



End-State Driven Performance-Based Management of Pump-and-Treat (P&T) Remedies

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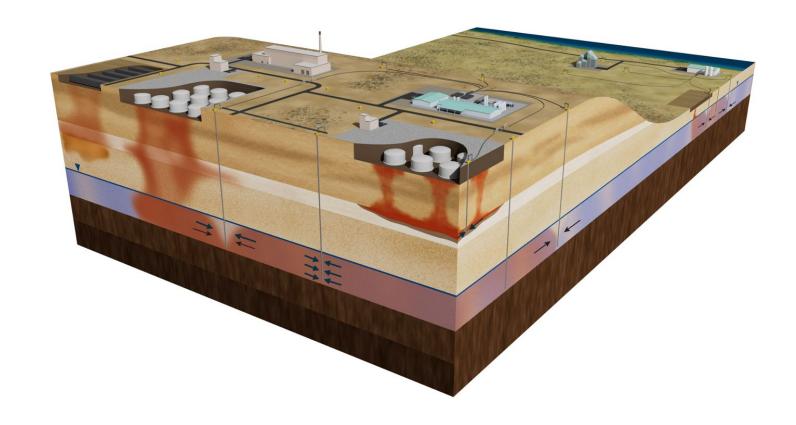
Outline



- Part 1: Performance-Based Optimization of P&T
 - Background
 - Performance-based optimization
 - Computational approaches
 - Pre-screening tool framework
- Part 2: Pre-Screening Tool Demonstration
 - Hanford 200 West P&T
 - Scenario evaluation
- Part 3: Use of Deep-Learning Approaches
 - Well performance predictions
 - Increasing model efficiency



Part 1– Performance-Based Optimization of P&T







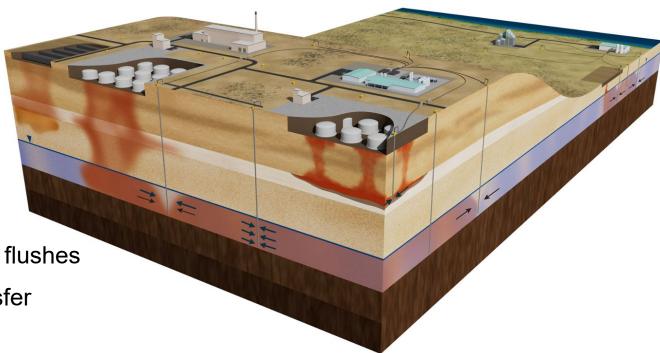
Pump-and-Treat (P&T) Systems



- Pump-and-treat (P&T) systems have been used for hydraulic containment and/or treatment of contaminated groundwater
 - A well network for groundwater extraction
 - Above-ground ex-situ treatment unit
 - Disposal system for the treated water
- Initial designs typically address large-scale containment and bulk treatment, and may not be an optimal design for mass removal and long-term effectiveness
 - Early goals focus on volumetric pumping
- Performance diminishes due to factors such as:
 - Heterogeneity
 - Large and dispersed plumes requiring multiple pore volume flushes
 - Presence of source zone and/or diffusion-limited mass transfer
 - Recalcitrant/competing contaminants



Pump-and-treat extraction well (adapted from PNNL-24741)

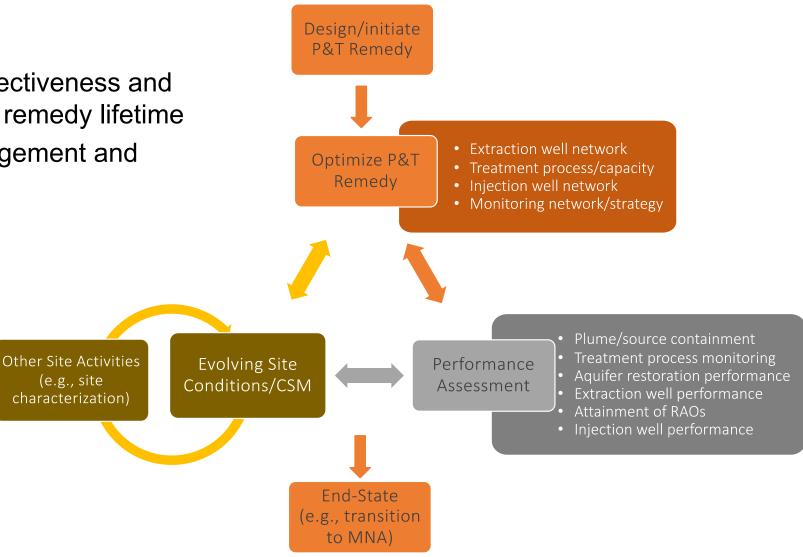








- Objective of periodic P&T optimization
 - Maintain/increase contaminant removal effectiveness and efficiency as much as possible throughout remedy lifetime
 - Well network and treatment capacity management and optimization
- Performance-based optimization approach relies on:
 - Continuous performance monitoring
 - Frequent updates to CSM based on the new data
 - Periodic evaluations of performance effectiveness and remedy lifetime
 - Computational optimization evaluations
 - Capacity & well network effectiveness



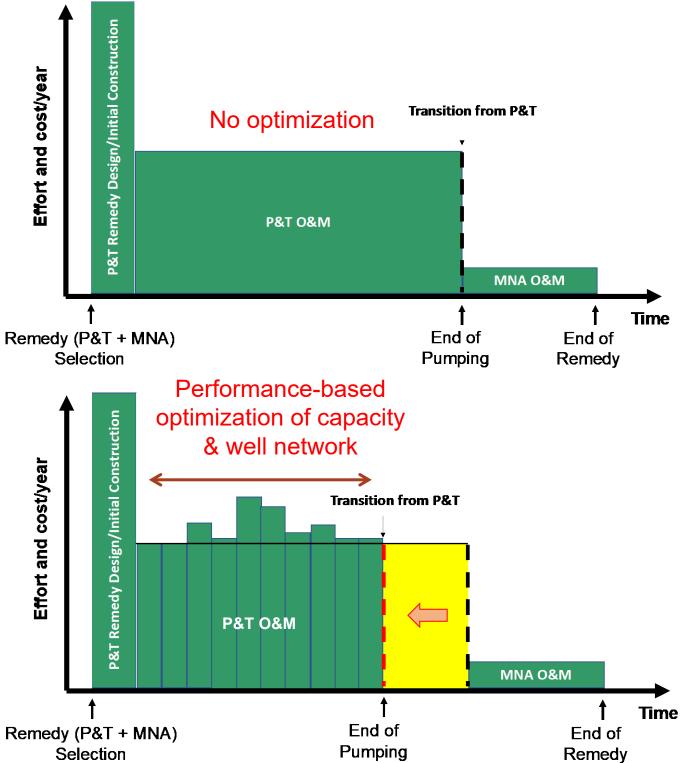


ITRC Guidance

- Interstate Technology Regulatory Council (ITRC)
- New performance-based P&T optimization guidance published in 2023











