

# HTCondor and Monte Carlo techniques to evaluate the uncertainty of water sources to groundwater wells



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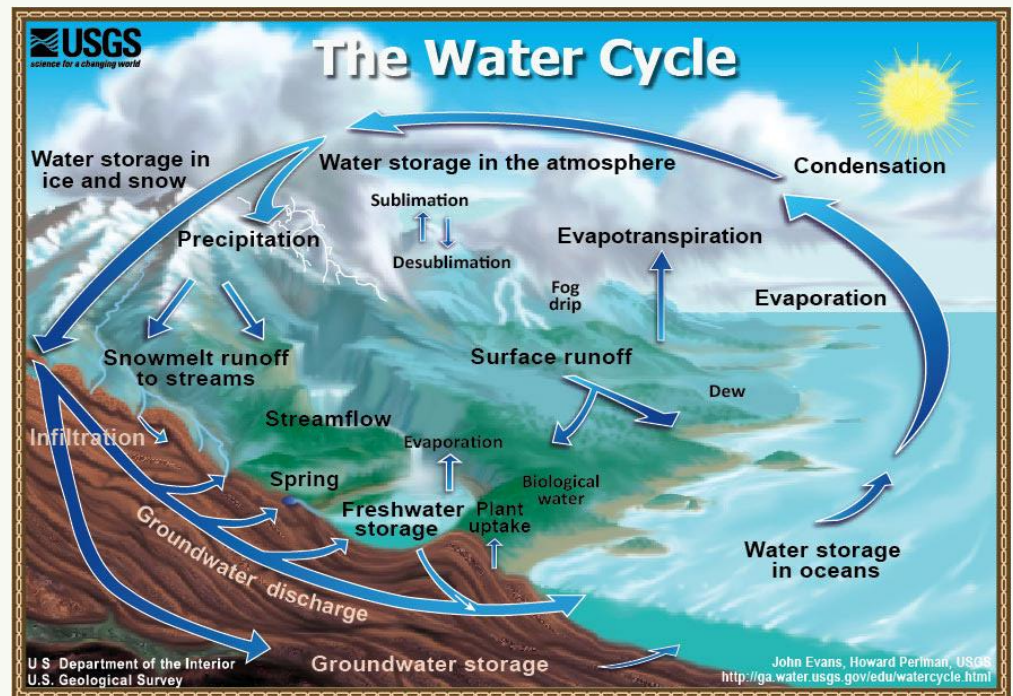
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Madison, Wisconsin



# The Source of Water to Wells

Groundwater is part of the hydrologic cycle

We can identify the surface expression of where water originates relative to a well



# Methods to determine source of water to a well

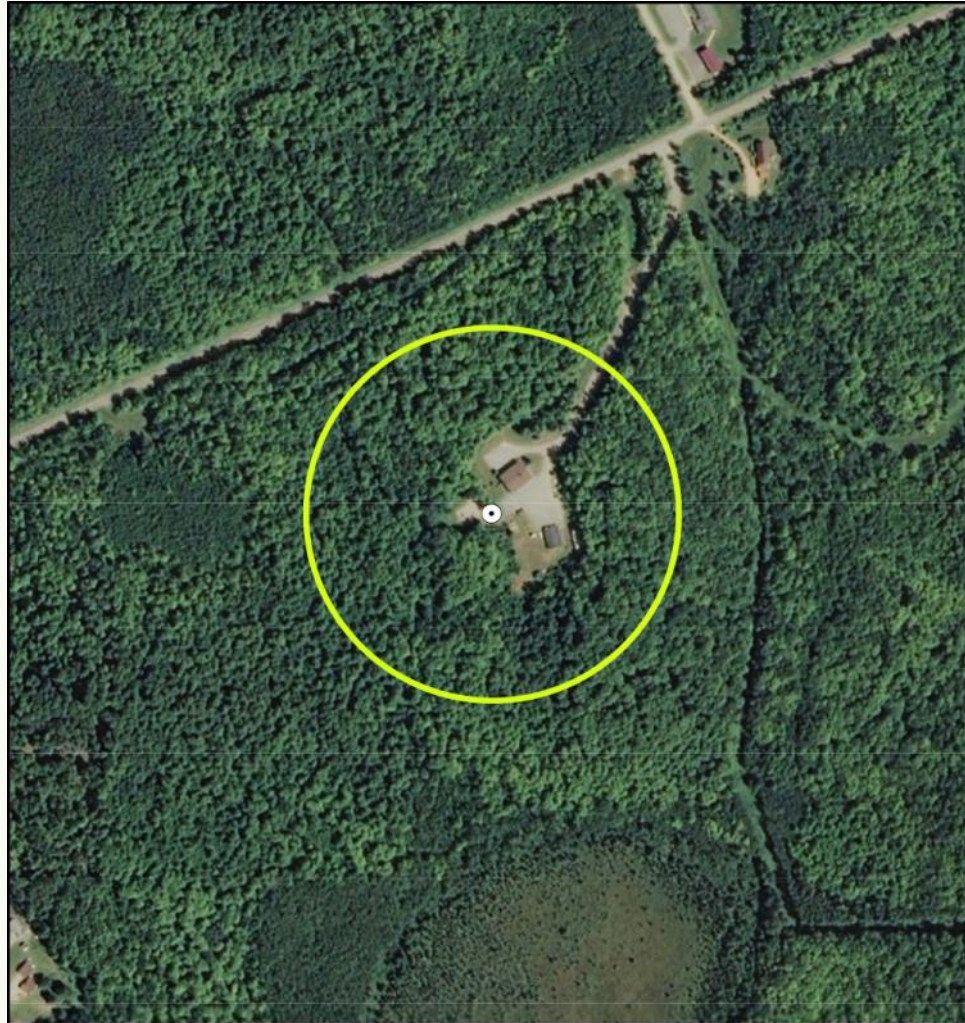
The classic method is the fixed-radius method

