

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

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National Energy Technology Laboratory
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Developing the Technologies and
Infrastructure for CCS
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Presentation Outline

- Project goals and benefits;
- Detailed project objectives & success criteria;
- Accomplishments to date;
- Summary of results;
- Appendix (organization chart; Gantt chart; supplemental results).

Benefit to the Program

Major goals:

Support industry's ability to predict CO₂ storage capacity in geologic formations to within $\pm 30\%$ 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.

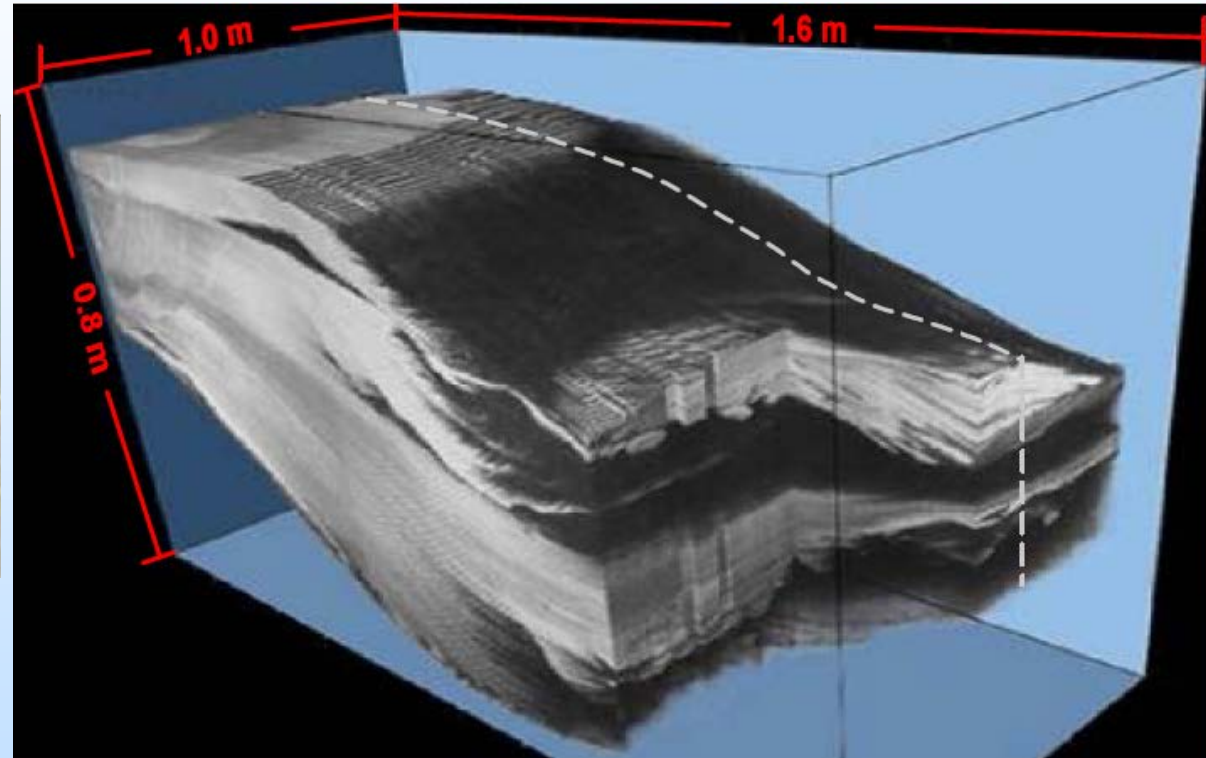
Project Overview: Success Criteria

- At increasing depth, for both weakly and strongly heterogeneous systems, the geologic models can capture the FHM CO₂ behaviors within the full parameter space; → **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 condition that give rise to gravity-stable flow. → **Enhanced storage security**.

Accomplishments to Date

- High-resolution reservoir k heterogeneity (3.2 M grid cells) & geologic models of decreasing k resolutions;
- Permeability upscaling & single-phase flow verification;
- CO₂ modeling with PFLOTRAN & performance scaling on the petascale Yellowstone supercomputer at NWSC;
- Model comparison & DoE/RS analysis;
- CO₂ modeling considering mineral reactions.

Sediment Experiment at SAFL

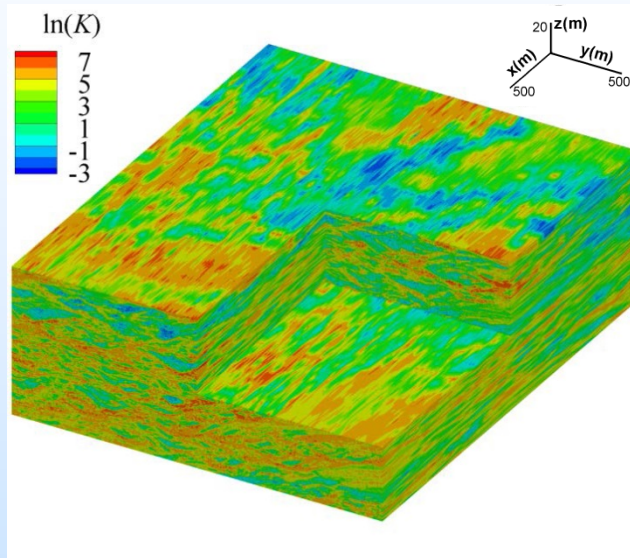


<http://www.safl.umn.edu/>

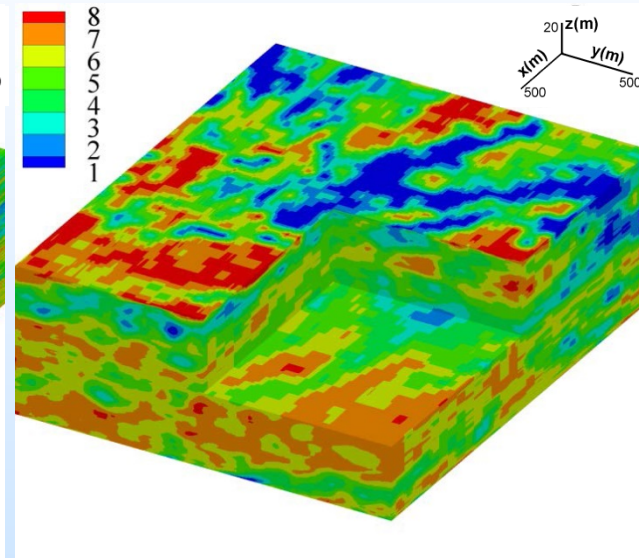
Project Leader: Prof. Chris Paola

Founding: NSF & oil industry consortium

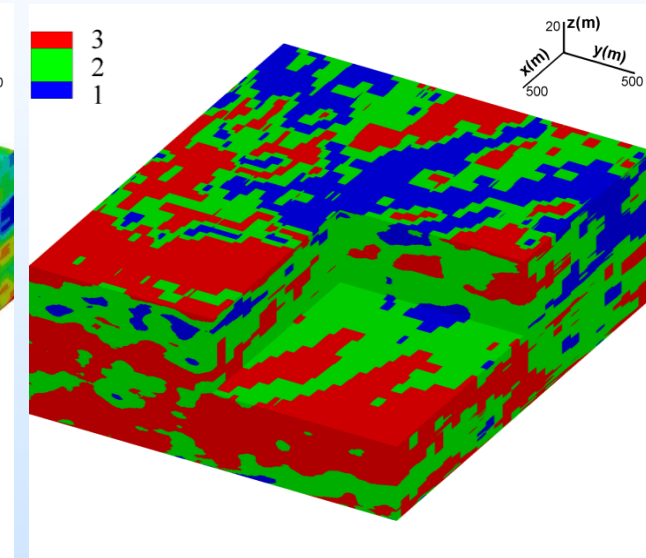
Reservoir Heterogeneity Vs Geologic Models



