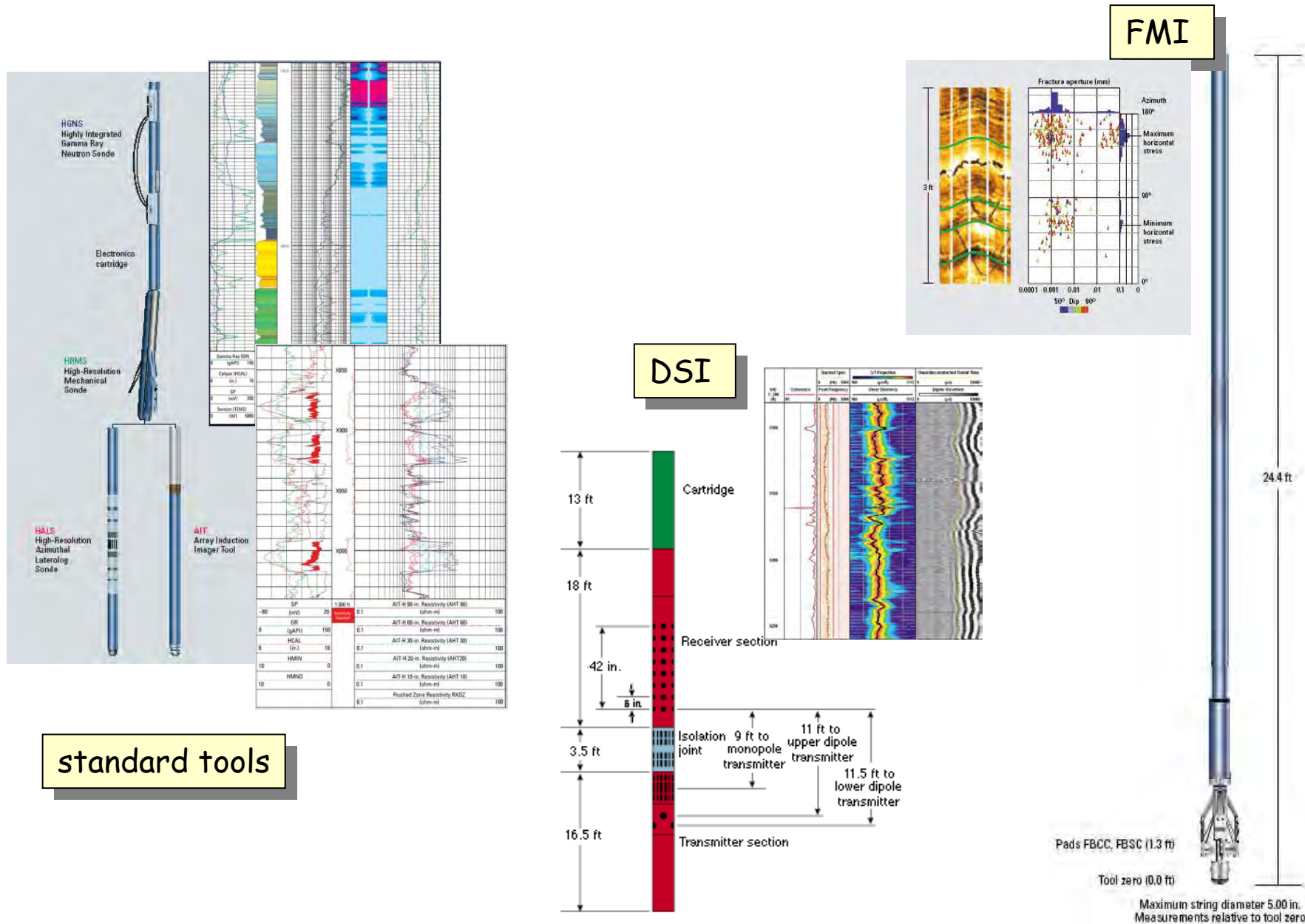
*triaxial test stand*



*"Brazilian test" measures the tensile strength*

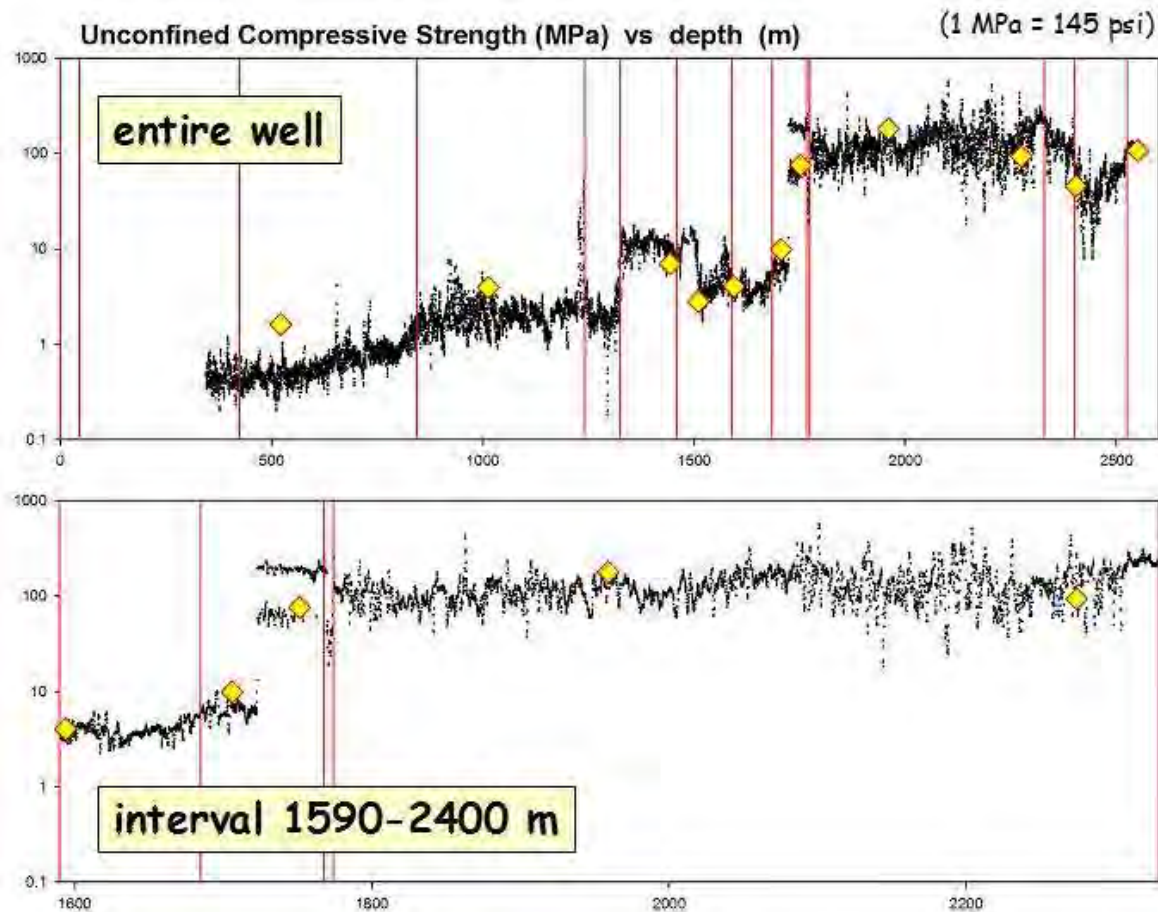


# Wireline well logs provided continuous coverage down the well.



The MEM provides a “log” of the rock mechanical properties down the well. For example, the unconfined compressive strength (UCS), here shown with

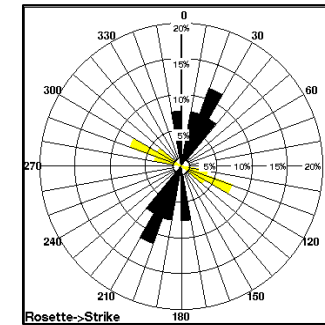
- o the geological interpretation (red vertical lines) and
- o the Core Lab measurements (yellow diamonds)



Well measurements allow us to estimate the orientation and magnitude of the background stress field.

