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Assessing the Efficiency of Mother-to-Child HIV Prevention in Low- and Middle-Income Countries using Data Envelopment Analysis

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Abstract

AIDS is one of the most significant health care problems worldwide. Due to the difficulty and costs involved in treating HIV, preventing infection is of paramount importance in controlling the AIDS epidemic. The main purpose of this paper is to explore the potential of using Data Envelopment Analysis (DEA) to establish international comparisons on the efficiency of implementation of HIV prevention programmes. To do this we use data from 52 low- and middle-income countries regarding the prevention of mother-to-child transmission of HIV. Our results indicate that there is a remarkable variation in the efficiency of prevention services across nations, suggesting that a better use of resources could lead to more and improved services, and ultimately, prevent the infection of thousands of children. These results also demonstrate the potential strategic role of DEA for the efficient and effective planning of scarce resources to fight the epidemic.

Keywords: HIV Prevention, DEA, Mother-to-Child HIV Transmission.

1. Introduction

The epidemic caused by the human immunodeficiency virus (HIV) is currently one of the greatest public health problems worldwide. It is estimated that in 2009 alone, 33.3 million people were living with HIV/AIDS, 2.6 million were newly infected with HIV and 1.8 million died from AIDS [1]. Although the majority of the disease burden remains in sub-Saharan Africa, the number of HIV-infected people is rapidly increasing in other regions, especially in Eastern Europe and Central Asia.

Due to the difficulty and costs involved in treating HIV, preventing infection is of paramount importance in controlling the AIDS epidemic. In particular, effective prevention measures can reduce new infections, the incidence of the disease and, consequently, the number of people that will require treatment. Whilst appropriate prevention measures can change the course of the epidemic, they require very substantial resources. It is estimated that the average cost of preventing an infection is US \$3,923 [2]. Given the limited resources available to healthcare systems in many of the most affected countries, the implementation of efficient and effective HIV prevention interventions is vital for policy makers and health care managers. It is therefore imperative to identify the most effective HIV prevention strategies as well as the most efficient ways of allocating the scarce resources available to these strategies.

Resource allocation models have been developed over the years intended to assist with allocation strategies that might improve the overall effectiveness of HIV prevention efforts. For some recent examples, the reader is referred to Lasry et al. [3], Earnshaw et al. [4] and Brandeau et al. [5]. However, as pointed out by Rauner and Brandeau [6], determining how best to spend limited resources on HIV prevention programmes in different regions of the world is a challenging task. One way of overcoming this challenge and obtain useful information for decision making consists of identifying best practices that can be shared to promote improvements. Benchmarking exercises to identify best practices amongst health care providers within a country or across countries have the potential to provide essential information for the improved allocation of HIV prevention funds.

Identifying best practice is not, however, straightforward. Although some international comparisons have been carried out in recent years, the approaches adopted have tended to rely on two major types of analysis: one that focuses on specific interventions and compares the relative costs and benefits of a particular prevention programme against the other alternatives (e.g. Beck and Miner [7]; Harling et al. [8]), and another that focuses on data aggregated at the country level and uses multiple performance indicators to show the progress that is being made by each country towards HIV prevention (e.g. UNAIDS [1]). Whilst valuable, these approaches are limited in carrying out comparative analyses and identifying best practice. The former is strongly impacted by the range of prevention strategies and geographic settings examined, ignoring the fact that one intervention might not be efficient in some countries but be efficient in others. The latter does not usually account for the fact that some countries might be better than others when using certain indicators, while poorer using others, making it difficult to derive a single aggregate measure of the overall performance of each of these countries in fighting the HIV epidemic.

The main purpose of this paper is to address these limitations by exploring the potential of using Data Envelopment Analysis (DEA) to establish international comparisons on the efficiency of HIV prevention, focusing on mother-to-child transmission, which is one of the strategies given high priority in battling the HIV epidemic. For an overview of the most important types of HIV/AIDS prevention and treatment interventions, the reader is referred to Rauner and Brandeau [6]. Mother-to-child transmission occurs when an HIV positive woman passes the virus to her baby during pregnancy, labour and delivery, or by breastfeeding. It is estimated that without appropriate prevention and treatment, around 15 to 30 percent of babies born to HIV positive women will contract HIV during pregnancy and delivery. A further 5 to 20 percent will become infected through breastfeeding [9]. It is, therefore, fundamental to ensure that mother-to-child-interventions which counsel and test pregnant women for HIV, offer antiretroviral medications to both the mother and infant to prevent HIV transmission, and

provide substitute feeding programmes to prevent transmission through breastfeeding, are as efficient and effective as possible. An efficient and effective use of resources in this particular context has the potential to save the lives of thousands of children each year.

2. Assessing Efficiency of Mother-to-Child HIV Prevention

In recent years, we have witnessed remarkable progress in the fight against the mother-to-child transmission of HIV, with the AIDS Annual Report [1] documenting a significant decrease in the infection rates among children born to mothers living with HIV. According to this report, in 2009, an estimated 370,000 children contracted HIV during the perinatal and breastfeeding period, down from 500,000 in 2001. Prevention of mother-to-child HIV transmission (PMTCT) has played a key role in this decrease. Unfortunately, however, this success has not been shared by all countries. Whilst in high-income countries mother-to-child HIV transmission has been virtually eliminated thanks to effective prevention and care, the number of children infected in low- and middle-income countries is still very high. This might be explained in part by the fact that coverage of antenatal care and other services for preventing mother-to-child transmission among women living with HIV varies considerably across nations. However, the fact that infection rates also vary considerably among low- and middle-income countries, seems to indicate that the resources allocated to prevention have been more efficiently and effectively used in some countries than in others.

The need to use the resources allocated to HIV prevention efficiently can hardly be overstated. As emphasised by Appleby and Thomas [10] and Flessa [11], an inefficient use of resources in the health care context means not only that money is wasted, but more importantly that the opportunity is lost to save lives, avert pain, and provide care to those in need. The search for efficiency and effectiveness in the use of resources has been a key driving force for many of the OR/MS science applications in the health care sector and for the emergence of important publications in this area (see, for example, Ozcan and Smith [12]). With this call for more efficiency and effectiveness in health care, a call for a further shift in the balance between prevention and treatment has also been made, with some authors acknowledging that OR/MS techniques have an important role to play in helping the shift towards more prevention (e.g. Royston [13]). Whilst these calls apply to all health care areas, they are particularly meaningful in the case of the HIV/AIDS epidemic. The Committee on HIV Prevention Strategies in the United States pointed out that decisions regarding the allocation of HIV prevention funds represent the single most important set of HIV prevention decisions made [14]. However, as emphasised by Kaplan and Merson [15: 1907] it is important to bear in mind that HIV "resource allocation is not simply an argument for how to divide the pie; some allocations are arguably better than others."

Considering that an efficient and effective allocation of resources is a key ingredient in controlling the HIV epidemic, and that there is some compelling evidence suggesting that the unit cost of HIV prevention programmes may be both higher and more variable than some estimates suggest (e.g. Marseille et al. [16]), it is fundamental for researchers to carry out comparative analyses across nations in order to identify best practices. The research we discuss in this paper aims to shed some light on this issue by comparing the resources spent by several countries on mother-to-child HIV prevention with the services provided.

Although there is an increasing and highly relevant body of literature on the efficiency and/or effectiveness of specific prevention, treatment and care programmes (see for example, Harling et al. [8]), the number of studies assessing and comparing how efficiently different countries are using the available funds to combat the epidemic is very rare. Furthermore, when comparisons are made, these tend to rely on the use of multiple performance indicators or ratios rather than on a single aggregate measure of the overall performance of each country. This problem could be addressed if it was possible to reach an agreement on the relative importance of the various performance indicators. This is, however, seldom the case as the importance of each indicator is dependent upon the perspective taken and upon the priorities of the countries under evaluation.

