

CENTER FOR APPLIED GEOSCIENCE

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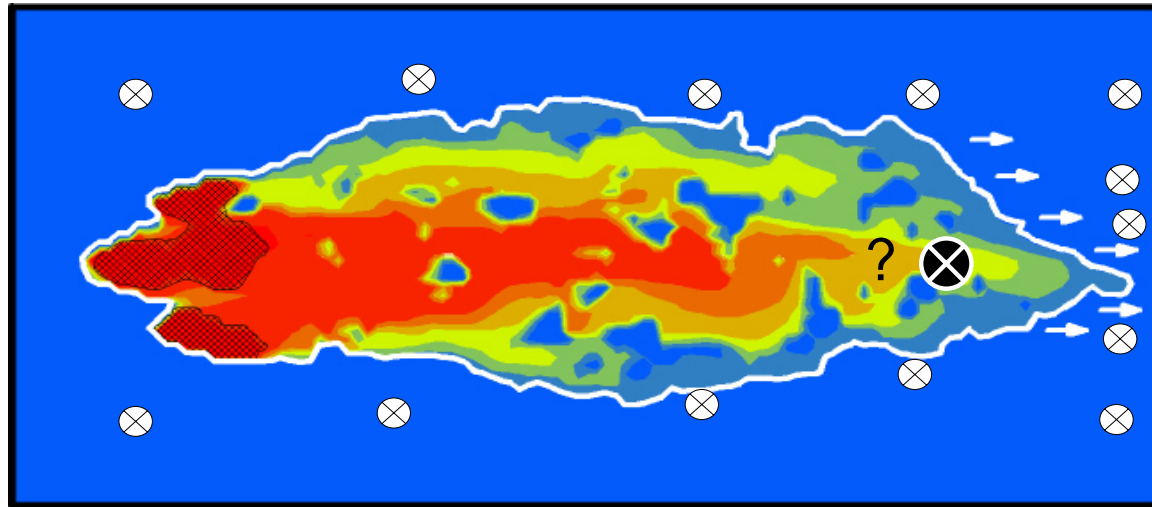
Implementation and performance of selected evolutionary algorithms ... for the tuning of groundwater remediation systems



Support for this project is provided by the German Department of Education, Science, Research and Technology (BMBF), Contract No. 02WR0195 .



- **Advective control**: heuristic optimization of particle tracking problems
- **Binary** and **real coded** decision parameters: SGA vs. **DES**
- Hydraulic/economic optimization of **Pump-and-Treat** and **Funnel-and-Gate** systems



“Concentration control”

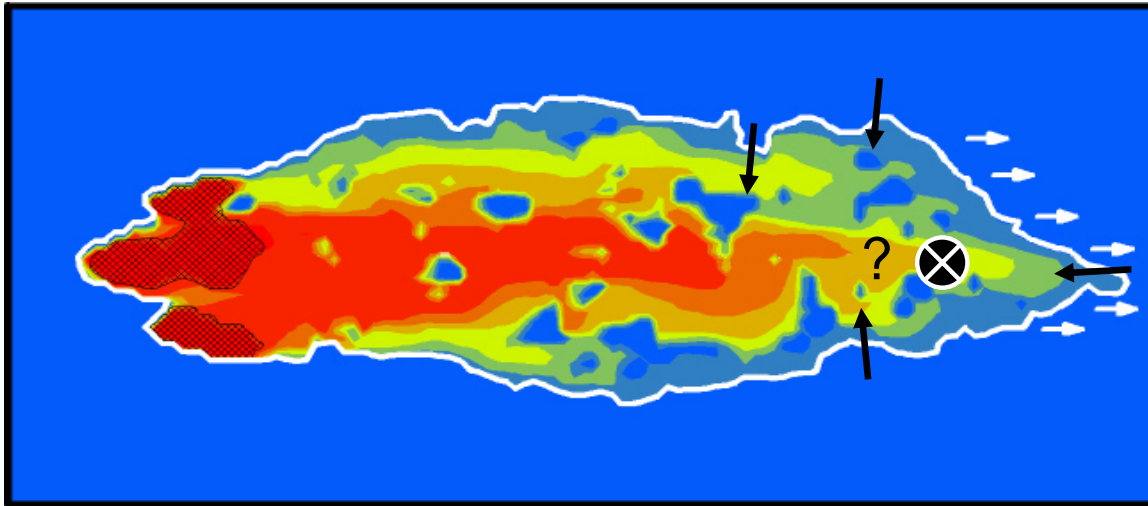
- (Accurate) Transport model of contaminant site required.
- Evaluation of system performance by logging contaminant concentrations at compliance areas or at hypothetical monitoring wells
- Adaptation of contaminant pumping wells: minimum pumping rate, minimum operation time.



Capture zone evaluation criteria for pump-and-treat

(cp. Mulligan & Ahlfeld 1999)

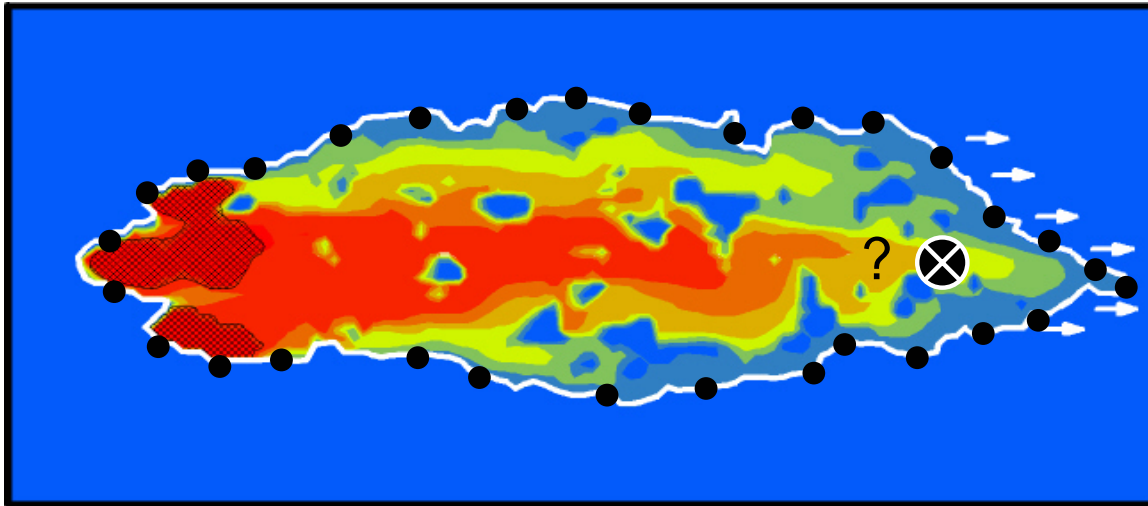
“Hydraulic control”



- Hydraulic and/or transport model of contaminant site required.
- Evaluation of system performance by ensuring gradient/velocity constraints at selected control points (**pre-definition** of optimal capture zone, downgradient control hardly possible).
- Adaptation of contaminant pumping wells: minimum pumping rate, **minimum operation time**.
- Relatively fast method.



“Advective control”



- Hydraulic model of contaminant site required.
- Evaluation of system performance by tracking particles at selected control points.
- Adaptation of contaminant pumping wells: minimum pumping rate.
- Relatively fast method.