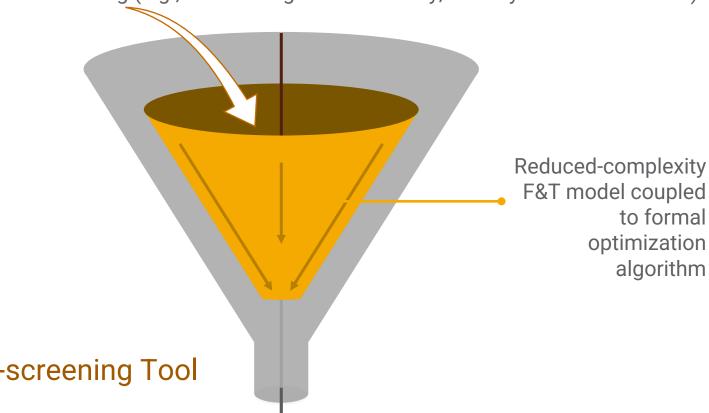


Computational Optimization and Pre-screening

- Formal optimization evaluations require:
 - Flow & transport (F&T) model of P&T system, coupled with optimization algorithms to run thousands of simulations
 - Resource intensive!
- Pre-screening (i.e., scoping) framework
 - Supports scenario evaluation to feed into decision tools
 - Reduced computational burden

Range of assumptions/scenarios

Goal setting (e.g., maximizing mass recovery, remedy lifetime reduction)







- Comparative assessment of scenarios
- Narrower set of potentially successful optimization approaches
- Uncertainty evaluation



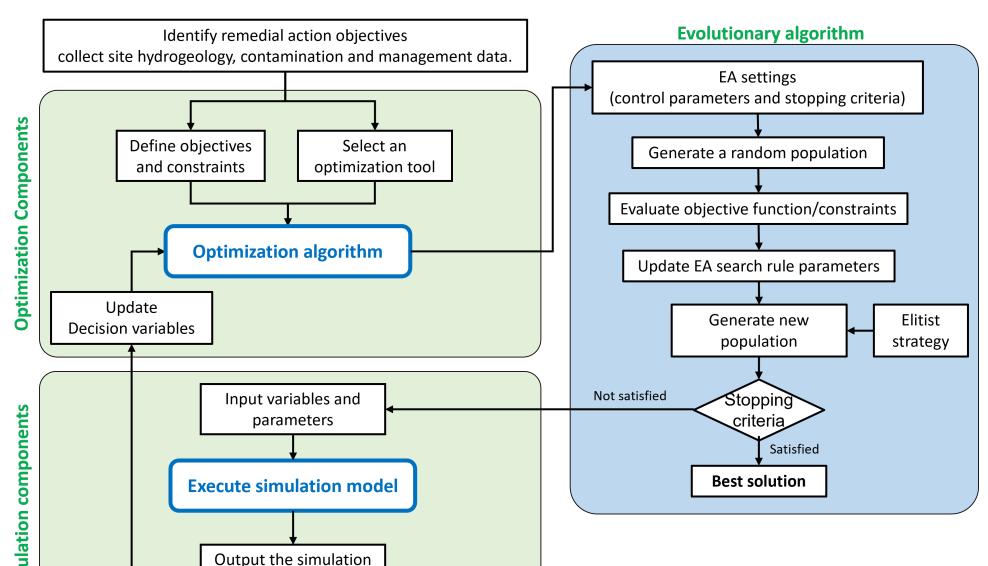


results



Formal optimization parameters

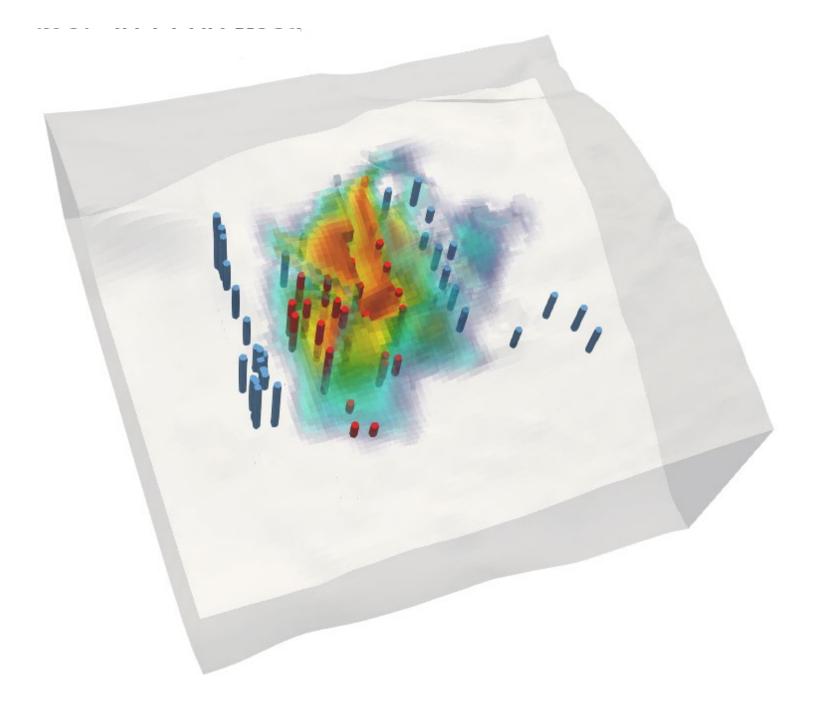
Reduced complexity fate and transport model



Stochastic
approach for
solving both
singleobjective and
multiobjective
problems



Part 2 – Pre-Screening Tool Demonstration



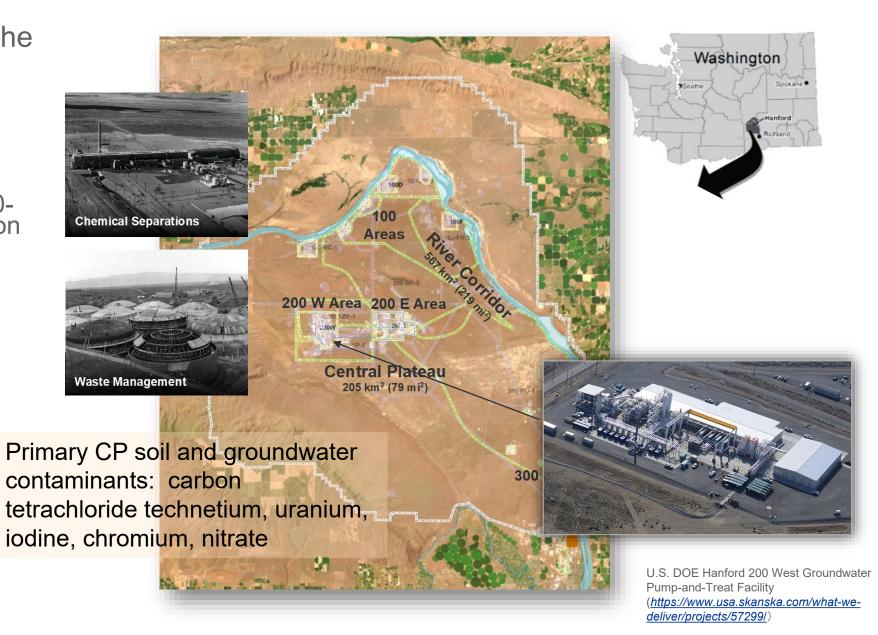




Hanford 200 West P&T



- Historical plutonium production for the Manhattan Project
- Operating since 2012 in the Central Plateau (CP) of the Hanford Site
 - Will be pumping for 25 years per 200-ZP-1 operable unit Record of Decision
- Addressing groundwater plumes:
 - Carbon tetrachloride (CTET)
 - Technetium
 - Uranium
 - Chromium
 - Nitrate*
- Current treatment capacity is 2500 gpm
 - 38 existing extraction wells
 - 30 existing injection wells