FHM



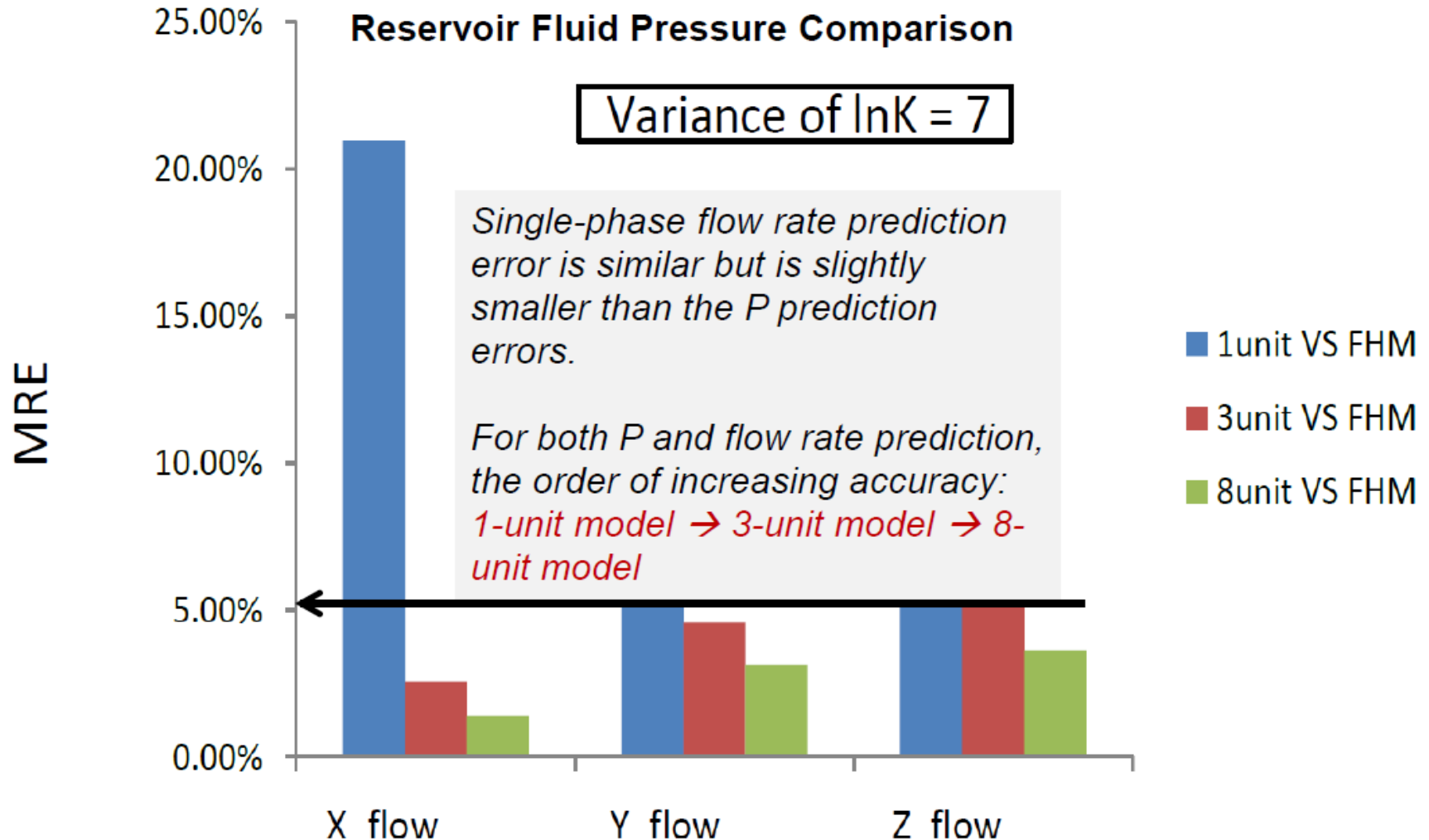
8-unit facies model



3-unit facies model

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

Upscaling Verification



Carbon Sequestration Modeling with Reactions

- Multicomponent-multiphase non-isothermal reactive flow and transport model;
- Massively parallel---based on the PETSc parallel framework;
Peta-scale performance
Highly scalable (run on over 265k cores)
- Supercritical CO₂-H₂O;
Span-Wagner EOS for CO₂ density & fugacity coefficient
Mixture density for dissolved CO₂ in brine (Duan et al., 2008)
Viscosity of CO₂ (Fenghour et al., 1998)
- Finite Volume Discretization;
Variable switching for changes in fluid phase
Structured/Unstructured grids
- Reactive transport modeling, including CO₂-mineral reactions with many degrees of freedom

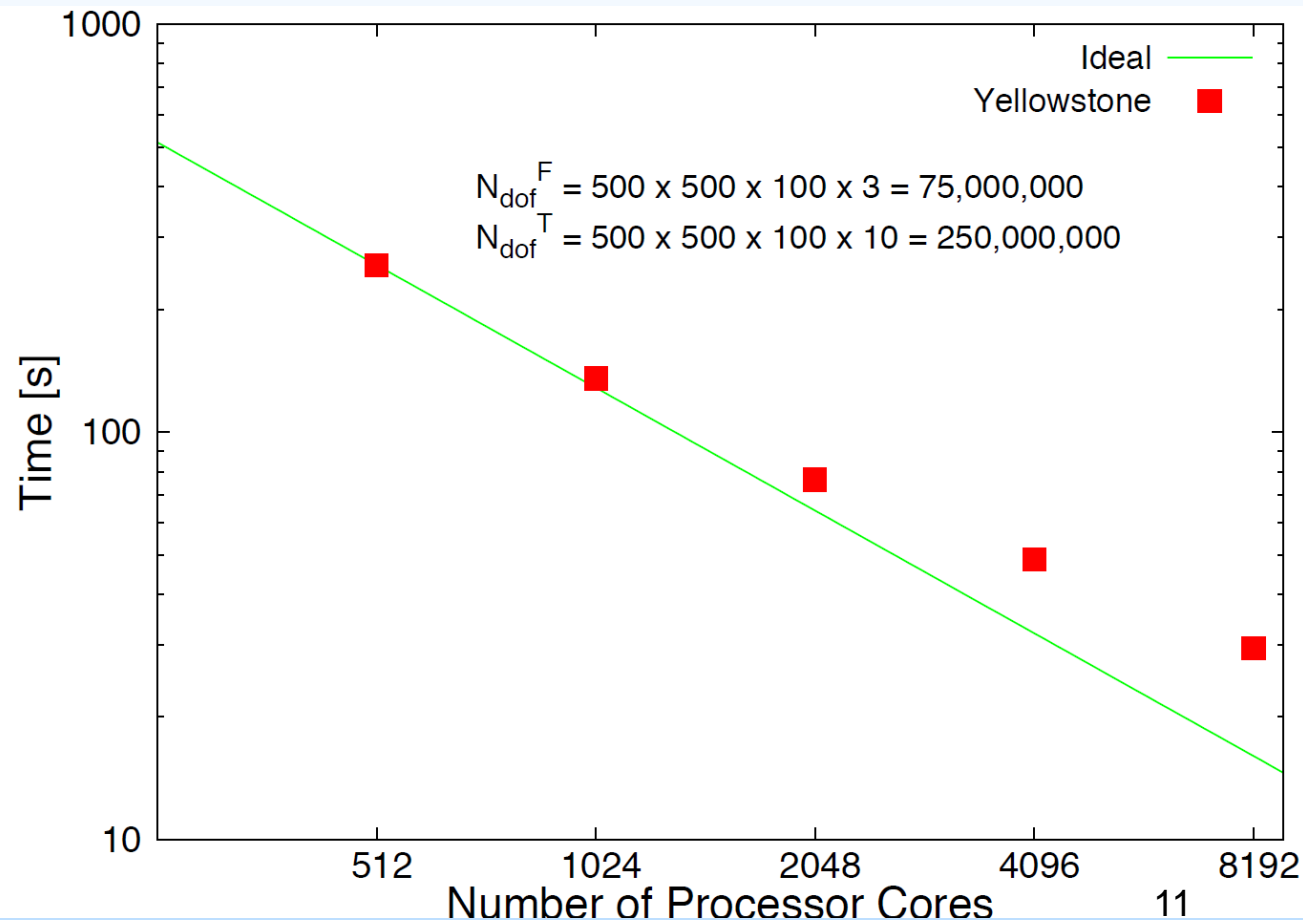
PFLOTRAN Scaling on Yellowstone

Yellowstone is a 1.5-petaflops supercomputer with 72,288 processor cores & 144.6 TB of memory.

