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Are Emissions Trading Policies Sustainable? A Study of the Petrochemical Industry in Korea

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Abstract: In 2015, Korea inaugurated an emissions trading scheme (ETS). In this regard, many studies have considered the sustainable performance and efficiency of industries that emit carbon; however, few have examined ETS at company level. This paper focuses on companies' data related to Korean ETS in the petrochemical industry. Based on the non-radial, nonparametric directional distance function (DDF), the paper evaluates the governance factors related to ETS policies and sustainable performance in terms of carbon technical efficiency (CTE), the shadow price of carbon emissions, and Morishima elasticity between the input and undesirable output of carbon emissions. Using a dual model, the paper shows that Korean ETS has huge potential for participating companies to improve CTE. If all companies consider the production possibility frontier, they could potentially improve efficiency by 52.8%. Further, Morishima elasticity shows strong substitutability between capital and energy, implying that green technology investment should bring a higher degree of energy-saving performance. Unfortunately, however, the market price of carbon emissions is far too low compared with its shadow price, suggesting that the Korean government's price-oriented market intervention has resulted in the ETS producing poor sustainable performance. As the title suggests, ETS of Korea is not sustainable at the current stage, but with more efforts on the transition period, all the developing countries should support the governance factors of the ETS in terms of the more effective green investment with easier access to the green technology.

Keywords: ETS; emissions trading scheme; carbon technical efficiency; CTE; shadow price; Morishima elasticity of substitution; governance

1. Introduction

Since the historic meeting of the United Nations Framework Convention on Climate Change (UNFCCC) in 1994, numerous efforts have been made to find the optimal path toward worldwide sustainable development. At Paris in 2015, the 21st session of the Conference of the Parties (COP) agreed to keep the increase in the global average temperature to below two degrees centigrade above the pre-industrial level. The session also agreed to pursue efforts to limit the temperature increase to 1.5 degrees centigrade above the pre-industrial level. These agreements recognized that temperature-control efforts would significantly reduce the risks and impacts of climate change. In order to achieve the UNFCCC's goals, all 195 member countries are required to make clear and measurable efforts for worldwide sustainable development. As one of the leading countries, Korea has promoted its version of green growth policies since the government hosted the G20 Seoul Summit in 2010. Since then, using one of the most powerful market-oriented frames, the Korean government has prepared and promoted its nationwide emissions trading scheme (ETS). This scheme became regulated in 2015; however, it has caused diverse conflict among politicians, industrial leaders, and even academic experts. Thus, the government has provided a cautious pilot scheme that involves

a weak and constrained ETS. Emissions trading rights are traded only among regulated companies. Moreover, emissions targets have been initially set at 95% of the regime. Because of this loose, yet uncomfortable, arrangement among ETS interest groups, it is easy for regulated companies not to invest in the sustainable performance of green productivity and to avoid the extra burden of carbon targets in the short run. Consequently, even though many trials have evaluated the effectiveness of ETS-related policies, it is difficult to measure the performance or efficiency of such policies from a field-oriented company perspective. This paper aims to analyze, from a company perspective, whether the milestone policy of ETS in Korea is successful in terms of decreasing carbon emissions efficiently. Since carbon emissions form most greenhouse gas (GHG) emissions, we evaluate the GHG emissions of companies based on their efforts to decrease ETS targets.

In 2012, total GHG emissions in Korea were 688.3 million tons. This figure represents a 133% increase compared with the base year of 1990. Carbon dioxide (CO₂) emissions overall were 593 million equivalent tons, a figure that places Korea as the seventh highest producer of such emissions in the world. Carbon emissions per person were 11.7 tons, which places Korea as the 18th highest producer of carbon emissions per person in the world and places it sixth among Organization for Economic Cooperation and Development (OECD) countries. These rankings imply that many other developing countries are more proactive than Korea; thus, Korea should be more aggressive in its attempts to decrease GHG emissions and carbon dioxide. The Korean government set its own target in November 2009, saying that it would reduce GHG emissions by 30% from BAU (business as usual) levels by 2020. In the Paris agreement of 2015, the Korean government decided to reduce GHG emissions by 37% from BAU levels by 2030. In order to achieve this target, the government published an emissions trading master plan (2015–2025) in 2014. It also implemented its ETS nationwide in 2015. Unfortunately, however, in spite of these proactive government efforts, the carbon trading market is regarded as a failure because very few transactions took place in the first year. Many experts blame the unreliable carbon price for this failure. They say that the carbon price is so low that companies with emissions permits are reluctant to sell them. Moreover, regulated companies are uninterested in green technology investment because of the low opportunity costs for such innovation. During most of 2015, the carbon price barely exceeded US \$10, resulting in the lack of trading. Thus, it is important to analyze whether this price is appropriate or not for achieving the future sustainable performance of the Korean ETS. This issue is the basic motivation of the current research.