$$r = \sqrt{\frac{Qt}{\pi nH}}$$

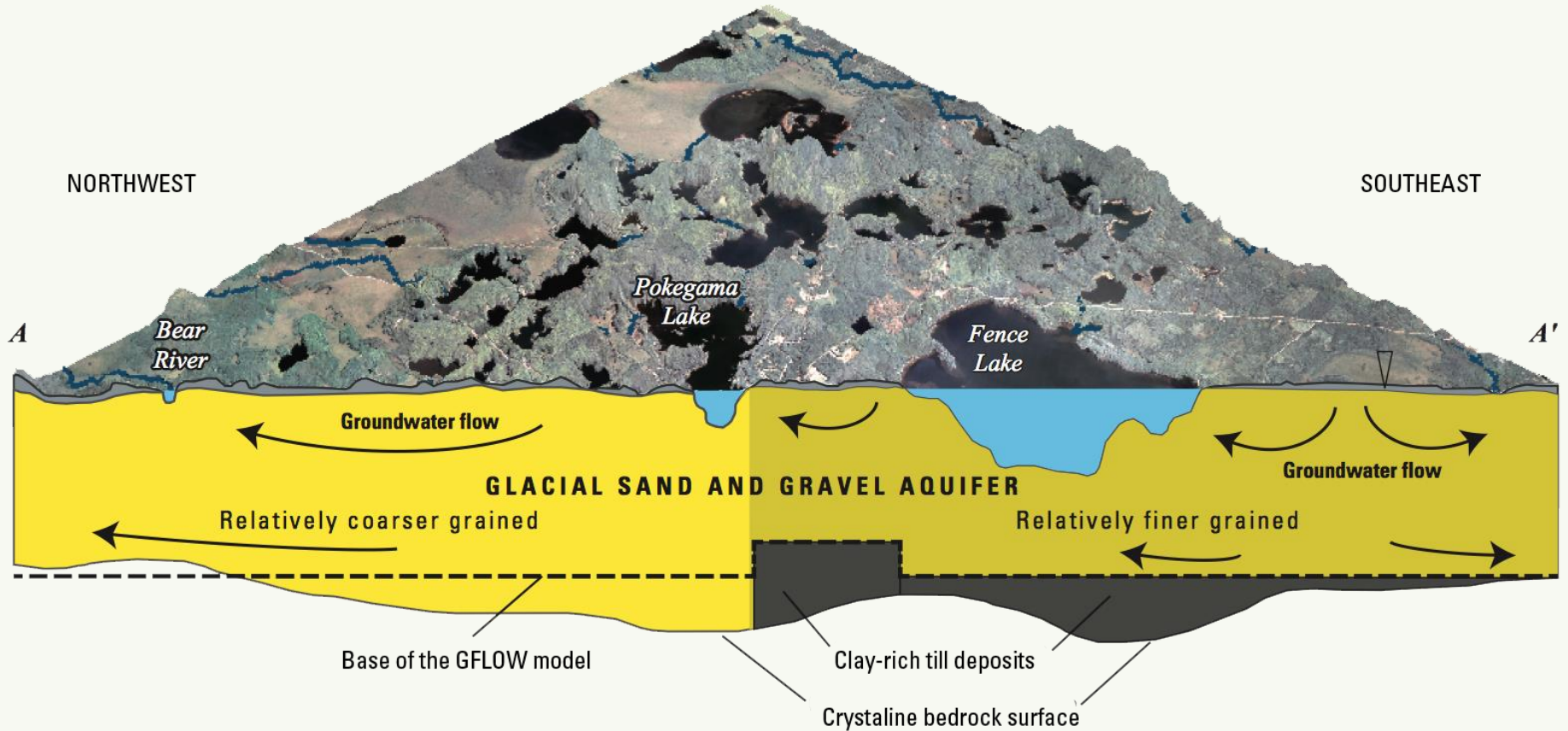
where

- $r$  is the radius,
- $Q$  is the pumping rate,
- $t$  is the time of pumping,
- $\pi$  is the mathematical constant,
- $n$  is the porosity, and
- $H$  is the aquifer thickness.

# Fixed radius results



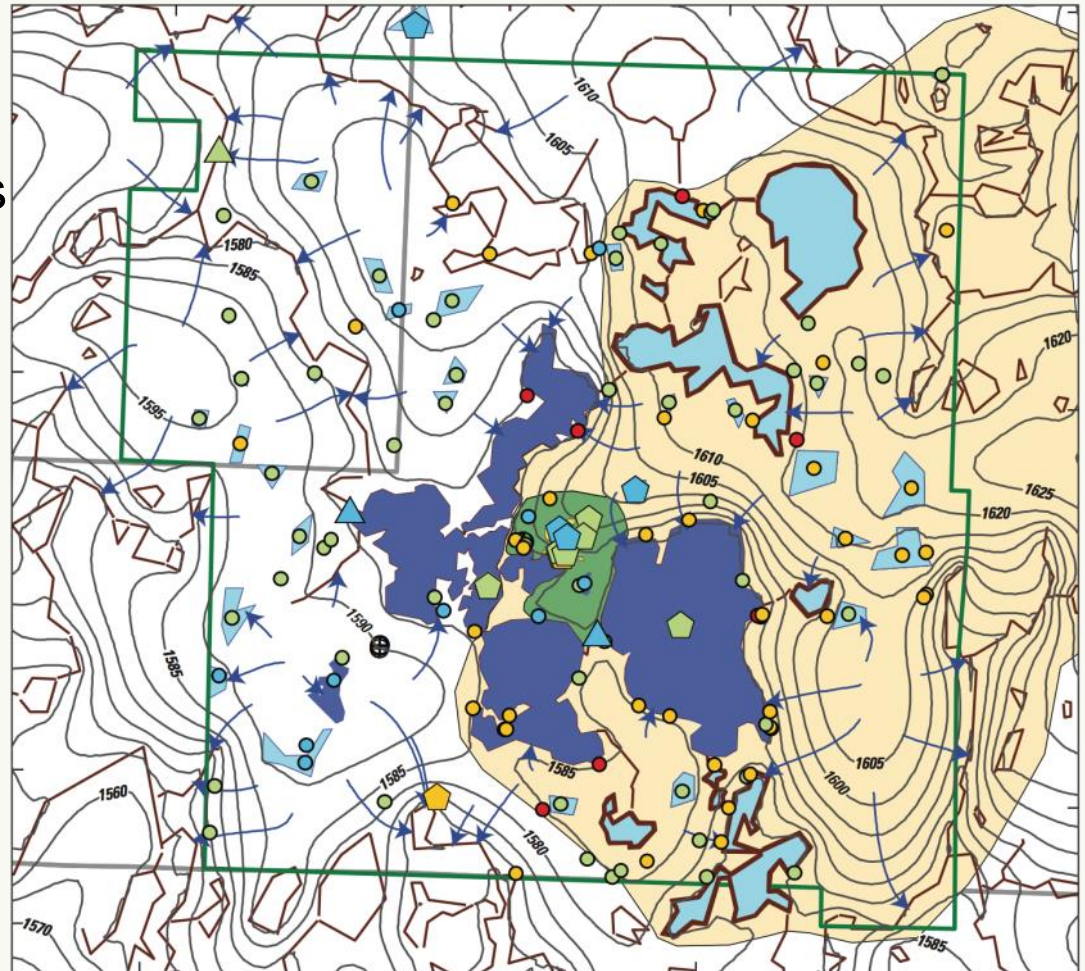
# Reality is a bit more complicated



# GFLOW is an analytic element model simulating groundwater and surface water\*

GFLOW superposes analytic  
solutions representing elements

No numerical grid (!)

