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Operational and Tactical Management of Water and Energy Resources in Pressurized Systems: the competition at WDSA 2014

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ABSTRACT

Optimal management of water and energy resources worldwide is a basis for environmental and socio-economic sustainability in urban areas, which has become even more relevant with the advent of the “smart” and “water sensitive” city paradigm. In water distribution networks (WDNs) water resource management is concerned with increased efficiency, which is mainly related to the reduction of leakages, while energy management refers to optimal pump, valve and source scheduling strategies considering the hydraulic system requirements. These management goals require planning of asset renewal and improvement works in the short time (operational) and medium time (tactical) horizons, considering the financial sustainability of relevant actions. The Battle of Background Leakage Assessment for Water

25 Networks (BBLAWN) was designed as a competition held at the 16th Water Distribution Systems
26 Analysis Conference, in Bari (Italy) in 2014 (WDSA 2014), to address the aforementioned management
27 goals. The teams taking part in the BBLAWN were asked to develop a methodology for both reducing
28 real water losses and saving energy in a real WDN considering the possibility of asset renewal and
29 strengthening. Fourteen teams from academia, research centers and industry presented their solutions at
30 a special session of the WDSA 2014 conference. This paper briefly describes the BBLAWN and presents
31 one of the solutions provided by the organizers to illustrate the ideas and challenges embedded in the
32 posed problem.

33 The overview of the solutions provided by the participants shows that management decisions need to be
34 supported by engineering judgment as well as with tools that combine computationally effective multi-
35 objective optimization and hydraulic models capable of assessing pressure-dependent background
36 leakages.

37

38 **Keywords**

39 Water distribution network; Integrated management; Leakages; Energy; Hydraulic models; Pressure Reduction Valves.

40

41 **Introduction**

42 The series of “Battle Competitions” date back to 1985 with the Battle of the Water Networks (BWN)
43 (Walski et al., 1987), and was created to stimulate academia, research centers and industry to provide
44 solutions and strategies for addressing complex practical problems in water distribution network (WDN)
45 analysis, design and management. More recently the Battle of the Water Sensor Networks (BWSN)
46 (Ostfeld et al., 2008) was held in Cincinnati (OH, USA); the Battle of the Water Calibration
47 Networks (BWCN) (Ostfeld et al., 2012) was held in 2010 in Tucson (AZ, USA); the Battle of the Water
48 Networks Design (BWN-II) (Marchi et al., 2014) was held in 2012 in Adelaide (Australia).

49 The Battle of Background Leakage Assessment for Water Networks (BBLAWN) was held at the 16th
50 Water Distribution Systems Analysis Conference, in Bari (Italy), in July 2014 (WDSA 2014), thus being
51 the fifth “Battle” on WDNs. The problem was designed to stimulate a discussion about the optimal
52 management of water and energy resources in WDNs. This is actually an emerging issue relevant from
53 environmental and socio-economic perspective worldwide, also pertaining to smart city paradigm.

54 The complexity of WDN analysis and management is increasing due to the growth of population in urban
55 areas and the increase of system size and interconnectivity. Real water losses in Europe range from 20
56 to 40% (and more in some environments) mainly due to the natural asset deterioration of aged WDNs.
57 For example, in Italy, it is estimated that over 64 billion euros will be needed in the next 30 years for
58 WDN rehabilitation (FederUtility’s Blue Book 2011). The asset deterioration and the consequent real
59 water losses are relevant water management issues because the inefficient use of water resources
60 exacerbates the impact of water scarcity due to socio-economic factors and/or climate changes.
61 Therefore, water companies seek management solutions and convincing/effective decision making
62 strategies to support real leakages reduction in short-medium and long time horizons and for managing
63 the rapid deterioration of assets which has an enormous public value. These facts make urgent for water
64 utilities to undertake actions in the short-medium time horizon, which need to be effective also in the
65 long time horizon. Optimal management of water resources in WDN actually means to minimize water
66 losses from deteriorated infrastructures and, more explicitly, the background leakages from pipes. These
67 type of distributed losses are less evident than major bursts and usually run for longer before repair
68 (Germanopoulos, 1985). In addition, in aged pipes the joint effect of both increased head losses (due to
69 increased internal roughness) and background leakages causes pressure drop through the system. A
70 commonly adopted countermeasure for this consists of increasing water pumping into the system in order
71 to provide sufficient pressure to deliver water to a service reservoir or directly into distribution. This, in
72 turn, results in increased water losses and energy consumption.

73 Thus, water and energy management are directly related and depend on WDN operation (e.g.
74 filling/emptying of tanks), pressure regime through the network and the total water demand, including
75 both customers' water requirements and leakages (Giustolisi and Walski, 2012).

76 On this premise, minimizing water and energy consumption is a complex problem that, in the short-term
77 horizon, requires effective *operational* strategies, as well as sustainable asset renewal plans for the
78 *tactical* planning (medium term horizon). In fact, the reduction of water leakages in the short time horizon
79 could be achievable by implementing optimal pumping (e.g. Giustolisi et al., 2013) as well as by
80 installing pressure control valves to avoid excessive pressure in some parts of the network. Nonetheless,
81 in real systems there is a range of technical asset management options including pipe renewal (e.g.
82 replacement, relining) or installation of new pipes in parallel to the existing ones, enlargement of existing
83 tanks or enhancement of pumping stations. The selection of the most effective alternative needs to be
84 evaluated in the medium term horizon, and in conjunction with optimal operation strategies. In addition,
85 each technically feasible solution needs to be evaluated in terms of financial sustainability, considering
86 total costs, i.e., both operational (OPEX) and capital (CAPEX) expenditure, in order to be readily
87 evaluated by water utilities.

88

89 **The Battle of Background Leakage Assessment for Water Networks - BBLAWN**

90 The BBLAWN called for teams from academia and industry to design a methodology for reducing water
91 losses due to background leakages, considering the cost for upgrading the hydraulic system capacity. The
92 intervention options available to the teams were pipe replacement or installation in parallel to existing
93 pipes, installation of new parallel pumps and enlarging tanks (i.e. addition of new cylindrical tanks
94 adjacent to the existing tanks), the installation of pressure control valves (PRVs), while considering also
95 the cost of energy and water losses (see Giustolisi et al. (2014) and BBLAWN webpage for further
96 details). The aim was to stimulate competing teams to deal with the conflicting cost objectives (i.e., asset

97 upgrading versus energy cost and leakage reduction versus system pressure reduction using costly control
98 valves).

99 Actually, devising strategies for water leakage management should encompass also environmental and
100 social sustainability criteria, beyond economic and technical objectives. The BBLAWN problem
101 statement accounts for such aspects in terms of “externalities” representing environmental and social
102 costs and benefits (Delado-Galvan et al., 2010; European Commission, 2013) like, for example, the
103 impact on water resources or the damages caused by leakages. The externalities are computed using the
104 cost of water as a proxy for the environmental and resource cost, beyond the operational costs (that are
105 part of the water tariff for customers and is related to the water company annual balance). Based on these
106 considerations, the cost of water lost volume is fixed at 2 €/m³.

107 The competition used C-Town (Ostfeld et al. 2012) whose network layout is reported in Fig. 1. To solve
108 the BBLAWN problem, it was assumed that the city has already commissioned the development of a
109 calibrated hydraulic model of the existing network to be used in evaluating its present state and future
110 improvements and performance. Therefore, the network model includes the network layout, the demand
111 patterns and the background leakage model parameters. It also contains existing pump and tank
112 characteristics and the controls of pumps and valves based on water level in tanks.