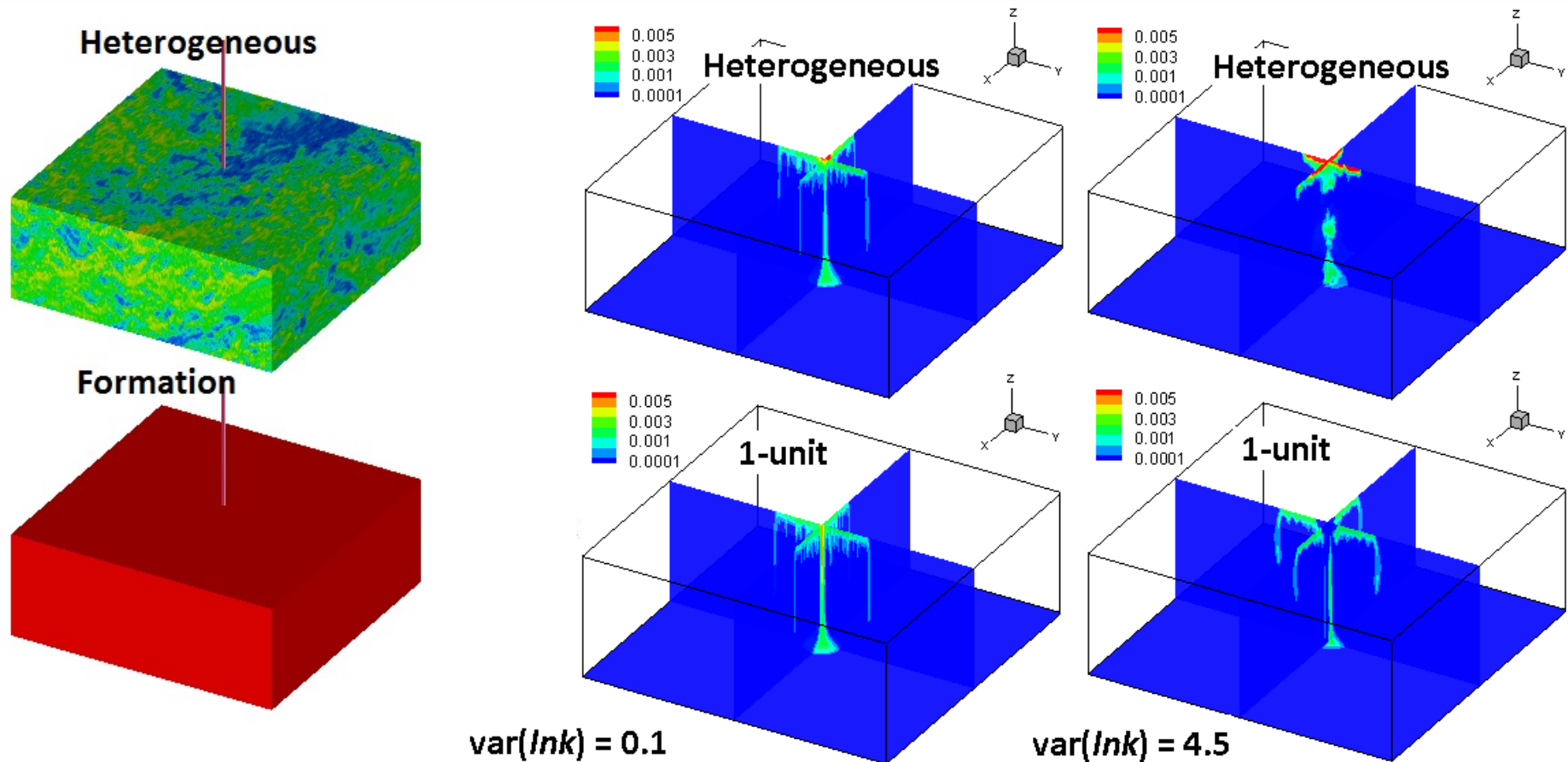
<http://www2.cisl.ucar.edu/resources/yellowstone>

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

1-unit model (25 M): CO₂ injection w/ reactive chemistry



Dissolved CO₂



- Under both low and high variance conditions, the 1-unit model can reasonably capture the plume footprint of the FHM.
- Base on results of the upscaling study, the 8-unit and 3-unit models (simulations are ongoing) should yield more accurate dissolved CO₂ predictions than the 1-unit model.

Design of Experiment (1-unit)

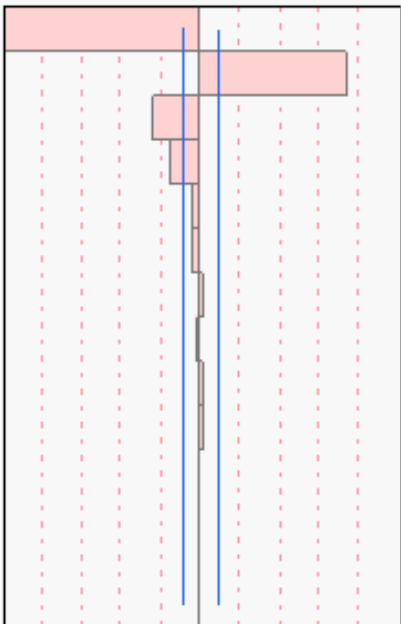
	Pattern	T_Gradient	Brin_Salinity	K_Cap	Inj_Rate	SR_EOI	SR_EOM	CO2_LK_EOI	CO2_LK_EOM	Brin_LK_EOI	Brin_LK_EOM
1	++00	1	1	0	0	0.10835058	0.296383093	2.45e-12	1.31e-10	0.0000000678	0.00000361
2	0+0-	0	1	0	-1	0.115352993	0.290742974	3.99e-12	9.84e-11	0.00000011	0.00000272
3	0000	0	0	0	0	0.142377741	0.538398262	1.7e-12	9.23e-11	0.000000466	0.00000253
4	-00-	-1	0	0	-1	0.153843493	0.55349839	2.48e-12	6.18e-11	0.0000000678	0.00000169
5	0--0	0	-1	-1	0	0.199386363	0.437694529	2.13e-11	9.38e-10	0.000000583	0.0000256
6	0+0+	0	1	0	1	0.100143073	0.284811298	7.88e-13	9.84e-11	0.0000000218	0.00000272
7	-0+0	-1	0	1	0	0.138510286	0.687346742	-1.77e-13	5.09e-12	-4.84e-9	0.00000139
8	--00	-1	-1	0	0	0.19363702	0.795624274	1.11e-12	6.17e-11	0.0000000302	0.00000168
9	+00-	1	0	0	-1	0.152782269	0.552892793	4.92e-12	1.22e-10	0.000000135	0.00000336
10	+0+0	1	0	1	0	0.138169204	0.672851482	-3.45e-13	1.01e-11	-9.48e-9	0.000000276
11	0--0	0	-1	1	0	0.187461554	0.930414357	-2.96e-13	7.44e-12	-8.08e-9	0.000000203
12	+0-0	1	0	-1	0	0.147983085	0.297567526	2.84e-11	1.25e-9	0.00000078	0.0000342
13	+00+	1	0	0	1	0.13057312	0.514362598	9.6e-13	1.22e-10	0.0000000264	0.00000336
14	00+-	0	0	1	-1	0.149784337	0.703780661	-2.01e-13	7.59e-12	-5.52e-9	0.000000208
15	0+-0	0	1	-1	0	0.113389545	0.221478542	2.28e-11	0.000000001	0.000000631	0.0000276
16	0000	0	0	0	0	0.142377741	0.538398262	1.7e-12	9.23e-11	0.000000466	0.00000253
17	00--	0	0	-1	-1	0.156476955	0.282904316	4.28e-11	9.39e-10	0.00000117	0.0000257
18	+--00	1	-1	0	0	0.192891704	0.787607997	2.21e-12	1.22e-10	0.0000000606	0.00000334
19	00+-	0	0	-1	1	0.134054747	0.306602194	1.05e-11	9.39e-10	0.000000288	0.0000257
20	00++	0	0	1	1	0.126123131	0.665760497	-2.57e-13	7.59e-12	-7.04e-9	0.000000208
21	0-0+	0	-1	0	1	0.176212246	0.794850016	6.9e-13	9.2e-11	0.0000000189	0.00000251
22	--+00	-1	1	0	0	0.108954684	0.28941944	1.22e-12	6.59e-11	0.0000000338	0.00000182
23	-00+	-1	0	0	1	0.131172432	0.515158456	4.69e-13	6.18e-11	0.0000000128	0.00000169
24	-0-0	-1	0	-1	0	0.148748884	0.296929128	1.42e-11	6.29e-10	0.000000389	0.00000172
25	0++0	0	1	1	0	0.106233377	0.319291597	-2.51e-13	8.19e-12	-6.94e-9	0.000000226
26	0000	0	0	0	0	0.142377741	0.538398262	1.7e-12	9.23e-11	0.000000466	0.00000253
27	0-0-	0	-1	0	-1	0.210735883	0.815954164	3.66e-12	9.2e-11	0.0000000999	0.00000251

Parameter Ranking (1-unit)

Outcome: dissolved CO₂ at End of Monitoring

Response SR_EOM

Sorted Parameter Estimates

Term	Estimate	Std Error	t Ratio		Prob> t
Brin_Salinity	-0.238335	0.009485	-25.13		<.0001*
K_Cap	0.1780224	0.009485	18.77		<.0001*
Brin_Salinity*K_Cap	-0.098727	0.016428	-6.01		<.0001*
K_Cap*K_Cap	-0.054452	0.014227	-3.83		0.0024*
Inj_Rate	-0.009852	0.009485	-1.04		0.3194
K_Cap*Inj_Rate	-0.01543	0.016428	-0.94		0.3661
Brin_Salinity*Inj_Rate	0.0037931	0.016428	0.23		0.8213
T_Gradient*K_Cap	-0.003783	0.016428	-0.23		0.8217
T_Gradient*Brin_Salinity	0.003745	0.016428	0.23		0.8235
Inj_Rate*Inj_Rate	0.0028854	0.014227	0.20		0.8427
T_Gradient	-0.001359	0.009485	-0.14		0.8884
Brin_Salinity*Brin_Salinity	0.0007546	0.014227	0.05		0.9586
T_Gradient*T_Gradient	0.0001758	0.014227	0.01		0.9903
T_Gradient*Inj_Rate	-4.757e-5	0.016428	-0.00		0.9977