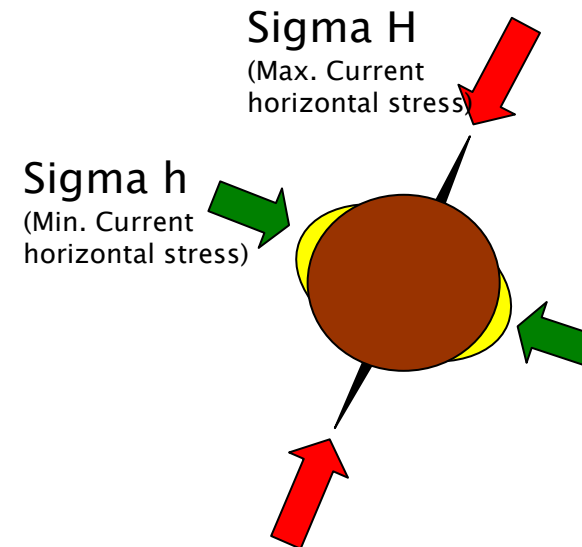
Breakouts and drilling-induced fractures observed in well logs indicate local **stress field orientation**

Schlumberger FMI log



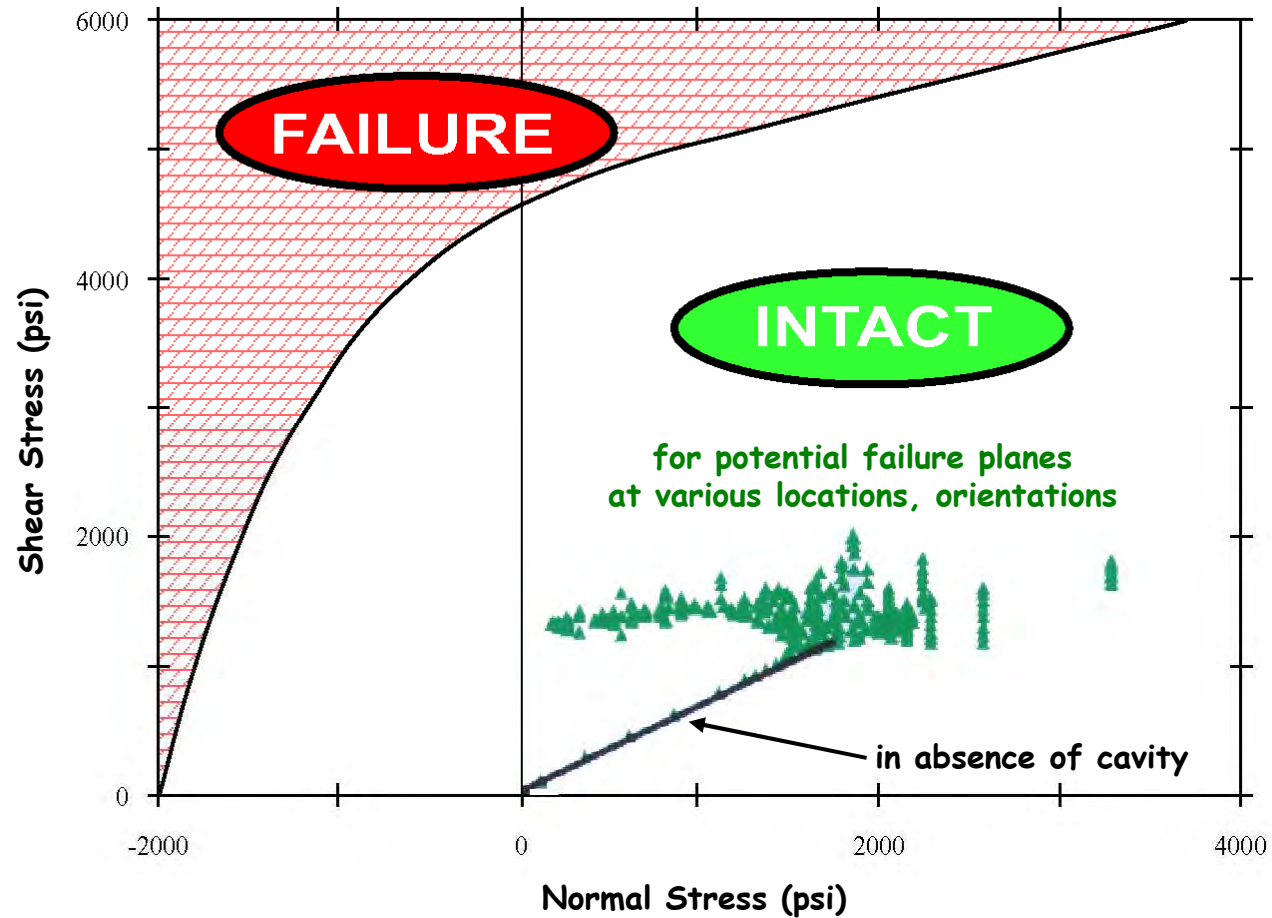
A hydrofrac test is the best available estimate of the **least horizontal stress**.

From Schlumberger's DSI log, we infer **greatest horizontal stress**.



From the density log, we estimate the **vertical stress**, which comes from the "overburden" (gravity acting on mass density).

We pass the test for failure on both cavities.  
Both are expected to remain stable for the duration of injection operations.





## Overview: stratospheric ozone depletion and global warming

Modeling is used by the world scientific community to unravel the mechanisms of ozone depletion and to predict the greenhouse warming caused by chemicals emitted on a large scale.

### global chemistry-transport model

chemical kinetics, gas phase and heterogeneous  
atmospheric dynamics: advective and diffusive transport  
radiative forcings  
Ozone Depletion Potential



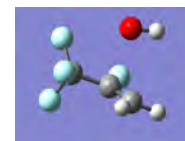
### warming model & emissions inventories

estimate/predict atmospheric loadings  
Global Warming Potential

$$GWP_i \equiv \frac{\int_0^{TH} RF_i(t) dt}{\int_0^{TH} RF_r(t) dt} = \frac{\int_0^{TH} a_i \cdot [C_i(t)] dt}{\int_0^{TH} a_i \cdot [C_i(t)] dt} = \frac{\int_0^{TH} a_i \cdot [C_i^0 \exp(-t/\tau)] dt}{\int_0^{TH} a_i \cdot [C_i(t)] dt}$$

### quantum chemistry models

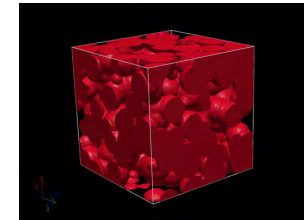
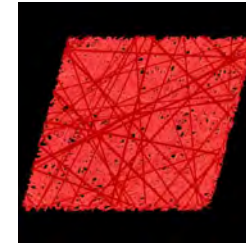
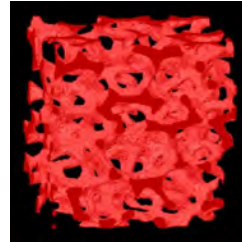
estimate energy balances due to molecular degradation  
of greenhouse gases  
provide inputs to the models above



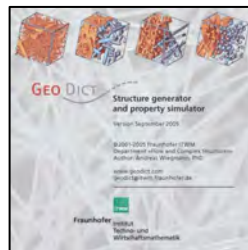
## Transport properties of disordered media (computational material science).

Many DuPont processes and products involve disordered media.

fuel cell components  
filter and barrier media  
thermal insulation, sound/noise control  
clothing and hygiene products  
composite materials  
packed beds  
aquifers



*Difficult for a finite-element approach!*

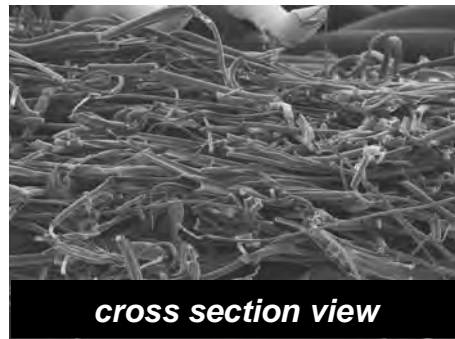
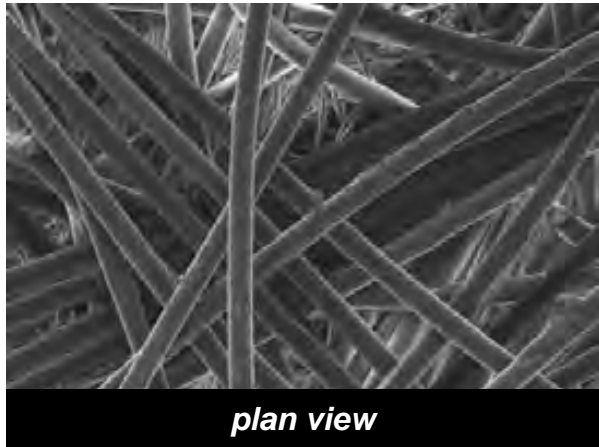


GeoDict can simulate a wide variety of structures.

