Optimization Framework for Improved CO₂ Injectivity, Storage Permanence, Monitoring, and Utilization

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Infrastructure for CCS
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Presentation Outline

- Benefit to the Program
- Project Overview
 - Task 1: Integrated Optimization Framework

The goal of this task is to develop adaptive optimization methods, and associated simulation tools, to address key challenges in CO₂ sequestration, see Tasks 2 through 5.

- Tasks 2 and 3: Injectivity, Storage Efficiency, and Permanence
 - The goal of these tasks is to utilize optimization methods to develop and provide new cost effective methods for improved injectivity, storage efficiency, and permanence for GCS projects
- Task 4: Monitoring

The goal of this task is to substantially improve current inversion capabilities to enable adaptive project control via monitoring, inversion, and optimization

- Task 5: Utilization
 - The goal of this task is to develop rigorous simulation capabilities for modeling GCS in oil reservoirs, and to use these capabilities in the optimization framework for maximum hydrocarbon recovery and maximum trapping of CO_2 (not presented today)
- Accomplishments to Date
- Project Summary

Benefit to the Program

- Tasks 1-5 provide technology that improves reservoir storage efficiency while ensuring containment
- Tasks 1-5 provide methodology that supports industry's ability to predict (or control) CO₂ storage capacity in geologic formations to within ±30 percent and develop technologies to ensure 99% storage permanence

Project Overview Task 1:

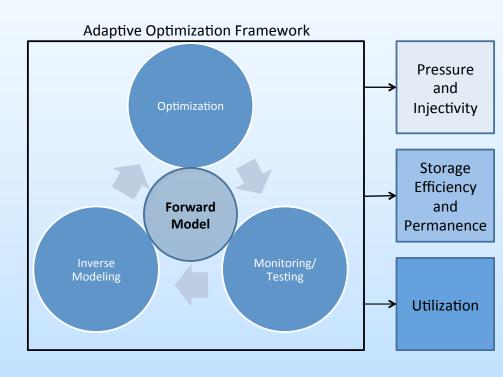
Optimization Framework and Integrated Management

Objectives

- Develop and apply an adaptive optimization framework for CO₂ storage projects which utilizes advanced automated optimization algorithms and suitable process models for coupled multi-phase behavior,
- Use this new framework to investigate process behavior and then develop optimal schemes for key technical challenges in CO₂ sequestration,
- Demonstrate applications for ongoing partnership or international projects and provide guidance for system optimization in future projects.

Technical Status

- Developed and demonstrated an efficient constrained global optimization methodology for realistic GCS systems
- Currently working on development of an integrated management system for adaptive control based on dynamic model updates using monitoring data



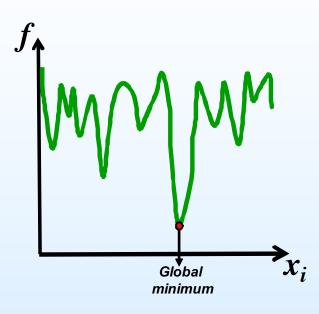
Constrained Global Optimization for Optimal Control and Placement of Wells

Motivation

- Due to complex reservoir geometry and heterogeneity in reservoir properties, objective functions tend to be highly irregular with multiple local optima in the solution space
- The irregularity of objective functions can easily cause gradient-based methods to be stuck in a local minimum

Objective

 Develop a constrained global optimization method, test for optimizing CO₂ injection and brine extraction (well placement and rates), and modify/apply to broad set of optimization scenarios



Approach

- Define specific reservoir performance criteria (e.g., maximum pressure, minimum AoR size, maximum storage efficiency, maximum trapping, minimum leakage rate)
- Via smart optimization algorithms, automatically find most suitable CO₂ injection and brine extraction solutions that meet performance criteria and constraints
- Optimization searches for optimum number of wells, well locations, and rates

Example Description of a Constrained Global Optimization Problem

A general constrained optimization problem for well placement, injection and extraction for maximum pressure control can be formally expressed as

Minimize
$$f(\mathbf{x}_{ext}, \mathbf{y}_{ext}, \mathbf{p}) = \frac{V_{ext}}{V_{inj}}$$

Brine extraction Ratio (if no brine extraction, f is chosen as $1/V_{inj}$)