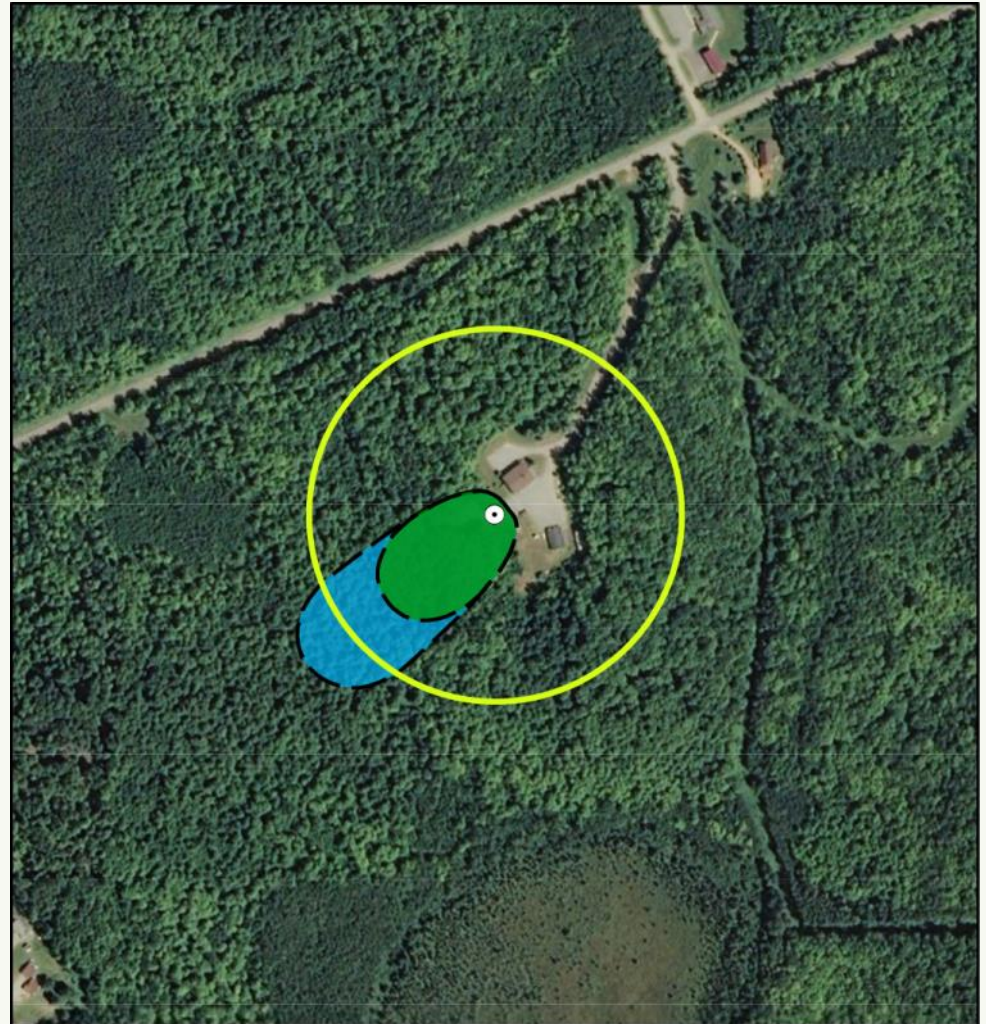


# Using GFLOW to estimate area of capture

This is clearly an improvement  
but what about uncertainty?

We can vary parameters about  
their calibrated values and see  
how that uncertainty propagates  
into the capture estimates

Monte Carlo with Latin Hypercube  
sampling strategy



# Monte Carlo with Latin Hypercube: Starting with calibration

## **PEST – *Parameter calibration and covariance***

Calibrate to heads, gradients, and flows with hydraulic conductivity (K), recharge (R), and sediment resistance (C)

Covariance matrices for calibrated parameters; estimated variances for non-calibrated parameters

## **Monte Carlo – *Uncertainty analysis***

Latin Hypercube sampling of covariance

(Starn and Bagtzoglou, 2012)

## **Custom Python codes**

End-point analysis for Areas Contributing Recharge to wells

Aquifer Mapping to synthetic grids for source tracking

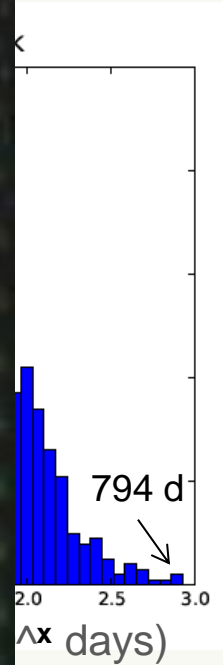
Evaluate convergence of the Monte Carlo simulations

