

1 **Estimating the Productivity of the Bosnian-**
2 **Herzegovinian Water Operators**

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7 **Abstract.** This paper deals with the issue of productivity of municipal water
8 operators in Bosnia-Herzegovina (BIH). It demonstrates that Stochastic
9 Frontier Analysis (SFA) could be a useful tool in assessing productivity and the
10 relative efficiencies of water operators.

11 The principal aim of this paper is to show that it is worth assessing the
12 productivity improvements that could result from better use of inputs,
13 primarily labor. In spite of severe scarcity of data for the BIH water sector,
14 collected data allowed us to develop a SFA model with a set of inputs (number
15 of connections, number of workers, electrical energy costs and chemical
16 costs) and outputs (water delivered) as required in the water sector
17 empirical literature.

18 The research could serve as a benchmark against which future quantitative
19 analysis of water operators' productivity can be measured. It could
20 additionally provide policy-makers with comparable quantitative evidence on
21 the functioning of water operators with the aim of regulating them more
22 effectively and improving the performances of one of the most poorly
23 functioning water sectors in Europe.

24 **Keywords:** Water Operators, Productivity, Stochastic Frontier Analysis (SFA),

25 **1 Introduction**

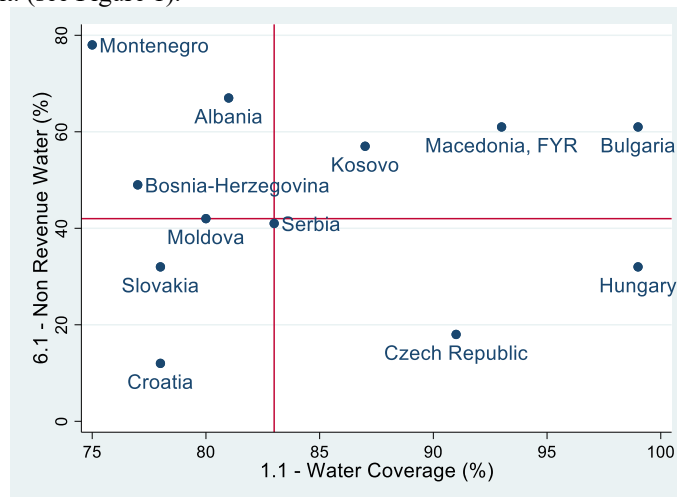
26 In BIH, responsibility for water service provision is decentralized and rests within
27 municipalities. There are about 130 municipal water companies in BIH serving the
28 needs of a population of 3.0 million.

29 There is only one type of ownership and operating in BIH i.e. there are only
30 municipally owned water operators. As in the Communist time and as a legacy from
31 that time, they operate as formally autonomous organizations, separated
32 organizationally, administratively and financially from the municipal authorities.
33 However, since the transfer of responsibilities to operator level is not clear-cut,
34 operators are capable of functioning as productive and efficient providers of services.

35 By being forced to follow social criteria (and consequently on practicing
36 overemployment and facing depressed tariffs), they amass costs and delay the
37 desperately needed investments on mains and other parts of infrastructure.

38 Due to the lack of financial resources for maintenance and investments and inefficient
 39 water management in most of operators, percentages of water coverage and non-
 40 revenue water are unsatisfactory. More than 24 years after the Dayton Peace
 41 Agreement, access to water services has been unacceptably low or unreliable.

42 Non-revenue water levels in the Balkan countries are the highest in Europe
 43 indicating worn-out water pipe networks, weak rule of law and poor water supply
 44 management. (see Figure 1).



45

46 **Fig. 1.** Non-revenue water and water coverage; BIH and comparators, 2017

47 Bearing in mind the current ranking of BIH, it would be useful to properly assess
 48 the productivity of water operators to find out what is behind them in order to improve
 49 the relative position of the country by introducing proper policy and regulatory
 50 measures.

51 **2 Materials and Methods**

52 This paper focuses on the examination of how productivity of operators has changed
 53 over time and decomposing it into its constituent parts. Such analysis can assist in
 54 examining the impact of regulation on the functioning of operators. Actually it can
 55 contribute to the idea of how to simulate market functioning within a non-market sector
 56 by comparing performances of operators and finding out how productive and efficient
 57 they are over time.

