



Introduction to RTC-Tools

Jorn Baayen

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- What is RTC-Tools?
- Applications
- Technology



RTC-Tools: Scope

RTC-Tools is an open source toolbox for control and optimization of environmental systems.

Delft-FEWS is an open data handling platform, used for the aggregation of (real-time) environmental data flows.

Together, they provide a platform for the development of decision support systems.



RTC-Tools is an open source toolbox for control and optimization of environmental systems.

- Interdisciplinary, object-oriented modeling using Modelica
- Mathematical framework designed for stable operation in environments that require consistent results
- Optimization under predictive uncertainty
- Multi-objective optimization
- Integration with Delft-FEWS
- Python scripting





Netherlands: Water board of 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.



Brazil: Decision support for Tres Marias dam (CEMIG)

A decision support system for the Tres Marias dam. The system helps to reduce flooding in Pirapora using stochastic optimization techniques.



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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Other packages step around the issue by using linear models.

RTC-Tools goes a step further by finding formulations that accommodate fundamental nonlinearities in the system.

Nonlinearity #1: Storage geometry

Storage volume is an <u>increasing</u>, but generally <u>nonlinear</u>, function of water level.

So this function cannot be included in the model.



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- However, accounting of volumes is linear:

$$\dot{V} = Q_{in} - Q_{out}$$

- Solution: Preprocess water level goals to volume goals.

Nonlinearity #2: Hydropower generation

Instantaneous power from a hydroelectric turbine: $P = \eta \rho g Q H$.



- Q and H are optimization variables.
- P nonlinear, even <u>nonconvex</u>, function of Q and H.
- **Solution**: Change of variables results in <u>nonlinear</u> but <u>convex</u> formulations for load balance and generation maximization goals.

Modelica is a language for object-oriented, declarative, equation-based modeling of dynamical systems.



- Open standard
- Independent of application domain
- Widely used in industry

Models can be written using a text editor (with a Python-like syntax), or using a GUI.

N.B. Steady-state init.

