

1 **Multiple Linear Regression Analysis of Pumps Performance in Water**
2 **Pumping Plants**

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9
10 **Abstract.** Energy use in water supply systems represents a consequent part of
11 global energy consumption across all sectors. This consumption is expected to rise,
12 due to the increasing demand and the recourse to unconventional water resources.
13 Regarding water utilities, most of their operating costs are related to energy
14 consumption, especially pumping systems consumption. In this context, the main
15 objective of this study was to model accurately the energy consumption of pumping
16 systems in order to optimize the whole water supply system, thus improving its
17 efficiency, especially in the case of a limited renovation. For this purpose, Multiple
18 Linear Regression was fitted to model the produced kWh/m³ ratio according to the
19 following parameters, active and reactive energies, the daily produced water
20 volume, the power factor (Cosφ), and the operating time of each pump. The final
21 model describes accurately the consumption per cubic meter produced (R²=0.91).
22 Therefore, this model could be considered a good estimator for the calculated ratio,
23 which was close to the experimental one. In addition, this approach considers the
24 system's behavior while most of the comparable studies focus on the pump
25 scheduling problem estimator for the calculated ratio which was close to the
26 experimental one. In addition, this approach considers the real-time-data behavior
27 while most of the comparable studies focus on the pump scheduling problem.

28
29 **Key Words:** energy efficiency, linear multiple regression, pumping systems, water
30 Supply.

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36 **1. Introduction**

37 In the upcoming decades, water demand is forecasted to significantly increase up to
38 35% in 2050 (UN-Water 2020). Therefore, energy needs that are related to the water
39 pumping and distribution processes could contribute to the expected increase in
40 energy demand (EIA 2019). Indeed, several studies have concluded that the energy
41 consumption related to these processes accounts for 7% of the total energy used
42 across the globe (Coelho et al., 2014). His share is expected to get bigger due to the
43 increasing distances between resources and populations, especially in water-scarce
44 countries, and the growing consumption per capita due to the industrialization and
45 improvement of living standards.

46 To reach the sixth Sustainable Development Goal (SDG6) that aims to
47 generalize the access to drinking water supply, the water production cost mustn't
48 impact its price which should stay affordable to the population. Therefore,
49 production costs should be reduced via the control of energy consumption which is
50 typically the largest marginal cost for the production (Helena et al., 2017).

51 Pumping processes consume the largest fraction of total energy (Plappally and
52 Lienhard, 2012). The pumps consumption often presents 80% to 90% of the total
53 energy consumption (Sarbu, 2016). However, this consumption may depend on
54 many factors such as surface water or ground water, transport differences, flat or
55 mountain regions, etc (Rothausen and Conway, 2011) (Plappally and Lienhard,
56 2012). Thus, by achieving energy efficiency improvements measures, we may
57 reduce this consumption by 25% (Moreira, et al., 2013). In this context, few studies
58 were interested in modeling pumping systems and evaluated the influence of
59 parameters such as the aging of the components, which can reduce the performance
60 of the pump by up to 12% (Durmus, et al., 2008).

61 Many factors impact the overall energy efficiency of a water pumping system.
62 One important factor is the operating point which should be as close as possible to
63 the maximum energy efficiency region (Carravetta, et al., 2020). Another factor is
64 the age of the system, especially the pumps which can lose 1.5% of their efficiency
65 every year (Zhou et al., 2021). Many attempts have been made to address the subject

66 of pumping optimization using different approaches to optimize either energy
67 consumption or the overall cost of energy. To optimize the energy consumption, a
68 model based on ant colony was utilized (Ostfeld and Tabultzev, 2008) to provide an
69 effective model that minimizes the design and the operating costs. This model act on
70 parameters such as: the diameters of the pipes, the pumping station's maximum
71 power, and the tank's storage. Another work targeting the minimization of energy
72 consumption (Giustolisi O et al., 2012) used a multiobjective strategy to minimize
73 losses by controlling pressure in the water distribution system and thus minimizing
74 the energy consumption.