58 We examine whether the change in productivity for a particular operator has been
 59 driven by improving its relative efficiency, by scale improvement, or by general
 60 technological improvement.

61 We consider a single-output production function, which, with panel data and output-
 62 oriented technical inefficiency, is written as:

$$63 \quad y_{it} = f(x_{it}, t) \exp(-u_{it}) \quad (1)$$

64 where y_{it} is the output of the i th operator ($i=1, \dots, N$) in period t ($t=1, \dots, T$), $f(\cdot)$ is the
 65 production technology, x_{it} is a vector of J inputs, t is the time trend variable, and $u_{it} \geq 0$
 66 is output-oriented technical inefficiency.

67 We simply focus on rate of change of output and examine how much of it is driven by
 68 (i) rate of changes in inputs, (ii) changes in technology, and (iii) changes in efficiency.

69 If we start from (1) and differentiate it totally with respect to t , we get

$$70 \quad \dot{y} = \sum_j \varepsilon_j \dot{x}_j + TC + TEC \quad (2)$$

71 where \dot{y} rate of change of output, $\sum_j \varepsilon_j \dot{x}_j$ inputs growth, TC is technical change
 72 ($\frac{\delta \ln(\cdot)}{\delta t}$), and TEC is technical efficiency ($-\frac{\delta u}{\delta t}$).

73 The equation (2) shows that output growth is driven by input growth, that technical
 74 change (TC) will affect output positively if there is technical progress, and that
 75 technical efficiency (TEC) will affect output positively if technical efficiency improves
 76 over time.

77 **3 Results**

78 The first step in our analysis is to estimate a Cobb-Douglas production function. From
 79 a statistical viewpoint, it yields a balanced panel of 21 operators but we estimated it
 80 as pooled cross section.

81 By dropping one by one the statistically insignificant variables whilst ensuring that the
 82 model passed a number of other diagnostic tests for the dataset we got:

$$83 \quad \text{lwd} = 5.08 + 0.85\text{lc} + 0.45\text{lm} + 0.09\text{le} - .06\text{lch} - 0.03\text{t} \quad (3)$$

84 The logarithmic specification is consistent with a standard multiplicative production
 85 function and enables the direct estimation of elasticities. The model appears to be a
 86 “good fit” with $R^2 = 0.934$. All the variables are statistically significant at 5% and
 87 they have their expected signs except for electricity costs. The sum of the coefficients
 88 on independent variables reflects, at a value of 1.3, the presence of scale economies (see
 89 Table 1).

90

Table 1. Regression

| Variables | Model |
|-----------|-----------------------|
| lc | 0.845*** (0.0658) |
| lm | 0.451*** (0.0878) |
| le | -0.0644** (0.0315) |

| | |
|--------------|-----------------------|
| lch | 0.0873*** (0.0135) |
| t | -0.0268** (0.0118) |
| Constant | 5.075*** (0.404) |
| Observations | 210 |
| R-squared | 0.934 |

91
92

Standard errors in parentheses
*** p<0.01, ** p<0.05, * p<0.1

93 The dependent variable in the regression model is the volume of water delivered
94 (measured annually in m³) (wd), and the independent variables are the number of
95 connections (c), the yearly average number of employees (in the hours of work) (m),
96 yearly electrical energy costs in BAM (e), and the yearly chemical costs (in BAM)
97 (ch). In addition, a time variable (t) is included in the model.
98 In the second step, we estimated the productivity (see Table 2). This is achieved by
99 running a stochastic frontier model (Kumbhakar, Wang and Horncastle, 2015).

100

Table 2. SFA

| Variables | Model |
|--------------|-----------------------|
| lc | 0.844*** (0.0652) |
| lm | 0.451*** (0.0866) |
| le | -0.0645** (0.0311) |
| lch | 0.0874*** (0.0134) |
| t | -0.0352 (0.0339) |
| Constant | 5.190*** (0.538) |
| Observations | 210 |

101
102

Standard errors in parentheses
*** p<0.01, ** p<0.05, * p<0.1

103 In the third step, we estimated productivity change and related it to efficiency (see
 104 Table 3). In order to derive components of changes in productivity we allow operators
 105 to be technically inefficient. (Kumbhakar, Wang and Horncastle, 2015).