Subject to
$$\Delta P_{th} - \max \left\{ \Delta P(x_f, y_f, t) \right\} > 0$$

Pressure buildup constraints, e.g. max buildup along a fault

$$V_{ext,CO_2} = 0$$

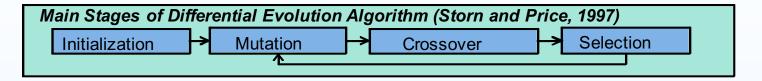
No CO₂ flow into extraction wells (for brine extraction schemes)

$$a_{1} \leq x_{inj} \leq b_{1}, a_{2} \leq y_{inj} \leq b_{2}$$

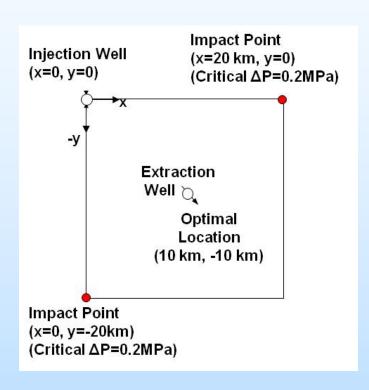
$$a_{4} \leq x_{ext} \leq b_{4}, a_{5} \leq y_{ext,j} \leq b_{5}$$

Range of Cartesian coordinates for vertical wells (for partially penetrating and horizontal wells, additional parameters can be included)

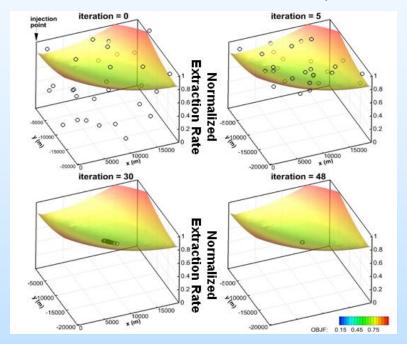
Constrained Differential Evolution Algorithm



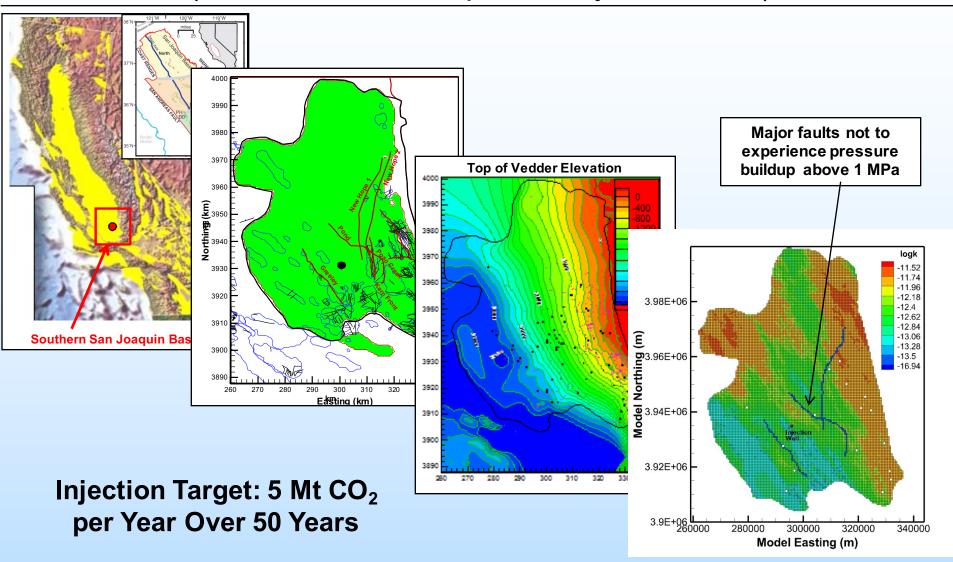
A Simple Optimization Problem for Demonstration



