

Define End-State and Optimize Monitoring Program Using High-Performance Computing

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DOE-EM Sites: Progress



Approximately \$6B /year
 107 major sites (1995) → 16 sites (2016)

Challenges

Remaining sites....

- **Complex contamination**
 - Multiple radionuclides, heavy metals (Hg)
 - VOC and other organic compounds
- **Hard/expensive to access**
 - Deep vadose zones
 - Increased drilling cost
- **Large volume with low contamination**
 - Not practical to remove soil (too much \$\$/waste)
 - Treatment/removal technologies are not effective

Environmental Monitoring

- Ensure public safety
- Prepare for liability issues

Beneficial for both residents and site operators

THE DENVER POST

OPINION

Activists ignore the science that says Rocky Flats National Wildlife Refuge is safe

By VINCENT CARROLL | The Denver Post
PUBLISHED: June 16, 2017 at 12:00 pm | UPDATED: June 16, 2017 at 2:36 pm




Andy Cross, Denver Post file

A herd of elk walks through a valley at the Rocky Flats National Wildlife Refuge on Sept. 25, 2015.

Good example: Monitoring data proves that the site is safe to dismiss false claims

thejapanimes



Reporters take photos in the basement of the planned relocation site for the Tsukiji fish market in Tokyo's Toyosu area in September. | KYODO

NATIONAL

Chemicals exceeding standards again are detected at new site of fish market in Toyosu

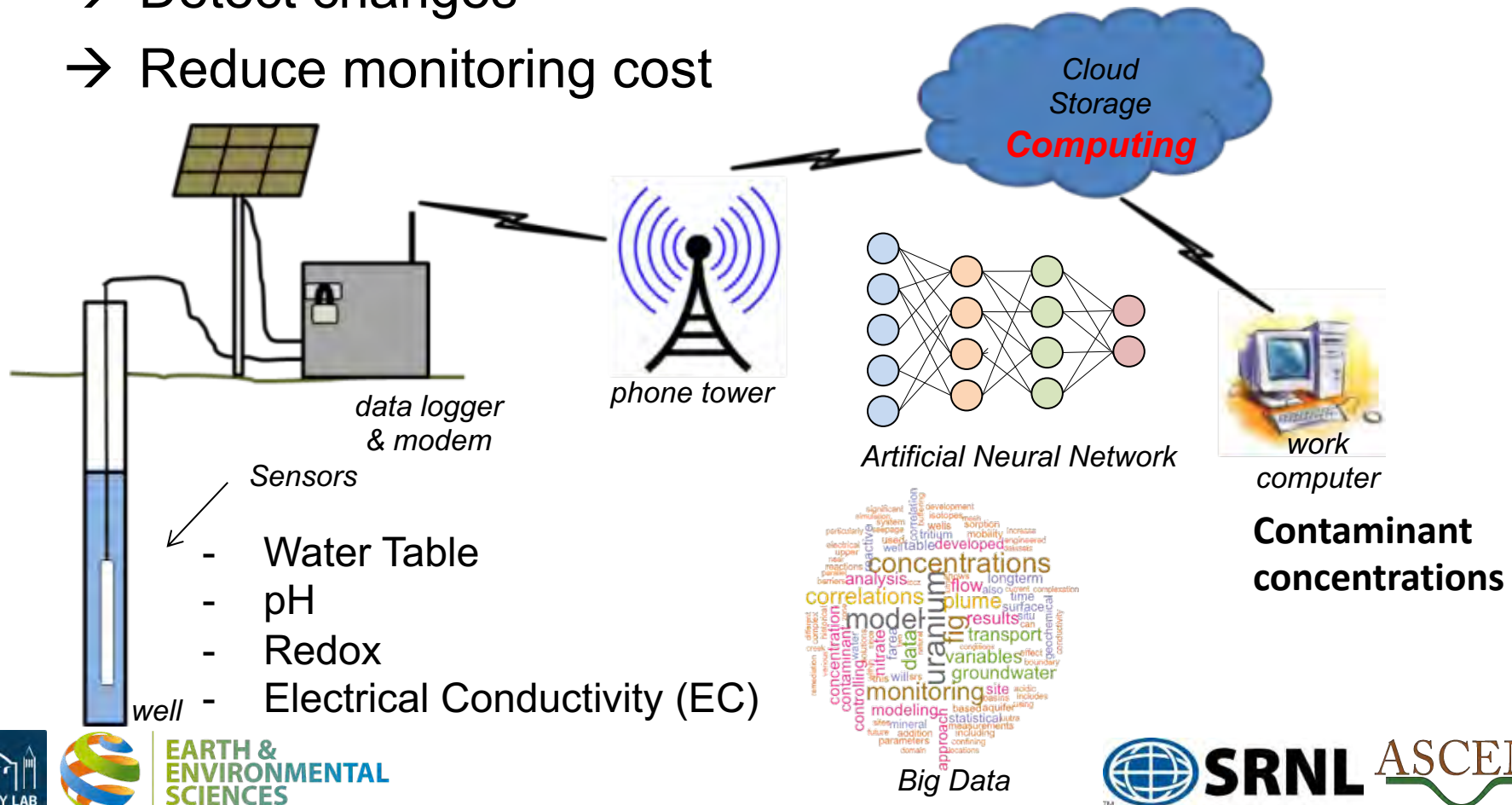
Bad example: Data anomaly cannot be explained → extra >\$100M

Research Goals

- Transition from **active** to **passive** remediation and **monitored natural attenuation**
 - SRS F-Area (2004) \$12M/yr → \$1M/yr
- Improve **long-term monitoring**
 - Great portion of life cycle cost (>\$10M/yr)
 - Detect new leaks/migration
- Ensure **long-term stability** of plumes
 - Climate change?

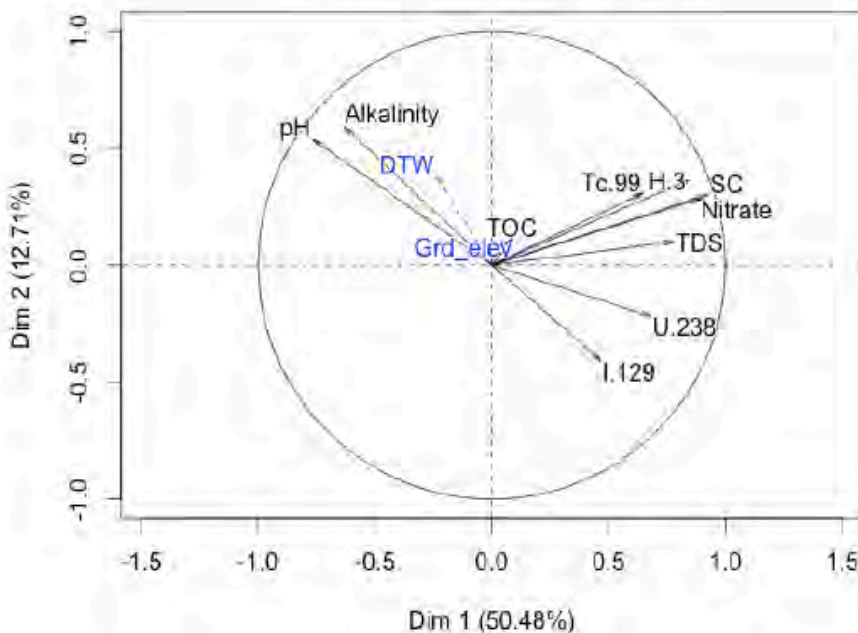
New Paradigm of Long-Term Monitoring

- In situ sensors, wireless network, cloud computing
 - Autonomous continuous monitoring
 - Detect changes
 - Reduce monitoring cost

