

Multi-Phase Flow Petrophysics: A Physics-Based Approach for Deriving Sw from NMR and Formation Sampling in Low Resistivity Pay Carbonates*

Nicolas Leseur¹ and Charles Smart¹

Search and Discovery Article #42547 (2020)**

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Abstract

In conventional petrophysics, resistivity logs are the main pay zone identifiers due to resistivity contrast between hydrocarbon and formation water. If, however, a pay zone exhibits saturation-dependent Archie exponents or conductive minerals, these logs become incapable of identifying the producing zones and providing further insights about water mobility. Because of these limitations, many potentially productive zones with high irreducible water saturations are overlooked in many fields in the Middle East and other fields around the world. In an attempt to leverage the best of both worlds, the multi-method presented in this article introduces an integration of physics-based and data-driven approaches to de-risk and quantify the initial production performance of low resistivity fine-grained carbonate formations.

To begin with, the pore architecture of these rocks is derived from NMR and automated MICP deconvolution. Reservoir fluid properties and initial fractional flows (fw) are then measured through formation testing and sampling. Then, the knowledge of pore architecture is propagated to data-scarce intervals and other wells by means of probabilistic machine-learning. Initial water saturation is subsequently calculated on the basis of the equilibrium between buoyancy and capillary forces after what, the Buckley-Leverett formalism is used to derive a formation-testing-calibrated fw continuous log, thus further informing Sw and Free-Water Level (FWL) elevation. The aforementioned multi-phase flow petrophysics method provides the practitioner with a number of critical reservoir insights unrivaled by single-tool or data-driven-only approaches. Probability distribution function of Sw, initial fw, transition zone evaluation and FWL identification are among the key outputs informing the final choice of completion strategy. The framework introduced here also enables the pore architecture understanding, constructed and ground-truthed at the well level, to be exported and scaled to 3D reservoir models.

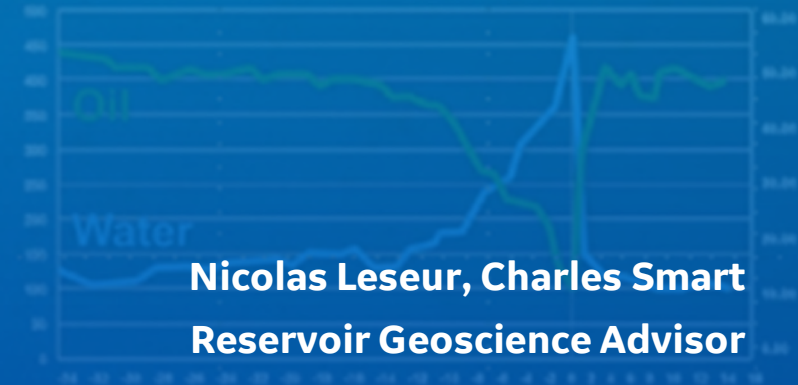


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Advancing the World of Petroleum Geosciences™

Multi-Phase Flow Petrophysics

A Physics-based approach for deriving S_w from NMR and Formation Sampling in low-resistivity pay Carbonates.



Nicolas Leseur, Charles Smart
Reservoir Geoscience Advisor

Setting the Scene

Problem Statement and Methodology

○ Micro-crystalline Calcite in the Arabian Gulf

High Salinity + High Capillarity + Surface Conductance = Excess Conductivity

○ A complementary range of solutions

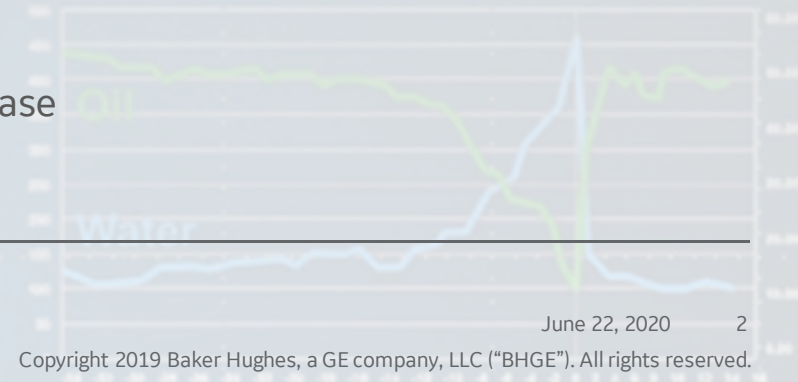
- Correcting Resistivity (Resistivity modeling, Apparent n , ...)
- Alternate Measurements (Dean-Stark, P_c , Sigma, Azimuthal Resistivity, DWT NMR, ...)
- Multi-Physics (multi-mineral, multi-phase, ...)

○ From T2 Distribution to Sw Distribution

- Reservoir Logging – Informing LWD logs with Contextual Data

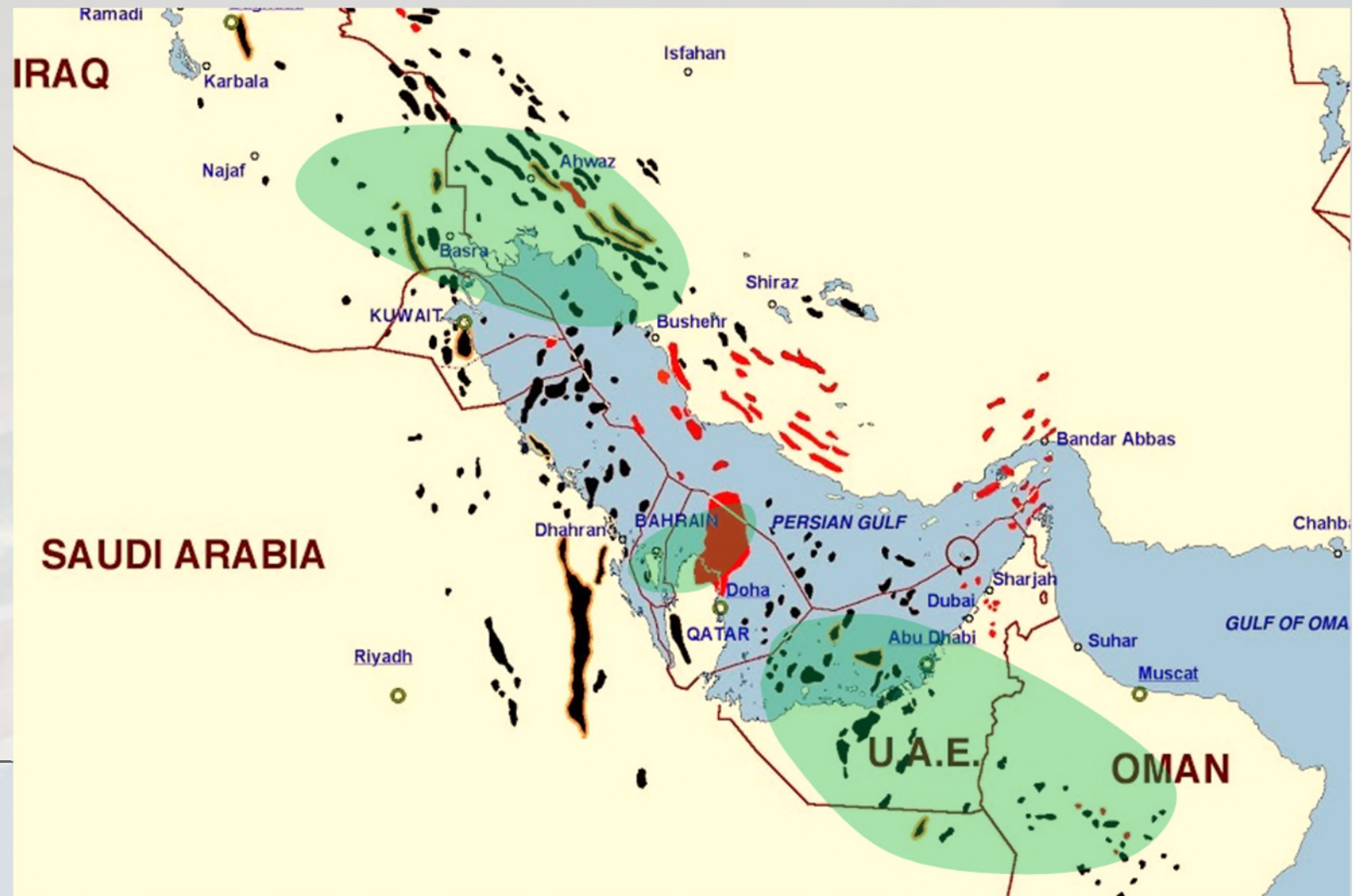
○ Multi-phase flow Petrophysics

- De-risk fluid typing and compartmentalization, calibrating multi-phase producibility and permeability



Carbonate LRP in the Arabian Gulf

- Oman
- UAE
- Bahrain
- Qatar
- Iran
- Iraq
- Saudi Arabia



1971


Pittman

“Large pores hold and transmit fluids whereas associated micropores may hold irreducible water. Analysis of borehole logs of micro porous carbonate rocks can result in misleadingly high calculated S_w and possibly bypassing of a potential oil or gas reservoir ”

1984

Asquith

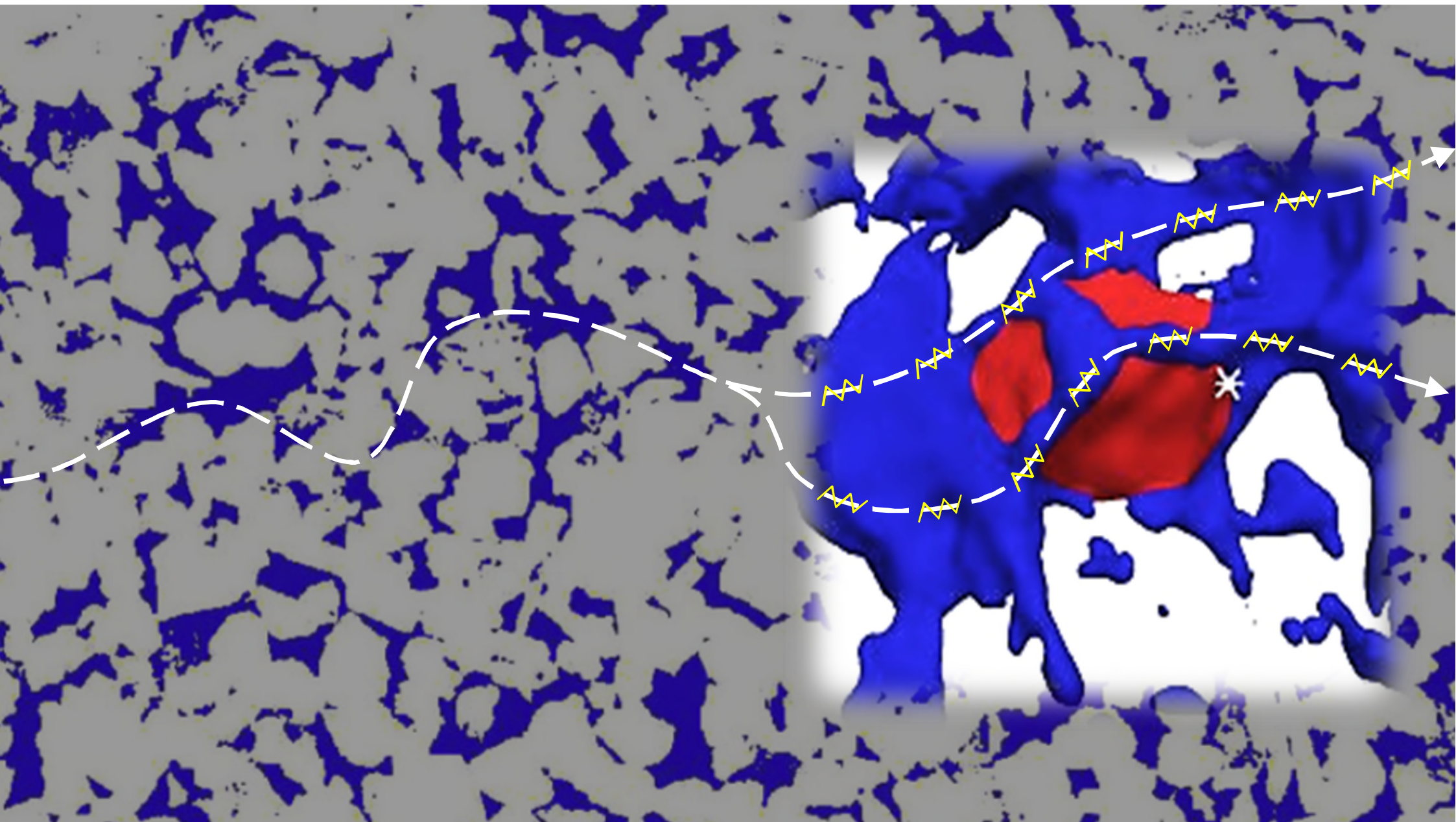
“Because of the high P_c associated with the micro porosity in ooids, hydrocarbons cannot enter the pores and so the micropores become saturated with immovable bound water. The less tortuous path of electrical flow caused by micro porosity results in the recording of abnormally low values on resistivity logs ”