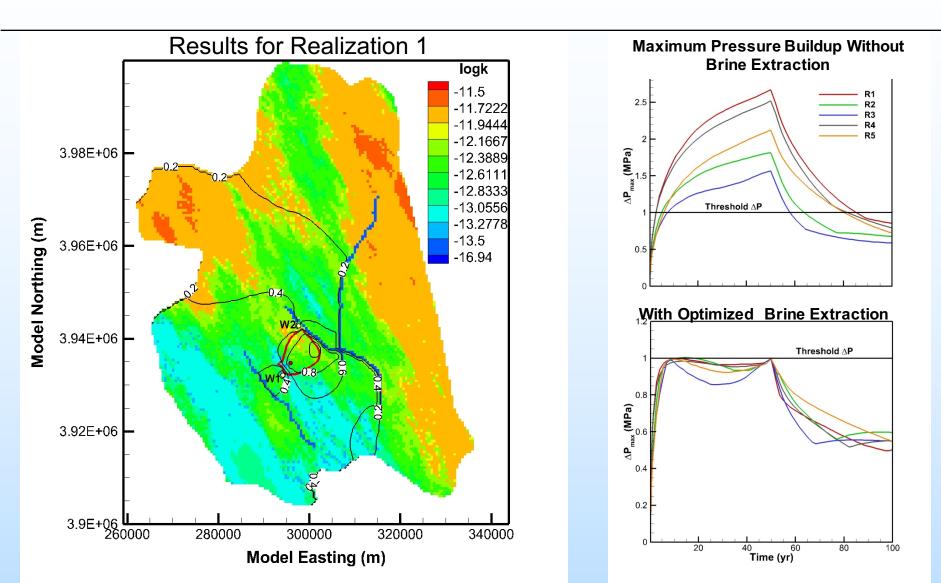
Evolution of the CDE solutions (the zones below the isosurface line are infeasible)



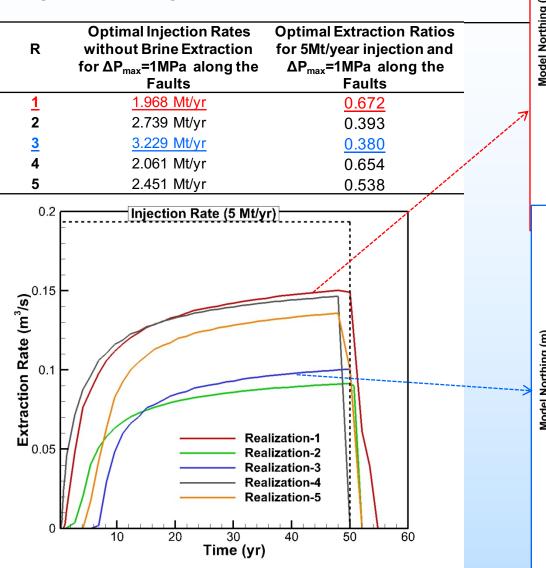
Application to Hypothetical CO₂ Injection in the Vedder Formation (Southern San Joaquin Valley, California)

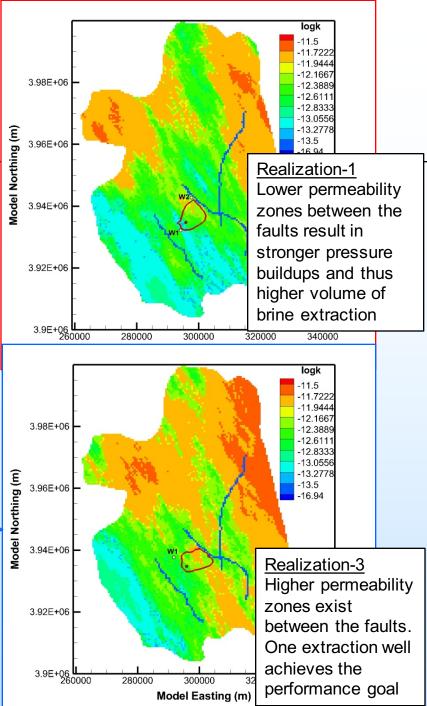


Task 1:
Optimal Well Placement and Pressure-Buildup Changes

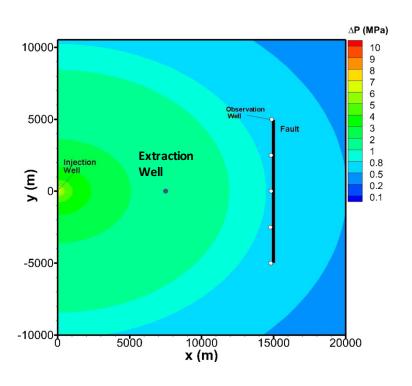


Heterogeneity plays a significant role for optimal well placement and extraction rate