*synthesize geometry  
analyze porosity  
calculate flow field  
simulate filtration/barrier performance*

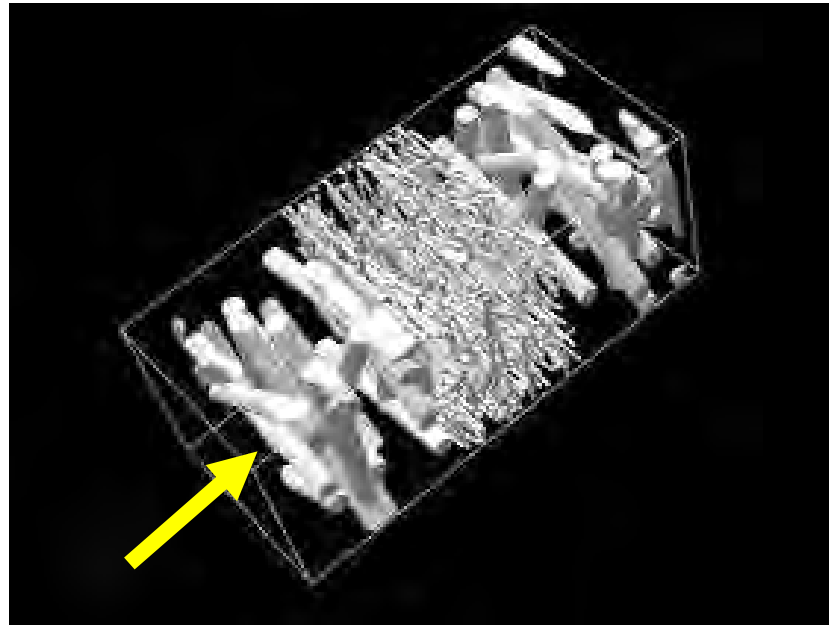
Case study. Predict the barrier performance of a 3-layer fibrous medium.

SEMs



Step one: generate a virtual sample.

model of medium

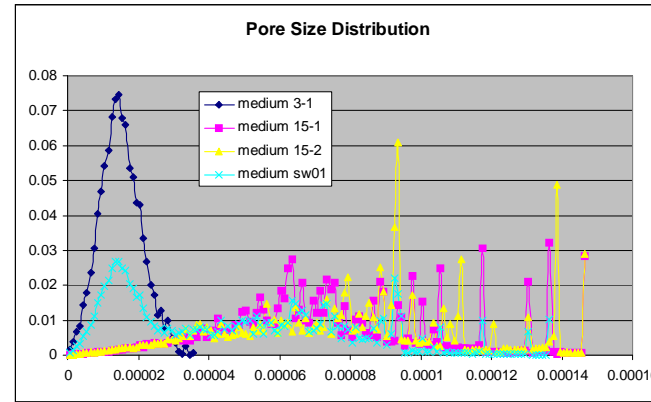


simulated SEMs

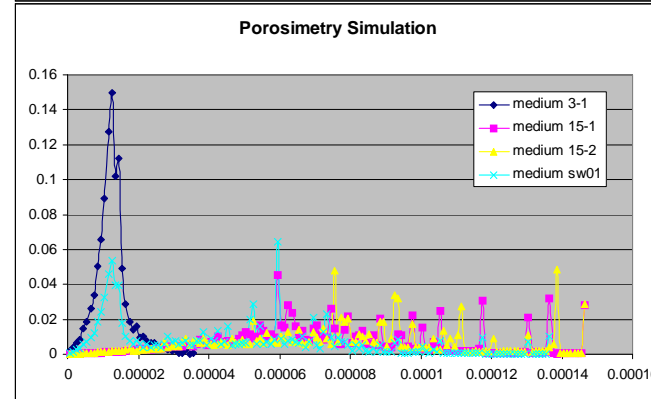


Given a synthetic structure, the code can predict its pore properties.

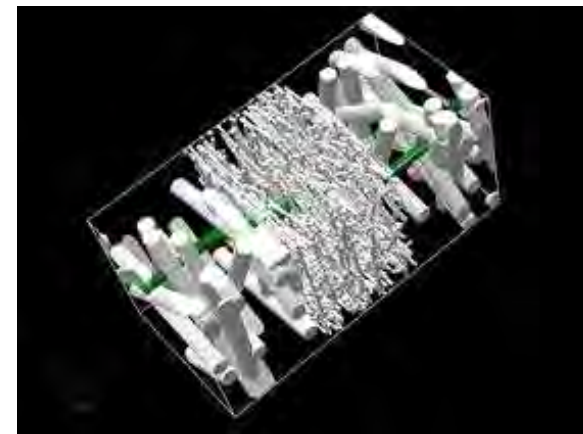
pore size distribution for  
all pores



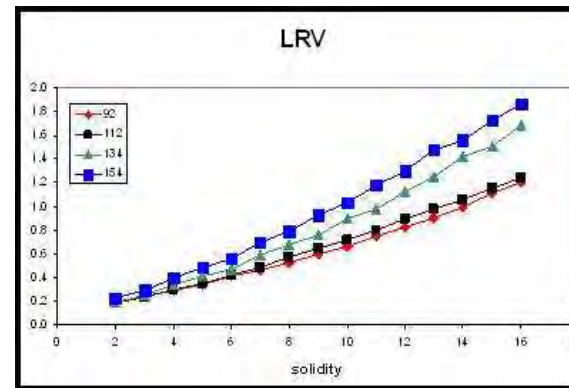
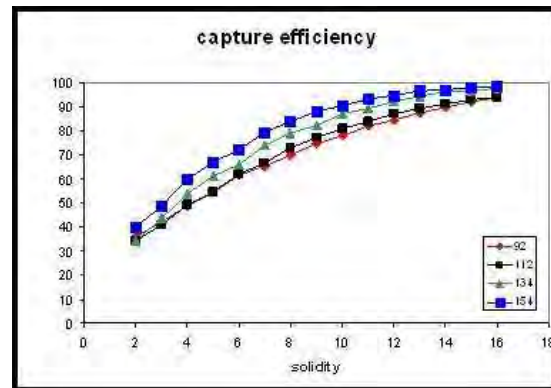
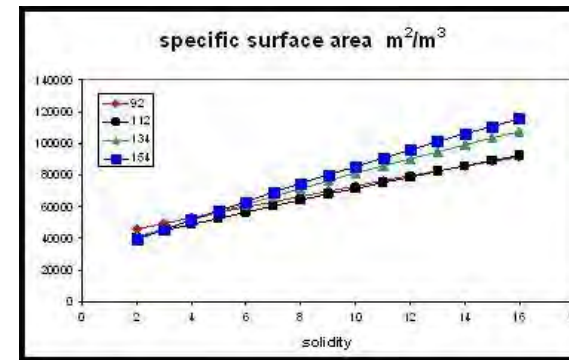
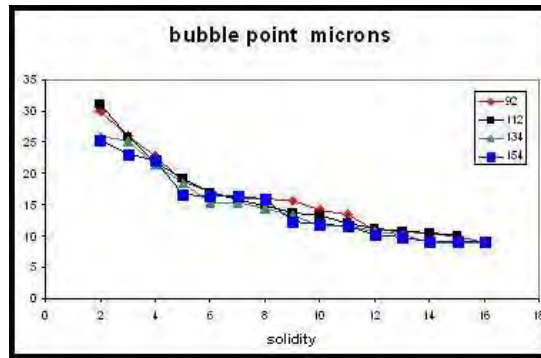
through-pores only



percolating  
sphere emulates  
bubble point



Scripting the runs allows a large number of number of models to be run efficiently with minimal intervention.



This procedure generates quantitative relationships, showing sensitivities to key parameters.

These can be used as properties in a finite-element model of the overall device.



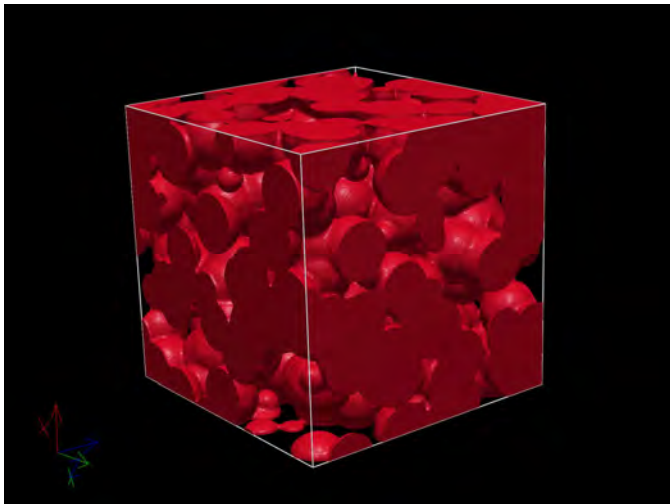


# Case study. The loss of permeability in a sandstone aquifer due to the presence of fine particles.

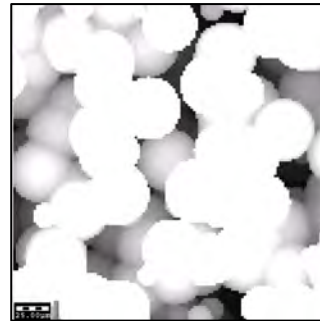


Consider, e.g., 80% / 20% mixture of 50 mm / 20mm sand grains:

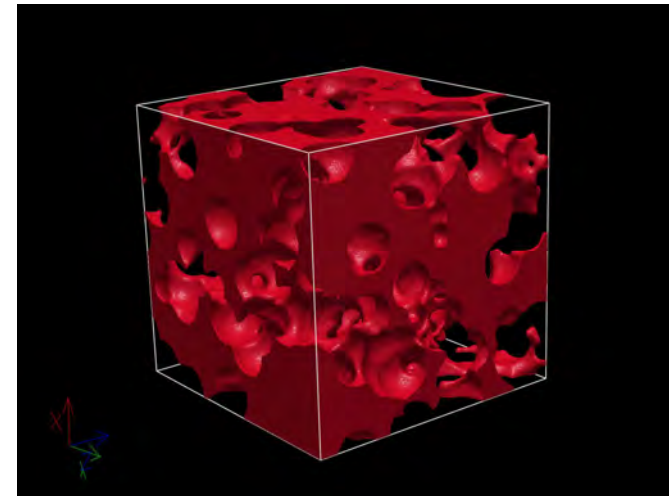
structure with 30% porosity



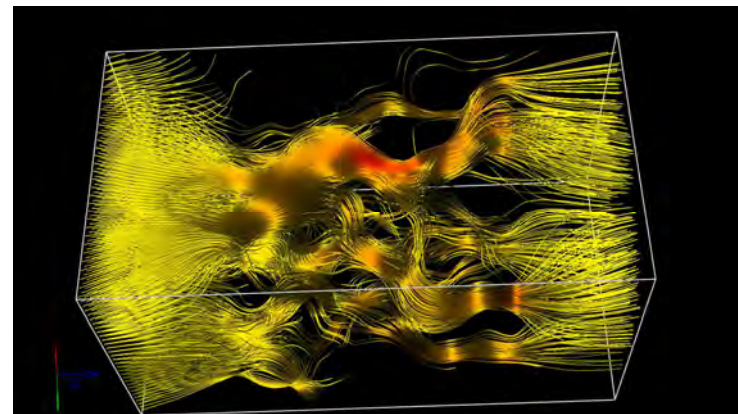
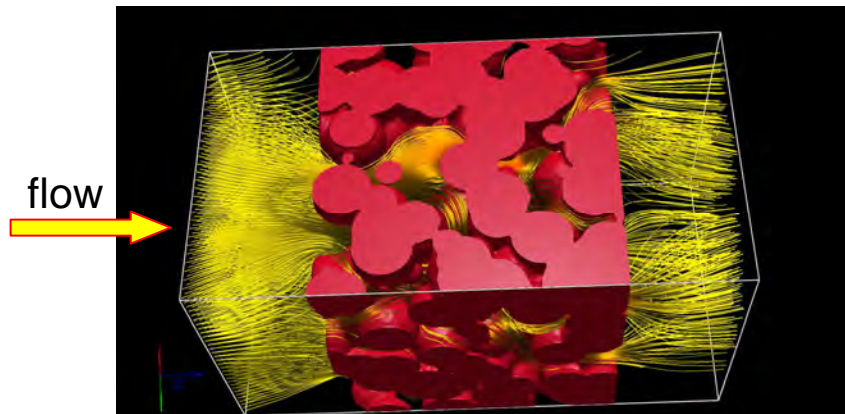
simulated SEM



pore structure

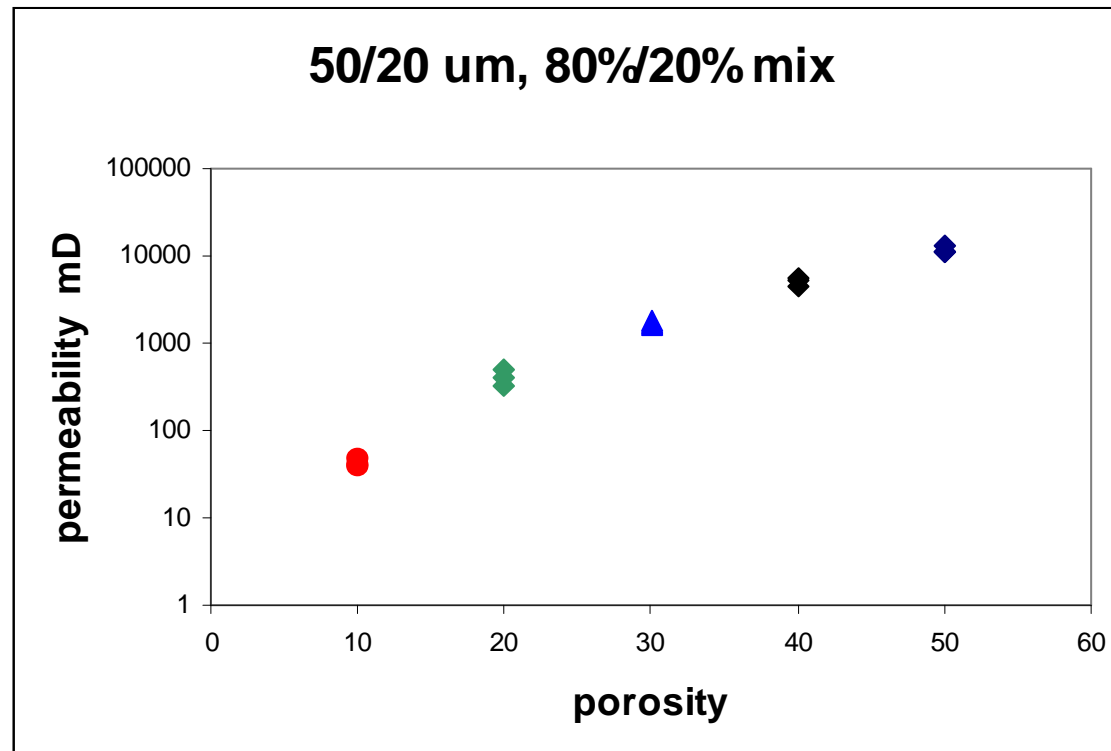


streamlines through the structure





A scripted suite of flow models gives us a quantitative relationship for permeability as a function of porosity -- with no adjustable parameters -- that agreed well with lab measurements.

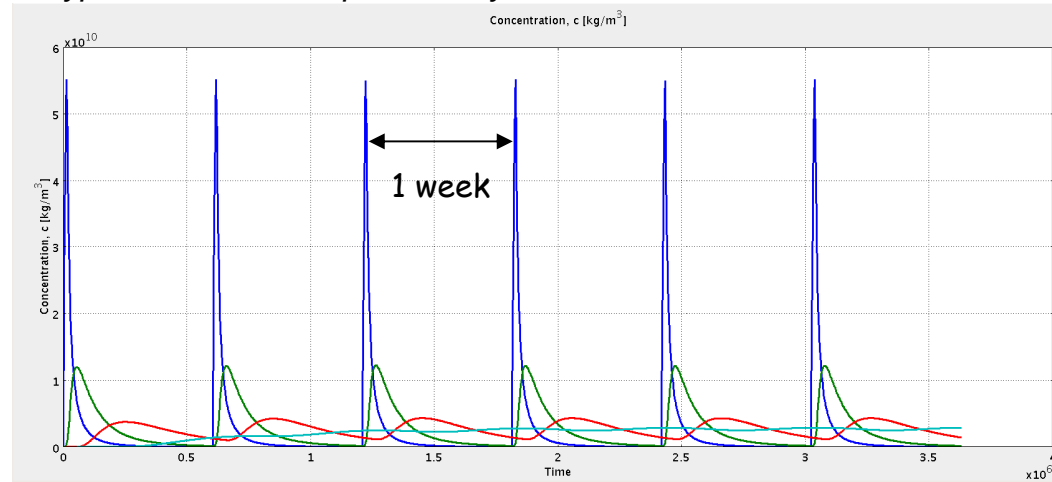


# Case study: Transport and dispersion of a train of treatment pulses injected into an aquifer.



goal: design dosing of the aquifer (concentration, pulse length, frequency)

typical result: a 4-hr pulse every week, seen at 1, 3, 6, 10 m from well



Two step procedure, using COMSOL:

## 1. Darcy's Law.

Pump pressure establishes a background steady-state pressure and flow fields (the well is pumping water for some time before our solute pulses are introduced).