Fractures Irreducible Sw Low Salinity
Pyrite
Connected
Glaucconite Vugs Saline Deep
Invasion Conductive
Mineral
Thin Micro-porous
beds lamination Cross
beddings Micro-
crystalline
High High angle
Capillarity Multi-modal
porosity Paramagnetic

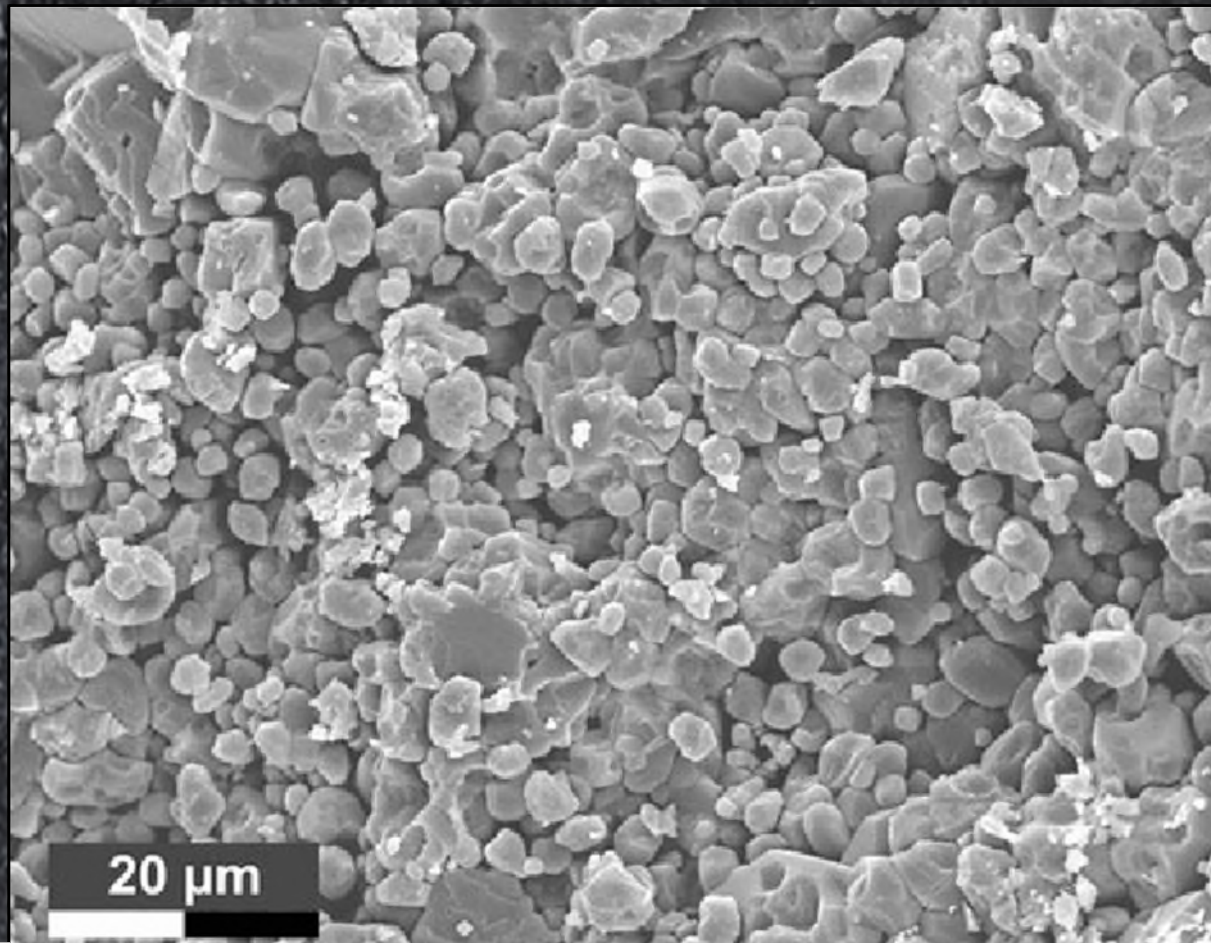


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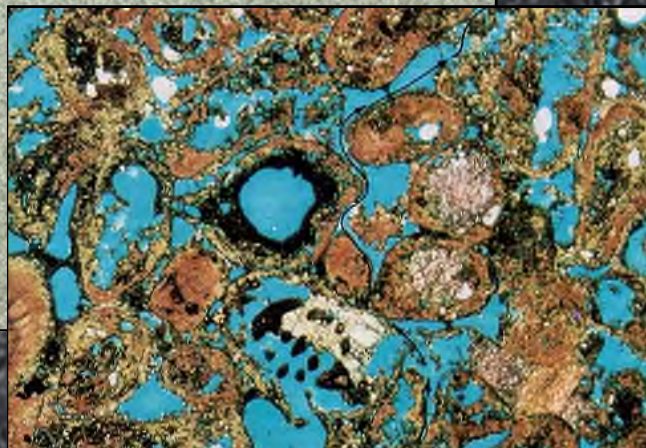
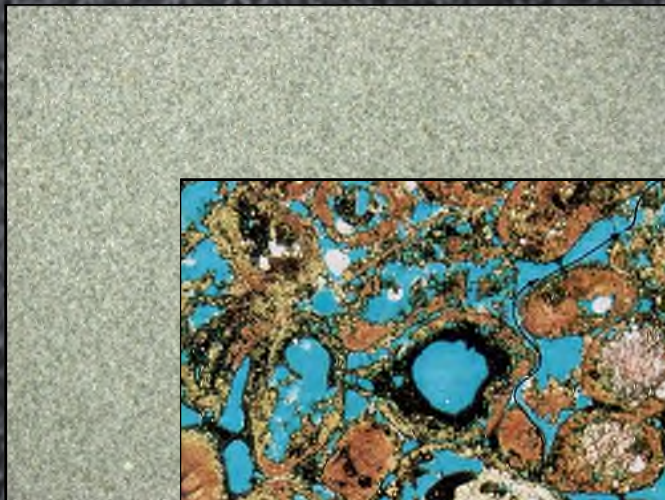


Micro-Crystalline Calcite

Mishrif, Iraq

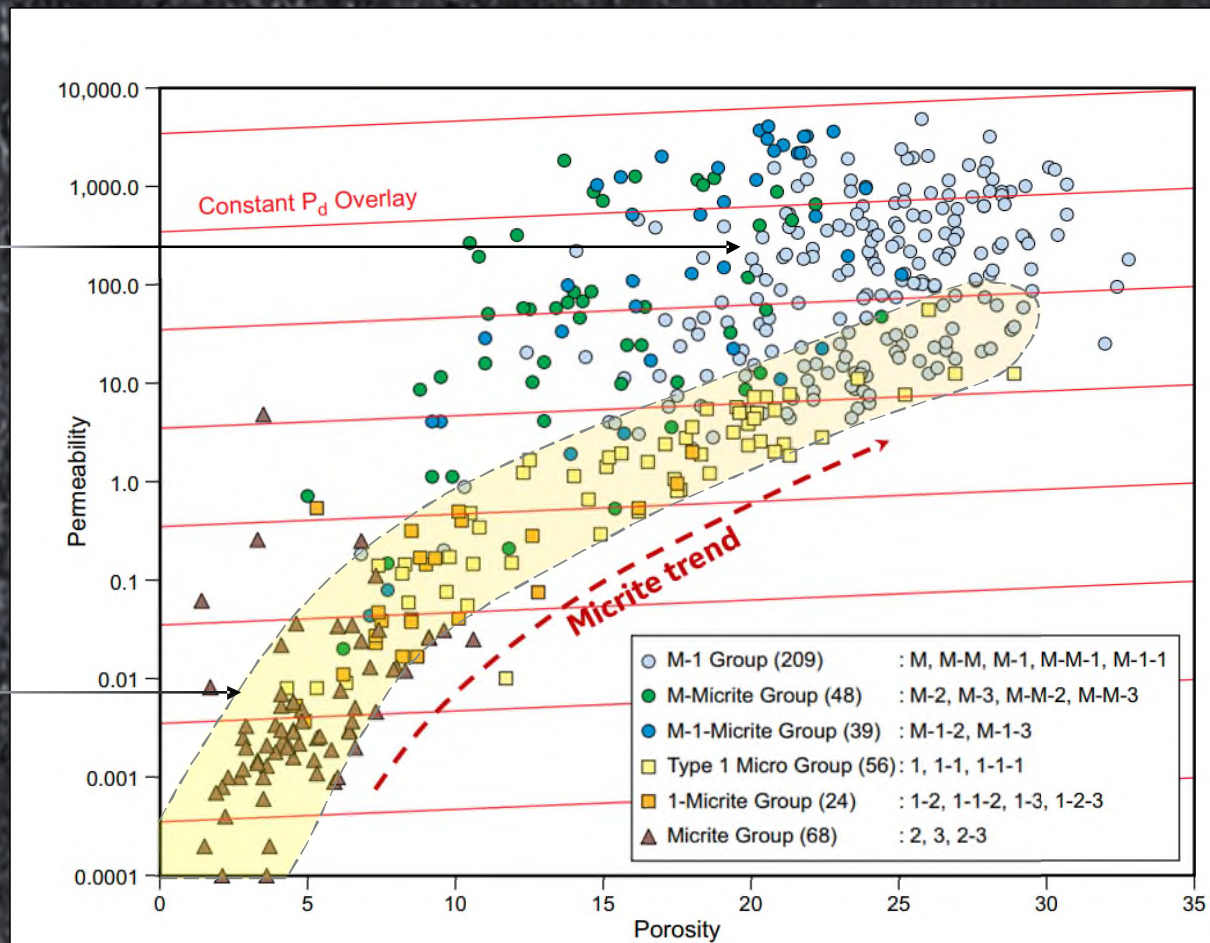


micrite

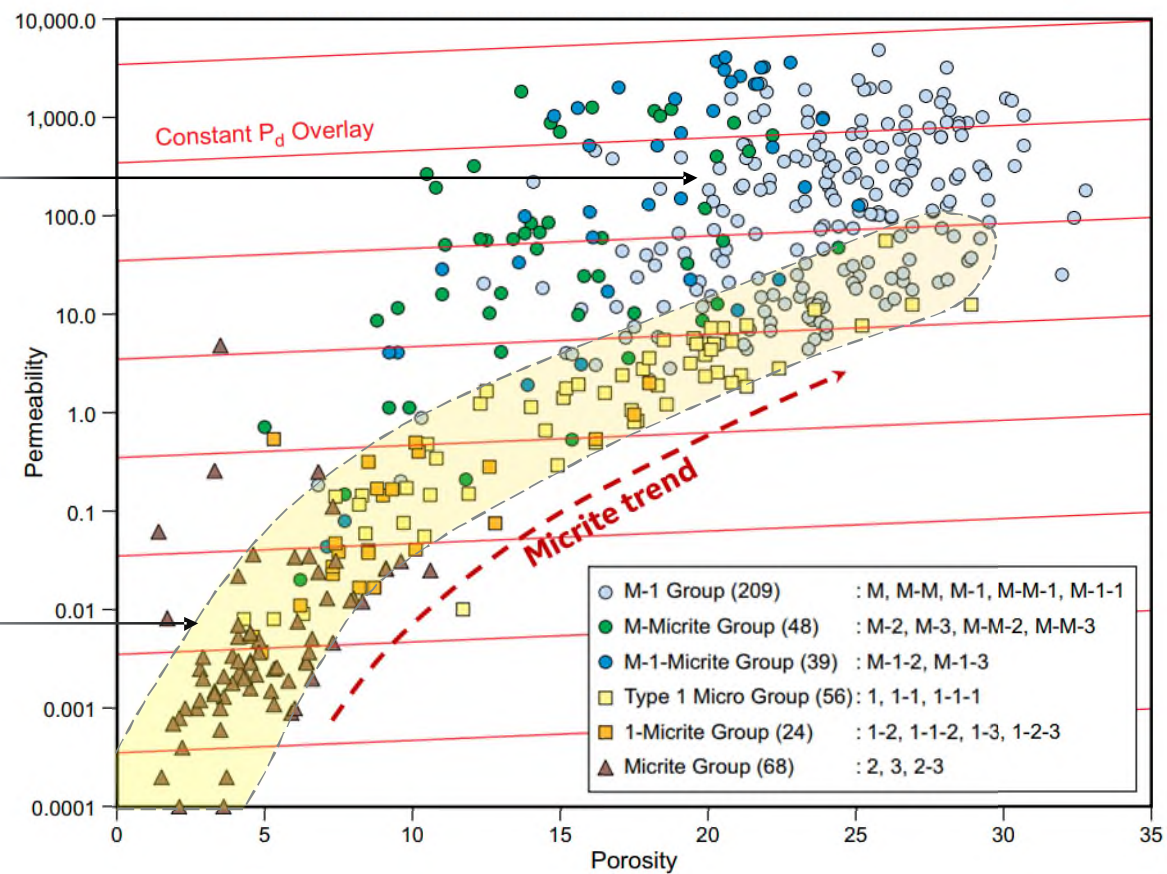
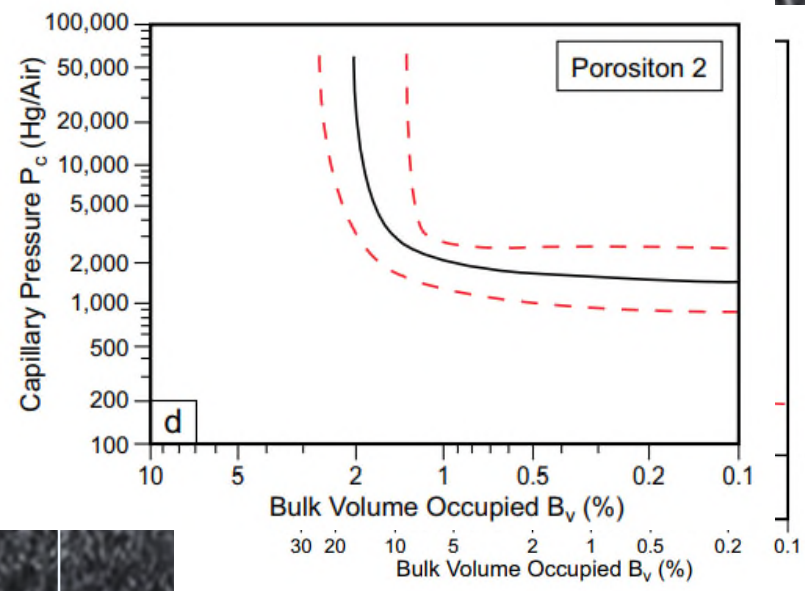