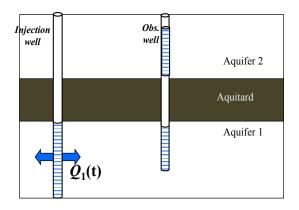




Task 1: An Example Application of Adaptive Approach to Project Control and Optimization

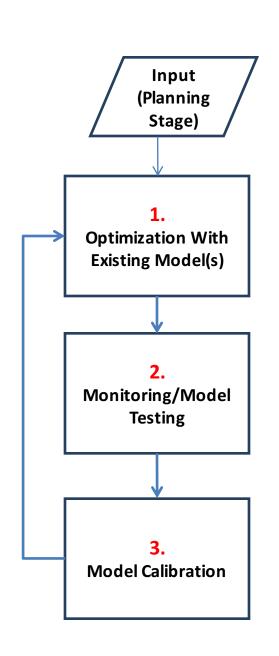


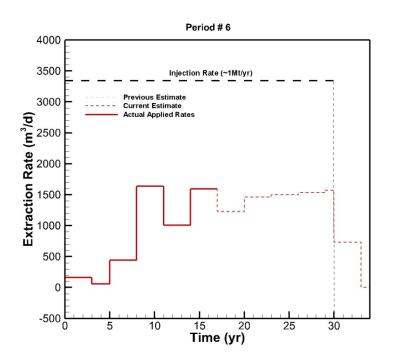
Parameters assumed to be Unknown	Actual Values
K1 (m/d)	1.015E-01
S1 (Specific Storage Coef, 1/m)	1.693E-06
Horizontal Anisotropy Ratio	2.000E+00
K2 (m/d)	1.037E-01
S2 (1/m)	1.704E-06
K' (aquitard)	1.028E-06
S' (aquitard)	1.988E-06

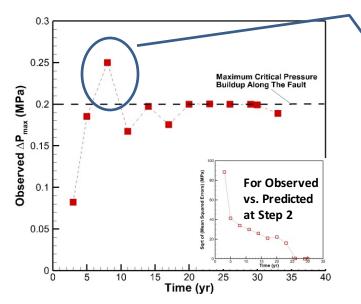


Objective:

- Critical pressure buildup is assumed to be 0.2 MPa along the fault
- Fluid injection rate is 1Mt/yr over 30 yrs
- A adaptive optimization algorithm involving optimization +monitoring + model update is applied to manage the system, minimizing extraction rates while attempting to reduce risk of fault slippage.







High pressure in violation of constraint indicates:

- Importance of site characterization at the planning stage
- Monitoring and model update need to be done at a higher temporal frequency at earlier times to detect higher pressures and change extraction rates accordingly

Project Overview Tasks 2 and 3:

Injectivity, Storage Efficiency, and Permanence

Objectives

- Investigate and optimize options of ensuring and improving injectivity, including well
 placement, number and design, novel adaptive injection schemes, combined CO₂
 injection and brine extraction, as well as localized stimulation to be able to inject target
 amount of CO₂ into reservoirs with natural heterogeneity and moderate permeability,
- Investigate and optimize injection/extraction schemes utilizing trapping mechanisms and natural heterogeneity for enhanced sweep efficiency and residual trapping of CO₂
- Apply adaptive injection/extraction control methods based on continuously collected field data for improving storage and developing mitigation plans
- Demonstrate the developed methods for suitable field sites/partners.