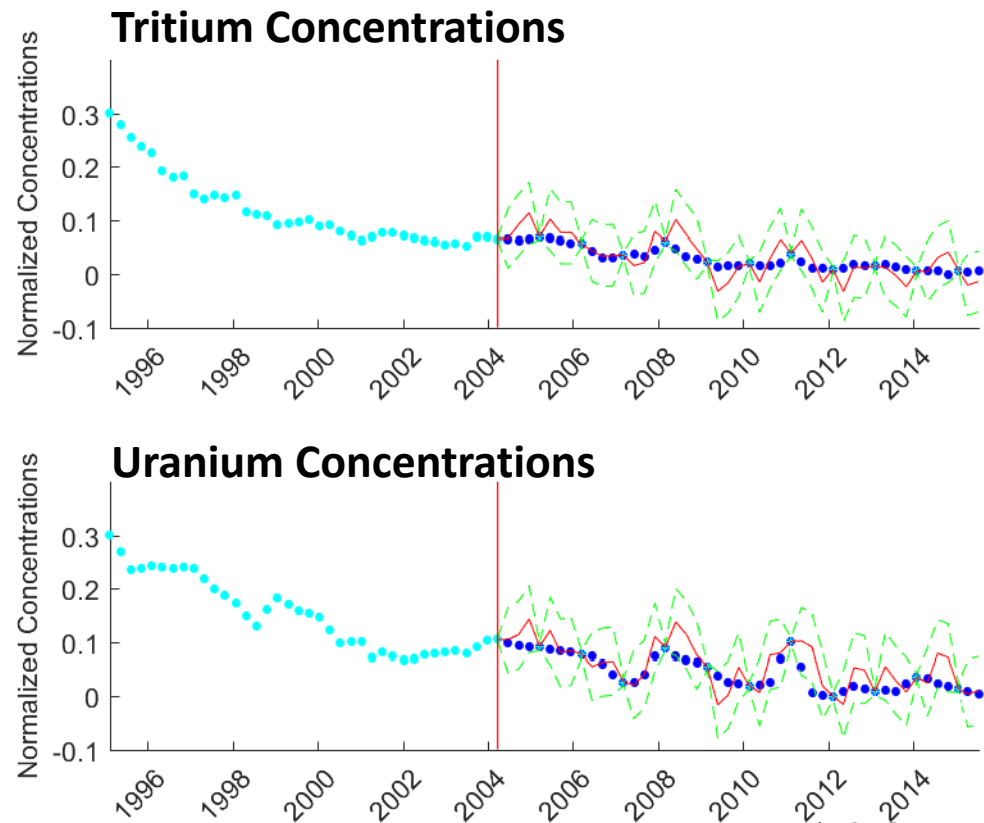


Data Analytics for Monitoring

- Big Data analytics
 - e.g., Principle component analysis (PCA)
 - System understanding
 - Master variables vs contaminant conc.



- Kalman filtering
 - In situ real-time estimation of contaminant concentration



Big Interest in Environmental Monitoring



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Article

In Situ Monitoring of Groundwater Contamination Using the Kalman Filter

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HOME NEWS MULTIMEDIA MEETINGS PORTALS ABOUT

PUBLIC RELEASE: 13-AUG-2018

Algorithm provides early warning system for tracking groundwater contamination

Berkeley Lab researchers devise system to monitor contaminant plumes
DOE/LAWRENCE BERKELEY NATIONAL LABORATORY



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New Algorithm Provides Real-Time Monitoring Of Groundwater Pollutants

Sam Benzera 8 Months Ago



NEWS EVENTS VIDEOS TV & PODCASTS

Efficiency & Environment

Scientists develop new to track groundwater pollutants in real-time

It is expected to reduce the frequency of manual groundwater sampling and lab analysis and therefore cut the monitoring cost



The Technology that Drives Government IT

AI & Automation Cybersecurity Cloud & Infrastructure Data & Analytics Smart Cities & IoT Emerg

Machine learning improves contamination monitoring

BY MATT LEONARD | AUG 14, 2018

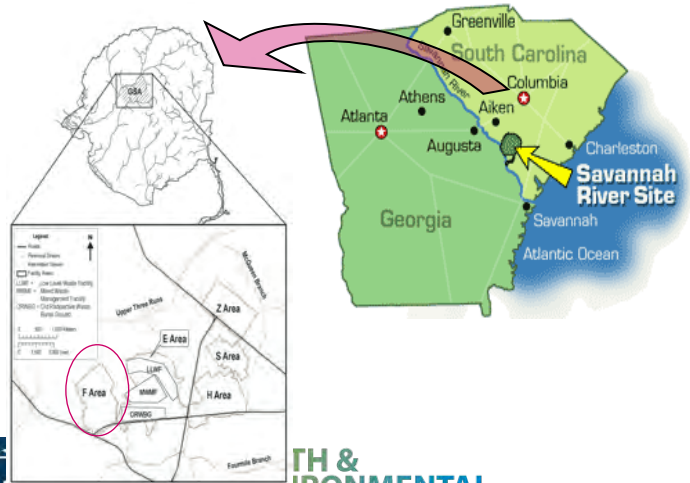
Because groundwater is susceptible to pollution from automotive fuel, fertilizer or naturally occurring substances like iron, the Environmental Protection Agency and its state-level counterparts conduct annual or quarterly sampling and analysis.

Modeling for Supporting Monitoring

- Confirm the correlations: Master variables vs contaminant concentrations
- Climate resiliency: how to place monitoring wells or what to expect in the response to climate changes
- (In development) Monitoring well placement based on simulated plume evolutions