113 The existing infrastructure is not able to meet the pressure performance target of 20 m at each node with
114 demand, and the situation is compounded by excessive background leakage. Therefore, the water utility
115 is interested in minimizing operational and capital costs.

116 The design problem must be solved as a one-stage intervention problem (i.e. both operational and capital
117 costs to be minimized are reported as annual cost, which account for the lifetime of the single component
118 and the discount rate), and the teams were asked to come up with a solution respecting other common
119 engineering considerations and operational constraints in order to propose a methodology and provide
120 one feasible solution from the utility standpoint. For this reason, the solutions were evaluated by the

121 organizers in terms of operational and capital costs, but also accounting for the engineering soundness
 122 of the methodology used and the technical justification for the choices taken by the teams (e.g. the
 123 consistency with the expected hydraulic WDN behavior over an operational cycle, the rationale behind
 124 the selection of candidate pipes for replacement or the location of PRV, and so on).

125 In fact, the BBLAWN competition was designed as close as possible to a real situation in terms of
 126 complexity and design/operational options. This was aimed at stimulating the discussion and exchange
 127 of information among the different teams about the use of optimization tools, the need for enhanced
 128 hydraulic modelling to predict the background leakages and the whole system behavior, as will be
 129 discussed in the next section.

130 Fig. 1. TOWN-C for BBLAWN composed of 444 pipes, a reservoir (R), seven tanks (Tx), eleven
 131 pumps (PM), a control valve (CV), a check valve (CH).

132

133 ***Hydraulic and Leakage Modelling***

134 Water leakage is caused by small or large breaks and openings in pipes, which occur at water mains and
 135 along the pipe connections to properties. The technical literature classifies leakages in background and
 136 burst leakage (unreported or reported) depending on the level of outflow. Germanopoulos (1985)
 137 proposed the following model for background leakages:

$$138 \quad d_k^{leaks}(P_{k,mean}) = \begin{cases} \beta_k L_k P_{k,mean}^{\alpha_k} & P_{k,mean} > 0 \\ 0 & P_{k,mean} < 0 \end{cases} \quad (1)$$

139 where k = index referring to the k th pipe; $P_{k,mean}$ = model mean pressure along the k th pipe in [m] (see
 140 next section for details); d_k^{leaks} = background leakages outflow along the k th pipe in [m³/sec]; α_k [-] and
 141 β_k [m^{2- α_k /s] = model parameters; L_k = length of the k th pipe, in [m].}

142 Background leakages are diffuse (spatially distributed) and low intensity losses (outflows) along pipes

143 (mains and connections), which depend on the asset condition, i.e., as related to the multiplier β in Eq.
144 (1). They run continuously over time and could cause significant losses from the system.

145 Bursts are the natural evolution of background leakages due to external forces/factors, which act on
146 deteriorated pipes. The model in Eq. (1) is aimed at predicting the outflows of diffuse leakages,
147 considering also unreported small bursts, thus it is useful for planning purposes. This is opposite to burst
148 modelling, which is much more suited for operational purposes, e.g., for outflow location and
149 consequence prediction. Therefore, the competing teams were asked to employ hydraulic modelling
150 considering background leakages (Giustolisi et al., 2008) because the hydraulic consistent prediction of
151 those outflows not only influences the computation of the water losses but also the assessment of the
152 system capacity, energy and water use.

153 The need for an accurate prediction of the system behavior is important to design an effective solution
154 for real systems. To this purpose, the teams were asked to compute the energy for pumping using the
155 following formulations involving expressions for variable head and efficiency:

$$\begin{aligned} H &= H^s - rQ^c \\ \eta &= -\frac{4\eta_{max}}{Q_{max}^2} Q^2 + \frac{4\eta_{max}}{Q_{max}} Q \\ Q_{max} &= \left(\frac{H^s}{r} \right)^{\frac{1}{c}} \end{aligned} \tag{2}$$

157 where η_{max} = maximum pump efficiency; H^s , r and c = parameters of the pumps. Eq. (1) represents a
158 parabolic function with the maximum value (η_{max}) at $Q_{max}/2$ (Giustolisi et al., 2013).

159

160 ***Background Leakages versus Burst Modelling***

161 Background leakage modelling, Eq. (1), for planning purposes is different from modelling a single burst
162 for operational purposes like, for example, for its detection and/or preliminary localization.

163 The model in Eq. (1) depends on the average pressure in pipes, because leakages along mains and pipe
 164 connections are dependent on pressure. Consequently, the average local pressure is a good indicator
 165 influencing the total leakage in a pipe. In fact, the model in Eq. (1) states that the overall leakage outflow
 166 (the volume of water losses), is proportional to the average, i.e. local, pressure in the hydraulic system
 167 where the exponent α is related to the pipe material (i.e. stiffness) (Giustolisi et al., 2008). From the
 168 hydraulic modelling point of view, it is important to remark that, given the k -th pipe whose end nodes
 169 are i and j , the model for background leakages in Eq. (1) is different from the model for pipe bursts (i.e.,
 170 outflows from nodes). The model in Eq. (1) states that the background leakages for pipe k are:

$$171 \quad d_k^{leaks}(P_{k,mean}) = \beta_k L_k \left(\frac{P_i + P_j}{2} \right)^\alpha \quad (3)$$

172 and, for modelling purpose, such background leakage outflow along the k th pipe is concentrated at two
 173 water withdrawal points at the end nodes, and divided equally:

$$174 \quad d_i^{leaks}(P_{k,mean}) = d_j^{leaks}(P_{k,mean}) = \frac{\beta_k L_k}{2} \left(\frac{P_i + P_j}{2} \right)^\alpha \quad (4)$$

175 Lumping the pipe level outflow at the end nodes preserves the mass balance while causes an error in the
 176 energy balance equation. The magnitude of the error can be evaluated as in Giustolisi and Todini, (2009)
 177 and Giustolisi (2010).

178 The strategy of using a concentrated outflows at pipe ending nodes characterized by the outflow
 179 coefficient $\beta_k L_k / 2$, (i.e., assuming a burst model surrogating the background leakage model), results in
 180 the following computed outflows from nodes i and j respectively:

$$d_i^{leaks}(P_i) = \frac{\beta_k L_k}{2} (P_i)^\alpha; \quad d_j^{leaks}(P_j) = \frac{\beta_k L_k}{2} (P_j)^\alpha \Rightarrow$$

$$d_k^{leaks}(P_i, P_j) = \beta_k L_k \frac{(P_i)^\alpha + (P_j)^\alpha}{2}$$

181

182 This assumption generates a modelling error, represented by the difference between d_k^{leaks} of Eq. (5)
 183 and Eq. (4), that is actually a function of asset (i.e. α, β, L) and nodal pressures,

$$\beta_k L_k \left(\frac{P_i + P_j}{2} \right)^\alpha - \beta_k L_k \frac{(P_i)^\alpha + (P_j)^\alpha}{2} = f(\alpha, \beta_k, L_k, |P_i - P_j|)$$

184

185 It is worth noting that nodal outflows computed by Eq. (4) and (5) are different even if $\alpha=1$ is used:

$$d_i^{leaks} = \frac{\beta_k L_k}{2} \left(\frac{P_i + P_j}{2} \right) \quad d_j^{leaks} = \frac{\beta_k L_k}{2} \left(\frac{P_i + P_j}{2} \right) \quad \text{background leakage model}$$

$$d_i^{leaks} = \frac{\beta_k L_k}{2} P_i \quad d_j^{leaks} = \frac{\beta_k L_k}{2} P_j \quad \text{burst model}$$

186

187 Indeed, Eqs (4) and (5) return different leakage outflows lumped at nodes causing different pressures
 188 through the network, which, in turns, change the background leakage outflows.