In order to monitor and measure efficient carbon reduction, many researchers have conducted carbon efficiency studies based on the distance function because of its reliable clarity about relative efficiency. Although there are numerous sustainable performance modeling methods, the distance function approach has become popular because it can simultaneously model joint-production technology with eco-friendly desirable and undesirable output. Another benefit of the distance function approach is that compared with other cost functions, it does not require any price-specific functional form that is relatively difficult to obtain. Given the quantity of data for input and output, which are comparatively easy to obtain, various critical environmental production characteristics can be formally analyzed. These include environmental technical efficiency, environmental productivity growth, the shadow prices of pollutants, and intra-factor substitution possibilities [1].

Even if Shepherd distance function provides a basic logical frame for the treatment of multi-input/output analysis [2], it is limited in the sense that it treats desirable and undesirable output proportionately; thus, it overlooks the possibility that it decreases undesirable output without any loss in desirable output. Consequently, most directional function research uses the directional distance function (DDF) introduced by Chambers, Chung, and Färe [3]. According to the literature review of Zhang and Choi [1], more than 100 papers have analyzed energy efficiency and the shadow prices of industrial emissions by using distance function models. Studies that use the DDF method employ either the parametric method [4] or the nonparametric method based on data envelopment analysis (DEA) [3,5]. The former can easily calculate the shadow price but has to compute the theoretic distance function coefficients; the latter does not need to preset functional forms and thus can analyze

the shadow prices of pollutants in all directions with significant flexible implications. Since it is not easy to accept a specific production possibility curve a priori without significant in-depth analysis on the characteristics of the decision units, this research is based on the DDF_DEA.

Most early research on the DDF is based on a radial model that estimates efficiency proportionately for all output [3]. However, several later studies have reported limitations regarding this radial approach. First, radial measures may overestimate efficiency when slack exists [6]. Second, the radial model has relatively weak discriminating power when ranking entities that require evaluation [7]. Third, a radial efficiency measure cannot provide a single-factor efficiency measure, such as energy efficiency, because the DDF can only give the same rate of inefficiency [8]. Because of these limitations, a number of studies have extended the conventional DDF to a non-radial DDF by considering slacks [6,9–11]. Thus, we employ the non-radial DEA-based DDF to examine the sustainable performance of the eco-friendly policies of Korea's ETS.

Most studies in this field have been conducted in the power plant sector [12–16], the steel and iron sector [17–19], the electricity sector [20–22], and other sectors such as the chemical, cement, and ceramic industries. Nearly all of these sectors emit a great deal of CO₂ [23–26]. Although these studies have employed the DDF to measure the performance of environmental policies, to the best of our knowledge no research has yet been conducted using company data related to real ETS trading. Moreover, Korea is the only country in the world to promote a nationwide, and not a regional or localized, ETS frame.

Since Korea is in a transitional stage between a developing and developed economy, this research has insightful implications about a stepwise approach to performance-oriented sustainable development. Moreover, this research has potential practical importance because it is based on a comparison between the shadow price and the price in the real world. Further, it evaluates the sustainability of the unduly cautious ETS policies in Korea. In this paper, the sustainability of ETS policies is analyzed from three perspectives. First, the Korean government's ETS policies are excessively price-oriented; thus, we need to check that the semi-regulated market price of carbon emissions permits is acceptable compared with their shadow prices. If the market price is too low, it may result in low investment in green IT technologies, which is contrary to the government's ultimate goal. Of course, the shadow price of carbon could be unrelated to the market carbon price for one important reason: the carbon allowances are given for free and all firms are allocated 100% of the allowances they need for at least in the first stage of three years. Under these conditions, it is reasonable for carbon allowances to have a low market value in the first few years of the ETS. Hence, it is hard to argue that the shadow prices reflect what the market price should be. However, the government should narrow this gap between the shadow price and market price as early and effectively as possible, and thus if the government stimulates companies to invest more on green technology, the price gap will be more effectively decreased. According to Oestreich and Tsiakas [27], during the first few years of the scheme, firms that received free carbon emission allowances on average significantly outperformed firms that did not in the European Union Emissions Trading Scheme (EU ETS). Even if this kind of transition trial and errors on the inappropriate carbon market price, however, they found that this kind of carbon premium is present for a specific period that commences about one year before the beginning of Phase I and disappears about one year into Phase II [27]. If this is true, then the Korean government should make this trial and error on the carbon market price as short as possible. In this paper, we will check the feasibility of this carbon premium in terms of the price gap between the shadow price and market price.

