

Optimizing Long Term Monitoring at a BP Site Using Multi-Objective Optimization Software

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June 15, 2004

*Conference On Accelerating Site Closeout,
Improving Performance, and Reducing Costs
Through Optimization*

Other Project Participants

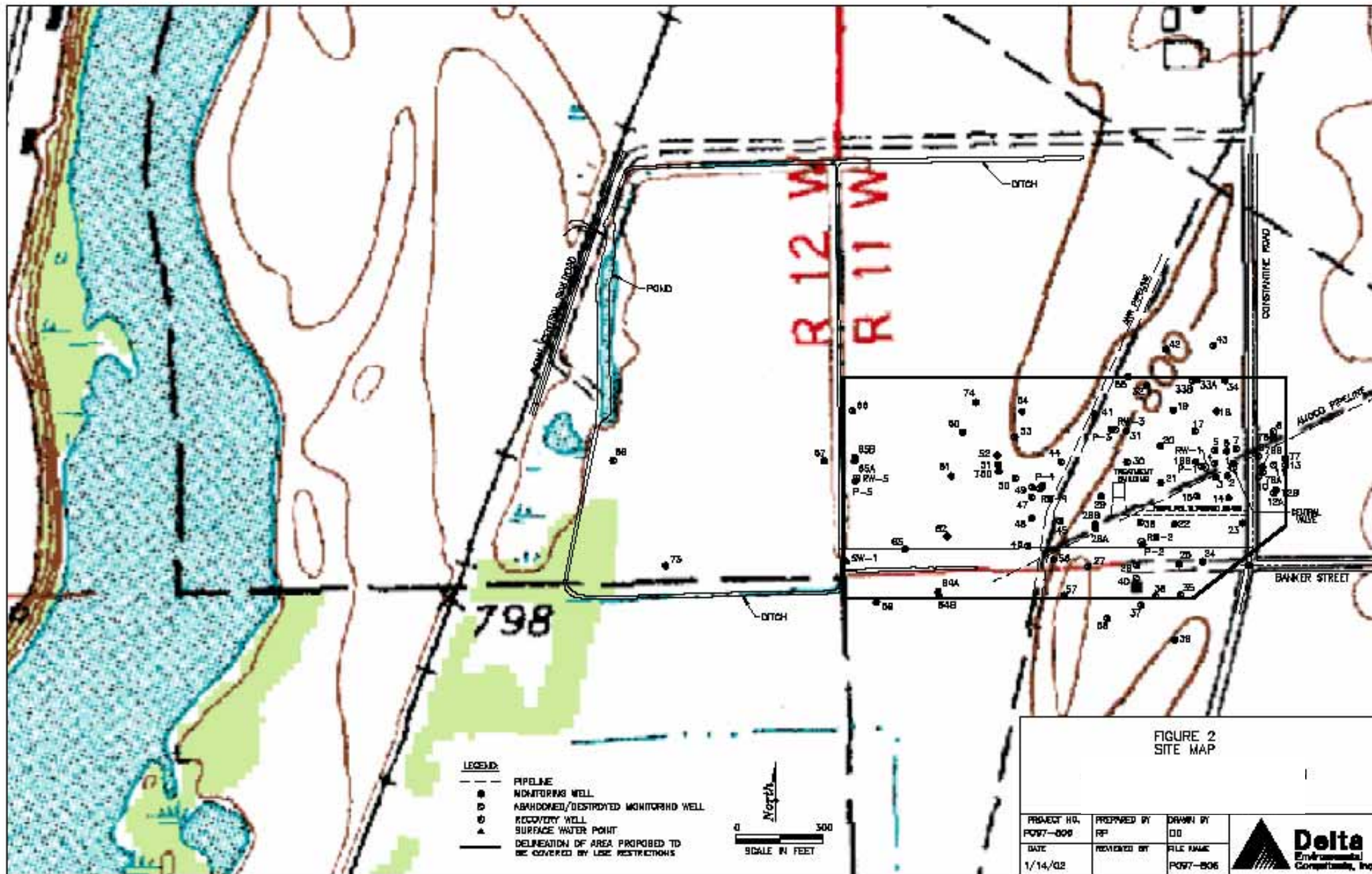
- **BP (formerly British Petroleum)**
 - Dennis Beckmann
- **Moiré, Inc.**
 - Peter Groves, Neil Kane, Tom Prudhomme
- **Delta Environmental Consultants, Inc.**
 - Jon Greetis
- **Meghna Babbar**

Outline

- **Introduction and project objectives**
- **Site background**
- **Optimization process**
- **Results**
- **Conclusions**

Introduction and Objective

- **Long-term monitoring (LTM) costs can be substantial**
- **Optimization to eliminate data redundancies can help reduce costs**
- **Objectives:**
 - Demonstrate how mathematical optimization can be used to reduce LTM costs by eliminating data redundancies.
 - Develop an optimized long-term groundwater-monitoring plan for a BP site in Michigan
 - Number and placement of monitoring wells.



- LEGEND:**
- PIPELINE
 - MONITORING WELL
 - ⊙ ABANDONED/DESTROYED MONITORING WELL
 - ⊕ RECOVERY WELL
 - ▲ SURFACE WATER POINT
 - DELINEATION OF AREA PROPOSED TO BE COVERED BY USE RESTRICTIONS

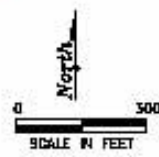


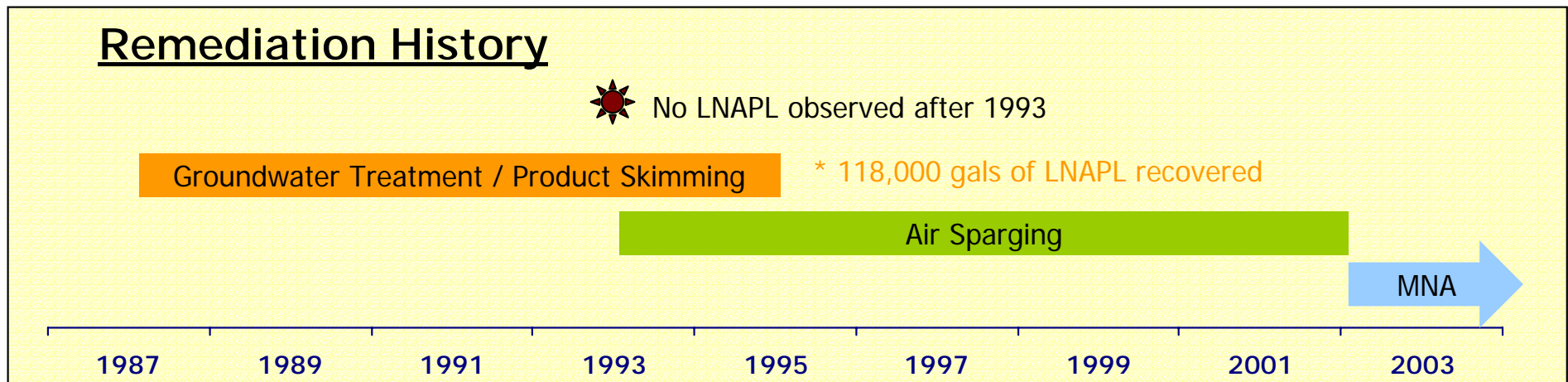
FIGURE 2
SITE MAP

PROJECT NO. PC97-006	PREPARED BY RF	DRAWN BY DD
DATE 1/14/02	REVIEWED BY	FILE NAME PC97-006



Release History

- Remedial Actions began in 1987 when a leaking pipeline gasket was discovered
- Catastrophic Release - estimates of the volume released are in the range of 350K gallons