Data envelopment analysis, developed by Charnes, Cooper and Rhodes in 1978 [17] is a non-parametric linear programming technique which has proved very effective in addressing this limitation and it is one of the most important and widely used approaches to measure the performance of homogeneous decision making units (DMUs) performing the same task. This technique uses a production metaphor. Each DMU is engaged in a transformation process, where by using some inputs (resources) it is trying to produce some outputs (goods or services). With the provision of these services the DMU achieves certain outcomes which measure the impact of the services on the users. One of the interesting features of DEA is that it allows each unit to identify a benchmarking group; that is, a group of units that are following the same objectives and priorities, but performing better. In this respect DEA aims to respect the priorities of each DMU by allowing each one of them to choose the weight structure for inputs and outputs that most benefits its evaluation. As a result, it aims to classify each unit in the best possible light in comparison to the other units.

The formulation of the DEA problem, under variable returns to scale, with output orientation, for DMU 0, as proposed by Banker et al. [18] is as follows:

Subject to:

$$\sum_{j=1}^{n} X_{ij} \lambda_j \leq X_{i0} \qquad (i = 1,...,m)$$

$$\sum_{j=1}^{n} Y_{rj} \lambda_j \geq \phi Y_{j0} \qquad (r = 1,...,s)$$

$$\sum_{j=1}^{n} \lambda_j = 1$$

$$\lambda_j \geq 0 \quad (j = 1,...,n)$$

Max ϕ

In this problem, *n* is the number of DMUs; *s* is the number of outputs and *m* is the number of inputs; Y_{r0} is the amount of output *r* generated by unit 0 and X_{i0} is the amount of input *i* used by unit 0; λ_j is the intensity variable for DMU *j*. The score

 ϕ obtained from the solution to this linear programming problem is the maximum rate of proportional expansion in all outputs of DMU θ , without decreasing its inputs. The efficiency rate of DMU θ can be obtained by calculating $1/\phi$, and will be equal to 1 if the DMU is efficient, and smaller than 1 if the DMU is inefficient when compared with the other DMUs. A comprehensive review of the DEA technique can be found in Cooper et al. [19, 20].

Since its advent in 1978, this method has been the subject of several theoretical developments and has been extensively applied in many sectors including education, finance, agriculture, sports, marketing and manufacturing, to name just a few. There have also been several documented applications of DEA in the health care sector. The Health Care Management Science journal, amongst others, has been an important forum for the dissemination of these applications. In two recent special issues of this journal dedicated to contemporary health care applications in OR/MS and to operational research applied to health services, the reader can find several good examples of DEA in this context (e.g. Ozgen and Ozcan [21]; O'Neill and Dexter [22]; and Langabeer and Ozcan [23]). Hollingsworth et al. [24], Hollingsworth [25], Chilingerian and Sherman [26] and more recently Hollingsworth [27], review several of the DEA applications in the health care context.

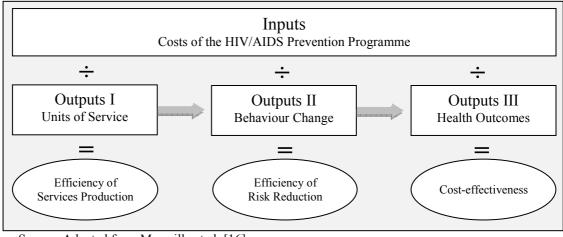
Whilst the use of DEA in the health care sector has been prolific, to the best of our knowledge, no studies have been documented on the use of DEA comparing the efficiency of countries in using HIV prevention resources. The only exception to this is the study by Zanakis et al. [28], which examines the impact of major socio-economic factors on HIV/AIDS indicators and assesses each country's efficiency in battling the epidemic. In so doing, it makes an important contribution to this under researched area, not only because it is the first one using DEA to assess the efficiency of nations in battling HIV but also because it identifies that research on HIV/AIDS should include both individual and socio-economic factors, to better understand the development of the global epidemic.

The research we discuss in this paper presents, however, some important departures from Zanakis et al.'s study. Firstly, the assessment of each country's

efficiency in preventing HIV infections is the main focus and objective of our research. In Zanakis et al.'s study this objective was part of a broader agenda and played a fairly marginal role. Secondly, our study focuses exclusively on the efficiency of HIV prevention, whilst theirs does not seem to distinguish prevention from treatment. However, our study, in line with the findings reported by Beck [29] and Piot et al. [30] acknowledges that the containment of the HIV epidemic requires a global strategy which combines effective prevention with treatment and care programmes. This implies that some variables related to HIV treatment and care will be considered in the analysis as they have a significant impact on HIV transmission. Thirdly, the model we propose is considerably different from the one they have used. In particular, we focus on a particular programme and use variables more directly related to HIV prevention than the surrogate general health variables used by Zanakis and colleagues. Finally, our study, by using data for 2008, aims to offer an updated overview of the efficiency of recent efforts in battling the epidemic whereas the analysis carried out by Zanakis et al. used data from 1998. Thus, our research complements this previous work and makes an important contribution to knowledge. Zanakis et al. recognised that their study was preliminary and that further additional multicountry studies in global efficient strategies for combating HIV/AIDS were needed. As emphasised by Harling et al. [8], establishing multi-country comparative analyses on the use, cost and outcome of HIV prevention services, especially in middle and lower income countries, is critical to fill some of the large gaps which exist in the scientific literature on HIV interventions in these countries. These gaps have limited the ability for the existing literature to guide policy makers in those settings where the epidemic is most intense. As pointed out by Harling et al. [8: 372] "increased research in such settings and dissemination of their findings is urgently required, especially given the need for intensified prevention strategies to complement the scaling up of HIV treatment and care services in these countries."

The framework proposed to assess the efficiency of mother-to-child HIV prevention and to contribute to the extant literature, draws on the principles of the Prevent AIDS Network for Cost-Effectiveness Analysis (PANCEA) approach [16]. PANCEA is a five-country study, funded by the U.S. National Institute of

Health, aimed at providing information for an improved allocation of HIV prevention funds in low- and middle-income countries. In particular, the PANCEA approach is designed to generate three indicators of efficiency for each prevention programme: cost per unit of service provided, cost per reduction in risky behaviours, and cost per HIV infection prevented. Figure 1 illustrates how these indicators build on each other.



Source: Adapted from Marseille et al. [16].

Figure 1. The PANCEA Conceptual Framework

The first indicator of efficiency measures the cost per unit of service provided by a particular programme or intervention (e.g. cost per pregnant woman who received voluntary counselling and testing). The second indicator of efficiency assesses the cost of reducing risky behaviours (e.g. cost per newly protected sex episode). The third indicator measures the cost effectiveness of the programme by comparing its costs with the outcomes achieved (e.g. cost per HIV infection averted).

The analysis we discuss in this paper focuses essentially on the Efficiency of Services Provision regarding mother-to-child HIV prevention. In spite of its limited scope, it can provide essential information for an improved allocation of HIV prevention funds as there is compelling evidence showing that the three indicators above are strongly linked. In particular, behavioural evaluations have shown that the more units of service provided the lower the risky behaviours tend to be. Epidemic modelling has shown, in turn, that a reduction in behavioural risk contributes to reducing HIV transmission. Therefore, it seems reasonable to assume that by comparing the efficiency of service provision across countries and identifying the best performers, we are able to identify those countries that are most likely using the resources in a more effective way. This is, however, the subject of further research. Given that our study focuses on the efficiency of nations in implementing a particular prevention programme whilst the PANCEA project focuses on the efficiency of particular programmes when compared with alternative interventions, we believe that these two approaches can, together, provide a richer understanding of efficiency in HIV prevention and, consequently, a better foundation for policy recommendations.