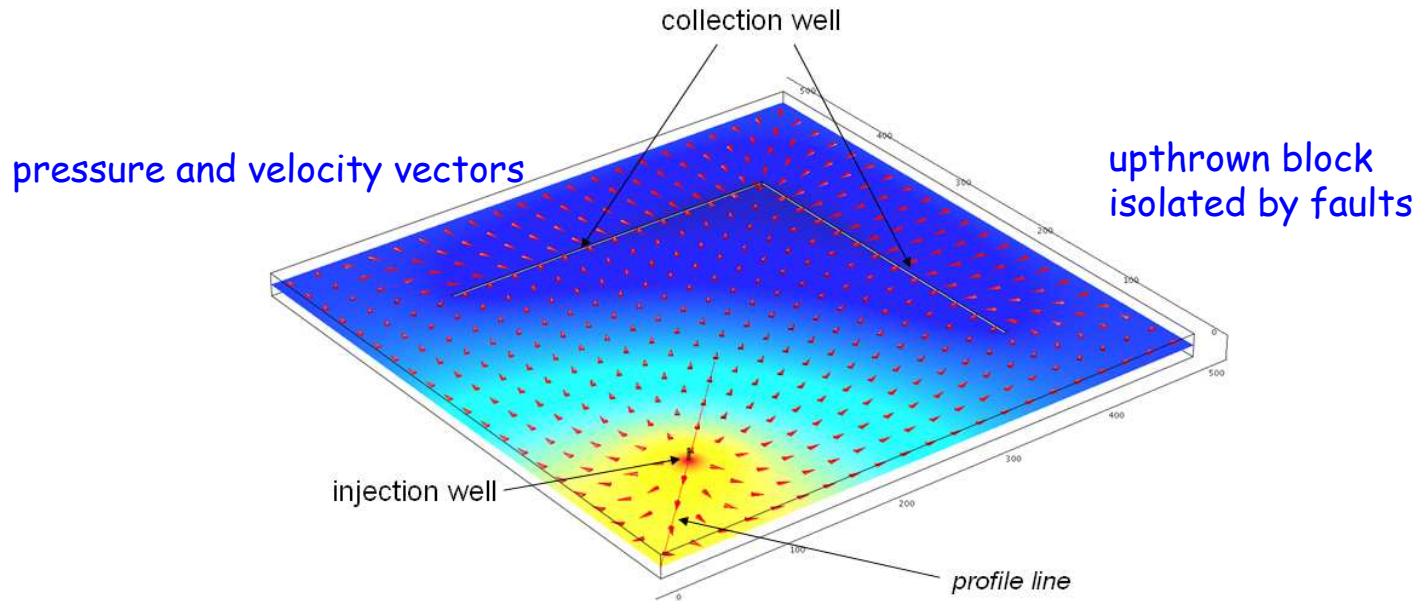
~~$$S \frac{\partial p}{\partial t} + \nabla \cdot \left[ -\frac{\kappa}{\eta} (\nabla p + \rho_f g \nabla D) \right] = Q_s$$~~

## 2. Advection-dispersion equation.

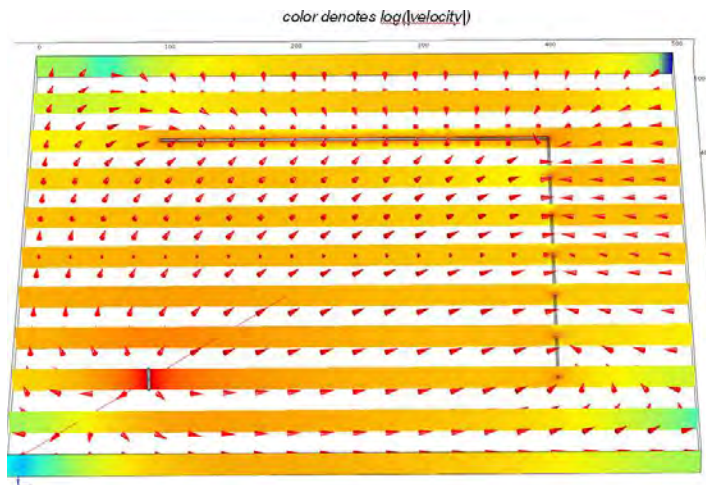
Pulses of solute are transported by the background flow field, dispersing as they travel.

$$\theta_s \frac{\partial c_i}{\partial t} + \rho_b \frac{\partial c_{Pi}}{\partial c} \frac{\partial c_i}{\partial t} + \nabla \cdot [-\theta_s D_{Li} \nabla c_i + \mathbf{u} c_i] = R_{Li} + R_{Pi} + S_{ci}$$

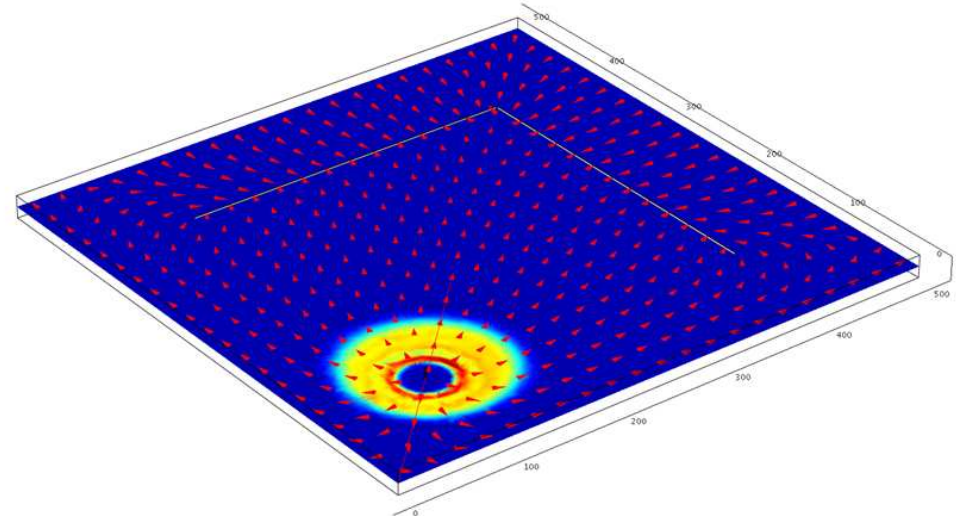
# A 3-D model shows the injector/producer geometry.