bio-clasts

Micrite, a **subset** of the
Carbonate world

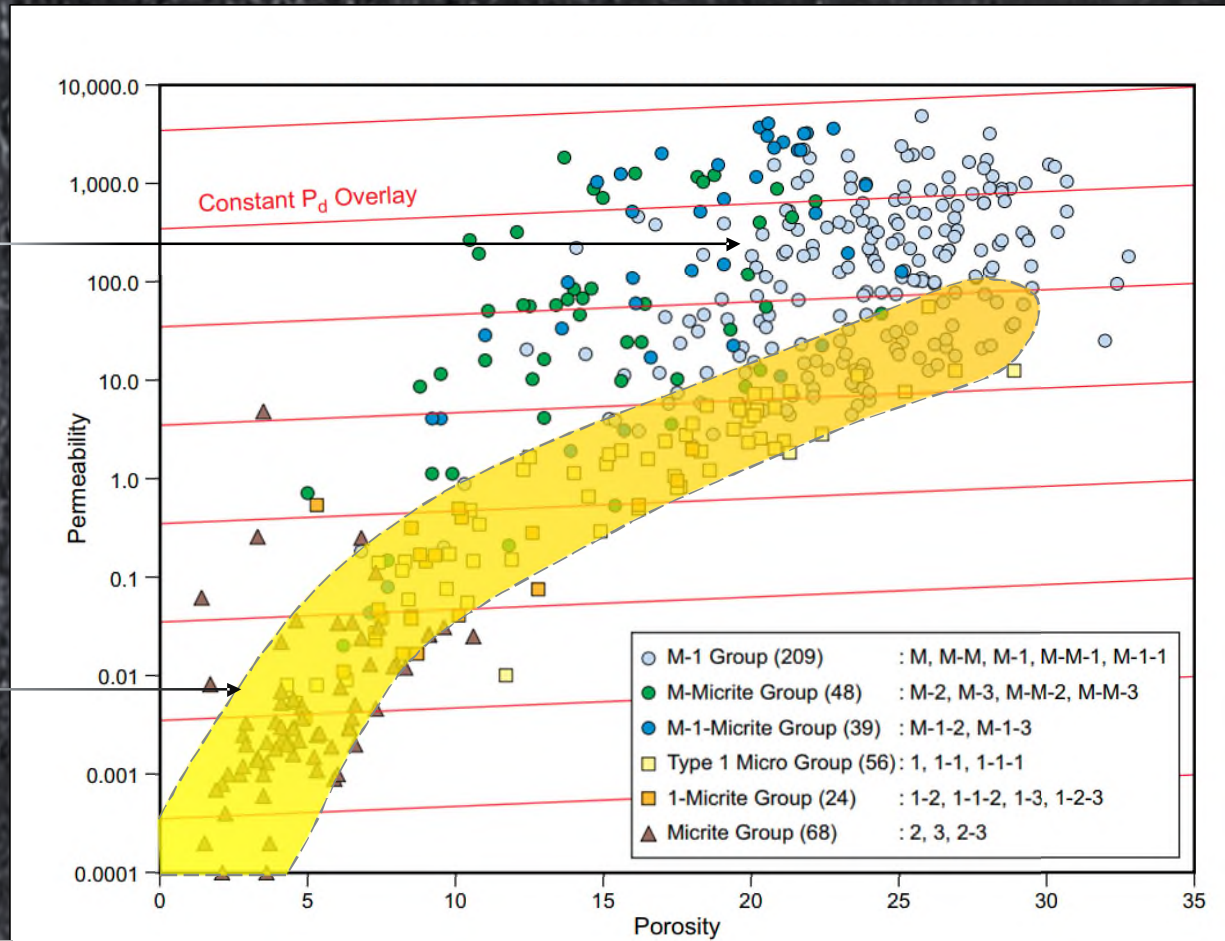
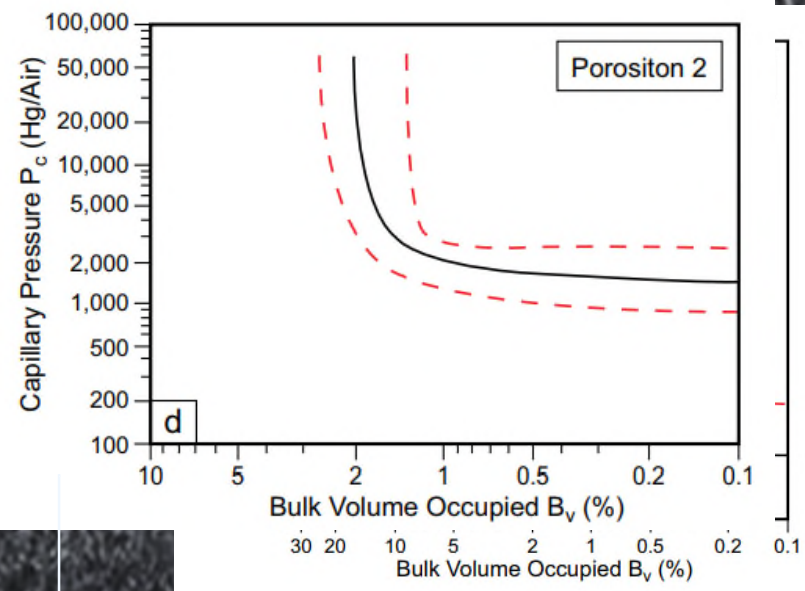


Clerke et al. (2008)



Pure micrite, is primarily
Monomodal

Clerke et al. (2008)



Pure micrite, is primarily
Monomodal

Clerke et al. (2008)

The Method

Logging only vs. Logging & Coring

Logging

NMR
Formation Testing
+
Capillary Pressure
Buckley-Leverett

Logging & Coring

MICP
Formation Testing
+
Capillary Pressure
Buckley-Leverett

The Method

Logging only vs. Logging & Coring

Logging

NMR

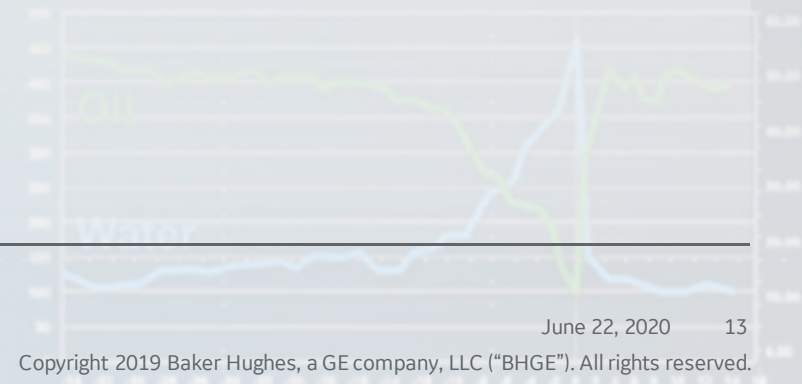
Formation Testing

+

Capillary Pressure
Buckley-Leverett

Logging & Coring

*From **T2** Distribution
to **Sw** Distribution*



The Method

Logging only vs. Logging & Coring

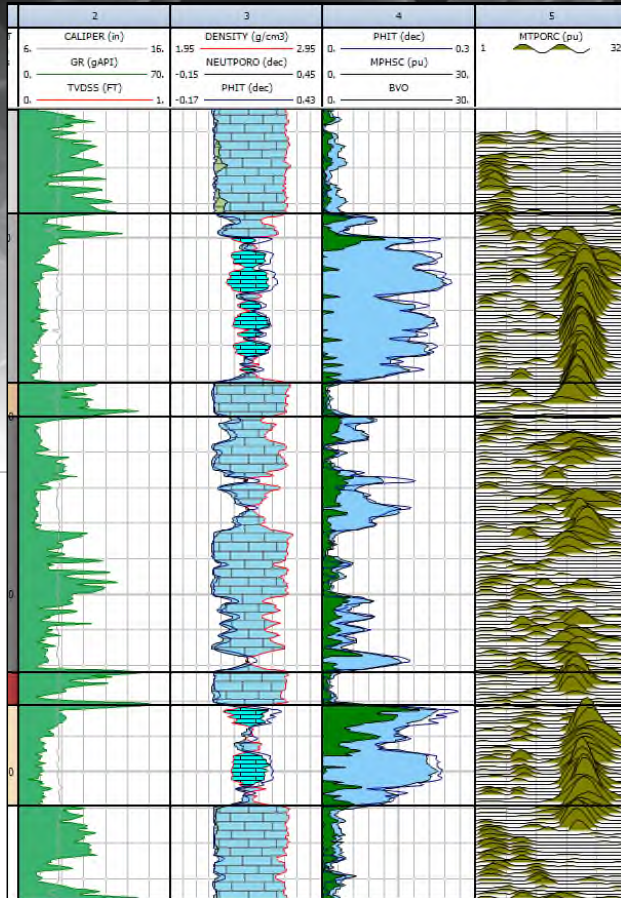
Logging

Multi-phase Flow
Petrophysics

Logging & Coring

MICP
Formation Testing
+
Capillary Pressure
Buckley-Leverett

From T2 Distribution to Sw Distribution



INPUTS

- Fluid Densities
- Free-Water Level
- Wettability

ASSUMPTIONS

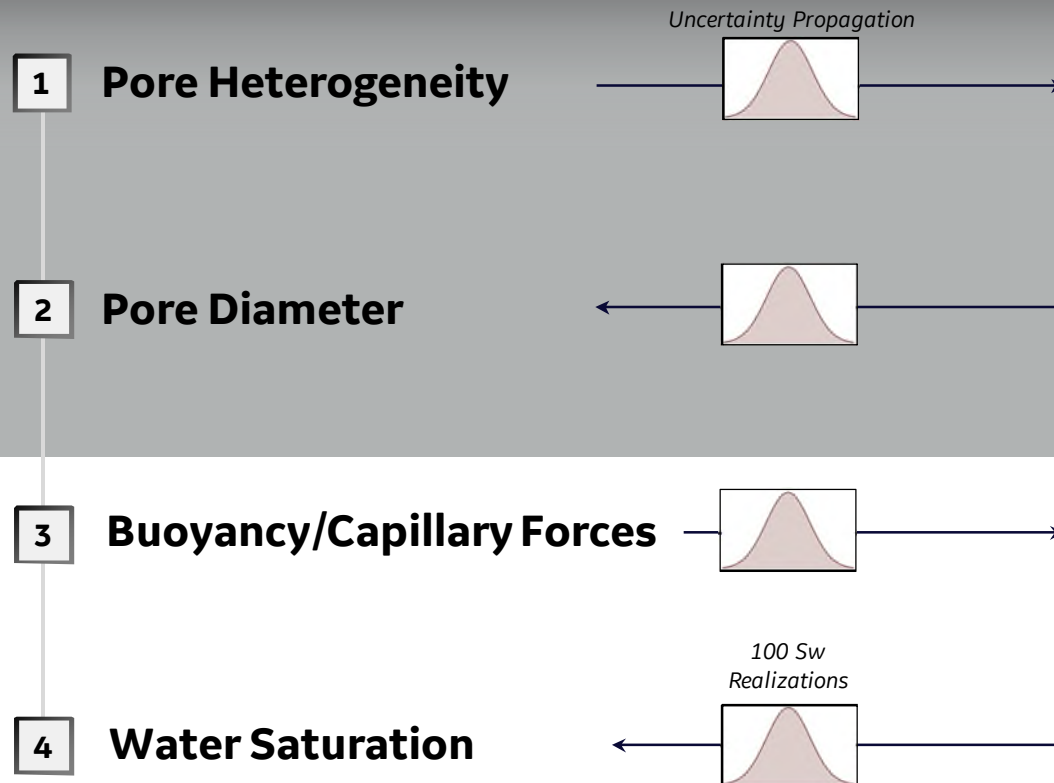
- **Monomodal Micritic Rocktype:**
Limited Pore Architecture
Heterogeneity.