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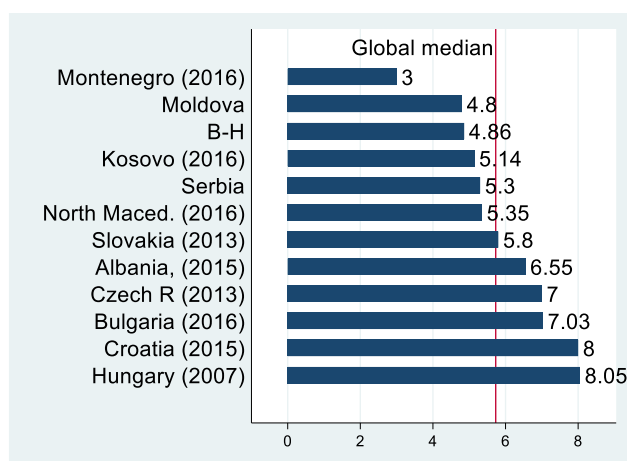
Table 3. SFA

| Variables | Results |
|---------------|----------|
| dot_wd | 0.009752 |
| TC | -0.03525 |
| input_driven | 0.032642 |
| TEC | -0.00806 |
| expl_output | -0.01067 |
| unexpl_output | 0.020422 |
| dot_wd | 0.009752 |

107 This demonstrates, that output growth, of around 1% percent, has been driven
 108 by increase in inputs of 3.3 percent, what is offset by reduction in technical
 109 progress of 3.6% and technical efficiency of 0.8%.

110 4 Discussion

111 The IBNET's Apgar score of 4.8 for the BIH operators is "fairly low". It places them
 112 among the worst within the group of comparators countries (small Central and South
 113 East European countries).



114

115 **Fig. 2.** Apgar score; BIH and comparators, 2017

116 The score is just below the median score of comparators countries (5.80) but even
 117 below the global median score (5.73) (see the Figure 2 at which years in brackets state
 118 the latest data available for the respective country, if data for 2017 is not available).

119 How much certain types of inputs contribute to such a “fairly low” position can be
 120 seen in the Table 4.

121 **Table 4.** Inputs of water production; BIH and comparators countries, 2017

| Countries | 13.1 Labor Costs | 3.2 Electrical Energy Costs | 100.13 Chemical Costs | 100.14 - Other Costs |
|----------------------|------------------|-----------------------------|-----------------------|----------------------|
| BIH | 64 | 9 | 1 (2016) | 54 (2016) |
| Albania | 41 (2015) | 38 (2015) | | |
| Bulgaria | 37 (2015) | 16 (2015) | | |
| Croatia | 39 (2015) | 9 (2015) | | |
| Czech R | 13 (2013) | 7 (2013) | | |
| Hungary | 42 (2007) | 17 (2007) | | |
| Kosovo | 62 (2016) | 21 (2016) | | 3 (2016) |
| North Macedonia | 51 (2016) | 8 (2016) | | 0 (2016) |
| Moldova | 45 (2017) | 26 (2017) | | |
| Montenegro | 39 (2016) | 39 (2016) | | |
| Serbia | 48 (2017) | 12 (2017) | | |
| Slovakia | 32 (2013) | 8 | | |
| Median-comparators | 40.5 | 16 | - | 1.5 |
| Median-all countries | 36 | 16 | 2 | 5 |

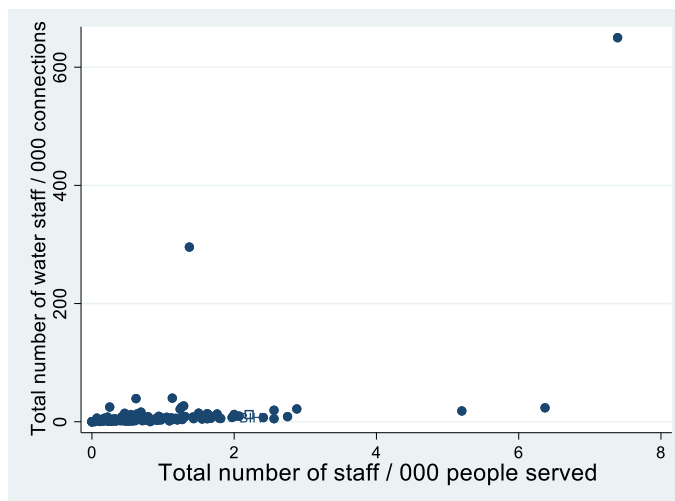
122 NB: Years for which the latest data available are stated in brackets.