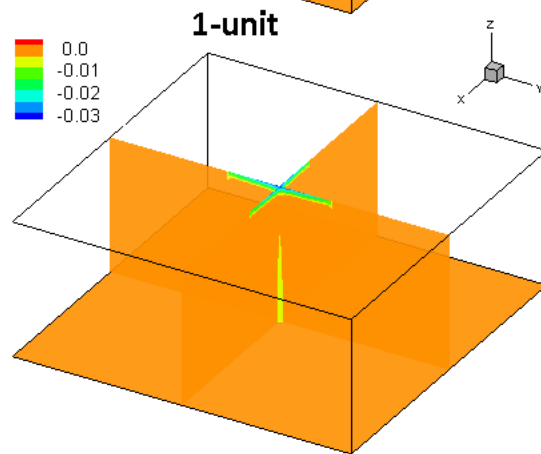
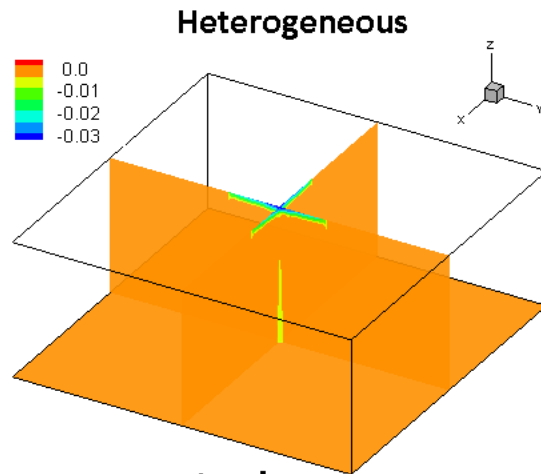
Mineral List

Mineral	Formula	Init VF (%)
Quartz	SiO_2	43.04213
Calcite	CaCO_3	4.21872
K-Feldspar	KAlSi_3O_8	15.77216
Kaolinite	$\text{Al}_2\text{Si}_2\text{O}_5(\text{OH})_4$	0
Albite	$\text{NaAlSi}_3\text{O}_8$	0
Plagioclase	$(\text{Na}_{0.75}, \text{Ca}_{0.25})(\text{Al}_{1.25}, \text{Si}_{2.75})\text{O}_8$	4.06691
Illite	$\text{K}_{0.6}(\text{Mg}_{0.25}, \text{Al}_{1.8})(\text{Al}_{0.5}, \text{Si}_{3.5})\text{O}_{10}(\text{OH})_2$	4.01098
Hematite	Fe_2O_3	1.598
Dawsonite	$\text{NaAlCO}_3(\text{OH})_2$	0
Chlorite	$(\text{Mg}_{2.5}, \text{Fe}_{2.5}, \text{Al})(\text{Al}, \text{Si}_3)\text{O}_{10}(\text{OH})_8$	7.191
Siderite	FeCO_3	0
Ankerite	$\text{Ca}(\text{Mg}_{1.3}, \text{Fe}_{0.7})(\text{CO}_3)_2$	0
Magnesite	MgCO_3	0
Na-Smectite	$\text{Na}_{0.290}(\text{Mg}_{0.26}, \text{Al}_{1.74})(\text{Al}_{0.03}, \text{Si}_{3.97})\text{O}_{10}(\text{OH})_2$	0
Ca-Smectite	$\text{Ca}_{0.145}(\text{Mg}_{0.26}, \text{Al}_{1.74})(\text{Al}_{0.03}, \text{Si}_{3.97})\text{O}_{10}(\text{OH})_2$	0
Dolomite	$(\text{CaMg})(\text{CO}_3)_2$	0

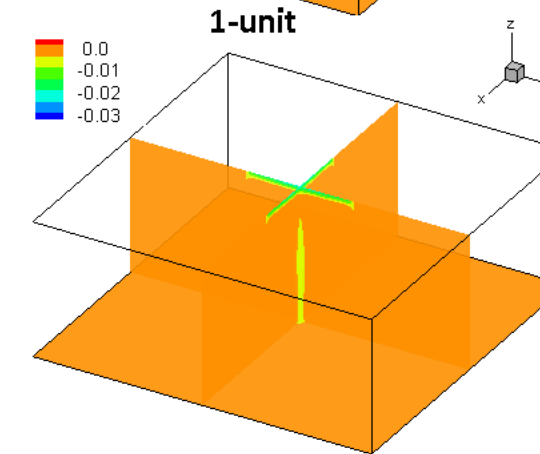
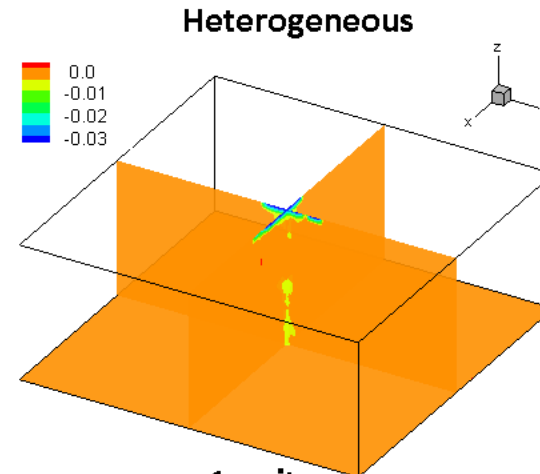
CO₂ Simulation: Mineral Trapping

- Chlorite can provide cations such as Mg²⁺ and Fe²⁺, which are essential chemical components for forming carbonate precipitates.
- The reactions between cations and CO₂ forms carbonate minerals (e.g., siderite, magnesite and ankerite) to trap CO₂ as precipitates.

Changes in Volume Fraction: Chlorite after 2000 years

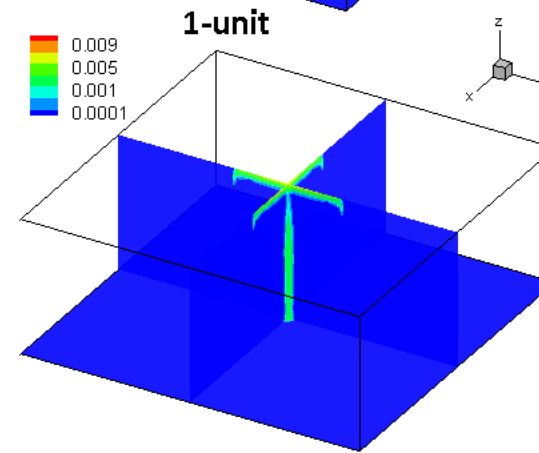
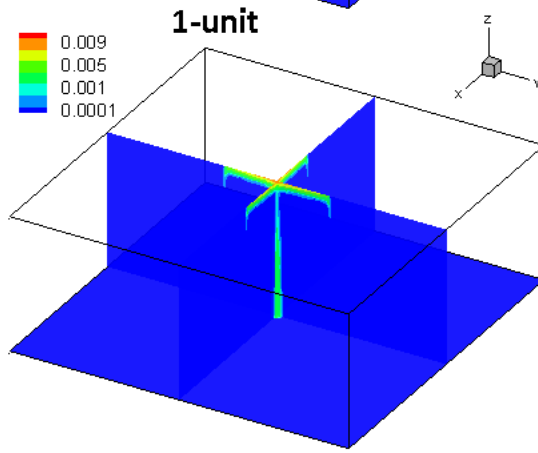
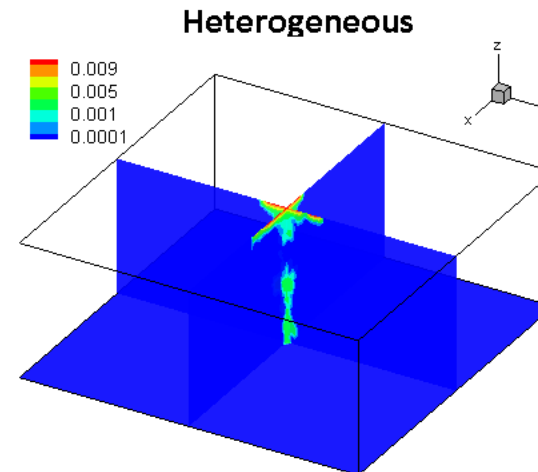
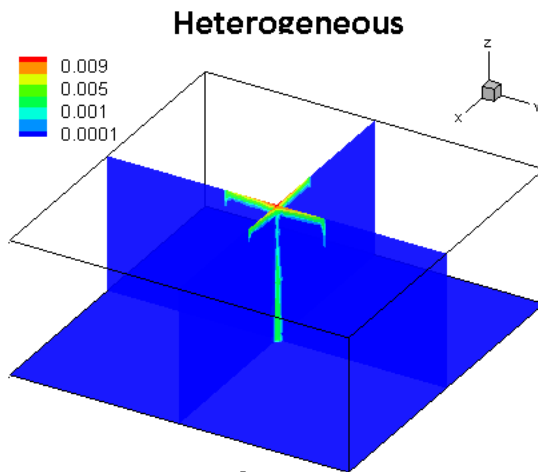


