

Article

# The Sustainability of Energy Substitution in the Chinese Electric Power Sector

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**Abstract:** The Chinese electric power industry, including its coal industry and other energy industries that are not efficient, contributes to China's serious energy shortages and environmental contamination. The governing authority considers energy conservation to be one of the most prominent national targets, and has formulated various plans for decarbonizing the power system. Applying the trans-log cost function, this paper examined the trans-log cost function to analyze the potential inter-factor substitution among energy, capital and labor. We also investigated what role human capital played in energy substitution for the electric power sector during the period from 1981 to 2017. Three key results were derived: (1) energy is price-insensitive, (2) there exists large substitution sustainability between both capital and labor with energy, and (3) human capital input not only enhances the extent of energy substitutability with capital and labor but also is a substitute to energy itself. These findings imply that the liberalization of the electric price mechanism is conducive to lessening energy use and augmenting non-energy intensiveness, and that energy conservation technology could become more sustainable by investing more capital in the electricity sector, thereby achieving a capital–energy substitution and a decrease of CO<sub>2</sub> emissions. We further suggest that the priority for the Chinese electric power industry should be to attach more importance to increasing human capital input.

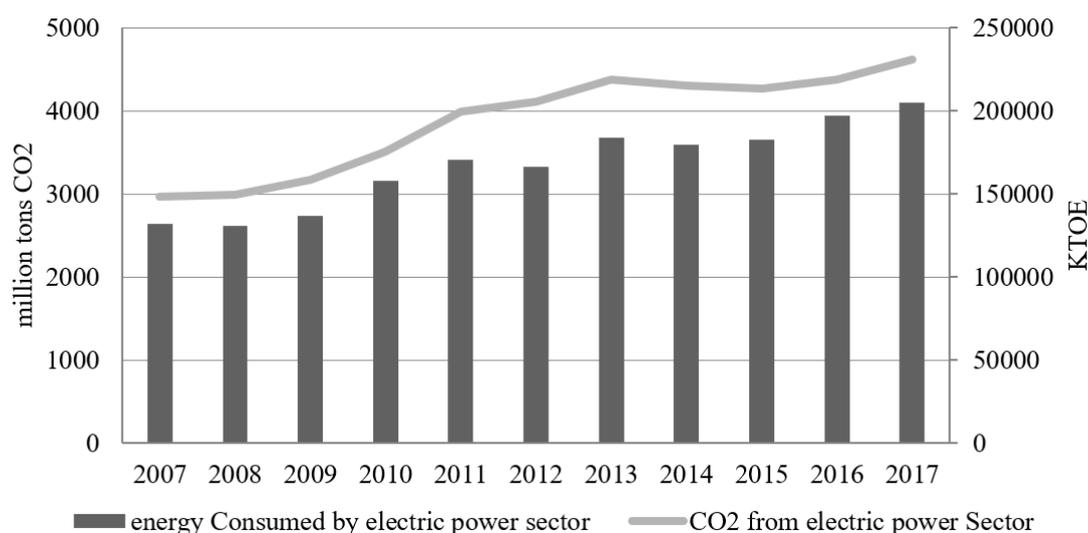
**Keywords:** energy substitution; inter-factor substitution; trans-log cost function; electric power industry; sustainability

**JEL Classification:** H32; L69; Q41

## 1. Introduction

China, the world's current largest energy consumer, is now confronting a big challenge in seeking equilibrium between sustainable development and economic growth. By the end of 2017, China consumed 3077 billion tons of oil equivalent (TOE), ranking first in the world, and this will likely continue soaring during the process of industrialization and urbanization, which will further intensify the imbalance between energy supply and demand. China has seven cities listed among the ten most contaminated international cities, implying that the pollution levels in China are higher than the World Health Organization (WHO) standards [1]. No other country depends on coal to the degree that China does, and no sector in China depends on coal to the degree that the electricity sector does. The electric power industry has a key role in economic development and people's welfare,

directly or indirectly linking with all significant sectors in China. However, electricity generation leads to several issues, which are mainly correlated with the emission of greenhouse gases (GHGs) and energy consumption, resulting in global environmental deterioration. The electric power industry, with a large proportion of inefficient generating units, is predominately coal-based, and contributes to most of China's serious energy shortages and environmental contamination. It is also the largest global source of GHG emissions (International Energy Agency, IEA, 2012). Figure 1 illustrates that the enhanced energy consumption of the Chinese electricity sector moved along with the consequential CO<sub>2</sub> emissions in the past ten years.