Long-Term Monitoring Scenarios / Drivers

- **Discontinuation of air sparging operation primarily based on:**
 - Technical impracticability
 - Planned use of groundwater use restrictions
- **Natural attenuation provides plume stability with institutional controls to address residual hydrocarbons in source area.**
- **14 years of monitoring data to support plume stability assertion.**

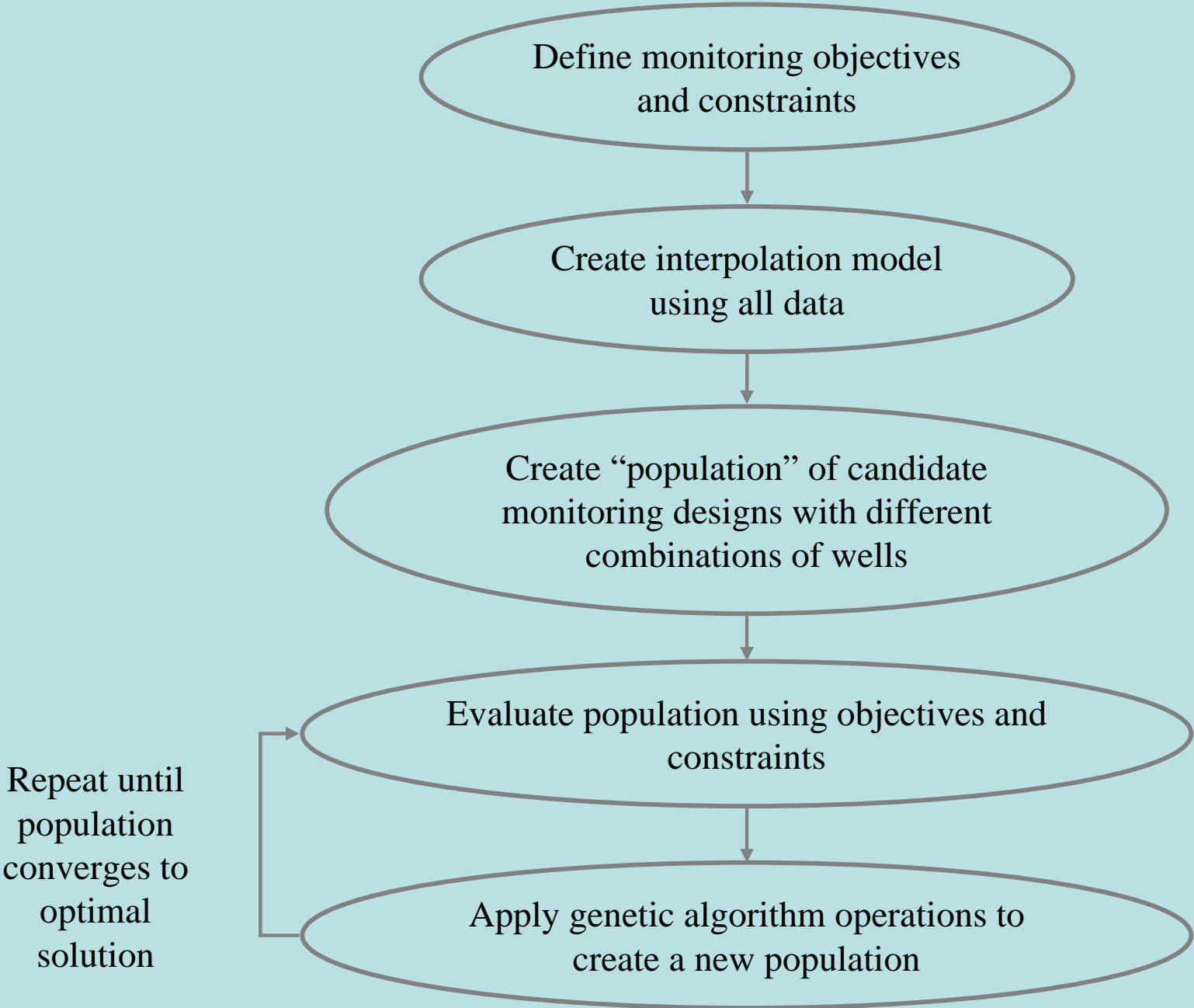
Long-Term Monitoring Scenarios / Drivers

- **MDEQ Response:**
 - MDEQ will require 30 years of post-closure monitoring
 - Costs could reach \$400,000 over 30 years
- **Optimization can be used to reduce costs of monitoring by eliminating data redundancy.**

Redundancy Analyses

- **Spatial**
 - Wells that are spatially redundant provide information (usually on concentrations) that can be obtained from other nearby wells without substantially increasing errors
- **Temporal**
 - Temporal redundancy analyses identify reductions in monitoring frequencies based on redundant information from the same set of wells
- **Spatial Redundancy (BTEX) was evaluated in this case**