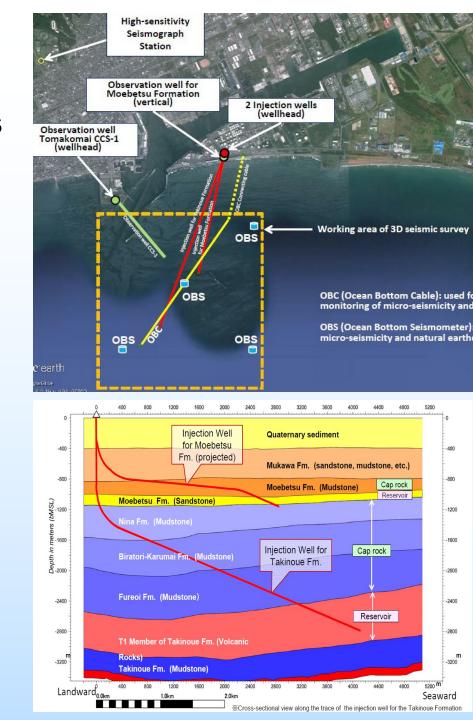
Technical Status

- Expanded the application of the CDE optimization framework for improved injectivity studies and tested for generic case studies.
- Developed new modeling capabilities for efficiently simulating hysteresis in two-phase flow and started investigating optimal control methods for enhanced storage, trapping and plume stability
- Selected Japanese CO₂ pilot field test at Tomakomai for demonstration of CDE-based optimization tools and initiated partnership with the Research Institute of Innovative Technology for the Earth (RITE) in Japan.

Tasks 2 and 3:

Field Demonstration of the Optimization Methods and Tools

- Tomakomai CCS demonstration project is planned to start at the beginning of 2016.
- Initially, the scCO₂ injection is planned for 3 years at a rate of 100Kt/yr. However, LBNL and RITE will investigate for future larger injection rates and durations using our optimization methods.
- The scCO2 will be injected into two storage layers
 - The upper injection zone sandstone layers of Moebetsu Formation, about 1000-1200 m depth below the seabed. Porosity ~ 0.2-0.4 and permeability ~ 9-25mD.
 - The lower injection zone volcanic and volcaniclastic rocks, about 600m thick and at 2400-3000m below the seabed. Porosity ~ 0.03-0.19 and permeability~ 0.01mD-2.6mD.
- The injection site is located nearby faults and the lower injection zone is known to have a very low permeability,
- Concerns about possible injection-induced seismic effects and low injectivity require reservoir pressure management methods for optimal control of the injection operation and brine extraction. Optimization studies will be conducted by LBNL.



Project Overview Task 4:

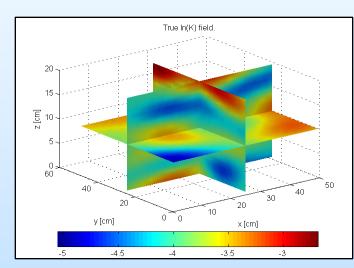
Monitoring and Inverse Modeling

Objectives

- Develop intelligent data analysis and inversion approaches that allow for fast interpretation and analysis
- As a first step, implement advanced stochastic inversion toolset into optimization framework and demonstrate applicability of toolset in real storage conditions
- Long-term goal is to enable adaptive project control involving monitoring, analysis, efficient joint inversion, and optimization

Technical Status

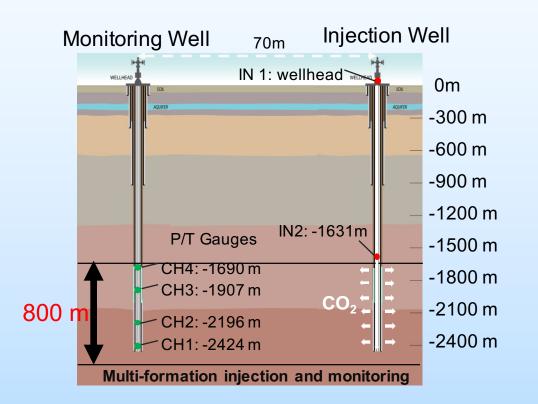
- Tested different stochastic inversion tools and demonstrated use for Ketzin Project (Liu et al., 2013, 2014); implementation into the optimization framework is ongoing
- Completed data collection and analysis for the Shenhua Ordos CCS Project in China
- Currently conducting different types of inverse modeling to demonstrate applicability and use of inverse modeling tools to the Shenhua Project

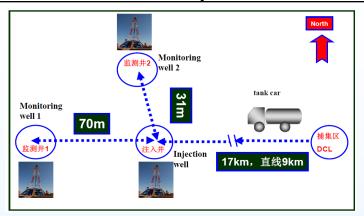


Task 4:

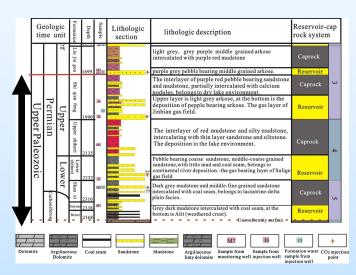
China's Shenhua Ordos CCS Project: Site Description

- Four main low-permeability storage units (< 5 md), with multiple layers, ranging in depth between 1680 m and 2520 m;
- Injection well perforated in 18 layers, with a total thickness of 88 m over the 800 m thick sandstone/mudstone sequence.