189 In summary, for any $\alpha \neq 1$, the difference between the background leakages prediction on a single pipe is
 190 evident as reported in Eq. (6), while for $\alpha=1$ the predictions become different because the demands and
 191 pressure distribution in the network are different.

192

193 **Solution of the Competition Organizers**

194 The organizers of the BBLAWN also solved the problem in order to verify its feasibility and provide a
 195 further contribution to the discussion. The solution is developed using a mix of engineering judgment,
 196 system optimization and extended period simulation (EPS) analysis aimed at supporting the decisions

197 step by step. The solution was designed in three steps that are summarized here and detailed in the
198 following.

199 Step 1. Pump scheduling optimization of the original hydraulic system is performed first without
200 upgrading any assets. The step is useful for the assessment of the initial level of leakage
201 (assuming optimal pumping) and the hydraulic capacity of the system. The EPS analysis of the
202 optimized system allowed the identification of critical nodes in terms of pressure requirements.
203 Together with the analysis of the hydraulic behavior of the WDN they were used to select
204 candidate pipes for replacement in the comprehensive system optimization of step 2.

205 Step 2. Hydraulic system optimization is performed considering the cost of: (i) pipe replacement;
206 (ii) tank enlargement; (iii) new installed parallel pumps; (iv) pump scheduling; and (v) water
207 loss reduction. Before optimization runs, some pipes of the WDN were closed at no cost (since
208 in the BBLAWN problem statement an isolation valve is assumed present on each pipe; these
209 pipes are reported as dotted lines in Figure 2). Indeed, closing a pipe allowed all the water
210 feeding a network segment to go through the pipes with a PRV. It was assumed that PRVs are
211 not installed yet in Step 2 but they would be installed in the future with the option of a multi-
212 stage intervention strategy.

213 Step 3. Pump scheduling optimization is performed by considering 25 PRVs already installed,
214 and the asset-intervention solution obtained in step 2. The pump scheduling problem was then
215 solved and the 25 PRVs were ranked based on their individual contribution to the reduction of
216 water losses. On the one hand, this strategy permitted to have the total cost of the intervention
217 together with the total expected reduction of energy and water loss costs (as requested by
218 BBLAWN rules). On the other hand, it supports the utility in selecting the most effective
219 sequence of valves to install considering the incoming of budget and the marginal advantage
220 of each installation.

221

222 *Step 1. Optimal pump scheduling of the original hydraulic system*

223 This stage provided a solution showing a small pressure deficit at two nodes (indicated with empty black
224 circles) in Figure 2, occurring at the first hour of the weekly operational cycle. The volume of water
225 losses during the week was 36,281 m³, corresponding to 26.05 % of the total water put into the system,
226 which corresponds to the weekly customer demand of 102,973 m³. The weekly energy consumption was
227 42,221 KWh, corresponding to a cost (given the energy tariff pattern) of 5,176 €. The solution of this
228 stage was helpful for understanding WDN behavior over time (EPS analyses). In addition, it represents
229 the maximum system performances achievable without any asset upgrade, thus being of direct relevance
230 for the water utility.

231

232 *Step 2. Hydraulic system optimization with upgrade of hydraulic capacity and closing pipes*

233 The engineering judgment and EPS analyses drove the system optimization mainly to upgrade the system
234 hydraulic capacity. To this purpose, the candidate pipes to be replaced were identified as those located
235 along the transmission lines (see blue segments in Figure 2). There are three basic motivations for
236 selecting the main transmission pipes.

- 237 1. The hydraulic capacity of the network was reduced by closing some additional pipes (dotted
238 lines in Figure 2) to prepare the system for the installation of PRVs (based on engineering
239 judgment). This affected the ability to deliver water from the pump system of DMA 1 (i.e., close
240 to the reservoir) to the tanks n.2 and n.6 (see Figure 1) and to the four inline pump systems of
241 DMAs 2-5.
- 242 2. As it is not hydraulically feasible to reduce the pressure along transmission pipes by installing
243 PRVs, it is better to replace these pipes in order to reduce the volume of water losses. In addition,
244 from system reliability perspective is better to renew transmission pipes whose failure would

245 reduce significantly the hydraulic capacity.

246 3. Interventions on transmission pipes are cost efficient for the utility considering a one-stage
247 intervention. Furthermore, this approach reduced the search space during the optimization stage,
248 which improved in terms of computational efficiency and effectiveness.

249 Consistently with the choice of increasing the system hydraulic capacity, six new parallel pumps were
250 assumed as candidates for upgrading the pump system of DMA 1 and two for each inline pump systems
251 of DMAs 2-5. Finally, tanks were considered as candidate for enlargement in order to reduce the energy
252 cost (through optimal pumping) and to increase the hydraulic capacity of the DMAs 2-5, where pipes
253 were not replaced, together with the possibility to increase the maximum power of the local pump
254 systems.

255 Fig. 2. TOWN-C pressure control valve (PRV) and node of pressure set (Pset).

256

257 In summary, the overall approach was to divide the network into smaller areas by using 25 PRVs (whose
258 settings will be defined in step 3) in order to reduce the pressure locally and increase the hydraulic
259 capacity by means of the replacements of DMA 1 transmission pipes. Additionally, upgrading the main
260 pump system (in DMA1) and tank n.2 was also considered. Furthermore, it is possible to increase the
261 local hydraulic capacity of the DMAs 2-5 by upgrading inline pump systems and by enlarging internal
262 tanks.

263 Figure 3 shows only the capital costs of Pareto solutions obtained by the multi-objective optimization
264 procedure, where separate costs (i.e., pipe and pump cost; energy and water loss cost; and tank enlarging
265 cost) were minimized simultaneously. This was achieved by using a dedicated function available in the
266 WNetXL system (Giustolisi et al., 2011) that permits to manage the entire problem using advanced
267 hydraulic simulation (e.g. accounting for pressure-driven simulation of background leakages and
268 remotely controlled PRV) and decision support (e.g. for optimal pump scheduling and asset upgrades)

269 functions developed in the latest technical-scientific research working in Microsoft-Excel® environment
270 (for details www.hydroinformatics.it). It is worth recalling that Figure 3 refers to the capital costs only,
271 since the main aim of step 2 is to support decisions on asset upgrade. The fifth solution from the left of
272 the Pareto front (see Figure 3) was selected based on engineering judgment. Indeed, this solution permits
273 the WDN hydraulic capacity to increase by replacing seven pipes and enlarging two tanks, with tank
274 water levels controlling the pumps. This entails cheap asset strengthening works, which could be
275 immediately implemented by the water utility, being also a good starting point for next optimizations.
276 The solutions results in 25.11 % of leakages and required 13,306 € for the replacement of pipes and
277 44,660 € for the enlargement of tanks T2 and T3 to the maximum volume of 1,693 m³ and 180 m³,
278 respectively.

279 Figure 2 reports a black solid circle on the seven replaced pipes of the transmission line and a square on
280 the enlarged tanks (i.e. T2 and T3). A pipe was also replaced (based on EPS analysis) in one segment of
281 DMA 1 that was prepared to allocate a PRV (indicated with “7” in Figure 2) (by closing two pipes).
282 Finally, the solution has one new pump (identified with a white square in Figure 2), at the cost of 4,339
283 €, to be installed for the DMA 2. The total weekly energy consumption for this solutions is 42,164 KWh,
284 corresponding to a cost (given the energy tariff pattern) of 5,074 €.

285 Fig. 3. Pareto front of solutions for the multi-objective optimization problem (pipe and pump cost vs.
286 energy and water loss cost vs. tank enlarging cost).