LTM Optimization Process



Interpolation Modeling Process

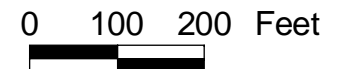
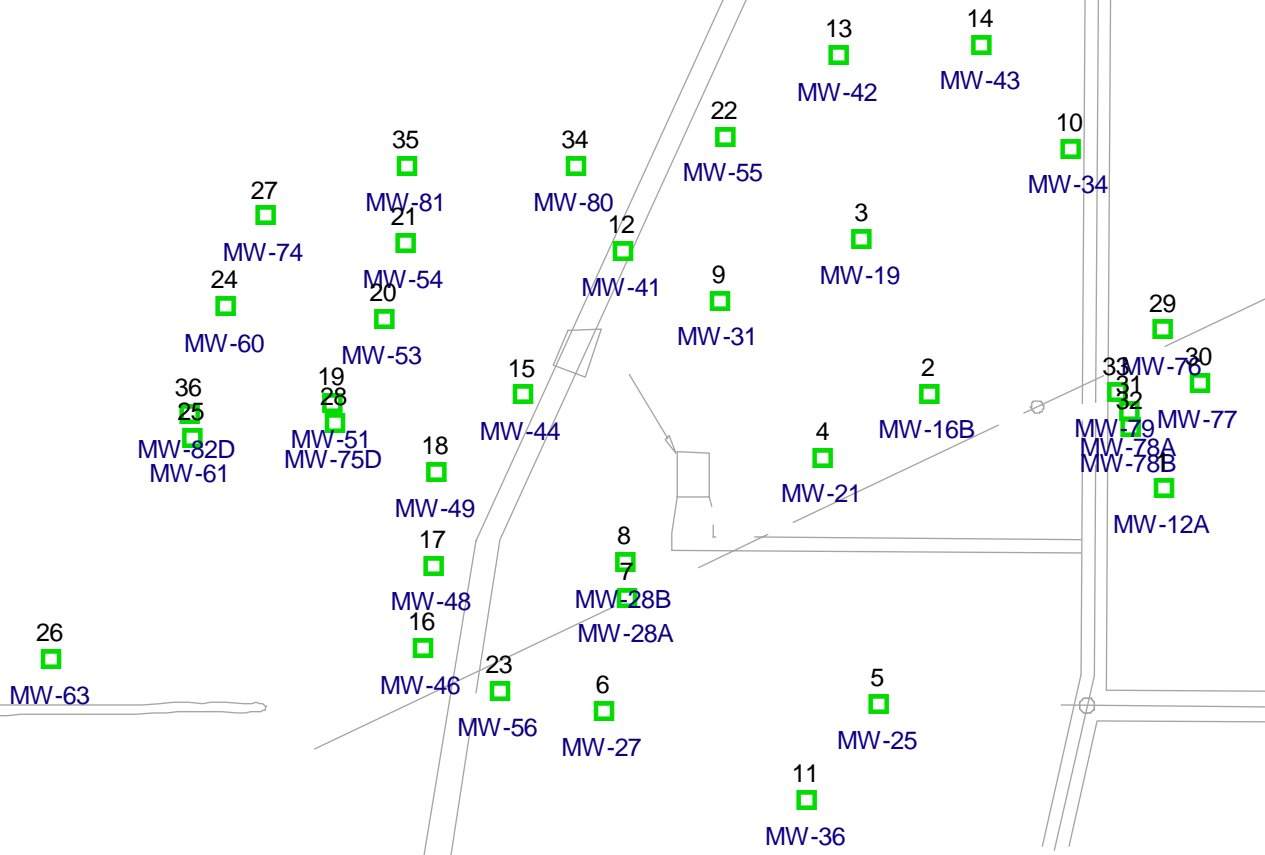
- **Identify key contaminants of concern (COC)**
- **Create spatial grid for interpolating COC concentrations**
- **Fit interpolation models**
- **Test interpolation model fit and choose model with best performance**

Spatial Interpolation Modeling



Legend:

- Current Monitoring Network - 7-Jan-04
- 14 Interpolation ID
- MW-21 Monitoring Well ID



Site Map

Date:
16 Jan 2004



Interpolation Model Evaluation

- **To test interpolation model, use cross-validation**
 - Eliminate data from well 1
 - Interpolate concentration at well 1 from data at all other wells
 - Compare interpolated concentration with measured concentration
 - Repeat for all other wells

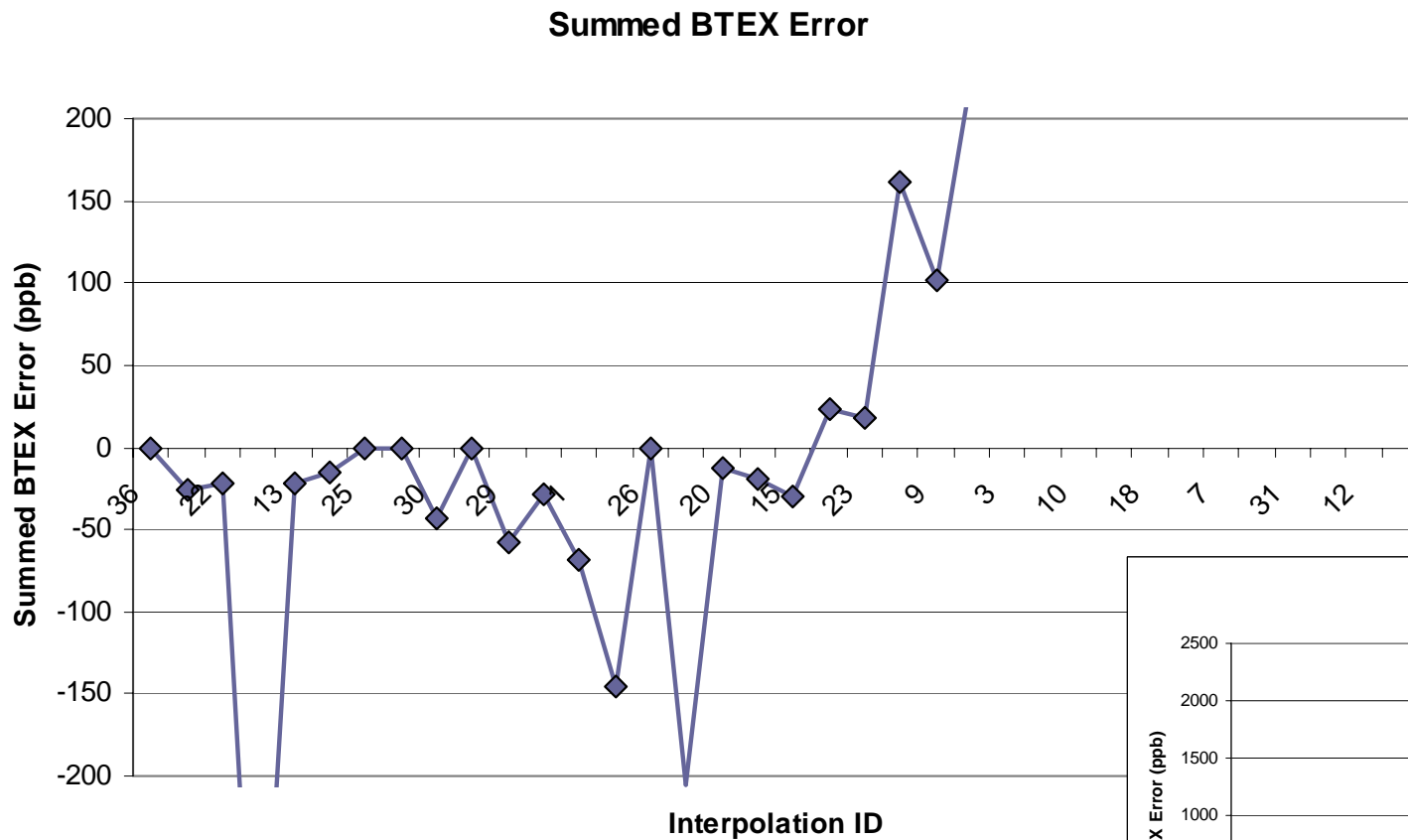
Interpolation Modeling Results Summary

- **A suite of interpolation approaches were tested**
 - Ordinary kriging
 - Quantile kriging
 - Inverse distance weighting } Most recent data only
- Neural network for detrending in time, with quantile kriging for residual - historical data
- **Quantile kriging performed best of first 3 approaches, with variograms fit to each BTEX constituent and then summed**
- **Detrending using historical data provided small increase in accuracy, but very large computational time increase**

Interpolation Modeling Results Summary (cont'd.)

