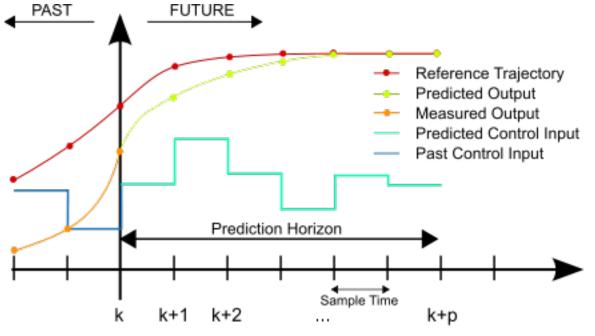


Nonlinear optimization of hydraulic systems

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Liège, October 13, 2017

Model predictive control



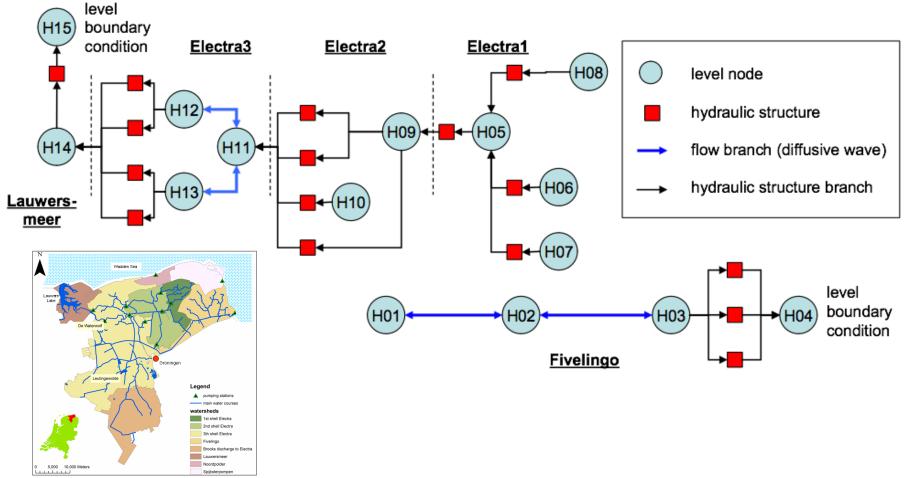
Source: Wikipedia. CC BY-SA 3.0.

- Predict system state based on model
- Compute control inputs that maximize performance over prediction horizon
- Implement first computed control input
- Repeat procedure at next time step

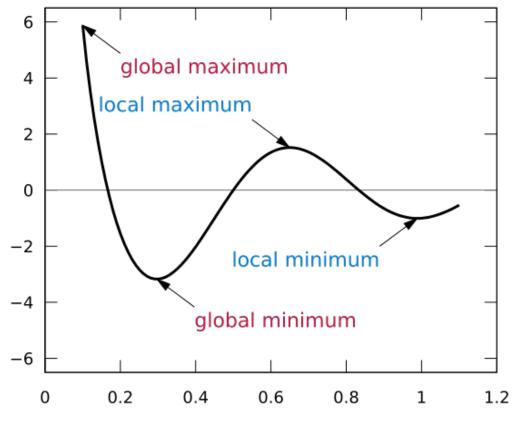
A good prediction model satisfies several requirements:

- *Accurate*: It captures the relevant physical processes with <u>sufficient</u> accuracy.
- Simple: It focuses on the <u>essential</u> processes. Details are left out. Optimizing for details is a bad idea, considering the inaccuracies inherent in any inflow forecast. Less = more.
- *Quick*: As it will need to be evaluated many times during optimization, a single run needs to be computationally <u>inexpensive</u>.

Prediction model



Local and global optima



Source: Wikipedia. GFDL 1.2.

Idea: Formulate optimization problems that only admit global minima.

The mathematical term for such formulations is that they are <u>convex</u>.



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Deltares

Unfortunately, we will see that convexity is too restrictive when it comes to modelling hydraulic processes with large discharge variations.

 $Q^2 \propto \Delta H$



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Problem: Both Q and ΔH vary throughout optimization.



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Problem: Both Q and ΔH vary throughout optimization.

- ➡ Nonlinear equality constraint.
- ➡ Nonconvex optimization problem.
- ➡ Finding even any feasible solution hard in general, and when one is found, no certificates on solution quality.

Solution: Homotopy

Idea: Interpolate between linear and nonlinear model.

 $M = (1 - \theta)M_L + \theta M_{NL}$

With continuation parameter $\theta \in [0, 1]$.

