Optimal Model Complexity in Geological Carbon Sequestration: A Design of Experiment (DoE) & Response Surface (RS) Uncertainty Analysis

Project Number: DE-FE-0009238

Mingkan Zhang¹, Ye Zhang¹, Peter Lichtner²

1.Dept. of Geology & Geophysics, University of Wyoming, Laramie, Wyoming 2.OFM Research, Inc., Santa Fe, New Mexico

U.S. Department of Energy National Energy Technology Laboratory Carbon Storage R&D Project Review Meeting Developing the Technologies and Infrastructure for CCS August 18-20, 2015

Presentation Outline

- Project goals and benefits;
- Detailed project objectives & success criteria;
- Accomplishments to date;
- Summary of results;

Benefit to the Program

Major goals:

Support industry's ability to predict CO_2 storage capacity in geologic formations to within ±30% accuracy;

Develop and validate technologies to ensure 99% storage permanence.

Project benefits:

Facilitate the development and implementation of efficient workflows for modeling field-scale GCS in a variety of geochemically reactive environments, where formations exhibit multiple scales of permeability (*k*) heterogeneity.

Project Overview: Goals and Objectives

- Develop, test, and verify the DoE and RS uncertainty analysis for a fully heterogeneous reference model (FHM) & increasingly lower resolution "geologic models" created from upscaling the FHM.
- Investigate the effect of increasing reservoir k variance and depth on the uncertainty outcomes including optimal heterogeneity resolution(s). At greater injection depths, investigate gravity-stable injection.
- Investigate the effect of mineral reactions on GCS, including mineral volume fractions, reactive rate constants, reactive surface areas, and the impact of different geochemical databases.

Project Overview: Success Criteria

- At increasing depth, for both weakly and strongly heterogeneous systems, the geologic models can capture the FHM CO₂ behaviors; → Reduced characterization cost;
- RS analytical models are successfully verified against full-physics reservoir simulations via HPC, thus prediction uncertainty of any outcome at any time can be assessed using the low-resolution model(s) running the efficient RS models. → Enhanced computation efficiency;
- Mineral storage analysis: seeking the most efficient composition for reactive storage
 → Enhanced storage;
- Greater injection depth: within the uncertainty analysis framework, identify the combination(s) of favorable parameters & reservoir conditions that give rise to gravity-stable flow. → Enhanced storage security.

Accomplishments to Date

- High-resolution reservoir k heterogeneity (3.2 M grid cells) & geologic (upscaled) models of decreasing k resolutions built;
- For multiple system lnk variances, permeability upscaling & singlephase flow verification;
- For multiple system lnk variances, dispersivity upscaling & verification;
- Parallel simulation of CO₂ storage with PFLOTRAN & performance scaling on supercomputer;
- Uncertainty analysis of dissolution storage and CO₂ leakage in heterogeneous and the geologic models;
- Uncertainty analysis of CO₂ modeling considering mineral reactions;
- Uncertainty analysis of CO₂ modeling in greater injection depths. 6

Sediment Experiment at SAFL





http://www.safl.umn.edu/

Project Leader: Prof. Chris Paola Founding: NSF & oil industry consortium

Reservoir Heterogeneity Vs Geologic Models



A 1-unit homogeneous "formation" model is also created (not shown);

Permeability Upscaling & Verification



MRE

Dispersivity Upscaling & Verification

The plume moments are employed to compare the transport upscaling results as shown in Figure 3. The zero, first and second plume moments are defined as:

$$M = \iiint_{\Omega}(\theta c) dx dy dz, \mathbf{L}_{p} = \frac{1}{M} \iiint_{\Omega} (\mathbf{X}_{p} \theta c) dx dy dz \quad \text{and} \quad s^{2} = \frac{1}{M} \iiint_{\Omega} (\mathbf{X}_{p} - \mathbf{L}_{p}) (\mathbf{X}_{p} - \mathbf{L}_{p}) \theta c dx dy dz$$

In addition, both the tailing behavior (Figure 4) and the breakthrough curve (Figure 5) of the FHM have been captured when the variance of ln(K) is low to modest.



Figure 3. Evolution of plume moments with time: (a) & (b) mean plume displacements for variance($\ln K$) = 0.1 and 4.5, respectively; (c) & (d) longitudinal plume covariances for variance($\ln K$) = 0.1 and 4.5, respectively. The black, red, blue and green lines represent FHM, 1-unit, 3-unit and 8-unit models, respectively.