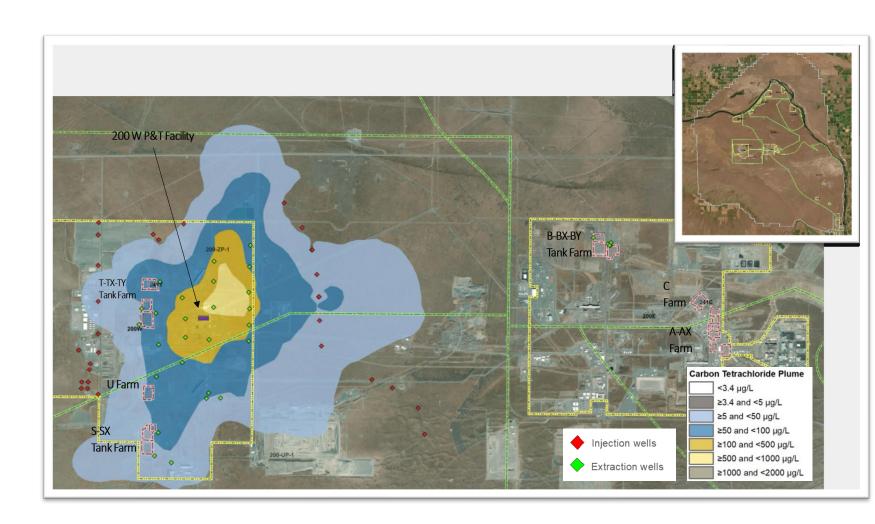








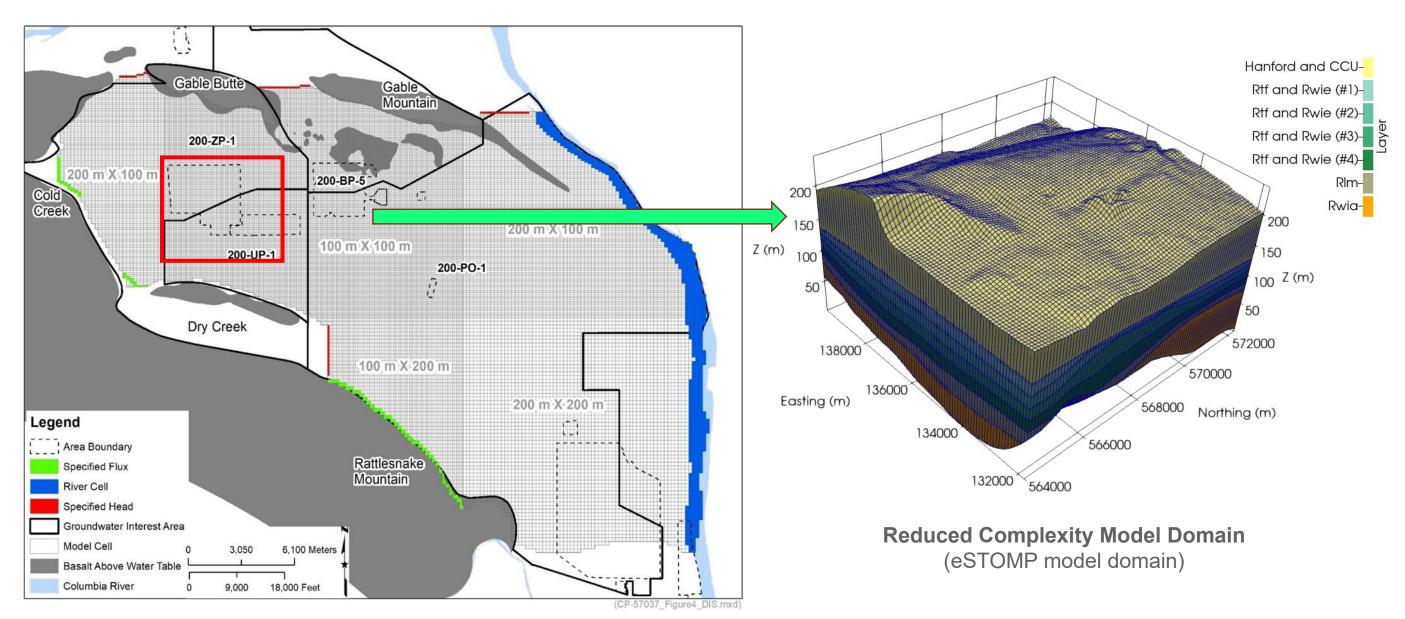
- ► Large carbon tetrachloride plume in the 200-ZP-1 Operable Unit (OU)
 - Slower CTET degradation rate
 - More contaminant mass in the aquifer
 - Diminishing performance at some wells
- ► 200-ZP-1 OU Optimization Study Plan to evaluate
 - Increasing carbon tetrachloride removal and treatment
 - Evaluating the transition to monitored natural attenuation (MNA) for nitrate, consistent with RAOs











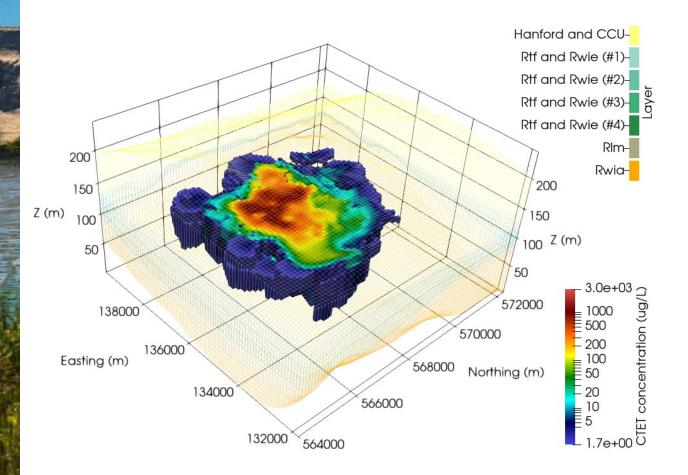
Plateau to River (P2R) Version 8.3 Model Extent and Groundwater Flow Boundary Conditions (source: ECF-HANFORD-22-0114-REV.0)



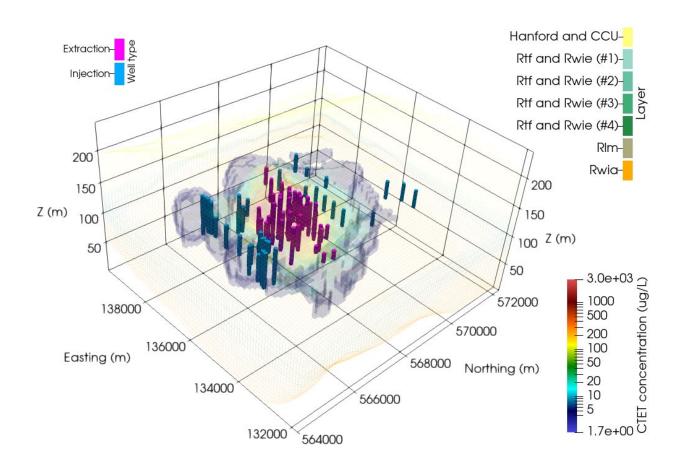




► Initial CTET plume (2015)