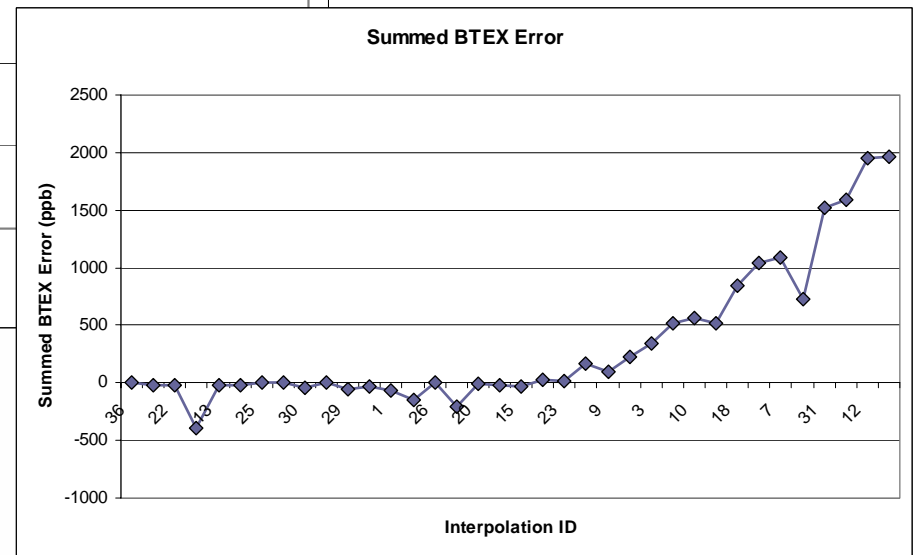
- **Of the 36 wells, the following numbers of wells were predicted sufficiently accurately during cross-validation:**
 - Benzene: 17 (within 5 ppb)
 - Toluene: 32 (within 100 ppb)
 - EthylBenzene: 28 (within 100 ppb)
 - Xylene: 23 (within 100 ppb)
 - BTEX: 19 (within 100 ppb)
- **Benzene performs quite well, but has a much stricter acceptability threshold.**
- **Summing the predictions of the components of BTEX gives a small boost in accuracy over predicting it directly.**

Cross-Validation Results for BTEX (summed from constituents)



Zoomed In

Zoomed Out



Interpolation ID (Well ID) sorted by true concentration

Optimization Process

- **Create Optimization Formulation**
 - Decision Variables
 - Objective Functions
 - Constraints (none for this site)
- **Use genetic algorithms to search for monitoring designs that best meet the objective functions and constraints**
 - When more than one objective exists, find optimal tradeoffs among objectives (e.g., cost vs. errors)

Decision Variables for This Site

$$x_i = \begin{cases} 1 & \text{if well } i \text{ is sampled} \\ 0 & \text{otherwise} \end{cases}$$

Optimization problem is to identify values of the x_i , for $i = 1$ to 36 wells

$2^{36} = 7 \times 10^{10}$ possible sampling plan designs

Objective Functions for This Site

- **Minimize Cost (no. of wells):**

$$\textit{Minimize} \sum_{i=1}^n x_i$$

- **Minimize maximum error between actual concentrations and those estimated with subset of K wells:**

$$\textit{Minimize} \left[\textit{Max}_K \left\{ \textit{Error} = \left| c_i^{\textit{actual}} - c_i^{\textit{est}} (K) \right| \right\} \right]$$

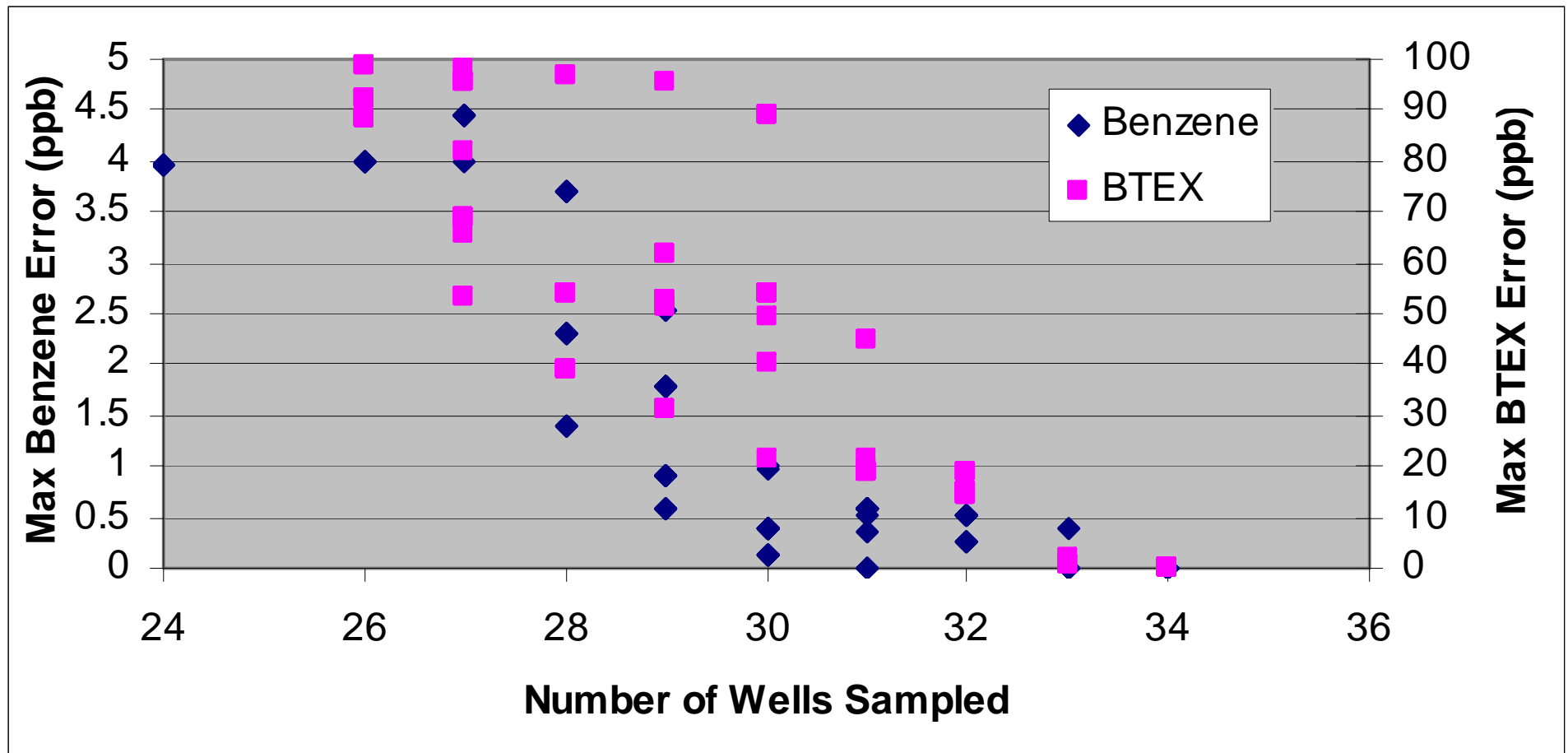
Error Objective Functions for This Site

- **One error objective for benzene and one for BTEX**
 - Scaled by maximum acceptable error (5 ppb for benzene, 100 ppb for BTEX)
- **Locations for measuring error are important**
 - At monitoring well locations only
 - Other locations in the interpolation grid have no data support, so could only compare predictions with modeled values that have errors themselves

