Nodal Importance Concept for Computational Efficiency in Optimal Sensor Placement in Water Distribution Systems

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Background & Motivation for Study

- Recent interest in protecting drinking water in water distribution systems (WDSs) in the event of terrorist attack by contaminant injection
- Human variables of uncertainty hinder definitive contaminant sensor placement in WDSs
- Methods documented to date
 - computationally-expensive algorithms
 - versimplifying assumptions
 - ➤ inability for implementation on larger systems
- Method needed to increase efficiency of search for optimal placement schemes without compromising WDS protection.





Study Problem

> Objective

Allocate contaminant sensors to WDS nodes in a <u>computationally efficient</u> manner to provide maximum WDS protection.

Performance Measures

- $\succ Z_{time}$ = expected detection time (minimized)
- $\succ Z_{vol}$ = expected contaminated water volume (minimized)
- $\succ Z_{lik}$ = detection likelihood (maximized)

Limiting Factor

- \succ *M* = fixed number of sensors available
- Attack Scenario
 - eligible injection node: any one node in WDS
 - eligible injection time: any 5-minute multiple of first 1/4 of study period
 - ➤ injection: constant mass flow of conservative contaminant





Study System

BWSN Network 1 *



▶ 129 nodes (126 junctions, 2 tanks, 1 reservoir)

178 links (168 pipes, 2 pumps, 8 valves)

localized flow behavior

large variance in hydraulic demand (63% of junctions with demand)

➢ 96-hour study period



* "Battle of the Water Sensor Networks," 8th Annual WDSA Symposium



Nodal Importance Concept (1)

Nodal Importance Defined

- degree to which an individual WDS node should be considered as a candidate for sensor placement
- related to potential on average for adverse effects to be experienced at an individual node under an unknown attack scenario

Use of Concept

- ➤ isolate a subset of "more important" nodes to confine search domain
- test different combinations of subset nodes with optimization algorithm to find sensor placement scheme providing maximum protection

Expected Advantages

- better-performing sensor placement schemes
- reduced computational runtimes





Nodal Importance Concept (2) ≻Nodal Importance Function

$$f_{is} = \alpha_1 \frac{V_{is}^{cont}}{\max_i (V_{is}^{cont})} + \alpha_2 \left(\frac{T_s - t_{is}^d}{T_s}\right) D_{is}$$

- V_{is}^{cont} = total contaminated volume associated with node *i* under scenario *s* assuming no contaminant detection at node *i*
- t_{is}^{d} = time after injection during scenario *s* when contaminant is first present at node *i* (if not present at any time, $t_{is}^{d} = 2$ x study period duration)
- T_s = time after injection at the end of study period for scenario s
- $D_{is} = 1$ or 0, indicating contaminant presence or absence, respectively, of contaminant at node *i* at any time during scenario *s*
- α_1 , α_2 : scalars in domain [0, 1] ($\alpha_1 + \alpha_2 = 1$)





Nodal Importance Concept (3)

Weighting of importance function terms

- Use array of (α₁, α₂) schemes to capture nodes according to different protection preferences
- $\succ (\alpha_1, \alpha_2) = (1, 0), (0, 1), \& (1/2, 1/2)$ for this study

Relative Importance

$$f_{is}^{rel} = \frac{f_{is}}{\max_{i} (f_{is})}$$

Expected Relative Importance

$$F_i = \frac{1}{S} \sum_{s} f_{is}^{rel}$$

S = number of Monte Carlo scenarios run

 \succ Used to rank nodes for a particular (α_1, α_2) scheme





Subset Creation (1)

Base Subsets to Total Subset

> each "base subset" of more-important nodes corresponds to a particular (α_1, α_2) scheme

"total subset" is the union of all base subsets generated





Only nodes in total subset tested for sensor placement
The optimal subset: smallest subset that includes optimal nodes for sensor placement





Optimization (1)