** This is valid for tight carbonates with permeability < 10mD*

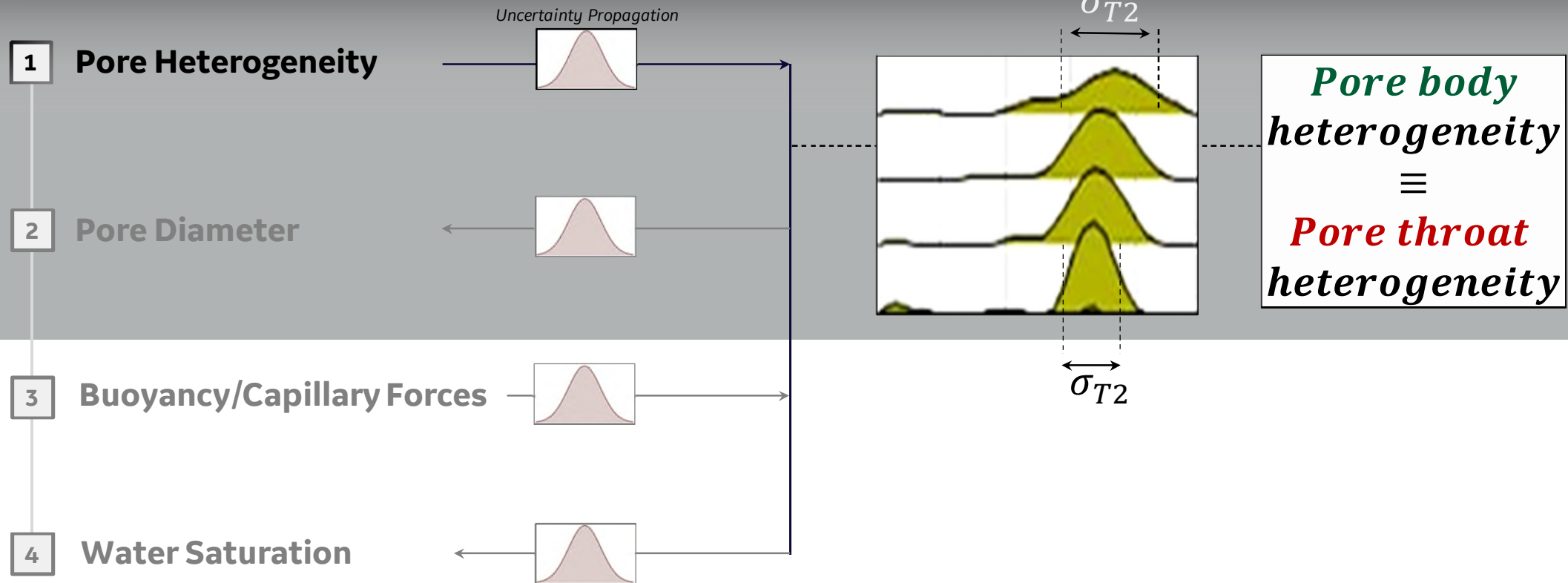
MODEL

- Poiseuille's Bundle of Tube
- Equilibrium between Capillary and Buoyancy Forces