Second, in order to boost investment in energy-saving green technologies by ETS member companies, there must be significant elastic substitutability between energy and other input. Third, it is necessary to check whether the potential sustainability of ETS policies is increasing the eco-friendly efficiency of participating companies. If not, such companies have no incentive to exert too much effort in proactively participating in ETS trading.

Thus, the purpose of this paper is to examine these three points regarding the sustainability of governmental policies. In this context, we use the ETS data of participating member companies.

However, the Korean ETS covers 525 companies from diverse industries. Consequently, since the clarity and credibility of the implications are important in order to check the sustainability of the ETS policies, this research focuses on the petrochemical industry. This industry is the largest in the ETS with 85 companies, which represent 16% of all ETS members. Thus, based on the arguments regarding an examination of the ETS policies' sustainability, this paper estimates the eco-friendly efficiency of the participating member companies in the major field of the petrochemical industry. It compares the shadow price with the market trading carbon price and considers the potential of green investment in terms of Morishima elasticity between capital and energy.

The rest of this paper is organized as follows: Section 2 presents a methodological framework of the non-radial DDF models with Morishima elasticity and describes the data collection. Section 3 presents the results of the empirical study of the petrochemical industry. Section 4 concludes with some policy implications for sustainable governance.

2. Methodology

In this section, the non-radial DDF model is presented. This model shows how to estimate carbon technical efficiency. Moreover, a dual model is introduced to calculate the shadow price of undesirable output. Then, the Morishima elasticity of input is articulated from the ratio of the shadow price.

2.1. Carbon Technical Efficiency

In order to introduce the non-radial DDF, the term “environmental production technology” should be defined. Assume that there are $j = 1, \dots, N$ decision-making units (DMUs). These DMUs are petrochemical manufacturing companies in our research. Suppose that each DMU uses input vector $x \in \mathbb{R}_+^M$ to produce jointly a good output vector $y \in \mathbb{R}_+^S$ and a bad output vector $b \in \mathbb{R}_+^J$. Environmental production technology is expressed as

$$T = \{(x, y, b) : x \text{ can produce } (y, b)\}, \quad (1)$$

where T is often assumed to satisfy the standard axioms of production theory. Inactivity is always possible, and finite amounts of input can produce only finite amounts of output. In addition, input and desirable output are often assumed to be freely disposable. With regard to regulated environmental technologies, weak disposability must be imposed on T [28]. The weak disposability assumption implies that reducing bad output, such as CO₂ emissions, is costly in terms of relative decreases in production. Further, the null-jointness assumption implies that CO₂ emissions are unavoidable in production and that the only way to remove CO₂ is to stop production. Mathematically, these two assumptions can be expressed as follows:

- (1) if $(x, y, b) \in T$ and $0 \leq \theta \leq 1$, then $(x, \theta y, \theta b) \in T$;
- (2) if $(x, y, b) \in T$ and $b = 0$, then $y = 0$.

The DEA piecewise linear production frontier is used to construct environmental production technology. Then, regulated environmental technology T_1 for N DMUs exhibiting constant returns to scale can be expressed as

$$T_1 = \{(x, y, b) : \sum_{n=1}^N z_n x_{mn} \leq x_m, m = 1, \dots, M, \\ \sum_{n=1}^N z_n y_{sn} \geq y_s, s = 1, \dots, S, \\ \sum_{n=1}^N z_n b_{jn} = b_j, j = 1, \dots, J, \\ z_n \geq 0, n = 1, \dots, N\}. \quad (2)$$

A formal definition of the non-radial DDF is proposed by Zhou et al. [9] with undesirable output as

$$\vec{D}(x, y, b; g) = \sup \{ \mathbf{w}^T \beta : ((x, y, b) + g \cdot \text{diag}(\beta)) \in T \}, \quad (3)$$

where $\mathbf{w} = (w_m^x, w_s^y, w_j^b)^T$ denotes a normalized weight vector relevant to the amount of input and output, $g = (-g_x, g_y, -g_b)$ is an explicit directional vector, and $\beta = (\beta_m^x, \beta_s^y, \beta_j^b)^T \geq 0$ denotes the vector of scaling factors. The value of $\vec{D}(x, y, b; g)$ under environmentally regulated technology can be calculated by solving the following DEA-type model:

$$\begin{aligned} \vec{D}^r(x, y, b; g) &= \max w_m^x \beta_m^x + w_s^y \beta_s^y + w_j^b \beta_j^b \\ \text{s.t.} \quad &\sum_{n=1}^N z_n x_{mn} \leq x_m - \beta_m^x g_{xm}, m = 1, \dots, M, \\ &\sum_{n=1}^N z_n y_{sn} \geq y_s + \beta_s^y g_{ys}, s = 1, \dots, S, \\ &\sum_{n=1}^N z_n b_{jn} = b_j - \beta_j^b g_{bj}, j = 1, \dots, J, \\ &z_n \geq 0, n = 1, 2, \dots, N \\ &\beta_m^x, \beta_s^y, \beta_j^b \geq 0. \end{aligned} \quad (4)$$