75 On the other hand, various studies have focused on optimizing the operation of
76 the existing water systems by using deterministic methods such as Dynamic
77 Programming (Fallside and Perry, 1975) which worked on minimizing the energy
78 cost by load shifting using a hierarchical optimization to reduce the problem to 10
79 control parameters. Given the difficulty of the deterministic approach, different
80 metaheuristic methods were developed, the widely used among them is Genetic
81 Algorithms (Wu et al., 2014) which used evolutionary computation to maximize the
82 efficiency of a system consisting of 2 pumps. Although these efforts employ
83 evolutionary optimization techniques. However, in literature, particular attention has
84 been paid to the applicability of linear regression (Puleo et al., 2014) and multiple
85 linear regression (Adamowski et al., 2012) to the pumping scheduling optimization
86 problem.

87 This work targets the pumping system of the Bab Louta drinking water
88 production located in the province of Taza, Morocco. In this perspective, a multiple
89 linear regression (MLR) was fitted to model the produced kWh/m³ ratio according to
90 the input parameters by using real-Time-data.

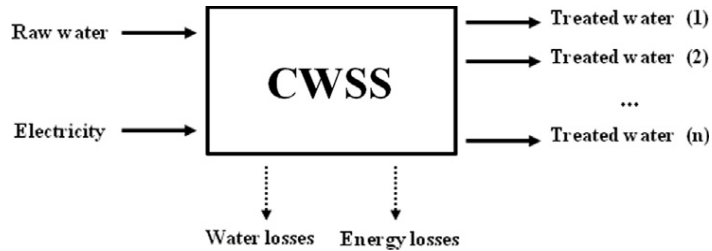
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92 **2. Materials and methods**

93 Conventional water supply systems (WSS) consist of sets of structures and
94 facilities providing products with a suitable quantity and quality for domestic and

95 industrial use (Luna et al., 2019). To develop energy and hydraulic model, WSS
96 systems should be evaluated in terms of mass and energy (Figure 1).

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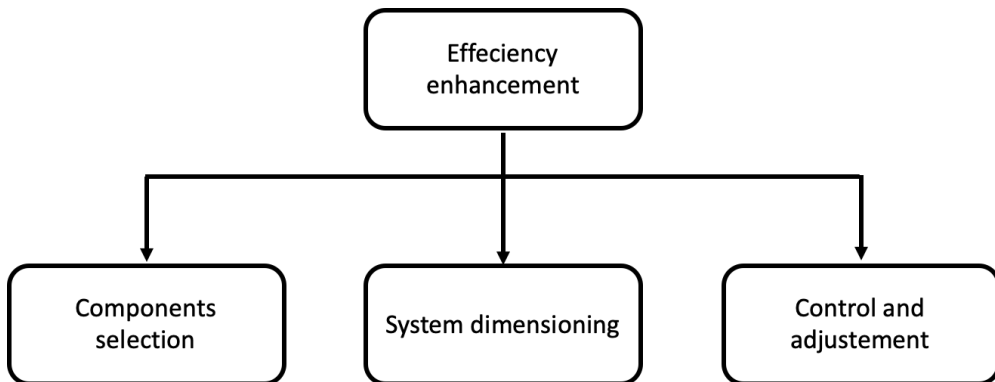


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99 **Figure 1:** Energy and Hydraulic flows in a WSS (Vilanova, et al., 2014).

100 There are various methods to enhance energy efficiency in WSS. These
101 methods are ranging from leakages control through simple monitoring operations to
102 massive investments by reviewing the design of the infrastructures, equipment
103 upgrade, pump system optimization, and real-time control. These Methods were
104 classified into 3 major sub-categories in Figure 2 (Shankar, et al., 2016).

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107 **Figure 2:** Efficiency enhancement opportunities in WSS.