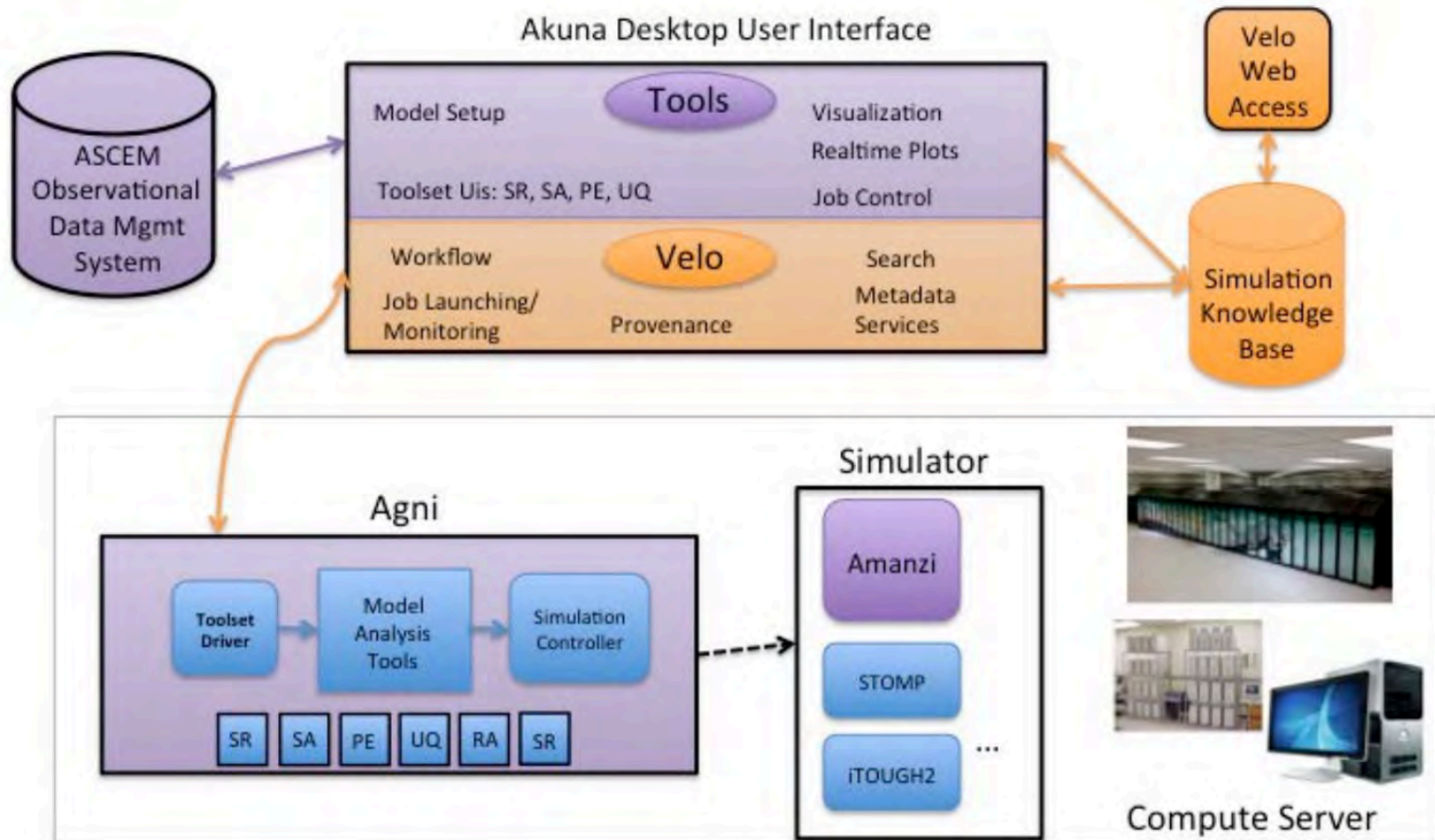
Demonstration: SRS F-Area

- **Disposal activities:**
 - Disposal of low-level radioactive, acid waste solutions (1955–1989)
 - Acidic plume with radionuclides (pH 3–3.5, U, ^{90}Sr , ^{129}I , ^{99}Tc , ^3H)
- **Remediation approaches**
 - Pump & treat (\$12M/yr) → Passive remediation (funnel-gate system for pH neutralization; \$1M/yr)
 - Natural attenuation: long-term remediation alternative



Virtual Test Bed: ASCEM Overview

Advanced Simulation Capability for Environmental Management



Geochemistry Development

- Complex geochemistry
 - pH Dependent
 - Aqueous complexation
 - Surface complexation
 - Mineral dissolution/precipitation
 - Cation exchange
 - Decay

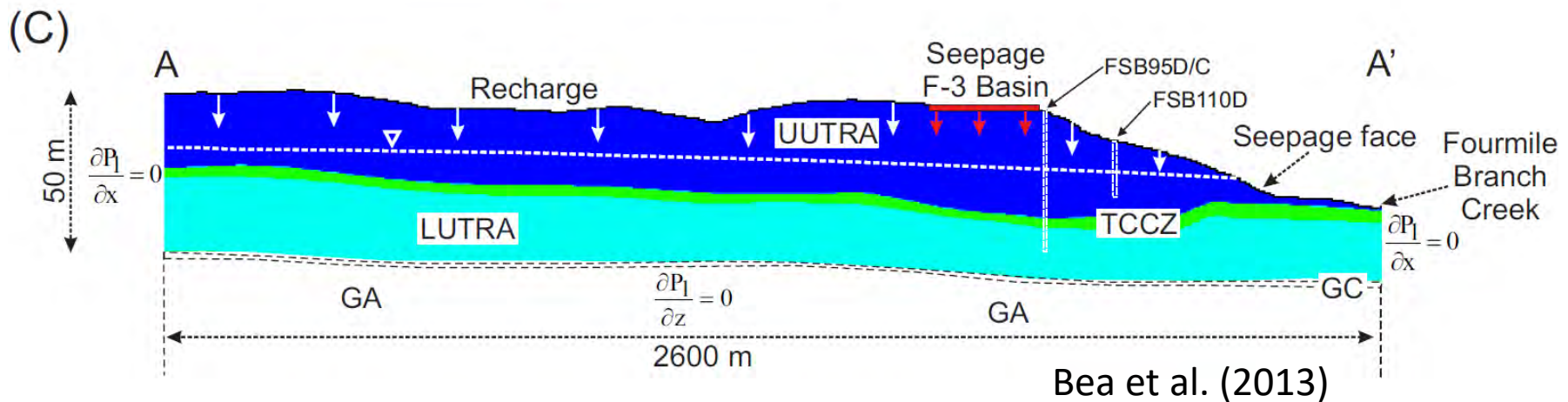
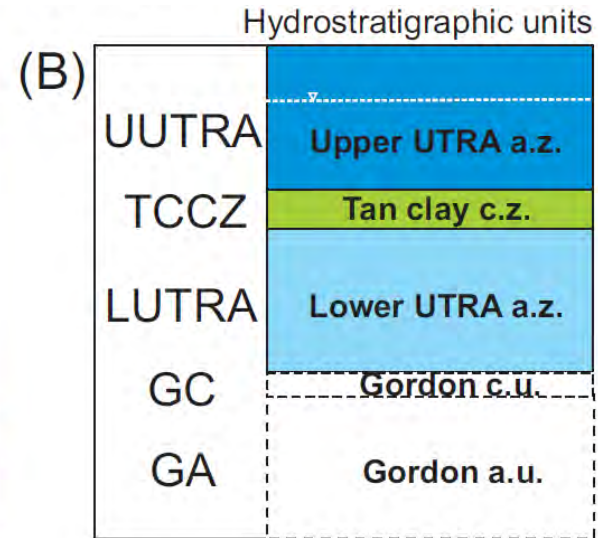
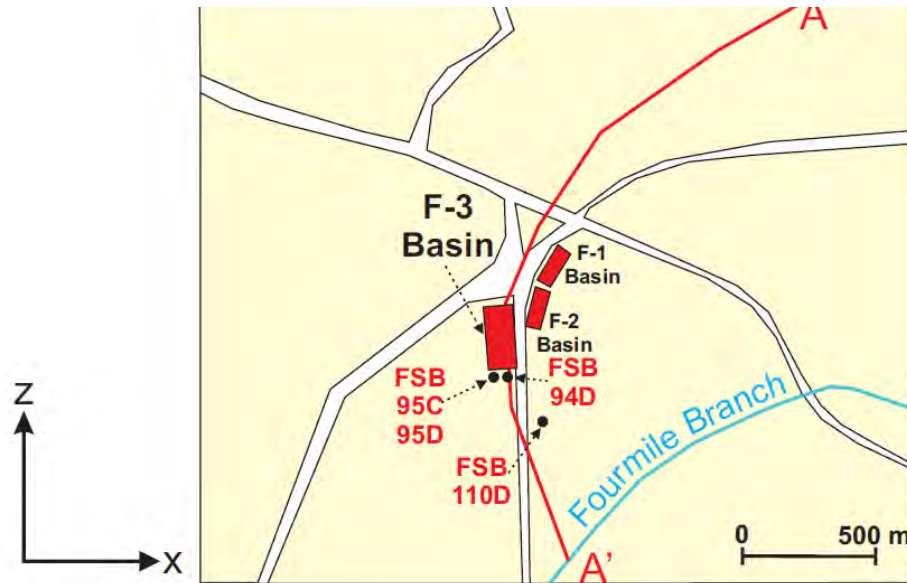
	$\log_{10} K$ (25° C)
⁽¹⁾ Equilibrium Surface Complexation	
$(>SO)UO_2^+ \leftrightarrow >SOH - H^+ + UO_2^{2+}$	-0.44
⁽²⁾ Cation Exchange	
$NaX \leftrightarrow Na^+ + X^-$	1.0
$CaX_2 \leftrightarrow Ca^{2+} + 2 X^-$	0.316
$AlX_3 \leftrightarrow Al^{3+} + 3 X^-$	1.71
$HX \leftrightarrow H^+ + X^-$	0.025

	$\log_{10} K$ (25° C)	Ref.
Quartz \leftrightarrow SiO ₂ (aq)	-3.7501	(1)
Kaolinite \leftrightarrow 2Al ⁺³ + 2SiO ₂ (aq) + 5H ₂ O - 6H ⁺	7.57	(2)
Goethite \leftrightarrow Fe ⁺³ + 2H ₂ O - 3H ⁺	0.1758	
Schoepite \leftrightarrow UO ₂ ⁺² + 3H ₂ O - 2H ⁺	4.8443	(1)
Gibbsite \leftrightarrow Al ⁺³ + 3H ₂ O - 3H ⁺	7.738	(3)
Jurbanite \leftrightarrow Al ⁺³ + SO ₄ ⁻² + 6H ₂ O - H ⁺	-3.8	(4)
Basaluminite \leftrightarrow 4Al ⁺³ + SO ₄ ⁻² + 15H ₂ O - 10H ⁺	22.251	(4)
Opal \leftrightarrow SiO ₂ (aq)	-3.005	(5)