$\text{var}(\ln k) = 0.1$



$\text{var}(\ln k) = 4.5$

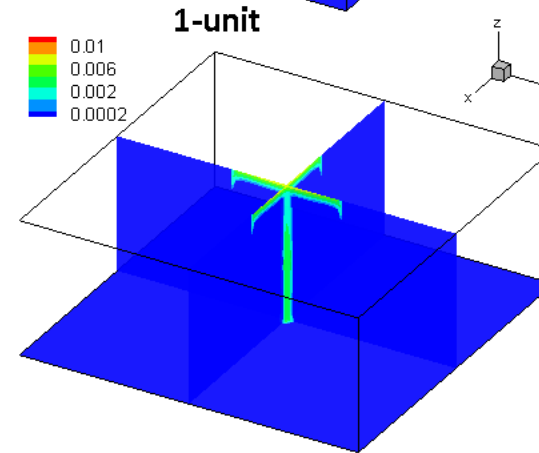
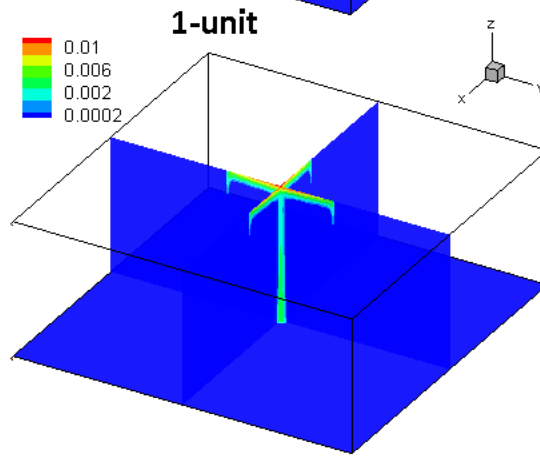
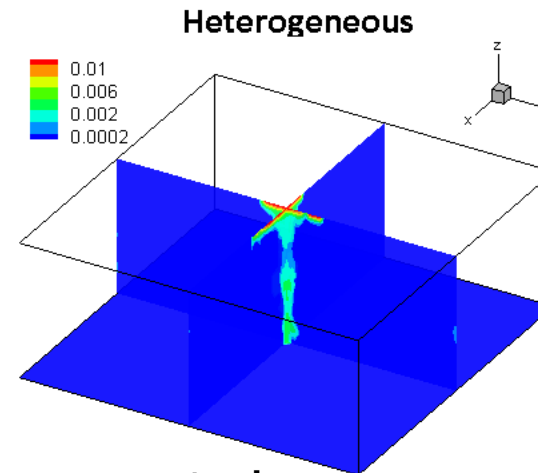
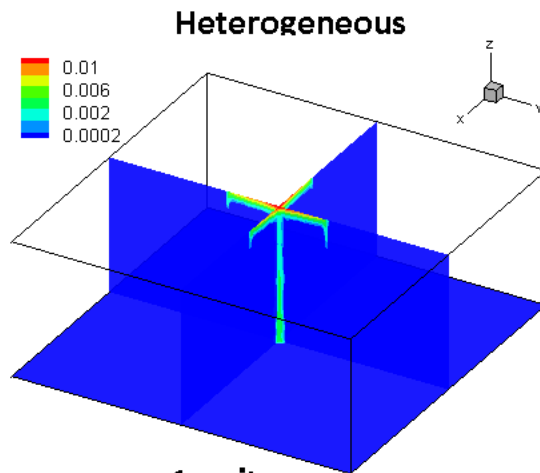
Changes Volume Fraction: Siderite after 2000 years



$\text{var}(\ln k) = 0.1$

$\text{var}(\ln k) = 4.5$

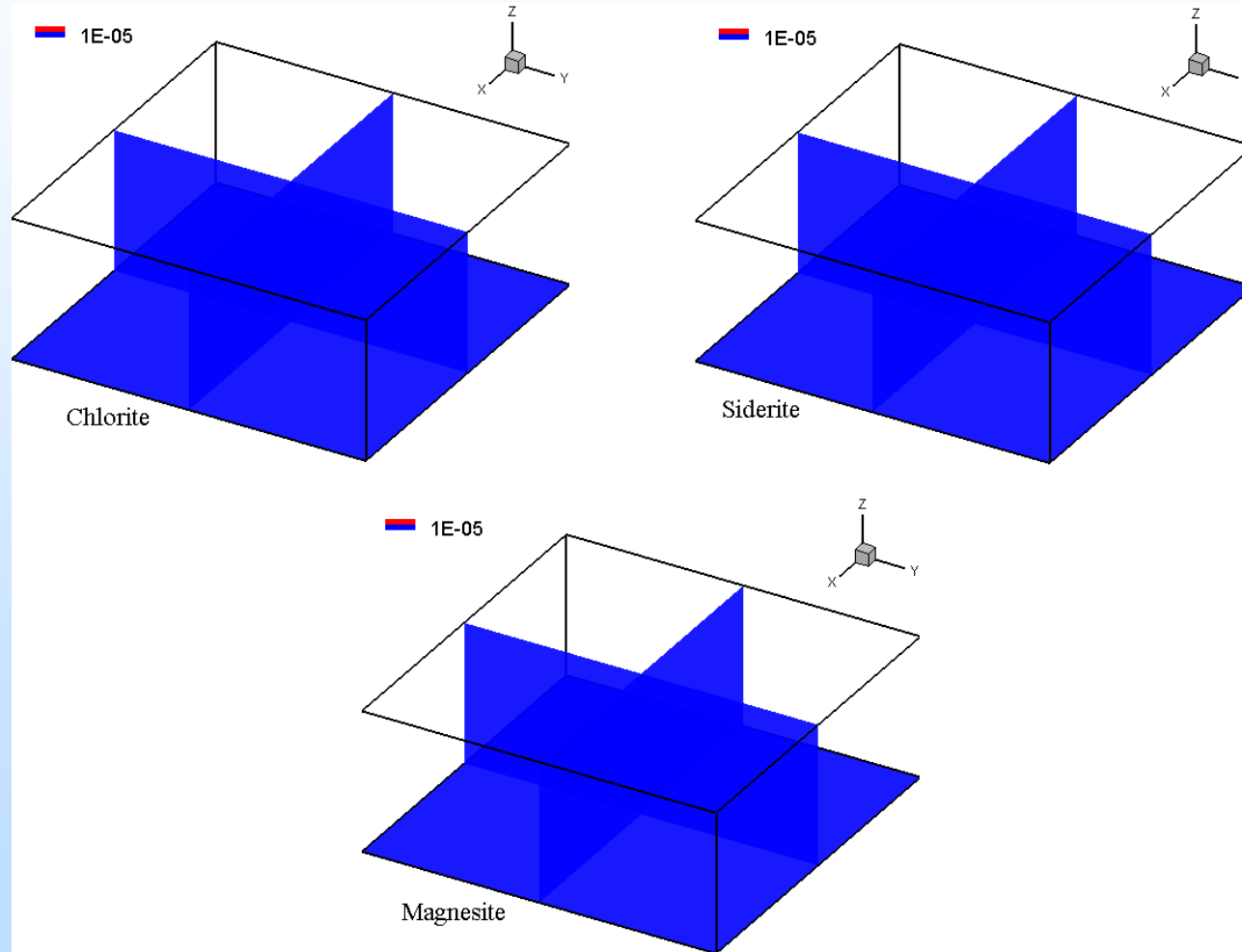
Changes Volume Fraction: Magnesite after 2000 years