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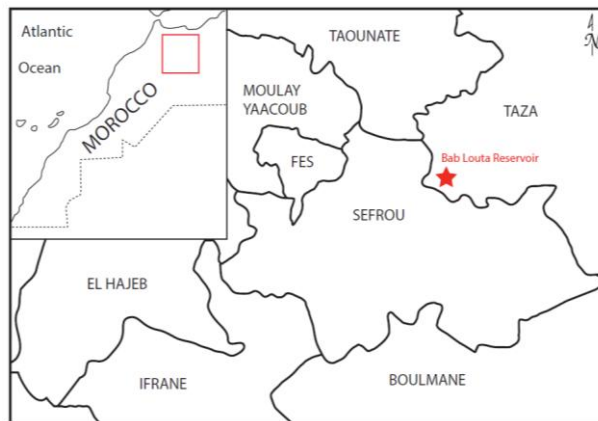
109 2.1 Study area and description of the drinking water production system

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111 The province of Taza is located in the center North of Morocco and it is one of
112 the 9 provinces of the region of Fez-Meknes with a population of roughly 530 000
113 inhabitants (Morocco 2014). The water treatment plant of Tahla provides a large
114 population of the province, mainly the urban areas. It is situated 60 km from the city
115 of Taza. The plant treats the raw water of the Bab Louta reservoir.

116 The production system consists of a pumping station SP0 for raw water, a water
117 treatment plant then another pumping station of SP3 to reach the city of Taza, which
118 is the chief town of the province.

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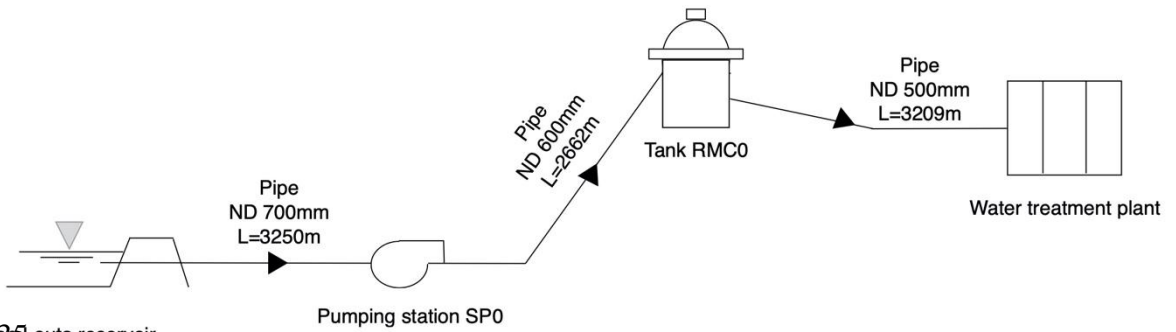
121 **Figure 3:** Location of the Bab Louta reservoir.

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123 2.2 Description of the pumping station

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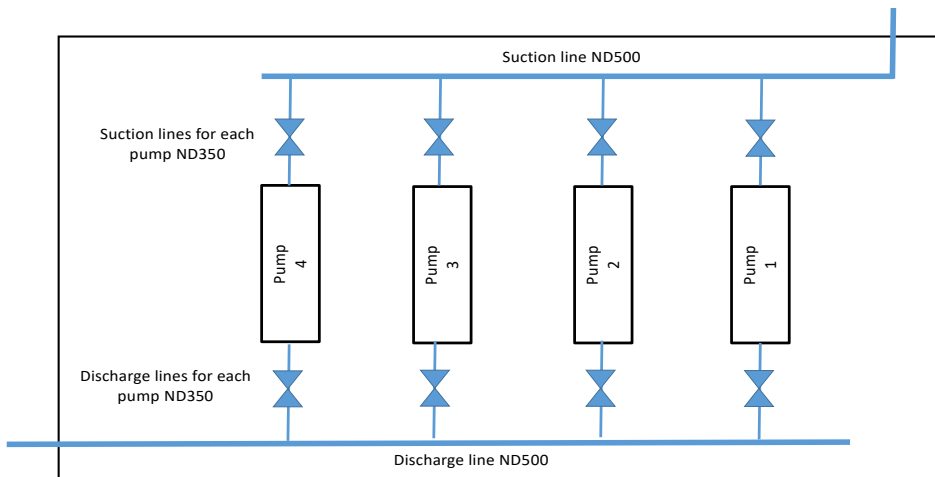
125 The pumping station SP0 is located about 3km from the reservoir. It is
126 responsible for overcoming the difference in altitude between the raw water intake
127 and the water treatment plant. The water is taken by gravitation from the reservoir to
128 the pumping station SP0 through a pipe of a nominal diameter (ND) of 700 mm
129 which converge to a diameter of 500 mm just before reaching the SP0. Then, the
130 water is pumped through a pipe of a nominal diameter of 600 mm to a tank destined
131 to provide the water treatment plant with raw water, it is called RMC0 and having a
132 capacity of 1000 m³, which provides the water treatment plant with raw water
133 through a pipe of a nominal diameter of 500mm (ONEE 2019).