- For a given variance, accuracy: 8-unit > 3-unit > 1unit model;
- For var(ln*k*) up to 4.5, 8- and 3unit models can accurately capture plume migration pathway, mass centroid, and size; Optimal resolution: 3-unit model.
- For var(ln*k*) = 0.1, all models can accurately capture solute transport BTC; Optimal resolution: 1-unit model.
- For var(lnk) = 4.5, only the 8unit model can capture some aspect of solute transport BTC;
 Optimal resolution: 8-unit or higher.
- Optimal heterogeneity depends on prediction goal and system variance.

PFLOTRAN Scaling on Yellowstone

Yellowstone is a 1.5petaflops supercomputer with 72,288 processor cores & 144.6 TB of memory. <u>http://www2.cisl.ucar.e</u> <u>du/resources/yellowsto</u> <u>ne</u>

1-unit model (3.2M): * 20 yr CO₂ injection + 2000 yr monitoring * 2048 cores: 9 hours



Uncertainty Analysis: CO₂ Storage & Leakage

Case	T Gradient (°C/m)	Brine Salinity (Molal)	k_{cap} (m ²)	Inject rate (kg/s)
1	-0.025	0	10 ^{-17.5}	4
2	-0.025	4	10 ^{-17.5}	4
3	-0.05	0	10 ^{-17.5}	4
4	-0.05	4	10 ^{-17.5}	4
5	-0.0375	2	10 ⁻¹⁶	2
6	-0.0375	2	10 ⁻¹⁶	8
7	-0.0375	2	10 ⁻¹⁹	2
8	-0.0375	2	10 ⁻¹⁹	8
9	-0.025	2	10 ^{-17.5}	2
10	-0.025	2	10 ^{-17.5}	8
11	-0.05	2	10 ^{-17.5}	2
12	-0.05	2	10 ^{-17.5}	8
13	-0.0375	0	10 ⁻¹⁶	4
14	-0.0375	0	10 ⁻¹⁹	4
15	-0.0375	4	10 ⁻¹⁶	4
16	-0.0375	4	10 ⁻¹⁹	4
17	-0.025	2	10 ⁻¹⁶	4
18	-0.025	2	10 ⁻¹⁹	4
19	-0.05	2	10 ⁻¹⁶	4
20	-0.05	2	10 ⁻¹⁹	4
21	-0.0375	0	10 ^{-17.5}	2
22	-0.0375	0	10 ^{-17.5}	8
23	-0.0375	4	10 ^{-17.5}	2
24	-0.0375	4	10 ^{-17.5}	8
25	-0.0375	2	10 ^{-17.5}	4



- When system *k* variance is low, the 1-unit model can accurately capture the scCO₂ plume footprint of the FHM.
- 8-unit and 3-unit models provide more accurate scCO₂ plume predictions than the 1-unit model, when system k variance is high,



- When system *k* variance is low, the 1-unit model can accurately capture the dissolved CO₂ plume of the FHM.
- 8-unit and 3-unit models provide more accurate dissolved CO₂ plume predictions than the 1-unit model.

Dissolved CO₂



- Increasing brine salinity decreases CO₂ dissolution.
- The 8-unit and 3-unit models yield great accurate dissolved CO₂ predictions.

Dissolved CO₂ at 2000 Years

		$\sigma^2 = 0.1$			$\sigma^2 = 1.0$)		$\sigma^2 = 4.5$	5
Model	Term	Scaled Estimate	Plot Estimate	Term	Scaled Estimate	Plot Estimate	Term	Scaled Estimate	Plot Estimate
1unit	T Gradient	-66525.68		T Gradient	1555.4917		T Gradient	-50947.11	
	Brin Salinity	-17979466		Brin Salinity	-17082310		Brin Salinity	-18280660	
	К Сар	1497108	: : : : : :	К Сар	1860613		К Сар	1969252	
	Inj rate	287260.65		Inj rate	241458.07		Inj rate	-104160.2	
3unit	T Gradient	-32093.85		T Gradient	27034.642		T Gradient	17782.292	
	Brin Salinity	-15783477		Brin Salinity	-11537619		Brin Salinity	-9257227	
	К Сар	1460367	: : : : : :	К Сар	1767155		К Сар	1770883	
	Inj rate	410148.22		Inj rate	421509.17		Inj rate	-82521.86	
8unit	T Gradient	16158.892		T Gradient	48589.475		T Gradient	60138.433	
	Brin Salinity	-14915585		Brin Salinity	-10880135		Brin Salinity	-8878195	
	К Сар	1776827		К Сар	1745176		К Сар	1735825	
	Inj rate	340980.92		Inj rate	450741.17		Inj rate	426151.34	
FHM	T Gradient	14264.942		T Gradient	39376.7		T Gradient	-6628.533	
	Brin Salinity	-14967318		Brin Salinity	-10393509		Brin Salinity	-8334250	
	К Сар	1765641		К Сар	1687305		К Сар	1611689	
	Inj rate	320034.78		Inj rate	488939.2		Inj rate	544649.55	



- Under low variance condition, the 1-unit model can reasonably capture the leakage plume of the FHM.
- The 8-unit and 3-unit models yield more accurate leaked CO₂ plume predictions than the 1-unit model.

