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A DEA MODEL WITH A NON DISCRITIONARY VARIABLE FOR OLYMPIC EVALUATION

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ABSTRACT. In recent years, a lot of work has been done dealing with alternative performance rankings for the Olympic Games. Almost all of these works use Data Envelopment Analysis (DEA). Generally speaking, those works can be divided into two categories: Pure rankings with unitary input models and relative rankings with classical DEA models; both output oriented. In this paper we introduce an approach taking into account the number of athletes as a proxy to the country investment in sports. This number is an input for a DEA model, and the other input is the population of the country. We have three outputs, the number of gold, silver and bronze medals earned by each country. Contrary to the usual approach in the literature, our model is not output oriented. It is a non-radial DEA model oriented to the "number of athletes" input, as our goal is not a countries' ranking. We intend to analyse whether the number of athletes competing for each country accords with the number of won medals. For this analysis, we compare each country with its benchmarks. The Decision Making Units (DMU) are all the countries participating in the Beijing Olympic Games, including those that did not earn a single medal. We use a BCC model and we compare each DMU's target with the number of athletes who have won, at least one medal.

Keywords: Data Envelopment Analysis, olympic games, sport evaluation.

1 INTRODUCTION

Performance evaluation has been widely studied in sport for the last 25 years (Nevill et al., 2008). In the case of the Olympic Games the rank of the nations is traditionally carried out with the socalled Lexicographic Multicriteria Method (Lins et al., 2003). In this very paper the drawbacks of the Lexicographic Method are pointed out and a new ranking is suggested.

There are already some other approaches using Data Envelopment Analysis (Charnes et al., 1978). The very first one was proposed by Lozano et al. (2002). They used population and GNP as inputs and the medals as outputs. In a similar approach, Lins et al. (2003) built a new

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model taking in account one more constraint: the total amount of medals is a constant. This resulted in the development of a new model, the so-called Zero Sum Gains DEA model (ZSG-DEA). Churilov & Flitman (2006) used DEA to establish a ranking, the inputs of which were some social economical variables (population, GDP, DEL index and IECS index). The outputs were some linear combination of the number of medals earned by each country. Li *et al.* (2008) have used GDP per capita and population as inputs and impose different weight restrictions for each DMU, according to a previous country categorization.

Wu *et al.* (2008, 2009b) use a cross Evaluation DEA model to establish an Olympic Ranking. They came across the same problem described in Soares de Mello *et al.* (2002): the existence of negative efficiencies in the BCC-DEA model. They solved the problem with the same approach used by Soares de Mello *et al.* (2008b). Wu *et al.* (2009a) have used the results of Wu *et al.* (2009b) to study a Nash equilibrium in a BCC-DEA Cross Evaluation for the Olympic Ranking.

Zhang *et al.* (2009) used lexicographic preferences instead of weight restrictions (Allen *et al.*, 1997) to establish a DEA based ranking for the Athens Olympic Games.

All the works mentioned hereabove take into account the results in the Olympics and the socio economical conditions of each country. Models using DEA and nothing but medals won by each country were presented by Soares de Mello *et al.* (2004), Hai *et al.* (2007), Soares de Mello *et al.* (2008a, 2009) and Soares de Mello *et al.* (2008c).

Despite their important differences, the papers cited hereabove have some common points. The first one is that those papers evaluate only countries that have earned medals. In other words, the units evaluated by DEA (Decision Making Units – DMU) are the countries that have won at least one medal. The other point in common in all those papers is the DMU's objective: to maximize some combination of the number of earned medals.