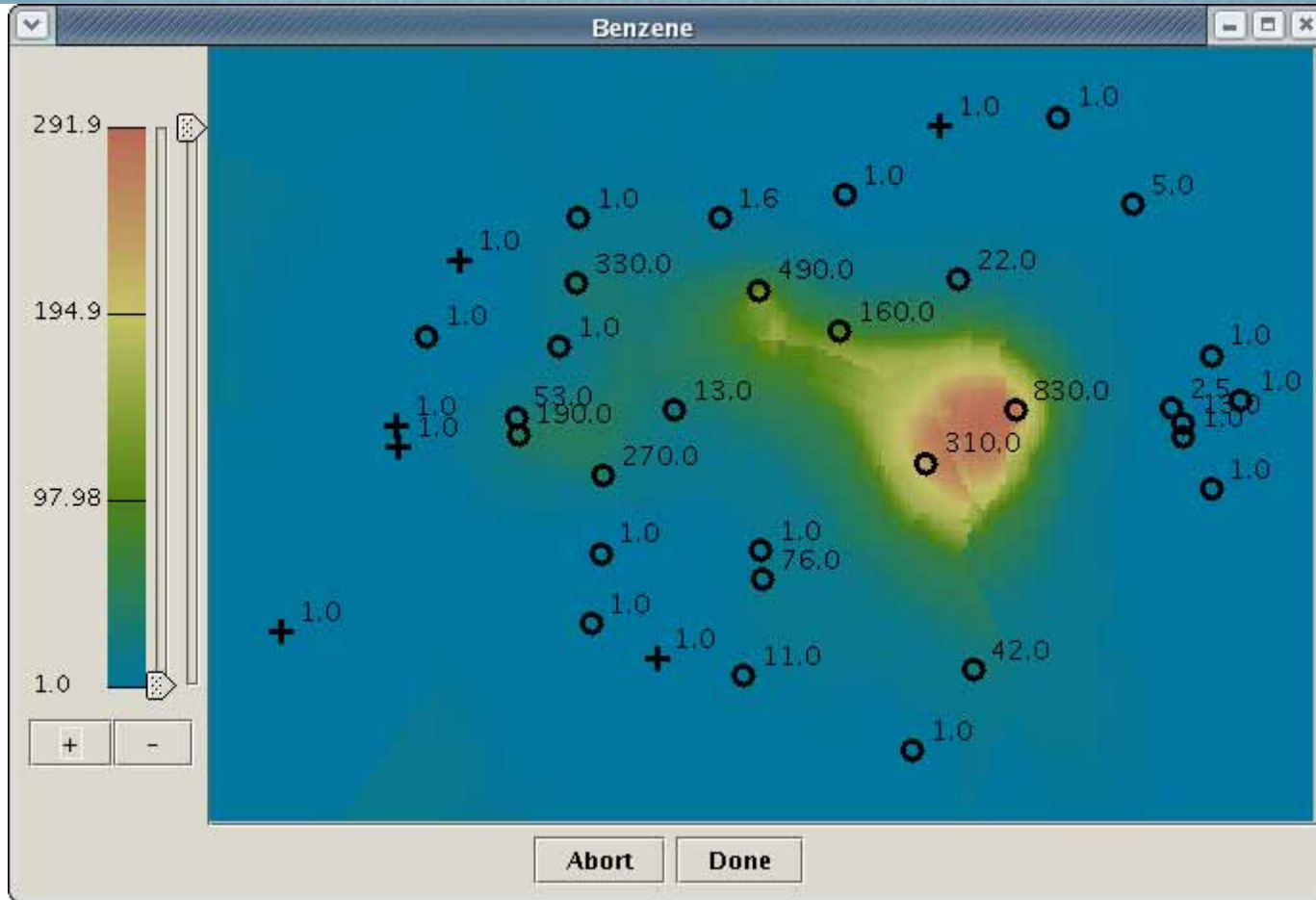
M-LTMO Software

- **Optimization process was implemented in Multi-objective Long Term Monitoring Optimizer Software (M-LTMO) developed at University of Illinois and Moire**
 - Automated interpolation model fitting and selection
 - Multiobjective optimization to find monitoring designs that best meet objectives
- **For more information and a demonstration of the software, come to the Long-Term Monitoring Optimization Methods and Software Workshop Wednesday evening from 6:30-9 PM**