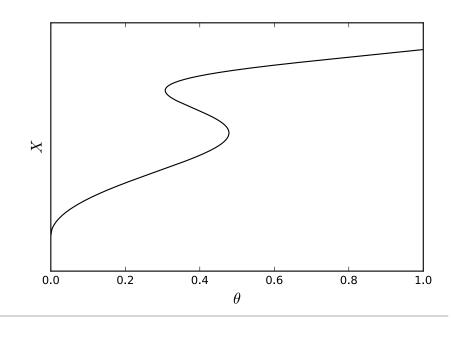




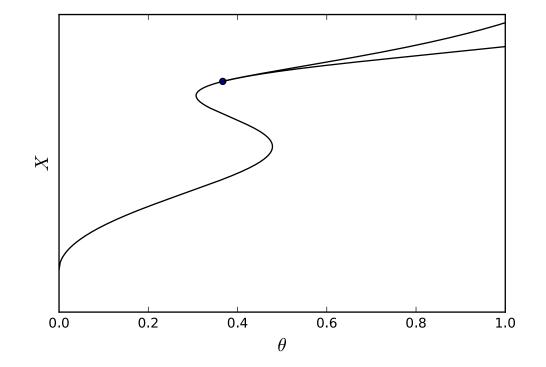
Continuation method

Start solving at $\theta = 0$ continuing step-by-step until $\theta = 1$.

At $\theta = 0$, the linear problem only admits global optima. We find a global optimum, and trace this through to a nonlinear solution at $\theta = 1$: Solution to nonlinear problem can be *traced back* to globally optimal solution of linearized problem.



Continuation method: Bifurcations



Bifurcations occur at so-called singular points.

Nonlinear hydraulic model

$$\frac{\partial Q}{\partial t} + \frac{\partial}{\partial x} \frac{Q^2}{A} + gA \frac{\partial H}{\partial x} + g \frac{Q|Q|}{C^2 RA} = 0$$
$$\frac{\partial A}{\partial t} + \frac{\partial Q}{\partial x} = 0$$

Idea:

- Discretize 1D shallow water (Saint-Venant) equations on staggered grid;
- Drop advection term (valid when $\frac{\partial A}{\partial x}$ and $\frac{\partial Q}{\partial x}$ are small); of minor importance for typical NL water board applications.

Idea: Take nonlinear discretization, and fix terms that vary little compared to other terms.

 $(\neq$ linearization, which does not work well for large variations in flow variable Q.)



Homotopy linear → nonlinear: Main result

Theorem A necessary condition for a singular point is for either of the following alternatives to hold:

- Wetting and drying $(H = H_b)$
- Flow reversal (Q = 0)

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Homotopy linear → nonlinear: Main result

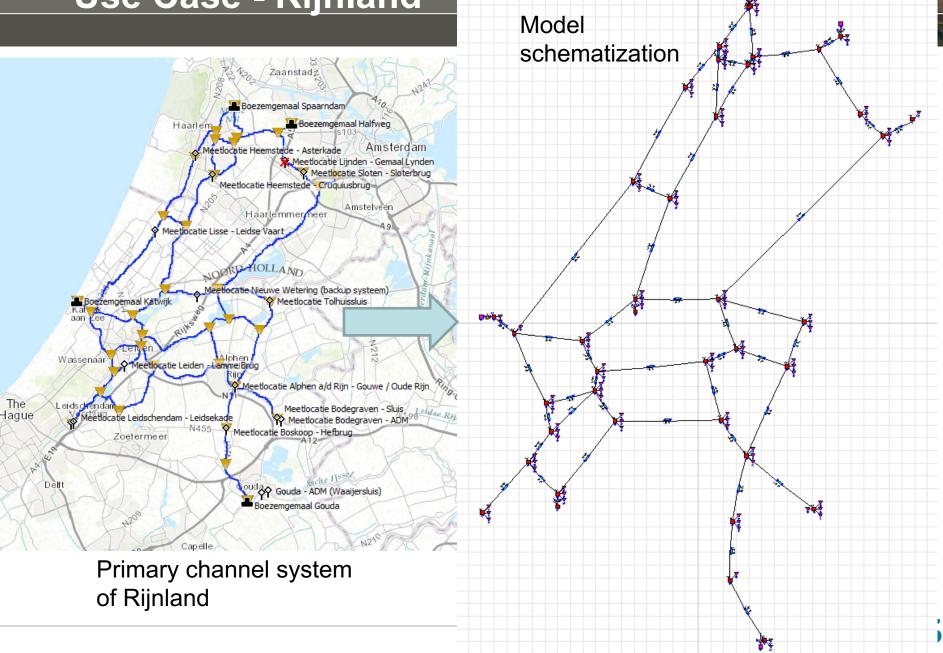
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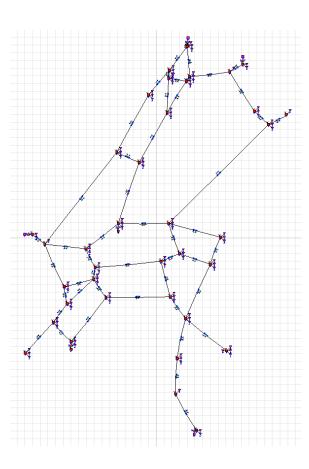
As long as the alternatives are avoided, a continuation method can be applied.