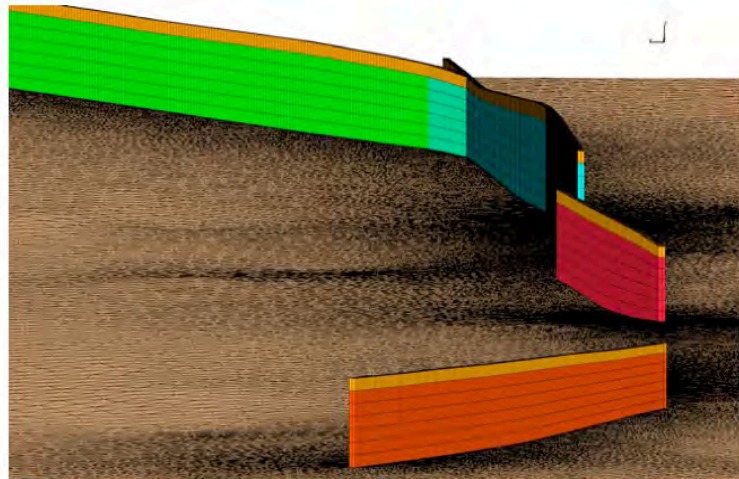
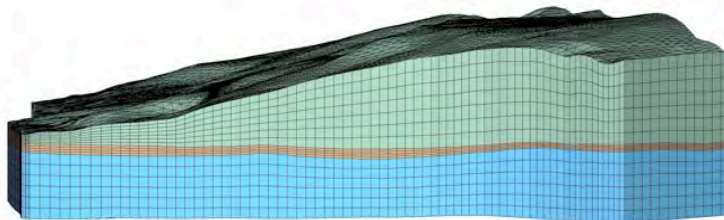
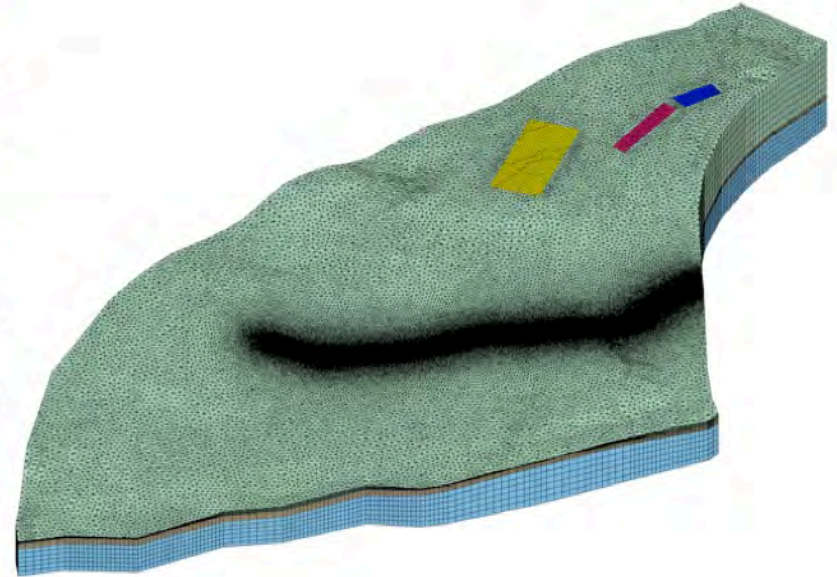
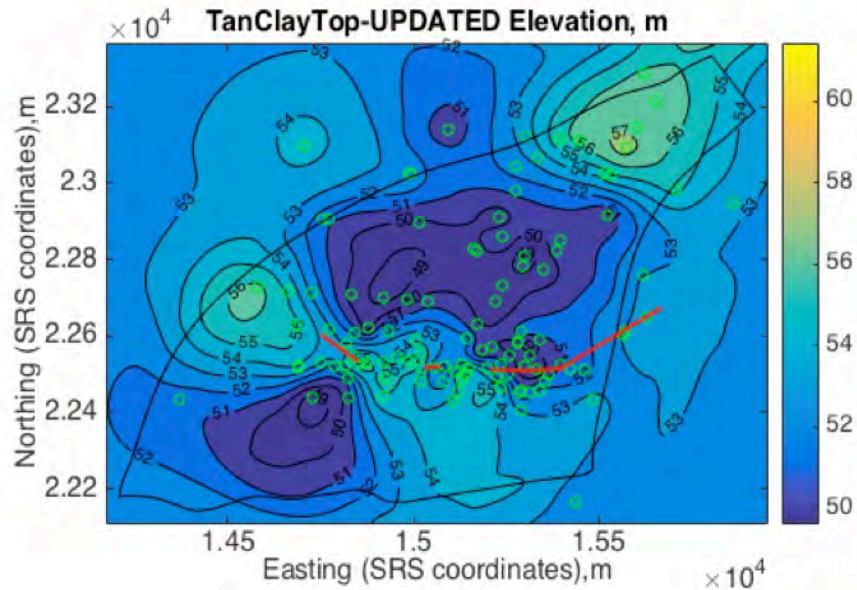
Aqueous complexation

Reaction	$\log_{10} K$ (25° C)
$OH^- \leftrightarrow H_2O - H^+$	13.99
$AlOH^{2+} \leftrightarrow Al^{3+} + H_2O - H^+$	4.96
$Al(OH)_2^+ \leftrightarrow Al^{3+} + 2H_2O - 2H^+$	10.59
$Al(OH)_3(aq) \leftrightarrow Al^{3+} + 3H_2O - 3H^+$	16.16
$Al(OH)_4^- \leftrightarrow Al^{3+} + 4H_2O - 4H^+$	22.88
(and more)	

Flow/Transport Model



3D Mesh Development

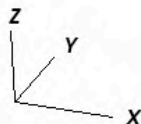
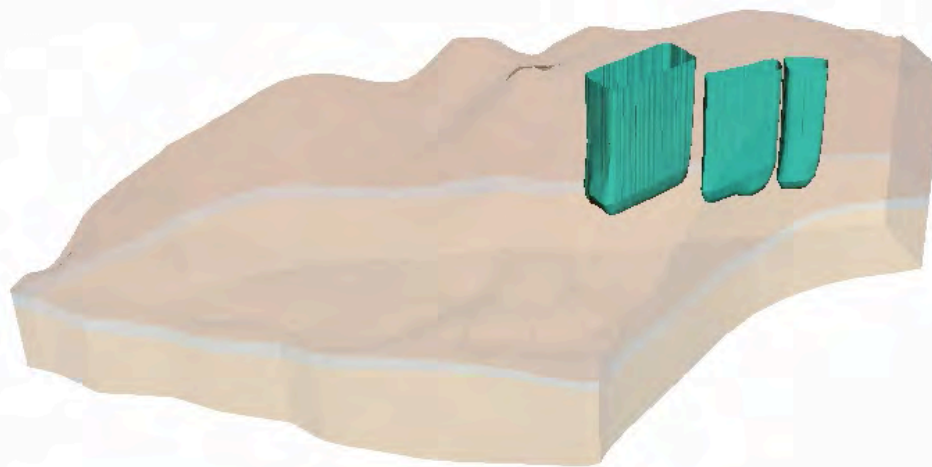