Water allocation in Citarum basin, Indonesia

odel citarum
<pre>import SI = Modelica.SIunits;</pre>
input SI.VolumeFlowRate inflow_Saguling_Q;
input SI.VolumeFlowRate inflow_Cirata_Q;
input SI.VolumeFlowRate inflow_Jatiluhur_Q;
input SI.VolumeFlowRate lateralLoss_Saguling_QLat1;
input SI.VolumeFlowRate lateralLoss_Cirata_QLat1;
input SI.VolumeFlowRate lateralLoss_Jatiluhur_QLat1;
input SI.VolumeFlowRate lateralLoss SagulingCirata Qlat1;
input SI.VolumeFlowRate lateralLoss_CirataJatiluhur_Qlat1;
input SI.VolumeFlowRate lateralLoss_JatiluhurDemand_Qlat1;
input SI.VolumeFlowRate terminal_Agriculture_Qin2;
input SI.VolumeFlowRate terminal_River_Qin2;
input SI.VolumeFlowRate lateralLoss_Saguling_Qin2;
input SI.VolumeFlowRate lateralLoss_Cirata_Qin2;
input SI.VolumeFlowRate lateralLoss_Jatiluhur_Qin2;
input SI.VolumeFlowRate lateralLoss_SagulingCirata_Qin2;
input SI.VolumeFlowRate lateralLoss_CirataJatiluhur_Qin2;
input SI.VolumeFlowRate lateralLoss_JatiluhurDemand_Qin2;
Deltares.Flow.SimpleRouting.Branches.LateralLoss lateralLoss_Saguling =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Inflow inflow_Saguling =;
Deltares.Flow.OpenChannel.Storage.Linear linear_Cirata(Area = 30000000, Htail = 103, Hloss = 4) =;
Deltares.Flow.SimpleRouting.Branches.LateralLoss lateralLoss_Cirata =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Inflow inflow_Cirata =;
Deltares.Flow.SimpleRouting.Nodes.Node node_Agriculture(nout = 2) =;
Deltares.Flow.SimpleRouting.Nodes.Node node_Drinking(nout = 2) =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Terminal terminal_Drinking =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Terminal terminal_Agriculture =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Terminal terminal_River =;
Deltares.Flow.SimpleRouting.Nodes.NodeHQPort_nodeHQPort_Saguling(nout = 1) =;
Deltares.Flow.SimpleRouting.Nodes.NodeHQPort nodeHQPort_Jatiluhur(nout = 1, nin = 2) =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Inflow inflow_Jatiluhur =;
Deltares.Flow.SimpleRouting.Branches.LateralLoss lateralLoss_Jatiluhur =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Terminal terminal_Saguling =;
Deltares.Flow.SimpleRouting.Branches.LateralLoss lateralLoss_JatiluhurDemand #;
Deltares.Flow.SimpleRouting.BoundaryConditions.Terminal terminal_Jatiluhur =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Terminal terminal_Cirata =;
Deltares.Flow.SimpleRouting.Branches.LateralLoss lateralLoss_SagulingCirata =;
Deltares.Flow.SimpleRouting.Branches.LateralLoss lateralLoss_CirataJatiluhur =;
Deltares.Flow.OpenChannel.Storage.Linear linear_Jatiluhur(Area = 63000000, Htail = 27, Hloss = 1) =;
Deltares.Flow.SimpleRouting.Nodes.NodeHQPort_nodeHQPort_Cirata(nout = 1, nin = 2) =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Terminal_terminal_CirataJatiluhur =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Terminal terminal_SagulingCirata =;
Deltares.Flow.SimpleRouting.BoundaryConditions.Terminal terminal_JatiluhurDemand =;
Deltares.Flow.OpenChannel.Storage.Linear linear_Saguling(Area = 20000000, Htail = 252, Hloss = 28) =;
quation
inflow_Saguling.Q = inflow_Saguling_Q;
inflow_Cirata.Q = inflow_Cirata_Q;
inflow_Jatiluhur.Q = inflow_Jatiluhur_Q;
lateralLoss_Saguling.QLat_control = lateralLoss_Saguling_QLat1;
lateralLoss_Cirata.QLat_control = lateralLoss_Cirata_QLat1;
lateralLoss_Jatiluhur.QLat_control = lateralLoss_Jatiluhur_QLat1;
lateralLoss_SagulingCirata.QLat_control = lateralLoss_SagulingCirata_Qlat1;
lateralLoss_CirataJatiluhur.QLat_control = lateralLoss_CirataJatiluhur_Qlat1;
lateralLoss_JatiluhurDemand.QLat_control = lateralLoss_JatiluhurDemand_Qlat1;
<pre>node_Drinking.QOut_control[1] = 0;</pre>
node_Agriculture.QOut_control[1] = terminal_Agriculture_Qin2;
<pre>nodeHQPort_Saguling.QOut_control[1] = lateralLoss_SagulingCirata_Qin2;</pre>
nodeHQPort_Cirata.QOut_control[1] = lateralLoss_CirataJatiluhur_Qin2;
nodeHQPort Jatiluhur.QOut control[1] = lateralLoss JatiluhurDemand Qin2;
connect (nodeHQPort Saguling.HQ, linear Saguling.HQ) =;
connect(terminal JatiluhurDemand.QIn, lateralLoss JatiluhurDemand.QLat) =;
connect(lateralLoss SagulingCirata.OLat, terminal SagulingCirata.OIn) =:





Suppose we have the following goals:

- Keep water levels within bounds as much as possible
- Maintain minimum spill flows for fish migration, if possible
- Apply best effort to track the generation request

Let $\{f_i : i \in I\}$ denote the set of functions encoding these goals. We have:

min $f_i \forall i \in I$ subject to $g(x) \le 0$ h(x) = 0

How to solve this?



Pareto optimality

A solution x^* of the problem

 $\min f_i \ \forall i \in I \text{ subject to} \\ g(x) \leq 0 \\ h(x) = 0$

Is *Pareto-optimal* if there is no x^{**} such that for a *j*

 $f_j(x^{**}) < f_j(x^{**})$

and for all $i \neq j$

In words: Pareto optimality implies that no goal can be improved without making another one worse.

 $f_i(x^{**}) \le f_i(x^{**})$

 $\min f_i \forall d$







The *Pareto front* is the set of all Pareto-optimal solutions.



Ensemble techniques



-[1] H.fas (ECMWF)

Multi-stage Stochastic Oprimization

Decision Uncertainty Resolution Decision



Multi-stage Stochastic Oprimization

Decision Uncertainty Resolution Decision







... RTC supports stochastic MPC with <u>multiple model</u> <u>parametrizations</u> or even <u>multiple models</u> in parallel.

All models are tied into the stochastic control tree.





Thank you for your attention!

Feel free to contact me any time at

jorn.baayen@deltares.nl