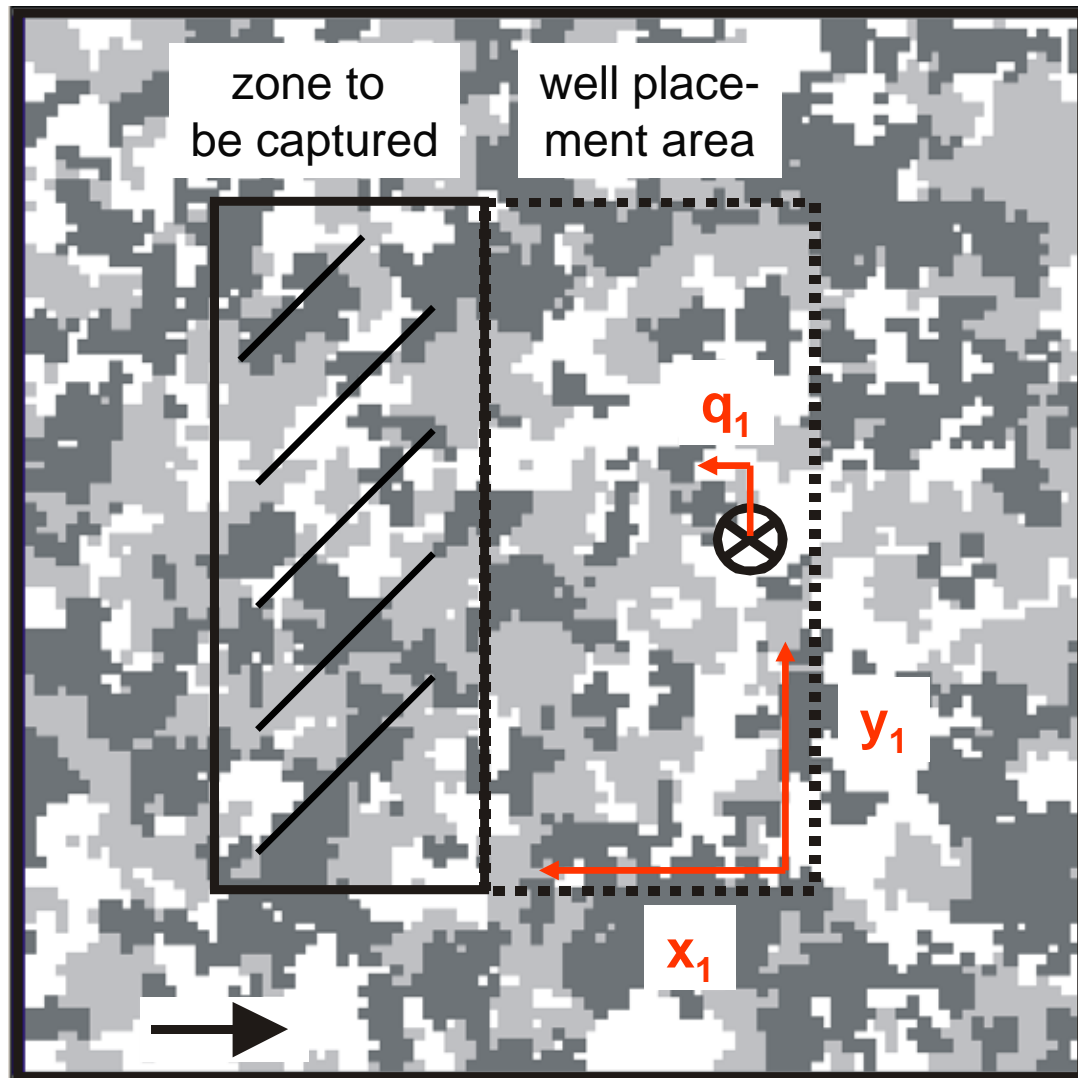


- Only direct method, optimal well configuration not predefined.
- Source of uncertainty: hydraulic description, delineation of contaminated zone.
- Can also be used for optimization of wells with time-related capture zones or recharge areas

Problem: natural heterogeneous aquifers lead to highly nonlinear relationship between well positions, pumping rates and the 'degree' of capture.



Example problem: heterogeneous aquifer, 1 well



hydr. conductivity distribution



$$K_f = 10E-2 \text{ m/s}$$



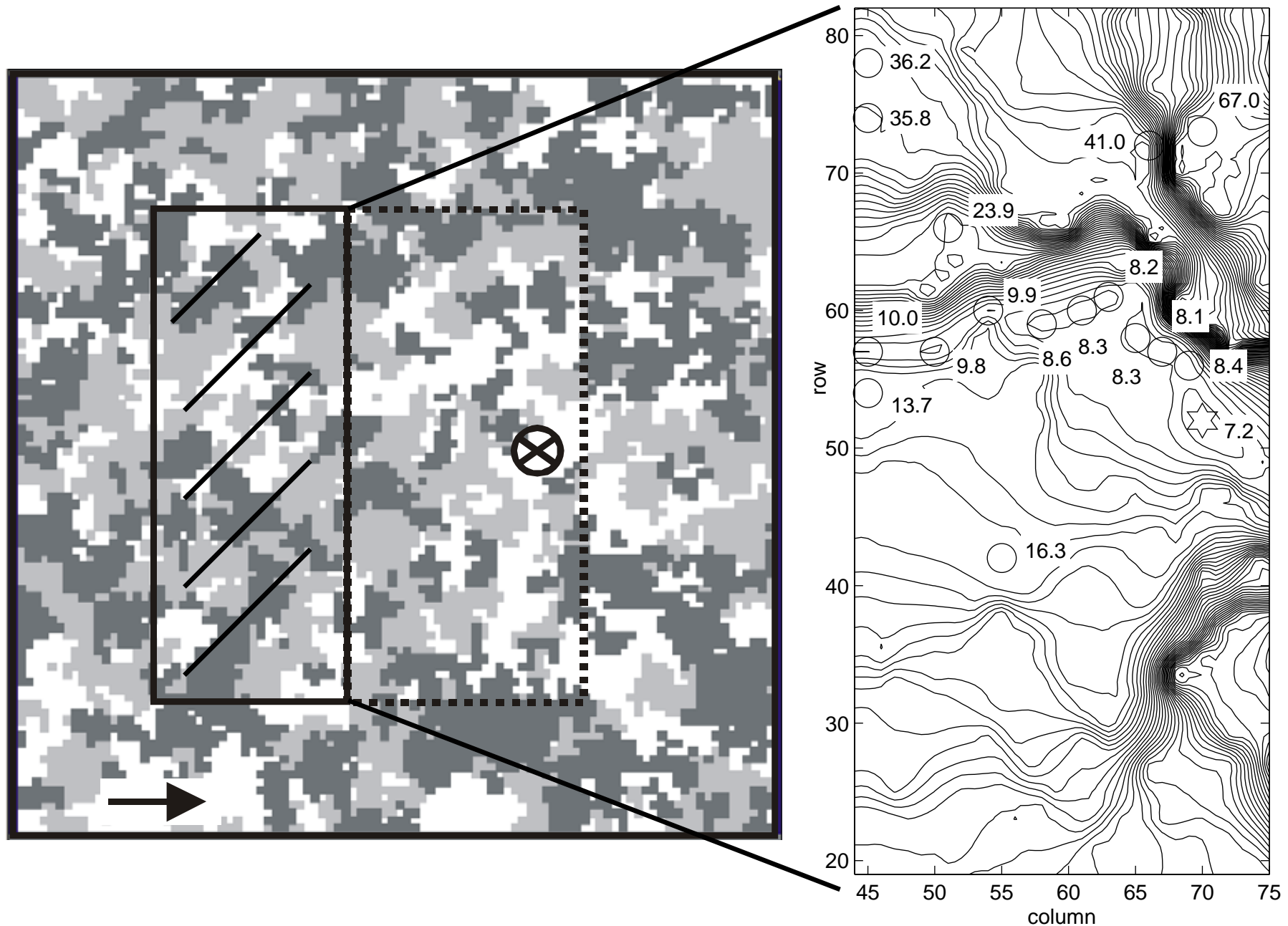
$$K_f = 10E-3 \text{ m/s}$$



$$K_f = 10E-5 \text{ m/s}$$



Minimum pumping rate distribution (1 well)