**Figure 1.** Energy consumption and CO<sub>2</sub> emission in the electric power sector of China. Source: the International Energy Agency.

China has a considerably imbalanced energy structure, with an abundance of coal and hydropower, but deficiencies in both oil and gas. Owing to reserve limitations, thermal power occupies a dominant position (67.3%) in the Chinese power generation mix, while the proportions of hydro, nuclear and wind power are relatively low. In general, thermal power generation uses coal with roughly 28% ash content, which is much higher than other power generation modes, thus making itself a dominant producer of pollutant emissions such as SO<sub>2</sub>, NO<sub>x</sub> and flue dust, which contaminate the ecosystem and atmosphere. Furthermore, though the energy efficiency of the electric power industry has increased substantially during the last decade, there is still a hard gap to surmount between China and other developed economic entities. The energy intensity of electricity generation is higher than the average level of other developed countries. As a result, the electricity sector is the largest consumer of coal resources and the key contributor to severe environmental problems. To specify, the electric power industry accounted for almost 50% of Chinese CO<sub>2</sub> emissions in 2017, according to the statistics in the IEA database.

The elasticities of energy substitution are involved in many important issues, including investment subsidies, carbon and gas emissions, and capital, carbon and fuel taxes [2]; thus, it is a major target and tool for environmental policymakers and planners to grasp the ability to convert production factors. In such an international environment, energy substitution research is vital for both the theoretical enrichment of energy economics and practical applications of national environmental and energy policies. The electric power industry is particularly crucial for the field because it accounts for almost half of the national greenhouse gas emissions, and an even larger proportion of the whole country's consumption of primary energy resources.

China has enacted a host of policies and laws targeted at energy conservation and environmental protection related to the electric power industry in order to attain sustainable development; for example, the Revised Energy Conservation Law, and China's plans for both energy and nuclear power

developments. Safarzadeh et al. [3] suggested that China, the USA and Sweden are the countries addressed the most on industrial energy efficiency programs and their applications. This further underlines the importance of investigating whether there is any possible energy substitution in this sector. However, as far as we know, there has been no study in the literature on China's inter-factor substitution sustainability with respect to the electric power industry until now. This paper helps fill the essential gap in this aspect. Employing the trans-log cost function, this research examines the substitution between non-energy and energy factors, and particularly explores the peculiar role of human capital input in moving the Chinese electric power industry towards a low-carbon system over the 1981–2017 period.

The remainder of the article is organized as follows. We give a brief review of the major empirical literature of energy and non-energy substitution in Section 2. Section 3 introduces the data sources, followed in Section 4 by a discussion of the methodologies employed in this study. Section 5 discusses the estimated results, provides the scenario analysis and investigates the unique impact of human capital input. Section 6 gives the conclusions and policy implications.

## 2. Literature Review

Since Berndt and Wood [4] wrote about the study of energy substitution, the sign and magnitude of energy elasticities of substitution have been broadly debated by researchers associated with numerous regions and industries, with divergent results [5]. The empirical literature in academia has so far discovered proof of both complementarity and substitutability for energy and other traditional inputs. Christopoulos and Tsionas [6] and Arnberg and Bjorner [7] insisted that energy and non-energy are strong complements, while Welsch and Ochsens [8], Koetse et al. [9] and Lin and Xie [10] found assertive evidence of energy and non-energy being substitutable. However, Halvorsen and Ford [11] and Field and Grebenstein [12] discovered 'mixed results' on energy and non-energy substitutability.

The relationships between non-energy and energy inputs vary across studies because academics applied different approaches, use different periods in their studies and use different datasets [12]. Some scholars attempted to explore the internal causes, including Apostolakis [2], who reviewed the early literature and noticed that most studies employing time-series data sorted energy and non-energy as complements, while most studies employing pooled data classified them as substitutes. He concluded that the principal explanation for this distinction was that the time-series mainly showed short-term effects, while pooled cross-section data mainly expressed long-term influence. This notwithstanding, Fondel and Schmidt [13] held the view that cross-price-elasticity depended on the factors' cost shares. Kander and Schon [14] estimated the energy—capital substitutability in Sweden over the period 1970–2000 and obtained a similar result to the former study.