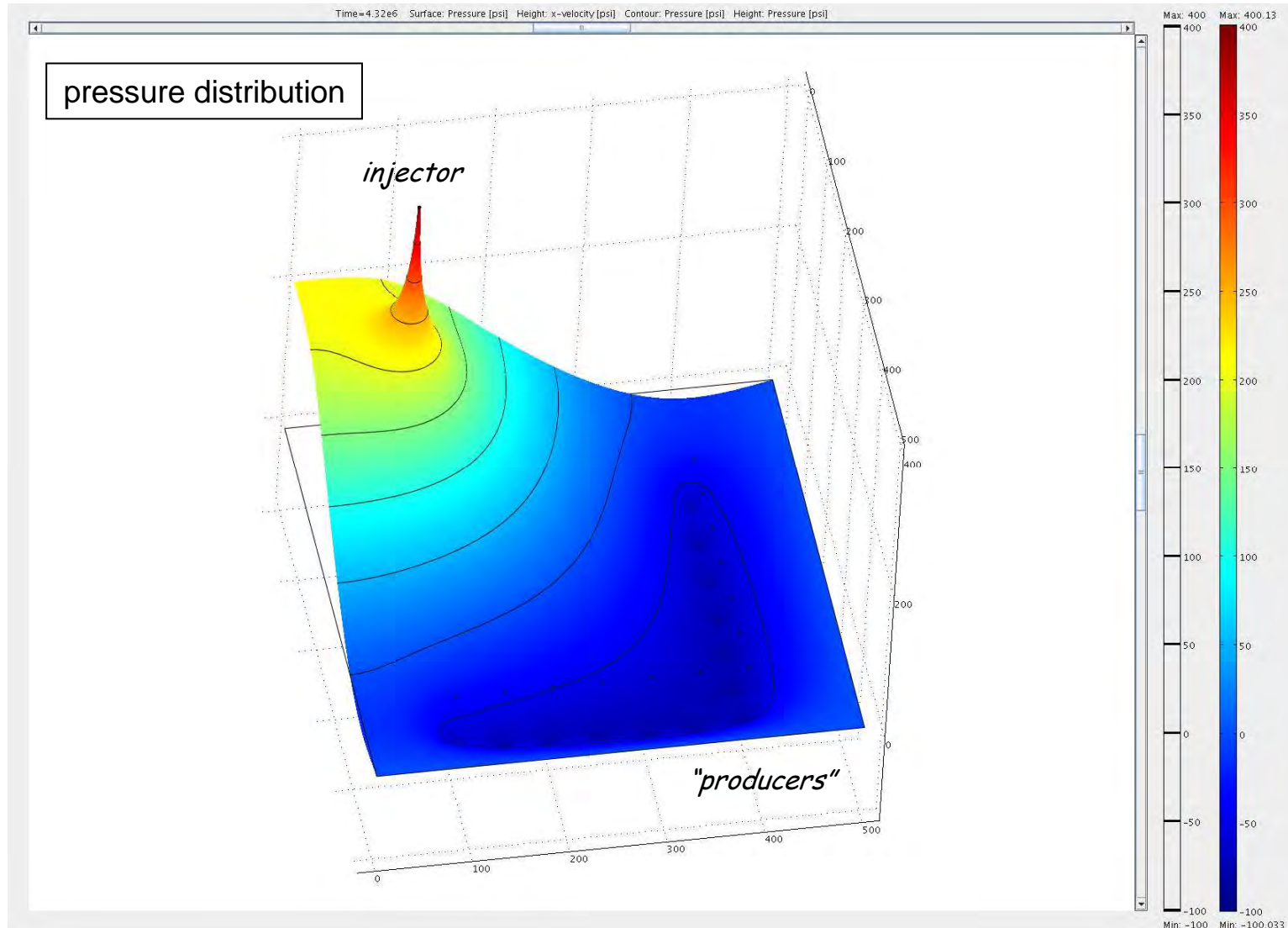
## 3-D velocity field



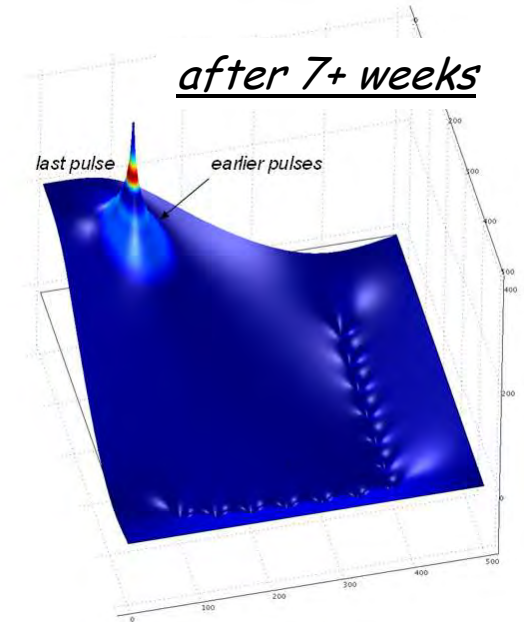
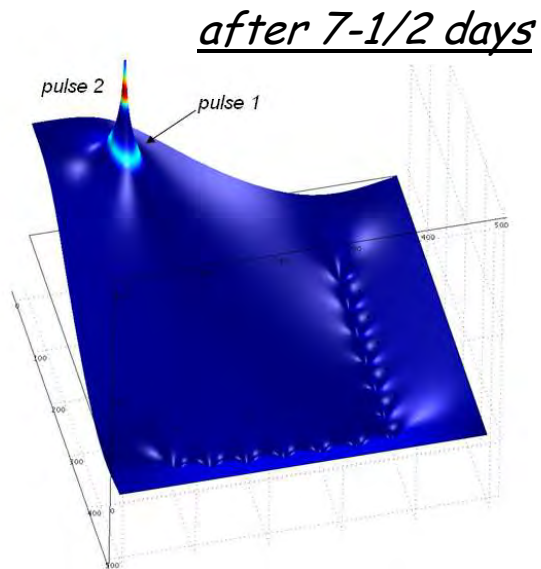
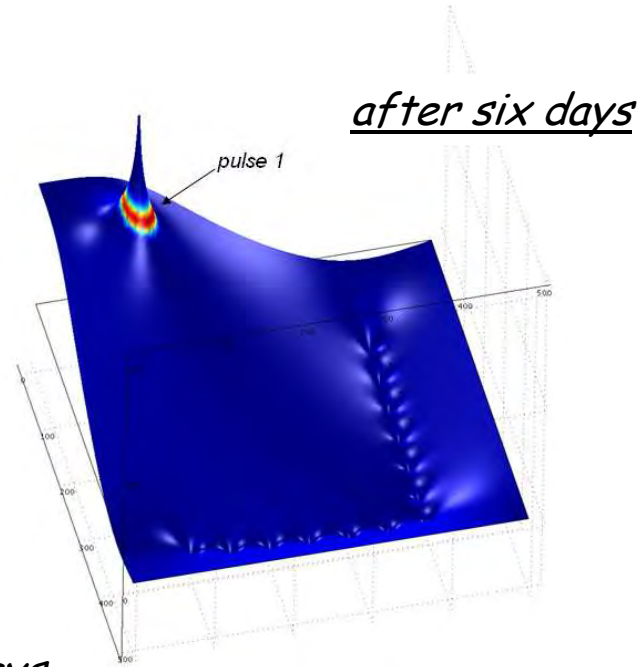
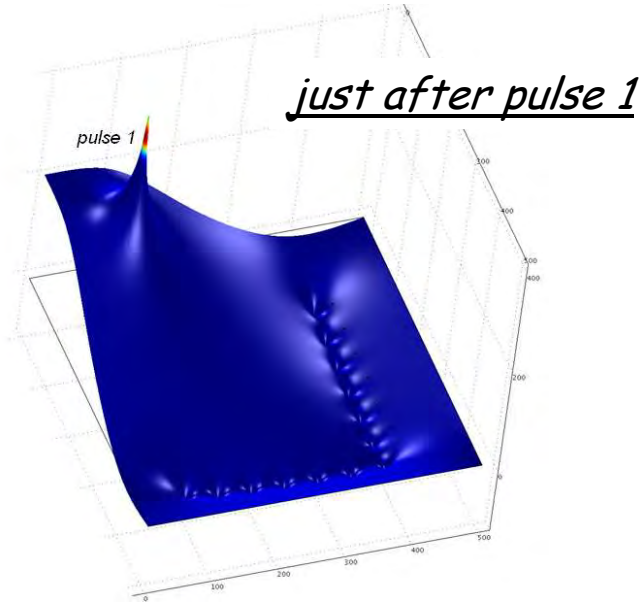
## Outward solute transport after five weekly pulses



We can do it more conveniently in 2-D, however, since the aquifer is thin.



The pulses move down-gradient, spread geometrically, and disperse.



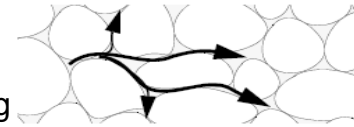


Case study. The numerical value of dispersivity is a key question in scaling up from lab to the field: empirically, its value depends on the scale size of the measurement.



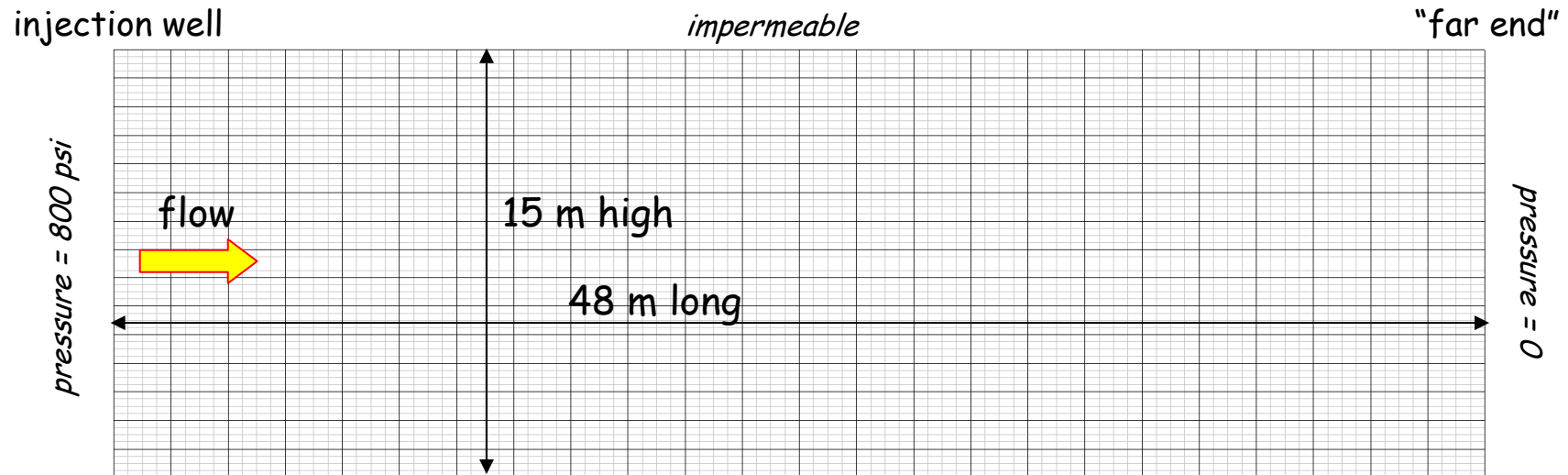
$$\theta_s \frac{\partial c_i}{\partial t} + \rho_b \frac{\partial c_{Pi}}{\partial c} \frac{\partial c_i}{\partial t} + \nabla \cdot [-\theta_s \mathbf{D}_{Li} \nabla c_i + \mathbf{u} c_i] = R_{Li} + R_{Pi} + S_{ci}$$

dispersivity -> hydrodynamic mixing



We show that heterogeneity of an aquifer can be important in dispersing a solute and can be the basis for a scale dependence of the dispersivity.