3. Empirical Analysis

3.1. The DEA Model

The choice of appropriate input and output variables is a fundamental step in DEA analysis. Whilst several input indicators may be used in this context, the costs of running mother-to-child HIV prevention interventions is, perhaps, the most relevant. These costs are usually grouped into three major categories: personnel expenses (e.g. salaries and other compensation paid to the personnel involved in the prevention programme), other recurrent goods and services expenses (e.g. medicines, laboratory tests, supplies, office and administrative expenses, and utilities), and capital expenses (e.g. medical equipment, furniture, vehicles, computers, and buildings). In this research we use the variable "Prevention of mother-to-child HIV transmission domestic spending from public and international financing sources (million US\$)" as the only controllable input.

The resources allocated to PMTCT can have several uses as various interventions are usually developed as part of the core response to achieve this objective. The choice of the most appropriate outputs should, therefore, aim to include in the DEA model those outputs thought to have the greatest potential to reduce motherto-child HIV transmission. Taking this principle into account, the model we propose has four main outputs, which follow.

The first output variable in our model is the "Reported number of pregnant women tested for HIV". As identified by the Joint United Nations Programme on HIV/AIDS [31], voluntary counselling and testing are an essential part of any PMTCT strategy. HIV testing is critical because pregnant women who do not know they are HIV positive cannot benefit from appropriate prevention interventions. The protective benefit of antiretroviral drugs is diminished when babies continue to be exposed to HIV through breastfeeding, therefore, HIV positive mothers are advised not to breastfeed whenever they have safe, affordable alternatives to breastfeeding. HIV testing and counselling also offer significant opportunities to provide important advice to HIV positive mothers on the risks and benefits of different infant feed options. This allows them to make better informed decisions about the most suitable options and, ultimately, contributes to lower mother-to-child HIV transmission. In addition, when testing is carried out and post-test counselling is given, valuable information and support is also passed to women not infected to ensure the avoidance of future infection.

The second output variable is the "Number of pregnant women living with HIV who received antiretrovirals for preventing mother-to-child transmission". The administration of antiretroviral drugs to pregnant women decreases the viral load in the mother and is highly effective at preventing mother-to-child transmission (see, for example, Bryson [32]). Pregnant women who are HIV positive are usually advised to begin antiretroviral treatment either immediately or after the first trimester. Under the 2010 World Health Organization guidelines, all HIV positive mothers identified during pregnancy should receive an extensive course of antiretroviral drugs to prevent mother-to-child transmission.

In order to lower the risk of transmission even further and/or reduce HIV-related morbidity and mortality, it is important, that newborn babies who have been exposed to the virus, are also given a course of prophylactic treatment for the first

few days or weeks with antiretrovirals and/or co-trimoxazole, respectively. Cotrimoxazole is a widely available antibiotic that substantially reduces HIV-related morbidity and mortality in both adults and children. As pointed out by Zachariah et al. [33: 686], "prophylaxis with co-trimoxazole is a recommended intervention of proven benefit that could serve not only as an initial step towards improving paediatric care in young children with limited access to antiretroviral treatment, but also as an important complement to antiretroviral therapy in resource-limited settings". Marseille et al. [34] have also provided evidence, more than a decade ago, suggesting that the administration of antiretrovirals to mothers and babies just before and soon after birth has a major potential impact on reducing vertical HIV transmission. In order to account for these factors, we included in our DEA model the following output variables: "Reported number of infants born to women living with HIV receiving antiretrovirals for preventing mother-to-child transmission" (i.e. Output 3) and "Reported number of infants born to women living with HIV receiving co-trimoxazole prophylaxis within two months of birth" (i.e. Output 4).

In a comparative analysis of countries battling the HIV/AIDS epidemic it is also important to bear in mind that there are socio-cultural, political and economic factors (i.e. exogenous conditions) not directly controllable by the prevention authorities that can explain differences in efficiency and ultimately help understand the development of the epidemic within each country. Consequently, it is important to include in the analysis nondiscretionary variables (inputs and/or outputs) that will assure comparable technologies. According to Quinn [35], to control AIDS, countries must not only promote changes in individual behaviour but also address socio-economic issues. In fact, considerable HIV/AIDS differences among nations have been attributed to exogenous conditions like nation's wealth, migration from rural to urban areas, education, access to health services and drug use. For example, Zanakis et al. [28], in exploring the effects that socio-economic factors have on the development of the HIV/AIDS epidemic, found evidence that countries with lower population density that manage to provide better health system performance, per capita support and better media information, without necessarily higher GNP, are more likely to exhibit lower HIV/AIDS indicators.

In the particular case of mother-to-child HIV transmission, several factors including education, degree of urbanization and access to health services have also been shown to impact the capacity of families to comply with prevention programmes.

It is acknowledged that good quality HIV/AIDS education and knowledge regarding mother-to-child-HIV transmission are fundamental to eliminating myths and misunderstandings, and to counter stigma and discrimination against pregnant women. It is likely that in countries with lower literacy rates, pregnant women will be less willing to carry out voluntary testing and counselling for HIV, which might result in poorer efficiency results when compared with nations with higher literacy rates. To account for this non-controllable factor, and allow fairer comparisons, we decided to include a proxy variable to capture the level of education in the population. We have used the variable "People aged 15 years and older who can, with understanding, both read and write a short simple statement on their everyday life (millions)". This variable resulted from the product of the literacy rate of each country and the respective population aged 15 years and older.

Easy access to health facilities is also an important motivator for mothers to adhere to prevention programmes. In particular, there is evidence suggesting that in rural areas health care facilities are less accessible, and therefore, that pregnant women living in these areas are at higher risk of not complying with PMTCT programmes (e.g. Jones et al. [36]). In order to account for this factor we have included in our DEA model the variable "People living in urban areas (millions)". This is an important variable to consider as the degree of urbanization also seems to have a significant impact on awareness among pregnant women regarding AIDS, HIV transmission and prevention methods. As pointed out by Msellati [37], when referring to mothers' access to PMTCT programmes in West Africa, the level of knowledge about the HIV/AIDS epidemic can be considered good in urban areas but worse in rural areas. Increased awareness by pregnant women is important as it improves their compliance with programmes targeted at preventing mother-to-child HIV transmission.

In addition to the socio-economic factors previously discussed, it is also known that weak health infrastructures exacerbate the impacts of the HIV/AIDS epidemic. As suggested by Coovadia and Hadingham [38] one of the major reasons for the apparent ineffectiveness of global interventions in battling the HIV/AIDS epidemic is the structural weaknesses in the health systems of underdeveloped countries, which contribute to bottlenecks in the distribution and utilisation of funds. As pointed out by these authors, robust health care systems play a critical role in channelling globally recognised prevention and treatment best practice for the mitigation of HIV/AIDS. It is important to bear in mind that investment in health infrastructures and in health care professionals can improve care provision, not only through improved detection and monitoring of HIV positive people but also through treatment of co-morbidities and nutritional support (Little et al. [39]). In order to account for this factor we have used the proxy variable "Total health expenditure (million US\$)". This variable is the sum of public and private health expenditure and it covers the provision of health services (preventive and curative), family planning activities, nutrition activities and emergency aid.