The directional vector g can be set up in different ways, based on given policy goals. If $\vec{D}(x, y, b; g) = 0$, then the specific unit to be evaluated is located on the frontier of best practices in the direction of g . Carbon technical efficiency (CTE) is defined based on the non-radial DDF. Here, there are three inputs, one desirable output, and one undesirable output. We set the weight vector of S , M , and J as one-third each for input, desirable output, and undesirable output respectively. We also set the directional vectors as $g = (-x, y, -b)$, based on Zhou et al. [9].

According to Zhang and Xie [23], the overall eco-friendly efficiency or CTE for industries is defined as the average efficiency of each factor. Suppose that β_x^* , β_y^* , and β_b^* represent the optimal solutions to Equation (5), then the CTE can be formulated as:

$$CTE = 1 - \frac{1}{M + S + J} \left(\sum_{m=1}^M \beta_{xm}^* + \sum_{s=1}^S \beta_{ys}^* + \sum_{j=1}^J \beta_{bj}^* \right) \quad (5)$$

2.2. The Shadow Price of Undesirable Output and the Morishima Elasticity of Substitution

In this subsection, a dual DDF model is used to estimate the shadow prices of CO₂ and the elasticity of substitution for input. The shadow cost function of a non-radial DDF may be defined as

$$\begin{aligned} &\min vx_0 - uy_0 + rb_0 \\ &\text{s.t.} \\ &vx - uy + rb \geq 0 \forall n \\ &v \geq \left[\frac{1}{g_1^x}, \dots, \frac{1}{g_m^x}, \dots, \frac{1}{g_M^x} \right] \\ &u \geq \left[\frac{1}{g_1^y}, \dots, \frac{1}{g_s^y}, \dots, \frac{1}{g_S^y} \right] \\ &r \geq \left[\frac{1}{g_1^b}, \dots, \frac{1}{g_j^b}, \dots, \frac{1}{g_J^b} \right]. \end{aligned} \quad (6)$$

In Equation (6), $v \in R^m$, $u \in R^s$, and $r \in R^j$ are the dual-variable vectors of the input ($x \in R^m$), the desirable output ($y \in R^s$), and the undesirable output ($b \in R^j$) respectively. The dual variables of the input, desirable output, and undesirable output can be estimated by the linear programming of Equation (6). The dual Equation (6) aims to minimize the virtual cost of the specific company concerned. In this regard, the dual DDF model is a type of product maximization model in which the virtual cost is at best zero (non-positive) when $\vec{D}^u(x, y, b; g) = 0$ for the DDF-efficient unit.

The dual variables $v \in R^m$ and $r \in R^j$ can be interpreted as the shadow prices of the input and undesirable output, respectively. $u \in R^s$ denotes the marginal virtual income of the desirable output. Assuming that the absolute shadow price of the undesirable output is equal to its market price (p^b), the relative shadow price of the undesirable output with regard to the desirable output (p^y) can be measured by

$$r = u \times \frac{p^b}{p^y} \quad (7)$$

In other words, the shadow price of the undesirable output can be interpreted as a marginal abatement cost that represents the marginal rate of transformation between undesirable and desirable output. Under the environmental regulations, the abatement of pollutants is not free but costly for companies because they incur an opportunity cost associated with reducing desirable output.

The curvature of the isoquant curve reflects the degree of substitutability of the input factors in the production function. Following Lee and Zhang [29], the elasticity of substitution between input x_i and x_j can be estimated by employing the idea of indirect Morishima elasticity of substitution, as shown in Equation (8). Morishima elasticity is defined as the shadow price ratio between the two factors [30]. As shown in Table 1, most research on Morishima elasticity is based on the distance function approach with input-orientation. This is because the substitution between traditional input, such as capital and labor, and energy input is crucial in order to find the potential of energy-saving efficiency enhancement through additional investment in energy-saving technologies. Based on the table, this paper analyzes the input elasticity of substitution among three types of input: capital, labor, and energy. It also examines the shadow price of the undesirable output of carbon emissions in accordance with the literature. In the simplest format, Morishima elasticity for input captures the degree to which the relative shadow prices of types of input should be altered to allow substitutability among such input along the isoquant curve. A high value for Morishima elasticity indicates low-level substitutability. It should be noted that $M_{ij} \neq M_{ji}$ because the ratios related to the two input shadow prices differ depending upon which input is used as the basis. The degree of substitution of x_j for x_i does not coincide with the substitution of x_i for x_j in general. Further, even if Morishima elasticity can be calculated for all three types of input and output perspective, we focus only on the elasticity between input and undesirable output. This is because we want to examine the potential of green technology investment for the sustainability of the ETS policies. It should also be noted that this application of Morishima elasticity is especially important in our research field of the capital-intensive petrochemical industry. Thus, it may provide useful insights with regard to the green investment promotion effect of the ETS policies. Hence,