0. Model is planar 2-D
1. Darcy's law establishes the pressure and flow
2. Advection-dispersion equation transports solute (injected once a week for 4 hours)



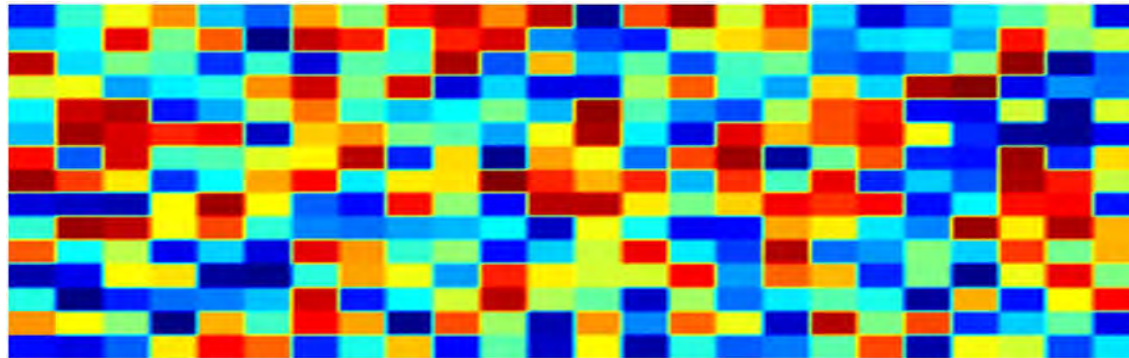
2m x 1m domains (large blocks)  
16 elements (small blocks) per domain



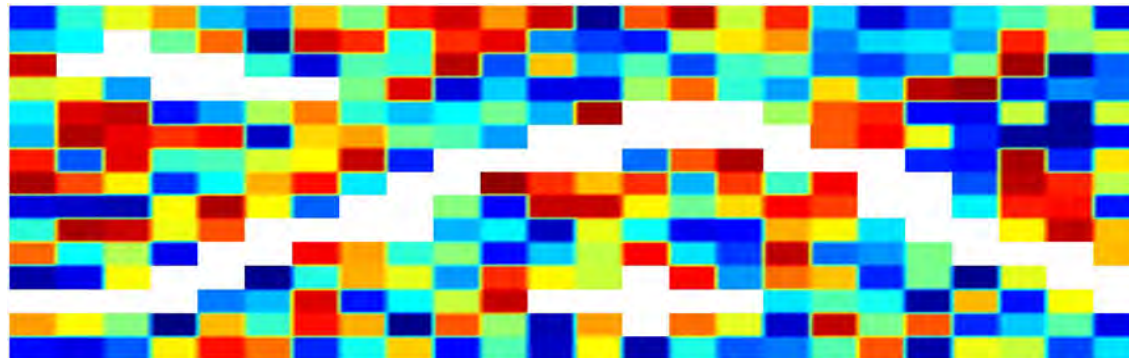


We look at two permeability distributions:

randomly assigned on [0, 100] milliDarcy



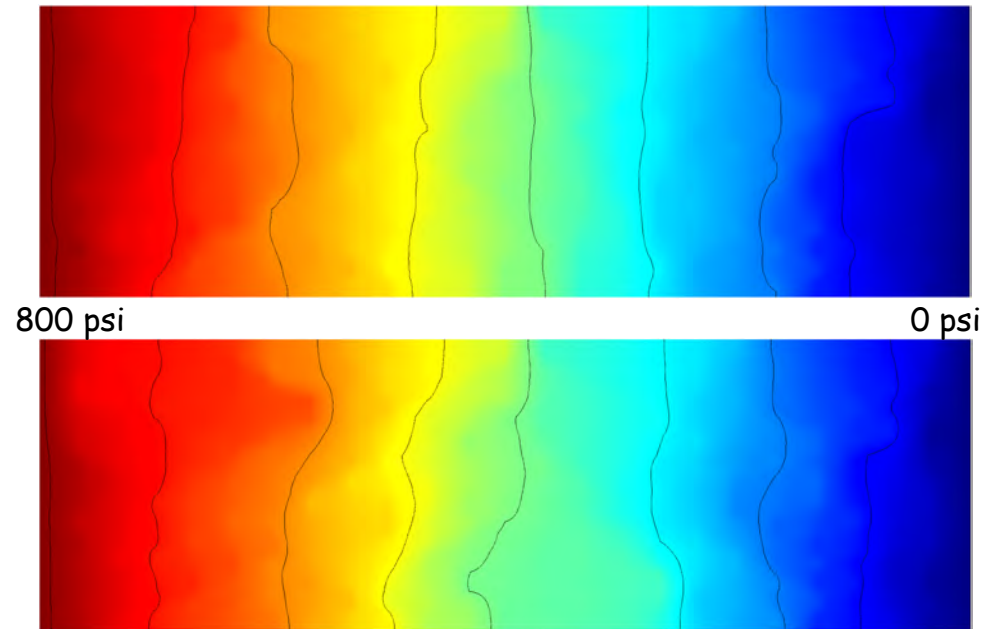
as above, with high perm zones (1 Darcy)



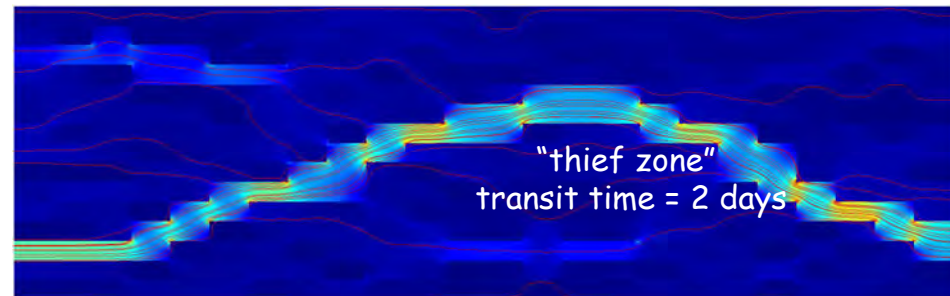
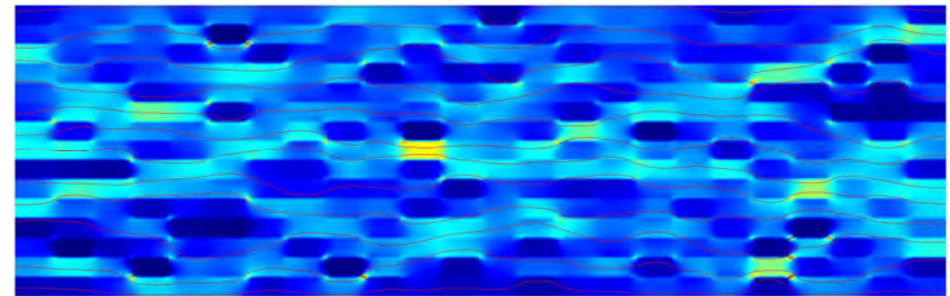


The pressure distributions  
are quite similar.

contour interval 100 psi

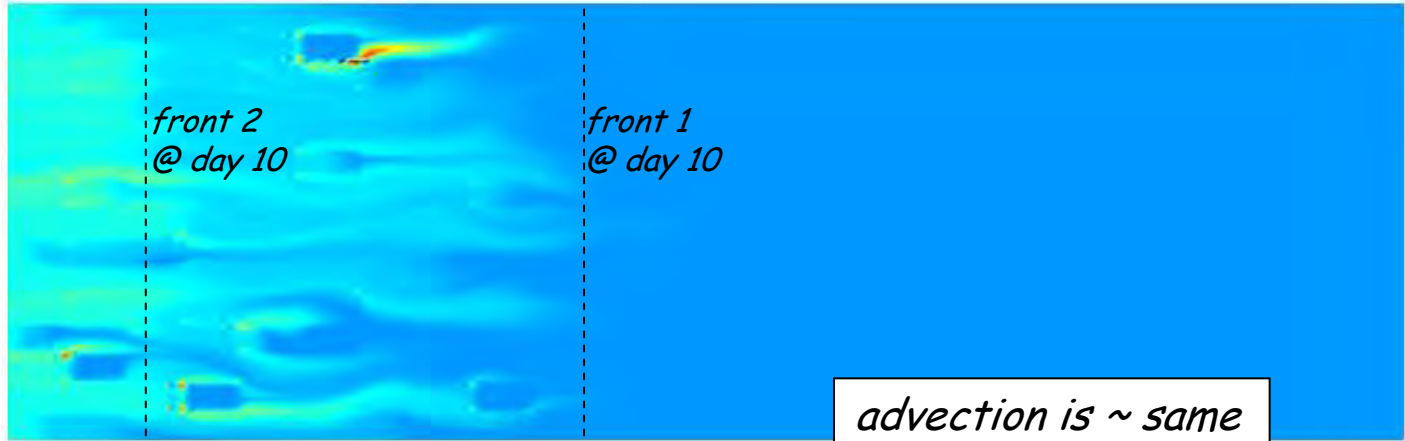


The high perm zones "attract" and  
channel the flow, as shown by  
|velocity| & streamline plots.