The literature on inter-factor substitution in developing countries is very limited (Presley et al. [15]), especially in China [12,16]. The restricted amount of work on this topic is due to the fact that few scholars have paid attention to the feasibility of substitution between energy and other classical factors, and the few existing papers appear quite unrealistic as there are many econometric mistakes, including model misspecification and a short period of data. As Ma et al. [17] stated, "new and more representative datasets and more appropriate robust econometric techniques are needed to explore the estimation of the elasticities of the substitution of energy—capital and energy—labor and the demand for energy in China." Further study on the interplay of energy and other traditional inputs is urgent. In recent years, scholars have filled this gap in the literature by analyzing inter-factor or inter-fuel substitution on an economy-wide level. For instance, Lu and Zhou [18] gauged the own-price elasticities and Allen elasticities of inter-factor substitution in the manufacturing sector, and found capital—labor and labor—energy to be strong substitutes, and capital—energy to be a strong complement. Guo et al. [16] calculated the own-price elasticities and Morishima elasticities and found strong evidence of energy—capital and energy—labor substitutability in the Chinese industrial sector. Su et al. [19] utilized the production function to examine the elasticities of substitution between capital stock, labor and energy for China. He and Lin [20] revealed both heterogeneity and asymmetric

effects in the energy–capital substitutability of the manufacturing sectors in China, while Wei et al. [21] examined inter-factor substitution in China’s high-tech industries, and found that the effect of factor substitution has gradually increased in recent years.

Whereas studies of China’s energy substitution are limited, academics have ignored the electricity sector, and the non-energy inputs have always been limited to capital and labor. To date, there has been no study available to researchers or policymakers and planners that has estimated inter-factor substitution whilst considering human capital for China’s electricity generation industry. This paper intends to fill this gap.

### 3. Data and Methodology

We conducted this study by employing the following inputs: capital stock ( $K$ ), energy use ( $E$ ), labor use ( $L$ ) and human capital stock ( $H$ ). The total cost series ( $C$ ) consisted of capital stock, aggregate energy use, labor use and human capital stock. The four-factor share series were estimated using four-factor inputs and the total cost series.

We followed the literature to construct the four-factor price indices as follows: the aggregate energy price index ( $P_E$ ) is counted as the fuel and power price index, which is provided by the China Statistical Yearbook (CSY). The capital stock price index ( $P_K$ ) given by [14] satisfies

$$P_K(t) = [r(t) + \delta - (dq_k(t)/dt)/q_k(t)] (1 - \tau) q_k(t) \quad (1)$$

where  $P_K(t)$  is the capital price,  $\delta$  is the depreciation rate,  $r(t)$  is the nominal interest rate,  $\tau$  is the marginal corporate income tax rate,  $(dq_k(t)/dt)/q_k(t)$  is the capital real price gradient, and  $q_k(t)$  is the capital real price. In particular, when  $\tau = 0$ ,  $q_k(t) = 1$ , and  $(dq_k(t)/dt)/q_k(t) = \pi(t)$ ,  $\pi(t)$  represents the actual inflation rate. We used the labor price index ( $P_L$ ) to represent the labor wage rate, which is the ratio of the total wage payment over total employment in the Chinese electric power industry. We used the GDP deflator index to represent the index ( $P_H$ ) of the human capital stock price. We used the fixed assets price index and the weighted index of the consumer price index to estimate the GDP deflator index to deflate GDP, because the GDP deflator is not available in the CSY, and capital costs and labor are the main components that form GDP in China. The base year for all four-factor price indices was 1981. All of the above indicators were obtained for the Chinese electric power industry, and for each year from 1981 to 2017, offering a time database with 36 observations.