123 Source: adapted from IBNET (2019)

124 As Table 4 shows labor costs of BIH operators amounts for 64 percent of total
 125 operational costs. Obviously, there is a mistake in the IBNET databases. To have a total
 126 of annual operational costs of 100%, the item *other costs* should be reduced to 26
 127 percent of total costs, or alternatively the item *labor costs* should be reduced to 36
 128 percent.

129 BIH’s total annual labor costs, expressed as a percentage of total annual operational
 130 costs, of 64% are pretty high in comparison to both the comparators countries and the
 131 world, since their respective median values are 40.5 and 36 percent. For instance, that
 132 size of the BIH operators’ labor costs are in line with the labor costs of Camerun (77%),
 133 Dominican Republic (71%), Jordan (65%), Papua New Guinea (64%), Argentina
 134 (64%), Kosovo (62%) and Ecuador (61%) (IBNET, 2019).

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136

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Fig. 3. Staff of water operators; BIH and world, 2017

138 Figure 3, showing staff productivity of operators, measured as the total number of
 139 staff members per 1000 connections (vertical axis) and per 1000 people served
 140 (horizontal axis), clearly demonstrates that by both indicators, BIH operators are
 141 overstuffed – global median of staff per 1000 connections is 5.12 and per 1000 persons
 142 served is 0.74 while BIH's respective medians are 7.24 and 1.97.

143 According to our dataset - which is not quite comparable with the IBNET dataset,
 144 since it relates to the period 2000-2009, and embraces 15% of total number of operators
 145 - ratios of respective costs i.e. labor, electrical energy, chemicals and other costs are
 146 37.5, 10.2, 0.2 and 48.0 percent respectively.

147 In addition, it was not possible to use the length of mains or network length, usually
 148 used in the production function as an indicator of capital, since data is missing or is
 149 not reliable in the case of many operators. Further, the number of operators was reduced
 150 to allow the dataset to establish balance panel data.

151 Regardless of these and other limitations, we did not expect that technical change
 152 would have a negative impact on productivity change. In our view it could stem from
 153 the worsening methods in using inputs (disembodied technical change) or
 154 alternatively and less likely through the use of quality with worsening quality
 155 (embodied technical change) in the respective period. If the former case, it could
 156 be due to the replacing of people leaving the sector due to retirement, shifting
 157 to other domestic sectors or emigrating abroad with less skillful workers. In
 158 addition, there are broadly spread negative comments coming from associations
 159 of employers about the worsening quality of employees.

160 Globally, the water sector is generally characterized by slowly changing
 161 technology. In BIH, the R&D expenditure is at a very low level (around 0.15 percent
 162 of GDP) and top managers are not appointed by professional but political criteria.

163 If prices of inputs were available, the estimation of the cost function and an another
 164 stochastic frontier model would have been carried out, shedding more light on the issue
 165 of technical progress.

166 **5 Conclusions**

167 In this paper we have examined how to estimate productivity and efficiency of
 168 Bosnian-Herzegovinian water operators.

169 In spite of their deficiencies, partial productivity indicators clearly state that BIH
 170 operators have weaker performances than their counterparts in comparators
 171 countries (small Central and South East European countries) not to mention other
 172 European countries.

173 The research demonstrated that BIH policy makers should focus on reforming the
 174 water system by introducing more economic criteria in their operating e.g. introducing
 175 benchmarking techniques with the purpose of stimulating the introduction of more
 176 innovative activities, particularly in the field of operational processes and
 177 organizational changes. Introducing private organizations e.g. through public-private
 178 partnership could contribute to the improvement of the currently very weak sector. The
 179 IBNET's Apgar score demonstrates that clearly.

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