Hydrocarbon Groundwater Plume Stability and Persistence: Lessons learned from 32 sites and 5 million data points

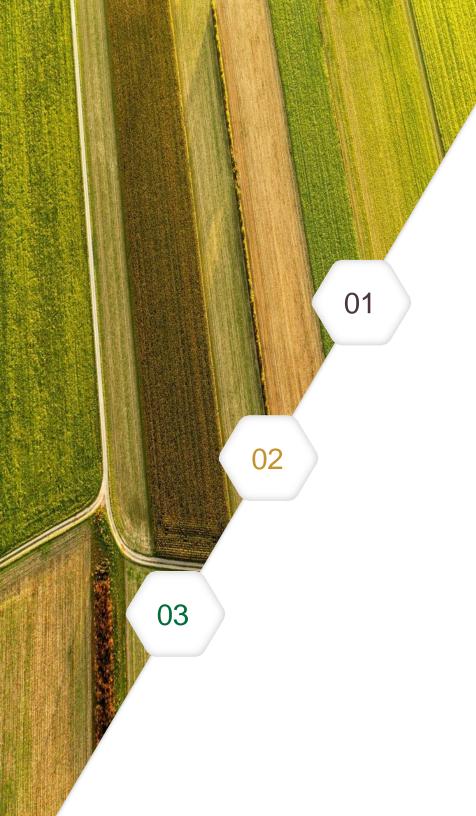
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Today's Roadmap:

- 1. GW sampling and variability
- 2. Autonomous sensors and benzene trends
- 3. Why GW models disagree and solutions
- 4. GW Models: LiORA Trends and Insights



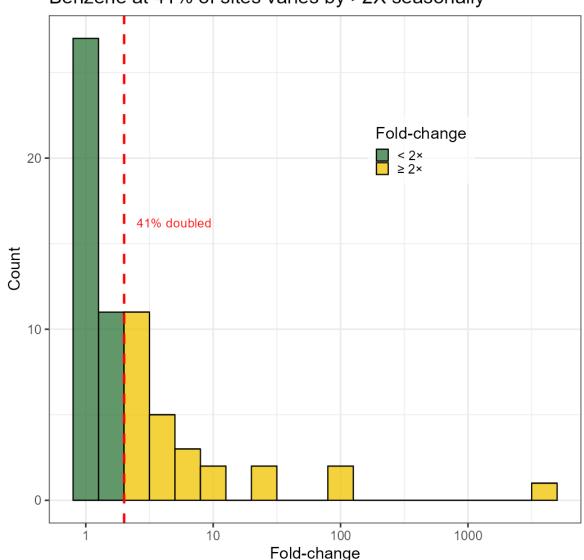
How do we measure groundwater plumes?

We measure dissolved concentrations once or twice per year.

Is that the best approach?

GW Sampling: Same well, same season, same result?

Benzene at 41% of sites varies by >2X seasonally



Benzene across 11 sites with groundwater wells sampled at least twice during a season

- Across 64 wells and 138 samples, only 28% were the same (no variance)
- Only half of subsequent samples were within 50% of the previous measurement
- 41% varied by a factor of two or more!



Improving Groundwater Models with Autonomous Sensors

LiORA's autonomous sensors continuously measure contamination levels in groundwater.