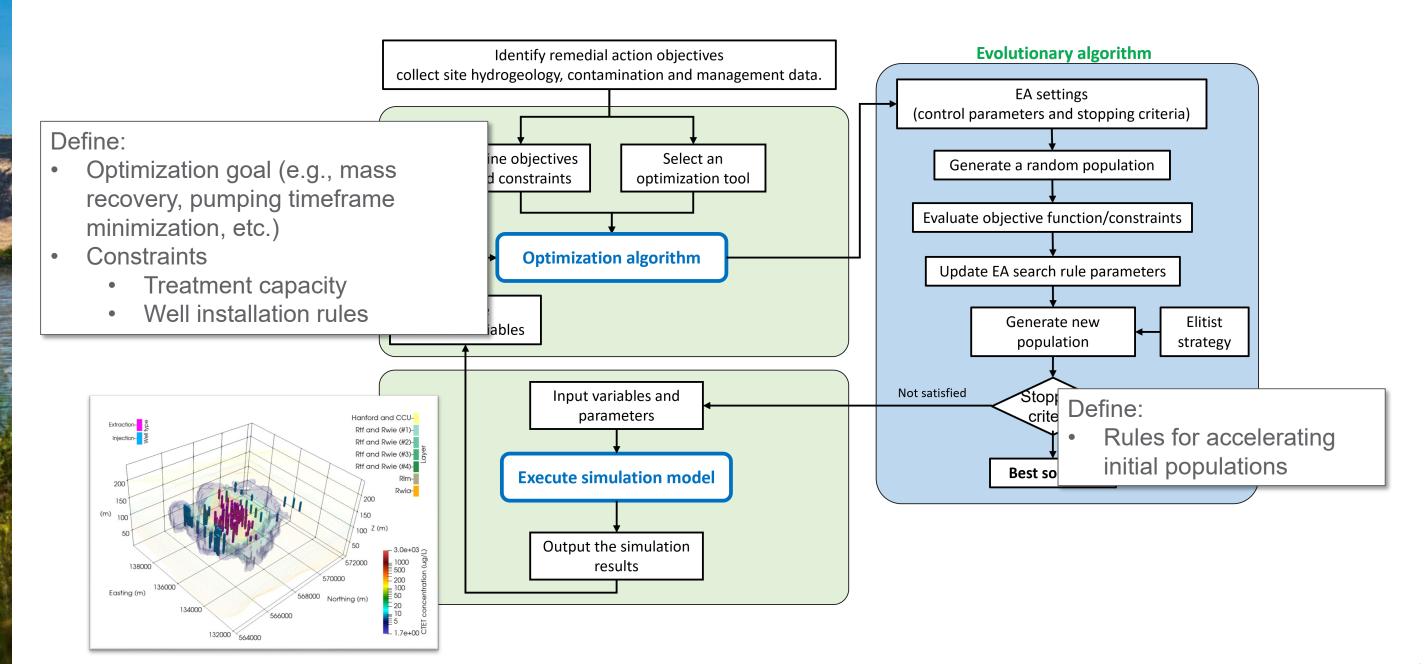
Existing extraction and injection wells







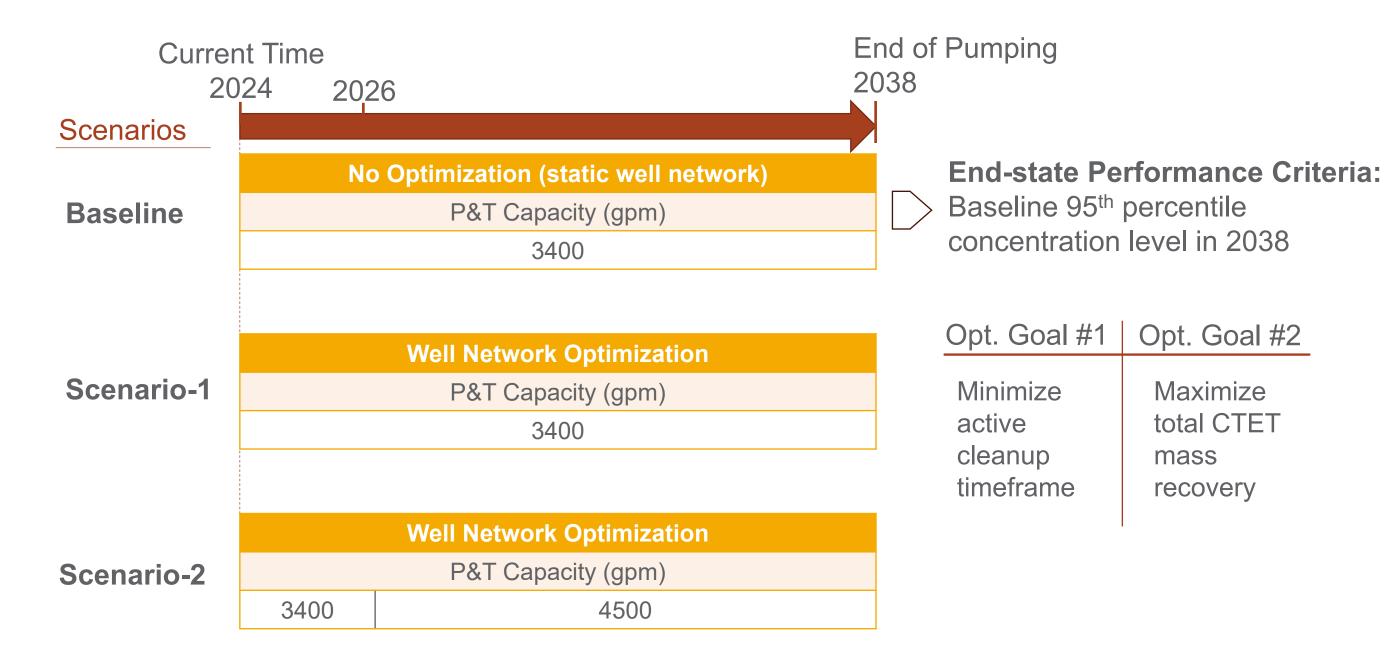
Pre-Screening Tool Optimization Setup





Optimization: Scenario Setup for Comparative Evaluation



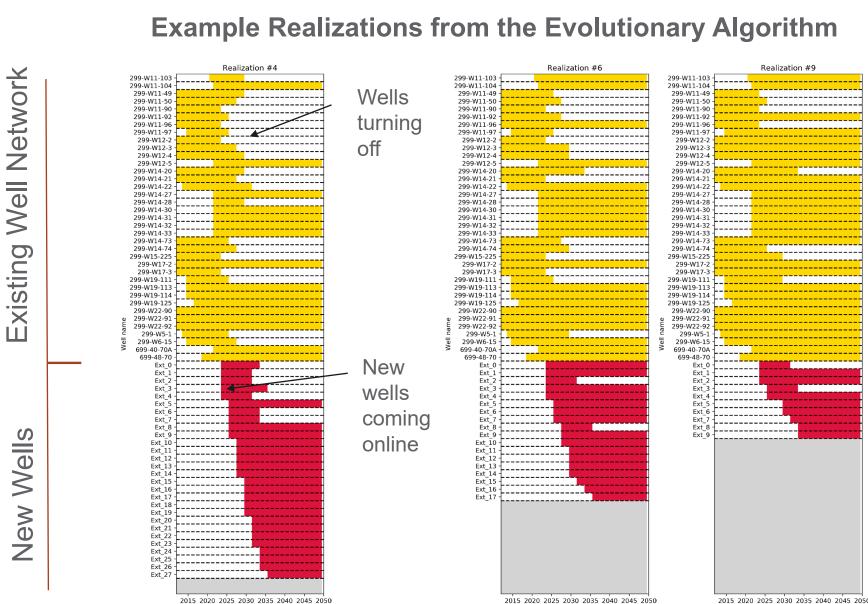




Optimization Constraints: Well Installation



- Parameterization rules:
 - Rule 1: One-to-one well replacement with a maximum number of active wells (based on the total capacity of the treatment plant)
 - Rule 2: Each well only has one operation period
 - Rule 3: Each well has a fixed pumping rate
 - Rule 4: When a new well replaces an old well, the new well inherits the pumping rates of the old one