Explicit formulation of multi-criterion objective function

$$J = \min f(q_{1..N}, x_{1..N}, y_{1..N})$$
$$f = f_q(q_{1..N}) \cdot p(q_{1..N}, x_{1..N}, y_{1..N})$$

‘cost term’

‘penalty term’

*calculation by
(non)linear/economic
model*

*containment ?
computation by numerical model*

$$p = v(q_{1..N}, x_{1..N}, y_{1..N})$$

v : relative number of captured particles

$$p = 10^{100v}$$

e.g., exponential penalty term



- ! **Objective function is not differentiable**
- ! **Objective function is supposed to have various local optima**
- ! **Flexible applicability to high dimensional problems (e.g. 4 wells: $D = 12$)**

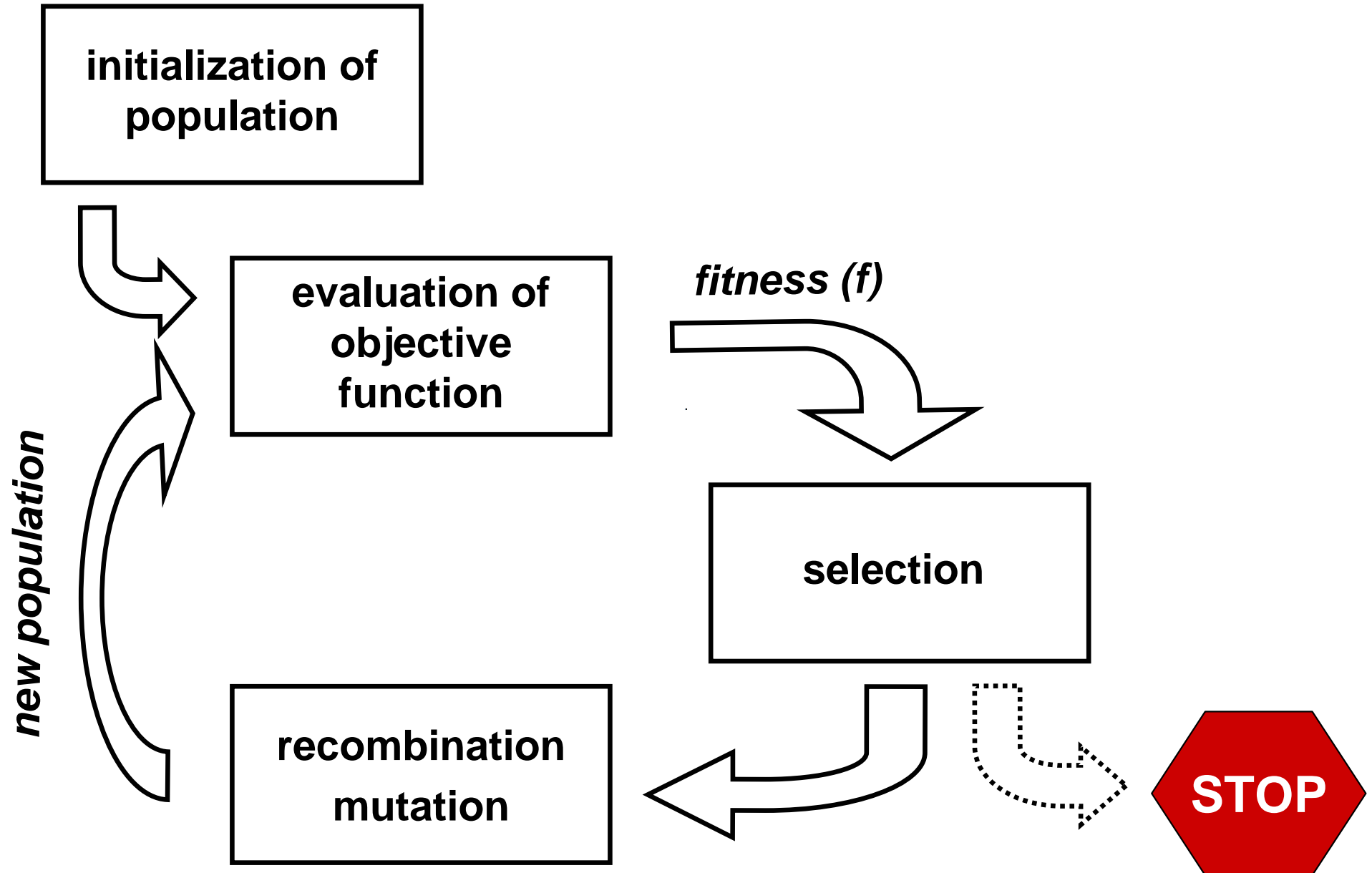
Heuristic techniques:

- solve problems with rules-of-thumb or common sense rules
- typically transformed from another science field

Evolutionary algorithms
analogies to evolutionary
processes

Simulated annealing
simulation of heating and
cooling of metal

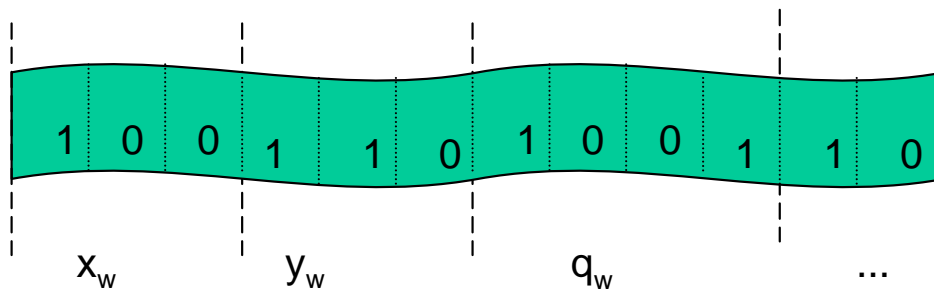
Tabu search
Inspired by random
elements of human behavior





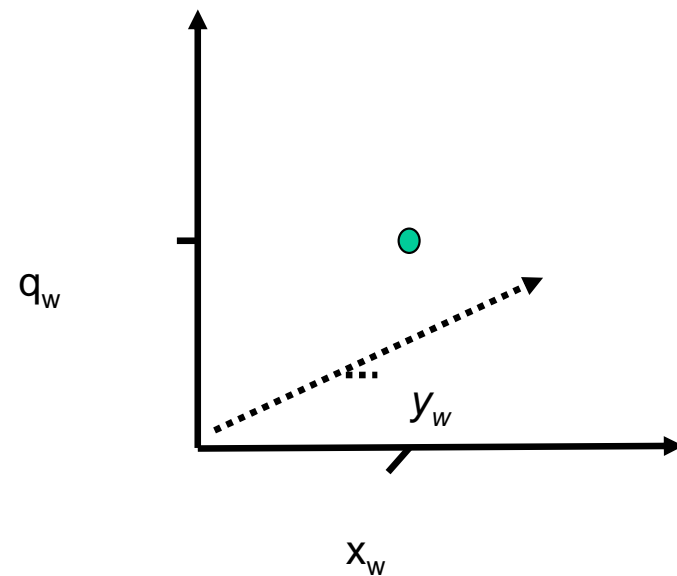
GENETIC ALGORITHM (SGA)

binary



EVOLUTION STRATEGY (DES)