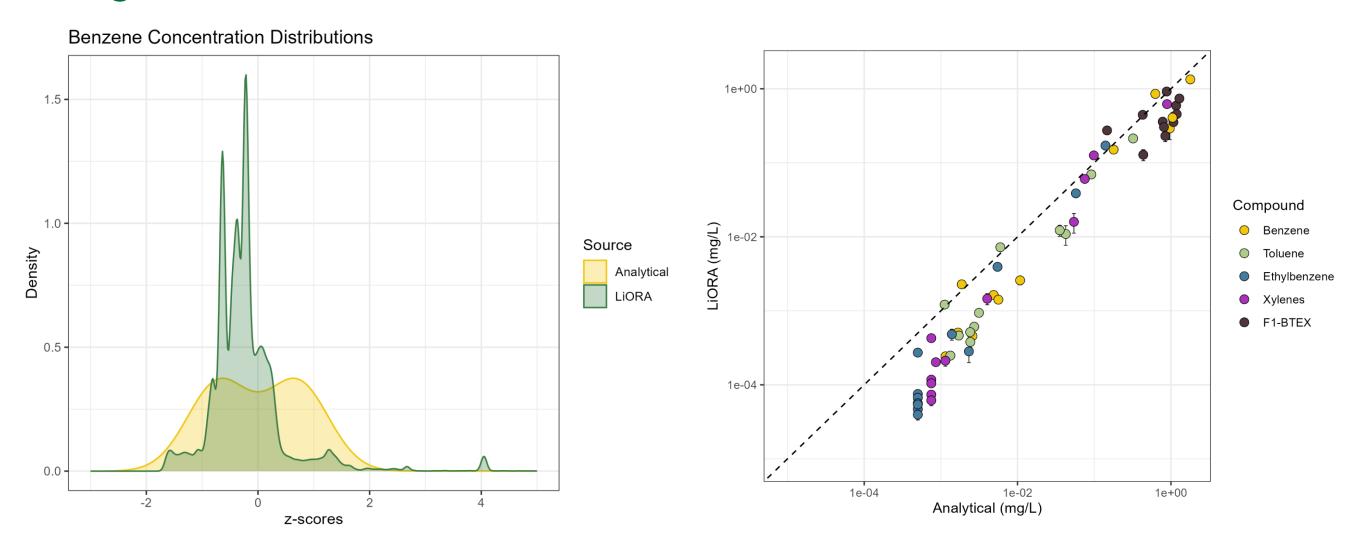
Data are measured, transferred, and stored automatically.

Continuous sensor data updates groundwater models for improved risk assessment.

LiORA Water Sensor measures: Vapour: CO₂, CH₄, total hydrocarbons Other:

Temperature and pressure.

Are sensors and analytical telling the same story?

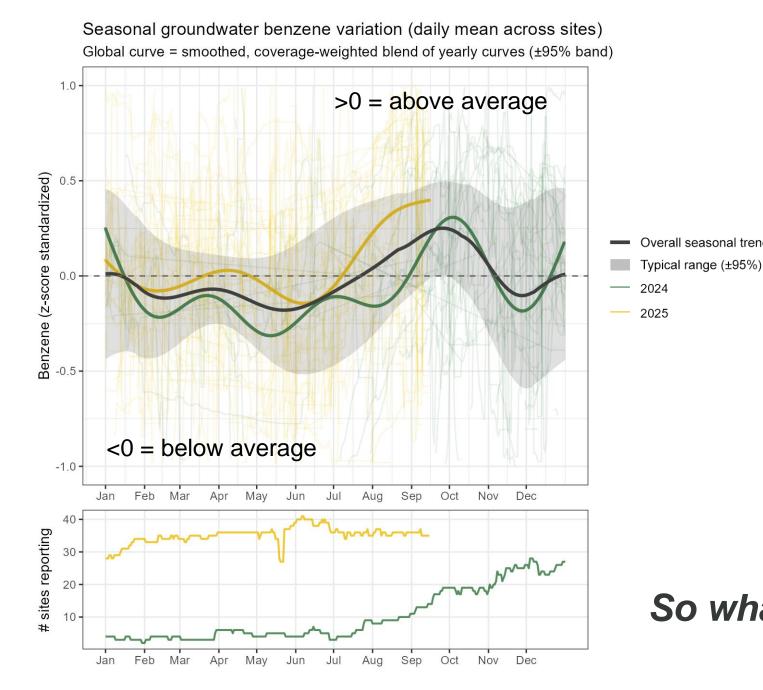


So what? Repeated, continuous measurements at each location = increased concentration confidence

Groundwater benzene varies seasonally

Overall seasonal trend

2024



How does groundwater benzene vary seasonally?