Finally, we decided to include a variable to account for political factors. There is evidence suggesting that political instability can undermine maternal and child health services (e.g. de Castella [40]) and compromise adherence to programmes targeted at preventing mother-to-child HIV transmission. Unstable governments tend to put HIV prevention at the bottom of their political agenda. In order to account for this factor and carry out a fairer comparison of countries, we have used the variable "Political stability and absence of violence/terrorism" as a non-controllable input in our model. This variable is from the Worldwide Governance Indicators (Kaufmann et al., [41]) and it reflects the compilation of responses on the quality of governance given by a large number of enterprise, citizen and expert survey respondents in industrial and developing countries. Considering that this variable ranges from approximately -2.5 (weak) to 2.5 (strong) governance performance, a transformation was undertaken in order to ensure that no negative values were present. Following Ali and Seiford [42], this input was transformed

using the following expression: $X_{io} = abs(Min X_{ij}) + X_{io} + c$. In this

expression, X_{io} is the transformed value for input *i* of DMU₀ and X_{ij} is the original value of input *i* for DMU_j, X_{io} is the original value for input *i* of DMU₀ and *c* is a constant.

In summary, the DEA model we propose to assess the efficiency of mother-tochild-HIV transmission interventions and which was developed based on the literature review and on the discussion presented above, is the one presented in Table 1:

Inputs	Outputs
Discretionary:	• Output 1: Reported number of pregnant
• Input 1: Prevention of mother-to-child	women tested for HIV;
HIV transmission domestic spending from	• Output 2: Number of pregnant women
public and international financing sources	living with HIV who received
(million US\$).	antiretrovirals for preventing mother-to-
	child transmission;
Non-discretionary:	• Output 3: Reported number of infants
• Input 2: People aged 15 years and older	born to women living with HIV receiving
who can, with understanding, both read	antiretrovirals for preventing mother-to-
and write a short simple statement on their	child transmission;
everyday life (millions);	• Output 4: Reported number of infants
• Input 3: People living in urban areas	born to women living with HIV receiving
(millions);	co-trimoxazole prophylaxis within two
• Input 4: Total health expenditure (million	months of birth.
US\$);	
• Input 5: Political stability and absence of	

Table 1: DEA model to assess efficiency of mother-to-child HIV transmission prevention

It is important to point out that a sequential production process may be in place here, rather than a simultaneous one, given that women must first be tested, and only afterwards can the other services be offered. However, considering that the available data does not allow distinction between the money that is spent on testing and that spent on targeted medication, it is important to include both types of outputs in the model. The DEA assumption of trade-offs between the output

violence/terrorism.

variables is still applicable in this context, as certain countries may choose, at first, to spend most of the money on developing programmes to reach women for opportunistic testing. Other countries, which may already have in place effective testing programmes may choose to spend more money on providing targeted medication. In the future, if disaggregated data is available regarding the resources used for HIV testing and the resources used for targeted medication, it may be interesting to analyse the efficiency of each of these production stages as well as the effectiveness of the programmes in order to assess the impact of the services delivered.

3.2. Data and Efficiency Results

The data used in this study refers to the activity of 52 low- and middle-income countries regarding mother-to-child HIV prevention in 2008. As previously discussed, PMTCT programmes in developed countries have almost eliminated vertical HIV transmission. However, progress in low- and middle-income countries has been slow, and therefore, it is fundamental to compare prevention programmes in these countries in order to identify best practice. Whilst lack of availability of data prevented us from considering all low- and middle-income countries in the analysis, we believe that the 52 nations assessed are reflective of the problems faced by many other low- and middle-income countries worldwide regarding mother-to-child HIV prevention. Our sample includes countries belonging to various geographic regions. In particular, it includes 27 countries from sub-Saharan Africa, 6 from East, South and South-East Asia, 9 from Europe and Central Asia and 10 from Latin America and the Caribbean. Of these, 28 are low-income countries and 24 are middle-income countries. Our sample also includes countries with various HIV epidemic levels. In 26 of the countries assessed the epidemic is considered concentrated, in 22 it is considered generalised and in 4 it is considered low.

The data used in our analysis was collected from a variety of international public sources including the Joint United Nations Program for HIV/AIDS, United Nations Development Program, World Health Organization and World Bank.

Although these are trustworthy sources, some of the information they provide is supplied by each individual country. Considering that some of these countries have very poor statistics, the analysis that follows has to be interpreted with caution.

Table 2 presents a descriptive summary of the data for the 52 countries considered in the analysis.

	Spending in prevention (10 ⁶ US\$)	Literate population (millions)	Urban population (millions)	Spending in health care (10 ⁶ US\$)	Political stability index	Pregnant women tested	Pregnant women that received antiretrovirals	Infants that received antiretrovirals	Infants that received prophylaction treatment	
	Input 1	Input 2	Input 3	Input 4	Input 5	Output 1	Output 2	Output 3	Output 4	
Low-income	e countries									
Average	2.59	29.29	19.80	2959.79	1.39	340155	7486	5404	1643	
St Dev	5.97	103.50	63.68	10167.21	0.78	793671	14925	10844	4270	
Max	29.33	551.00	336.75	53485.00	2.42	4234401	59601	41253	21841	
Min	0.00	0.08	0.10	15.00	0.01	6281	15	11	0	
Middle-inco	ome countries									
Average	2.75	30.28	28.77	24007.08	1.68	361629	2635	2379	947	
St Dev	3.47	45.48	44.03	49086.92	0.61	553201	3742	3465	2271	
Max	13.90	157.71	163.92	195414.00	3.01	2381280	11971	10308	8315	
Min	0.00	0.35	0.26	124.00	0.24	5335	6	0	0	
Total										
Average	2.67	29.74	23.94	12674.92	1.52	350066	5247	4008	1322	
St Dev	4.93	81.27	55.15	35406.58	0.71	686742	11411	8366	3479	
Max	29.33	551.00	336.75	195414.00	3.01	4234401	59601	41253	21841	
Min	0.001	0.08	0.10	15	0.01	5335	6	0	0	

Table 2 - Summary statistics for the variables used in the study

From this table, one can see that there are considerable discrepancies across countries. For example, whilst the average spending on prevention of mother-to-child HIV transmission is approximately US\$2.7 million per country, some have spent considerably more than others. For instance, Sao Tome and Principe reports the lowest annual spending on this prevention programme (US\$1,035), whilst Kenya the highest (US\$29.3 million). From another angle, we can see that the estimated number of people aged 15 years and older who can, with understanding, both read and write is also highly variable between nations. This is explained in part by the fact that some countries like India are highly populated compared to other, small, countries (e.g. Sao Tome and Principle). Total health expenditure,

political stability and urbanization are also highly variable as can be seen from Table 2.

Considering that there are remarkable discrepancies at the level of the inputs, it is not surprising that these will also exist at the outputs level. As shown in Table 2, the number of pregnant women tested for HIV and the number found to be HIVpositive and receiving antiretroviral therapy also varies considerably among the 52 nations in our sample. The same applies to the number of infants receiving antiretrovirals and/or co-trimoxazole. In this particular case, it is important to mention that the zero value, shown as the minimum, indicates that the countries are not providing these medicines to infants or, most likely, that data were not reported, and therefore, it was assumed as being zero. Given that there is no consensus in the DEA literature on how best to handle the issue of missing data, that the model we propose has several outputs, and that no weight restrictions are imposed on the outputs, we decided to adopt the approach proposed by Kuosmanen [43], who suggests replacing missing outputs by zero. This ensures that countries with no data available do not benefit in the performance assessment in relation to their counterparts. Furthermore, this approach does not bias the calculation of the efficient frontier as long as the countries with the missing data are inefficient, even if these data had been reported correctly. However, this approach, like alternative approaches, also has its own limitations. In particular, in can impact the DEA results in two ways. Firstly, it may bias the efficient frontier if the countries with the missing data would have been efficient had the missing data been reported correctly. Secondly, it may bias the efficiency scores for the countries who fail to report the data, because it may make them appear less efficient than they actually are.