135 Bab Louta reservoir
 136 **Figure 4:** Location of the Bab Louta reservoir.

137 The pumping station, the object of the current study, is situated at an altitude of
 138 498.20 m and tank RMC0 is situated at an altitude of 737.80m. SP0 consists of four
 139 pumps (Figure 5) with a flow of 457 m³/h each and a manometric pressure of 209 m.
 140 They are manufactured by the company HIDROTECAR powered by an electric
 141 motor which is manufactured by LEROY SOMER with a nominal power of 455 kW
 142 (ONEE, 2019).

143



144
 145 **Figure 5:** Hydromechanical scheme of the pumping station SP0.

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 147 2.3 Experimental procedure

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149 To obtain a reliable model of our system, measurements of the key parameters
150 influencing this system were conducted. These measurements were taken in the
151 period between January 2015 and December 2018 daily for the parameters:

- 152 • P: the active energy consumed by the pumping station (measured by a
153 wattmeter),
- 154 • Q: the reactive energy consumed by the pumping station (measured by a
155 wattmeter),
- 156 • V: the daily produced volume (measured by an electromagnetic
157 flowmeter),
- 158 • $\text{Cos}\varphi$: the power factor,
- 159 • HMG1: the pump operating time “1”,
- 160 • HMG2: the pump operating time “2”,
- 161 • HMG3: the pump operating time “3”,
- 162 • HMG4: the pump operating time “4”.

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164 2.4 Modelling

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166 This study aimed to use Multiple Linear Regression analysis to predict an output
167 from a range of inputs. MLR model with multiple input variables can be expressed
168 as follows (Longo et al., 2016):

169

$$170 \quad Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_n X_n \quad \text{Eq 1}$$

171

172 where :

- 173 • Y is the output variable;
- 174 • β_i are the regression parameters;
- 175 • X_i are the input variables.

176

177 In order to assess the influence of the included parameters on the cubic meters
178 ratio produced by the pumping station, the following parameters were considered:

179 the active energy (EP), the reactive energy (EQ), the daily produced volume (V), the
180 power factor (Cosφ), and the operating time of each pump (HMGi).

181 The effect of eight variables on the produced Kwh/m³ ratio was evaluated (Table
182 1). Of note, 1388 experiments were conducted during 4 years.

183

184 **Table 1:** Problem characteristics

| Objective of the study | The effect of the variables on the ratio KWh/m ³ |
|----------------------------|--|
| Number of Variables | 8 |
| Number of experiments | 1388 |
| Number of the coefficients | 8 |
| Number of responses | 1 |