- 42 sites (2024 & 2025)
- 5.7M half-hourly data points
- Lowest in spring, greatest in autumn

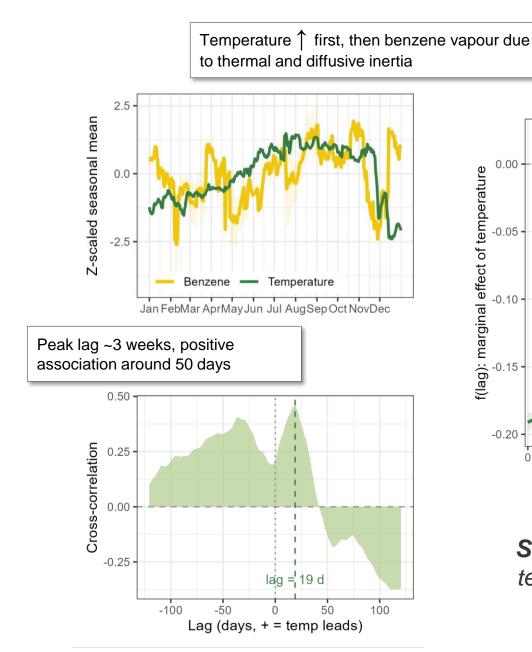
Implications?

Depending on when you sample, may not be directly comparable to previous measurements without seasonal context

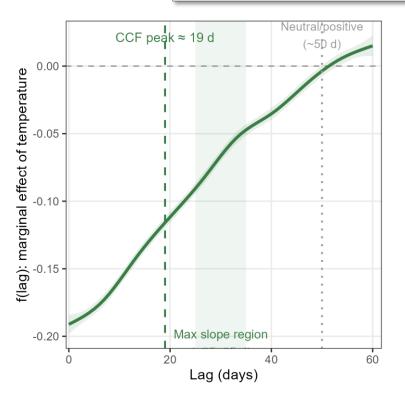
So what? Clear seasonal cycle confounds point in time sampling

Benzene lags temperature

LiORA GW Sensors



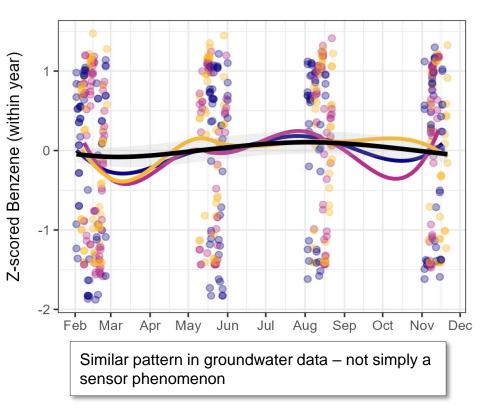
Lagged temperature effect confirms delayed benzene response



Groundwater PIT Data

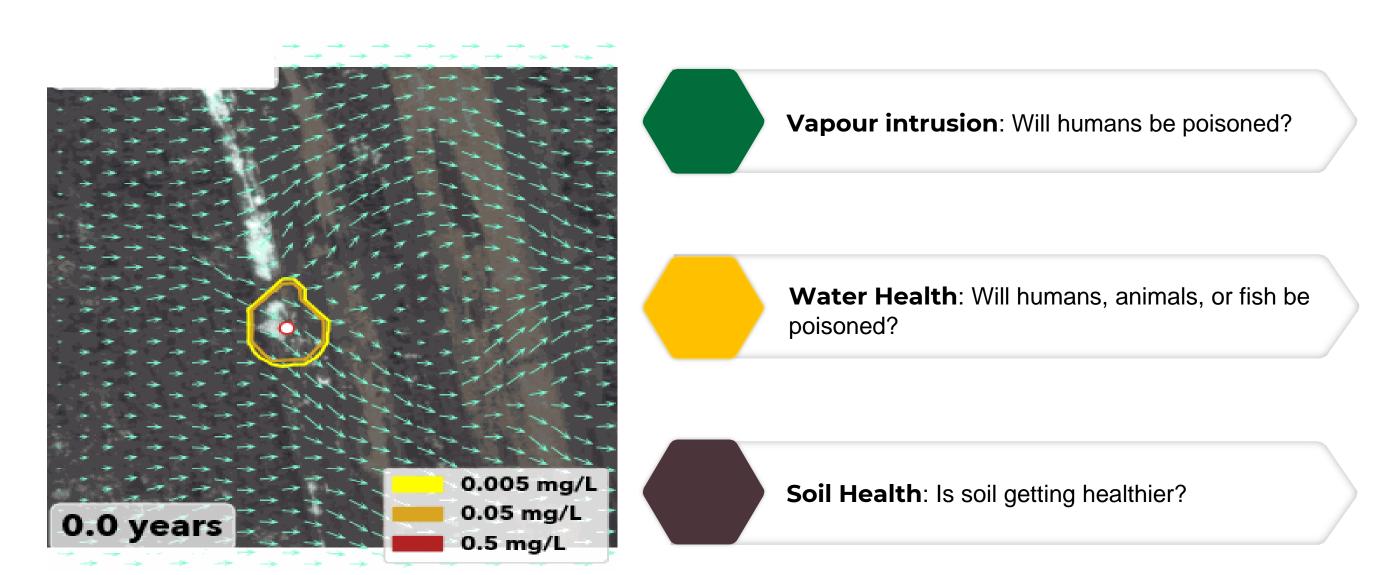
Seasonal variation in dissolved benzene
Black line = multi-year mean