287

288 *Step 3. Pumping optimization considering all the PRVs and ranking of their installation*

289 Once the upgrading of assets was completed, the EPS analysis was performed to locate critical nodes for
290 controlling PRVs. Remotely controlled pressure devices were used and critical nodes were selected based
291 on the elevation and the hydraulic distance from the valves (remote set control points of PRVs are
292 reported as red triangles in Figure 2). The selection of the critical nodes in a DMA (i.e., experiencing

293 minimum pressure) to control PRVs allows setting the pressure at 20 m (minimum pressure for a correct
294 service) which does not change over time (Giustolisi and Walski, 2012). This way the optimal control of
295 the degree of valve opening does not require modulating the pressure based on the node immediately
296 downstream from the PRV, which needs to be predicted by the model based on assumptions about
297 demand variation over time. Of course, such solution requires that the hydraulic model to be used for
298 assessing system performances is capable of simulating remotely controlled PRVs.

299 Furthermore, the pressure in the segment with no demand (see shadowed area in Figure 2) was kept low
300 by setting it at 2 m at the critical node (i.e. as per BBLAWN rules). The pumping schedules with the
301 setting of 25 PRVs was then optimized achieving a solution with the 18.60% of leakages (23,531 m³ of
302 water loss) and 37,430 KWh of energy consumption corresponding to a reduced cost of 4,438 €.

303 The above optimal pumping schedule was set and the EPS analysis was performed assuming the
304 installation of one PRV at a time. The 25 PRVs were ranked in descending order based on leakage of
305 reduction achievable by installing each PRVs. This was followed by analyzing the cumulative effect of
306 the sequential installation of 25 PRVs. Table 1 reports the results in terms of weekly water losses,
307 percentage of leakages and energy consumption expected by progressively adding PRVs.

308 Table 1 could be used as a multi-stage intervention support system allowing the user to assess the residual
309 water losses and energy reduction. It is possible to optimize pumping for each new installation as the
310 control of pumps by tank levels is robust with respect to small variations of demand and/or leakages
311 (Giustolisi et. al, 2014). Finally, Table 2 summarizes the relevant data considering the original and the
312 optimized solutions.

313 Table 1. Ranking of the PRVs

314 Table 2. Relevant data of the initial and final status of the network. Operational costs are weekly-based.

315

316 The solution obtained by organizers has an annualized capital cost of 62,305 € + 26,182 € (i.e. for the

317 investment upgrading the asset and for the installation of PRVs), while the reduction of the weekly-based
318 operational costs with respect to the initial condition is about 26,000 € (although that cost is not merely
319 based on economic evaluations regarding the water losses but also financial consideration, as it accounts
320 for the savings achievable as PRVs are progressively installed). If the cost of the lost water was assumed
321 to be 0.5 €/m³, thus neglecting “externalities” in the water cost (e.g., the impact of socio-environmental
322 factors), the reduction in the weekly operational costs is about 7,000 €, which becomes about 37,000 €
323 when calculating it on annual basis to be compared with the investment. Therefore, the leakage reduction
324 could be less significant if the environmental value of water losses is not considered. However, leakages
325 are indicators of general deterioration and pressure in the system. Therefore, the economic impact of
326 unplanned interventions caused by the natural progress of deterioration, should be considered when
327 performing a cost-benefit evaluation of the reduction of water losses.

328

329 **Brief presentation of methodologies proposed by the participant teams**

330 Fourteen teams from academia, research centers and companies provided their solutions for the
331 BBLWAN at WDSA 2014. Here they are briefly presented in the order they were submitted to the
332 conference website; thus such order does not reflect any judgment on the methodologies. Further details
333 on the single approaches and solutions are reported in individual papers authored by each competing
334 team.

335 Morley and Tricarico (2014, 2015) presented a methodology based mainly on the use of population-
336 based optimization algorithm. They formulated the problem as a constrained single and multiple-
337 objective optimization, implementing a generic hydraulic optimization and benchmarking software
338 application (Acquamark – see Morley and Tricarico (2014, 2015) for details). To permit multiple
339 solutions to be executed and evaluated in parallel a distributed computing architecture was implemented.
340 A pressure-driven demand extension to the EPANET2 (Rossman, 2000) hydraulic model is employed to

341 assist the optimization techniques in accurately ranking near-feasible solutions and to dynamically
342 allocate leakage demand to the end nodes of each pipe.

343 Roshani and Filion (2014) presented a methodology based on a multi-objective optimization approach to
344 minimize capital and operational costs of the network, employing NSGA-II (Deb et al., 2002). The
345 optimization includes all the decision variables involved, e.g., pipes, valves, pumps and tanks, subject to
346 pressure and water level in tanks constraints. The EPANET2 network solver is used to evaluate pipe
347 leakages (simulated as pressure-dependent by means of the orifice discharge coefficient reflecting the
348 leakage model coefficient in Eq. (3)), as well as to evaluate the hydraulic constraints (i.e., nodal pressures,
349 tank levels, etc.). The C# programming language was used to couple the EPANET2 network solver with
350 the NSGA-II engine. Multi-threading (parallel processing) was used to reduce the computational time.

351 Iglesias-Rey et al. (2014, 2015) presented a methodology combining the use of engineering judgment
352 and an optimization model based on a pseudo-genetic algorithm. The methodology consists of two stages:
353 an analysis of marginal costs of pipes considered for replacement, followed by the network topological
354 analysis to study the pipes that could be potentially closed in order to facilitate pressure control.
355 Additionally, a methodology for studying branched areas was also developed, determining possible
356 location for pressure reducing valves. This approach was aimed at reducing the number of decision
357 variables, thus reducing the domain of the specific optimization model in the second stage. Network
358 hydraulic analysis has been performed using the EPANET2 network solver using emitters at nodes to
359 simulate leakages.

360 Creaco et al. (2014, 2015) proposed a multi-objective optimization approach considering three objective
361 functions (i.e., minimization of installation cost, operational cost and PRVs cost). The approach consists
362 of four steps. First, some feasible solutions are identified based on engineering judgment. Then, for step
363 two and three, the NSGAI optimizer was implemented to find an optimal set of solutions: firstly
364 considering only to capital and operational costs, and then considering operational and pressure reducing

365 valves costs. Finally, by grouping the solutions found at the end of previous optimization steps the final
366 three-objective Pareto surface was derived and the best solution selected. The methodology implements
367 the EPANET2 hydraulic solver simulating leakages with emitters first, and then assessing leakages using
368 a sub-routine that applied the Germanopoulos' formula.

369 Price and Ostfeld (2014, 2015) proposed a methodology based on the successive Linear Programming
370 by minimizing costs. A linear representation was solved successively for the non-linear constraints of
371 headloss, leakage, pump energy consumption and pipe sizing. The optimization model returned minimal
372 cost pump scheduling and pipe sizing while minimizing leakage and maintaining minimum service
373 pressures to the consumers. The problem is divided into four main parts: PRV positioning, pumping
374 station and water tank sizing, pipe sizing and pump scheduling for minimum leakage and operational
375 cost. The resulting optimal pump scheduling was not controlled by the water levels in the tanks (as
376 required by the main BBLAWN rules) as the pumps are operated to maintain minimum water pressures
377 at the consumer nodes while utilizing minimum electrical tariff periods. For this reason the solution
378 provided was not accepted for the competition since it was not comparable with other teams that complied
379 with the rules.