$\text{var}(\ln k) = 0.1$

$\text{var}(\ln k) = 4.5$

Changes Volume Fraction: without Chlorite after 2000 years



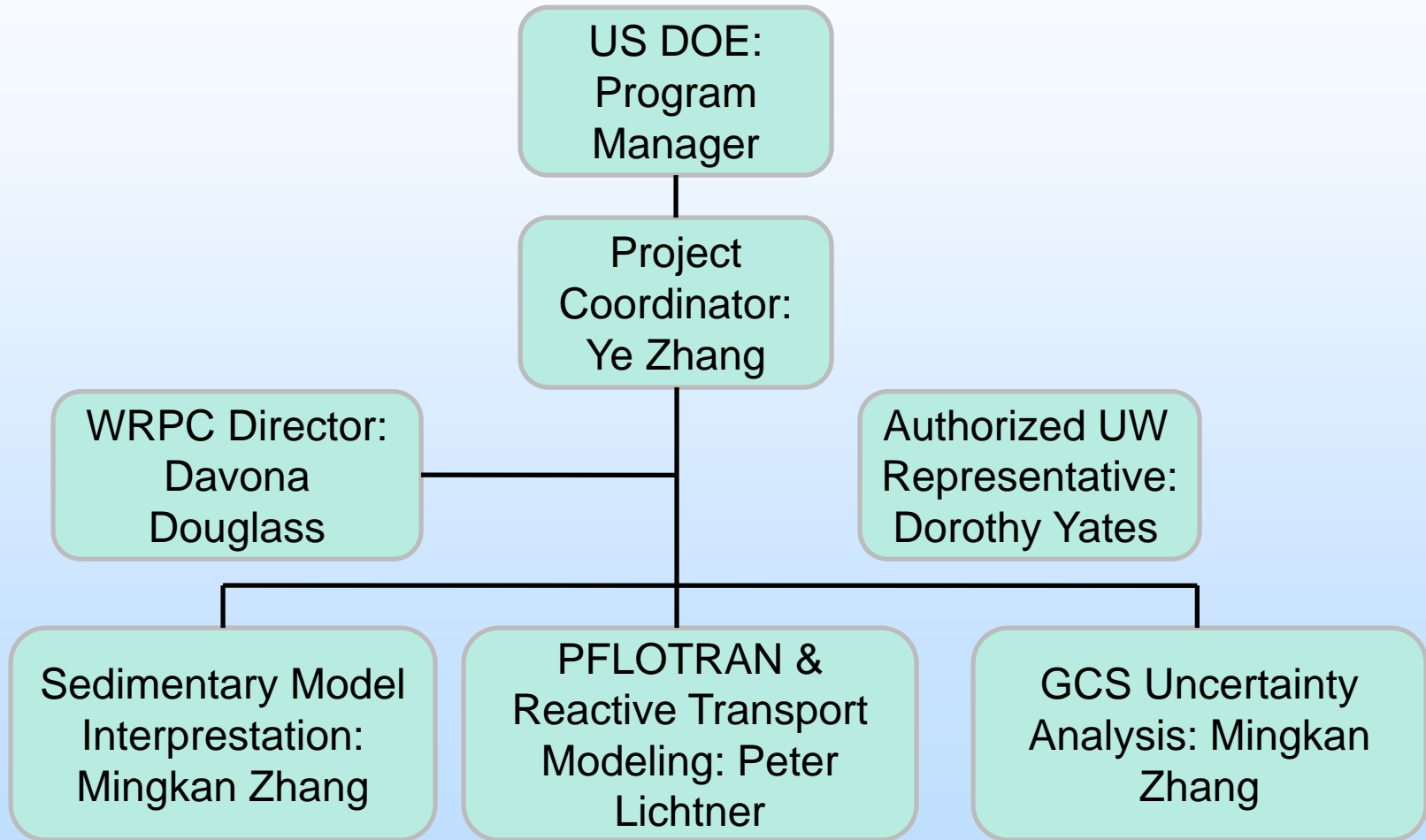
Summary

- Global upscaling computes equivalent k s 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 fingering as the FHM. When the variance of $\ln(k)$ is high, the 1-unit predicts more dissolution fingering per unit time (more optimistic dissolution storage estimate).
- Experimental design analysis suggests that brine salinity is the single most influential factor impacting CO_2 dissolution storage.
- Reactions between cations and CO_2 forms carbonate mineral precipitates (i.e., Siderite and Magnesite), leading to mineral storage. But, high degree of uncertainty exists in its prediction.
- **Next step: For low and high variance systems, complete the DoE and RS analysis for all models with reactions to compare their parameter sensitivity & prediction uncertainty.**

Appendix

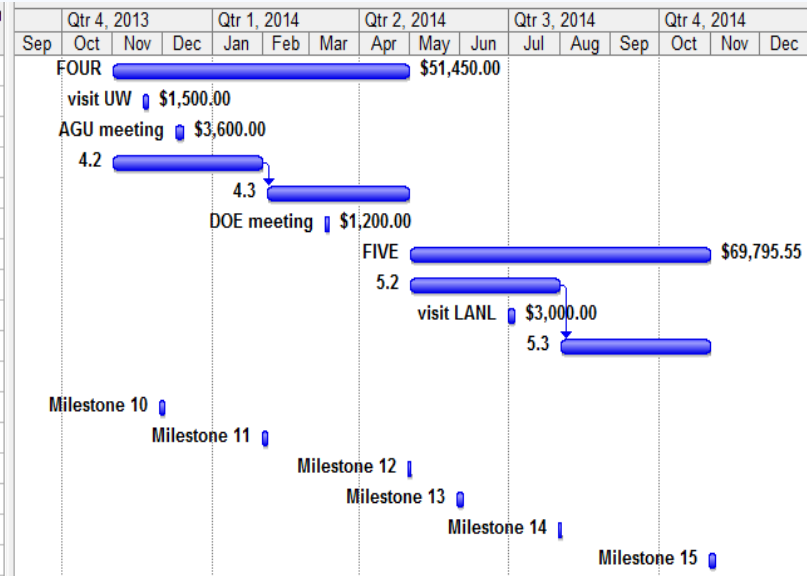
- These slides will not be discussed during the presentation, **but are mandatory**

Organization Chart

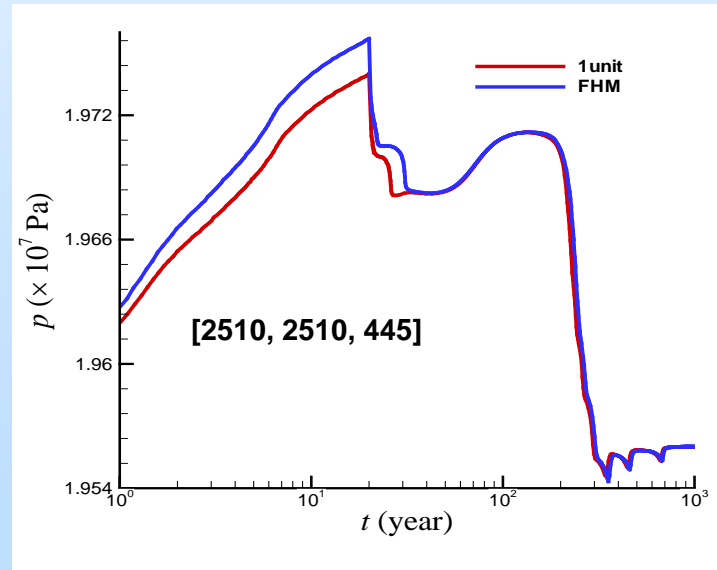
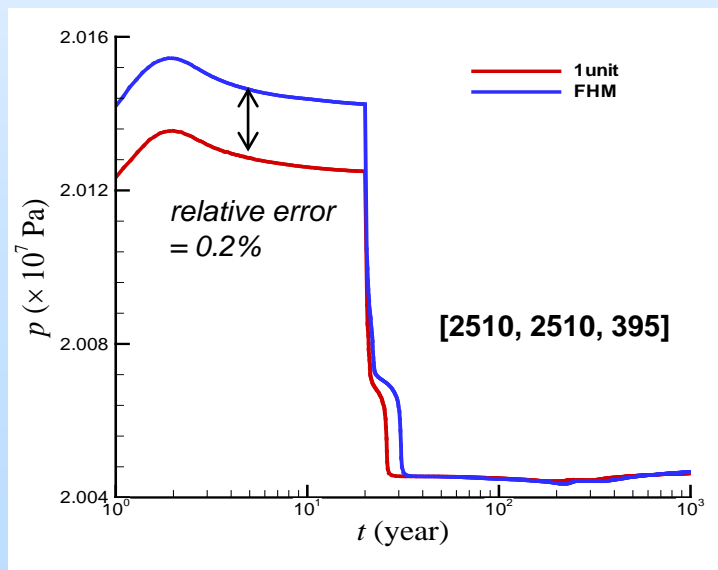
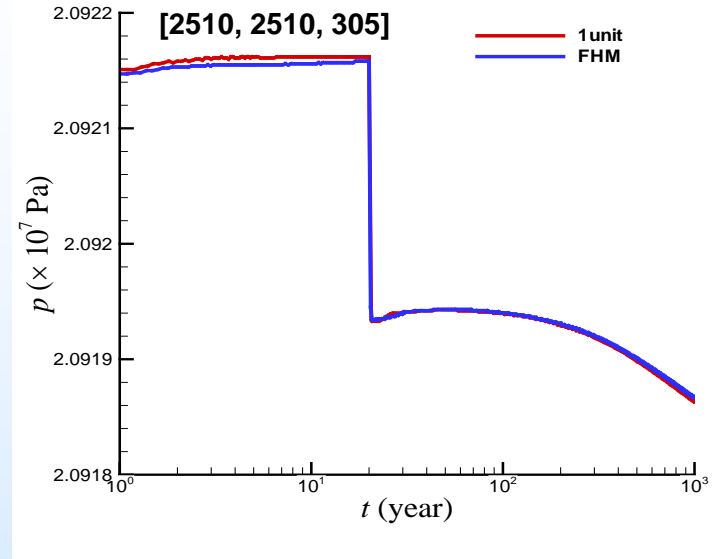
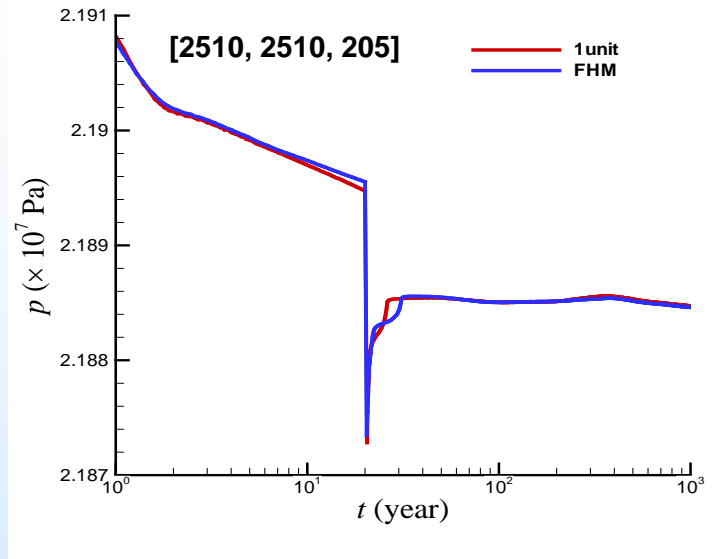