In this paper we are not interested in Olympic Rankings. We wish to study whether each country's number of competitors is reasonable when compared to all the other countries. Moreover, we will evaluate all participating countries in the Beijing's Olympic Games including those that have won no medals. This is possible because of a mathematical property of the DEA model with variable returns to scale, the so-called BCC DEA model (Banker *et al.*, 1984). In that model, a DMU may still be efficient even if all its outputs have no value. Likewise the majority of models concerning Olympic evaluation, our model will have three outputs: number of gold, silver and bronze medals. The model will have two inputs: the population and the number of competitors for each country. As we intend to evaluate the dimension of each country's team we will orient the DEA model only to input. This leads to the study of non radial DEA models. These models will be summarized in section 2. Section 3 presents the DEA model used in our study. The results are presented in section 4 and in Section 5 we present some final comments.

2 DATA ENVELOPMENT ANALYSIS MODELS WITH NON DISCRETIONARY VARIABLES

N.C. 1

The classical BCC input oriented model (Banker *et al.*, 1984) in the envelopment formulation is presented in (1).

$$\begin{aligned} \min n_0 \\ \text{subject to} \\ h_0 x_{i0} &- \sum_{k=1}^n x_{ik} \lambda_k \ge 0, \quad \forall i \\ &- y_{j0} + \sum_{k=1}^n y_{jk} \lambda_k \ge 0, \quad \forall j \end{aligned}$$

$$\begin{aligned} \sum_{k=1}^n \lambda_k &= 1 \\ \lambda_k \ge 0, \quad \forall k \end{aligned}$$

$$(1)$$

where h_0 is the efficiency of DMU θ , the DMU under evaluation; x_{ik} is the input *i* of DMU *k*; y_{jk} is the output *j* of DMU *k*; λ_k is the share of DMU *k* for DMU θ 's target. According to the model, when an inefficient DMU achieves efficiency, all its inputs are proportionally reduced. Should we wish to achieve efficiency changing only the value of a single input, we ought to use model (2). This model was introduced by Banker & Morey (1986) and can be seen as a particular case of the MORO model (Lins *et al.*, 2004; Quariguasi Frota Neto & Angulo-Meza, 2007), when all but one factor are not allowed to be changed.

$$\begin{aligned} &\operatorname{Min} h_{0} \\ &\operatorname{subject to} \\ &h_{0}x_{v0} - \sum_{k=1}^{n} x_{vk}\lambda_{k} \geq 0, \\ &x_{i0} - \sum_{k=1}^{n} x_{ik}\lambda_{k} \geq 0, \quad \forall i \neq v \\ &- y_{j0} + \sum_{k=1}^{n} y_{jk}\lambda_{k} \geq 0, \quad \forall j \\ &\sum_{k=1}^{n} \lambda_{k} = 1 \\ &\lambda_{k} \geq 0, \quad \forall k \end{aligned}$$

$$(2)$$

In model (2), x_{vk} is the variable that can change.

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3 MODELLING

As mentioned earlier, the main goal of our study is to evaluate the performance of the Olympic teams in Beijing 2008 Games, taking into account the number of athletes for each and every country. We use as outputs the number of gold, silver and bronze medals won by each country. The controlled input is the number of athletes in each team, which is a proxy for the country investment in sports. This input has been proved to be an important contextual variable to explain the performance of the countries in the Olympics (Gomes & Souza, 2009). As an uncontrolled input we use each country's population. We have restrained the weight of each medal as in Soares de Mello *et al.* (2008a, 2008c, 2009), *i.e.* the gold medal weight is no lower than the silver medal weight, which again is no lower than the bronze medal weight. Moreover, the difference of weights between the gold and the silver medals is not lower than the difference of weights between the silver and bronze medals. Model (3) is the linear programme for a given country.