Use Case - Rijnland



Use Case – Primary channel system Rijnland

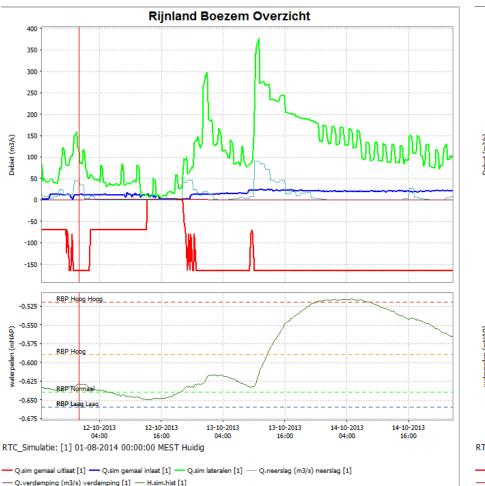
- Multi-objective optimization for safety, salt flushing, and energy use.
- Primary goal: Optimization of water level in primary channel system (the *boezem*)
 - Water level range as well as target level
- Salt and wind dynamics are relevant
 - Wind has an impact on relative water level differences and hence on circulation pattern
 - Multiple sources of salinity, flushing strategy required

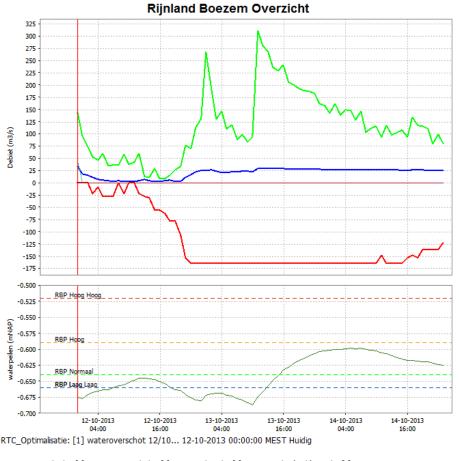


Feedback control vs. Optimization – wet event

Feedback control results

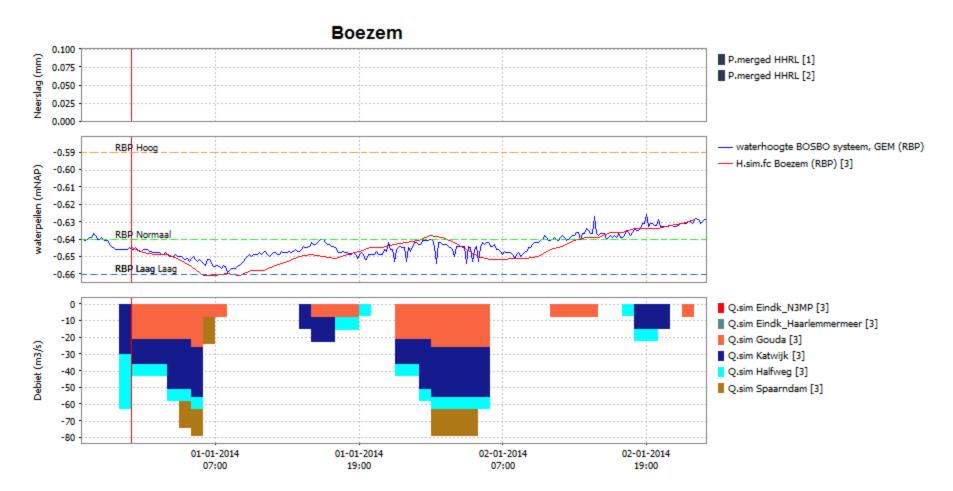
Optimization results





Q.sim gemaal uitlaat [1] — Q.sim gemaal inlaat [1] — Q.sim lateralen [1] — Q.neerslag (m3/s) neerslag [1]
Q.verdamping (m3/s) verdamping [1] — H.sim.fc [1]

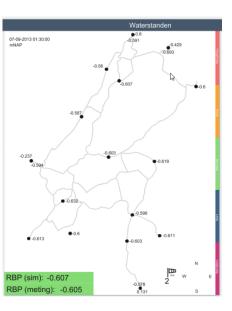
Measured vs. Optimized – similar pump action

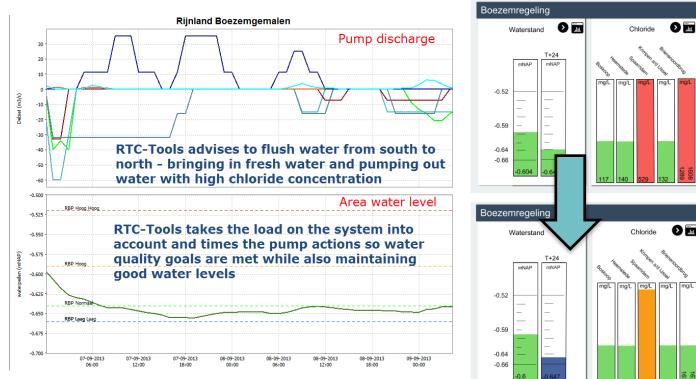


Optimization prefers nighttime pumping (lower price)

Rijnland

A decision support and control system for the water board of Rijnland was brought online earlier this year. The system provides advice on the dispatch of pumping stations, taking into account the operational objectives of flood control, <u>water quality</u>, and cost savings.







Thank you for your attention!

Feel free to contact me any time at

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