$$M_{ij} = \frac{v_i}{v_j} \quad (8)$$

Table 1. Literature that considers Morishima elasticity.

Researchers	Sector	Finding	Model
Färe et al. [31]	Electric utilities	Output elasticity of substitution between electricity and sulfur dioxide (SO ₂), shadow price of SO ₂	Distance function
Lee and Zhang [29]	Manufacturing industry	Input elasticity of substitution among energy and non-energy shadow price of CO ₂	Distance function
Lee and Jin [30]	Electric power generation industry	Input elasticity of substitution between thermal capital and nuclear capital and their shadow prices	Distance function
Zhang et al. [32]	Poyang Lake	Input and output elasticity of substitution, shadow price of carbon	Slacks-based measure (SBM) DEA
Zhang and Xie [23]	Health information management (HIM) industry	Input and output elasticity of substitution, shadow price of carbon	Distance function

3. Characteristics of Data and Empirical Results

3.1. Data

At the initial stage of the ETS, the Korean government selected 525 ETS member companies from all industrial fields, as shown in Table 2. These companies are substantial carbon emitters with more than 1.25 billion tons emitted per business, or 25,000 tons per factory site, as calculated from three-year averages of 2012 to 2014. During the initial ETS stage, from 2015 to 2017, all these companies are allowed 100% free permits. As long as the companies maintain carbon production at the same levels as the prior three years, they have no obligation to buy carbon allowance permits. However, because production volume increases every year, the companies must emit less carbon from the increased production processes. Since the free allowance permit is excessively generous at this initial stage, there has been no significant carbon trading. In the first year (2015), the volume of carbon trading was less than 1%, and the average price was about US \$5 per carbon ton. The reason is that many ETS member companies are export-oriented; thus, they are concerned that because of the unilateral ETS burden, they may lose their price competitiveness. Such companies believe that the government should consider this issue; thus, the government has established an initial stage goal at a modest average of the BAU level. In order to avoid unexpected third-party manipulation, only ETS member companies can buy and sell the carbon allowance permits. However, in the second stage from 2018 to 2020, member companies will be allowed 97% of emissions targets free of charge; then, certified third parties can participate in carbon trading.

Table 2. The industries and the numbers of companies in the Korean ETS.

Industry	Numbers	Industry	Numbers
Construction	40	Textiles	15
Mining	2	Water	3
Machinery	19	Cement	25
Wood processing	7	Glass	24
Electronic, displays, and semiconductors	45	Food and drink	23
Power and energy	38	Car manufacturing	24
Nonmetallic	24	Oil refining	4
Petrochemical	85	Paper	44
Shipbuilding	8	Steel	40
Telecommunications	6	Waste	44
Aviation	5	Total	525

In order to examine this ETS policy, we collected data from 63 companies in the petrochemical industry in 2014. We chose the petrochemical industry because it is the main group in the ETS that has homogeneity. We used the data to evaluate CTE, the shadow prices of carbon emissions, and the substitutability of types of input by using Morishima elasticity.

With regard to the output variables, we selected sales turnover (T) as a sole desirable output and carbon (C) as an undesirable output. We also selected two basic types of input, labor (L), and capital (K), and added energy (E) as the third input for environmental performance. The data for labor, capital, and turnover were derived from DART (the Data Analysis, Retrieval, and Transfer System) and Alio (Public Information Online). Energy and carbon data were taken from the Greenhouse Gas Inventory & Research Center of Korea. The CO₂ data was unavailable for the research period; thus, we interpolated the numeric values from the GHG emissions data, which includes other gases such as methane (CH₄), nitrogen (N₂O), hydrofluorocarbons (HFCs), perfluorinated compounds (PFCs), and sulfur hexafluoride (SF₆). In general, studies on the energy and environment (E&E) field have extracted pure CO₂ values under the International Panel on Climate Change (IPCC) guidelines by using a macro type of data such as fuel [13,23,33], consumption rate, and so on. However, we focus on the company level; and in Korea, only GHG data is available. Thus, we interpolate GHG data into a company's carbon value because carbon consists of 80% of GHG emissions worldwide, a percentage that according to the UNFCCC is the same in Korea. Because of the scarcity of some data, only 63 of 85 companies were included in the analysis. The basic data statistics for these companies are shown in Table 3.