380 Diao et al. (2014, 2015) proposed a methodology based on a clustering-based hierarchical decomposition.
381 The network is decomposed into a twin-hierarchy pipeline structure consisting of backbone mains and
382 community feeders. The method consists of three steps: clustering analysis; vulnerability analysis; and
383 identification of backbone mains and community feeders. The system was topologically decomposed
384 into backbone mains and 28 communities. Optimal pressure control strategies for each cluster is
385 addressed in a sequential manner based on the cluster hierarchy with constraints on network performance.
386 Considering such simplified topology, the most cost effective PRV placement strategy and pipe
387 upgrading options for each branch cluster were identified.

388 Eck et al. (2014, 2015) proposed a methodology that decomposes the problem according to the type of

389 intervention, considering and assessing each type in sequence. Initially, a diagnosis of the network is
390 performed through simulating its hydraulic behavior with no infrastructure or operational modifications.
391 An optimization technique is then developed to recommended improvements of a particular type, such
392 as pipes to replace. The presented technique is applied sequentially to yield a list of suggested
393 improvements for the network. The leakage simulation problem was transformed into an equivalent
394 formulation for which EPANET can be applied. To simulate the leakage equations, an iterative technique
395 was developed using the emitters feature in EPANET.

396 Tolson and Khedr (2014; Khedr and Tolson, 2015) propose to rely on engineering judgment with limited
397 use of optimization to generate an approximation of the Pareto-optimal front without intensive
398 computational requirements. A simple heuristic approach consisting of a five-stage approach based on
399 enumeration and trial-and-error (WDN modeler expert judgment) was used to identify and prioritize
400 potential decisions variables (i.e., pipe replication, PRV installation, tank installation, etc.). The decision
401 variables are ranked based on their operational savings per unit of capital cost expenditures with those
402 variables with the highest ratio being implemented. The system hydraulics and objective functions were
403 recalculated after each successive change to ensure feasibility and all intermediate solutions were used
404 to generate a trade-off curve. Finally, the quality of the Pareto-optimal curve generated using engineering
405 judgment, was compared to one created using a heuristic global search optimization algorithm. A
406 background leakage modelling methodology in EPANET was adopted for approximating the leak
407 assessment methodology provided by the competition organizers.

408 Saldarriaga et al. (2014, 2015) presented a methodology that used the Unit Headloss to select pipes to
409 rehabilitate, the Flow-Pressure concept to locate valves and GA for the pump optimization process. The
410 methodology is composed of different steps, starting from the application of a leakage model to the initial
411 network using EPANET model with emitters. The network was then sectorized according to DMA's
412 demand patterns and a rehabilitation process was conducted to meet pressure requirements. An

413 infrastructure optimization process was carried on allowing for improvements, such as installation of
414 new pipes, pumps and tanks, and a pump optimization was iteratively performed together with the
415 estimation of leakage parameters. Finally, the whole network improvement was considered to evaluate
416 the final cost of the proposed solution.

417 Matos et al. (2014, 2015) proposed an evolutionary approach that operates in an exclusively discrete
418 solution space and is intended to require as little engineering judgment and time as possible while
419 attaining acceptable and informative results that are useful for decision-making. Its main features are
420 custom crossover and mutation operators, being the latter guided by specific network and simulation
421 parameters. The developed operators, specific for water distribution network optimization tasks, are
422 applicable to single- and multiple-objective genetic algorithms as well as to other evolutionary
423 algorithms.

424 Thus, authors presented two implementations: the first consisted of a single-objective (i.e., minimization
425 of the total operational and capital cost) genetic algorithm whose mutation operator was designed to find
426 increasingly parsimonious solutions as the optimization unfolds. The second was a multiple-objective
427 approach: the objectives were the minimization of investment and operational costs. A simple post-
428 processing greedy algorithm to locally refine pipe replacements is also presented as a means of
429 complementing the evolutionary approach.

430 Computations have been carried out in a Java version of EPANET aiming at increased computational
431 efficiency, greater platform portability, and improved flexibility regarding optimization software.

432 Rahmani et al. (2014, 2015) presented a methodology based on a three-stage multi-objective optimization
433 model. At the first stage, the optimal design of pipeline rehabilitation, pump scheduling and tank sizing
434 is formulated and solved on the skeletonized network by optimizing the costs of pipes, upgrading of
435 pumps and tank and the cost of water losses and energy. The second stage employs the best Pareto front
436 obtained from the first stage to solve the previous two objectives optimization problem for the full

437 network. The third step employs a three-objective optimization model by adding the number of PRVs as
438 the third objective and PRV settings are also added to the decision variables. This stage employs three
439 solutions on the Pareto front of the second stage to seed the optimization on the full network.

440 The optimization model used in all stages is non-dominated sorting genetic algorithm (NSGA-II) and the
441 simulation model is the EPANET software tool.

442 Sousa et al. (2014, 2015) proposed two optimization models supported by engineering judgment to help
443 in choosing the best strategies to follow, starting with the optimization of the pump controls, followed
444 by the installation of PRVs and the replacement of existing pipes. The first optimization model used is a
445 least-cost design model to identify the pipes to be replaced and size them; the second is an optimal
446 operation model to define the pump controls and the PRV settings. Both models are solved by linking a
447 hydraulic simulation model (WaterNetGen - a pressure driven EPANET extension) with a simulated
448 annealing algorithm. The selection of final optimal solutions was done using engineering judgment.

449 Vassiljev et al. (2014, 2015) proposed an approach based on a trial-and-error methodology using heuristic
450 methods coupled with hydraulic simulation. To find the optimal solution, customized research tools were
451 developed for WDN optimization. These tools, based on the EPANET2 toolkit, were employed for the
452 optimization of water tanks levels to switch pumps on/off; the estimation of the influence of PRVs on
453 leakages to decide adding a PRV to a pipe or not; the calculation of leakages under different conditions.

454 Commercially available tools are also used carrying out comparison of various network structures
455 (parallel pipe alternatives). The analyses were carried out in four major stages: (a) the elimination of
456 bottlenecks (in terms of small pipe diameter and/or low pipe roughness coefficient C); (b) the installation
457 of PRVs to reduce the pressure at leak nodes; (c) the examination of pump efficiencies; and (d) the
458 optimization of water levels in tanks.

459 Finally, Shafiee et al. (2014, 2015) implemented a genetic algorithm approach within a high-performance
460 computing platform to select tank sizes, pump placement and operations, placement of pressure control

461 valves, and pipe diameters for replacing pipes. Multiple problem formulations are solved that use
462 alternative objective functions and allow varying degrees of freedom in the decision space. The original
463 framework is based on a genetic algorithm that was written in Java and calls functions from the EPANET
464 toolkit to simulate network hydraulics. The framework is implemented on a parallel cluster and was
465 modified for the BBLAWN application, incorporating additional functions from the EPANET toolkit for
466 manipulating pressure control valves and created new functions for calculating hydraulics based on
467 leakage across pipes.

468

469 **Discussion**

470 All the approaches proposed by teams brought interesting contributions to solving the complex
471 BBLAWN problem. The proposed strategies range from those strongly based on a multi-objective
472 optimization including all the conflicting cost objectives and the involved decision variables (pipes,
473 valves, pumps and tanks) proposed by the organizers (Morley and Tricarico ; Roshani and Filion), to the
474 approaches based on successive stages in which the engineering judgment has the main role, thus
475 resulting in a limited use of optimization procedures (Tolson and Khedr).

476 Most of the proposed methodologies are structured as multi-stage approaches combining it with the use
477 of engineering judgment/expertise, which has been aimed at reducing the size of the optimization
478 problem and driving towards the selection of intermediate and final solutions. The use of engineering
479 judgment is very important for the extension of the proposed approaches to real-network problems,
480 because it allows the inclusion of other types of knowledge and expertise in the technical and decision-
481 making process.