The four principal data sources in this study are the State Development Planning Commission of China (SDPC), the China Energy Yearbook (CEY), the CSY, and the China Industry Economy Yearbook (CIEY). The data of capital investment, employment, wages, the consumer price index, and a fixed assets price index are obtained from CSY. However, we cannot get any statistics for both capital stock and human capital stock from CSY. Thus, we used the perpetual inventory method to estimate the capital stock/human capital stock series, as shown in the following equation:

$$K_t = K_{t-1}(1 - \delta_t) + I_t \quad (2)$$

which estimates the capital stock/human capital stock  $K_t$  at time  $t$ , where  $\delta$  stands for the capital/human capital depreciation rate and  $I_t$  is the current capital/human capital investment. Our estimates of both capital stock and human capital stock are consistent with Jiao and Jiao [22]. The CEY offers detailed data about the consumption of primary energy in the electric power industry by year, and the conversion coefficient between each energy source and standard coal. The SDPC offers fuel price data from 150 city price bureaus, but we only selected the coal price and converted it to the standard coal price. The CIEY provides detailed data on the major economic indexes of China’s electric power industry. The cost shares and price index for each factor during the study period are shown in Figures 2 and 3. Figure 2 shows the time series plots of the cost shares of factors for the electric power industry in China. The change in the share of capital factor is strongly opposed to the change in the share of both labor factor and energy factor. This implies that energy–capital and energy–labor are substitutable. On the other hand, Figure 3

shows the changes in the indices of factor prices for the electric power industry in China. From the figure, it can be seen that the prices of China’s capital factor were low, and increased significantly in recent years. The prices of other factors have risen steadily also. The increases in both labor and energy factor prices have been the highest in recent years. This reflects to a certain extent that China’s factor marketization is relatively low, and that factor prices, especially capital prices, are subject to certain distortions.

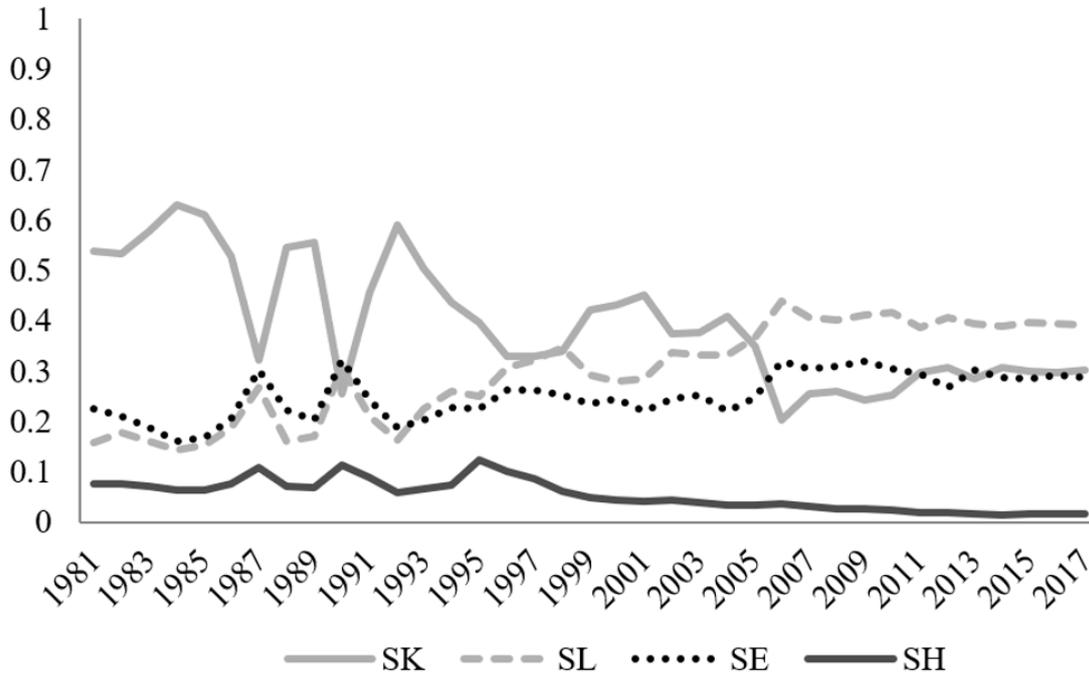


Figure 2. Cost shares of factors for the electric power industry in China.