Optimal Tradeoffs Between Errors and Sampling Levels

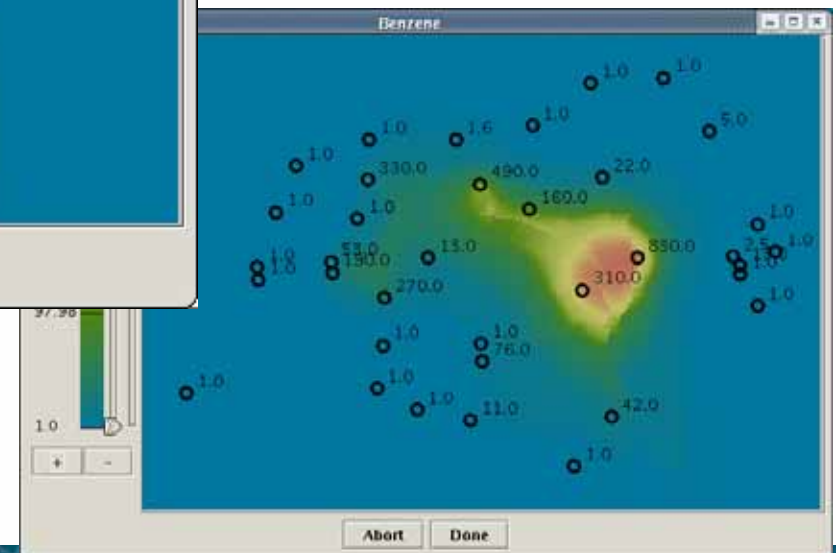


Benzene Concentrations for 30-Well Design



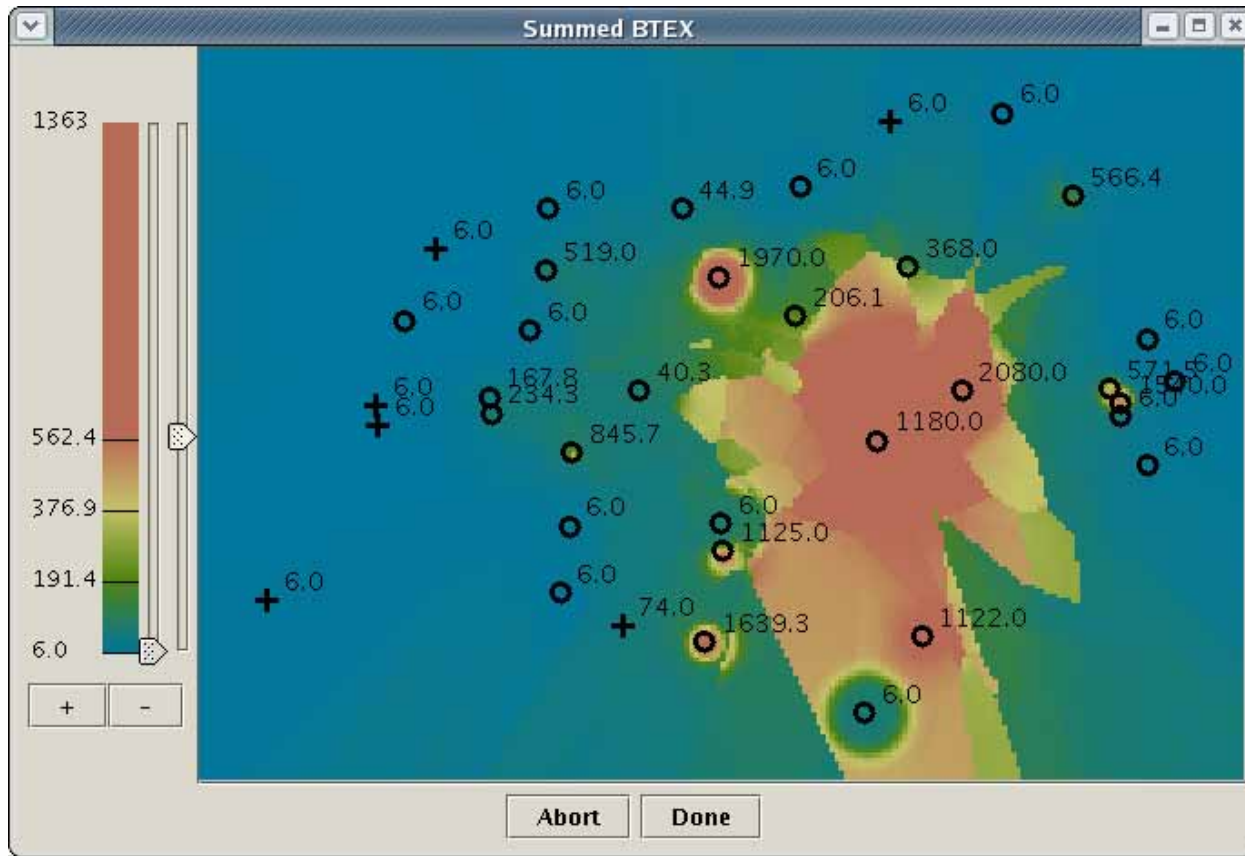
30-Well Predictions

All-Well Predictions



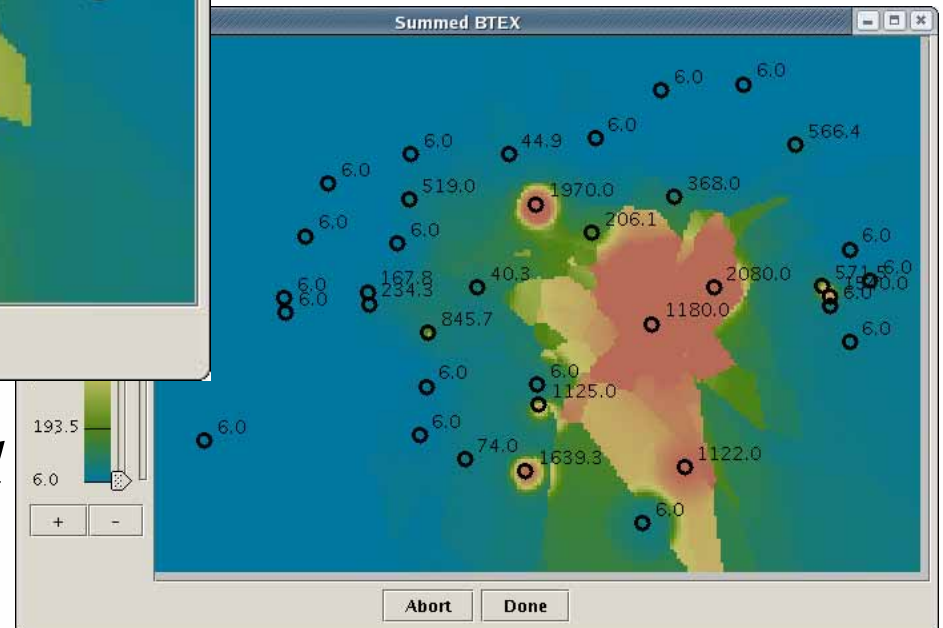
+ = *locations that are not sampled*
O = *locations that are sampled*