Historical Optimization

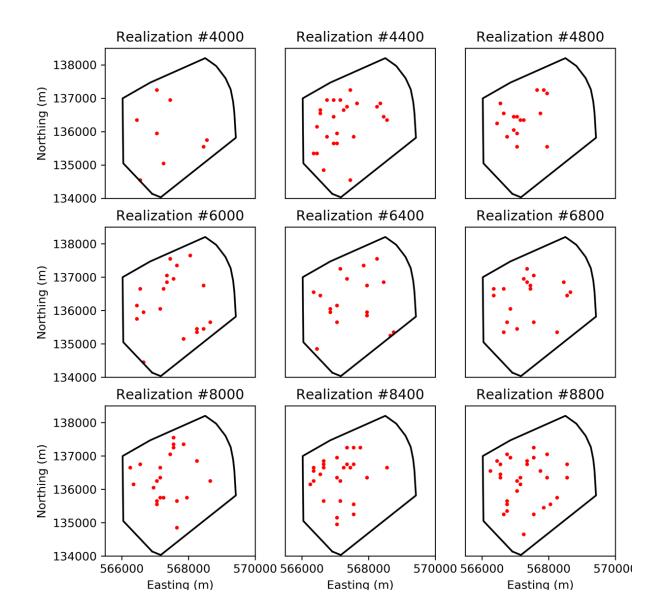
Simulation Period Simulation Period

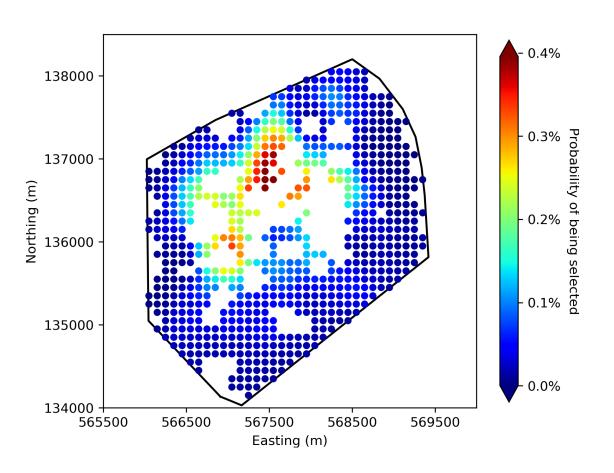


Optimization Setup: Well Locations



Concentration-weighted sampling to create the initial population for the optimization simulation





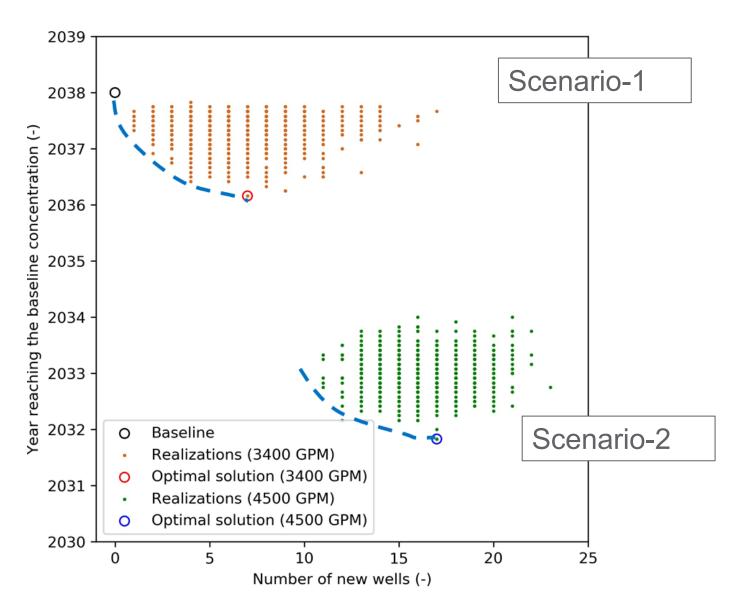
► High concentration locations are more likely to be chosen for installing new wells in initial realizations



Results for Goal #1 (Minimizing active pumping timeframe)



- Scenario-1 with well network optimization only (i.e., constant P&T capacity at 3400 gpm) achieves ~ 8% reductions in active remediation timeframe
 - Total of 7 new wells are added to the network
 - Achieves the same 95th percentile concentration level as the baseline with 2 fewer years of pumping
- ➤ Scenario-2 with well network optimization and increasing P&T capacity is found to have relatively more reduction in remedy lifetime, ~24%
 - Total of 17 new wells added to the network
 - Achieves the same 95th percentile concentration level as the baseline with 6 fewer years of pumping



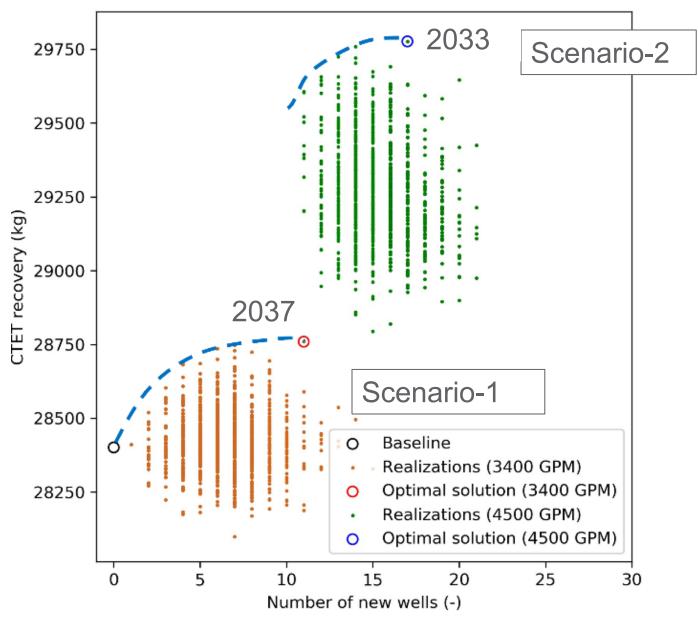
Relation between # new wells added versus pumping timeframe



Results for Goal #2 (Maximizing CTET Mass Recovery)



- Scenario-1 provides about ~ 4% reduction in pumping timeframe
 - A total of 11 wells added to the network
 - Achieves the same 95th percentile concentration level as the baseline with 1 year less pumping
- Scenario-2 provides about ~20% reduction in pumping timeframe
 - A total of 13 wells added to the network
 - Achieves the same 95th percentile concentration level as the baseline with 5 fewer years of pumping