185

186 **Table 2:** Measurements Summary

| | E _P (kWh) | E _Q (kVar) | Cosφ | Production (m ³) | HMG1 (m) | HMG2 (m) | HMG3 (m) | HMG4 (m) | Ratio (kWh/m ³) |
|--------------------|----------------------|-----------------------|------|------------------------------|----------|----------|----------|----------|-----------------------------|
| Mean | 14 837.47 | 8 339.58 | 0.87 | 18 181.40 | 7.95 | 7.63 | 10.43 | 9.01 | 0.82 |
| Standard deviation | 9 158.63 | 8 529.85 | 0.05 | 5 081.13 | 8.90 | 5.88 | 11.51 | 8.85 | 0.43 |

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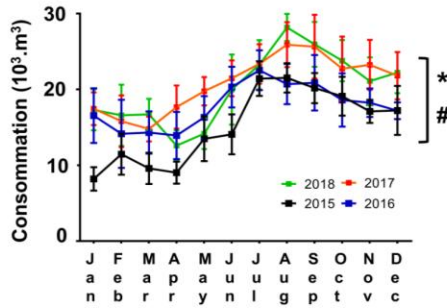
188 In Table 2 there is a preview of descriptive statistics of the eight parameters, the
189 mean and the standard deviation.

190 In the graphs below the distribution of the parameters is represented.

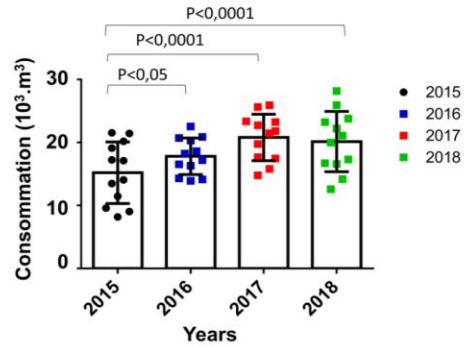
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A



B



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194 **Figure 6:** Representation the dataset collected of the variation of the consumptions through
 195 the months during period 2015 and 2018, (A) consumption variation through the months (*:
 196 $p < 0.05$, consumption variation through the years; #: $p < 0.05$, consumption variation through
 197 the months), (B) consumption variation through the years ($P < 0.05$ is considered as
 198 statistically significant. Each point represents the consumption average of a given month
 199 during the year).

200

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202 The dataset collected of the water production across the year for each year from
 203 2015 to 2018 presented in figure 6-A, and it the trend shows that clearly
 204 demonstrates a higher production which reflects a higher consumption during the
 205 summer months in province of Taza. On the other hand, figure 6-B shows the
 206 evolution of the production through the years 2015 to 2018 and it has also the same
 207 increasing trend due to the continuous commissioning of new networks leading to a
 208 growing number of consumers.

208

209 3. Results and discussion

210 3.1 Statistical interpretation

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From the correlation matrix (Table 3), it was found that the studied variables are strongly correlated:

- 214 • The active energy consumed by the pumping station was dependent on the
215 production, reactive energy, pumps operating hours (4;1), power factor
216 ($\text{Cos}\phi$), HMG2, and HMG3 respectively.
- 217 • The reactive energy consumed is highly was dependents on the production,
218 active energy, $\text{Cos}\phi$, pumps operating hours (4;1;2;3) respectively.
- 219 • $\text{Cos}\phi$ was dependent on the reactive energy, the production, active energy,
220 pumps operating hours (4;1;3;2) respectively.
- 221 • Production was dependent on the active energy, the reactive energy pumps
222 operating hours (4;1;2;3).