Table 3. Descriptive statistics.

Variable	Type	Unit	Mean	Std. Dev.	Max.	Min.
Sales turnover	Desirable output	US \$ Billion	2,102,894	3,800,331	18,076,229	59,824
Carbon	Undesirable output	CO ₂ equivalent tons	756,249.6	1,436,648	7,063,768	23,461
Capital	Input	US \$ Billion	110,867	195,610	1,125,781	3666
Labor	Input	Per person	1236.742	2198.649	2669	57
Energy	Input	Terajoules (TJ)	13,438.59	2,626,7.12	134,604	356

Sources: Greenhouse Gas Inventory & Research Center of Korea (<http://www.gir.go.kr/>) [34]. DART: Data Analysis, Retrieval, and Transfer System (<http://dart.fss.or.kr/>) [35]. Alio: Public Information Online (<http://www.alio.go.kr/>) [36].

Table 4 presents the input and output correlation matrix. The results show that the correlation between the types of input and output is positive. Capital and labor are significantly related with desirable output, and energy influences carbon significantly. Thus, an overall increase in input causes an increase in output, which strongly suggests that our approach is feasible for an empirical study.

Table 4. Input and output correlation matrix.

Variable	Capital	Labor	Energy	Turnover	Carbon
Capital	1				
Labor	0.900	1			
Energy	0.565	0.424	1		
Turnover	0.913	0.837	0.491	1	
Carbon	0.554	0.410	0.984	0.478	1

3.2. CTE and the Shadow Price of Carbon

Based on Equations (4) and (5), CTE is calculated in Table 5. The CTE scores range from 0.094 to 1 and the average CTE score is approximately 47.2%. The latter result implies that it is possible to accomplish a 52.8% efficiency enhancement in the petrochemical industry when it operates on the frontier of environmental production technology. Of the ETS group, 12.7% (eight companies) exhibit best practice in CTE, while a further eight companies have CTE levels that are lower than 20%.

This result implies that there is a strong tendency for bipolarization among the leading companies in Korea; thus, the government should take measures to support (or regulate more strictly) those companies with low CTE scores in order to improve their efficiency. Since there is a potential of more than 50% to improve eco-friendly efficiency, more proactive regulation is urgently needed for companies with lower CTE levels to encourage them to make greater efforts to enhance such levels.

Table 5. Carbon technical efficiency and the shadow prices of carbon emissions.

DMU	Efficiency	Shadow Prices	DMU	Efficiency	Shadow Prices
DMU1	0.240	40.56	DMU33	0.203	15.20
DMU2	0.704	57.66	DMU34	1.000	8.94
DMU3	0.108	0.71	DMU35	0.371	5.67
DMU4	0.903	14.80	DMU36	0.226	2.71
DMU5	0.548	0.24	DMU37	0.351	11.20
DMU6	0.157	19.45	DMU38	0.223	38.77
DMU7	0.233	73.15	DMU39	0.581	8.92
DMU8	0.147	28.22	DMU40	1.000	24.11
DMU9	0.217	43.93	DMU41	0.276	7.20
DMU10	0.199	37.18	DMU42	0.608	0.24
DMU11	1.000	53.37	DMU43	0.353	73.46
DMU12	0.273	27.42	DMU44	0.251	69.85
DMU13	0.505	45.66	DMU45	0.232	17.37
DMU14	0.395	18.30	DMU46	0.537	24.27
DMU15	0.195	2.09	DMU47	0.618	6.09
DMU16	0.255	3.05	DMU48	0.260	53.55
DMU17	0.435	13.12	DMU49	0.231	29.40
DMU18	1.000	0.57	DMU50	0.094	4.37
DMU19	0.243	4.48	DMU51	0.547	2.86
DMU20	0.701	23.89	DMU52	1.000	15.12
DMU21	0.723	7.23	DMU53	0.107	0.48
DMU22	0.460	2.78	DMU54	0.361	1.64
DMU23	0.283	18.72	DMU55	0.480	22.09
DMU24	0.337	13.80	DMU56	1.000	6.32
DMU25	0.409	3.66	DMU57	0.294	8.93
DMU26	1.000	18.19	DMU58	0.650	3.62
DMU27	0.438	0.10	DMU59	0.175	1.10
DMU28	1.000	2.60	DMU60	0.223	23.80
DMU29	0.196	1.19	DMU61	0.855	2.14
DMU30	0.450	4.52	DMU62	1.000	0.51
DMU31	0.348	12.81	DMU63	1.000	2.89
DMU32	0.555	1.59	Average	0.472	17.27

Table 5 also shows another shadow price result. The concept of the shadow price is widely used to measure the abatement cost caused by environmental regulations [29,32]. The shadow price not only presents an estimated value of the opportunity cost for undesirable output but also provides such output with meaningful guidelines to formulate regulatory policies for public decision-makers. In this study, the shadow price is equal to the carbon price for each unit of carbon emissions abatement.