In order to assess each nation's efficiency, we have used the PIM DEA software [44] and a DEA model with output orientation. This is justified by reasoning that countries should aim to increase the level of service provision, given spending on prevention and in health care in general, the population's education, the degree of urbanization and the level of political stability observed. Furthermore, we have used an assumption of variable returns to scale (VRS), as proposed by Banker et al. [18]. According to Marseille et al. [45] the assumption of constant unit costs in

the context of HIV prevention may result in substantial inaccuracies as they found compelling evidence that efficiency increased (unit costs decreased) with scale, across all countries and interventions examined. Brandeau et al. [5] and Brandeau and Zaric [46], also discuss the issue of scale in HIV prevention programmes, pointing out that the relationship between investment in HIV prevention and HIV infections averted may not be linear, which indicates that increased spending on a prevention programme may not always be cost effective. Our choice of a VRS assumption is also consistent with the publication of Hollingsworth and Smith [47], which warns that when ratios are used, as is the case of the variable political stability, the BCC formulation [18] should be adopted. We have also used a weight restriction on the input side, in order to prevent countries from attributing a null weight to the variable related to spending on PMTCT. This restriction imposes that the sum of the virtual weights given by each country to the noncontrollable inputs cannot exceed the virtual weight given to the controllable input. The development of this virtual weight restriction follows the approach proposed by Sarrico and Dyson [48]. Please refer to the appendix for more details on the DEA model used.

Table 3 presents the preliminary results from the DEA model discussed above. It presents the efficiency scores for each one of the 52 countries during 2008 as well as the returns to scale for each country.

Due to the exploratory nature of this research and the limitations of the data, caution needs to be taken when we analyse these results. In spite of this, the results shown on Table 3 corroborate some previous findings reported in the literature and highlight some relevant aspects about the efficiency of each country in preventing mother-to-child HIV transmissions.

Table 3: Countries efficiency scores for 2008

	Technical	Pure Technical	Scale	Returns to
	Efficiency	Efficiency	Efficiency	Scale
	(%)	(%)	(%)	
Low-income countries		, ,		
Benin	83.15	85.46	97.30	DRS
Burkina Faso	70.54	89.46	78.85	DRS
Burundi	30.6	76.62	39.94	IRS
Cambodia	39.64	51.91	76.36	DRS
Central African Republic	100	100	100.00	CRS
Chad	95.53	100	95.53	IRS
Congo	20.35	20.37	99.90	DRS
Côte d'Ivoire	100	100	100.00	CRS
Democratic Republic of the	100	100	100.00	CRS
Eritrea	100	100	100.00	CRS
Ghana	49.85	87.23	57.15	DRS
Guinea-Bissau	27.3	28.48	95.86	IRS
India	100	100	100.00	CRS
Kenya	100	100	100.00	CRS
Kyrgyzstan	100	100	100.00	CRS
Madagascar	100	100	100.00	CRS
Malawi	100	100	100.00	CRS
Mali	100	100	100.00	CRS
Mozambique	100	100	100.00	CRS
Myanmar	58.46	67.94	86.05	DRS
Niger	100	100	100.00	CRS
Nigeria	100	100	100.00	CRS
Rwanda	100	100	100.00	CRS
Sao Tome and Principe	100	100	100.00	CRS
Senegal	38.04	51.65	73.65	DRS
Tajikistan	21.86	26.55	82.34	DRS
Togo	58.82	60.04	97.97	IRS
Viet Nam	29.38	45.65	64.36	DRS
Middle-income countries*				
Angola	27.74	36.36	76.29	DRS
Argentina	37.01	60.01	61.67	DRS
Armenia	23.76	27	88.00	DRS
Azerbaijan	48.68	71.39	68.19	DRS
Belarus	27.86	47.47	58.69	DRS
Bolivia (Plurinational State of)	13.17	18	73.17	DRS
Botswana	100	100	100.00	CRS
Brazil	68.8	100	68.80	DRS
Cameroon	100	100	100.00	CRS
Chile	16.33	27.85	58.64	DRS
Colombia	39.8	43.7	91.08	DRS
El Salvador	26.37	43.59	60.50	DRS
Equatorial Guinea	66.11	71.25	92.79	IRS
Georgia	32.94	35.16	93.69	DRS
Honduras	36.02	48.8	73.81	DRS
Indonesia	56.73	100	56.73	IRS
Lesotho	100	100	100.00	CRS
Mexico	22.57	35.38	63.79	DRS
Paraguay	31.1	33.08	94.01	DRS
Peru	43.62	79.19	55.08	DRS
Republic of Moldova	36.5	36.66	99.56	DRS
Russian Federation	41.86	62.17	67.33	DRS
Thailand	77.64	100	77.64	DRS
Ukraine	49.46	69	71.68	DRS
Average	62.5	71.9	84.5	
St Dev	32.0	29.0	17.1	
Max	100.0	100.0	100.0	
Min	13.2	18.0	39.9	

Note: IRS=increasing returns to scale; CRS=constant returns to scale; DRS=decreasing returns to scale. *Includes lower-middle income and upper-middle income countries. As can be seen from Table 3, we have broken down the technical efficiency score into pure technical efficiency and scale efficiency. In 2008, the average total efficiency score of the 52 countries was only 62.5%, with a standard deviation of approximately 32.0%, which indicates that there are considerable differences between countries and remarkable potential for improvement in some of them. In our study, the spending on preventing vertical transmission of HIV is classified CRS efficient in eighteen countries: Botswana, Cameroon, Central African Republic, Côte d'Ivoire, Democratic Republic of the Congo, Eritrea, India, Kenya, Kyrgyzstan, Lesotho, Madagascar, Malawi, Mali, Mozambique, Niger, Nigeria, Rwanda and Sao Tome and Principe. When we allow VRS, which seems to be a more appropriate assumption to adopt in this context, as explained previously, the average efficiency score increases to 71.9% and the number of efficient countries increases to twenty two. This means that we have only four countries (Brazil, Chad, Indonesia and Thailand) that are technically efficient while operating under sub-optimal scale sizes. Chad and Indonesia seem to operate under increasing returns to scale (IRS), which suggests that a bigger scale of operation would lead to a greater ratio of outputs over inputs. Brazil and Thailand seem to operate under decreasing returns to scale (DRS). In fact, with regard to most of the countries that are technically inefficient, the DEA results suggest that they operate under DRS, indicating that they could benefit from decreasing their size of operation. These results seem to counter the findings of Marseille et al. [45], who found evidence that efficiency increased with scale. The only countries which could benefit from operating at a higher scale are, according to the results in Table 3, Burundi, Chad, Equatorial Guinea, Guinea-Bissau, Indonesia and Togo. Whilst scale seems to play an important role in the efficiency of some countries (e.g. Burundi, Ghana, Indonesia and Thailand), others (e.g. Benin, Congo, Georgia, Guinea-Bissau, Republic of Moldova and Togo) could benefit mostly from a better use of the resources allocated to preventing motherto-child HIV transmission, whilst maintaining their size of operation.

In analysing the efficiency results it is also important to identify the optimal weight structure chosen by each country. Despite the fact that the DEA model may yield alternative optimal solutions for the weights of the countries classified as efficient (Cooper et al. [49]), the different optimal weights identified can

provide useful information. For instance, the different optimal output weight profiles are useful to identify different service delivery strategies. Table 4 illustrates some of these different strategies.