Uranium Plume Evolution

DB: plot_data.VisIt.xmf
Time: 1956

Uranium Plume: Residual contaminants

- Under the basins
- Within Tan Clay

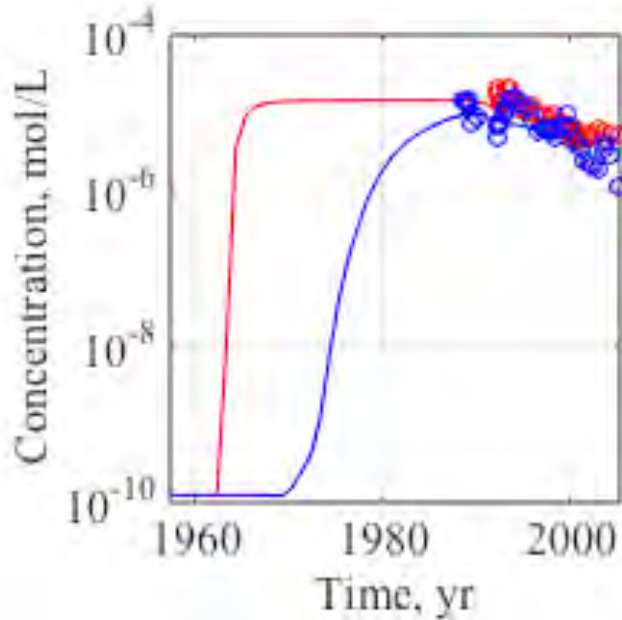


ASCEM Modeling Results

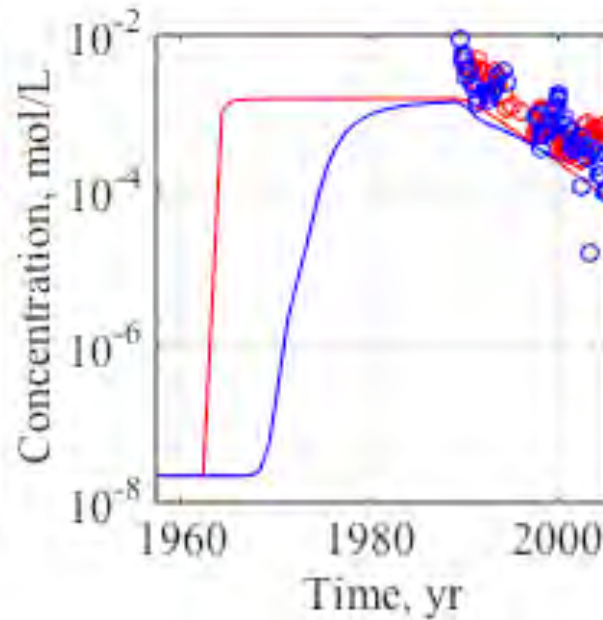
user: user
Sun Apr 14 10:34:15 2019

Validation with Observations

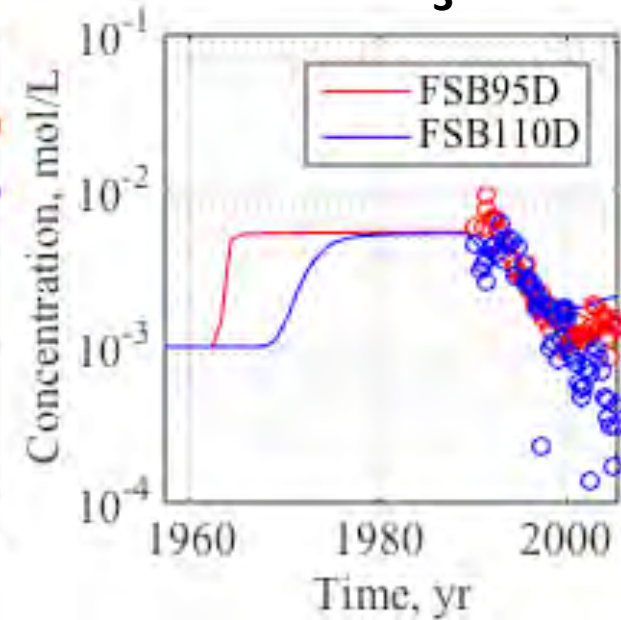
Uranium



Al³⁺



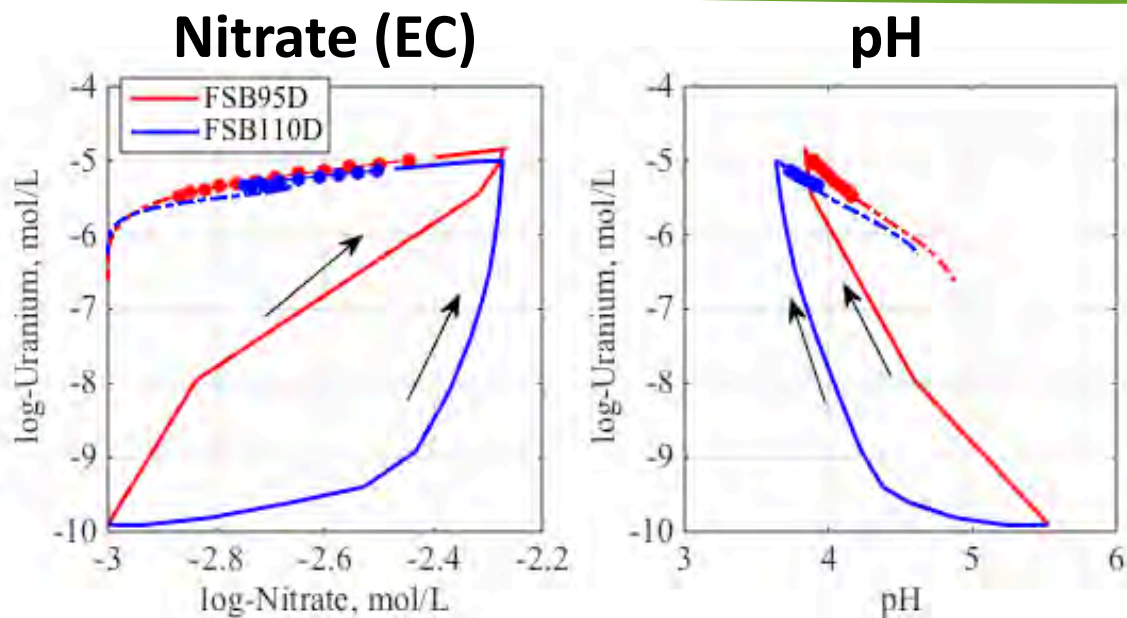
NO₃⁻



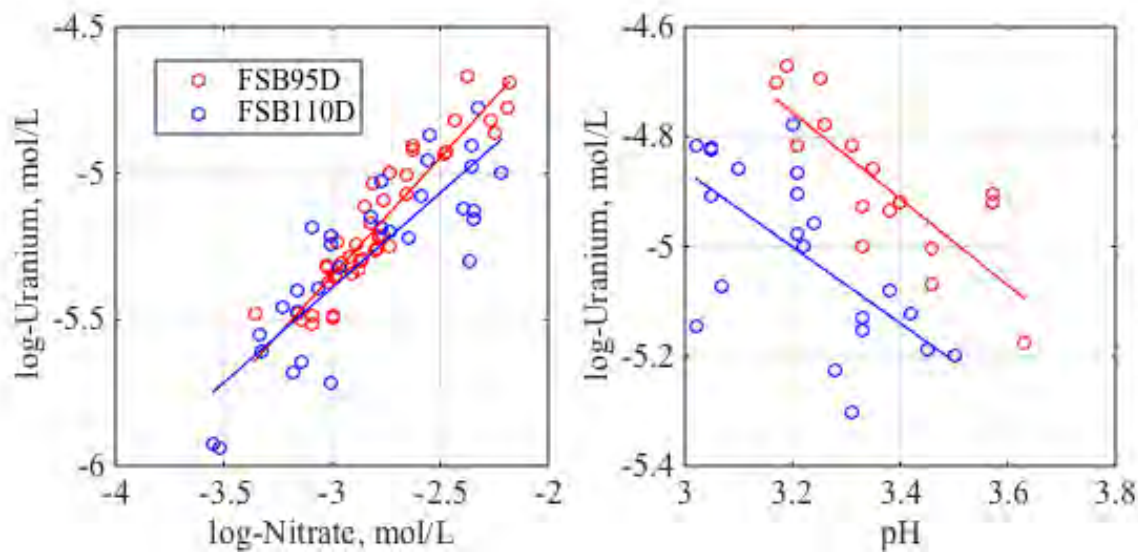
Good agreement with observations

In situ Monitoring: Master Variables vs U Conc.