482 From the optimization standpoint, most of the teams implemented population based techniques (i.e.,
483 genetic algorithms) in a multi-objective setting, including, in different combinations, the conflicting cost
484 objectives proposed by the organizers. The only exceptions are Price and Ostfeld, who solved the

485 problem using Linear Programming, and the approach by Sousa et al. that implemented a simulated
486 annealing algorithm. Some other teams (Diao et al. , Saldarriaga et al. , Rahmani et al.), tried to reduce
487 the space of solutions of the “main” multi-objective optimization by means of network
488 clustering/sectorisation/skeletonization, thus dealing with a larger number of smaller (and simpler)
489 optimization problems.

490 From the computational point of view, all teams used the EPANET hydraulic solver with some of them
491 implemented a pressure-driven version in order to enhance the simulation of background leakages.
492 Interestingly, Matos et al. implemented a Java version of EPANET. Some teams, (Morley and Tricarico
493 , Roshani and Filion , Shafiee et al.) , have also made use of parallel processing in order to reduce the
494 computational time of their applications.

495 As reported by many teams, the adoption of the EPANET2 model, although well-known and used
496 worldwide, showed major limitations in dealing with the BBLAWN real problem. First, it required some
497 modifications/post-processing of results in order to consistently assess the background leakages from
498 pipes according to Eq. (3); otherwise the simulation is affected by errors as explained above. Second,
499 EPANET2 does not model pressure reduction valves controlled by remote set points (i.e., far from the
500 downstream PRV node). This limitation actually prevented all teams from using the remote control
501 option of valve that was allowed in BBLAWN rules. However, this is a preferred option due to control
502 solutions currently available to water utilities. Using remote controlled PRVs is likely to provide
503 solutions that are technically more reliable than “classical” PRVs. In fact the pressure at remote set point
504 (e.g., the critical node in the controlled area) better reflects the real network hydraulic behavior than the
505 one immediately downstream of the PRV. For example, in case of abnormal water demand (e.g. for
506 firefighting or during pipe burst events) resulting in pressure drop at the control node (which is usually
507 the most critical node due to node elevation and/or building height), the PRV opens to reach the set
508 pressure value. Vice versa, the set point of a “classical” PRV needs to be modulated over time based on

509 some prediction of network hydraulic behavior, which relies heavily on predicted demands and model
510 calibration (and related uncertainties).

511 In this regard, the solution proposed by Price and Ostfeld suggested that a more realistic problem
512 formulation, maybe in future “Battle” editions, could also include remote control of pumps and, also,
513 variable speed pumps.

514 Depending on the particular strategy adopted, the solutions presented different trade-offs between capital
515 (parallel pumps, tank enlargement, pipe renewal/doubling) and operational (energy, water losses) costs.

516 Table 3 summarizes the key decision variables. The solutions showing lower capital costs, are also those
517 requiring the highest operational costs. In fact, keeping the existing water infrastructures intact (i.e.
518 without any investment on asset renewal) is likely to result in large volume of water losses and pumping
519 energy requirements. On the other hand, a significant reduction in water losses can be achieved by
520 strategically investing in renewal of pipes, enlargement of tanks and/or new pumps. Some of the solutions
521 with the lowest capital costs are also those requiring implementation of the largest number of PRVs to
522 control as much as possible pressure through the network. Nonetheless, the need for providing water to
523 customers that satisfies the minimum pressure requirement, does not permit further reduction of leakages
524 via PRVs only.

525 Such a variety of solutions further demonstrates the need for engineering judgment as well as the
526 knowledge of water utilities’ management strategies to take effective and sustainable decisions in such a
527 complex multi-objective problem encountered in a real networks.

528 Table 3. Comparison among different BBLAWN approaches

529

530 **Conclusions**

531 The Battle of Background Leakage Assessment for Water Networks (BBLAWN) was designed to follow
532 the tradition of the “battle” competitions” held during the Water Distribution Systems Analysis (WDSA)

533 Conferences. The BBLAWN problem was about the optimal management of water and energy resources,
534 as relevant environmental and socio-economic issue worldwide. The competition considered asset
535 renewal planning and strengthening, as well as optimal operation, including possible installation of
536 PRVs. All the participant teams performed well in the competition, producing interesting results and
537 some innovative ideas worthy of future exploration. Most of the proposed methodologies were able to
538 suggest sensible solutions in both short time (operational) and medium time (tactical) horizons.

539

540 The review of all contributions clearly shows how challenging the BBLAWN problem is from
541 engineering perspective since it involves a real size network where multiple conflicting objectives need
542 to be considered and realistic technical constraints accounted for. Management decisions can and should
543 be supported by tools that combine hydraulic models capable of assessing pressure-dependent
544 background leakages with computationally effective multi-objective optimization strategies. In order to
545 promote the discussion inside the technical/scientific community, the rules of BBLAWN did not compel
546 the use of any specific software for hydraulic modeling and only provided the management objectives to
547 be fulfilled.

548 Due to the number of decision variables and the size of the search space, the WDN design process cannot
549 be fully automated. Engineering judgment can and should provide invaluable support to the formal
550 optimization approaches in the search for feasible alternative solutions. A multi-step approach was
551 preferred by most of the teams since it permits the progressive evaluation of the improvements in WDN
552 performance achievable at each step. The overview of proposed solutions demonstrated that many
553 alternatives are compatible with the problem in hand, ranging from massive network renewal (at lower
554 operational cost) to minimal interventions (requiring high cost for energy and pumping). If the same
555 approach was adopted for real life applications, the selection of the optimal strategy and of the most
556 effective solution, should take into account the possibility of planning different interventions over time,

557 thus reflecting the budget available. This would make preferable, for example, in the short term horizon
558 the optimal control of pumps rather than more expensive renewal of asset.

559 The overview of the proposed strategies also emphasized the need to overcome current limitations of
560 WDN simulation models in order to permit more realistic assessment of background leakages as well as
561 the modelling of remotely controlled devices. This would permit more reliable simulations to support
562 WDN management, allowing also the assessment of the impact of effective ICT solution for WDN
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564

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743 Table 1. Ranking of the PRVs

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745 Table 3. Comparison among different BBLAWN approaches

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Table 1. Ranking of the PRVs

Pipe ID	Pipe ID of PRV	Water Lost [m ³]	Leakage [%]	Energy [KWh]
	original	36,281	26.05	42,221
	solution 5	34,533	25.11	42,063
P122	48	32,140	23.79	41,315
P758	276	30,639	22.93	40,870
P789	299	29,458	22.24	40,365
P5	234	28,637	21.76	39,849
P305	163	27,976	21.36	39,346
P1000	441	27,395	21.01	39,379
P115	40	26,807	20.66	39,202
P1033	20	26,342	20.37	38,956
P125	51	26,049	20.19	38,898
P1002	443	25,794	20.03	38,679
P937	368	25,575	19.90	38,628
P786	296	25,240	19.69	38,548
P16	79	24,943	19.50	38,539
P772	286	24,801	19.41	38,418
P794	301	24,580	19.27	38,401
P72	267	24,370	19,.14	38,365
P344	187	24,170	19.01	38,106
P1001	442	24,075	18.95	38,004
P329	175	23,915	18.85	37,668
P1042	28	23,852	18.81	37,667
P633	255	23,823	18.79	37,695
P781	292	23,696	18.71	37,490
P1024	10	23,632	18.67	37,489
P811	316	23,583	18.63	37,474
P10	2	23,531	18.60	37,430