BTEX Concentrations for 30-Well Design



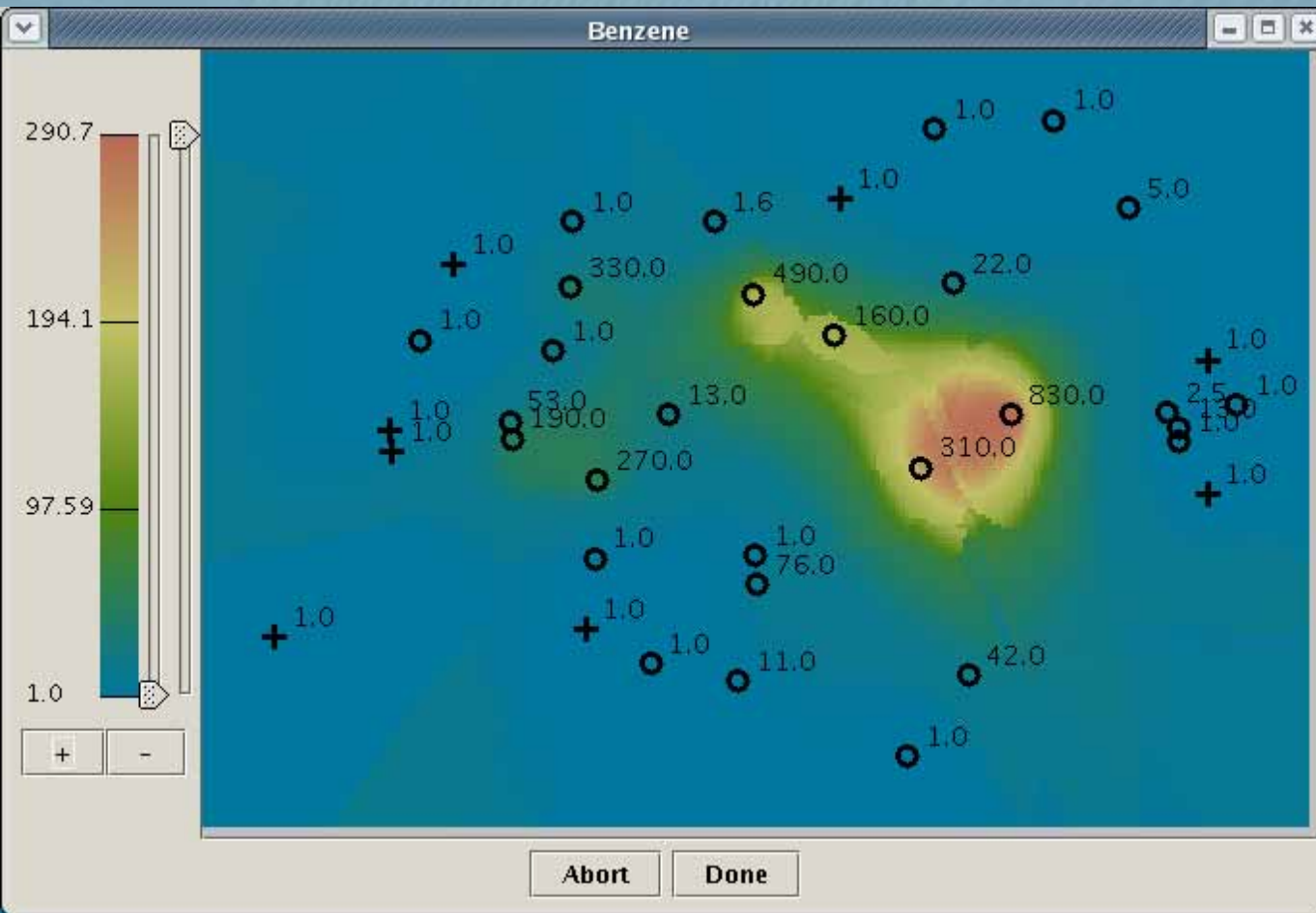
30-Well Predictions

All-Well Predictions



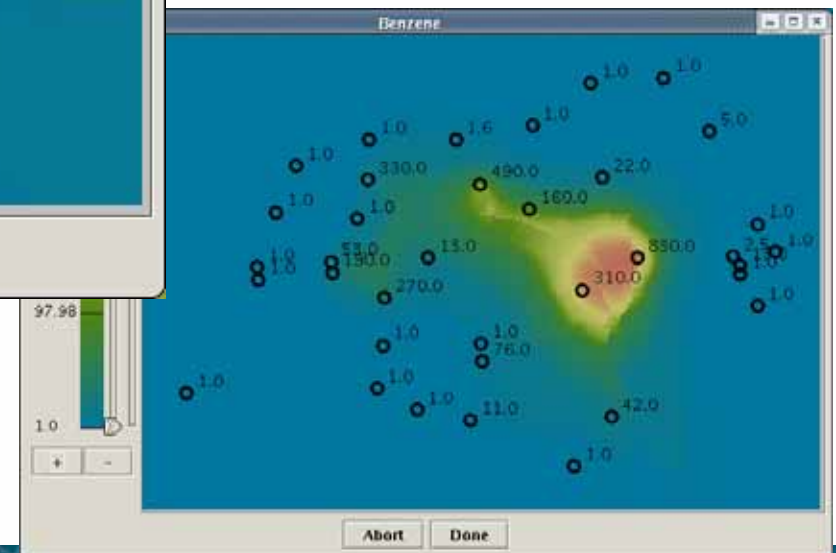
+ = *locations that are not sampled*
O = *locations that are sampled*