Simulated



Measured



Resiliency to Climate Disturbances

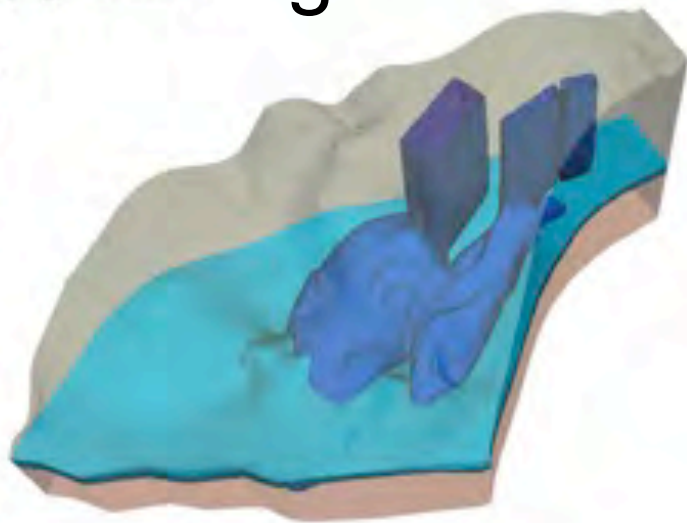
Extreme Events

- Flooding
- Drought



Savannah River Flooding, 2016

What will happen to residual contaminants?



(f) 2050

Technical Initiative in SURF and ITRC

- How to prepare for climate change in sustainable remediation

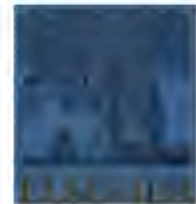
Resiliency to Climate Disturbances

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Climate change impact on residual contaminants under sustainable remediation



Arianna Libera^{a,*}, Felipe P.J. de Barros^a, Boris Faybishenko^b, Carol Eddy-Dilek^c, Miles Denham^d, Konstantin Lipnikov^e, David Moulton^e, Barbara Maco^f, Haruko Wainwright^b

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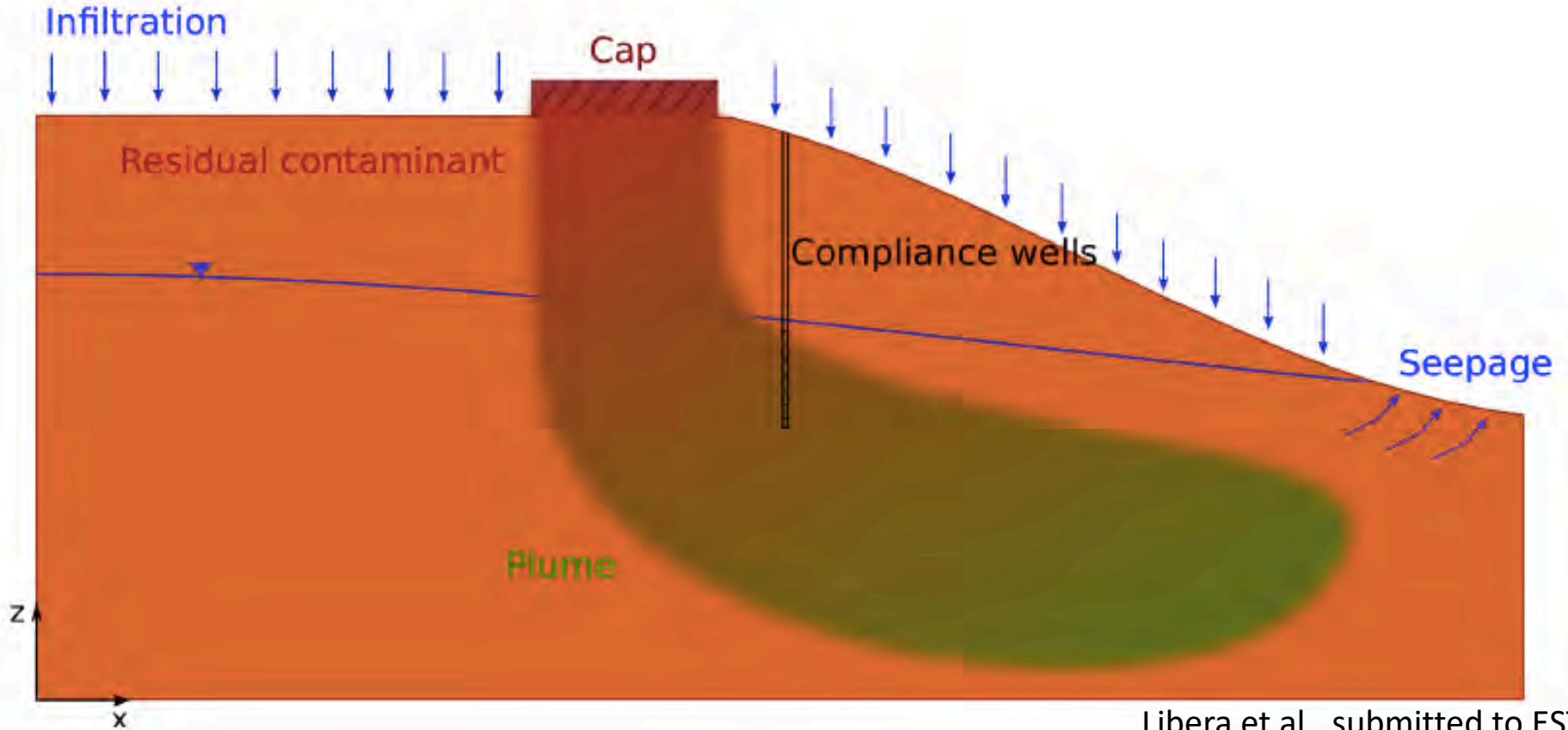
^c Savannah River National Laboratory, Aiken, SC, USA