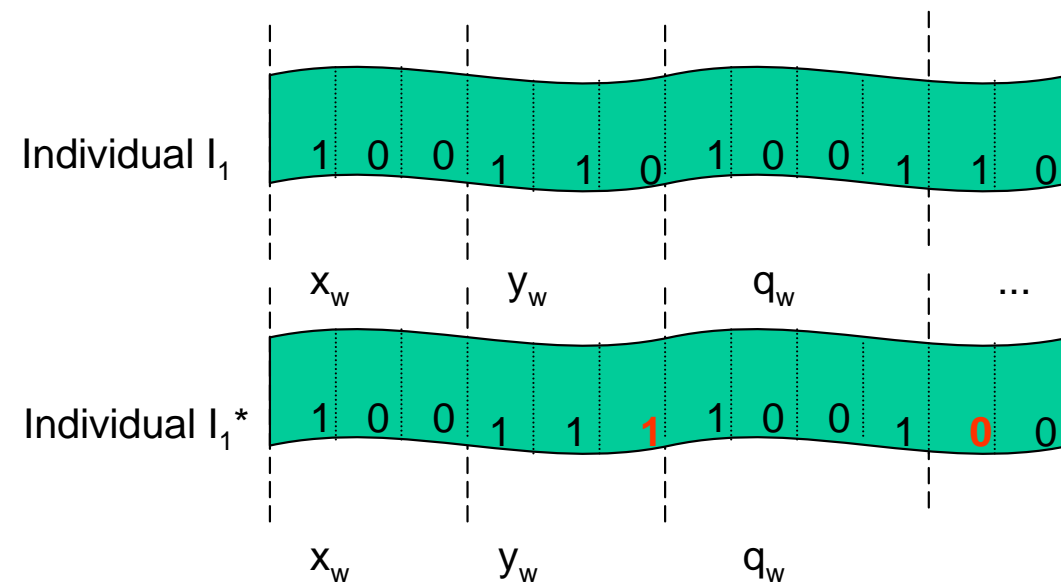
real valued





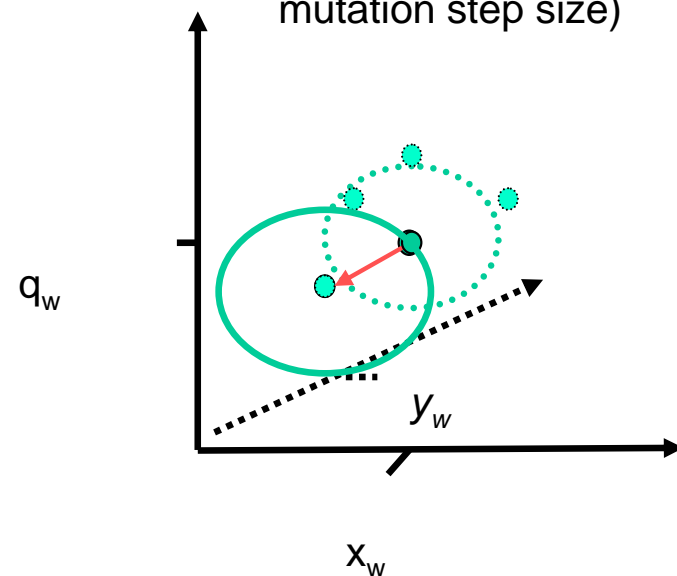
GENETIC ALGORITHM (SGA)

flipping bits



EVOLUTION STRATEGY (DES)

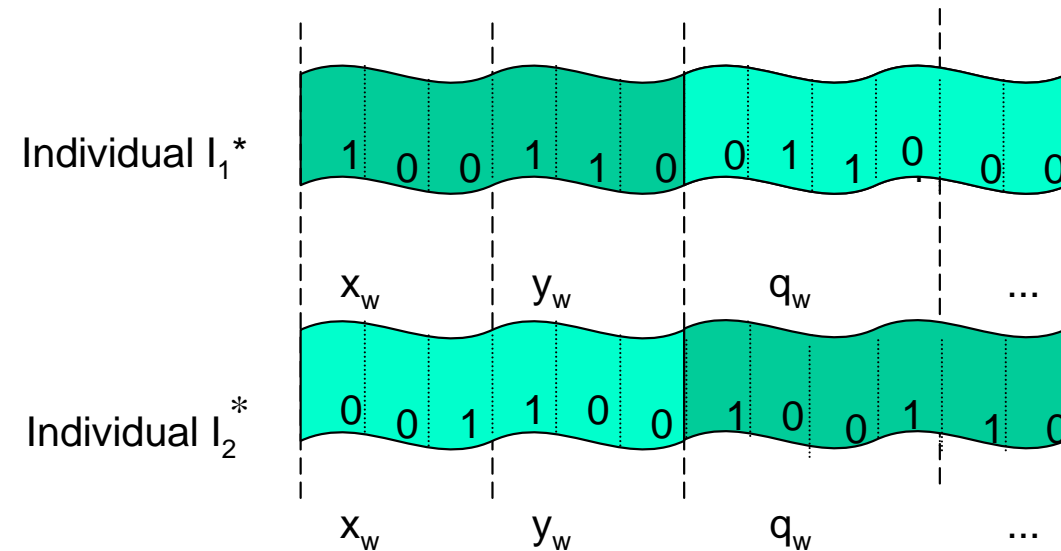
Mutation follows n-dimensional probability density function with adapted standard derivation (= mutation step size)





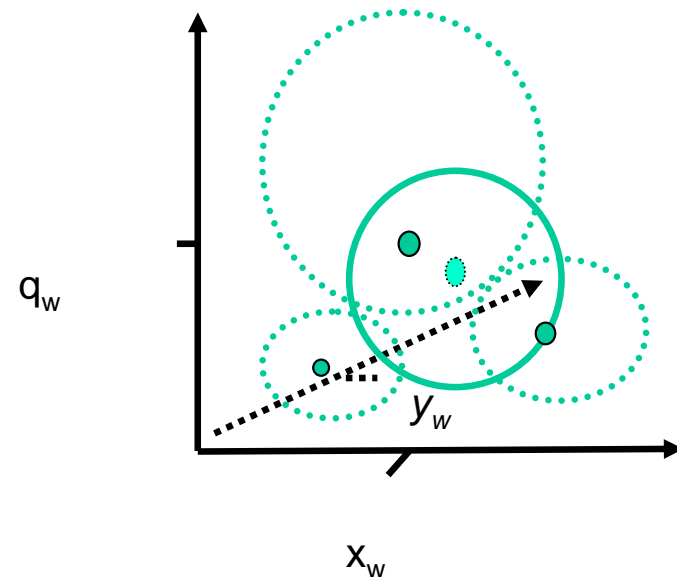
GENETIC ALGORITHM (SGA)

binary



EVOLUTION STRATEGY (DES)

real valued, multi-parent

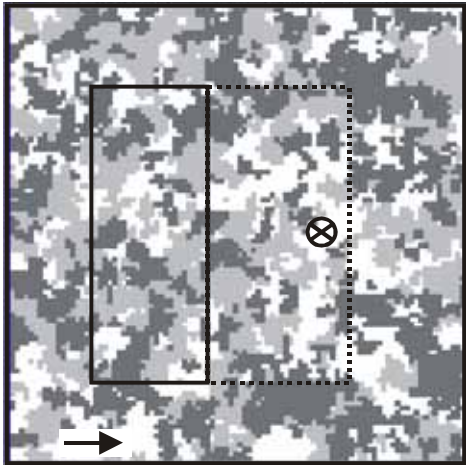




- **Features:** Self-adaptive mutative step size control; weighted/intermediate multi-parent recombination, `cumulation`; `CMA`
- **Applicability:** Efficient for non-linear, badly scaled, high dimensional problems
- **Questions:** Suitable for advective control problem with both discrete & real valued object parameters? Maximum number of iterations? Stop criterion?