The calculated average shadow price of all 63 companies is US \$17.27. This result is far from the real value of US \$10 at the end of 2014. In other words, the market price reflected just 58% of the shadow price. This situation indicates that the carbon price in the ETS market is too low to encourage companies to participate actively in the market. The small trading volume supports the hypothesis that the low carbon price prevents the ETS from being proactively pursued. As shown in Figure 1, the market price of carbon emissions remained at approximately US \$10 dollars for almost a year and then began to rise, increasing to US \$18 when each company had to register its carbon emissions performance at the end of March, 2016. This trend implies that companies do not regard carbon emissions regulation seriously; thus, they do not have a strong motivation to invest in

energy-saving technologies or less energy-intensive production processes. Indeed, the market price only increased, almost doubling from its initial price, when each company had to register its carbon emissions performance. An ETS member company could and should buy a carbon emissions permit at the doubled price, otherwise it should pay triple the price as a penalty for its unfulfilled carbon emissions target. Moreover, most companies did not participate in carbon trading because of the 100% free allowance; thus, the market price could not accurately reflect the willingness of companies to engage in the additional abatement of carbon emissions.

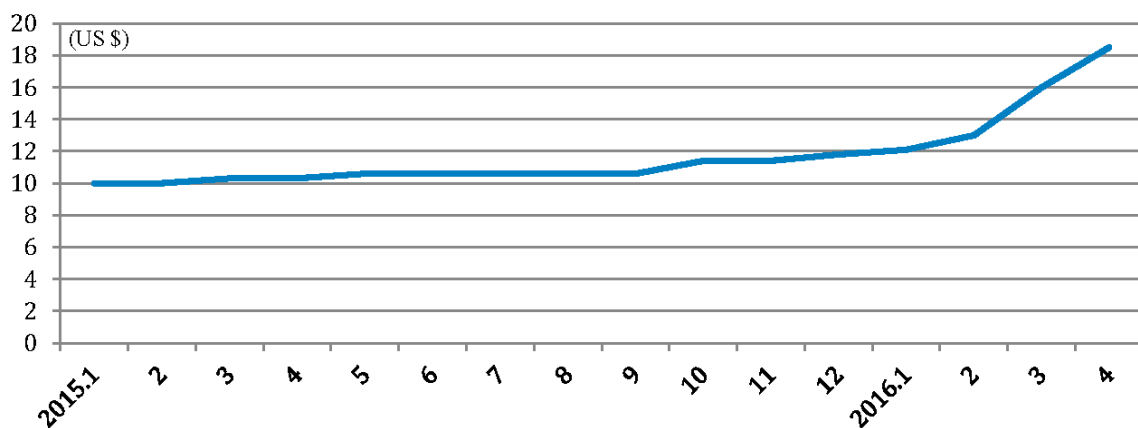


Figure 1. Market price trend of the carbon emissions allowance under the ETS.

3.3. Morishima Elasticity of Substitution

As shown by the CTE results, the Korean government should regulate the ETS more proactively because most companies show a very lower level of eco-friendly efficiency. Moreover, the market price for the carbon emissions allowance is much lower than its shadow price. This difference implies that even if a significant gap exists between the market price and the shadow price of carbon emissions, companies are not very willing to enhance their CTE and proactively participate in carbon trading. So that these companies take strategic decisions about investment in energy-saving green technology more seriously, capital-intensive enhancement of CTE performance should be more feasible. In order to evaluate the feasibility of green investment, we examine input/output-oriented Morishima elasticity. Table 6 presents the results of the Morishima elasticity of substitution that we employ to evaluate the sustainability of ETS policies for green investment promotion. The figures are calculated from Equation (8) and evaluated based on the shadow price ratio of inputs. As shown in Table 6, M_{kl} is 2252.9, which is much higher than an M_{lk} of zero. It should be noted that a higher Morishima value implies lower substitutability. The findings indicate that there is no difficulty in capital substituting for labor, although labor cannot substitute for capital. Such a result is to be expected in the capital-intensive petrochemical industry.

Table 6. Morishima-type elasticities of substitution (input).