^d Panoramic Environmental Consulting, LLC, Aiken, SC, USA

^e Los Alamos National Laboratory, Los Alamos, NM, USA

^f Wactor & Wick LLP Environmental Lawyers, Oakland, CA, USA

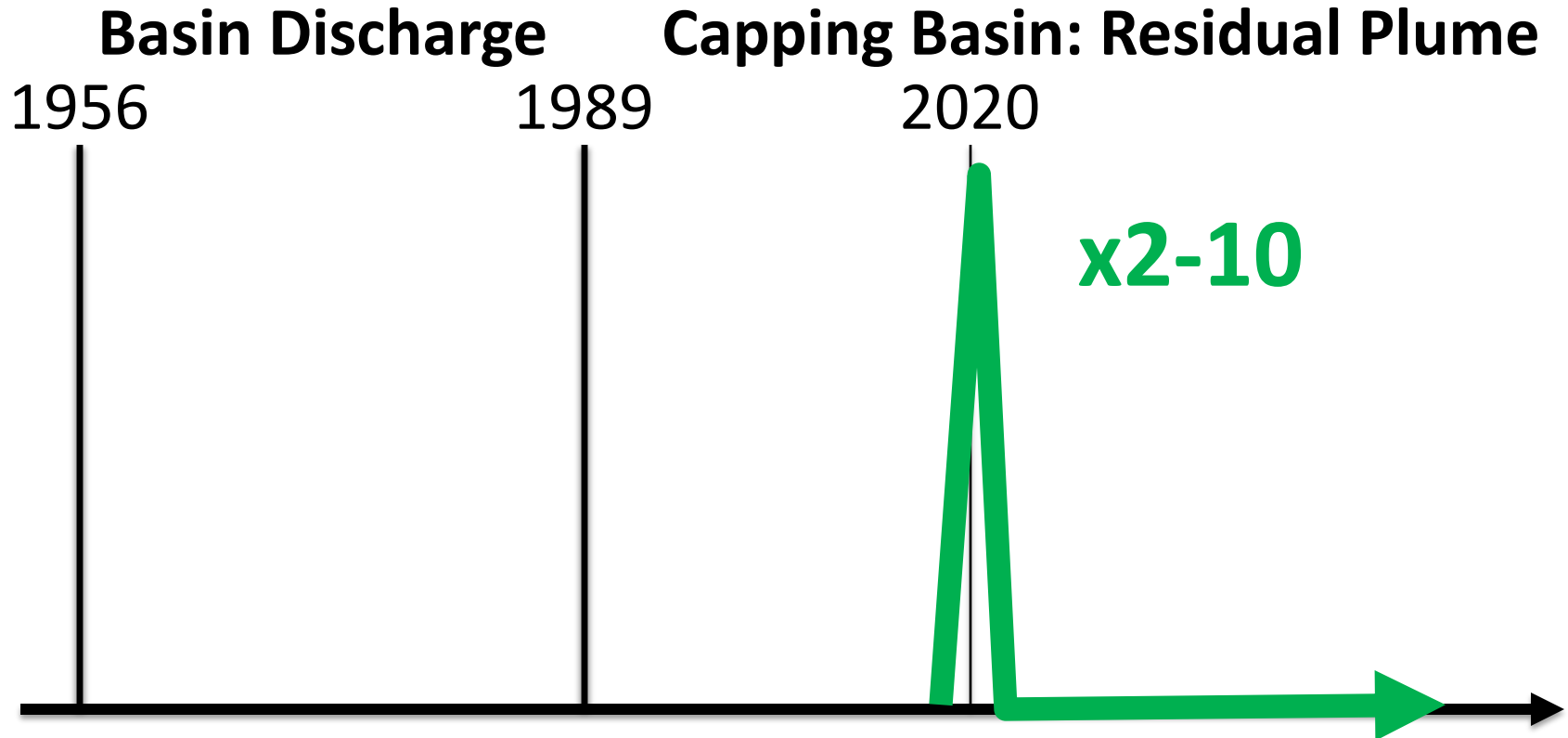
Flooding, Drought Impact



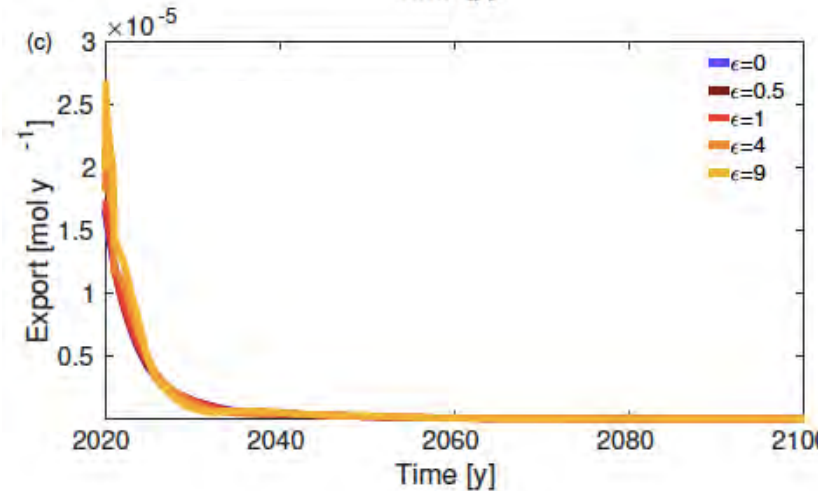
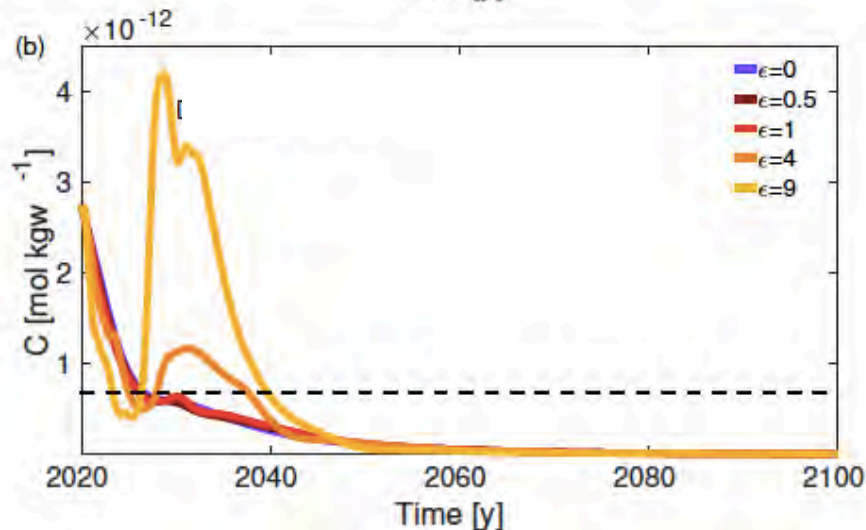
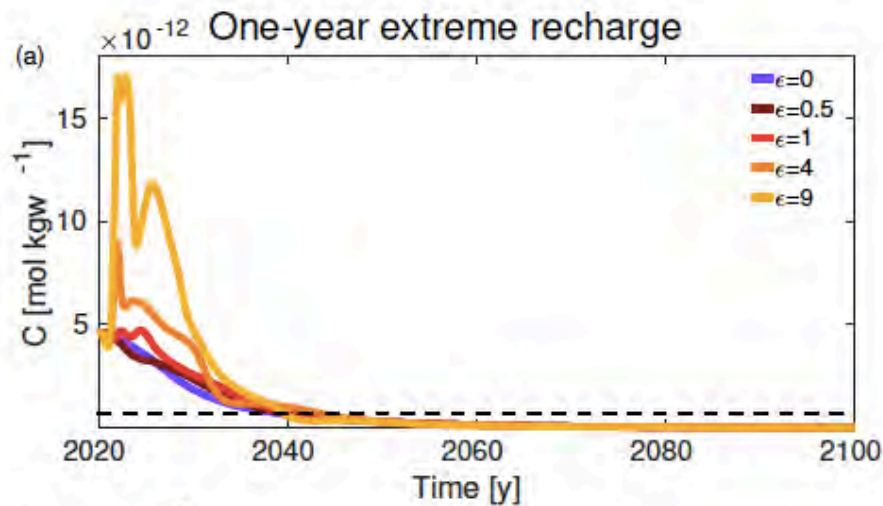
Libera et al., submitted to EST

+/- Precipitation/Temperature → Infiltration, ET
Trade off: Mobility vs Dilution

Climate Scenarios: Flooding



Flooding Event Effect



- Increase in precipitation of ONE year: x1.5 – x 10 in 2020
- Dilution then Increase
- Effect can linger for two decades
- Source zone wells important to detect remobilization
- Export to the river doesn't change significantly