Table 3: Correlation Matrice

| Variable | Correlations (MATRICE CR without outliers) Marked correlations are significant at $p < 0,05000$ N=1388 (Case wise deletion of missing data) | | | | | | | | | | |
|--------------|---|-----------|---------------|-----------|-----------|-----------|-----------|-----------|-----------|---------------|---------------|
| | Means | Std. Dev. | P | Q | Cos phi | Prod | HMG1 | HMG2 | HMG3 | HMG4 | Ratio |
| P | -0.022423 | 0.487229 | 1.000000 | 0.883071 | -0.190305 | 0.939746 | 0.333340 | 0.202424 | 0.169873 | 0.625872 | 0.463030 |
| Q | -0.027636 | 0.345041 | 0.883071 | 1.000000 | -0.585775 | 0.883506 | 0.332525 | 0.170169 | 0.170505 | 0.612964 | 0.298339 |
| Cosφ | 0.026763 | 0.761882 | - 0.190305 | -0.585775 | 1.000000 | -0.316528 | -0.197066 | 0.024943 | -0.076562 | -0.271961 | 0.193819 |
| prod | 0.000741 | 0.973866 | 0.939746 | 0.883506 | -0.316528 | 1.000000 | 0.340309 | 0.222533 | 0.175107 | 0.613597 | 0.150594 |
| HMG1 | -0.001400 | 1.002171 | 0.333340 | 0.332525 | -0.197066 | 0.340309 | 1.000000 | -0.164919 | -0.016355 | 0.236341 | 0.111733 |
| HMG2 | -0.001865 | 0.998758 | 0.202424 | 0.170169 | 0.024943 | 0.222533 | -0.164919 | 1.000000 | -0.086627 | -0.141591 | 0.006575 |
| HMG3 | 0.001362 | 1.003257 | 0.169873 | 0.170505 | -0.076562 | 0.175107 | -0.016355 | -0.086627 | 1.000000 | 0.074751 | 0.036371 |
| HMG4 | -0.017745 | 0.708400 | 0.625872 | 0.612964 | -0.271961 | 0.613597 | 0.236341 | -0.141591 | 0.074751 | 1.000000 | 0.025731 5 |
| Ratio | -0.051987 | 0.197427 | 0.463030 | 0.298339 | 0.193819 | 0.150594 | 0.111733 | 0.006575 | 0.036371 | 0.025731 5 | 1.000000 |

224 From the regression table (Table 4), it was concluded that the factors
225 influencing the kWh/m³ ratio in descending order are:

- 226 • Ratio is positively correlated with the active energy consumed by the
227 pumps;
- 228 • Ratio is negatively correlated with the production;
- 229 • Ratio is positively correlated with Cosφ;
- 230 • Ratio is negatively correlated with the reactive energy consumed by the
231 pumps;
- 232 • Ratio is positively correlated with the operating hours of pumps 1 and 4.

233

234 **Table 4:** Regression summary for dependent variable

| N=1388 | b* | Std. Err. of b* | B | Std. Err. of b | t (1379) |
|------------------|-----------|------------------------|-----------|-----------------------|-----------------|
| Intercept | | | -0.026073 | 0.001608 | -16.2112 |
| Prod | -2.52837 | 0.026164 | -0.512563 | 0.005304 | -96.6366 |
| HMG ₁ | 0.02116 | 0.008985 | 0.004168 | 0.001770 | 2.3546 |
| HMG ₂ | 0.01693 | 0.009327 | 0.003346 | 0.001844 | 1.8150 |
| HMG ₃ | -0.00080 | 0.008299 | -0.000157 | 0.001633 | -0.0960 |
| HMG ₄ | 0.04287 | 0.011275 | 0.011948 | 0.003142 | 3.8023 |
| P | 2.85701 | 0.043662 | 1.157669 | 0.017692 | 65.4347 |
| Q | -0.08315 | 0.040932 | -0.047577 | 0.023421 | -2.0314 |
| Cosφ | 0.09614 | 0.020445 | -0.024913 | 0.005298 | -4.7023 |

235

236 From the variance analysis (Table 5):

- 237 • Regression mean squares was superior to the residual mean squares;
- 238 • The Fischer value is very high;
- 239 • P-value is strictly inferior to 0.05.

240 These results validated our variance analysis.

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246 **Table 5:** Analysis of Variance Table

| Effect | Analysis of Variance; DV: Ratio | | | | |
|-----------------|---------------------------------|------|--------------|----------|---------|
| | Sums of Squares | df | Mean Squares | F | p-value |
| Regress. | 49.31283 | 8 | 6.164104 | 1790.044 | 0.00 |
| Residual | 4.74865 | 1379 | 0.003444 | | |
| Total | 54.06149 | | | | |

247

248 Multiple Linear Regression has shown that most of the experimental results are
 249 highly adjusted and the model is explanatory, considering the value of R^2 and the
 250 value standard error estimate which is very low (table 6).