Elasticity of Substitution	Estimate
M_{kl}	2252.9
M_{lk}	0
M_{ke}	33.70317
M_{ek}	0
M_{le}	0.34207
M_{el}	52.17643

In addition, M_{ke} is 33.7, implying that capital can easily be substituted by energy. Thus, it is strongly recommended that energy-intensive companies should expand their investment in facilities

and energy-saving technology. Finally, M_{el} is higher than M_{le} , which means that energy can be substituted by labor but not vice versa. Hence, the results show that in a capital-intensive industry, labor is easily substituted with energy; thus, green technology also leads to labor-saving operations.

4. Conclusions

One year has passed since the Korean government inaugurated its ETS in 2015; consequently, it may be too early to evaluate the sustainability of the government's ETS policies. Although there have been problems such as a low volume of carbon trading and unduly low semi-fixed prices, such policies must continue successfully. The reason is that ETS policies are necessary for optimal control of future challenges and for a low carbon society, both of which are goals that the Korean government should pursue because the achievement of low carbon emissions is an engine for future development. From the beginning, there have been excessive claims and opposition from diverse interest groups such as politicians, industrial leaders, business managers, and experts. Because of the complaints from these people, the government has set its carbon-reduction targets at much lower levels than originally planned, resulting in a lack of governance for sustainable performance. In this research, we examined three governance factors of sustainability with regard to Korean ETS policies.

First, the eco-friendly efficiency of CTE shows that companies have significant potential to enhance their efficiency. The average CTE for the 63 ETS member companies in our sample was 0.472. This figure suggests that there is potential for a 52.8% enhancement of such companies' CTE. In other words, the proactive participation of companies in the ETS should lead to an efficiency improvement that more than doubles the current average. If such companies can obtain information about the best practices of companies on the frontier, they can certainly improve their CTE. Moreover, in order to avoid bipolarization of carbon-reduction performance, the government should make cross-learning from best practices easier and more profitable for the ETS member companies.

Based on this empirical result, we examined the second governance factor of market price appropriateness. Using the shadow price of carbon emissions, the empirical result showed that the market price was approximately half that of the shadow price for almost a year. This result implies that an excessively regulated market price prevents companies from proactively participating in carbon trading. Further, it is clear that there is a missing link in the carbon trading policy of the ETS. In order to find this missing link, we used a third stage of empirical testing in terms of Morishima elasticity between the input and undesirable output of carbon emissions. The findings showed that there is significantly high substitutability between capital and energy input, implying that greater investment in green technology results in higher levels of energy-saving efficiency. This improvement is the ultimate goal of the Korean government's ETS policies. However, unfortunately, the current severely distorted restriction of unduly generous carbon emissions targets has resulted in poor performance from efficiency and shadow price perspectives. Thus, the empirical results strongly suggest that the Korean government should encourage the ETS member companies to make greater efforts to decrease their carbon emissions. Such encouragement should not rely on a voluntary approach but should be in the form of a stricter regulated system with much lower free allowances for carbon emissions. In particular, CTE is bipolarized, implying that there is significant potential for companies with low CTE to improve easily and effectively. Such improvements can originate from intra-industry technological transfers or the learning effects of best practice taken from those companies on the frontier of the DDF production possibility curve.

This research used duality theory with a non-radial, non-parametric DDF approach to examine diverse production characteristics from the sustainable development perspective. Using input-oriented DDF, CTE was measured in the first stage. Further, by using the dual model of non-radial DDFs, we obtained the shadow prices of carbon emissions and input substitutability. Because the methodology could not reflect the dynamic effect of the ETS, it may have limited implications. Moreover, the empirical test was undertaken in the petrochemical industry only. Thus, even if the research provides clearer insights because of the homogeneity of the DMUs, it still needs to expand

its empirical base to encompass diverse industries by using the meta-frontier approach. All the data in the paper is based on the one year of 2015, and it may be too short to infer the significant implications and suggestions. However, since the Korean government initiated the target management system (TMS) five years before ETS, all the covered companies are ready to participate in the ETS and their efforts are relevant to be analyzed even in the short term. Of course, our research need to be complemented with the dynamic changes of the policy effects over time for the future research issue.

Are emissions trading policies sustainable in Korea? This paper showed clearly the lack of governance for a sustainable ETS policy in Korea; thus, the Korean government should engage in greater proactive efforts to upgrade the future sustainable performance of the ETS [37]. Especially, for the sustainable performance of ETS, the covered companies should make efforts to invest in green technology to improve its environmental efficiency, and the paper showed that there is huge potential for benefits from green investment to enhance environmental efficiency. However, the government should support for the companies to invest on the green technology as early as possible to fill the missing link between the target and the current huge lack of the environmental efficiency.

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