## Ensuring CO<sub>2</sub> Storage and Groundwater Protection Through Water Monitoring: Applying Geochemical and Statistical Tools



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# **Carbon Sequestration**

# **Importance of Monitoring**

- Ensures CO<sub>2</sub> storage
- Protects valuable assets
- Provides assurance to community

# **Geochemical Monitoring for MVA**

- Utilizes existing infrastructure
- Employs developed tools
- Provides source attribution







# **Field Area**

- CO<sub>2</sub> EOR as an analog ٠
- Sampled oil field in Permian Basin, West Texas ٠
  - June 2013 to April 2018 ٠
    - Eight field sampling events ٠
  - CO<sub>2</sub> EOR: October 2013
    - One pre-CO<sub>2</sub> injection event





32°44'30"N

32°44'0"N

32°43'30"N

32°43'0"N

0.25 hard and a second se 0 0.25 0.5

1 Kilometers

C1

# Field Area

## Samples

- Ogallala Fm. groundwater
- Santa Rosa Fm. groundwater
- San Andres Fm. produced water





- Developed analytical process for fossil fuel wastewaters
  - Miller et al., 2022



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# **Results: Producing Formation**

- Focus on alkalinity and dissolved inorganics
- Following CO<sub>2</sub> Injection
  - Increases in certain parameters
  - Statistical significance











# **Results: Producing Formation**

- Statistical analysis
  - Generated prediction intervals
  - Significant change: alkalinity, TDS, Na<sup>+</sup>, Cl<sup>-</sup>, and SO<sub>4</sub><sup>2-</sup>
- Interpretation
  - CO<sub>2</sub> injection caused significant changes in produced water parameters
  - However, CO<sub>2</sub> injection did not result in significant carbonate dissolution
  - Therefore, reservoir integrity was preserved during injection







# **Results: Groundwaters**



- Ogallala and Santa Rosa Fm. groundwaters
  - Did not see significant change following CO<sub>2</sub> injection







# **Results: Groundwaters**



- Ogallala and Santa Rosa Fm. groundwaters
  - Generated hypothetical mixing curve
    - Scenario: produced water migration into Santa Rosa Fm.







# **Application of Field Results: GILD**

- Problem definition
  - Need for a low-cost, easily implementable monitoring strategy for carbon storage reservoir leak detection
- Proposed solution
  - Geochemically Informed Leak Detection (GILD)







# **GILD** Overview

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## Geochemically Informed Leak Detection (GILD)

- 1) Assess fluid chemistry and mineral compositions of monitoring formation
- 2) Simulate leakage events with a geochemical reaction path model
- 3) Identify CO<sub>2</sub> leakage with a Bayesian Belief Network (BBN)



## Bayesian Belief Network (BBN)

- Decision support tool
- Probabilistic inference from multiple sources of evidence
- Application for leak detection: given monitoring parameters, compute the probabilities of the presence of leakage



# Bayesian Belief Network (BBN) for Leak Detection

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- Upstream node
  - CO<sub>2</sub> added concentration
- Arrows
  - Causal effect
- Downstream nodes
  - Monitoring parameters
- Bars of each node
  - Probability of a particular range
- Conditional probability
  - Probability of downstream given upstream
- Purpose
  - Back inference





- Observation parameters
  - Inorganic geochemistry
    - Na<sup>+</sup>, Ca<sup>2+</sup>, Mg<sup>2+</sup>, K<sup>+</sup>, Cl<sup>-</sup>, SO<sub>4</sub><sup>2-</sup>, Br<sup>-</sup>, alkalinity as mg/L HCO<sub>3</sub><sup>-</sup>
- Analytical techniques
  - Fit response functions
  - Identify sensitive parameters
  - Include uncertainty
  - Predict possible outcomes from potential leakage
  - Compare with observed fluid network
  - Compute probability of CO<sub>2</sub> leakage







## Construction of BBN-Response Functions & Sensitive Parameters



- Response functions
  - Output of the geochemical model
  - Relationship of ion concentrations (mol/kg) and CO<sub>2</sub> added concentrations (mol/kg)

- Sensitive Parameters
  - Selection Criterion
    - Monotonic relationship





# **Application of BBN-Combining Evidence**

 Evidence from high HCO<sub>3</sub> and low SO<sub>4</sub>







VerySmall

Small

Large

Medium

CO2

2.11

6.09

4.20

0.005 to 0.035

0.035 to 0.065

0.065 to 0.095

0.095 to 0.125

0.125 to 0.155

mean HCO3

1.74

0.0097217 to 0.0177152

0.0177152 to 0.022547

0.022547 to 0.0248275

CO2 Leakage Category

2.76

28.9

58.0

14.4

25.4 79.8 1.8

42.0

43.7

SD HCO3

0.0114 20.0042

\$D\_\$04

0.00119 20.00062

192381e-5 to 0.00807538

0.00807538 to 0.0118462

1.43033e-5 to 8.50817e-4

8.50817e-4 to 0.00126413

0.00126413 to 0.0024996

0.0118462 to 0.0183833

# Conclusions

- Geochemical monitoring provides insight into groundwaters and target formation reactions
  - No significant dissolution of storage reservoir
  - No produced water migration detected in groundwaters
- Geochemical-statistical model (GILD) can provide CO<sub>2</sub> leakage detection via robust statistical analysis
  - Current model applies user input via licensed software and researcher knowledge
  - Goal is to create standalone version that groundwater observers can use











### **References**

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# NETL Resources

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# **Supplemental Slides**



# **Field Sampling**





Fluid from the well (e.g. oil, gas condensate, produced water,  $H_2S$  gas) flows into the carboy and excess gas flows out.



# **Construction of BBN-Regression Results**

- Mean-predicted value from the response function for each CO<sub>2</sub> level
- SD-standard deviation of residuals (geochemical model output-predicted values)
- Ion concentration-normally distributed with mean and SD