HTCondor HTCondor



# Monte Carlo with Latin Hypercube: a few more details

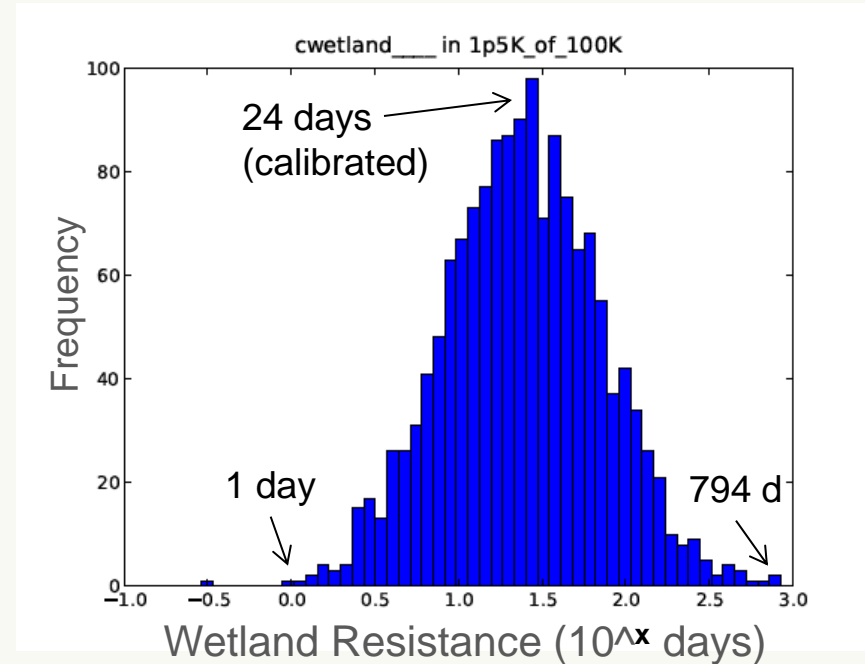
Latin H  
samplin  
covaria



# Monte Carlo with Latin Hypercube: a few more details

Latin Hypercube: Equiprobable sampling that honors parameter covariance and correlation

Model is run for each parameter set (1200 realizations)



Model runs distributed on 150 Linux CPUs using Wine  
with HTCCondor

**HTC**Condor



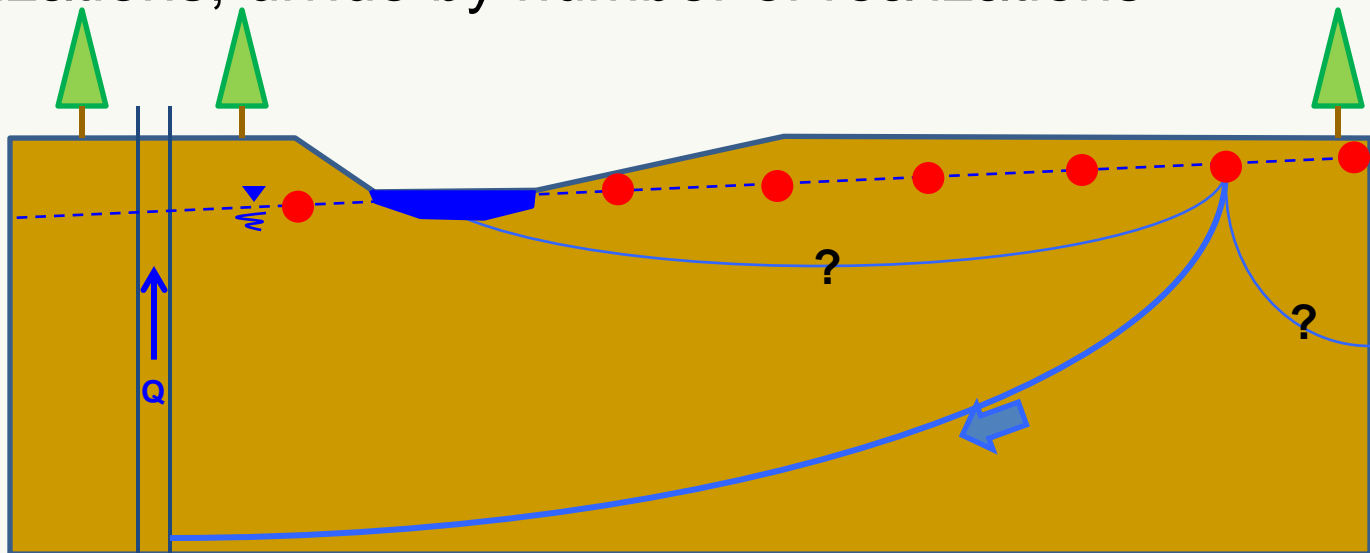
# Areas Contributing Recharge to Wells (end-point analysis)

Forward particle tracking

Read pathline output & search for the well ID

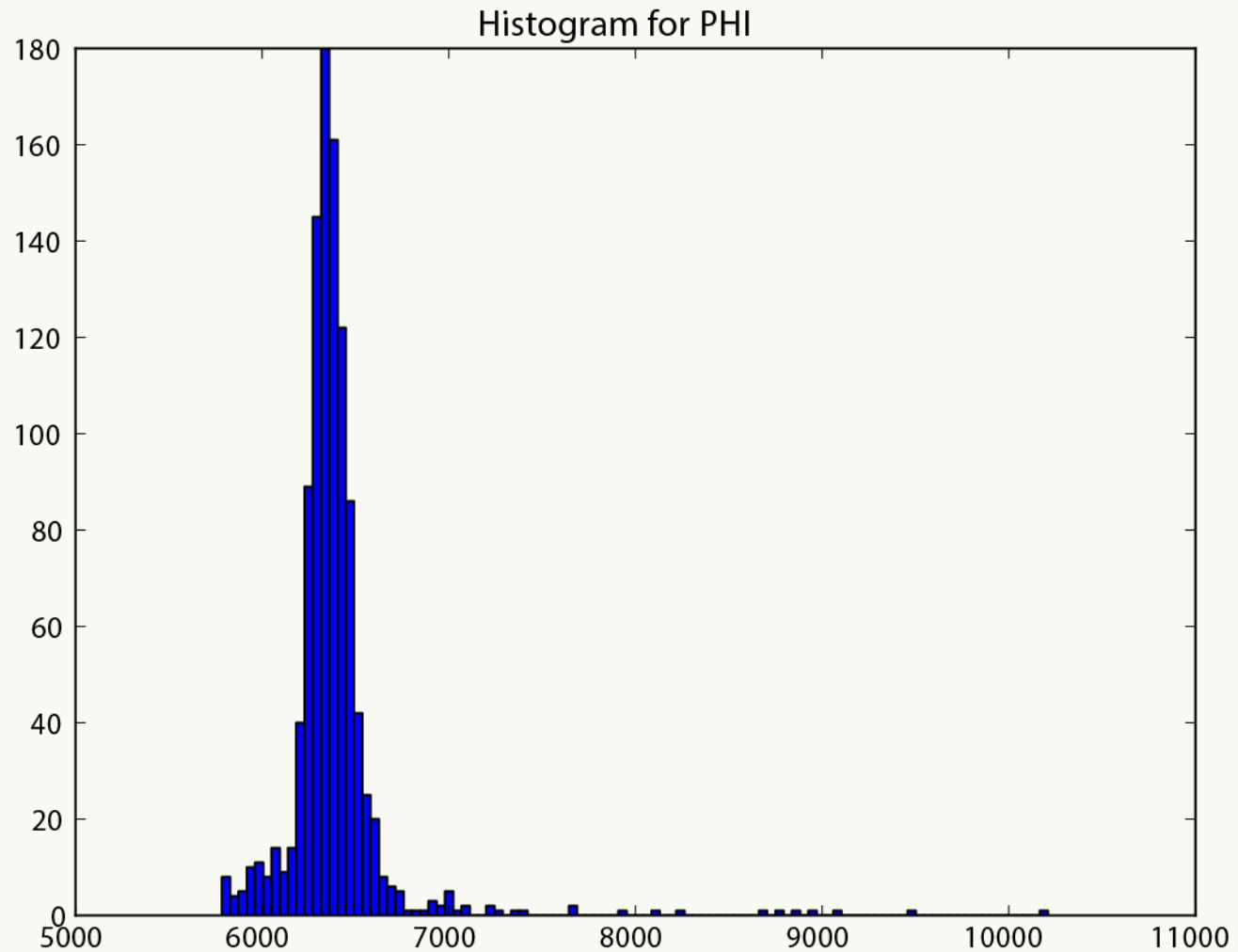
Captured by the well? Yes = assign 1 to the particle starting location for this realization