	Spending in prevention (10 ⁶ US\$)	Literate population (millions)	Urban population (millions)	Spending in health care (10 ⁶ US\$) Input 4	Political stability index Input 5	Pregnant women tested	Pregnant women that received antiretro- virals	Infants that received antiretro- virals Output 3	Infants that received prophy- lactic treatment Output 4
	Input 1	Input 2	Input 3			Output 1	Output 2		
Azerbaijan	50.0	0.0	50.0	0.0	0.0	99.9	0.0	0.0	0.1
Belarus	50.0	42.6	7.4	0.0	0.0	98.4	0.0	0.0	1.6
Botswana	50.0	38.7	3.8	3.8	3.8	4.0	4.0	67.5	24.6
Brazil	50.0	50.0	0.0	0.0	0.0	100.0	0.0	0.0	0.0
Cameroon	50.0	12.5	12.5	12.5	12.5	21.7	14.4	14.4	49.6
Cambodia	50.0	0.0	50.0	0.0	0.0	97.5	0.0	2.5	0
Côte d'Ivoire	50.0	0.0	0.0	0.0	50.0	27.7	72.3	0.0	0.0
Chile	50.0	28.7	21.3	0.0	0.0	100.0	0.0	0.0	0.0
El Salvador	50.0	31.0	19.0	0.0	0.0	100.0	0.0	0.0	0.0
Equatorial Guinea	75.4	0.0	24.6	0.0	0.0	0.0	100.0	0.0	0.0
Ghana	50.0	39.7	0.0	10.3	0.0	100.0	0.0	0.0	0.0
Guinea-Bissau	50.0	13.8	0.0	10.9	25.3	56.5	43.5	0.0	0.0
Honduras	53.0	23.0	24.0	0.0	0.0	100.0	0.0	0.0	0.0
India	50.0	12.5	12.5	12.5	12.5	50.0	16.7	16.7	16.7
Indonesia	50.0	1.4	1.4	13.1	34.1	25.0	25.0	25.0	25.0
Kenya	50.0	5.9	5.9	5.9	32.2	66.3	6.2	21.2	6.2
Madagascar	50.0	0.0	0.0	1.3	48.7	100.0	0.0	0.0	0.0
Malawi	50.0	12.5	12.5	12.5	12.5	14.6	14.6	14.6	56.2
Mali	63.7	35.3	0.3	0.3	0.3	0.3	0.3	0.3	99.0
Mozambique	50.0	0.0	0.0	0.0	50.0	0.0	0.0	100.0	0.0
Nigeria	50.0	12.5	12.5	12.5	12.5	29.5	8.4	32.0	30.2
Sao Tome and Principe	50.0	12.5	12.5	12.5	12.5	25.0	25.0	25.0	25.0
Senegal	50.0	50.0	0.0	0.0	0.0	100.0	100.0	100.0	100.0
Togo	55.8	1.0	0.0	43.2	0.0	2.2	0.0	97.8	0.0
Average from the 52	50.9	16.5	17.0	7.3	8.3	68.9	8.2	14.3	8.6
St. Dev from the 52	4.0	15.7	17.8	10.7	12.5	37.5	18.6	26.0	18.1
Max from the 52	75.4	50.0	50.0	50.0	50.0	100.0	100.0	100.0	99.0
Min from the 52	50.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0

Table 4: Virtual weights attributed to the variables by some of the countries (%)

Note: The input weights of the countries classified as inefficient have been standardised to add to 100%, in order to allow for easier interpretation.

As shown in Table 4, several different output weight profiles can be identified. Whilst some countries placed all the output weight in one of the outputs (e.g. Brazil, Equatorial Guinea, Mozambique), other countries distributed the weights between several or all outputs (e.g. Botswana, Cameroon, Kenya, Nigeria). The analysis of the results for the 52 countries shows that a large number of countries (i.e. 18 countries) have placed all the output weight on the output variable: "Reported number of pregnant women tested for HIV". Brazil and Madagascar are two examples of these countries: they have been classified as efficient, and placed 100% of the output weight on this output. A minority of countries also present an unbalanced profile of output weights, by placing all their weight on one of the other three outputs. For example, whilst Equatorial Guinea has put all its output weight on the variable "Number of pregnant women living with HIV who received antiretrovirals for preventing mother-to-child transmission", Mozambique has placed all its output weight on the variable "Reported number of infants born to women living with HIV receiving antiretrovirals for preventing mother-to-child transmission". Contrasting with this, we find four countries classified as efficient (Central African Republic, Chad, Indonesia and Sao Tome and Principe) that present a totally balanced weight profile regarding service provision, giving 25% of the weight to each output. The remaining countries present weights unequally distributed between two, three or all outputs.

Table 4 shows that different input weight profiles can also be identified. At this stage it is important to recall that virtual weight restrictions were imposed to prevent countries from attributing less than 50% of the total input weight to the total spending on prevention of HIV transmission from mother-to-child. It is possible to verify that the weight attributed to this input varies between 50% and 75.4%, with the large majority of countries attributing the minimum weight required. The remaining part of the input weight is distributed between the four environmental variables which account for socio-cultural, political and economic factors. In this respect, it is possible to find different weight profiles within these environmental variables, suggesting that some countries share some sociocultural, political or economic commonalities. For example, Cameroon, India, Malawi, Nigeria, and San Tome and Principe distribute the environmental input weight evenly between the four variables. Opposite to this, some countries place all their environmental weight on a single variable. For example, Senegal and Brazil place all their environmental weight on the input capturing the literacy rate, whilst Azerbaijan and Cambodia place all their environmental weight on the number of people living in urban areas. As expected, countries with considerable social and political unrest, such as Chad, Democratic Republic of Congo, Côte d'Ivoire, Central African Republic and Colombia, place a considerable amount of weight on the input capturing this aspect.

Another interesting piece of information provided by DEA relates to the identification of benchmarks for each of the inefficient countries. Benchmarks are units that are classified as efficient when applying the optimal weight structure of the inefficient units under analysis. For example, as illustrated on Table 5, Honduras and El Salvador, two neighbouring countries who share several sociocultural and economic commonalities, have similar sets of benchmarks for learning. These two countries were classified with below average efficiency rates suggesting great potential for improvement in this context. On one side, Honduras achieved 48.8% in terms of pure technical efficiency and its benchmarks for learning are Côte d'Ivoire, Niger, Madagascar and Malawi. On the other side, El Salvador achieved 43.59% in terms of pure technical efficiency and its benchmarks for learning are Côte d'Ivoire, Niger and Madagascar. In developing their strategies, policies and programmes for HIV prevention, Honduras and El Salvador can get inspiration from the policies and strategies developed by Côte d'Ivoire, Niger and Madagascar. These countries were identified as benchmarks for Honduras and El Salvador because they share commonalities in their social and economic structures, suggesting that similar policies can be effective in fighting vertical transmission of HIV. Honduras and El Salvador are two countries characterised by a high percentage of the population living in rural areas, below average literacy rates, low rates of contraception usage and high rates of teenage pregnancies, combined with fragmented, inefficient and inequitable health care systems which limit the access to services (World Health Organization [50, 51]). With specific relevance to the context of HIV prevention is the fact that these two countries have low coverage of family planning and low pre-natal, childbirth and post-natal care, as well as deficient training in public health, resulting in low levels of maternal education (World Health Organization [50, 51]). If we inspect the optimal input weight structure of Honduras and El Salvador, we can confirm that these two countries distribute all their environmental weight between the variable capturing literacy rate and the variable measuring urban population. From another side, if we inspect the output weight structure of these countries, we conclude that all the output weight was put on the number of women tested for HIV. This means that, even with deficient coverage, this is the service in which

these countries appear to be focusing. Their relative performance in terms of the other three outputs (related with access to targeted medication) is even worse.