One injection well and two monitoring wells and CO₂ source from a DCL plant

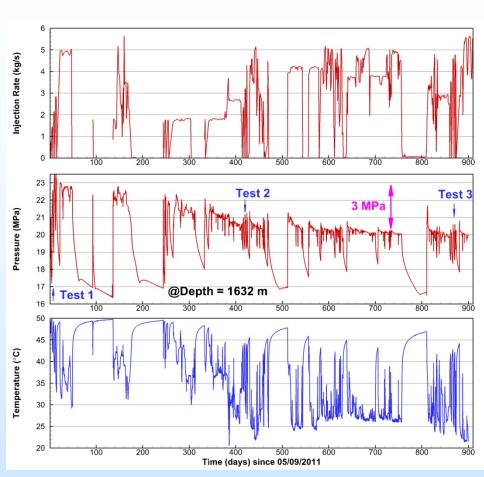


Storage complex: multiple layers for injection (88 m) and storage (800 m)

Task 4:

CO₂ Injection: Unique Features

- Injection of liquid CO₂ at 0 °C or less at ~0.1 Mt/year, with cumulative injected CO₂ mass of 0.280 Mt from May 2011 to August 2014
- Bottomhole injection pressure decreases with time by more than 30 bar, indicating significantly enhanced injectivity;
- Bottomhole temperature decreases by 30 °C under injection conditions;
- Three step-rate injection/shut-in tests were conducted, once each year, to characterize the dynamics of the storage system.

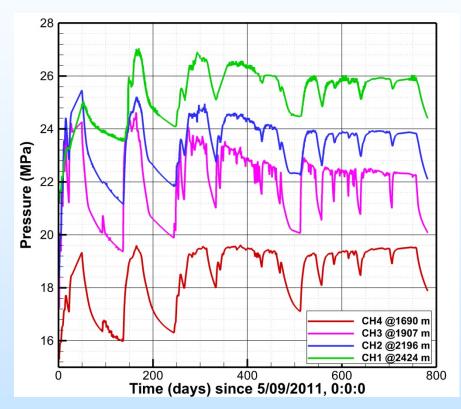


Bottomhole (@1631 m) pressure and temperature at the injection well (data shared from Beijing Normal University, China)

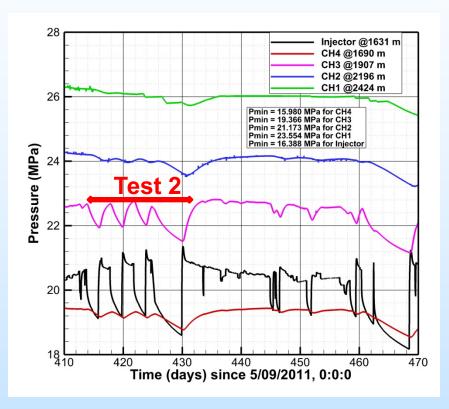
Task 4:

Multi-formation Pressure Monitoring: Operation and Tests

- Multi-formation pressure monitoring during injection operation and step-rate injection/shut-in tests provides useful information
- Inverse modeling is under way to (1) estimate formation permeability and dynamics of two-phase flow, and (2) explore causes of enhanced injectivity



Multi-formation pressure responses during injection operation at the monitoring well 70 m from the injector



Multi-formation pressure responses to injection pressure (black) during the 2nd test

Accomplishments to Date

Task 1: Integrated Optimization Framework

- Developed a new constrained global optimization method involving injection/extraction well placement and rates, and applied the global optimization methodology to realistic complex problems involving pressure control near faults and other management objectives (e.g., improved injectivity and storage)
- Adaptive optimization capabilities for integrated optimization framework currently under development

Tasks 2 and 3: Injectivity, Storage Efficiency, and Permanence

- Expanded the application of the optimization method for improved injectivity and tested its applicability successfully for generic case studies
- Developed new modeling capabilities for efficiently simulating hysteresis in two-phase flow to investigate optimal control methods for enhanced storage, trapping and plume stability
- Identified Japanese CO₂ pilot field test at Tomakomai for demonstration of CDE optimization tools and initiated partnership with RITE in Japan