Sum results for a grid of particles and 1000s of realizations; divide by number of realizations



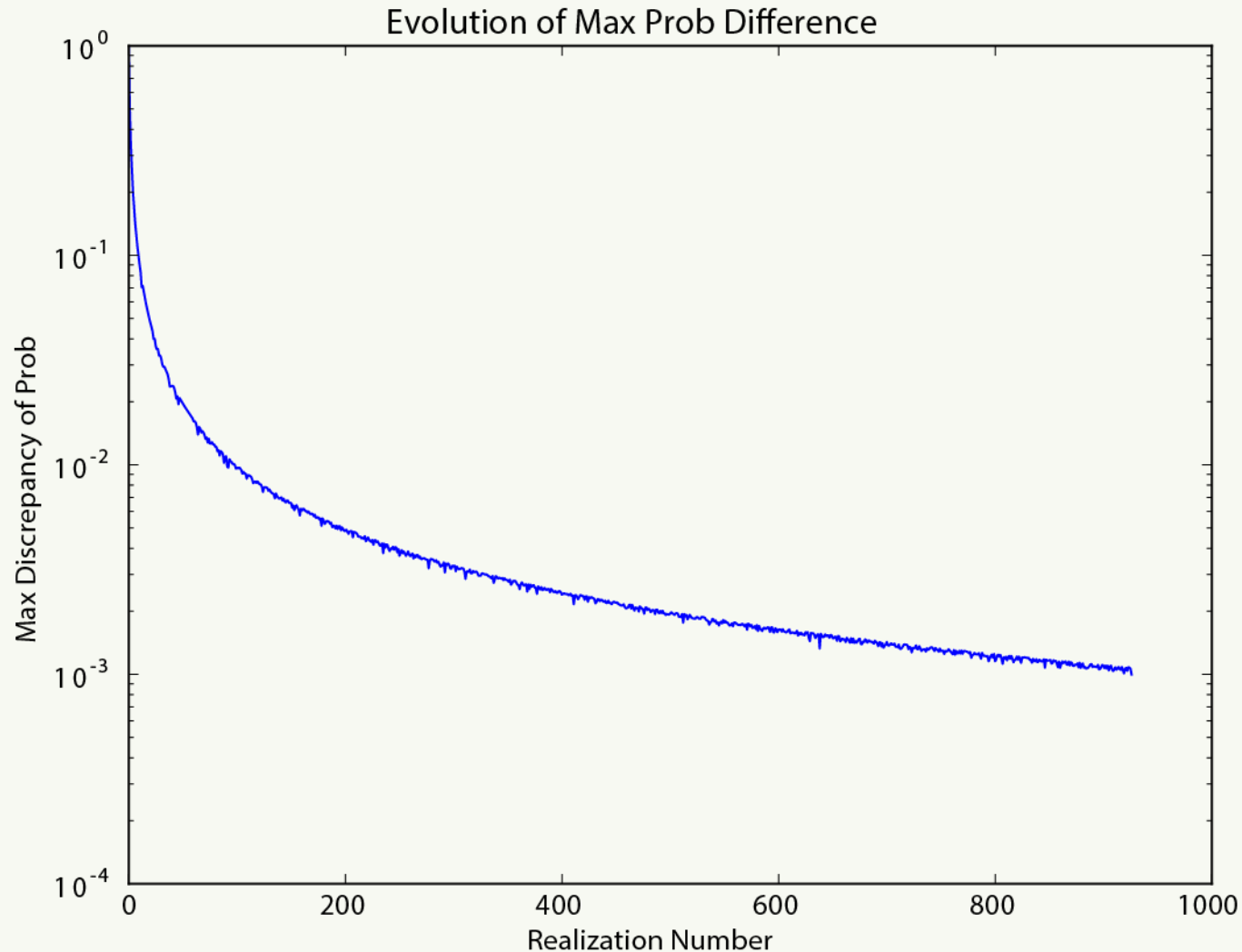
# Evaluate whether realizations kill our calibration

Throw out a few  
realizations that either  
failed or where PHI  
was too high or  
“uncalibrated”



# Declare convergence when probability from one realization to the next is below a threshold

Threshold here was  
0.001 or 0.1%



# Performance

Each forward run takes about 45 minutes

GFLOW has no grid, so the particle tracking work is done by an external Python script – tack on another 15 minutes