Leakage of CO₂ into caprock



- The decreasing caprock permeability reduces CO₂ leakage.
- Base on results of the upscaling study, the 8-unit and 3-unit models yield great accurate Leaked CO₂ predictions.

Uncertainty Results for CO₂ Leakage

		$\sigma^2 = 0.1$			$\sigma^2 = 1.0$			$\sigma^2 = 4.5$	5
Model	Term	Scaled Estimate	Plot Estimate	Term	Scaled Estimate	Plot Estimate	Term	Scaled Estimate	Plot Estimate
1unit	T Gradient	469420.96		T Gradient	-388487.8		T Gradient	-1419857	
	Brin Salinity	-15384250		Brin Salinity	-14242776		Brin Salinity	-8039491	
	К Сар	-76663063		К Сар	73456018		К Сар	81043291	
	<u>Inj</u> rate	-2662427		Inj rate	-3308578		Inj rate	-3977240	
3unit	T Gradient	543991.63		T Gradient	-194952.5		T Gradient	-544753.2	
	Brin Salinity	-18842939		Brin Salinity	-21237658		Brin Salinity	-20641268	
	К Сар	-76407790		К Сар	70965164		К Сар	75600868	
	<u>Inj</u> rate	-2558673		Inj rate	-2872086		Inj rate	-4702174	
8unit	T Gradient	-283710.1		T Gradient	-108156.7		T Gradient	-162012.8	
	Brin Salinity	-19111777		Brin Salinity	-21175701		Brin Salinity	-20683088	
	К Сар	-70108544		К Сар	69845490		К Сар	72073598	
	Inj rate	-2376895		Inj rate	-2710601		Inj rate	-4294708	
FHM	T Gradient	-307085.2		T Gradient	-91501.07		T Gradient	121540.03	
	Brin Salinity	-19035197		Brin Salinity	-20825214		Brin Salinity	-20438426	
	К Сар	-70111854		К Сар	67276691		К Сар	64498465	
	Inj rate	-2453283		Inj rate	-2954034		Inj rate	-3344799	

Mineral Storage Modeling

Mineral	Formula	Init VF (%)
Quartz	SiO ₂	43.04
Calcite	CaCO ₃	4.22
K-Feldspar	KAISi ₃ O ₈	15.77
Kaolinite	$AI_2Si_2O_5(OH)_4$	0
Albite	NaAlSi ₃ O ₈	0
Plagioclase	(Na _{0.75} ,Ca _{0.25})(Al _{1.25} ,Si _{2.75})O ₈	4.07
Illite	K _{0.6} (Mg _{0.25} , Al _{1.8})(Al _{0.5} , Si _{3.5})O ₁₀ (OH) ₂	4.01
Hematite	Fe ₂ O ₃	1.60
Dawsonite	NaAICO ₃ (OH) ₂	0
Chlorite	(Mg _{2.5} , Fe _{2.5} , AI)(AI, Si ₃)O ₁₀ (OH) ₈	7.19
Siderite	FeCO ₃	0
Ankerite	Ca(Mg _{1.3} , Fe _{0.7})(CO ₃) ₂	0
Magnesite	MgCO ₃	0
Na-Smectite	Na _{0.290} (Mg _{0.26} , Al _{1.74})(Al _{0.03} , Si _{3.97})O ₁₀ (OH) ₂	0
Ca-Smectite	Ca _{0.145} (Mg _{0.26} , Al _{1.74})(Al _{0.03} , Si _{3.97})O ₁₀ (OH) ₂	0
Dolomite	(CaMg)(CO ₃) ₂	0

Changes in Volume Fraction after 2000 Years



- Reactive minerals in sandstone such as Chlorite can provide cations such as Mg²⁺ and Fe²⁺, which are essential chemical components for forming carbonate precipitates during GCS.
- The reactions between cations and CO₂ forms carbonate minerals (e.g., siderite, magnesite and ankerite) to trap CO₂ as precipitates.
- Uncertainty analysis evaluating important uncertainty factors is ongoing.

Deep Injection

- Deep injection trapping of CO₂ in deep marine sediments provides additional storage to existing onshore capacity.
- CO₂ can be trapped through 'selfsealing' gravitational and hydrateforming mechanisms under suitable temperature and pressure conditions.
- Uncertainty analysis of ocean storage in the Gulf of Mexico is about sediment property, thermal gradient, sea water depth, etc.