$$\begin{array}{l} \operatorname{Min} h_{0} \\ \operatorname{subject to} \\ x_{POP0} - \sum_{j=1}^{n} \lambda_{j} x_{POPj} \geq 0 \\ h_{0} x_{ATHL0} - \sum_{j=1}^{n} \lambda_{j} x_{ATHLj} \geq 0 \\ y_{G} \leq \sum_{j=1}^{n} \lambda_{j} y_{Gj} - \gamma_{1} - \gamma_{3} \\ y_{S} \leq \sum_{j=1}^{n} \lambda_{j} y_{Sj} + \gamma_{1} - \gamma_{2} + 2\gamma_{3} \\ y_{B} \leq \sum_{j=1}^{n} \lambda_{j} y_{Bj} + \gamma_{2} - \gamma_{3} \\ \sum_{j=1}^{n} \lambda_{j} = 1 \\ \gamma_{j} \geq 0, \quad \forall j \end{array}$$

$$\begin{array}{l} \end{array}$$

$$\begin{array}{l} \operatorname{Min} h_{0} \\ \operatorname{Subject} t_{0} \\$$

In model (3), h_o is the efficiency of the observed country; γ_j is the share of country *j* for observed country *o*; x_{POP0} is the observed country's population; x_{ATH0} is each country's number of athletes; y_G , y_S and y_B are the observed country's number of gold, silver and bronze medals; y_{Gj} , y_{Sj} and y_{Bj} are country *j*'s number of gold, silver and bronze medals. Variables γ_1 , γ_2 and γ_3 are the dual variables corresponding to the weight restrictions in the primal problem (multipliers formulation).

4 RESULTS

Table 1 presents the results for some countries, using model (3). In this table, the "ideal" number of athletes is the non-radial target for each country. It is obtained by multiplying the efficiency by the number of athletes. The Number of Medal Winners, or number of medallists, represents how many athletes did a country need to win all its medals. For instance, if a country wins eight medals with only one athlete it would have number of medal winners equal to one. On the other hand if a country wins a medal in a collective sport, for instance, soccer, it would lead at least eleven athletes to win that medal.

We may observed that for some countries the ideal number of athletes is not an integer. Almost all researchers who work with output oriented models in DEA for Olympic Evaluation do not take into account that the number of medals cannot be a fractionary number. The only exception is the work of Wu *et al.* (2010). In our case, we will not deal with the problem of non integer values. We assume that in the case of a non integer target, the DMU may use the nearest integer.

Obviously it is impossible to know beforehand who will win an Olympic medal and who will not. So the "Ideal Team" must not be equal to the number of medallists. The "Ideal Team" is, in most cases, larger than the number of medal winners. So, there are no winners in the "Ideal team". There are some exceptions. For Iceland, Brazil, Argentina, Nigeria and Serbia the "Ideal Team" is smaller than the number of medallists. For such countries it is impossible to reduce the number of athletes down to the ideal number and still win the same number of medals. This result might vindicate the idea that the sports investment policy might not be the best policy for those countries.

As China has won the largest number of gold medals, a well known property of the BCC-DEA model (Ali, 1993) ensures that this country is efficient irrespective of input levels.

In Table 1, Niger, Burundi, Central African Republic, Gabon, Belize, Dominica, Nauru and Tuvalu have won no medals yet they are DEA efficient. This is because Tuvalu has the smallest population and the others have only one competing athlete. Due to another well known BCC-DEA model property (Ali, 1993), these DMUs are efficient irrespective of output levels.

China's fielding many non-winning athletes makes this analysis render larger other countries' ideal teams; several countries having won no medals, confirms it. Both enhance the Olympics ideal.

As mentioned, ideal teams need not necessarily include winners.

Table 2 shows the benchmarks for the countries for which the "Ideal Team" is smaller than the actual number of medallists. In this table, the most important benchmark is the DMU with the biggest contribution to the non-radial target, *i.e.*, the biggest λ in model (3).

Except for Iceland, those countries' benchmarks are African. China is in fact Brazil's secondary benchmark. However, she profited from the home advantage phenomenon (Balmer *et al.*, 2001; 2003), which may render her not really a true benchmark despite being a mathematical one.