Program

$$\max Z_{comb} = \gamma_1 \frac{\max_s(t_s^d) - Z_{time}}{\max_s(t_s^d)} + \gamma_2 \frac{\max_s(V_s^{cont,d}) - Z_{vol}}{\max_s(V_s^{cont,d})} + \gamma_3 Z_{lik}$$

s.t.
$$\sum_{a=1}^A m_a = M \quad \forall all \ a$$
$$m_a \in \{0,1\} \quad \forall all \ a$$

 t_s^d = time after attack when contaminant is first present under scenario *s* at any node *i* with an assigned sensor

 $V_s^{cont,d}$ = total volume contaminated under s at all *i* until t_s^d

 $m_a = 0$ or 1, indicating the absence or presence, respectively, of a sensor at total subset node a

A = number of nodes in total subset

 $\gamma_1, \gamma_2, \gamma_3$: scalars in the domain [0, 1] $(\gamma_1 + \gamma_2 + \gamma_3 = 1)$





Optimization (2)

"Simple" Genetic Algorithm



> chromosome: binary string; bits represent m_a values

population: initialized in uniform, random manner

→ **crossover**: one-point, $p_{cross} = 0.95$; parents chosen through roulette-wheel selection according to F_i values under $(\alpha_1, \alpha_2) = (1/2, 1/2)$

mutation: uniform, random "bit-flipping", $p_{mut} = 0.05$

post-handling: uniform, random "bitflipping" to satisfy sensor availability constraint

elitist selection: according to objective function





Performance Testing (1)

Decision Variables

- variables kept constant
 - > M = 5
 - **> S** = 3,000
 - > GA population size = 500
 - \succ number of GA generations = 500
- designated weighting schemes
 - $\succ (\boldsymbol{\alpha}_1, \boldsymbol{\alpha}_2) = (1, 0), (0, 1), (1/2, 1/2)$

$$\succ$$
 ($\gamma_1, \gamma_2, \gamma_3$) = (1/4, 1/2, 1/4)

➤ subset size

- ➤ critical variable for testing use of importance concept allowed to vary
- ➤ base subset size candidates: 5, 10, 20, 30, 40, 50
- Computational Runtimes
 - ➢ less than 1 hour for all tests employing nodal importance concept





Performance Testing (2)

Method Results

Method	Base Subset Size	Total Subset Size	Sensor Nodes ("JUNCTION-x")	Z _{time} (min)	Z _{vol} (gal)	Z _{lik} (%)	Z _{comb} x 10 ²
random placement				7382	10994	37.6	20.1
GA-only		129	17, 49, 84, 100, 122	3783	2269	70.5	73.7
ranking-GA (a)	5	9	17, 30, 68, 83, 126	4162	3949	66.9	64.9
ranking-GA (b)	10	18	30, 68, 83, 102, 118	3686	1985	71.1	75.3
ranking-GA (c)	20	37	17, 68, 83, 100, 103	3272	2635	75.1	74.6
ranking-GA (d)	30	47	68, 83, 100, 102, 118	3034	3038	77.2	74.1
ranking-GA (e)	40	59	45, 68, 83, 103, 118	3236	2665	75.2	74.6
ranking-GA (f)	50	73	68, 83, 100, 103, 117	3222	2954	75.6	73.5





Performance Testing (3)

Sensor Placement: Method "ranking-GA (b)"



Performance Testing (4)

> Performance vs. GA Generation

Georgia Tech

School of





Performance Testing (5)

Benefit-Cost Ratios

- Benefit: gained performance for method beyond baseline from random placement
- B/C Ratio: benefit averaged over all scenarios up to GA generation of convergence <u>for all</u> <u>methods</u>



Total Subset Size





Observations & Conclusions

- Using nodal importance concept can lead to heightened efficiency in the optimization of contaminant sensor placement without compromising WDS protection goals.
- A subset of more-important nodes too small in size may not provide enough diversity for finding a sensor placement scheme of acceptably high performance.
- As the size of a subset increases toward the total number of WDS nodes, performance reaches a peak value then converges to a value resulting from optimization without using the importance concept.





Future Work

- Applying nodal importance concept to optimization of sensor placement in larger systems
- Resolving ambiguities
 - definitiveness of importance functions & corresponding variables
 - Inumber of Monte Carlo scenarios to run
 - ➤ subset sizes
- Developing means of faster WDS simulation