Figure 22. Oil and gas pipelines with diameters greater than or equal to 20 inches.

oil and gas pipe line in GOM (Richardson et al., 2004, OCS Report MMS 2004-021)

scCO₂ plumes at 1km water depth



scCO₂ plumes at 3km water depth



Dissolved CO₂ plumes at 1km water depth



Dissolved CO₂ plumes for 3km case



Summary

- Global upscaling computes equivalent ks for the geologic models with decreasing k resolution; for increasing reservoir ln(k) variances (0.1, 1.0, 4.5), FHM pressure and flow rate are captured well by the geologic models, but errors increase with variance.
- When the variance of ln(k) is low, the 1-unit model yields similar dissolution and leakage plumes as the FHM. When the variance of ln(k) is high, the 3-unit and 8-unit models provide more accurate predictions on CO₂ dissolution and leakage.
- Experimental design analysis suggests that for the uncertainty factors evaluated, brine salinity
 is the single most influential factor impacting CO₂ dissolution storage, while caprock
 permeability is the most influential factor impacting CO₂ leakage to the caprock.
- Reactions between cations and dissolved CO₂ forms carbonate mineral precipitates (i.e., Siderite and Magnesite), leading to mineral storage. High degree of uncertainty exists in its prediction.
- When the water depth is 1km, it is too shallow to develop gravity stable flow. When the water depth is 3km, it is sufficient to develop gravity stable flow. However, the magnitude of sediment permeabity can impact storage security: when k<10⁻¹⁵ m² (clay sediment), CO₂ is also gravity neutral for all water depths, and for all geothermal gradient.

Appendix

 The following slides will not be discussed during the presentation, but are mandatory

FHM v. 1-Unit Model: σ_{lnk} =0.1



Dept. of Geology & Geophysics, University of Wyoming

FHM v. 1-Unit Model: σ_{lnk} =4.5



Dept. of Geology & Geophysics, University of Wyoming

An *example 1-Unit model* run for CO2 storage modeling simulated on the Yellowstone supercomputer. The problem domain is 7000 m x 7000 m x 250 m. Shown at 100 years for an isosurface of 0.0125 (mole fraction) of dissolved CO2. CO2 is injected at a depth of 50 m below the top at the center of the xy-domain for 20 years. The grid is $160 \times 160 \times 25 = 0.64$ million cells.



PFLOTRAN Scaling on Yellowstone



PFLOTRAN formulations

To model GCS, the following mass and energy conservation equations are solved:

$$\frac{\partial}{\partial t} \left[\varphi \sum_{\alpha} (\rho_{\alpha} s_{\alpha} X_{i}^{\alpha}) \right] + \nabla \cdot \sum_{\alpha} (\rho_{\alpha} X_{i}^{\alpha} \vec{q}_{\alpha} - \varphi \rho_{\alpha} s_{\alpha} \tau_{\alpha} D_{\alpha} \nabla X_{i}^{\alpha}) = S_{i}$$
(1)
$$\frac{\partial}{\partial t} \left[\varphi \sum_{\alpha} (\rho_{\alpha} s_{\alpha} U_{\alpha}) + (1 - \varphi) \rho_{r} C_{p,r} T \right] + \nabla \cdot \left[\sum_{\alpha} (\vec{q}_{\alpha} \rho_{\alpha} H_{\alpha}) - \lambda \nabla T \right] = Q$$
(2)

 φ denotes porosity, and ρ_{α} , s_{α} , τ_{α} , D_{α} , U_{α} , H_{α} refer to the density, saturation, tortuosity, diffusion coefficient, internal energy, and enthalpy of fluid phase α , respectively. Two fluid phases (CO₂, brine) will be modeled. The quantities X_i^{α} denote the mole fraction of component *i* in phase α . The quantities $C_{p,r}$ and λ denote the rock heat capacity and conductivity, respectively. The summation is carried out over all fluid phases present in the system. The system is assumed locally to be in thermodynamic equilibrium with temperature $T(\vec{x}; t)$ at position \vec{x} and time *t*. The quantity Q denotes an energy source/sink term.

The quantity *Si* denotes a source/sink term for the *i*th primary species describing reaction with minerals given by $S_i = -\sum_m v_{im} I_m$, with stoichiometric reaction coefficients v_{im} and kinetic rate I_m for the *m*th mineral, taken as positive for precipitation and negative for dissolution.

The flow rate \vec{q}_{α} of fluid phase α is given by the extended Darcy's law: $\vec{q}_{\alpha} = -\frac{\bar{k}k_{\alpha}}{\mu_{\alpha}}(\nabla p_{\alpha} - \rho_{\alpha}gz)$, with intrinsic permeability \bar{k} , relative permeability k_{α} , fluid viscosity μ_{α} , and pressure p_{α} of phase α .