Gantt Chart

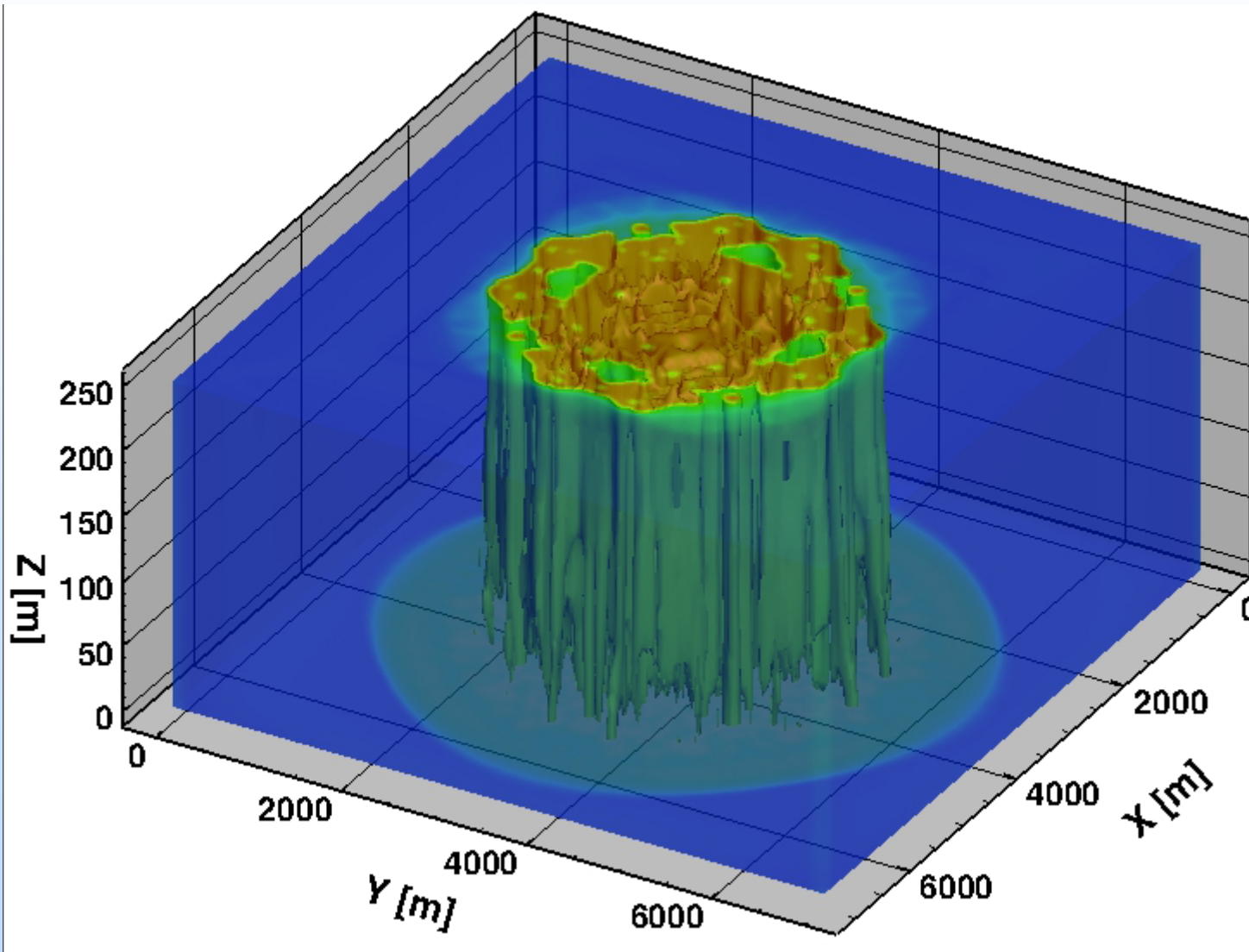
Budget Period	Task Name	Duration	Start	Finish	Travel	PI + Co-PI	Postdoc	Personnel+Tr
2								
FOUR	Uncertainty Analysis -- Fast reacting gec	129 days	Fri 11/1/13	Wed 4/30/14	\$0.00	\$15,000.00	\$36,450.00	\$51,450.00
visit UW	Co-PI visit UW	4 days	Tue 11/19/13	Fri 11/22/13	\$1,500.00	\$0.00	\$0.00	\$1,500.00
AGU meeting	Project team convene in AGU	5 days	Mon 12/9/13	Fri 12/13/13	\$3,600.00	\$0.00	\$0.00	\$3,600.00
4.2	Conduct a screening DoE analysis	66 days	Fri 11/1/13	Fri 1/31/14	\$0.00	\$0.00	\$0.00	\$0.00
4.3	Conduct a RS analysis, verification, MC :	63 days	Mon 2/3/14	Wed 4/30/14	\$0.00	\$0.00	\$0.00	\$0.00
DOE meeting	PI and/or Co-PI visit DOE NETL	3 days?	Mon 3/10/14	Wed 3/12/14	\$1,200.00	\$0.00	\$0.00	\$1,200.00
FIVE	Uncertainty Analysis -- slow reacting gec	132 days	Thu 5/1/14	Fri 10/31/14	\$0.00	\$33,345.55	\$36,450.00	\$69,795.55
5.2	Conduct a screening DoE analysis	66 days	Thu 5/1/14	Thu 7/31/14	\$0.00	\$0.00	\$0.00	\$0.00
visit LANL	PI & postdoc visiting Co-PI	5 days?	Mon 6/30/14	Fri 7/4/14	\$3,000.00	\$0.00	\$0.00	\$3,000.00
5.3	Conduct a RS analysis, verification, MC :	66 days	Fri 8/1/14	Fri 10/31/14	\$0.00	\$0.00	\$0.00	\$0.00
Milestone 10	Updated PMP	2 days	Fri 11/29/13	Mon 12/2/13	\$0.00	\$0.00	\$0.00	\$0.00
Milestone 11	Completion of Task 4.2	2 days	Fri 1/31/14	Mon 2/3/14	\$0.00	\$0.00	\$0.00	\$0.00
Milestone 12	Completion of Task 4.3	2 days	Wed 4/30/14	Thu 5/1/14	\$0.00	\$0.00	\$0.00	\$0.00
Milestone 13	Updated PMP	2 days	Fri 5/30/14	Mon 6/2/14	\$0.00	\$0.00	\$0.00	\$0.00
Milestone 14	Completion of Task 5.2	2 days	Thu 7/31/14	Fri 8/1/14	\$0.00	\$0.00	\$0.00	\$0.00
Milestone 15	Completion of Task 5.3	2 days	Fri 10/31/14	Mon 11/3/14	\$0.00	\$0.00	\$0.00	\$0.00



