A branch and bound method for globally optimising valve locations in water distribution networks

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Outline

Introduction

Problem definition

Global optimization framework

Numerical experiments

Challenges in water supply: growing water demand



Imperial College London Challenges in water supply: climate change (drought and flooding)

Water supply outages and infrastructure deterioration



Challenges in water supply: ageing infrastructure

UK urban water infrastructure > a century old (Ofwat)



Pressure management in water networks

Hydraulic pressure in pipes is a critical control variable for WSNs:

- Leakage losses
- Pipe bursts frequency



Pressure control valves

- Control pressure at their outlets
- Advanced forms of flow and pressure modulation



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Imperial College London

Optimal placement of control valves in WSNs

- Objective
 - Minimize Average Zone Pressure (AZP)
- Continuous variables
 - Node hydraulic heads
 - Pipe flow rates
- Discrete variables
 - Binary variables used to model the placement of valves
- Nonconvex constraints
 - Frictional energy losses within hydraulic conservation laws
- \Rightarrow Nonconvex Mixed Integer Nonlinear Programming (MINLP)



Problem formulation

$$\begin{array}{ll} \text{minimize} & c^T u \\ \text{subject to} & f_i(x_i) - y_i = 0, \quad \forall i = 1, \dots, m \\ & Ax + By + Cu + Dz \leq d \\ & x \in Q \\ & z_j \in \{0, 1\}, \quad \forall j = 1, \dots, n \end{array}$$
 (MINLP(Q))

f_i(·) is a nonlinear function modelling frictional energy losses *Q* is a rectangle representing variable upper and lower bounds.

Previous approaches

Heuristics:

▶ ...

- Genetic algorithms
- Simulated annealing
- Harmony search

Mathematical optimisation:

- Branch and bound on binary variables
- Penalty method
- Linear approximation

No guarantee on the global optimality of solutions

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Imperial College London Polyehdral relaxation for frictional energy losses constraints



Imperial College London Polyehdral relaxation for frictional energy losses constraints



Defined by

$$R_i x_i + E_i y_i \leq r_i, \quad \forall i = 1, \dots, m$$

where matrices R_i , E_i , and vector r_i depend on the rectangle Q.

Presolving : domain reduction

idea: Improve the polyhedral relaxation by solving

 $\begin{array}{ll} \mbox{minimize}/\mbox{maximize} & x_{l} \\ \mbox{subject to} & R_{i}x_{i}+E_{i}y_{i} \leq r_{i}, \quad \forall i=1,\ldots,m \\ & Ax+By+Cu+Dz \leq d \\ & x \in Q \\ & z_{j} \in [0,1], \quad \forall j=1,\ldots,n \end{array}$

for each l = 1, ..., m. **output:** A tightened rectangle Q^{tight}

Branch and bound: basic idea

goal: Find a global optimal solution for $MINLP(Q^{tight})$ to within some prescribed accuracy ε

- Define a partition \mathscr{Q} of Q^{tight}
- For each Q' ∈ 2, compute lower and upper bounds on the optimal value of MINLP(Q'):

 $L(Q') \leq y(Q') \leq U(Q'))$

- ► If $\min_{Q' \in \mathscr{Q}} U(Q') \min_{Q' \in \mathscr{Q}} L(Q') < \varepsilon$, quit
- else, refine partition $\mathcal Q$ and repeat

Ingredients for branch and bound

The algorithm needs

- methods to compute lower and upper bounds
- strategy to select the next region to split
- rule for choosing how to split

Lower bound

Given $Q' \in \mathcal{Q}$, a lower bound is computed solving

$$\begin{array}{ll} \text{minimize} & c^T u \\ \text{subject to} & R_i x_i + E_i y_i \leq r_i, \quad \forall i = 1, \dots, m \\ & Ax + By + Cu + Dz \leq d \\ & x \in Q' \\ & z_j \in \{0,1\}, \quad \forall j = 1, \dots, n \end{array}$$
 (MILP(Q'))

Let $(\hat{x}, \hat{y}, \hat{u}, \hat{z})$ be the solution of MILP(Q')

Upper bound

Given $Q' \in \mathscr{Q}$, an upper bound is computed solving minimize $c^T u$ subject to $f_i(x_i) - y_i = 0, \quad \forall i = 1, ..., m$ $Ax + By + Cu \le d - D\hat{z}$ $x \in Q'$ (NLP(Q'))

Ingredients for branch and bound

The algorithm needs

- \blacktriangleright methods to compute lower and upper bounds \checkmark
- strategy to select the next region to split
- rule for choosing how to split

Branching strategy

▶ Select the rectangle $Q^b \in \mathscr{Q}$ with the best lower bound, i.e.

$$L(Q^b) = \min_{Q' \in \mathscr{Q}} L(Q')$$

Split Q^b at x̂, along coordinate k corresponding to the largest error

$$|f_k(\hat{x}_k) - \hat{y}_k| = \max_{i=1,...,m} |f_i(\hat{x}_i) - \hat{y}_i|$$

Ingredients for branch and bound

The algorithm needs

- \blacktriangleright methods to compute lower and upper bounds \checkmark
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Case studies

Optimal placement of 1 to 5 pressure control valves in

- PescaraNet
 - 365 cont. var.
 - 198 bin. var.
 - 1591 lin. constr.
 - 99 nonconvex terms

- ► Net25
 - 3192 cont. var.
 - 74 bin. var.
 - 9762 lin. constr.
 - 888 nonconvex terms





Numerical results

Comparison with solvers BARON (v18.8.23) and SCIP (v3.2.1).



Large operational water network

Optimal placement of 1 to 5 pressure control valves in

BWFLnet

- 28251 cont. var.
- 2620 bin. var.
- 96599 lin. constr.
- 7107 nonconvex terms



Numerical results

n _v	Time (s)	No. Iter.	LB (m)	UB (m)	Gap(%)
1	86400	783	42.48	47.41	11.61
2	86400	29	35.54	39.31	10.62
3	86400	1	32.44	36.19	11.58
4	> 86400	-	-	-	-
5	> 86400	-	-	-	-

Table: Optimization results on BWFLnet (86400 s = 1 day)



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For further information

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