Application to example site DES vs. SGA



evolutionary search is stochastic:

50 randomly initialized equal **optimization runs**
with predefined number of model calls

setting of decision variable limits:

pumping rate maximum is set case-specific or
automatically (**boundary update**)

objective values:

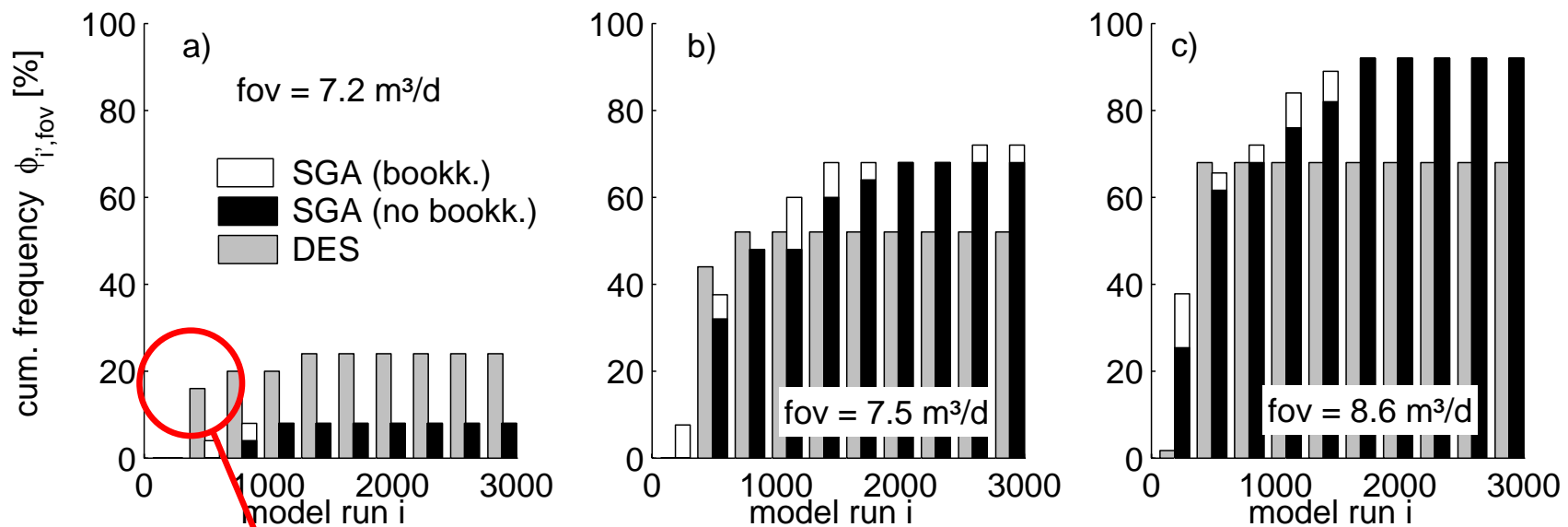
pre-selected optimal values of
pumping rate are set as **objective**
values fov