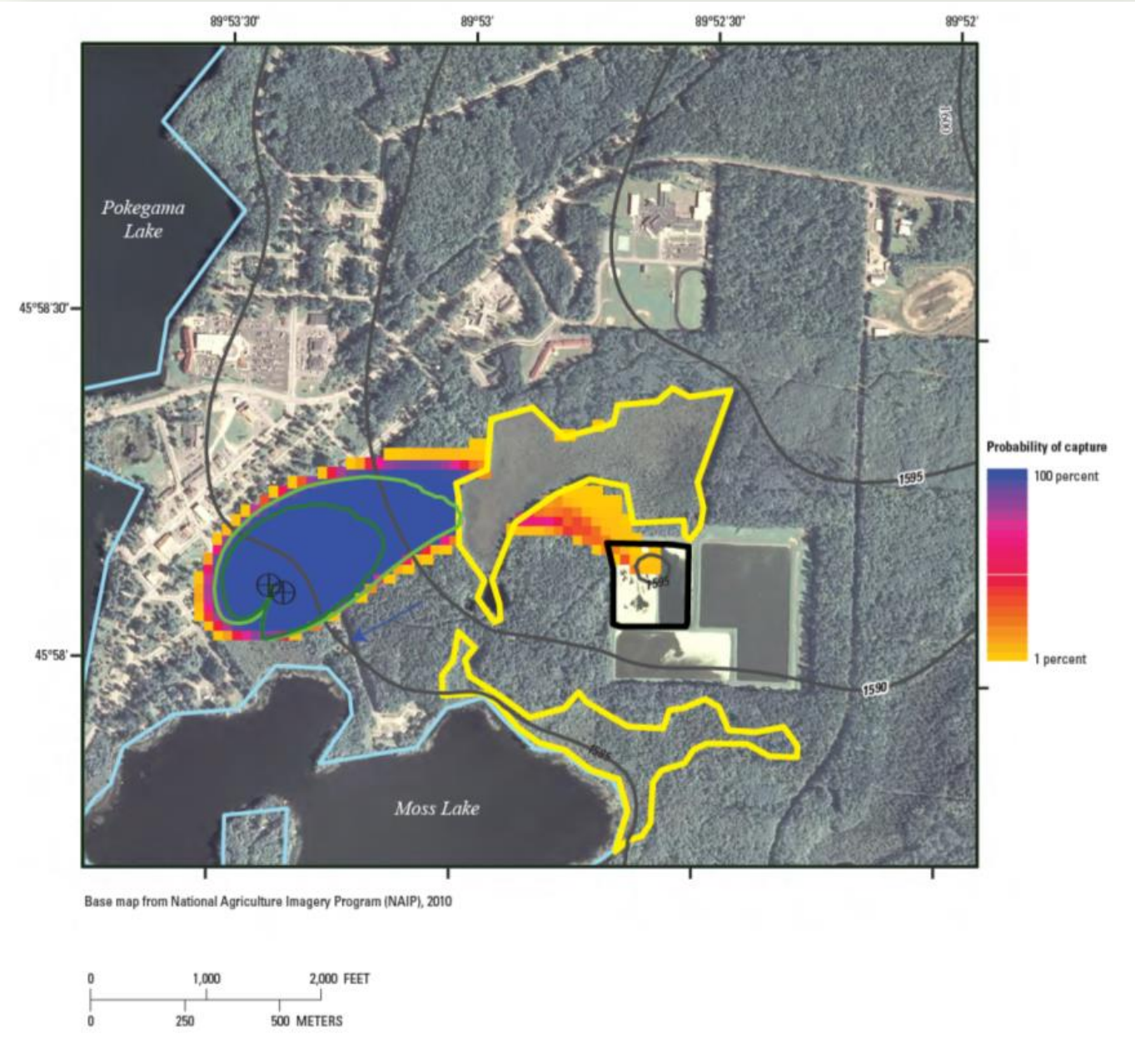
Need roughly 1,200 runs to be sure of convergence

HTCondor makes this all possible!

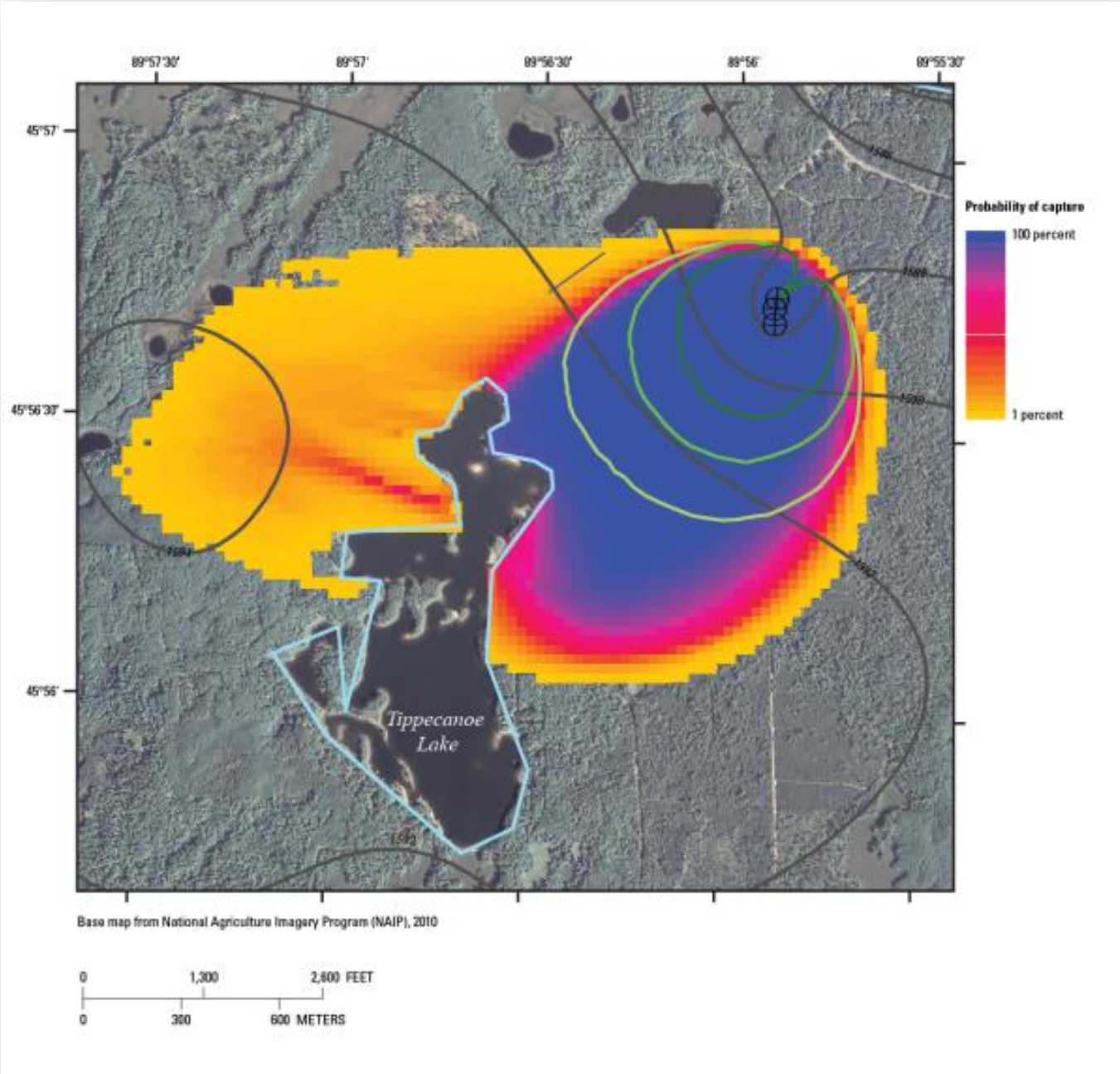
Use Python scripts to assemble files, push to a submit node, launch the HTCondor jobs, and postprocess results.