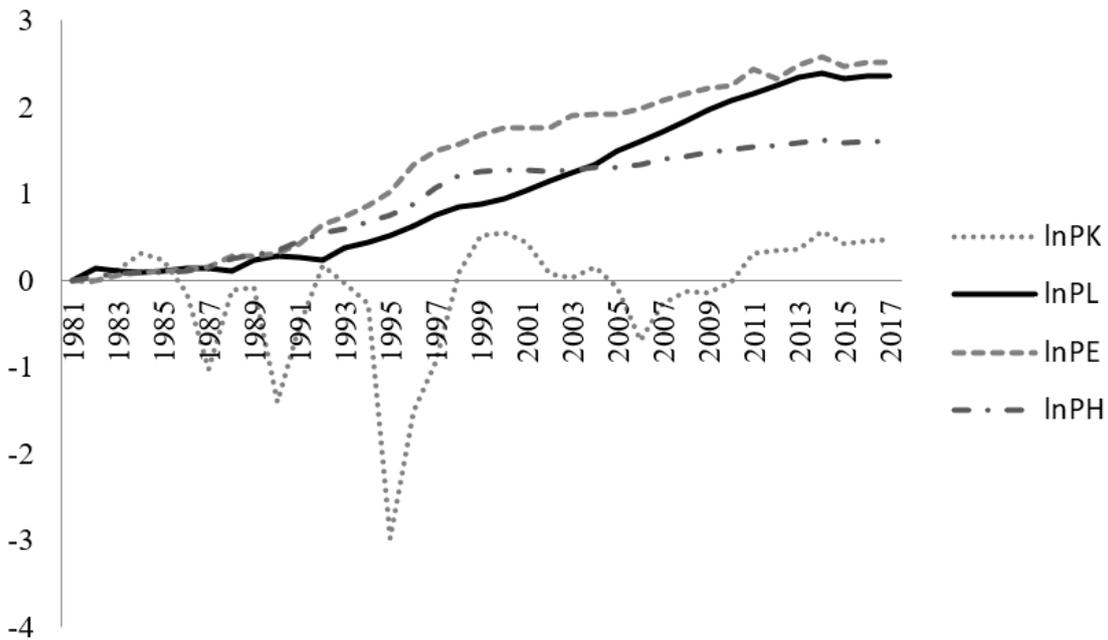


Figure 3. Indices of factor prices for the electric power industry in China.

#### 4. Methodologies

There is generally a dilemma when we estimate a production function, because input factors are probably endogenous and thus violate the basic necessary condition that the ordinary least squares must be unbiased. This problem can be circumvented by using factor prices in the cost function, and thus it is common in the energy economics literature to use a trans-log cost function to quantify energy demand elasticities [23]. Furthermore, as a second-order approximation, the trans-log cost function allows us to avoid specifying a particular production function and supposing constant or equal elasticities of substitution [24].

The three-factor production function is

$$Y = f(K, L, E) \quad (3)$$

where  $Y$  is the output, and  $K, L$  and  $E$  denote the capital input, labor input and energy input, respectively. If both the factor price and output level are determined by an exogenous element, then Equation (3) could be described by the following cost function:

$$C = f(Y, P_K, P_L, P_E) \quad (4)$$

where  $C$  is the total cost and  $P_K, P_L$  and  $P_E$  indicate the factor price indices of capital, labor and energy, respectively. If we apply the trans-log cost function model, Equation (4) is depicted as follows:

$$\ln C_t = \alpha_0 + \ln Y_t + \sum \alpha_i \ln P_{it} + 0.5 \sum \sum \beta_{ij} \ln P_{it} \ln P_{jt} + \gamma_{it} + \varepsilon_t \quad i, j = K, L, E \quad (5)$$

where both  $i$  and  $j$  represent indices for the inputs, while  $P_{it}$  and  $\varepsilon_t$  are the price of input factor  $i$  and the unobserved error term at time  $t$ , and the parameter  $\alpha_0$  is the intercept. By applying Shephard's lemma, we obtained the firm's system of cost-minimizing demand functions, which are the conditional factor demands obtained by differentiating Equation (5) with respect to the following input price:

$$X_i = \partial C / \partial P_i, \quad i = K, L, E \quad (6)$$

where  $X_i$  represents input  $i$ . We assume constant returns to scale and no neutral technological change. Then, the following three-factor cost-share equation can be derived:

$$S_{it} = X_{it} P_{it} / C = \frac{\partial \ln C}{\partial \ln P_{it}} = \alpha_i + \sum_{j=1}^m \beta_{ij} \ln P_{jt} + \gamma_{it} + \varepsilon_{it} \quad i, j = K, L, E \quad (7)$$

where  $S_{it}$  indicates the cost-minimizing cost share of factor  $i$  at time  $t$ . In order to explore the impact of human capital input, we incorporated the human capital input, giving us the following four-factor cost-share equation:

$$S_{it} = \alpha_i + \sum_{j=1}^m \beta_{ij} \ln P_{jt} + \gamma_{it} + \varepsilon_{it} \quad i, j = K, L, E, H \quad (8)$$

where  $H$  represents the human capital input. Equation (3) can be applied to approximate any unknown cost and any production function when we choose the proper special set of parameters. We set the symmetry restrictions to follow

$$\beta_{ij} = \beta_{ji} \text{ for all } i \neq j \quad (9)$$

which means that the cross-derivatives are the same. The following regularity conditions are required to satisfy the linear homogeneity at price:

$$\sum_{i=1}^m \alpha_i = 1, \sum_{i=1}^m \gamma_{it} = 0, \sum_{j=1}^m \beta_{ij} = 0, i, j = 1, \dots, m \quad (10)$$

The Allen partial elasticities of substitution at time  $AES_{ij}$  can be obtained by

$$AES_{ij} = \frac{\partial \ln X_i / X_j}{\partial \ln P_j / P_i} = \frac{CC_{ij}}{C_i C_j} = (\beta_{ij} + S_i S_j) / S_i S_j \quad i \neq j \quad (11)$$

$$AES_{ii} = \frac{\partial \ln X_i / X_i}{\partial \ln P_i / P_i} = (\beta_{ii} + S_i^2 - S_i) / S_i^2 \quad (12)$$

while the own-price elasticities  $OPE_{ii}$  and the cross-price elasticities  $CPE_{ij}$  of factor demand can be obtained by Equations (13) and (14), respectively, by using estimated parameters from Equation (5) [23]:

$$CPE_{ij} = \frac{\partial \ln X_i}{\partial \ln P_j} = \frac{\beta_{ij} + S_i S_j}{S_i} = AES_{ij} S_j \quad i \neq j \quad (13)$$

$$OPE_{ii} = \frac{\partial \ln X_i}{\partial \ln P_i} = \frac{\beta_{ii} + S_i^2 - S_i}{S_i} = AES_{ii} S_i \quad (14)$$

Starting from the two-factor inputs Hicks-type elasticity of substitution [25], related studies continued to develop three-factor inputs models. For example, cross-price elasticity (CPE), Allen elasticity substitution (AES), and Morishima elasticity substitution (MES) [26,27] are among the most popular methods being used for substitution elasticity. After AES was proposed, it has been widely used in empirical studies of multi-factor substitution [28]. Since it is difficult to use the AES to measure the relative price changes in a multi-factor production system, Morishima elasticity appears to be a more useful tool for the measurement of energy to non-energy substitution [29]. There are numerous empirical studies, for example, Sickles and Streitwieser [30] and Nguyen and Reznak [31], applying the MES model to estimate the inter-factor elasticities of substitution. In this paper, there are more than two input factors in the production systems. Thus, we employ the MES model to modify AES and apply two different measuring methods; that is, the Morishima elasticity method and the cross-price elasticity method. Therefore, the CPE measures the percentage change of the non-energy demand induced by a one percent change of energy price, while the MES requires the price of factor  $j$  to be flexible and holds other prices constants, and measures the percentage change of the non-energy ratio if the energy price increases by 1%. This approach is more appropriate because it can be used to measure the income effect of factor inputs, which cannot be estimated if one applied the Allen model. It is defined as [10]

$$MES_{ij} = \frac{\partial \ln(X_i / X_j)}{\partial \ln P_j} = \frac{\partial \ln X_i - \partial \ln X_j}{\partial \ln P_j} = CPE_{ij} - OPE_{jj} \quad i \neq j \quad (15)$$