some configurations are called more than once

for the SGA **bookkeeping** is conducted

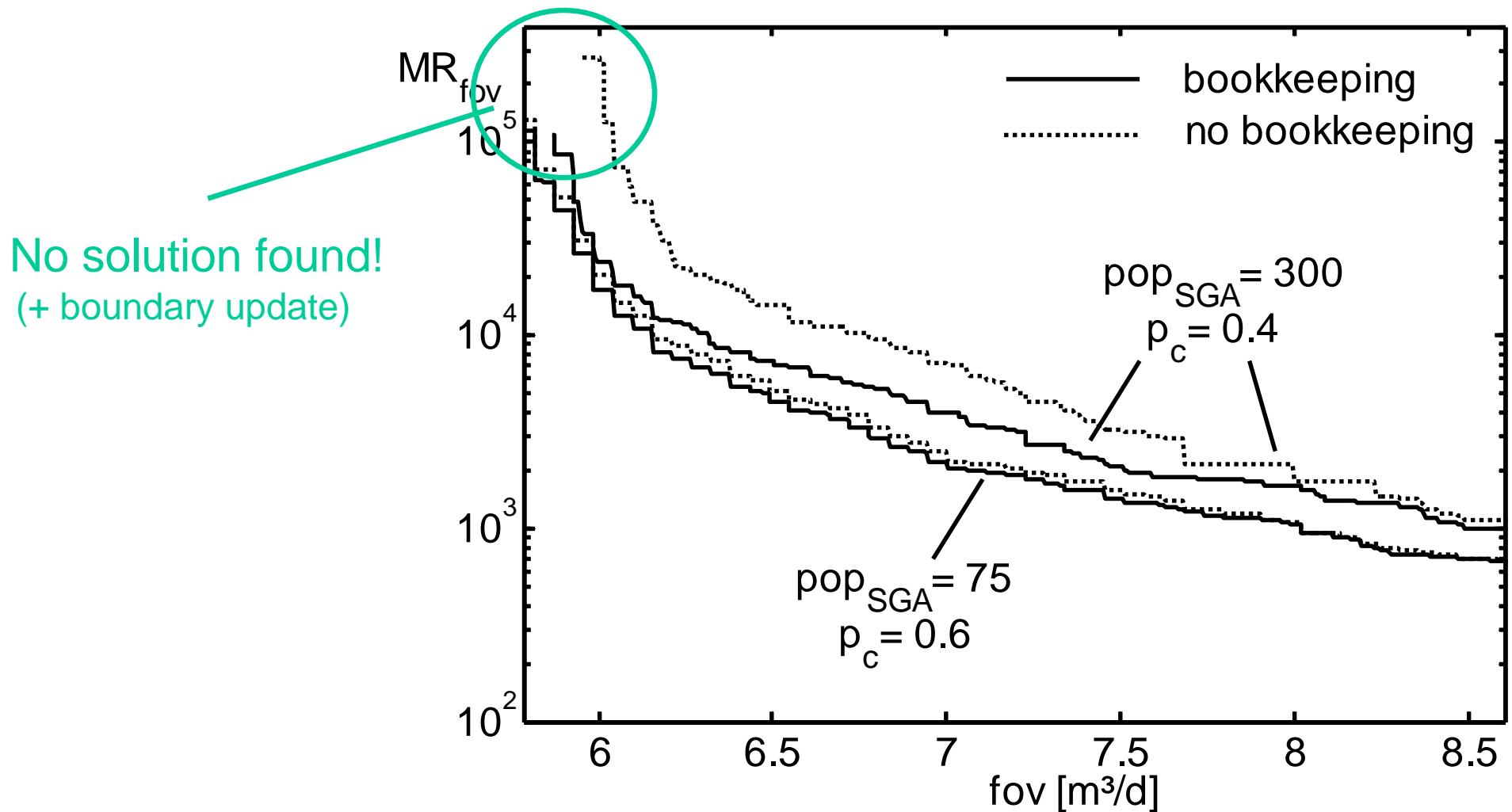


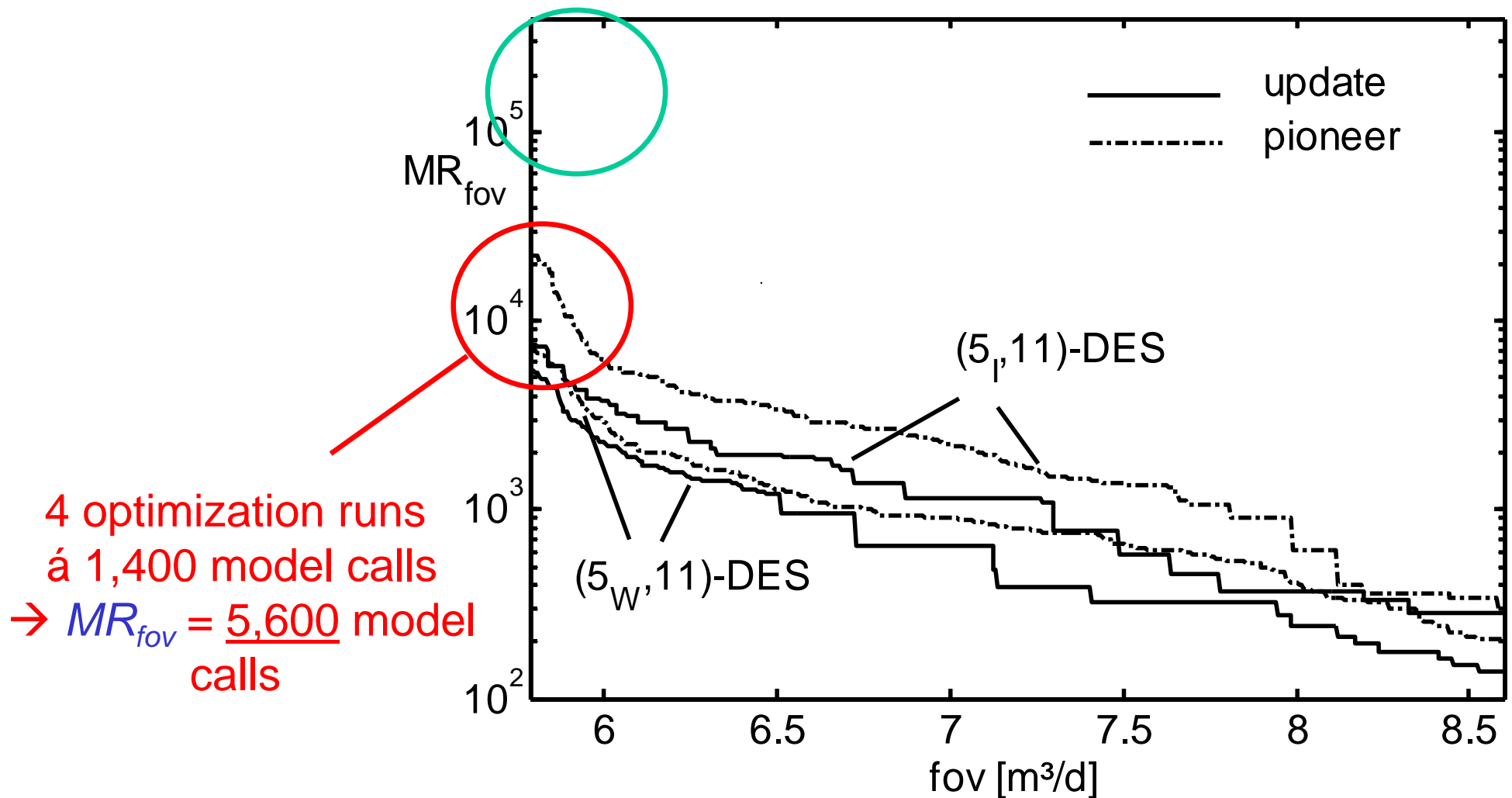
“ MR_{fov} ” : calculated average model runs required to find a solution for a selected objective value fov .



5 optimization runs
à 350 model calls

→ $MR_{fov} = 1750$ model calls

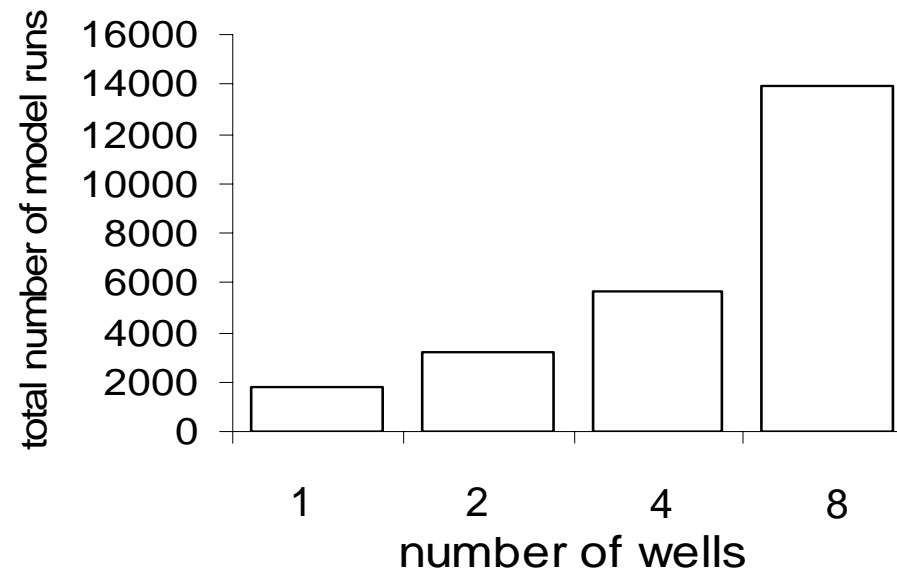
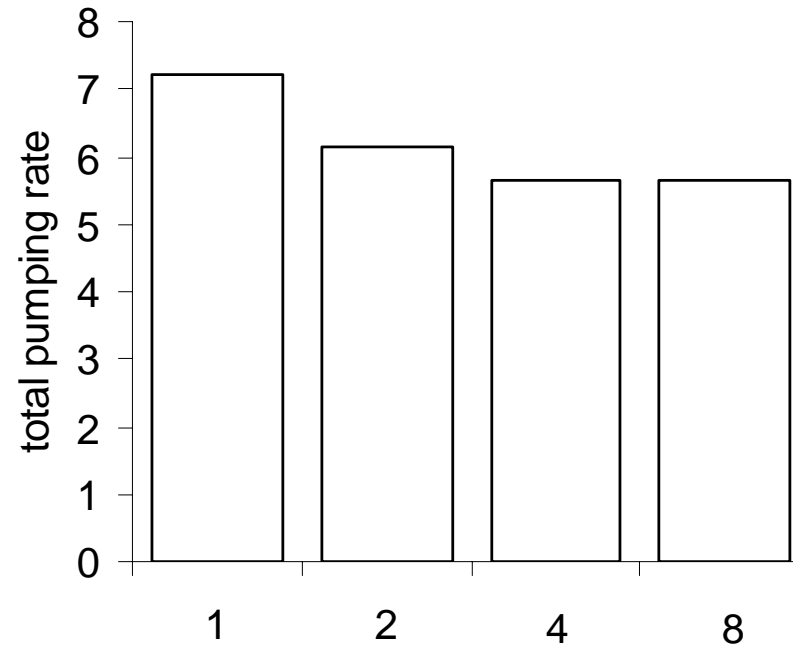