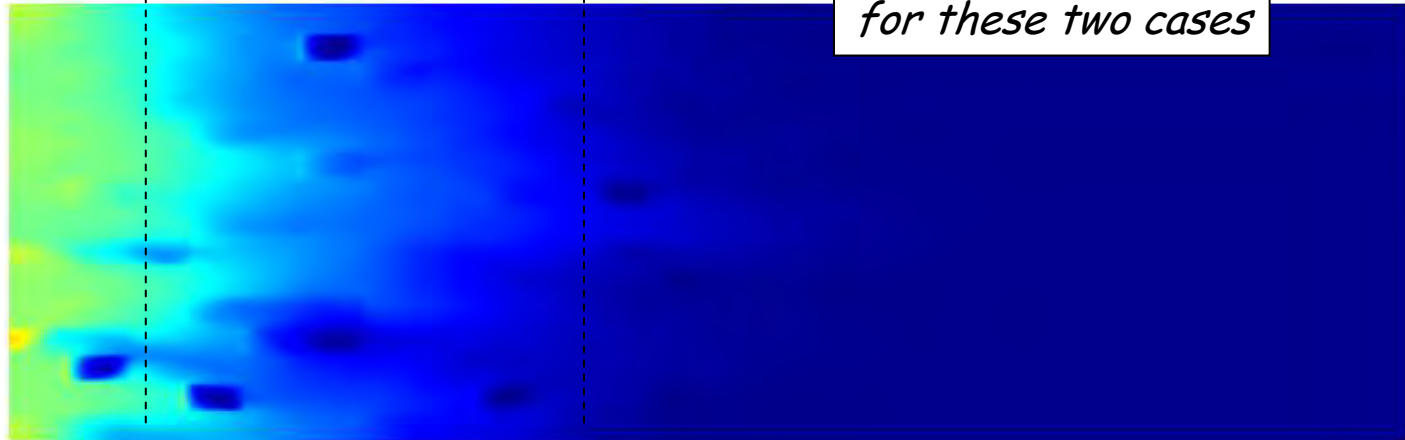


For comparison, here are the solute concentrations after ten days:

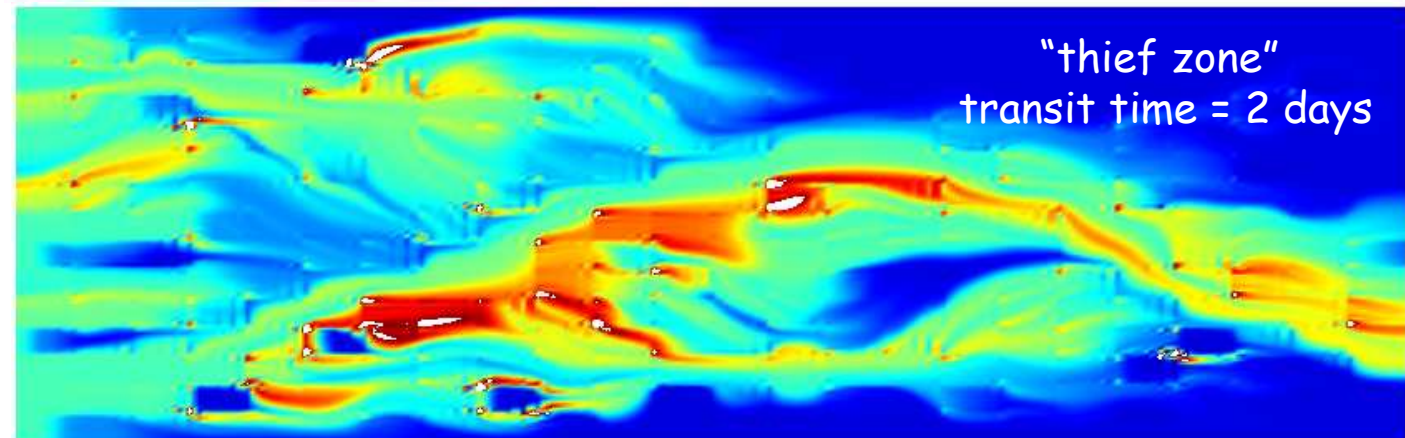
Solute dispersal for random perm, low dispersivity (0.05 m)



Solute dispersal for random perm, high dispersivity (1 m)



Solute dispersal for random perm with hi perm zones, low dispersivity (0.05 m)





## Challenges:

Many real life problems are ill defined at the outset.

Often, getting good input data is the most difficult task.

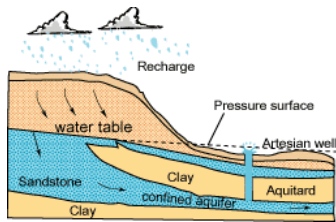
Many problems are "multiphysics."

Many systems are disordered in some sense.

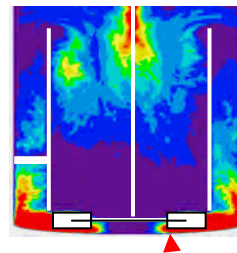
Many problems are structured across multiple scale sizes.

It is important to understand sensitivities to model parameters.

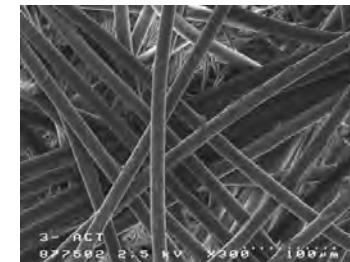
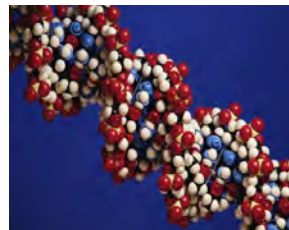
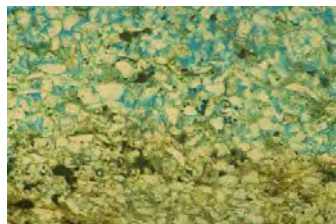
macro-scale



meso-scale



micro-scale

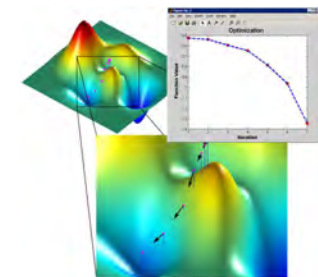
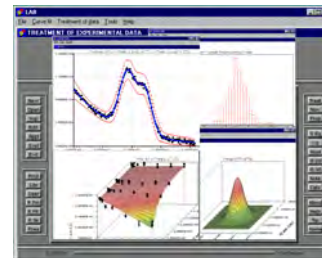
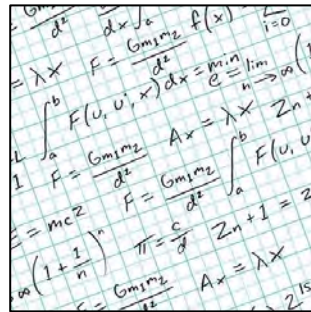


## Advice to modelers (IMHO, of course):

Use "empirical" and "fundamental" modeling early on -- when it can really make a difference:



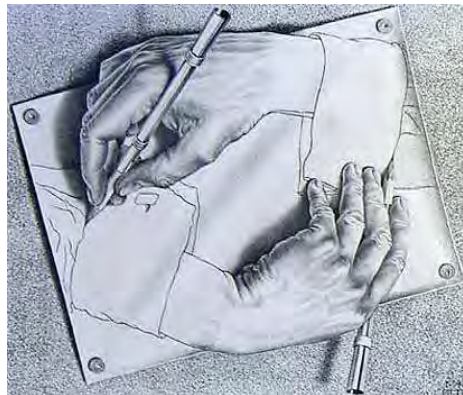
help define the problem  
guide experiments and interpret and codify data  
capture fundamental understanding  
suggest product/process improvements



Use "empirical" and "fundamental" modeling early on -- when it can really make a difference:

- help define the problem
- guide experiments and interpret and codify data
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- suggest product/process improvements

Do hand calculations. Perform a scale analysis of your problem to identify what is important and what is not. Draw things to scale.





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Do hand calculations. Perform a scale analysis of your problem to identify what is important and what is not. Draw things to scale.

Develop tools and techniques to solve funky, non-standard equations.

$$z(1-z) \frac{d^2 y}{dx^2} + [c - (\alpha + \beta + 1)z] \frac{dy}{dx} - \alpha \beta y = 0.$$



$$\frac{\partial \eta}{\partial t} = \frac{3}{2} \sqrt{\frac{g}{h}} \left( \eta \frac{\partial \eta}{\partial x} + \frac{2}{3} \frac{\partial \eta}{\partial x} + \frac{1}{3} \sigma \frac{\partial^3 \eta}{\partial x^3} \right)$$

$$\frac{d^2 U}{du^2} - [\alpha - 2q \cosh(2u)] U = 0$$

$$\frac{1}{3} u_{\tau\tau} - u_{\tau} u_{xx} - \frac{3}{2} u_x^2 u_{xx} + u_{xxxx} = 0$$

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- guide experiments and interpret and codify data
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Do hand calculations. Perform a scale analysis of your problem to identify what is important and what is not. Draw things to scale.

Have tools to solve funky, non-standard equations.

Use "multiphysics" tools which couple diverse and complex phenomena without requiring a major development effort.



*(my personal favorite, but there are others, too)*



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Have tools to solve funky, non-standard equations.

Use “multiphysics” tools which couple diverse and complex phenomena without requiring a major development effort.

**Maintain connections within the company and with vendors, academics, and government labs to gain access to the latest developments in modeling technology.**





Many insightful statements have been made about modeling:

"There is nothing more practical than a good theory."  
*primarily attributed to Kurt Lewin, but also to Maxwell, Einstein, Hilbert, ...*

"A model should be as simple as possible, but not simpler."  
*Albert Einstein*

"All models are wrong, but some are useful."  
*George Box*

"The purpose of computing is insight, not numbers."  
*Richard Hamming*

"But at some point you need numbers, too."  
*Rick Nopper*