750 Table 2. Relevant data of the initial and final status of the network. Operational costs are weekly-based.

Solution	Water Loss [m ³]	Leakages [%]	Energy [KWh]	Operational cost [€]	Capital cost [€]	PRVs Cost [€]
initial	36,281	26.05	42,221	77,738	0	0
Final	23,531	18.60	37,430	51,500	62,305	26,182

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Table 3. Comparison among different BBLAWN approaches

Solution by:	Hydraulic simulation and leakage model <i>Complies with Eq.(3)</i>	Strategy approach	PRV [-]	new/parallel pipes [-]	new/parallel pumps [-]	Enlarged tanks [m ³]
Morley and Tricarico (2014)	A pressure-driven version of EPANET, incorporating the leakage model. <i>Yes</i>	NSGA-II and Omni-Optimizer were used to run to completion on the full-scale multi-objective (MO) optimization.	6	373	3	-
Roshani and Filion (2014)	EPANET2; emitters at nodes to simulate leakages; orifice discharge coefficient reflecting the leakage model coefficient. <i>No</i>	Use of NSGAI to optimize (i) the capital cost and (ii) operational costs. Decision variables identification based on technical considerations and constraints (e.g. only pipes with a diameter of 300 mm or greater are considered for possible duplication).	14	409	8	T3: +500 T4: +1000 T7: +1000
Iglesias-Rey et al. (2014)	EPANET2; emitters at nodes to simulate leakages; dummy nodes close to tanks to allow leakage calculation. <i>No</i>	Single Objective (SO) Optimization strategy based on technical considerations (e.g. only pipes and pumps as decision variables after a cost analysis; sub-areas optimized separately; analysis of minimum pressure to change a pipe) Post-processing to improve the solution through fine adjustments based on best management practices	61	416	2	-
Creaco et al. (2014)	EPANET2; emitters at nodes to simulate leakages; second refinement to make leakage simulation compliant with the BBLAWN rules <i>Yes</i>	Three successive optimization considering 2 objectives for each optimization; final refinement of PCV setting to reduce leakage; no parallel pipes were allowed, based on cost analysis; parallel pumps location based on engineering judgement.	44	256	-	T5: +500
Diao et al. (2014)	EPANET2; emitter nodes to simulate leakages; iterative update of nodal demand using simulated pressure. <i>No</i>	Step-by-step optimization approach, based on hierarchical initial classification into trunk clusters and branch clusters. Next, optimal pressure control strategies for each cluster is addressed following the hierarchical sequence.	61	349	2	-
Eck et al. (2014)	Modified EPANET2 code; emitter coefficients at nodes are updated through iterations to simulate leakages. <i>No</i>	Sequential assessment of intervention types. An optimization technique is developed and applied sequentially to yield a list of suggested improvements for the network. Optimization solvers based on Bonmin and Ipopt techniques to solve mixed integer nonlinear programming problems.	22	345	2	-
Tolson and Khedr (2014)	EPANET2; emitters at nodes to simulate leakages; dummy node added to simulate leakages close to tanks.	Bi-objective optimization: maximizes the operational savings and minimizes the total capital costs. Engineering judgment and cost analysis is heavily relied upon to identify candidate and priority decision variables (i.e., PRV valve configurations).	27	23	-	T2: +500

	<i>No</i>	The design is fine-tuned to ensure pressure and tank level constraints.				
Saldarriaga et al. (2014)	EPANET2; leakages along pipes are simulated with emitters at pipe downstream nodes only using an approximated emitter coefficient.	The Unit Headloss concept supports rehabilitation interventions; the Flow-Pressure concept support the location of valves; GA support the pump optimization process. Sequential approach: leakage parameters estimation; sectorization; rehabilitation; PRV location; pump optimization with GA; Union of all DMAs; final leakage parameters estimation; final pump optimization; final cost evaluation.	66	12	-	-
	<i>No</i>					
Matos et al. (2014)	Java version of EPANET2.	Preliminary engineering analysis; MO-Optimization: minimization of investment and of operational costs. A modified version of the NSGA-II was employed in order approximate the Pareto front.	12	203	2	T1: +5000 T4: +5000
	<i>No</i>					
Rahmani and Behzadian (2014)	EPANET2 to get approximate solutions; a posteriori application of the Geranopoulos' leakage model.	Three-stage multi-objective optimization model (NSGAI). First stage: optimal design of pipeline rehabilitation, pump scheduling and tank sizing using a skeletonized WDS model; minimizing capital and operational costs. Second stage: the same optimization using the full network. Third stage: the same optimization including 168 hr simulation.	2	270	1	-
	<i>Yes</i>					
Sousa et al. (2014)	WaterNetGen permitting to compute burst and background leakages, consistently with BBLAWN rules.	The methodology comprises two optimization models: a least cost design model to identify the size of pipes to be replaced and optimal pump controls and the PRV settings. A simulated annealing algorithm was used to solve the optimal WDN design and operation problem.	41	429	0	0
	<i>Yes</i>					
Vassiljev et al. (2014)	Customized tool based on EPANET2 toolkit to simulate background leakages.	Use of customized research tools developed for tank parameter optimization (optimal volume); estimate the profit of each PRV and cost of exploitation. Four stages analysis: (i) elimination of bottlenecks, (ii) installation of PRVs, (iii) examination of pump efficiencies, (iv) optimization of tanks (pump on/off levels and tank diameter).	80	5	2	0
	<i>Yes</i>					
Shafiee et al. (2014)	Modified EPANET toolkit by creating new functions for calculating hydraulics based on background leakage across pipes.	SO-Optimization using a GA to select tank sizes, pump placement and operations, PRV locations, and new pipe diameters. Multiple problem formulations are solved that use alternative objective functions and allow varying degrees of freedom in the decision space.	28	29	0	T4: +1000
	<i>Yes</i>					

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758 pumps (PM), a control valve (CV), a check valve (CH).

759 Fig. 2. TOWN-C pressure control valve (PRV) and node of pressure set (Pset).

760 Fig. 3. Pareto front of solutions for the multi-objective optimization problem (pipe and pump cost vs.
761 energy and water loss cost vs. tank enlarging cost).