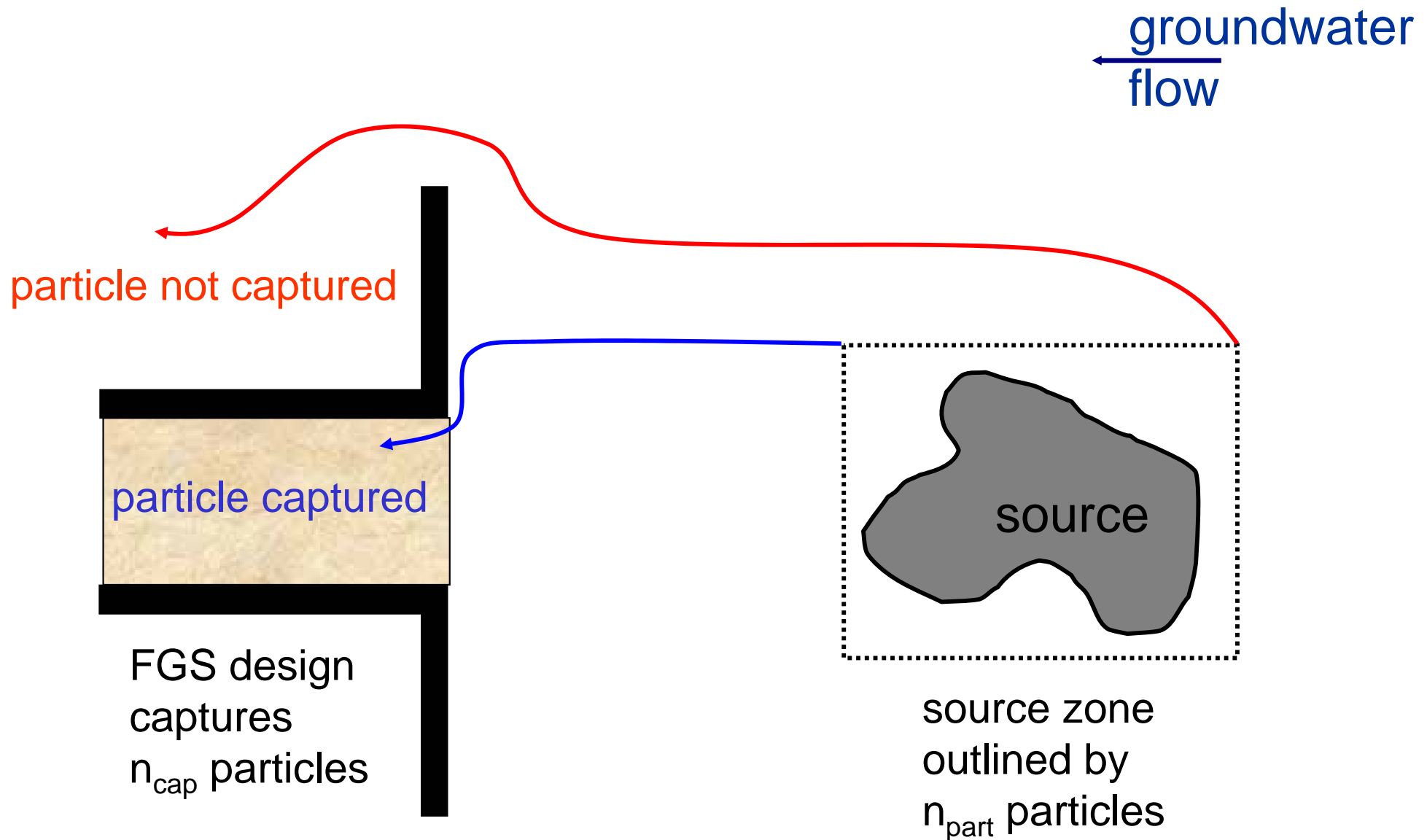
→ Increasing the number of wells reduces the minimum total pumping rate

→ Increasing the complexity leads to a ~ linear increase of required model runs





Funnel-and-gate: Plume capture by advective control





$$\text{Cost} = (2 \times \sum l_g + \sum w_f) \times C_f + 2 \times \sum w_g \times C_g + \sum (w_g \times l_g) \times C_{\text{reac}}$$

where

C_f : unit funnel costs

C_g : unit gate costs

C_{reac} : unit reactive material cost

plume capture constraint honoured by exponential penalty term:

$$\text{fit} = \begin{cases} \text{Cost} \times a^{((b \times ((n_{\text{part}}/n_{\text{cap}}) - 1))^c), & \text{if } n_{\text{cap}} \neq 0 \\ \text{inf}, & \text{otherwise} \end{cases}$$

where

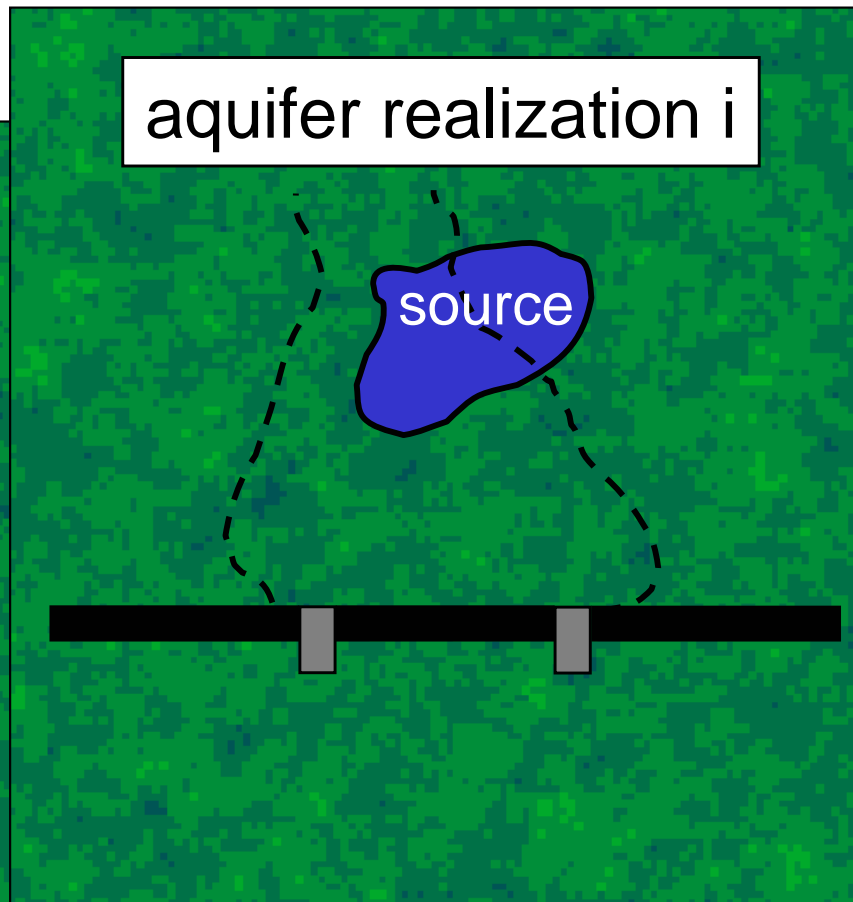
$a, b, c > 0$: to be selected according to the range of the unit costs