Maybe this is a good candidate for DAGMan?

# Area contributing recharge considering uncertainty



# Area contributing recharge considering uncertainty





# Wastewater Source Tracking

(where did the pooky go? pathline projection to grids)

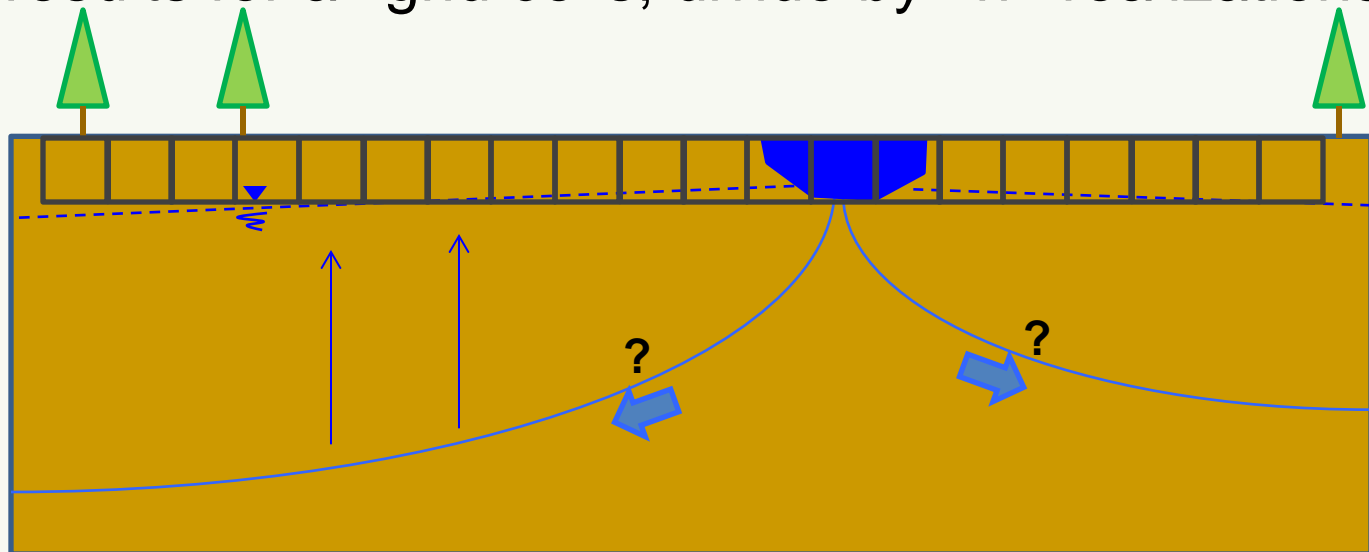
Generate synthetic grid

Forward particle tracking (track the source)

Project pathline onto grid

Does a pathline traverse a specific cell? Yes = assign 1 to the cell for this realization

Sum results for all grid cells; divide by “n” realizations

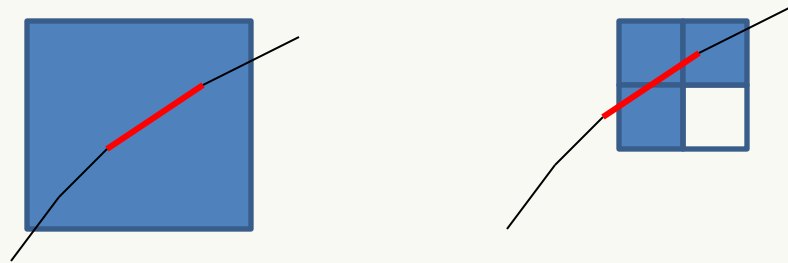


# Projecting a Pathline to a Grid

Process sequentially along each pathline

Identify nearby grid nodes

Make a map of which cells are traversed by a pathline



# Wastewater Source Tracking

Grid spacing = particle step size (5m)

1000 particles  
(forward tracked)

Probability of at least  
one particle intersecting  
a cell – “probability of  
plume extent”