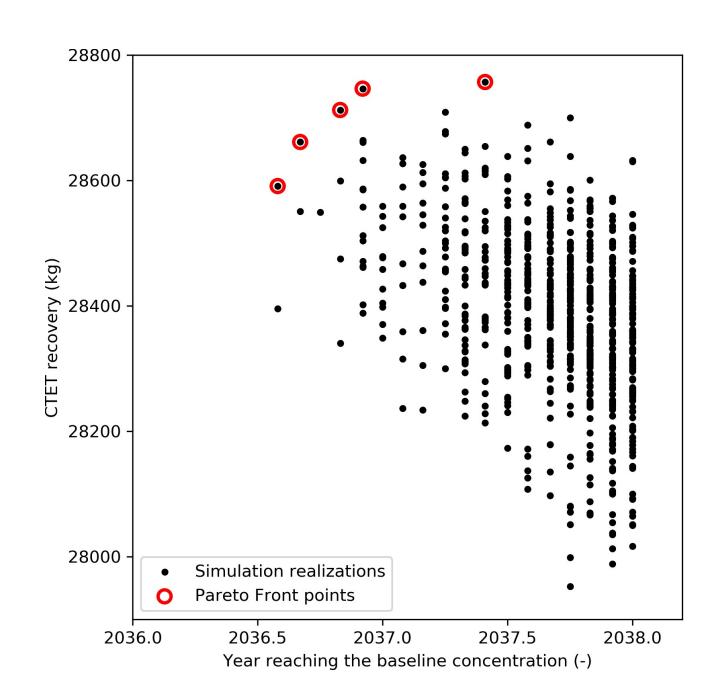


Relation between simulated total CTET mass recovery and the number of new wells added



Optimization Outcome: Pareto Front Analysis for Scenario1



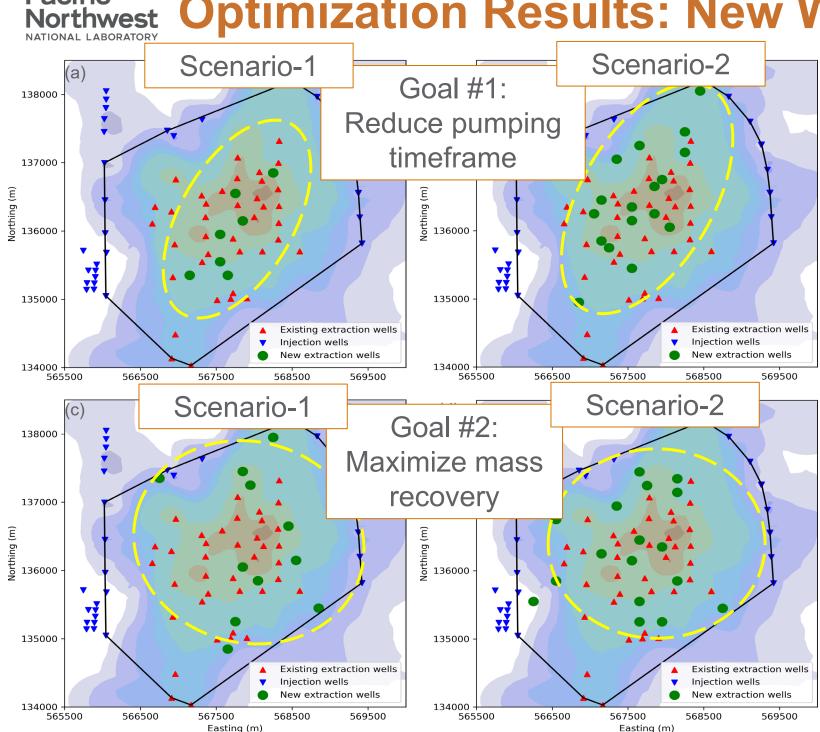


- ► The Pareto front plot provides the relationship between the two competing objectives:
 - 1. The time taken to reach the baseline CTET concentration; and
 - 2. The efficiency of CTET mass recovery
- ▶ Pareto front points: improvement in one objective will cause degradation in the other ("non-dominant" solutions)
- ► The curvature in the Pareto front indicates the rate at which these tradeoffs change



Optimization Results: New Well Locations





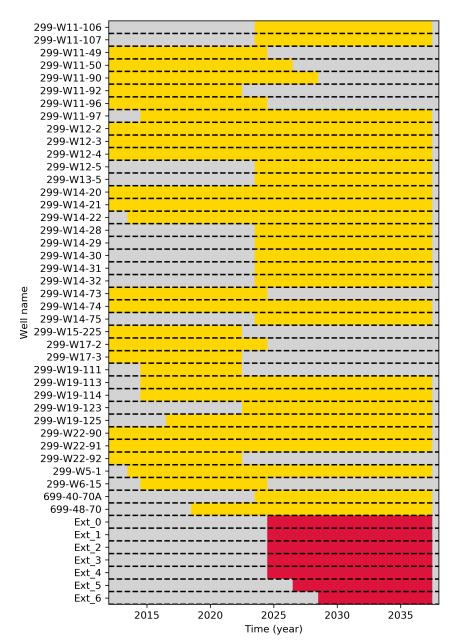
- Different optimization objectives result in varying strategies for selecting new well sites.
- ➤ To minimize cleanup time, wells are placed in high-contamination zones to lower the concentrations
- ➤ To maximize mass recovery, wells are spread out for thorough coverage, especially as concentrations decrease