If we inspect the strategies, policies and programmes developed by their benchmark countries, we can get inspiration for improvement in Honduras and El Salvador HIV prevention systems. Côte d'Ivoire, Niger and Madagascar (the three common countries which are benchmarks for Honduras and El Salvador) are also characterised by a high level of rural population and low literacy rate, resulting in challenges regarding access to ante-natal care (World Health Organization [52]). However, these three countries have developed inspiring policies and programmes to address these challenges. Due to the effectiveness of these programmes, steady progress has been achieved in terms of PMTCT (especially in terms of HIV testing). These good results allowed them to be identified by our DEA model as robust best practice examples. These policies and programmes have in common the focus on improving health coverage through the involvement of several actors in the health sector and the community. In these countries, innovative programmes were developed to address existing inequalities in access to services, such as the 'mother and child health week' introduced in Madagascar in October 2007 which provides an opportunity for female HIV testing throughout the country (UNICEF [53]). Examples of other strategies and innovative programmes are discussed in detail in the next section.

In addition to the information regarding the benchmarks from which the inefficient countries can learn, Table 5 also provides useful information to define the targets to be achieved by these countries. This information is contained in the lambda values displayed in each of the rows of Table 5. For example, the values of the lambdas associated with El Salvador (i.e. 0.382, 0.468 and 0.15) represent the proportion of the service levels of Cote d'Ivoire, Madagascar and Niger that El Salvador is required to achieve to become efficient.

Table 5: Peers and lambdas for some countries

	Brazil	Cameroon	Côte d'Ivoire	India	Madagascar	Malawi	Mozambique	Niger	Sao Tome and Principe	Thailand
Angola	0.017					0.91	0.073			
Benin		0.166	0.008		0.056			0.77		
Burkina Faso			0.266			0.115		0.62		
Chile	0.112		0.223		0.665					
El Salvador			0.382		0.468			0.15		
Equatorial Guinea			0.013			0.019			0.968	
Ghana			0.884	0.017	0.099					
Honduras			0.128		0.046	0.3		0.526		
Mexico	0.8	0.154		0.046						0.223
Senegal			0.868					0.132		
Vietnam				0.056	0.369					0.575
Number of times country <i>i</i> is used as a peer	9	14	12	9	17	11	2	10	6	3

Whilst some countries serve as benchmarks for a large number of countries, others are classified as efficient but do not serve as a reference for other countries. From the 52 countries compared, Botswana, Chad, Democratic Republic of the Congo, Indonesia, Kenya, Lesotho, Mali, Nigeria and Rwanda have been classified as efficient but do not serve as references to any other country. These countries have chosen very peculiar weight structures, which means that they have been classified as efficient because there are no other countries to which they can be compared. In contrast, as presented in Table 5, there are five countries that serve as benchmarks to 10 or more countries, presenting robust efficiency status. These countries are: Madagascar, Cameroon, Côte d'Ivoire, Malawi and Niger. These results indicate that the policies and service delivery strategies used by these countries should be identified in order to provide guidance for other countries. We will now discuss some of these strategies.

3.3. The practical and policy implications of the results

Despite the exploratory nature of this research and data limitations, some important lessons can be derived from the results previously discussed. In particular, the results highlight that there is considerable potential to save the lives of thousands of children each year by improving PMTCT services across countries. In addition, the results constitute a valuable source of information to policy makers, programme planners, managers, and implementers. By identifying best practices, the study points out those countries whose HIV prevention strategies ought to be investigated in order to develop sound strategies that can be replicated in those nations where HIV prevention strategies have been less successful.

Two of the countries that represent important learning peers for other nations are Cameroon and Côte D'Ivoire. These countries have been previously identified as important best practices in the area of PMTCT (Family Health International [54], Welty et al. [55]) and some of their service delivery strategies have been documented (Bolu et al. [56]). According to our results, together, these two countries serve as benchmarks to more than 20 of the nations assessed. Considering that the prevention programmes followed by these two countries have proven their effectiveness in battling HIV transmission from mothers to their infants, we believe that useful insights can be derived from their analysis.

Cameroon, which the DEA results indicate to be a benchmark to nations like Benin, Congo and Togo, can be an important example and inspiring model to these countries on how to reduce the incidence of HIV infection in children by involving the community in the scaling up of PMTCT services across the countries, including rural areas. By adopting a bottom-up approach Cameroon ensured the full involvement of the community and increased consensus, ownership and commitment towards successful PMTCT implementation. As identified by the AWARE-HIV/AIDS Project (Family Health International [54]), the success achieved by Cameroon is due, however, to several interrelated and complementary strategies. One of these strategies consists of using trained volunteers as community educators to help disseminate information regarding HIV/AIDS. This community education has been found to be an important forerunner of PMTCT as it is easier to provide counselling and testing to women who are knowledgeable about HIV/AIDS. Another important strategy adopted by the Cameroonian health authorities consists of the training of health workers in order to allow integration of PMTCT services in routine ante-natal care. This strategy, together with voluntary HIV counselling, an opt-out approach to HIV testing, and same day results for pregnant women in antenatal care, have

significantly increased the number of those who carry out the tests and who receive their results. This has proved fundamental in discovering the HIV status of pregnant women and assisting health care authorities to provide appropriate care to HIV-positive women and their newborns. The provision of antiretroviral therapy to HIV-positive pregnant women and the provision of counselling on infant feeding have proved critical to reduce the risk of mother-to-child transmission of HIV. Another strategy which has been fundamental in battling HIV transmission from mothers to children consists of involving local communities in service delivery through 'trained birth attendants' and through support groups involving HIV infected women (Wanyu [57]). This strategy has been particularly valuable and effective in the scaling up of PMTCT, by allowing psychosocial support and infant feeding counselling to reach more women in need.

These strategies have proved very effective in Cameroon, and therefore, we believe that their dissemination and replication in other countries could help them improve the efficiency and effectiveness of their prevention policies (see, for example, Msellati [31], Bolu et al. [56]). However, as demonstrated by the experience of Cameroon, for these strategies to produce the intended results it is fundamental that a clear and strong political leadership commitment exists to fight against HIV/AIDS. Furthermore, PMTCT needs to be seen as a major priority. Unless strong and available leadership gets involved early enough in the coordination of the implementation process, supervises the efforts on the ground through regular and effective monitoring and evaluation of PMTCT activities, and ensures the quality of the services provided, the strategies discussed previously are likely to fail.

The second example of a successful implementation of a PMTCT strategy is offered by Côte D'Ivoire and it illustrates how to integrate PMTCT training for health service providers in the curricula of training institutions in order to allow health care workers to take proper care of pregnant women, and in this way, battle HIV transmission from mother-to-child (Ministère de la Santé et de l'Hygiène Publique [58], Ministère de la lutte contre le SIDA [59]).

It is widely acknowledged that information, education and training are critical components of any programme targeted at preventing mother to child HIV transmission. Côte D'Ivoire has been successful in promoting these components, and therefore, the strategies adopted in this country also offer considerable scope for replication. Whilst the lessons learned from the experience of Côte D'Ivoire may be useful for many countries, we believe they are particularly relevant to those countries which the DEA analysis identified as having Côte D'Ivoire as a benchmark. These countries include, among others, Benin, Burkina Faso, Equatorial Guinea, El Salvador, Ghana, Honduras and Senegal.