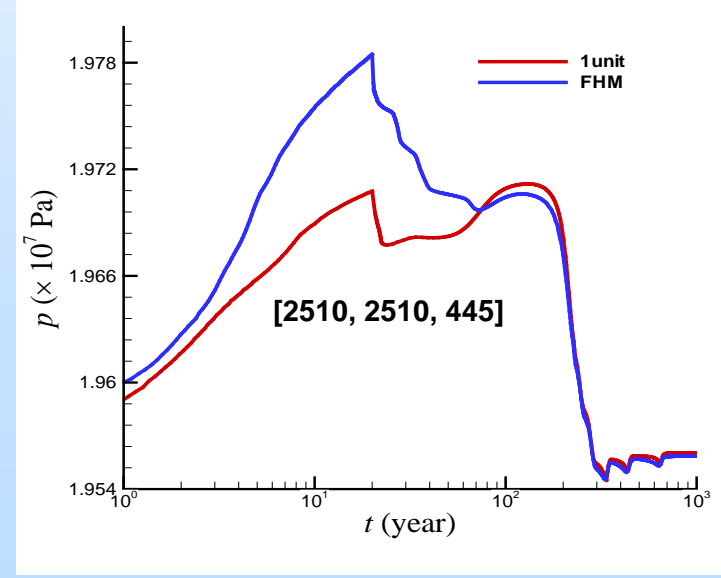
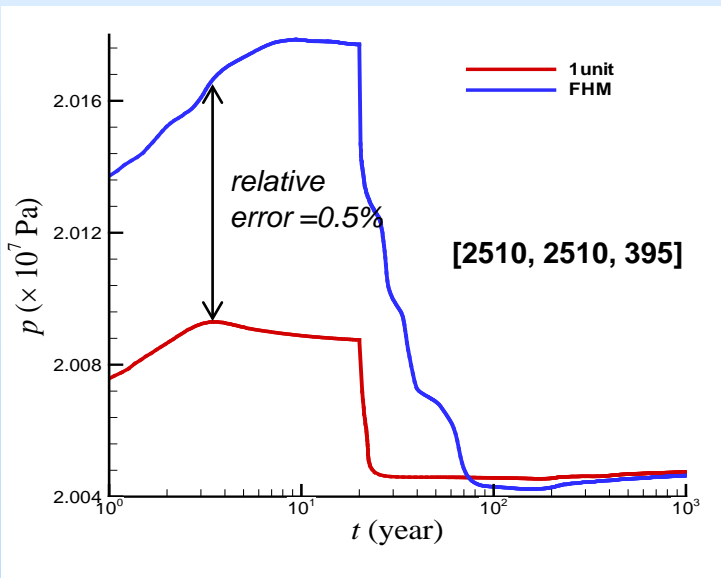
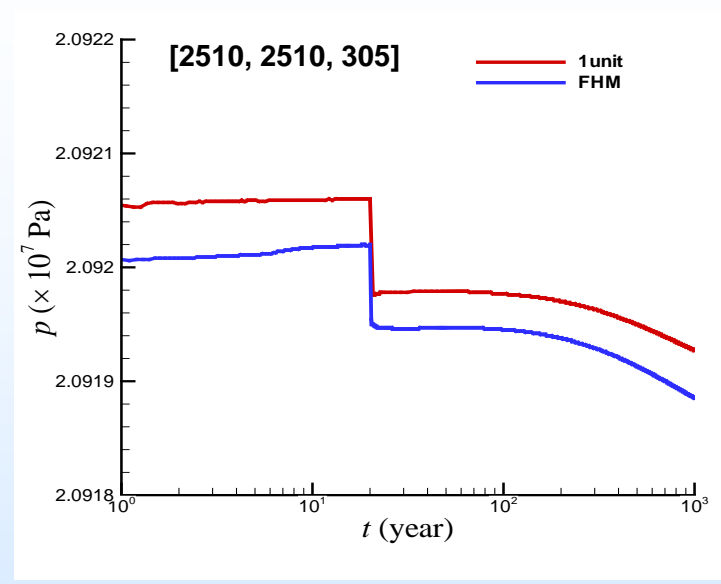
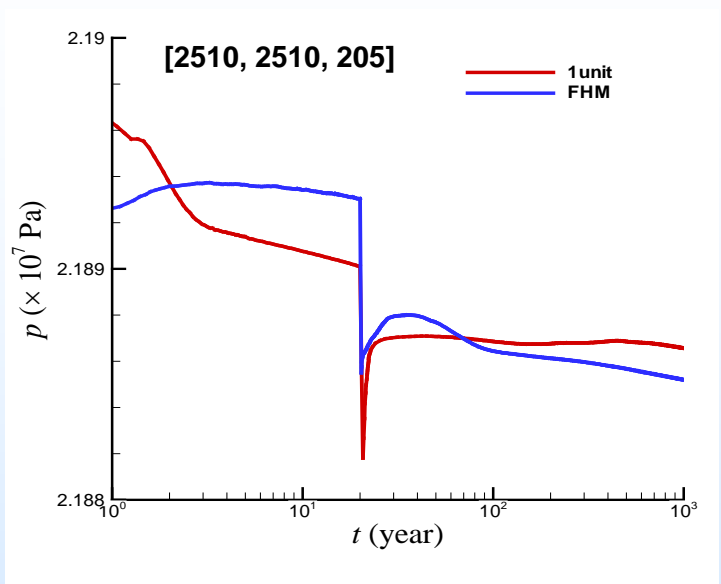
FHM v. 1-Unit Model: $\sigma^2_{\ln k}=0.1$



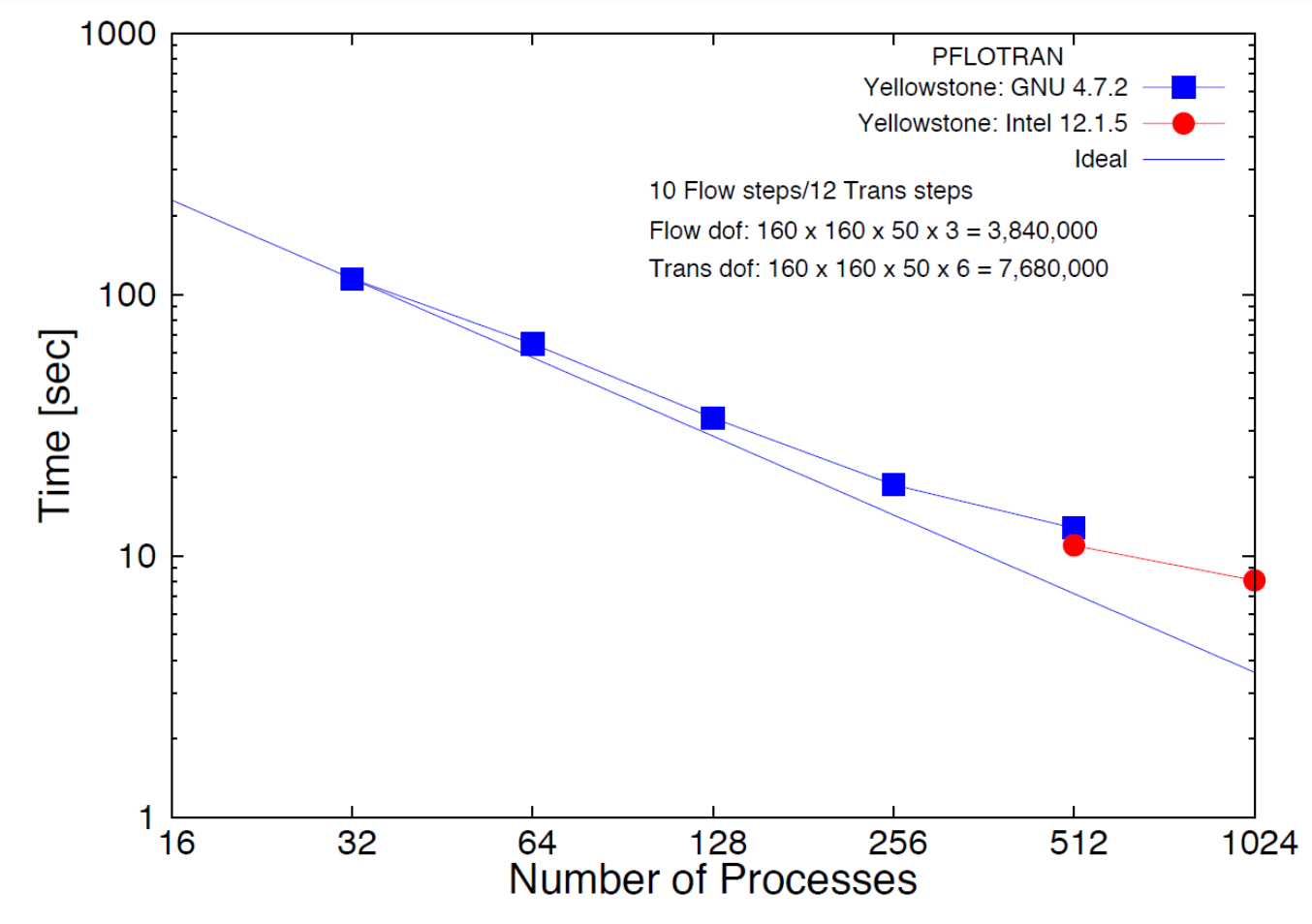
An *example 1-Unit model* run for CO₂ storage modeling simulated on the Yellowstone supercomputer. The problem domain is 7000 m x 7000 m x 250 m. Shown at 100 years for an isosurface of 0.0125 (mole fraction) of dissolved CO₂. CO₂ is injected at a depth of 50 m below the top at the center of the xy-domain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



FHM v. 1-Unit Model: $\sigma^2_{\ln k} = 4.5$



PFLOTRAN Scaling on Yellowstone



PFLOTRAN formulations

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

$$\frac{\partial}{\partial t} \left[\varphi \sum_{\alpha} (\rho_{\alpha} s_{\alpha} X_i^{\alpha}) \right] + \nabla \cdot \sum_{\alpha} (\rho_{\alpha} X_i^{\alpha} \vec{q}_{\alpha} - \varphi \rho_{\alpha} s_{\alpha} \tau_{\alpha} D_{\alpha} \nabla X_i^{\alpha}) = S_i \quad (1)$$

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

φ denotes porosity, and $\rho_{\alpha}, s_{\alpha}, \tau_{\alpha}, D_{\alpha}, U_{\alpha}, H_{\alpha}$ 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 S_i 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: $\vec{q}_{\alpha} = -\frac{\bar{k} k_{\alpha}}{\mu_{\alpha}} (\nabla p_{\alpha} - \rho_{\alpha} g z)$, with intrinsic permeability \bar{k} , relative permeability k_{α} , fluid viscosity μ_{α} , and pressure p_{α} of phase α .