Monitoring Optimization

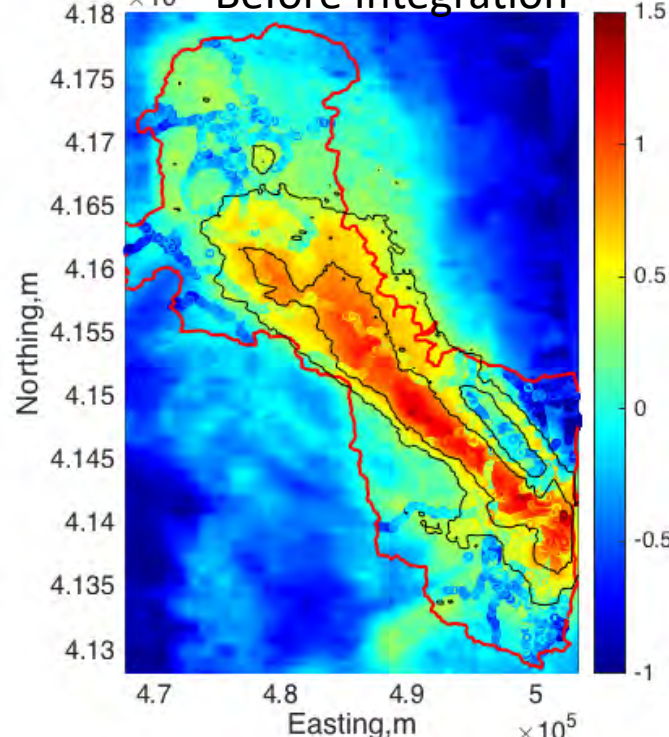
- How can we identify key monitoring locations, using increasingly available spatially extensive data?
 - Geophysical plume mapping
 - Simulated plume evolution
 - Airborne gamma mapping

Fukushima Radiation Mapping

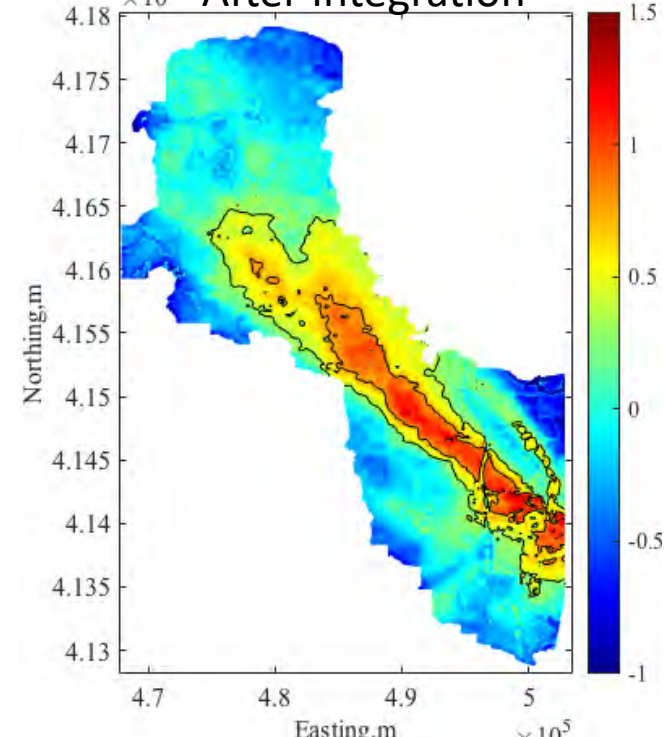
- Integrate various types/footprints of data
- Uncertainty quantification
- Adopted by Nuclear Regulatory Agency



Before Integration

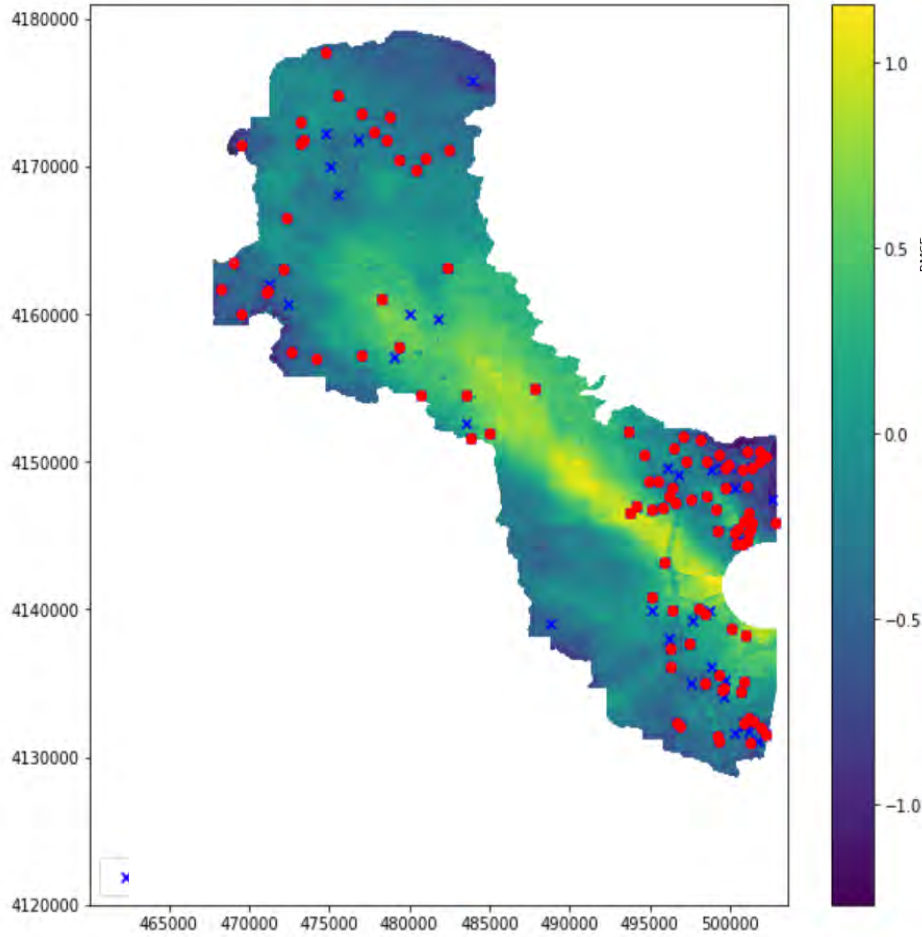


After Integration

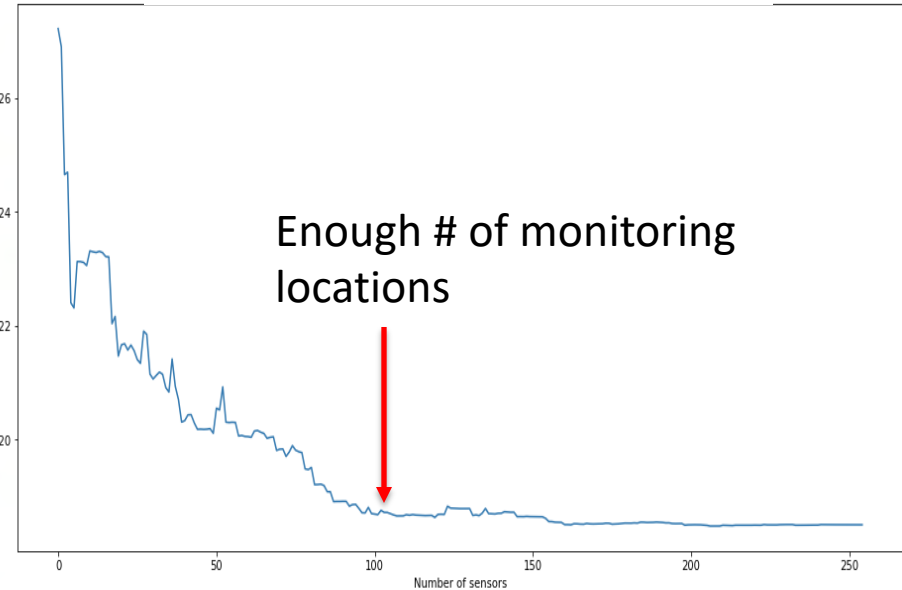


Monitoring Post Optimizations

log10(microSv/hr)



Interpolation Error Reduction



- Identified 100 locations that capture the variability of air dose rates
- Extending to simulated plume at the F-Area

Summary

- **Cost effective strategies for long-term monitoring**
 - In situ sensors for **continuous monitoring**
 - Reduce cost while enhancing the safety
 - **Data analytics**: Kalman filter etc
- **Modeling for supporting monitoring**
 - Confirming in situ monitoring strategies
 - Correlations between master variables and contaminant concentrations: Now and future
 - Climate change: what to expect, where to monitor?
 - Optimizing monitoring locations based on spatially extensive data (mapping data or simulated data)