Task 4: Monitoring and Inverse Modeling

- Implemented and tested different stochastic inversion tools into the unified optimization framework
- Completed data collection and analysis for the Shenhua Ordos Project

Summary

Key Findings / Lessons Learned from Modeling Tasks

- The constrained differential evolution algorithm is very well suited for solving complex and challenging storage management optimization problems
- Heterogeneity plays a significant role for optimal well placement and injection/extraction rate control, which points to the importance of site characterization and use of adaptive management strategies combining monitoring+inversion+optimization in an integrated framework
- Advanced stochastic inversion algorithms have notable potential for joint analysis of large data sets and will be useful for adaptive optimization frameworks

Future Plans

- Long-term plan is to develop and demonstrate a broad and flexible GCS optimization framework for improved injectivity, storage efficiency and permanence, monitoring, and utilization for ongoing or planned real project sites
- This planned work builds on developments made in recent years, such as the advanced global and local search algorithms for pressure management, and the advanced stochastic joint inversion methods for fast and intelligent analyses/processing of field data
- The optimization framework, including constrained global and local optimization methods, will allow for adaptive control, in the sense that storage management decisions can be revised from time to time using feedback from monitoring data

Appendix

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

Appendix: Organization Chart

- "Optimization Framework for Improved CO₂ Injectivity, Storage Permanence, Monitoring, and Utilization" is a subtask of LBNL's Core Carbon Storage and Monitoring Research program
- "Optimization Framework" has five main tasks with principal investigators identified as PI
 - Task 1: Jens Birkholzer (PI), Abdullah Cihan, Stefan Finsterle
 - Task 2: Abdullah Cihan (PI)
 - Task 3: Abdullah Cihan (PI)
 - Task 4: Quanlin Zhou (PI), Xiaoli Liu
 - Task 5: Curtis Oldenburg (PI) and Lehua Pan

List of scientific staff

Name	Title	Role in Task/Subtask		
Jens Birkholzer	PI and Senior Scientist	Lead scientist for Modeling CO ₂ Processes		
Abdullah Cihan	PI and Research Scientist	Lead scientist working on storage management and optimization		
Curtis Oldenburg	PI and Senior Scientist	Lead scientist for CO ₂ -EOR simulations		
Quanlin Zhou	PI and Research Scientist	Lead scientist working on stochastic inversion		
Xiaoyi Liu	Postdoctoral researcher	Main scientist working on stochastic inversion		
Stefan Finsterle	Research Scientist	Main scientist working on integrated optimization		
		framework		
Lehua Pan	Research Scientist	Main scientist working on CO ₂ -EOR simulations		

Appendix: Gantt Chart for FY15

Fiscal Year	FY15			
Task/Milestone Quarter	Q1	Q2	Q3	Q4
Task 1: Integrated Optimization Framework				
Identify suitable field sites/partners		G		
Develop methodology for adaptive management				Н
Task 2: Injectivity				
Develop optimization methods for improved injectivity				
Test and demonstrate methods for generic and field scenarios				
Task 3: Storage and Permanence				
Improve models for simulating hysteresis for storage and trapping optimization				
Investigate methods for injection/extraction schemes for enchanced storage				
Task 4: Monitoring and Inverse Modeling				
Implement and test advanced stochastic inversion tools				
Apply inverse modeling tools to site-specific CCS projects				
Task 5: Utilization				
Implement approaches for high-pressure solubilities of water-oil-CO ₂ mixtures				
Testing of the new T2 CO2-EOR code against published results of solubility				20

Appendix: Milestone Log for FY15

Task 1: Integrated Optimization Framework

- Milestone 4-1 (G), Q2 (3/31/15)

Title: Identify suitable field sites/partners for demonstration of optimization toolset.

- Milestone 4-2 (H), Q4 (9/30/15)

Title: Identify suitable field sites/partners for demonstration of optimization toolset.

Task 2: Injectivity

- No milestone in FY15

Task 3: Storage and Permanence

- No milestone in FY15

Task 4: Monitoring and Inverse Modeling

- No milestone in FY15

Task 5: Utilization

- No milestone in FY15

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