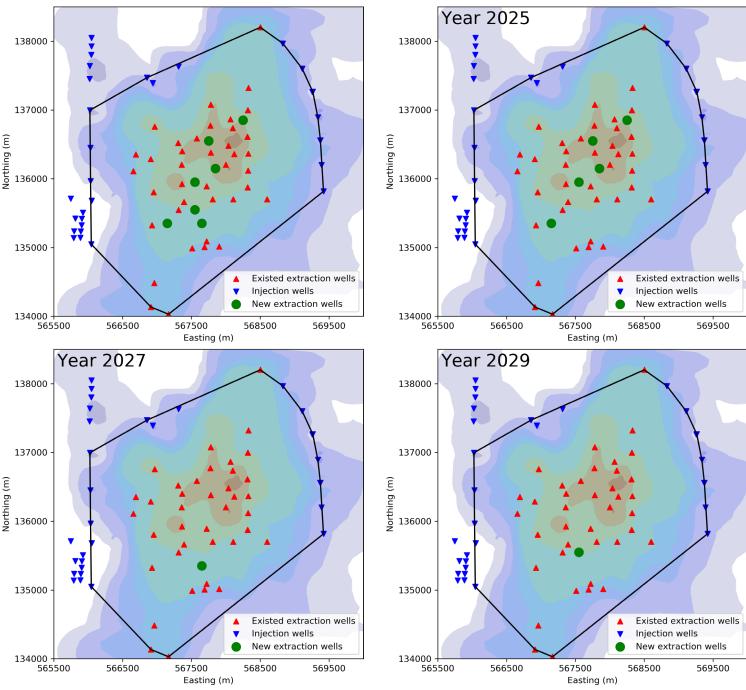


Scenario1: Optimal Solution at 3400 GPM



► Installation schedule and new well locations



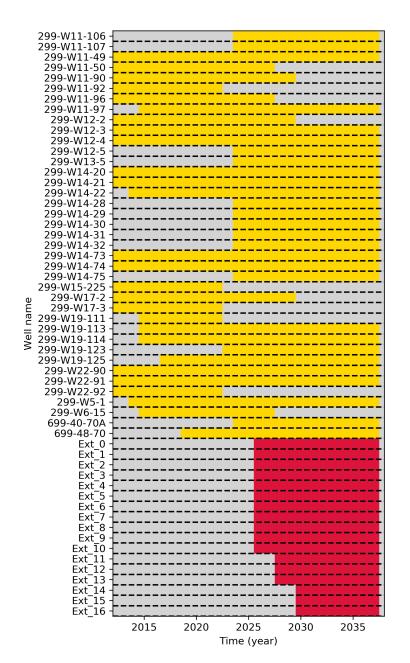


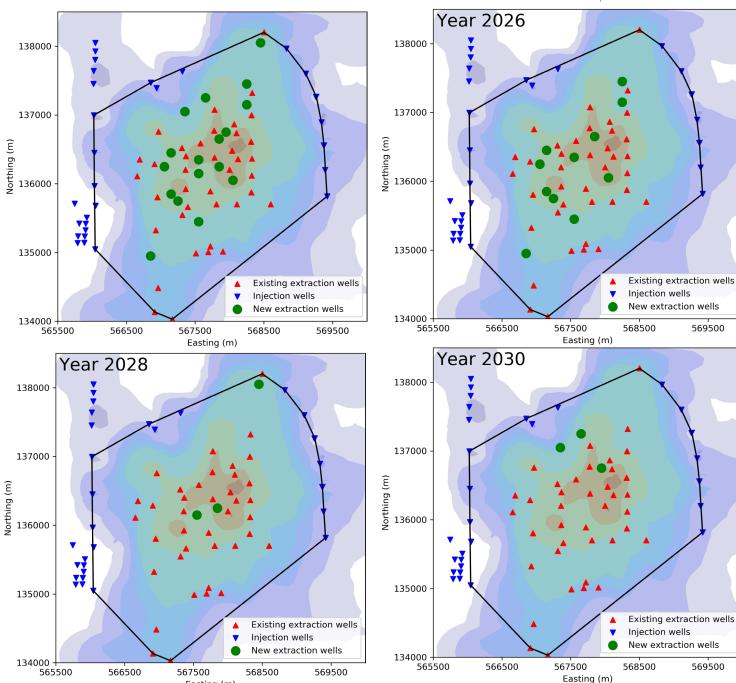


Scenario2: Optimal Solution at 4500 GPM



► Installation schedule and new well locations





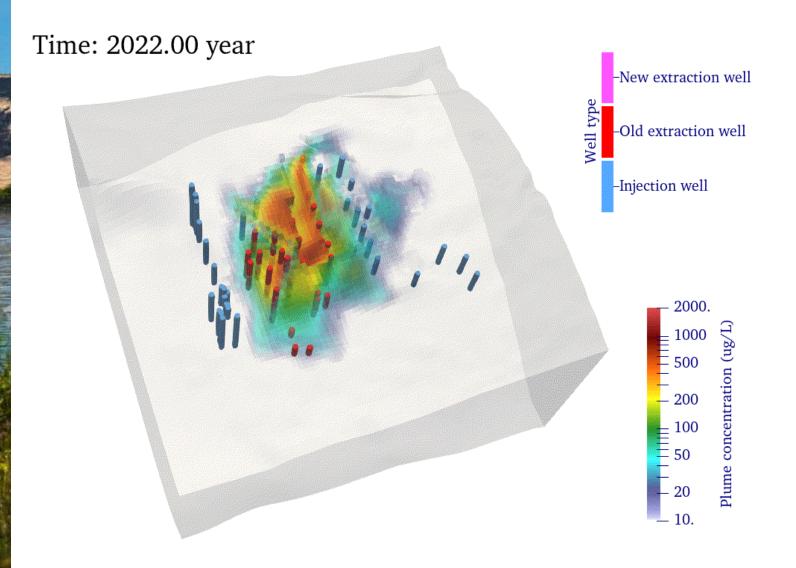
Easting (m)



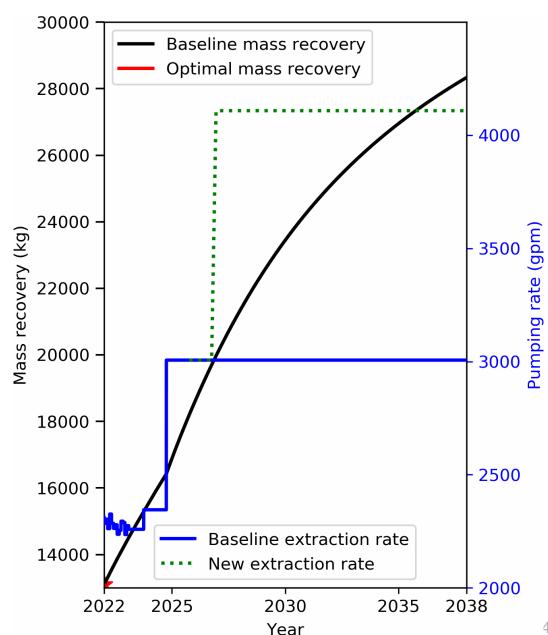
3D Animation of the Optimal Case



Predicted plume dynamics

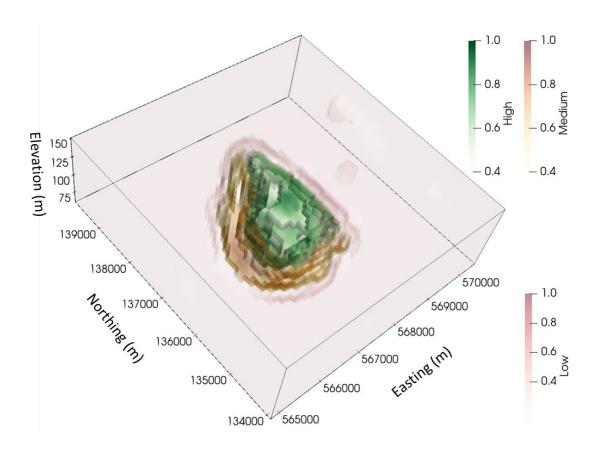


Predicted mass recovery





Part 3 – Use of Deep-Learning Approaches in the Pre-Screening Tool Framework



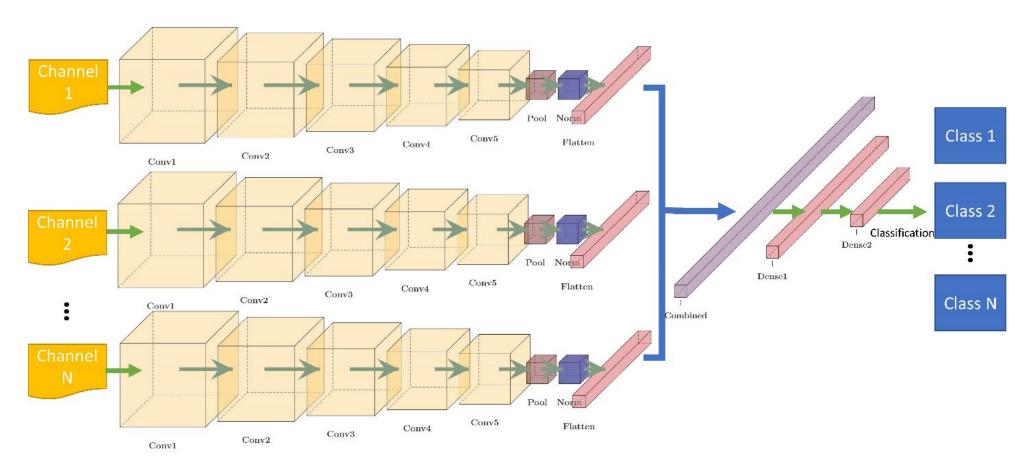






Alternative Approach: Well Location Selection

- ► A deep-learning model was developed for predicting extraction well performance for a given location in the model domain
 - Model relies on existing well performance data (2012-2023) and the data on site geology