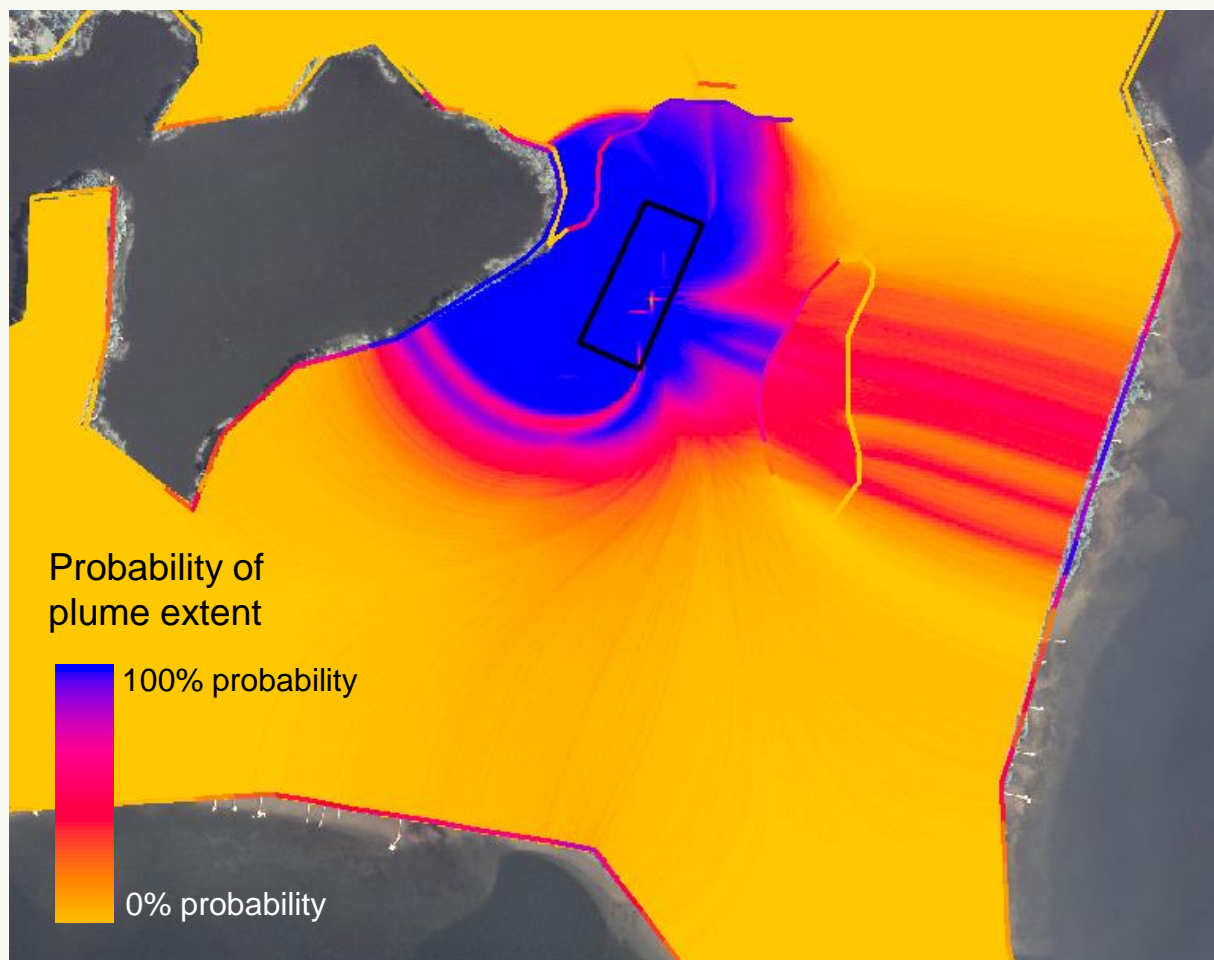
Considerations:

Number of particles

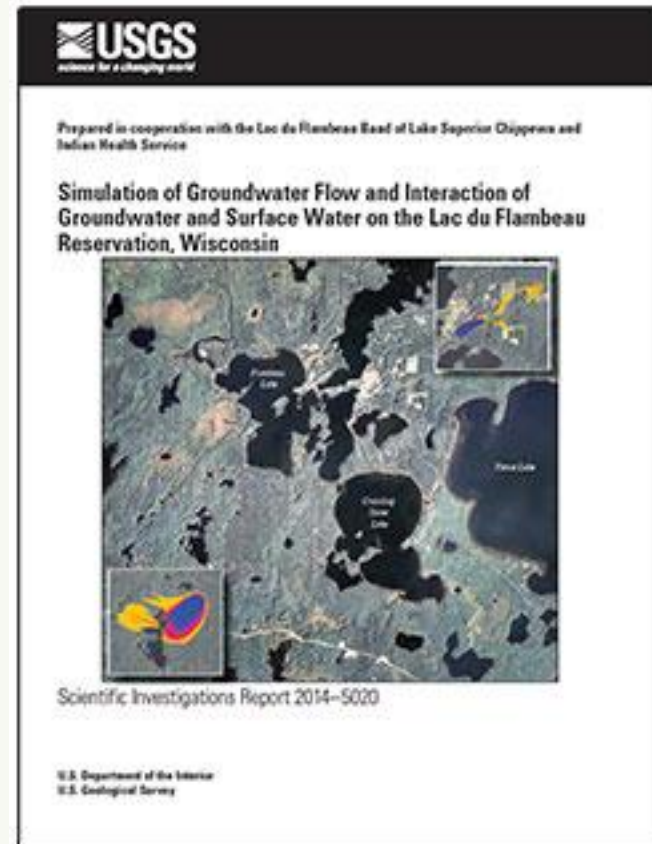
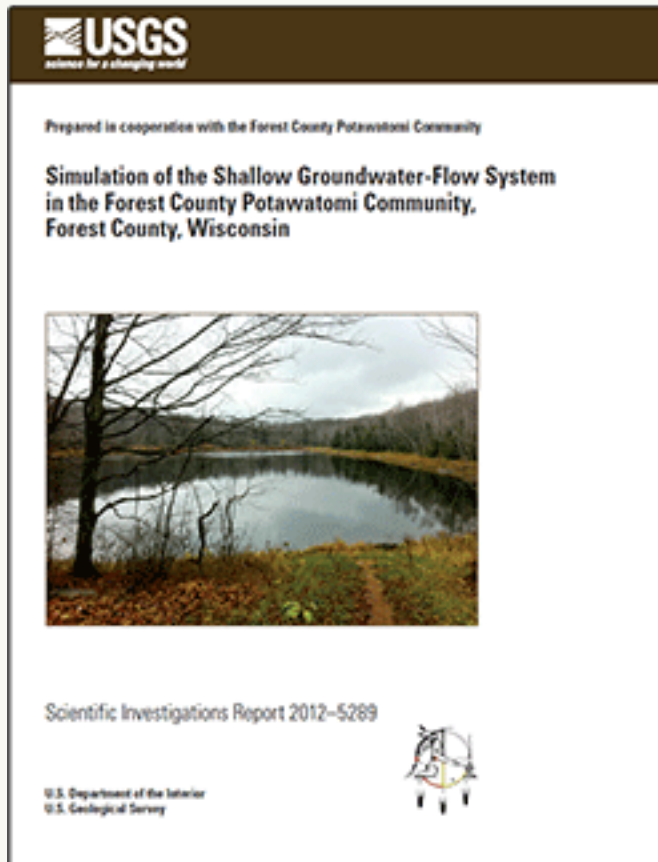
Particle step size

Grid spacing

Weak sinks



# Any Questions?



For more information:

<http://pubs.usgs.gov/sir/2012/5289/> <http://pubs.usgs.gov/sir/2014/5020/>

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