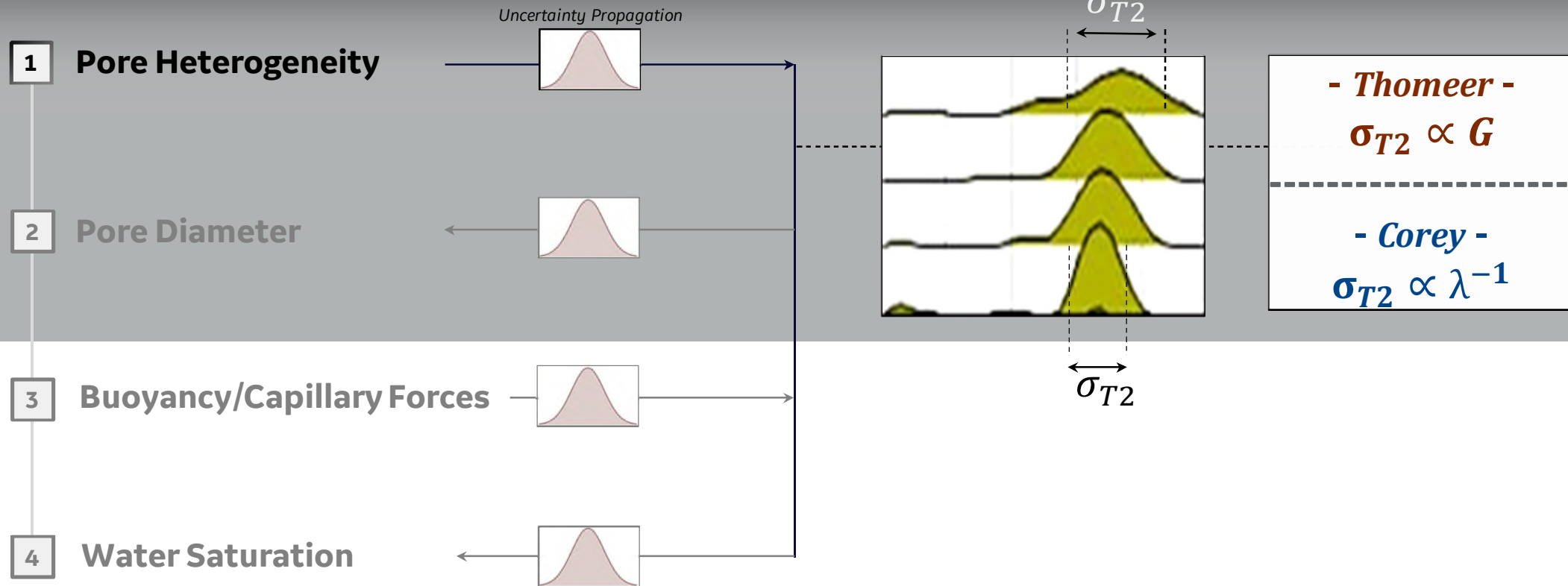
From T2 Distribution to Sw Distribution



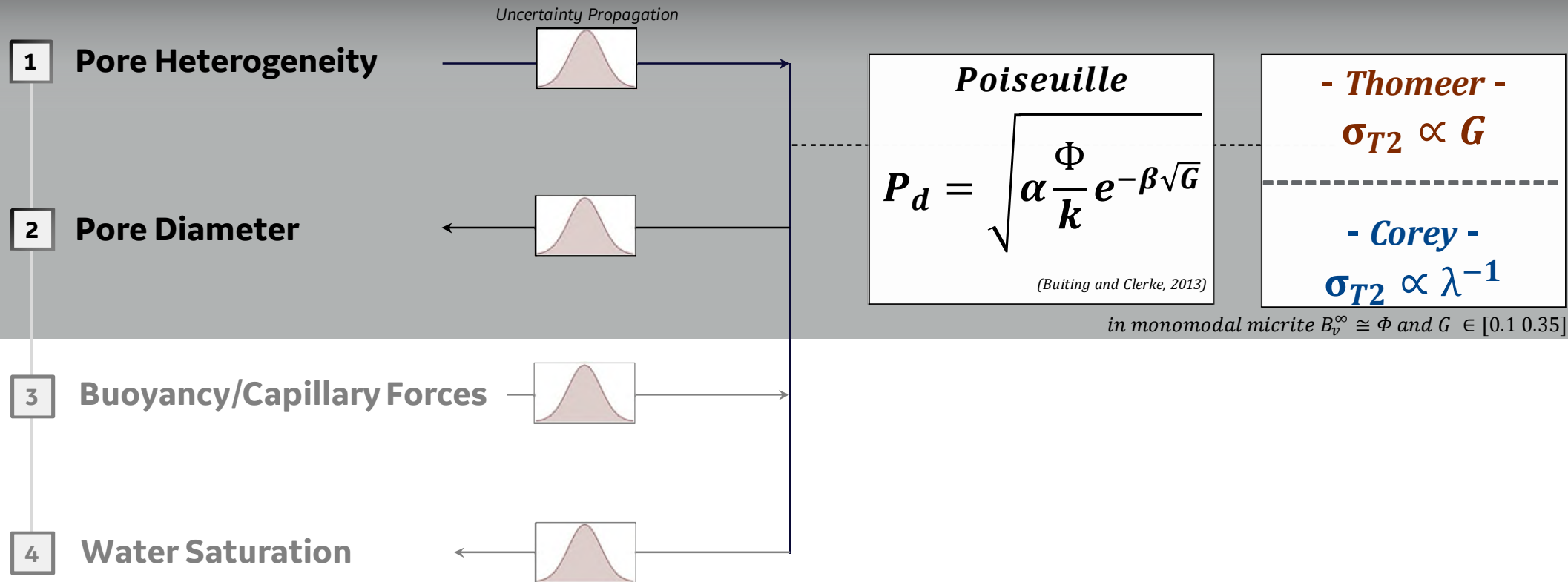
From T2 Distribution to Sw Distribution



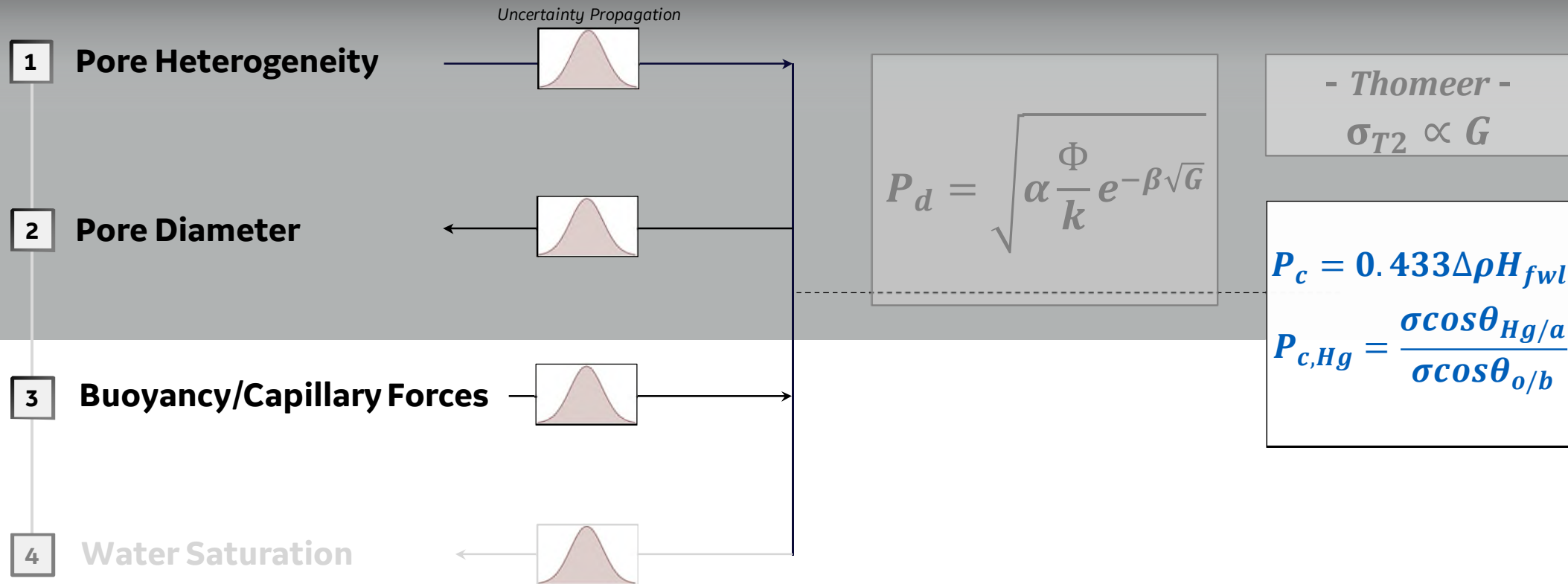
From T2 Distribution to Sw Distribution



From T2 Distribution to Sw Distribution



From T2 Distribution to Sw Distribution



From T2 Distribution to Sw Distribution

Uncertainty Propagation

1 Pore Heterogeneity



2 Pore Diameter



3 Buoyancy/Capillary Forces



4 Water Saturation



$$P_d = \sqrt{\alpha \frac{\Phi}{k} e^{-\beta \sqrt{G}}}$$

- Thomeer -
 $\sigma_{T2} \propto G$

$$P_c = 0.433 \Delta \rho H_{fwl}$$

$$P_{c,Hg} = \frac{\sigma \cos \theta_{Hg/a}}{\sigma \cos \theta_{o/b}}$$

$$S_o = \frac{B_v^\infty}{\Phi} e^{\frac{-G}{\log \frac{P_{c,Hg}}{P_d}}}$$

From T2 Distribution to Sw Distribution

Uncertainty Propagation

1 Pore Heterogeneity



2 Pore Diameter



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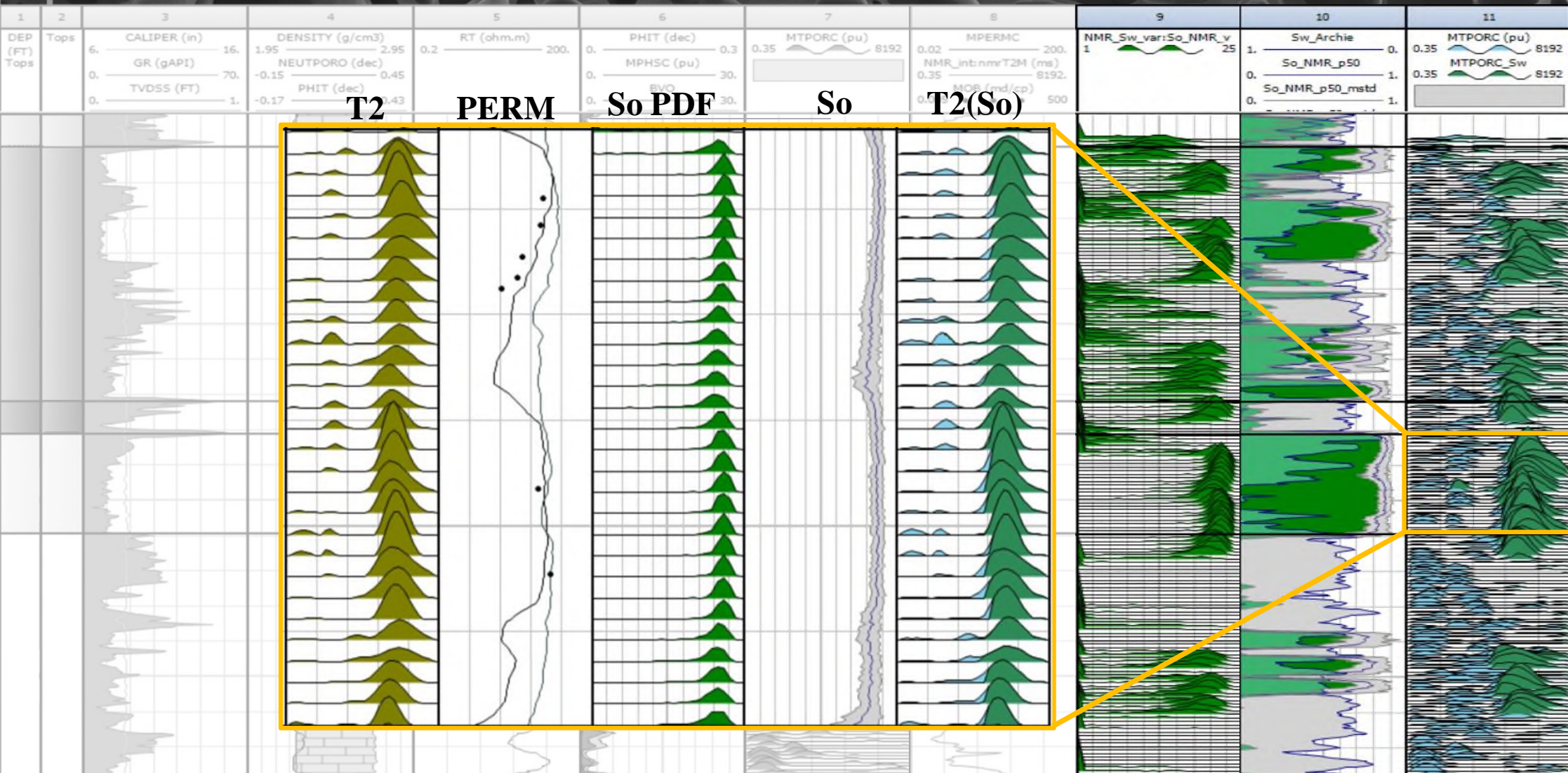
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$$S_o = \frac{B_v^\infty}{\Phi} e^{\frac{-G}{\log \frac{P_{c,Hg}}{P_d}}}$$

Uncertainty Propagation via Monte-Carlo Simulation

From T2 Distribution to Sw Distribution



The Method

Logging only vs. Logging & Coring

Logging

NMR

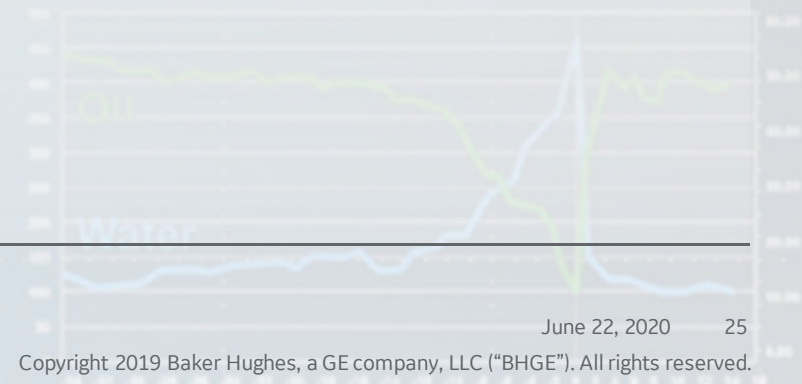
Formation Testing

+

Capillary Pressure
Buckley-Leverett

Logging & Coring

*From **T2** Distribution
to **Sw** Distribution*



The Method

Logging only vs. Logging & Coring

Logging

Multi-phase Flow
Petrophysics

Logging & Coring

MICP
Formation Testing
+
Capillary Pressure
Buckley-Leverett

MultiPhase Flow Petrophysics

A 3-step Approach

**Machine
Learning**

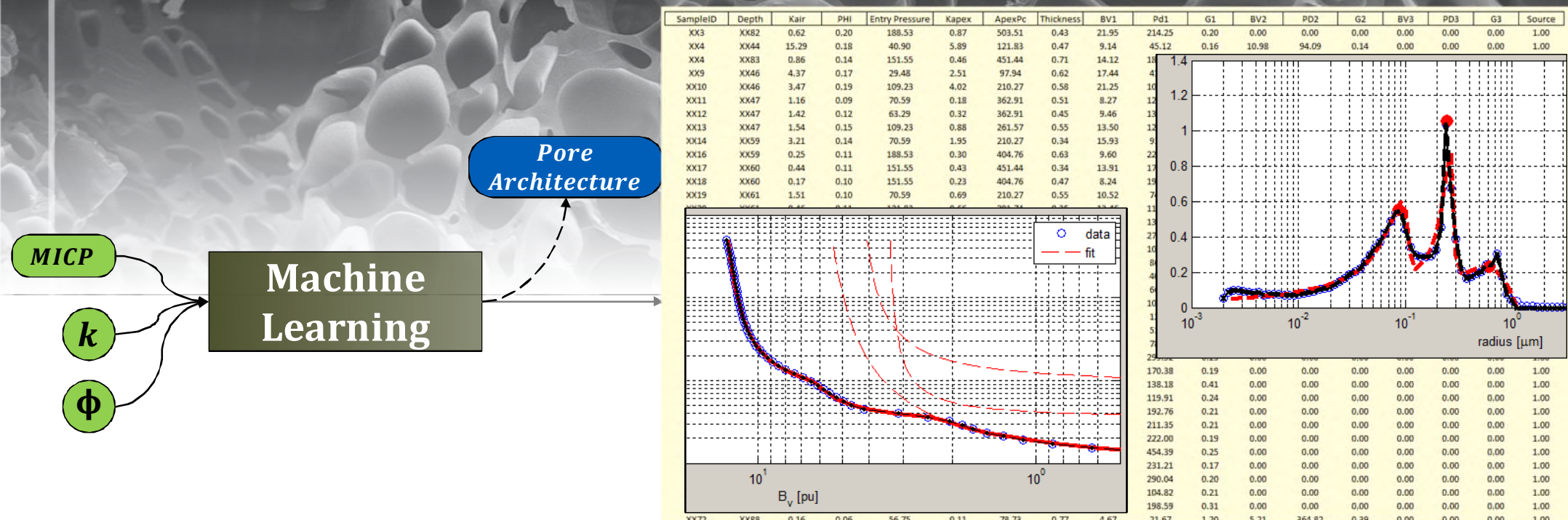
**SatH. Function
& Land**

**Buckley
Leverett**

MultiPhase Flow Petrophysics

Pore Architecture Modeling

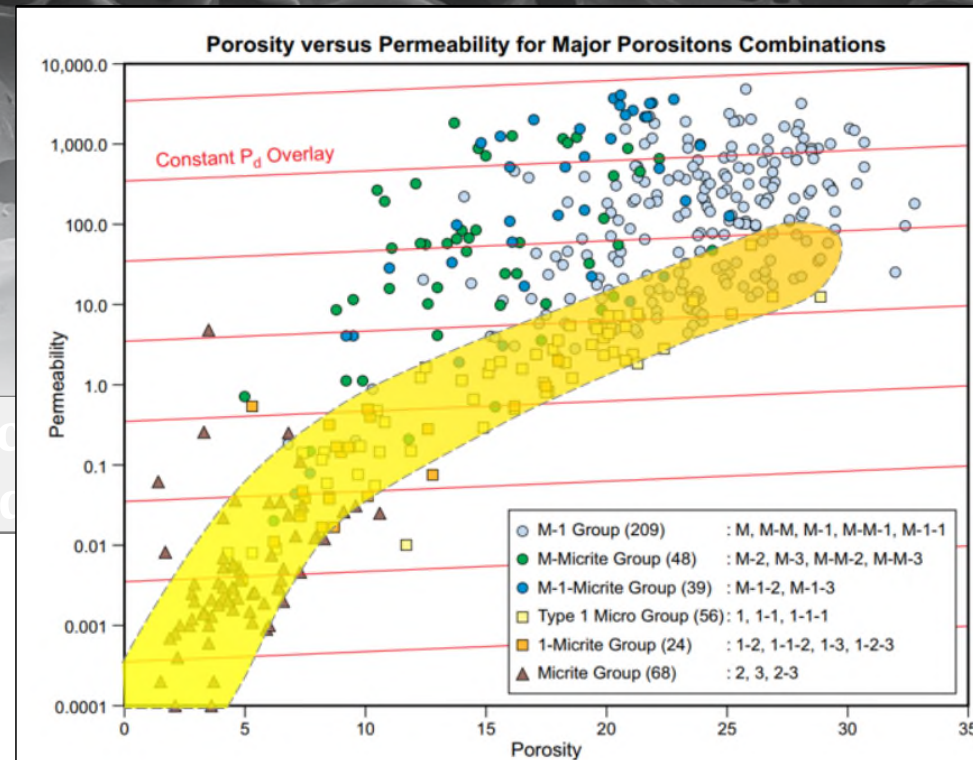
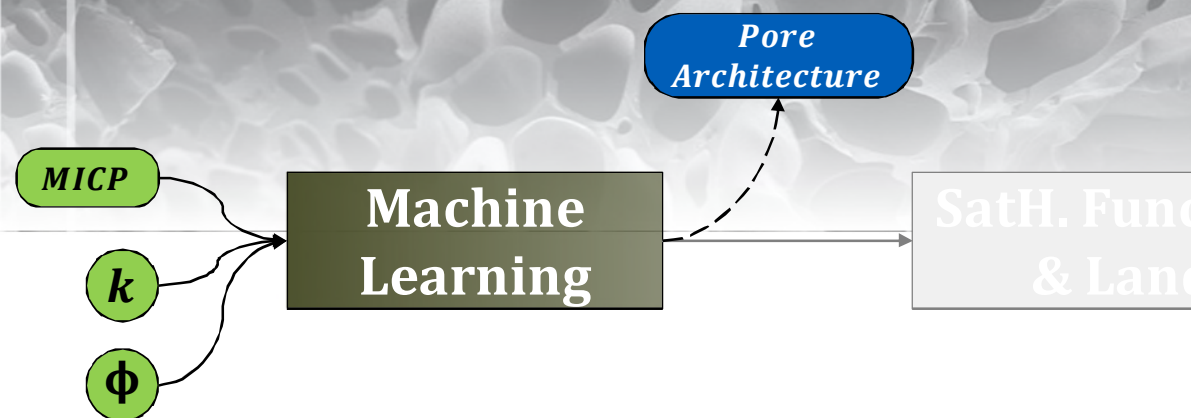
Automated Type Curve Fitting



Based on global non-linear optimization
1875 samples in 10 minutes

MultiPhase Flow Petrophysics

Pore Architecture Modeling



The $k - \phi$ relationship of Tight monomodal, micritic Carbonates is generally well behaved