n_{part} : number of tracked particles

n_{cap} : number of particles captured



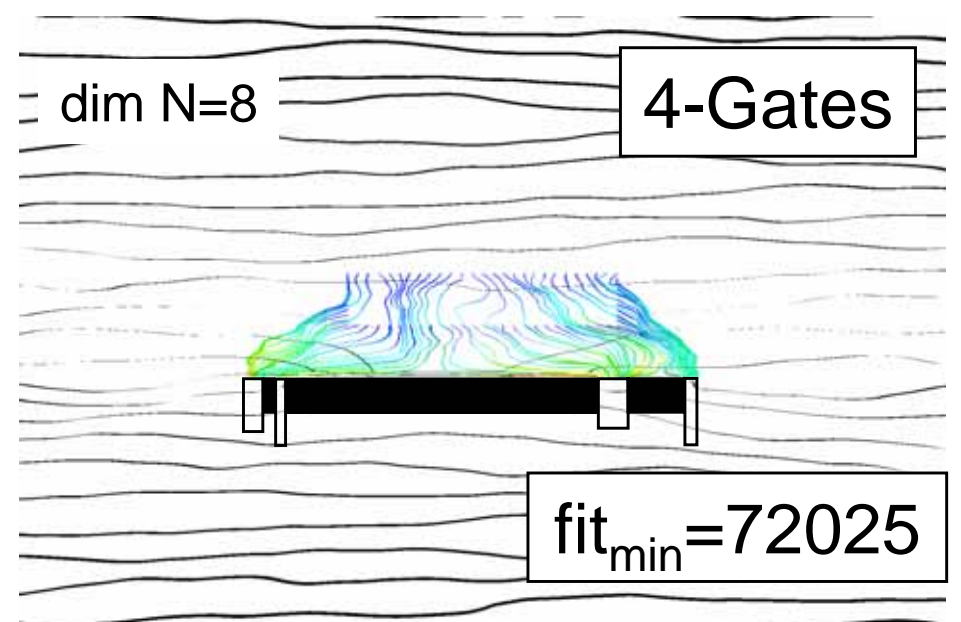
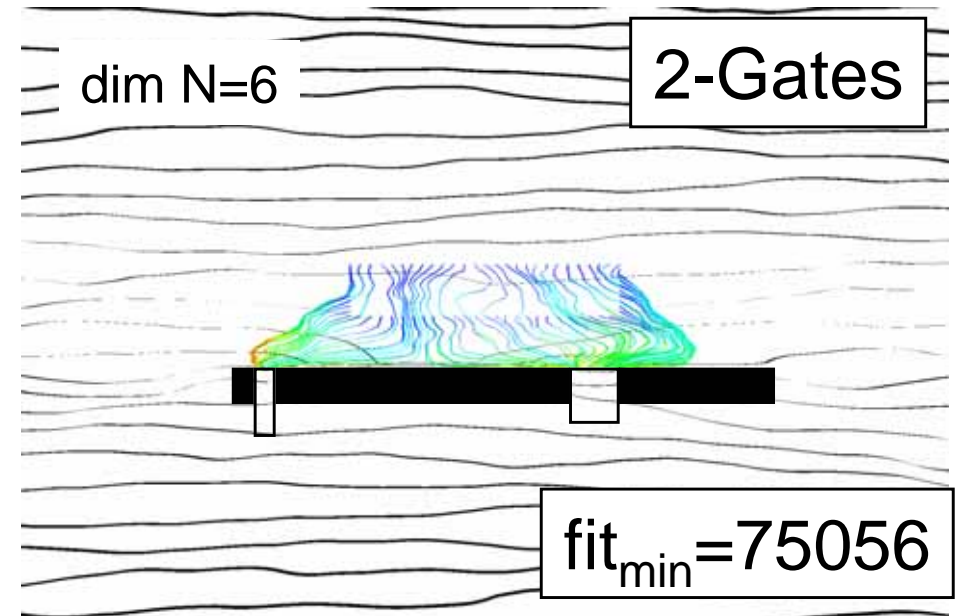
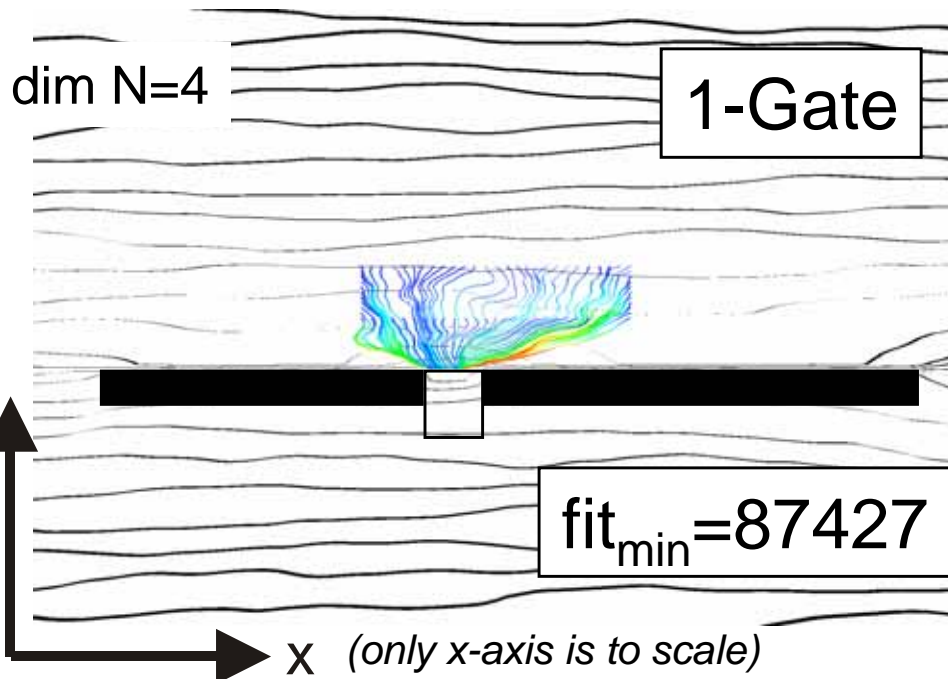
Additional constraint: One design fits all?



- natural aquifer heterogeneity creates uncertain description of flow field
- multiple equiprobable aquifer realizations as part of the objective function
- $\text{fit} = \max (\text{fit}_i, i = 1, \dots, 5)$



DES results:





Advantages of DES

- More robust & efficient, less stochastic & less model runs than SGA
- More suitable for in case of real valued object parameters, applicability in case of discrete object parameters
- Relevance of object parameter limits reduced by self adaptation

Lessons learnt + Outlook

- Number of model runs scales ~linear with problem dimension for DES
- Efficiency of boundary update: additional use of available information
- Limitations for DES in case of „hard“ discretized problem formulations?
→ Further inspection of applicability for Funnel&Gates Systems
- Missing: Requirement of stop criterion for practical use