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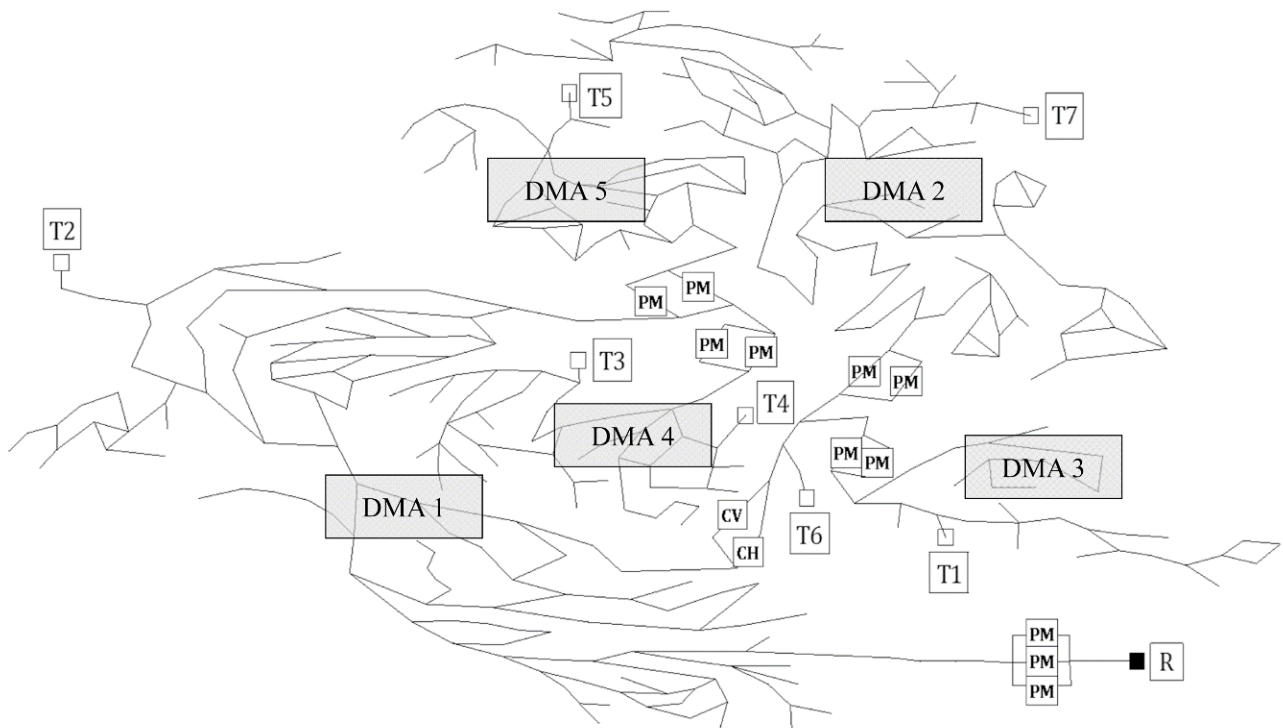
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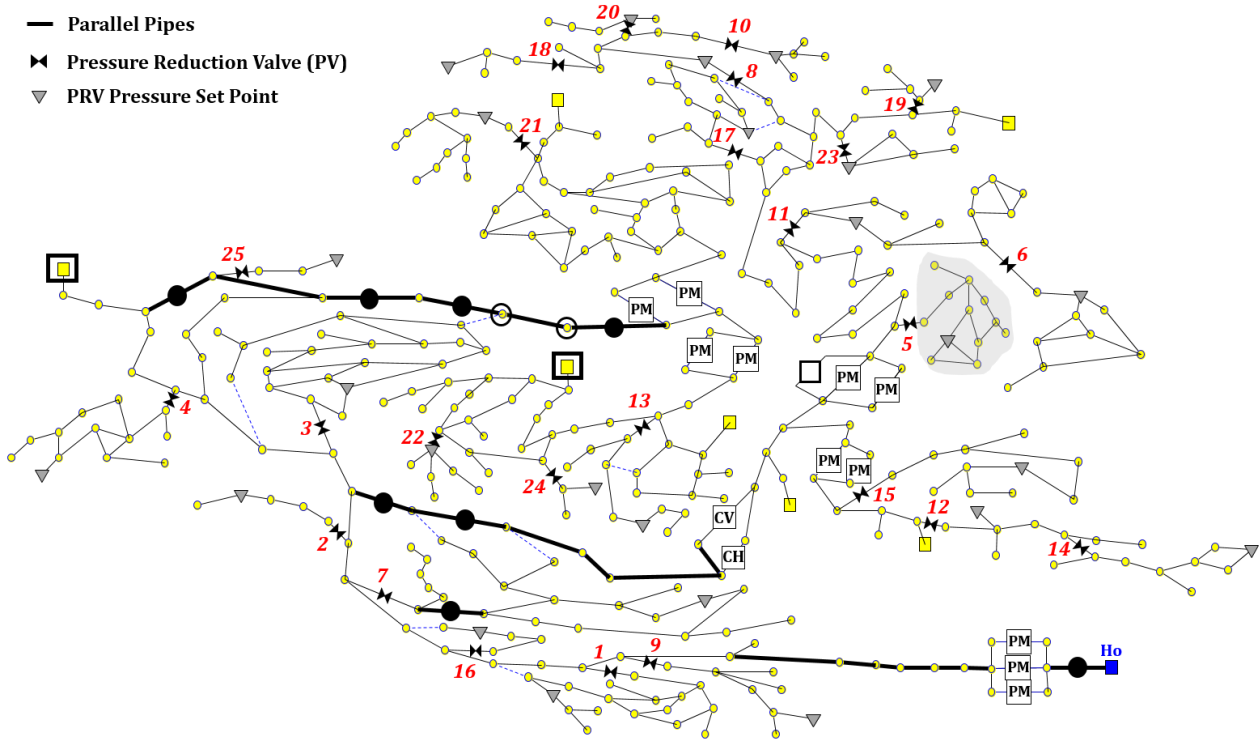
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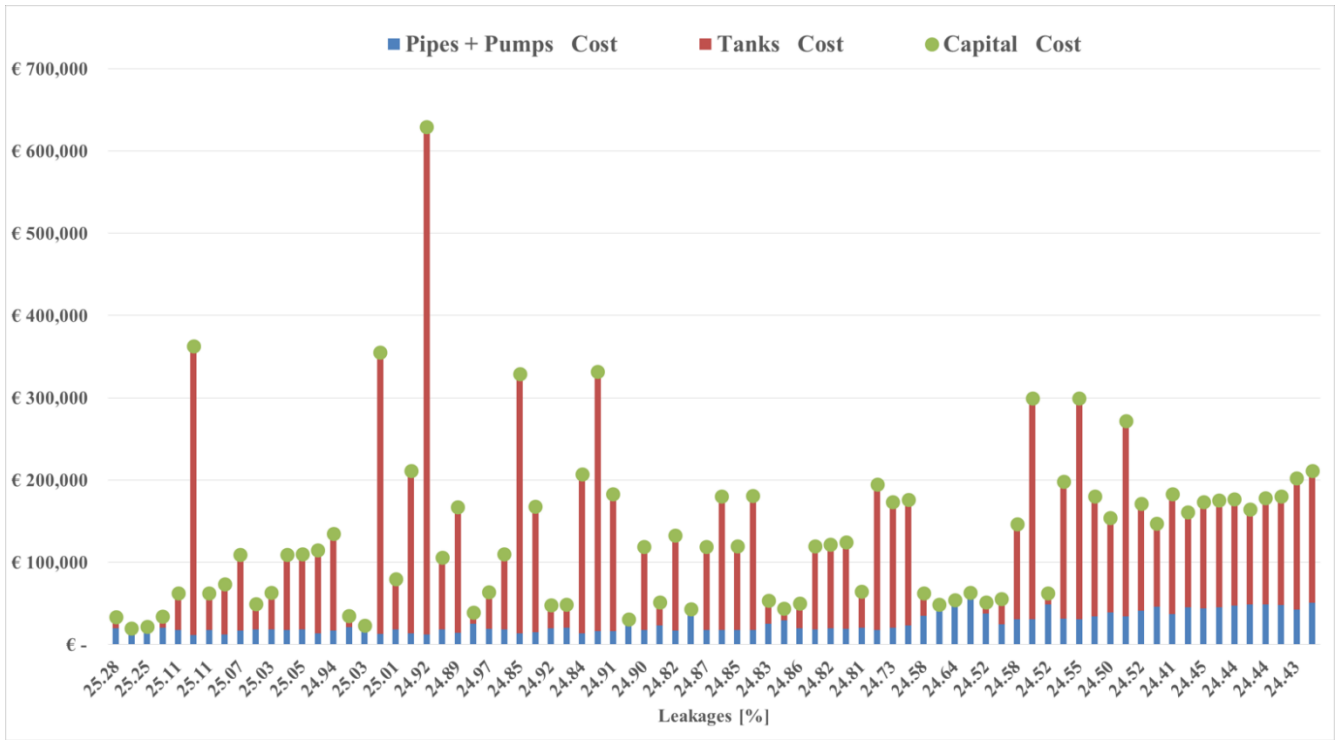
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