MultiPhase Flow Petrophysics

Pore Architecture Modeling

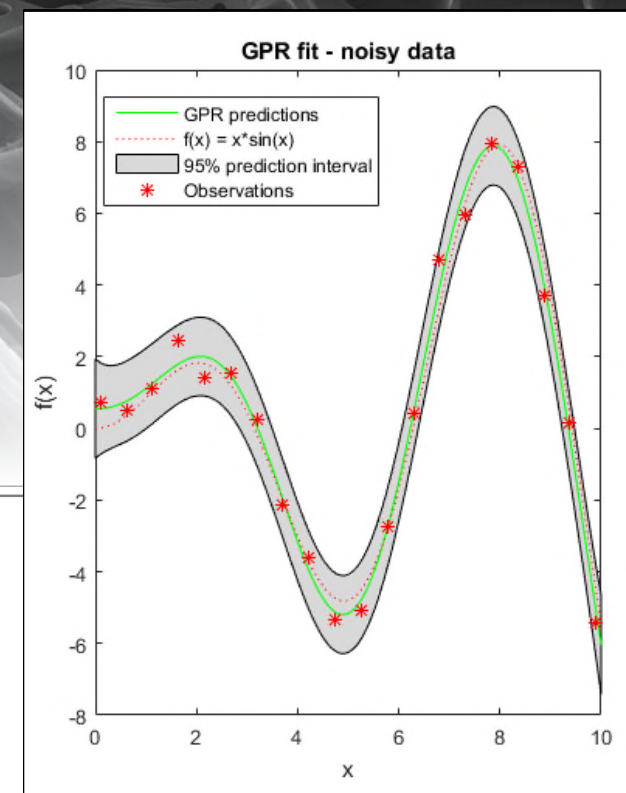
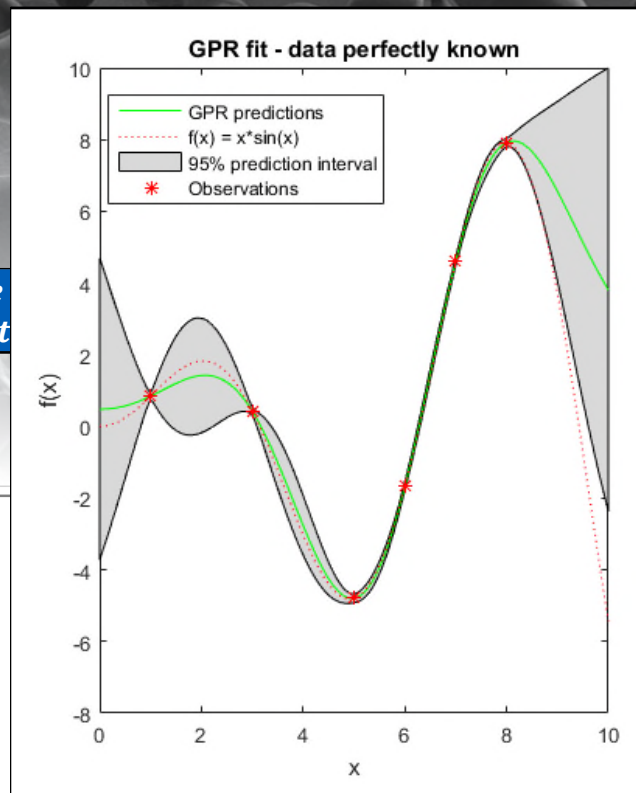
MICP

k

ϕ

Machine Learning

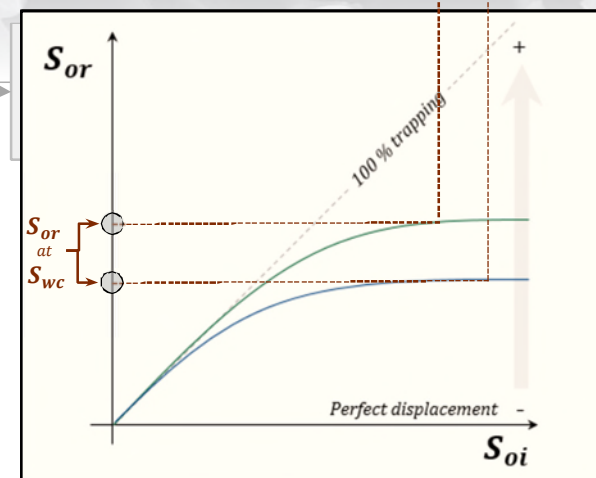
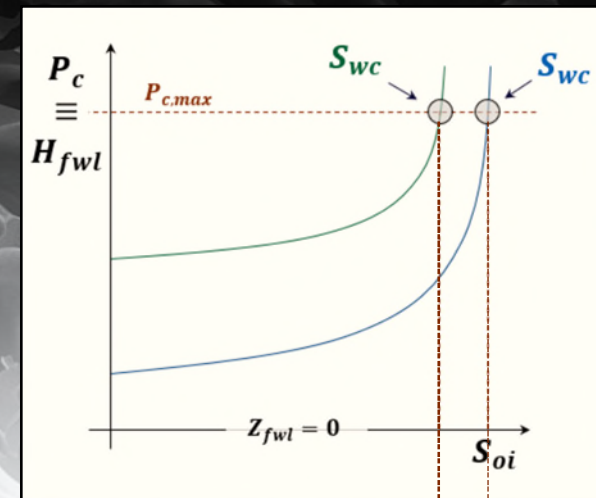
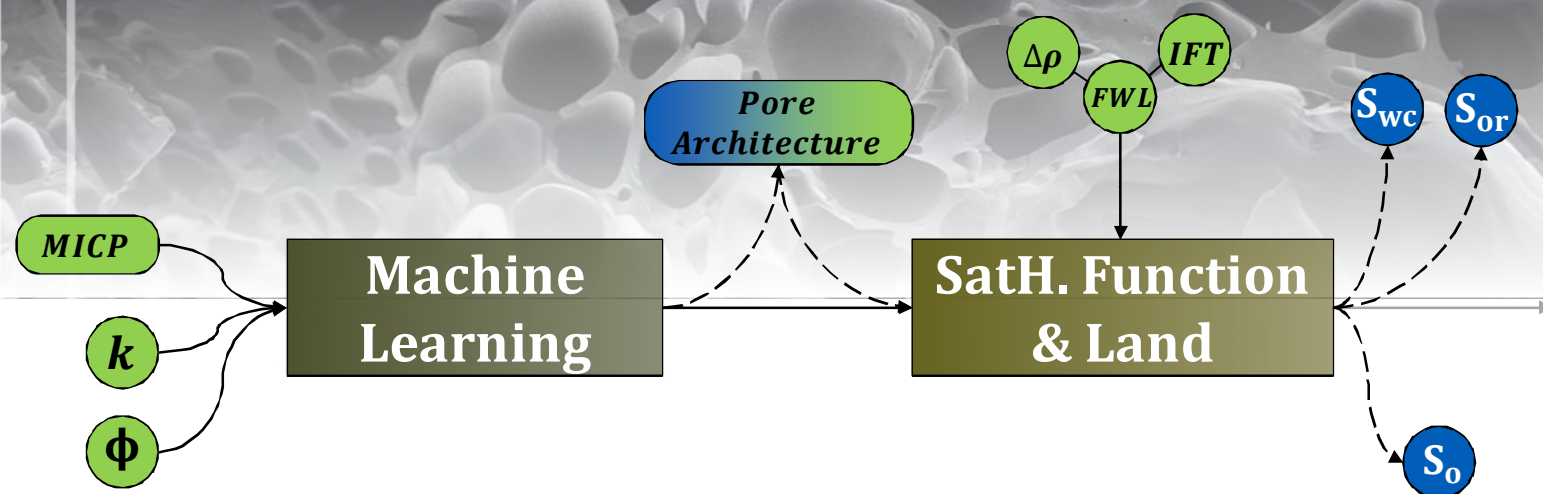
Pore Architecture