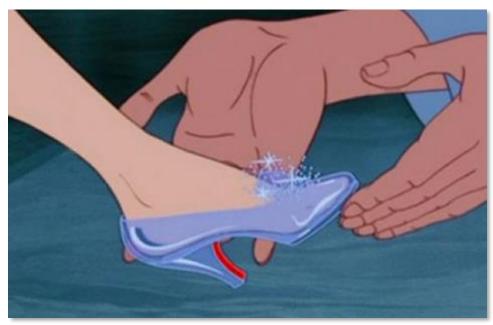
So what? Recognizing this lag helps separate true concentration changes from temperature-driven effects, improving how we interpret seasonal data, calibrate models, and schedule sampling.

Why do we care about plumes?



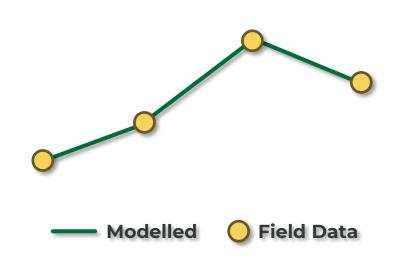
Cinderella Syndrome





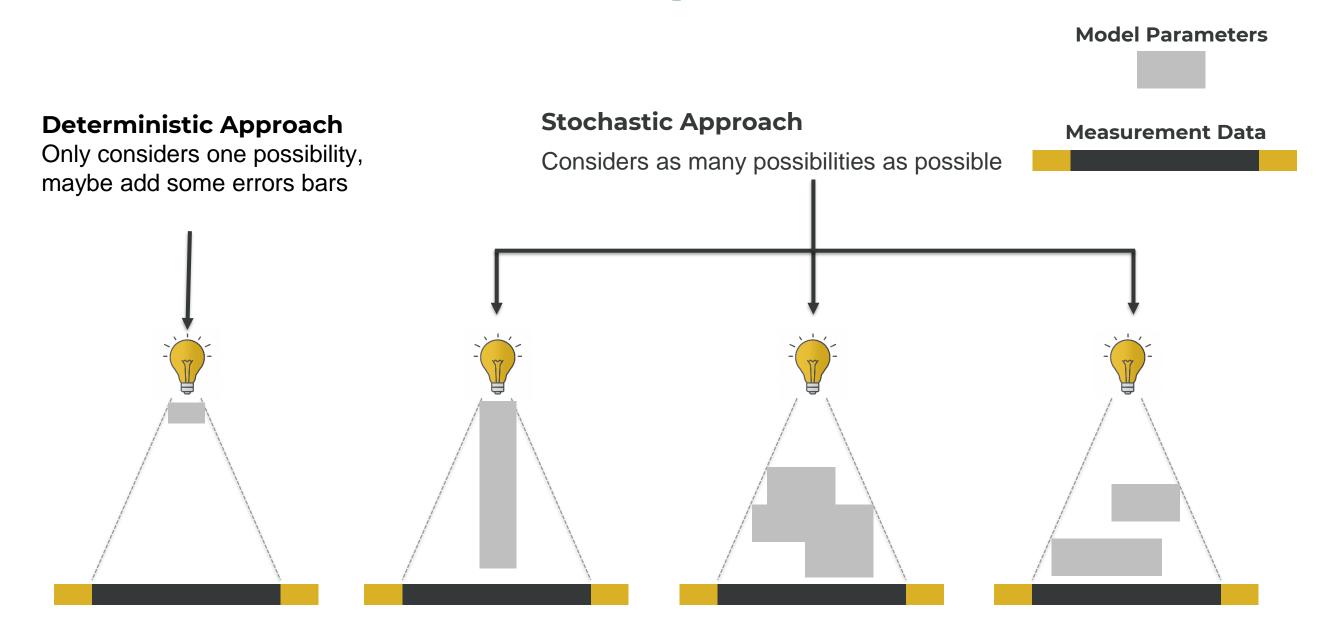








Non-Uniqueness

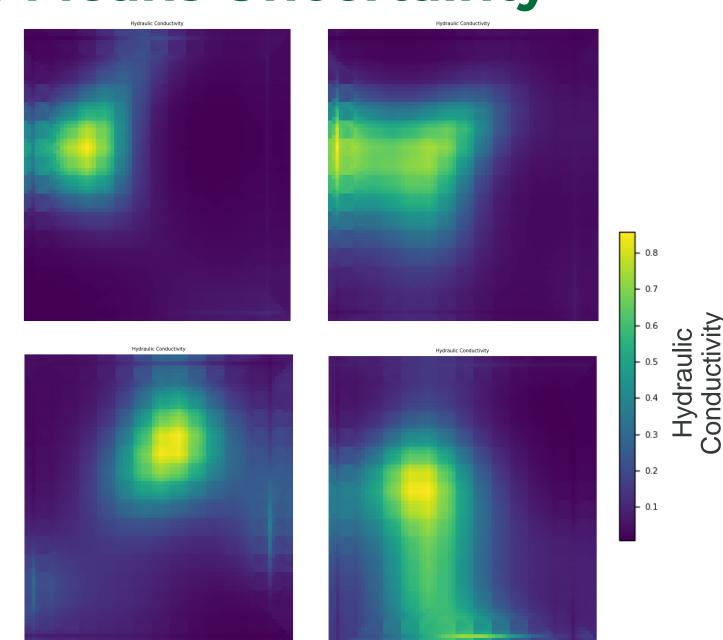


Non-Uniqueness Means Uncertainty

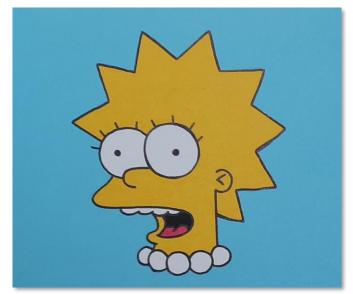
X Coordinates

Same site. Same data.

Many Models



Uncertainty Means Poor Predictive Power









Age 11....

Age 14....

Age 17....

And finally, age 18



LiORA Trends

- Adaptive groundwater models that evolve with new data.
 - MODFLOW, HYDRUS-1D, & PEST++

Inputs Engine Assimilation Outputs

The property of the proper

Data: DEMs
Hydrostratigraphy
HYDRUS-1D
Monitoring data
LiORA sensor streams

MODFLOW: GW/solute flow
PEST++: automated
calibration and uncertainty
reduction

Iterative Ensemble
Smoother: updates models
as sensor data arrive

Dynamic Plume extent/Persistence:

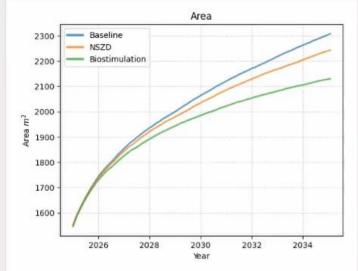
Risk forecasts

So what? Deliver decision-critical, uncertainty-quantified predictions of plume behavior and remediation performance.