Benzene Concentrations for 28-Well Design



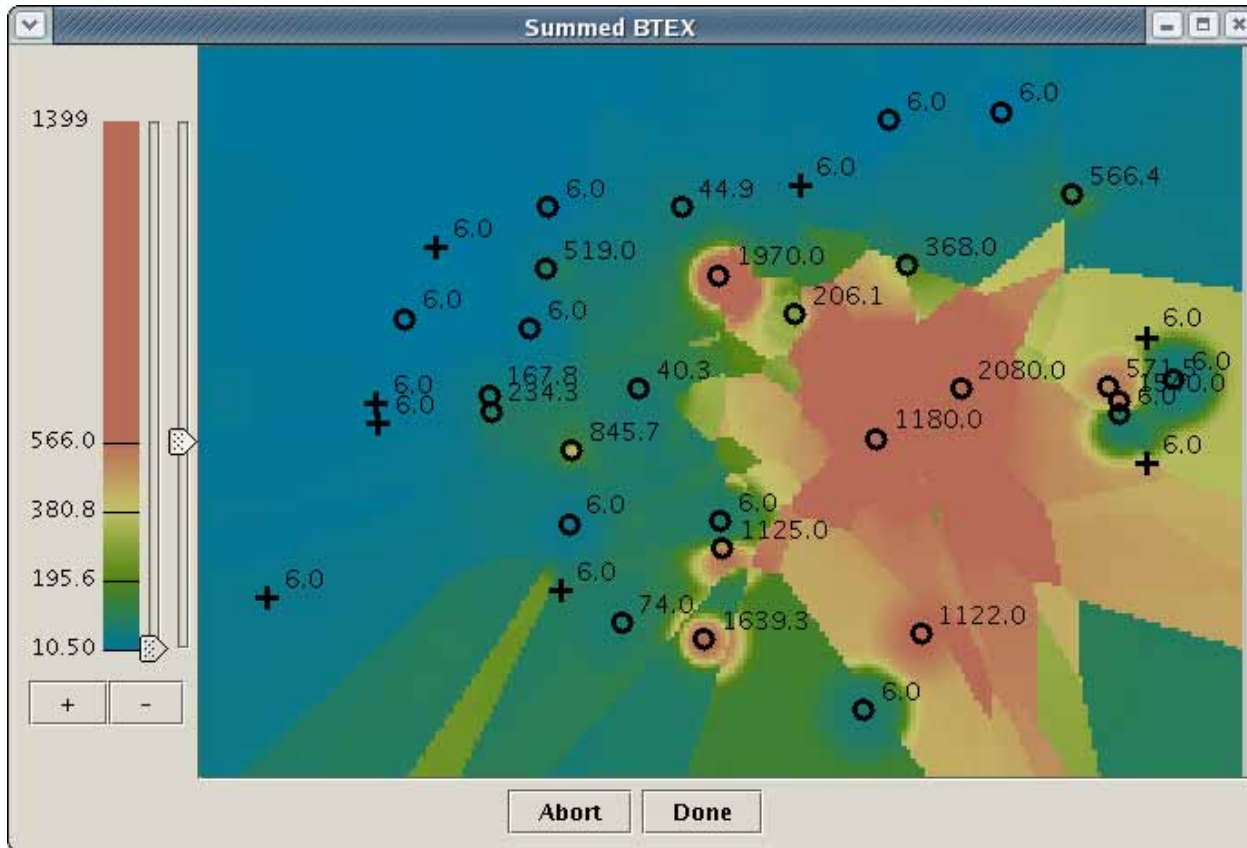
28-Well Predictions

All-Well Predictions



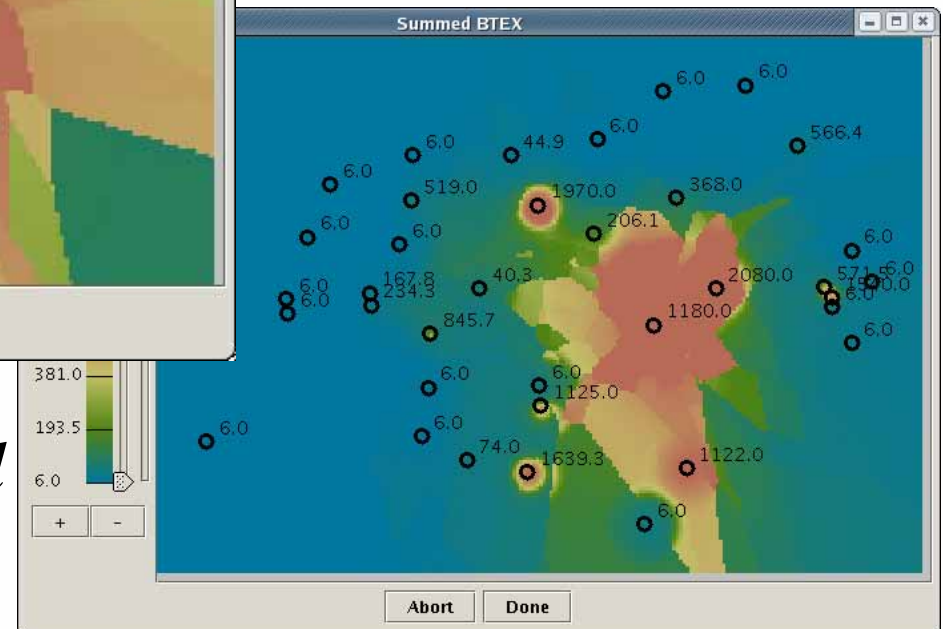
+ = *locations that are not sampled*
O = *locations that are sampled*

BTEX Concentrations for 28-Well Design



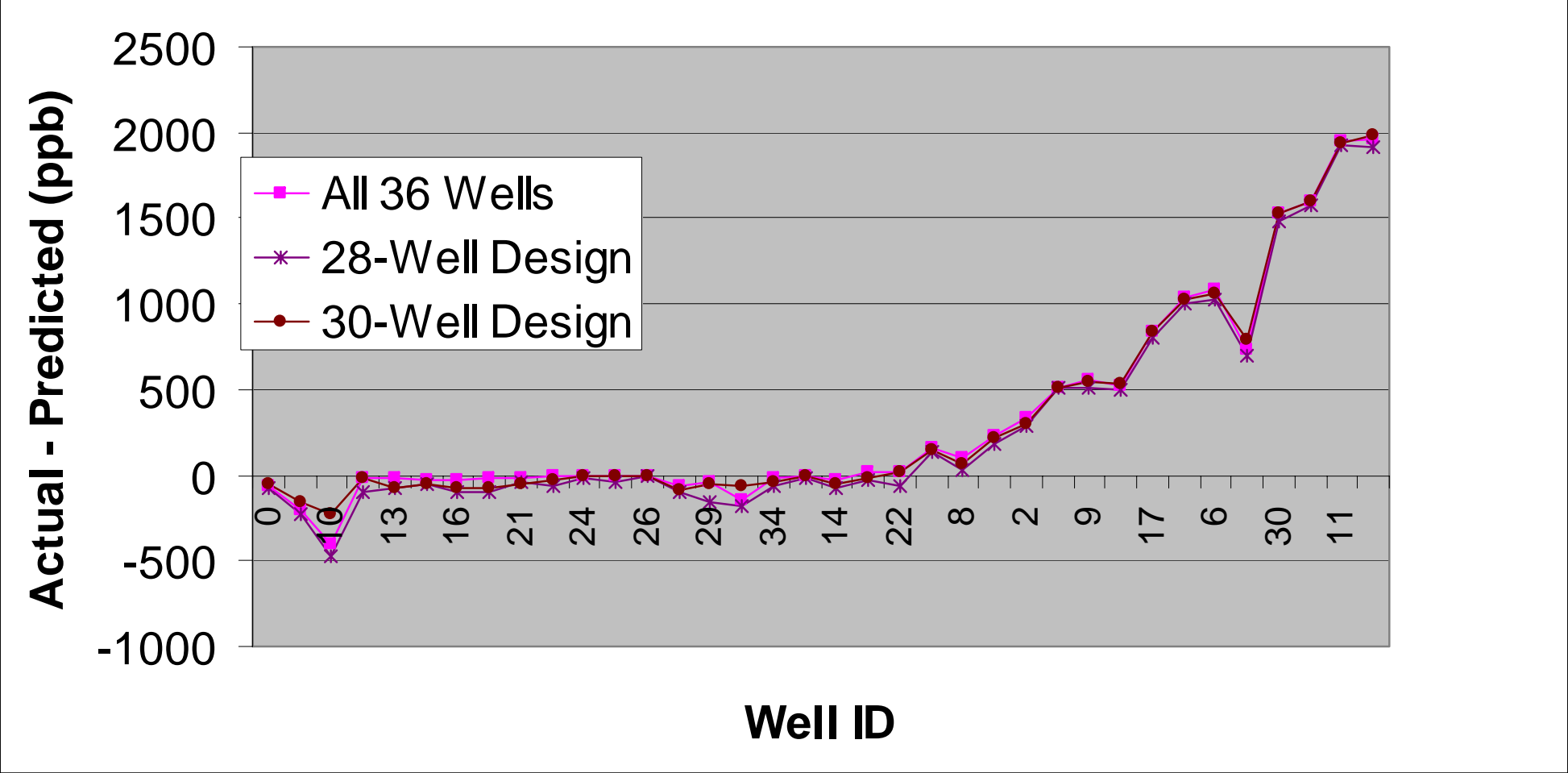
28-Well Predictions

All-Well Predictions



+ = *locations that are not sampled*
O = *locations that are sampled*

BTEX Cross-Validation Comparisons



Optimization Findings

- **Found good predictions at all well locations using 28-30 wells**
 - 17 to 22% reduction in sampling costs possible
- **28-well solution has more difficulty interpolating correctly in the southeast corner, although this area is of much less concern than the leading edge of the plume**
- **M-LTMO software is useful tool for identifying data redundancies**
- **Further testing at a New Jersey terminal site with more wells is underway**