Probabilistic Machine Learning approach
Includes measurement uncertainty

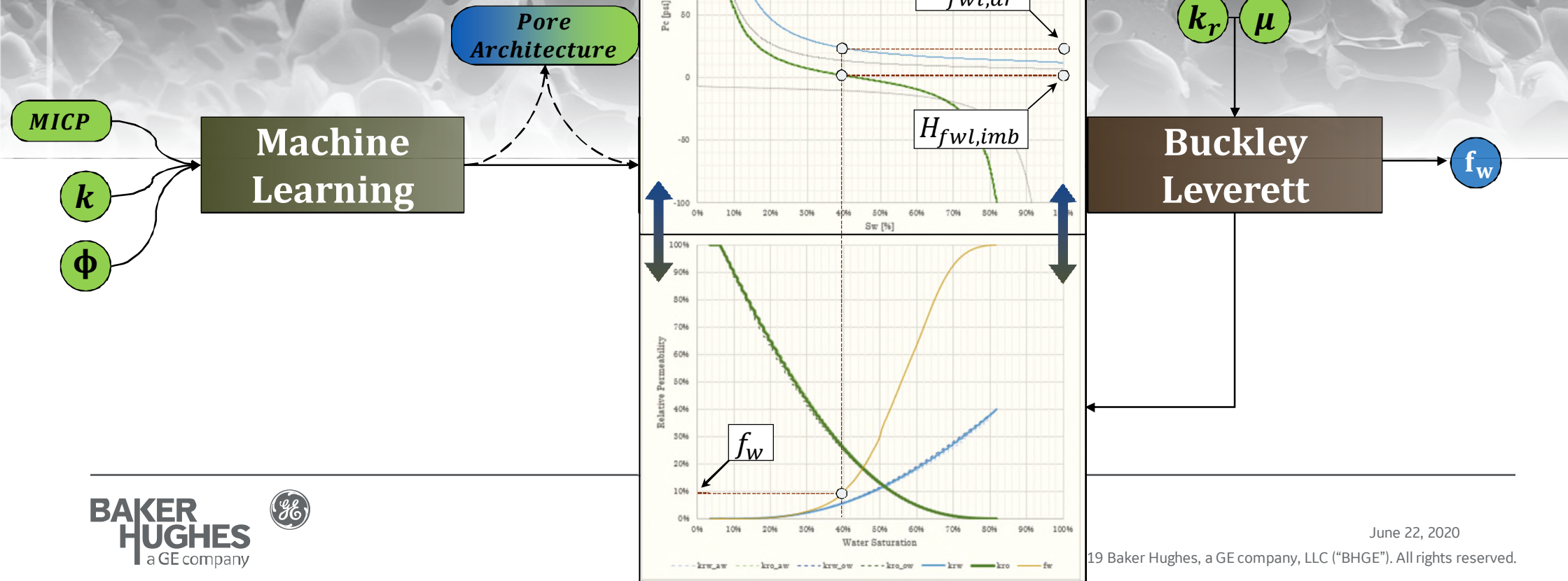
MultiPhase Flow Petrophysics

Saturation Modeling



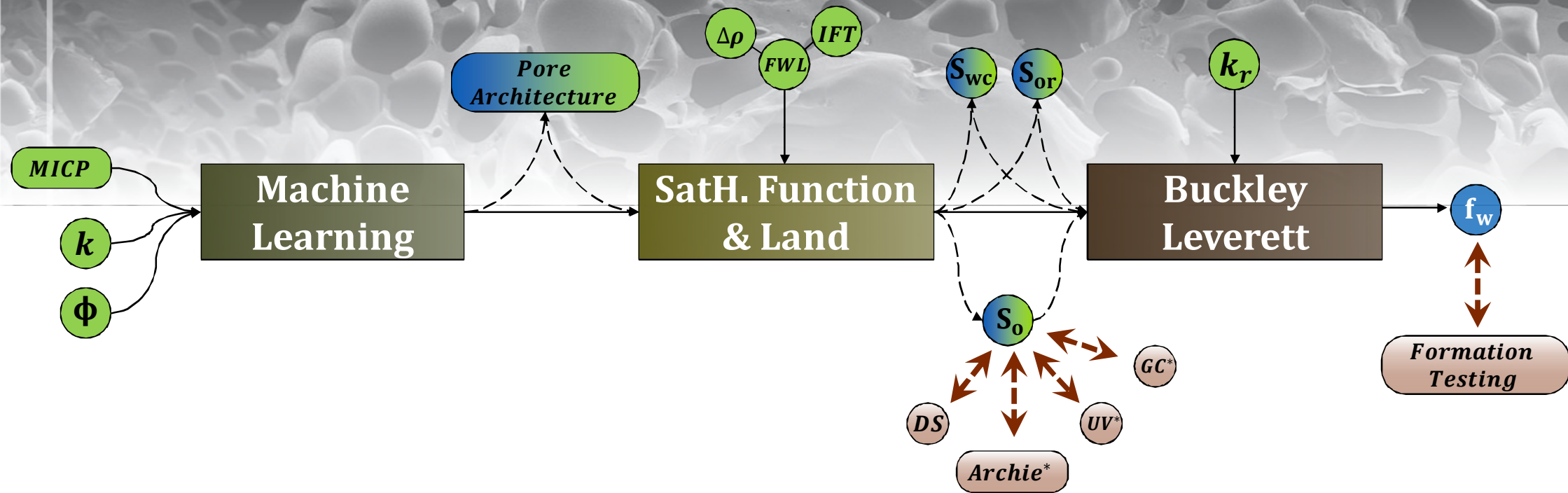
MultiPhase Flow Petrophysics

Fractional Flow Modeling



MultiPhase Flow Petrophysics

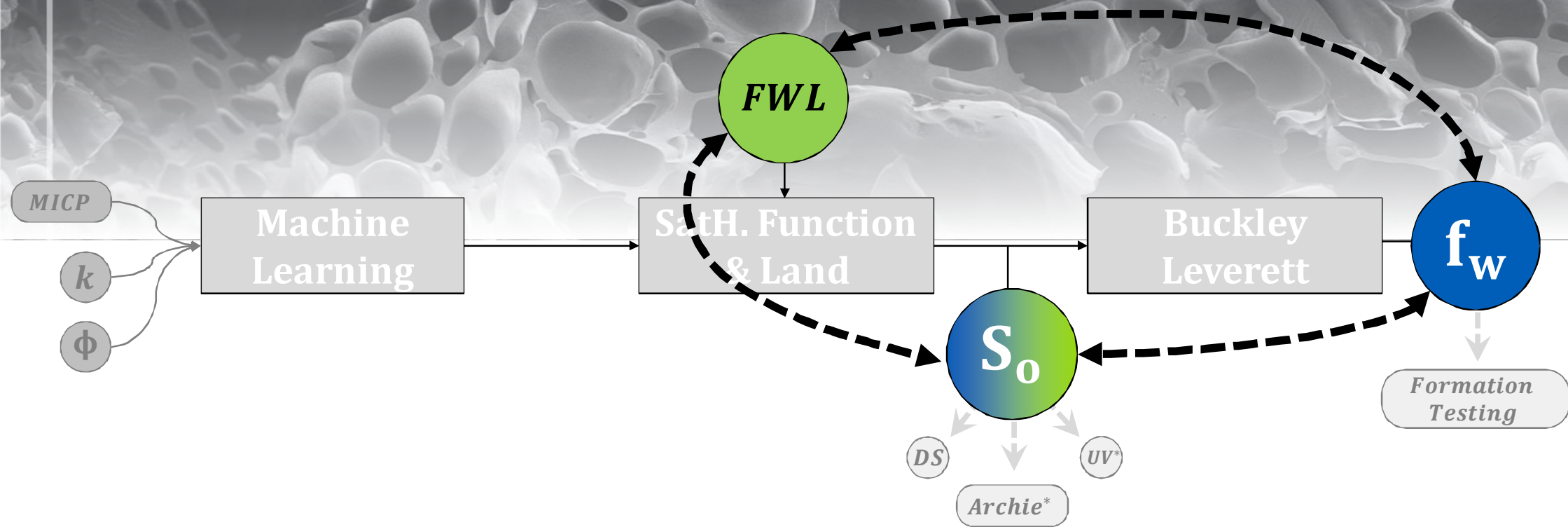
Fractional Flow Modeling



* qualitative

MultiPhase Flow Petrophysics

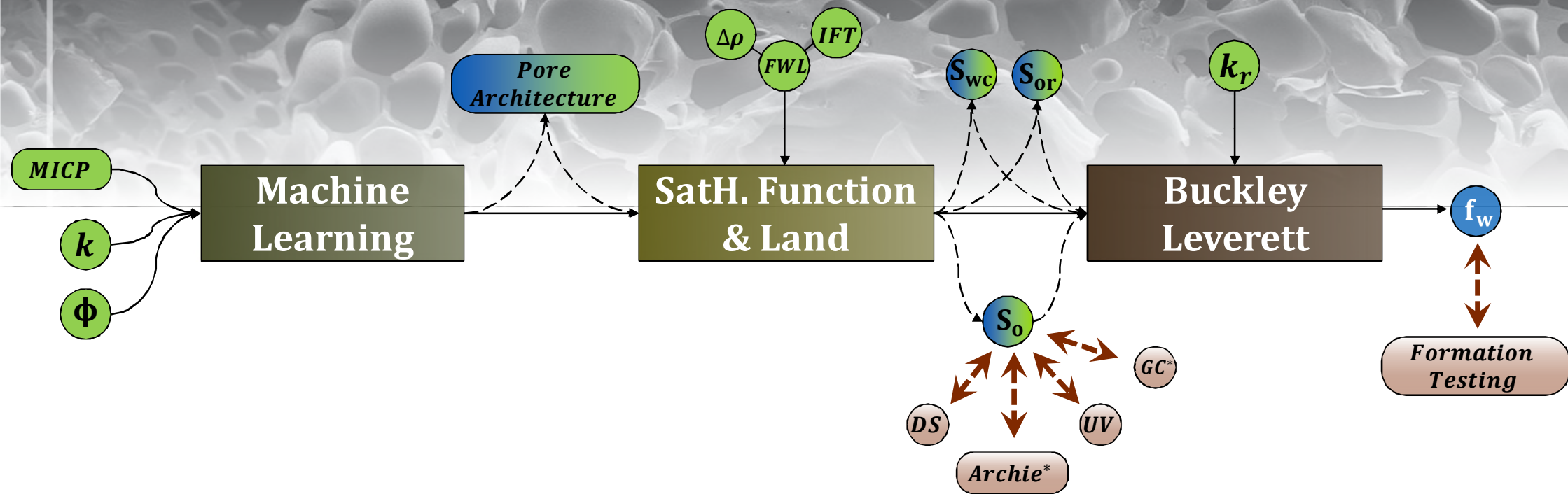
From Pore Architecture to Fractional Flow



* qualitative

MultiPhase Flow Petrophysics

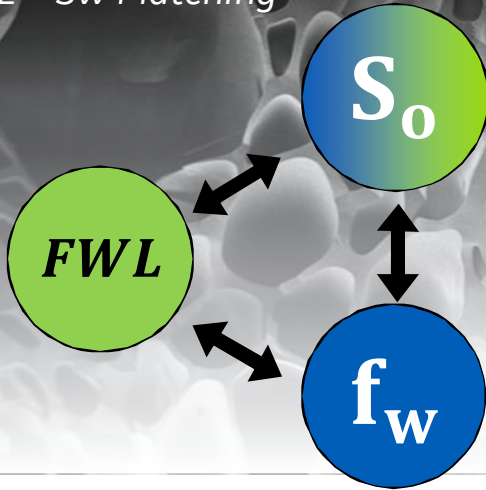
From Pore Architecture to Fractional Flow