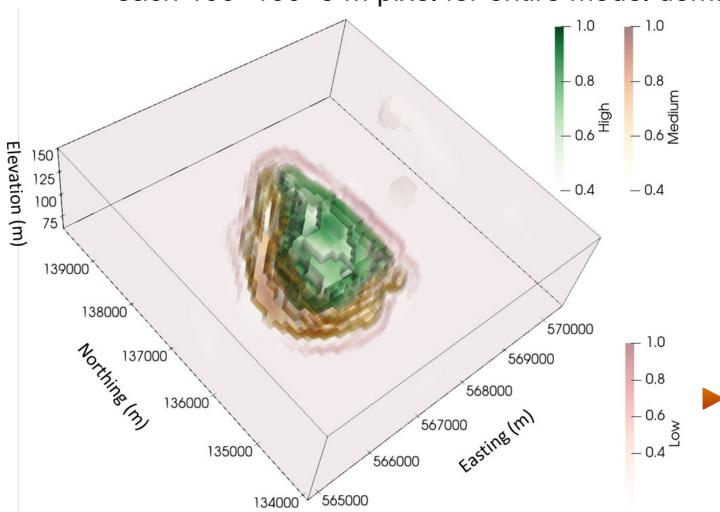
Multi-Channel Three-Dimensional Convolutional Neural Network (MC3D-CNN) framework



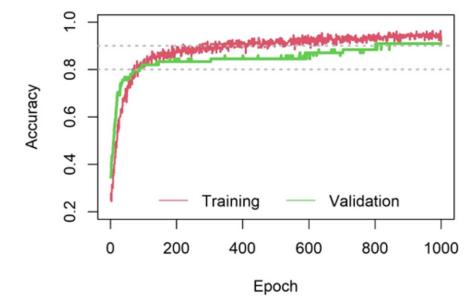
Deep Learning Model #1: A MC3D-CNN Framework for Well Performance Prediction



► The trained deep learning (DL) model was used to predict future well performance ranking on each 100×100×5 m pixel for entire model domain.



3D performance ranking map. Green, orange, and red colors indicate high, medium, and low performance rankings.



MC3D-CNN model accuracy during model training

Application of the MC3D-CNN DL model

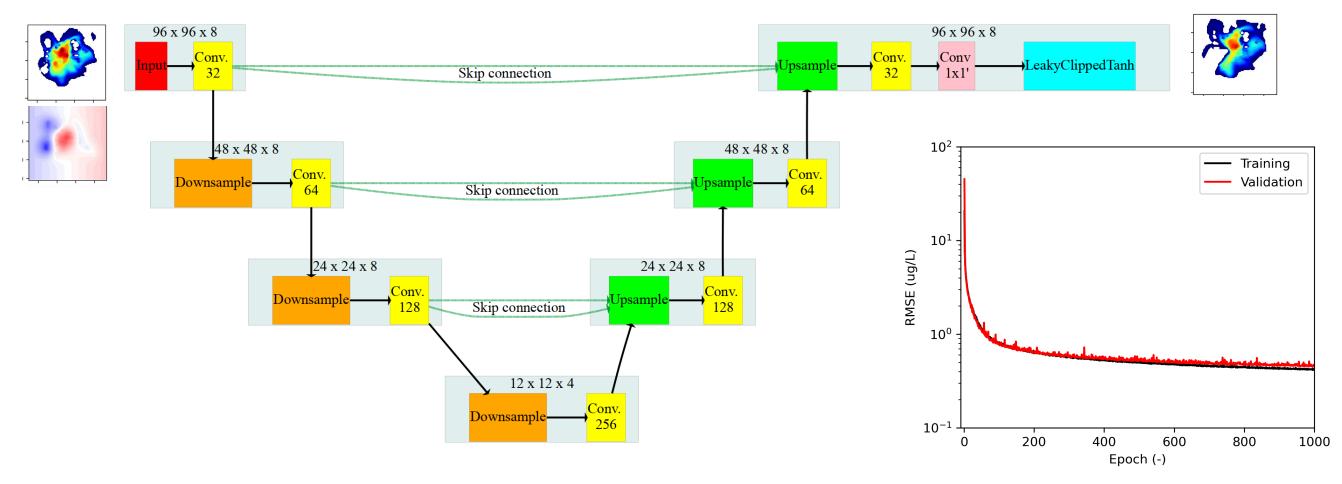
- Rebalance pumping rate of existing wells
- Reduce number of candidate well locations for P&T optimization simulation
- Integrate into flow and transport models to provide on-the-fly optimization of pumping rate





Alternative Approach: 3-D Plume Model

► An alternative deep-learning model is currently being developed to replace parts of the F&T model role in the pre-screening tool framework



U-Net Convolutional Neural Network Model Structure

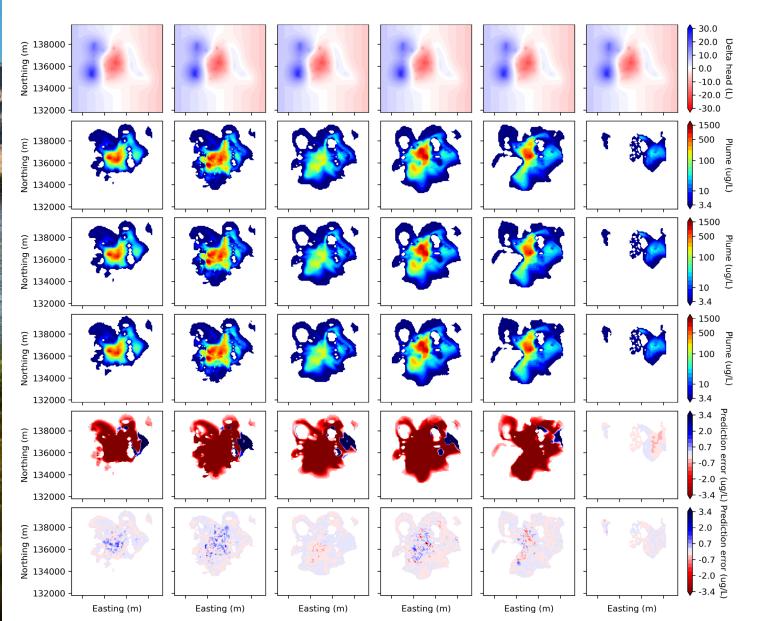
U-Net model accuracy during model training



Deep Learning Model #2: U-Net Architecture for 3-D Plume Prediction



► 100~1000 times faster surrogate model



Input #1: Thiem solution

Input #2: plume at year N

Target: plume at year N+1

Prediction: plume at year N+1

Plume change over year N (row 3-row 2)

Prediction error (row 4-row2)



Conclusions



- ► P&T optimization pre-screening tool
 - Readily allows evaluation of system behavior for multiple scenarios
 - Leads to proposed active management strategies to achieve the defined optimization goals
- ► Formal optimization of a P&T well network was demonstrated
 - Well network size, well locations, and pumping operational strategy to meet optimization goals
 - Can include treatment capacity considerations
 - Results show potentially to reduce cleanup timeframe and increase mass recovery
- ► Optimal outcomes vary with optimization objectives and constraints
 - Maximum mass recovery selected as the optimization goal may not provide the shortest cleanup timeframe
- ▶ Deep-learning approaches can significantly improve computational application of the framework



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Questions?