Country	Gold	Silver	Bronze	Number of	Number of	"Ideal" number	DEA
e cumu y			Dionet	athletes	medal winners	of athletes	efficiency
China	51	21	28	564	159	564.00	1.0000
USA	36	38	36	629	272	629.00	1.0000
Russia	23	21	28	411	134	411.00	1.0000
United Kingdom	19	13	15	297	59	297.00	1.0000
Australia	14	15	17	439	85	439.00	1.0000
Jamaica	6	3	2	56	13	56.00	1.0000
Kenya	5	5	4	52	14	52.00	1.0000
Ethiopia	4	1	2	21	5	21.00	1.0000
Cuba	2	11	11	152	47	152.00	1.0000
Zimbabwe	1	3	0	13	1	13.00	1.0000
Panama	1	0	0	5	1	5.00	1.0000
Bahamas	0	1	1	22	5	22.00	1.0000
Armenia	0	0	6	23	6	23.00	1.0000
Togo	0	0	1	2	1	2.00	1.0000
Niger	0	0	0	1	0	1.00	1.0000
Burundi	0	0	0	1	0	1.00	1.0000
Central African Republic	0	0	0	1	0	1.00	1.0000
Gabon	0	0	0	1	0	1.00	1.0000
Belize	0	0	0	1	0	1.00	1.0000
Dominica	0	0	0	1	0	1.00	1.0000
Nauru	0	0	0	1	0	1.00	1.0000
Tuvalu	0	0	0	3	0	3.00	1.0000
Mongolia	2	2	0	21	4	20.96	0.9979
Belarus	4	5	10	123	28	113.50	0.9228
Iceland	0	1	0	19	14	5.81	0.3056
Brazil	3	4	8	265	74	57.95	0.2187
Argentina	2	0	4	137	53	19.29	0.1408
Nigeria	0	1	3	82	24	11.50	0.1402
Serbia	0	1	2	88	15	9.33	0.1060
Angola	0	0	0	27	0	1.00	0.0370
Hong Kong	0	0	0	35	0	1.00	0.0286
South Africa	0	1	0	133	1	3.50	0.0263
Egypt	0	0	1	76	1	2.00	0.0263
Venezuela	0	0	1	92	1	2.00	0.0217

Table 1

Country	Most important	Other	
	benchmark	benchmarks	
Iceland	Nauru	Jamaica and Armenia	
Brazil	Kenya	China	
Argentina	Zimbabwe	Ethiopia and Kenya	
Nigeria	Ethiopia and Togo		
Serbia	Zimbabwe	Togo and Ethiopia	

Table 2

5 FINAL COMMENTS

We have presented in this paper a BCC-DEA model to evaluate countries at the Olympic Games. It is important to point out that we are not interested in rankings, but in sports evaluation. We have obtained the Ideal number of athletes for each country. This Ideal number includes winners and non-winners, preserving the "De Coubertin" Olympic spirit. However, the main result is that this does not apply for some countries. That is the case of Brazil, Nigeria, Serbia, Argentina and Iceland. As stated earlier, if those countries succeeded in reducing the number of their athletes, they would not be able to win the same number of medals. Countries whose "ideal" team is smaller than their number of all medal-winning athletes might have an inefficient policy of investment in sport (poor resource allocation). Those countries invest probably too much in team sports such as football (Argentina), volleyball (Brazil) or water polo (Iceland).

The above mentioned "inefficient" policy obtains only when the Olympic Evaluation is based on the total number of medals. Evaluations based on the number of medals won in different sports (Soares de Mello *et al.*, 2009) might have different results. It may also be noticed that qualifying a sports policy investment as inefficient does not take into account "national pride" ascribed to some sports in some countries and of course we ought not to neglect the important of the national pride in the Olympics (van Hilvoorde *et al.*, 2010).

It is interesting to notice that those benchmark countries are mainly African. These are particularly strong countries in individual sports.

Future studies may explore these conclusions and try to identify which are the most profitable sport investments for each country. And we may take also into account that the number of athletes is an integer value and we will use FDH models (Deprins *et al.*, 1984) or integer valued DEA models (Kazemi Matin & Kuosmanen, 2009; Lozano & Villa, 2006; Wu *et al.*, 2010).

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