251

252 **Table 6:** Summary statistics

| Statistic | Value |
|----------------------------|--------------|
| R | 0.95507172 |
| R^2 | 0.91216198 |
| Adjusted R^2 | 0.9116524 |
| Fischer (8.1379) | 1790 |
| P | 0 |
| Standard error of estimate | 0.0586817579 |

253

254 The final model is expressed by the equation below:

255

$$R_{\text{ratio}} = 2.85701 P - 2.52837 \text{ Prod} + 0.09614 Q - 0.08315 \text{ Cos}\varphi + 0.04287 \text{ HMG1} + 0.02116 \text{ HMG2.} \quad \text{Eq 2}$$

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3.2 Technical Interpretations

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From this analysis, this study concluded that in order to improve the ratio per produced cubic meter, a reduction of the active energy consumed by the pumps should be considered. As the function point of the pumps is already set by the system characteristics, only active energy could be reduced by the improvement of the pumps efficiency.

266

267

The ratio is negatively correlated with production means; i.e. there is an economy of scale. It means the most the production increases the least the ratio is.

268

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The operating hours of the pumps 1 and 4 are positively correlated, which means that more these pumps are used higher more the ratio is higher. Therefore, it

270 is recommended to use the pump 3, and if there is an operation of renovation of the
271 pumping station, it is recommended to start with upgrading the pumps 1 and 4. In
272 the event of a new investment, the company can increase the capacity of the RMC0
273 storage tank which, according to the model, will decrease the significantly the ration
274 and also allows a load shifting to the off peak hours

275 The model which was elaborated in this study has a standard error of estimate
276 of 0.05 and due to the lack of previous studies using multiple linear regression, we
277 compared the results with a previous study involving five data-mining approaches
278 (Kusiak et al., 2013). This study had for objective to model the energy consumption
279 in a comparable case of a wastewater pumping station that has 3 pumps that transfer
280 the wastewater to a treatment plant. Although there are differences between the flow
281 capacities and the pressure with drinking water supply facilities but the approach
282 remains the same. The five data mining approaches are the multi-layer, perceptron,
283 neural network (MLP), the boosted-tree (regression) algorithm (BT), the random-
284 forest algorithm (RF), the support-vector machine (SVM), and the k-nearest
285 neighbor algorithm. These approaches had all provided more than 90% of accuracy
286 which is the case in the model of this study. The benefit of our method goes beyond
287 the control methods used in most of the optimization approaches which only
288 provide a method to operate the system in an efficient way but don't account for
289 other factors such as the aging of the pumps, factors that are crucial when upgrading
290 the system.

291 **4. Conclusions**

292 A Linear multiple regression was conducted to assess and study the influence of
293 multiple parameters on the energy consumption ratio per cubic meter water,
294 involved in a water pumping station.

295 This unique approach has allowed evaluating the real response of the system
296 relying on data that is measured over a 4 years period. Modelling the ratio will be a
297 tool to take decisions on which pump should the work be done first. This method
298 combined with a cash flow analysis, can help to take decisions on establishing
299 priorities in case of renovations, to change the pumps 1 and 4 with more efficient

300 pumps. The model would be useful when the next renovation will be undertaken by
301 the office which will conduct a replacement of the pumps in the year 2024.

302 The above findings demonstrate the potential of method for solving real-time
303 pump scheduling problems in large water distribution systems with many pumps.
304 This, however, requires further work with other metaheuristic methods such as
305 Genetic Algorithms before relevant conclusion can be made.

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307 **Data availability.** The source data used for the illustrations of the cases are
308 available upon request.

309 **Conflict of interest.** All authors declare no conflicts of interest in this paper.

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