* qualitative

MultiPhase Flow Petrophysics

$F_w - FWL - S_w$ Matching



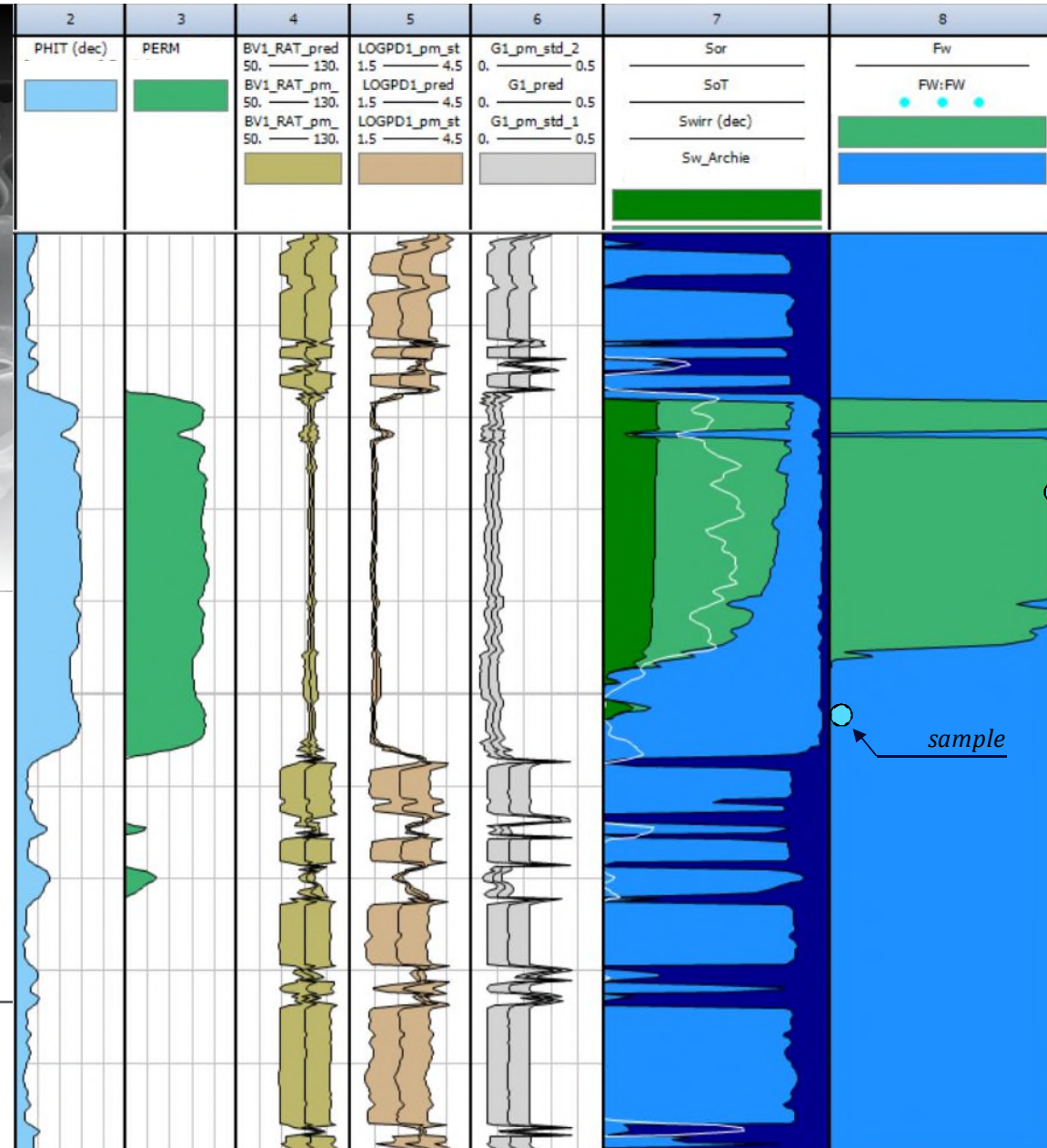
Optimization under Constraint

CASE #1

☐ Match the F_w from formation sampling

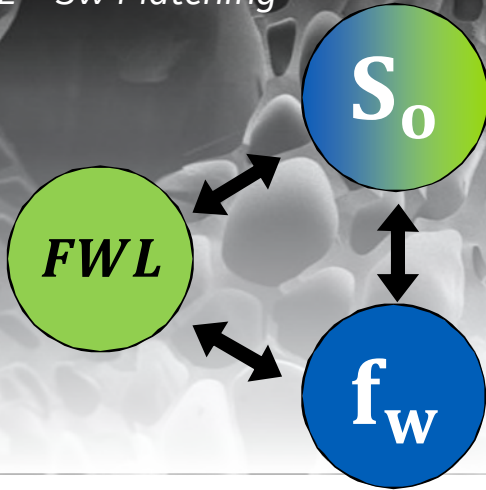
☒ While honoring the S_w deflection

**BAKER
HUGHES**
a GE company



MultiPhase Flow Petrophysics

$F_w - F_{WL} - S_w$ Matching



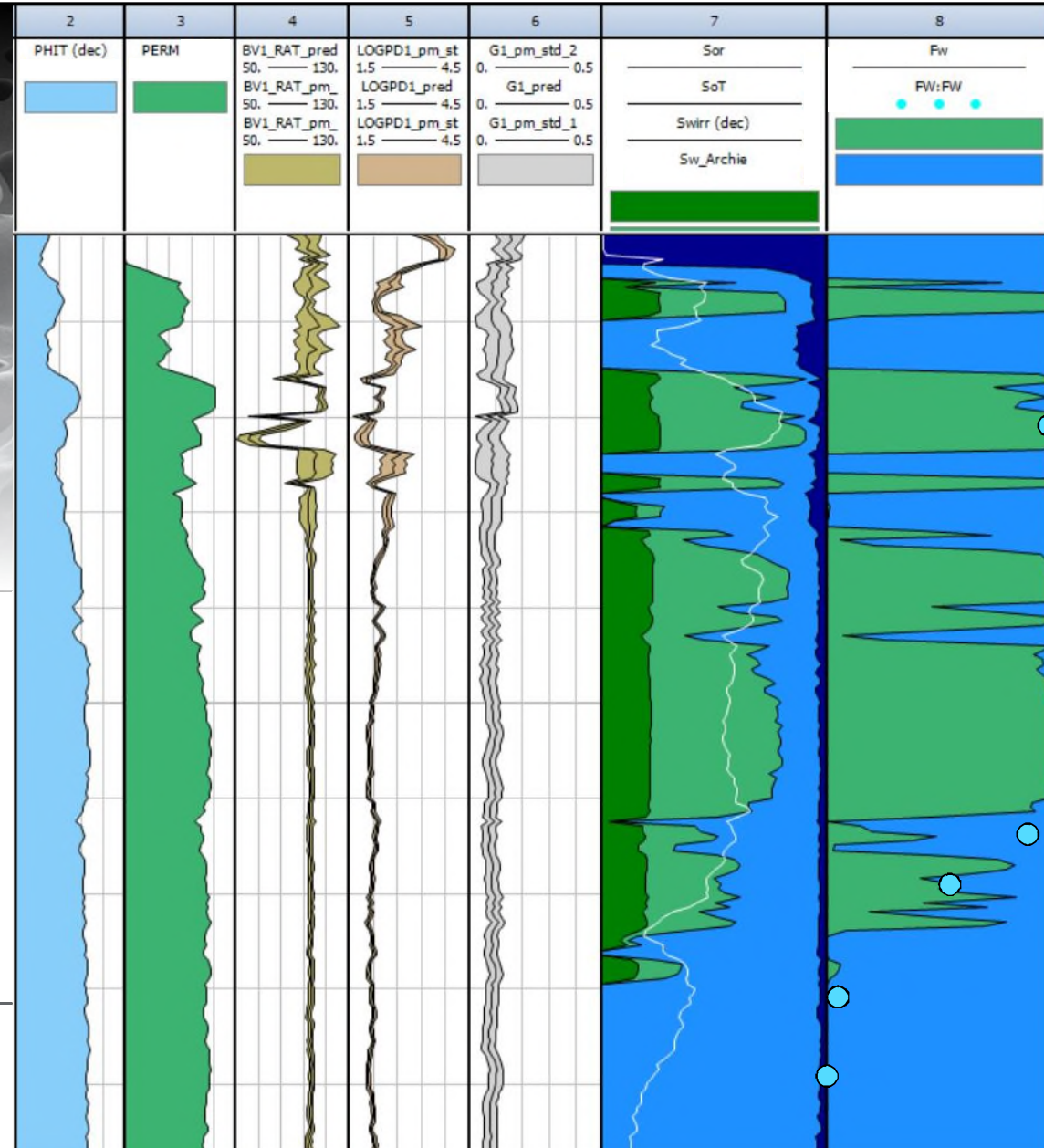
Optimization under Constraint

CASE #2

○ Match the F_w from formation sampling

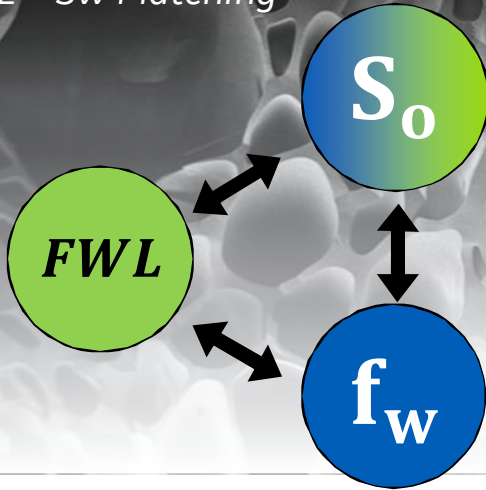
○ While honoring the S_w deflection despite an ROZ (remaining oil zone)

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HUGHES**
a GE company



MultiPhase Flow Petrophysics

$F_w - FWL - S_w$ Matching



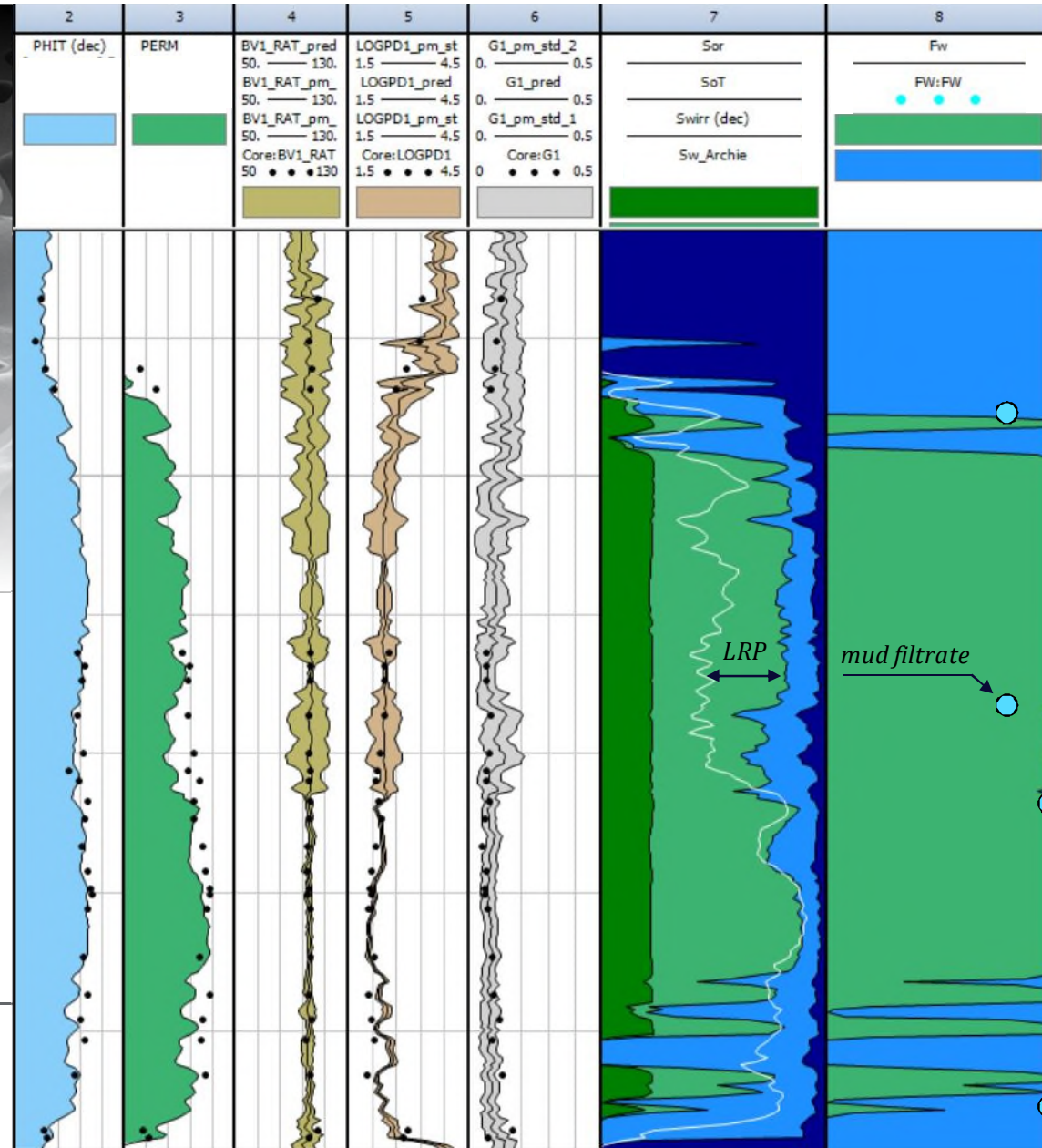
Optimization under Constraint

CASE #3

○ Match the F_w from formation sampling

○ While honoring an ODT

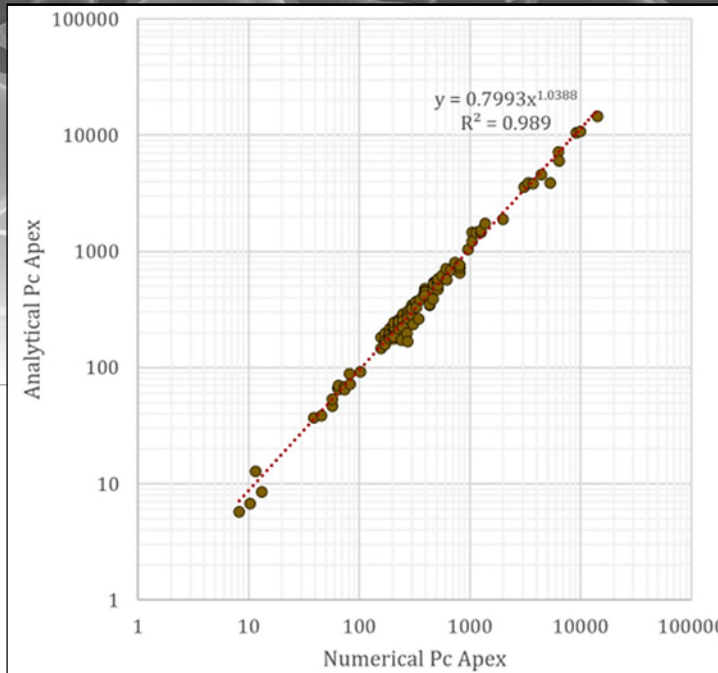
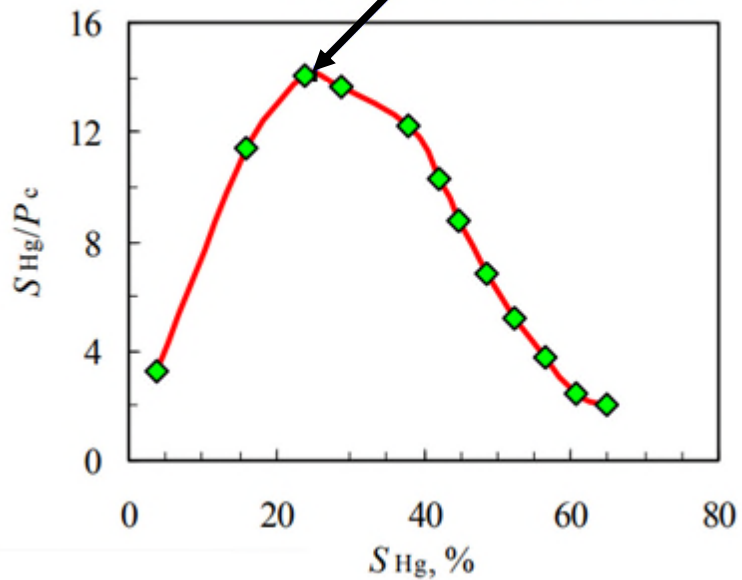
**BAKER
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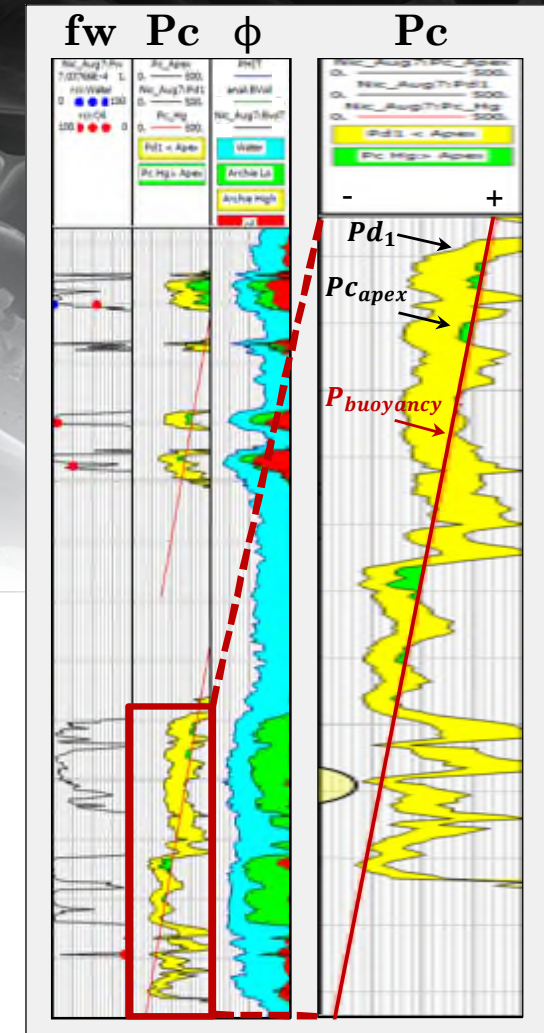
MultiPhase Flow Petrophysics

Using the Swanson Apex Heuristics

The Swanson parameter



$$P_{c,apex} \approx 2P_{d1} e^{-\frac{\sqrt{G_1}}{2}}$$



Wrap-up

A physics-based approach for real-time in-situ Sw in LRP

- **No full proof logging solution for OH in-situ Sw determination in non-Archie rocks on the market. Only indirect methods.**
- **The proposed approach has a well-defined domain of applicability is physics-driven, forward-modelling and has the potential to be applied real time.**
- **The NMR-based Sw processing provides in-situ water saturation, associated uncertainties as well as a single-phase producibility index. However, Sw only does not guarantee producibility**
- **The combination with Formation Sampling enables to de-risk compartmentalization and fluid typing challenges, calibrate NMR permeability while providing real-time multi-phase flow petrophysics actionable results.**

Take Away

A physics-based approach for real-time in-situ Sw in LRP

- No full proof logging solution for OH in-situ Sw determination in non-Archie rocks on the market. Only indirect methods.
- The proposed approach has a well-defined domain of applicability is **physics-driven**, forward-modelling and has the potential to be applied real time.
- The NMR-based Sw processing provides in-situ water saturation, associated uncertainties as well as a single-phase producibility index. However, **Sw only does not guarantee producibility**
- The combination with **Formation Sampling** enables to de-risk compartmentalization and fluid typing challenges, calibrate **NMR** permeability while providing real-time multi-phase flow petrophysics actionable results.