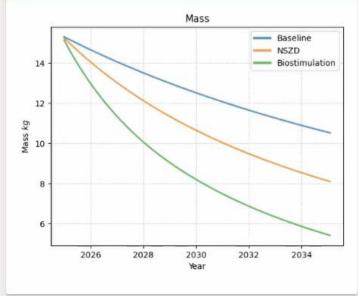
Plume by Method of Remediation



Plume Area



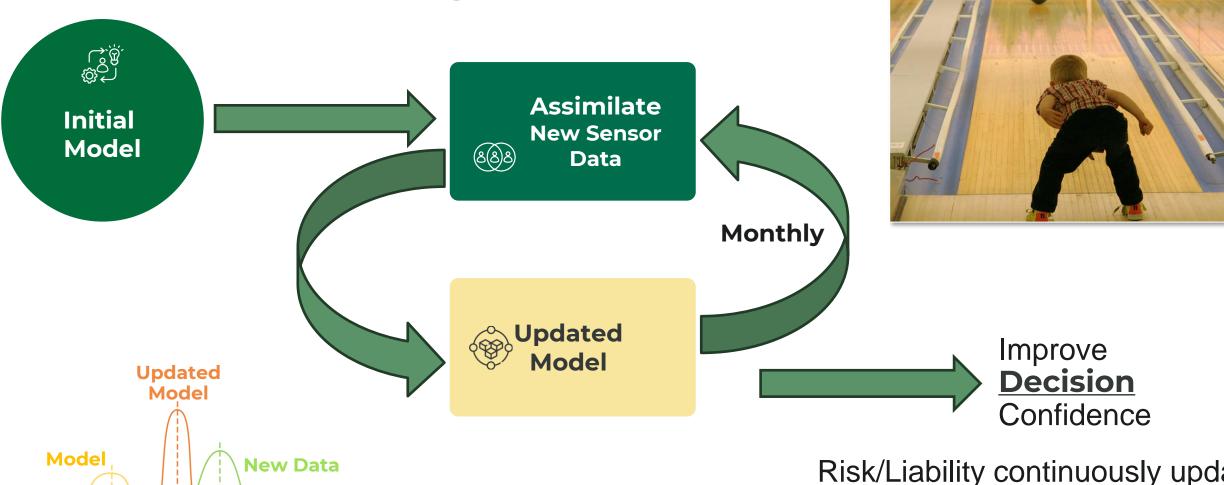
Plume Mass



Plume Risk Assessment



We can now build Machine **Learning Bumpers**



PEST++ and LiORA Data Assimilation into PlumeFutures Risk/Liability continuously updated:

- √ Vapour Risk
- ✓ Water Risk
- ✓ Pollutant Persistence

LiORA Trends and Insights







Current Site

Overview

Plume Stability

(§) Well Influence

00 Data Hub

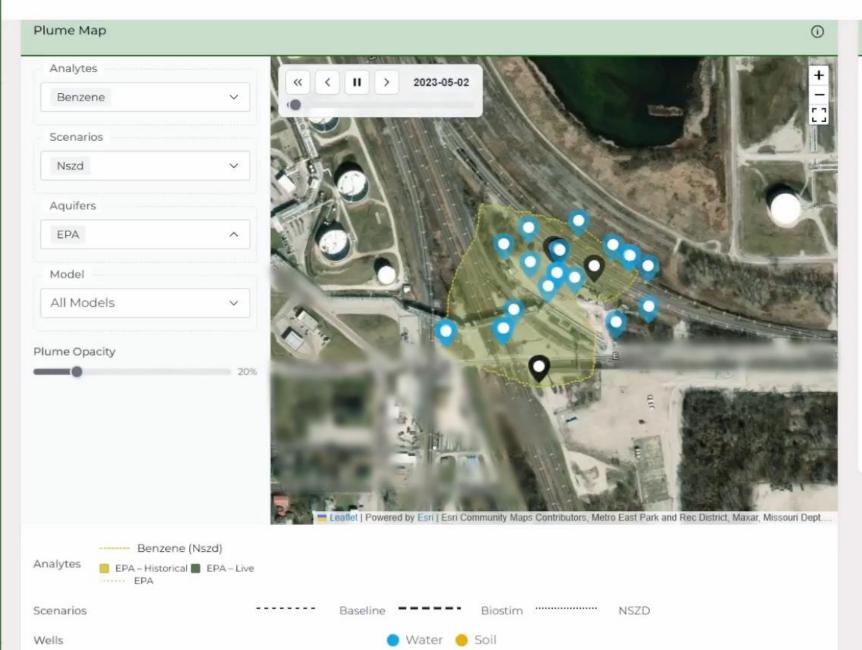
Portfolio

Companies

Background Jobs

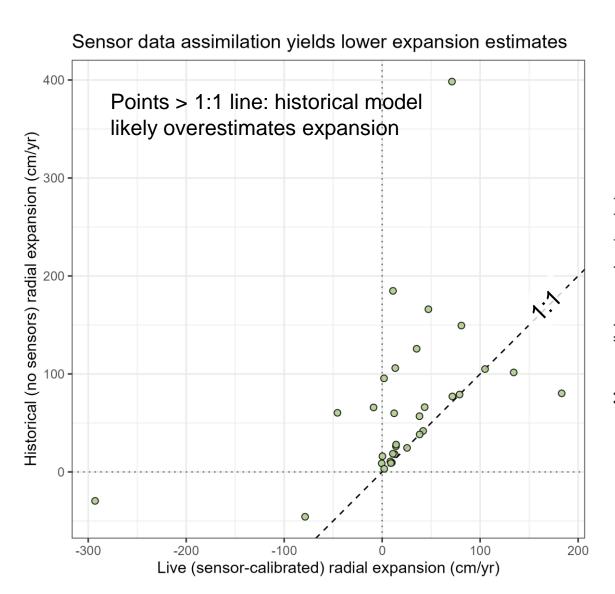
⇔ Oban

Dashboard

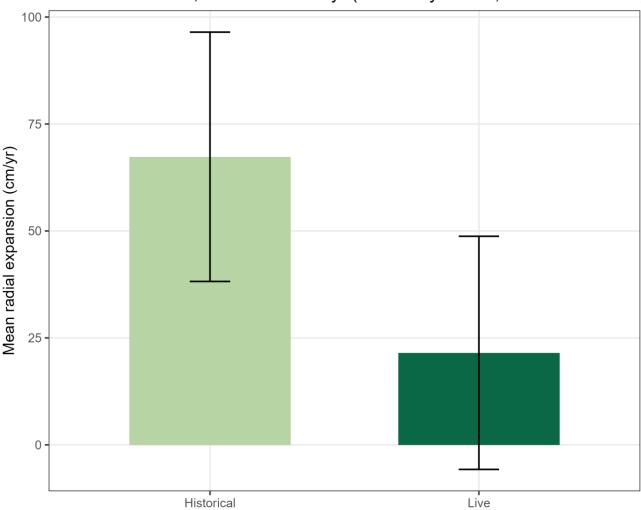


Stability Analysis Select analyte(s) and scenario(s) to view directional 6 analytes available expansion and migration. A Benzene Nszd East (-0.7°) Dominant expansion Migration Eastward, Southward East E-W expansion rate N-S expansion rate 0.01 m/period 0.03 m/period

Historical data over-estimated plume expansion in 22 of 32 sites



Mean radial expansion: Historical = 67.3, Live = 21.5 cm/yr (45.8 cm/yr lower; 68.1% reduction)



Key Takeaways

Traditional groundwater sampling misses key dynamics

- >40% of repeated samples from the same well, same season varied by >2X.
- Seasonal and thermal cycles can mimic or mask real concentration changes.
- Temperature and recharge drive predictable lags and cycles.

Adaptive modeling (LiORA) turns data into foresight

- Continuous data assimilation improves confidence and reduces overestimation.
- Historical models overpredicted plume growth in 22 of 32 sites.

Decision value

- Know what's changing, when, and why.
- Reduce/strategize expenditures.
- Prioritize interventions that matter.

So what? Continuous sensing and data assimilation turn uncertainty into foresight.



Want to take advantage of what you learned today?

But with your historical data?

Talk to me or Carolyn about LiORA Trends?

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Steve D Mamet Steve.Mamet@joinliora.com

Carolyn Inglis CarolynInglis@joinliora.com

Come chat with me at booth 6!

Technologies

Symposium 2025

Site insights for a better world.

Benzene lags temperature

Q1: Why this pattern?

- Driven by groundwater seasonality:
 - Spring → recharge and dilution
 - Fall → concentration and reduced mixing

Q2: Is it temperature, and how?

- Mostly yes. The lag reflects delayed mass transfer:
 - GW → membrane → vapour phase
 - Thermal and diffusive inertia cause benzene to respond weeks after temperature rises

Q3: Would we see this same pattern in intensive GW sampling?

- Yes, but with a shorter since there's no membrane resistance:
 - GW→ mass transfer

So what? Recognizing this lag helps separate true concentration changes from temperature-driven effects, improving how we interpret seasonal data, calibrate models, and schedule sampling.