Côte D'Ivoire experience, which is discussed in further detail in Family Health International [54], can help these countries develop and implement HIV sustainable PMTCT approaches grounded on a participatory process and targeted at promoting long-term quality training in health care training institutions. In fact, the process followed by Côte D'Ivoire, which consists of several steps, can work as a guide for other countries. The first step consists of carrying out a needs assessment to evaluate existing PMTCT pre-service and in-service training materials for health care workers and to determine how best to integrate PMTCT into pre-service and in-service health training curricula. Once this assessment has been conducted, good practice consists of holding a stakeholders meeting to share the findings and to create a steering committee to oversee, monitor progress and provide leadership in the development and standardisation of PMTCT in-service and pre-service training initiatives. It is then the responsibility of this steering committee to mobilise trainers and opinion leaders, to involve them in the PMTCT roll-out through a participatory process, and to sensitise them to the importance of the prevention plan and training needs. Once universities and training institutions are sensitised, the next step consists of developing PMTCT training materials for service providers. These initiatives, together with the creation of a pool of expert trainers representative of all the pre-service and inservice institutions, can assist countries in providing over the course of time a critical mass of health care professionals capable of mastering PMTCT practices throughout nations.

These two examples illustrate that DEA has strong potential to contribute to performance improvement in the context of PMTCT if used with formative purposes. By using DEA, policy makers and health care managers can identify which countries have been most successful in using the limited resources available to prevent the transmission of HIV from mothers to children. However, the identification of best practice is only the first step in the process. It is then fundamental that these practices are carefully analysed and documented in order to be disseminated and replicated.

Although our empirical analysis covers only a single year and a limited number of countries, it is important to emphasise that the results achieved still provide useful information for policy making, which can contribute to the development of programmes aimed at improving the efficiency of PMTCT.

4 – Conclusion

In spite of the remarkable progress that has been made in recent years in battling the HIV/AIDS epidemic, this disease still poses huge challenges for individuals, households and nations. It is estimated that more than 1,000 children are newly infected with HIV every day, and of these more than half will die as a result of AIDS because of a lack of access to HIV treatment [1]. Over 90 percent of the children living with HIV were infected through vertical transmission from mother to the baby during pregnancy, delivery or breastfeeding. Considering that there is compelling evidence that appropriate prevention and care can reduce the rate of these transmissions significantly, it is imperative that nations ensure that this prevention and care reaches all pregnant women and children in need and that the resources are used as efficiently as possible.

In the academic literature there have been important contributions exploring the efficiency and cost-effectiveness of different HIV prevention programmes, including mother-to-child transmission. Most of these contributions have, however, been narrowed to a small region, a single country or parts of it, and have

tended to rely on the use of a set of individual indicators detailing different aspects of service delivery. In this paper, we have explored the potential of using DEA to complement the existing literature in the area. To this effect we have proposed a model, using data from 2008, with five inputs and four outputs to compare the efficiency of 52 low- and middle-income countries in preventing mother-to-child HIV transmission.

Despite the exploratory nature of the research, there are some important empirical findings from our study. In particular, we found wide variation in terms of the efficiency of service provision across nations, indicating that some countries have been considerably more efficient in allocating resources to preventing HIV transmission from mother-to-child than others. This variation ought to be investigated as our results suggest that if all countries were using their resources efficiently, the level of service provision could increase by around 40 percent. Considering that the analysis reported in this paper also allows the identification of suitable learning peers for each nation, as well as targets for performance improvement, we believe it can be an important catalyst for a deeper formative assessment aimed at identifying the causes of poor performance, exploring the practices of the best performers; and based on this assessment to develop sound strategies for HIV prevention improvement.

The formative implementation of DEA in this context, however, faces some challenges. One of the most relevant is, perhaps, the lack of availability of data and/or poor data quality. There are many countries for which there is no data, and even when data exists it does not cover all the important variables, does not cover more than a year, or does not guarantee accuracy of information. Data unavailability might prevent the specification of appropriate models or might prevent the inclusion of countries in the analysis which could eventually change the efficiency frontier. As emphasised by Pedraja-Chaparro et al. [60], a shortage of observations can bias the DEA results as some parts of the feasible activity domain may not be properly modelled. When this happens, some countries may be regarded as efficient simply because there are no peers with which to compare them. Furthermore, when no panel data exists, it is not possible to perform dynamic assessments targeted at analysing eventual changes in the performance of

nations over time. In turn, poor data quality can also significantly impact the DEA results. As the technique relies on outlier DMUs to construct an efficient frontier, a single error of measurement can lead to a significant adjustment in the estimated frontier. In order to detect particularly influential countries, and therefore, those countries that might require careful scrutiny of the data to prevent measurement errors, we can use the super efficiency measure as suggested by Anderson and Petersen [61], or we can count the number of times an efficient country appears in the peer group of inefficient countries as indicated by Pedraja-Chapparo et al. [60]. When doubts exist regarding the quality of the data, the analyst or decision maker can also carry out a careful sensitivity analysis in order to construct data ranges within which results remain unchanged.

Another important challenge relates to the need to include weight restrictions in the models. In developing DEA models to assess the efficiency of countries in implementing particular HIV prevention policies it is important to decide whether it is acceptable to allow full flexibility in the choice of input and output weights. Whilst several approaches to define weight restrictions exist, the specification of appropriate weight restrictions can be a very challenging task. Research on how to express weight restrictions that are economically valid and that derive from the preferences of the stakeholders is very scarce.

In the particular context under study, it is our belief that an early involvement of the key stakeholders in the process, together with the procedures suggested above to address data quality issues, are fundamental aspects to consider to strengthen the DEA analysis and allow firmer policy lessons. Furthermore, whenever data permits, it can be insightful to form clusters with countries sharing socio-cultural, political and economic commonalities and perform disaggregate analyses for these clusters.

It is important to emphasise that despite the fact that the successful application of DEA can be a challenge, DEA has strong potential to contribute to performance improvement in this context. In particular, DEA is well suited to complement efficiency analyses undertaken by means of other methodologies.

In conclusion, we believe that DEA is a strong analytical technique upon which to build and which can play an important role in the development of efficient HIV prevention interventions. This paper has discussed some of the insights that derive from the use of DEA in this particular context. However, one of the premises of the authors of this paper is that DEA has to be broadened in order to contribute to performance improvement in practice. One of the main aspects of this broadening relates to the need to move away from the 'black box' type of evaluation, by incorporating the DEA exercise into case studies and context-driven research projects in order to facilitate the development of appropriate models and the implementation of the results in practice. Unless the results of these models are taken into consideration by decision makers and incorporated into the policy making process, their purposes are not served. Further research is being carried out by the authors in order to meet these objectives.

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Appendix

Multiplier version of the DEA model used in the analysis (includes one weight restriction)

$$\frac{1}{E_0} = Min \sum_{i=1}^5 V_i X_{i0} - w_0;$$

Subject to:

$$\sum_{r=1}^{4} U_r Y_{r0} = 1$$

$$\sum_{r=1}^{4} U_r Y_{rj} - \sum_{i=1}^{5} V_i X_{ij} + w_0 \le 0 \qquad (j = 1,...,52)$$

$$V_1 X_{10} - \sum_{i=2}^{5} V_i X_{i0} \ge 0$$

$$U_r; V_i \ge \varepsilon > 0$$

$$w_0 \text{ free in sign} \qquad (i = 1,...,5); (r = 1,...,4)$$

Where E_0 is the relative efficiency of DMU₀, X_{i0} is the quantity of *Input i* used by DMU₀ (with $i = 1 \dots 5$) and Y_{r0} the quantity of *Output* r produced by DMU₀ (with $r = 1, \dots, 4$), V_i is the weight attributed to *Input i*, U_r is the weight attributed to *Output r*, and w_0 is the scale variable for DMU₀.