Changes in demand for other factors are caused by changes in factor prices mainly based on the substitution effect and income effect. The difference between Morishima elasticities and CPE is that Morishima elasticities consider both substitution and income effects, but CPE only considers the substitution effect. The difference between CPE and Morishima elasticities is similar to the distinction between economic measures and engineering measures of substitution [32]. Different research objectives will prefer to adopt different measures. For instance, energy policy planning that requires information on actual changes will prefer to use cross-price elasticities, while academics studying the substitution potential in industries will be inclined to use Morishima elasticities.

## 5. Results and Discussion

Using the models presented in Section 3, we exhibit their estimates in this section. We first exhibit and discuss the estimated parameters of the models, and then we display and interpret both the derived elasticities and the scenario analysis.

### 5.1. Estimation Results

We exhibit the estimated coefficients of the share equations and the trans-log cost function in Table 1 for the electricity generation industry. We first remove the labor share equation from two factor share equations including both aggregate energy and capital share due to the adding-up restriction, and estimate the one total factor cost equation and two factor share equations. The  $R^2$  for the aggregate energy share equation, the total factor cost equation and the capital share equation are 0.92, 0.99 and 0.81, respectively. All estimates for the parameters are statistically significant and have the right signs, inferring that the estimated total factor cost function is appropriate and behaves properly because the input demand function is concave in price and strictly positive (Berndt and Wood [4]).

**Table 1.** Estimation results of the cost share equation.

Coefficient	Regression Estimate	Standard Error
$\alpha_K$	0.6687 **	0.0140
$\alpha_L$	0.1122 <sup>a</sup>	-
$\alpha_E$	0.2191 **	0.0160
$\beta_{KK}$	0.1268 **	0.0270
$\beta_{LL}$	0.0036 <sup>a</sup>	-
$\beta_{EE}$	0.0746 **	0.0170
$\beta_{KL}$	-0.0279 <sup>a</sup>	-
$\beta_{KE}$	-0.0989 **	0.0177
$\beta_{LE}$	0.0243 <sup>a</sup>	-

Note: \*\* indicates that the coefficient is significant at a 5% level; <sup>a</sup> due to adding up, standard errors are not provided.

### 5.2. Derived Demand Elasticities

We then apply Equations (13)–(15) based on the estimated parameters presented in Table 1 to compute the own- and cross-price elasticities and Morishima elasticities of factor demand. The own- and cross-price elasticities are plotted in Figures 4 and 5. We observe several notable features, as follows:

- (i) Consistent with all the existing economic theories, all own-price elasticities have negative signs. Each factor is responsive to the change in its price. The median own-price elasticities for energy, capital and labor are 0.4089, 0.221 and 0.8216, respectively. The demands in both labor and energy are more sensitive to the change in their price than that of capital. The magnitude and direction of the elasticity of own-price have been used in some empirical studies, including Costantini et al. [33] and Sharimakin [3].
- (ii) When considering the cross-price elasticities of capital demand, we find positive elasticities with respect to the price of energy. The estimated results of the cross-price elasticities imply that both capital and energy have substantial substitution sustainability, demonstrating a slightly upward trend. This indicates more potential for alleviating energy supply shortages with higher capital investment in China's electric industry, thus discovering an effective way of saving energy.
- (iii) Likewise, we find positive elasticities with respect to the price of energy regarding the cross-price elasticities of labor demand, inferring that both energy and labor appear to be significantly substitutable. Additionally, both labor and capital are substitutes with values for the cross-price elasticities of substitution at 0.0873. Substitution between energy and labor necessarily arises from technical innovation, in the context that technological development brings the mechanization and automation of the electric industry and enables many things that were originally done manually to be accomplished with more energy consumption, setting some surplus labor free.
- (iv) As can be noted, labor demand is more sensitive to the energy price change than capital, for a median CPE(KE) of 0.1337 and a median CPE(LE) of 0.5336. Compared with overseas countries such as the US, there is less potential for energy substitution when the price of energy rises [34].

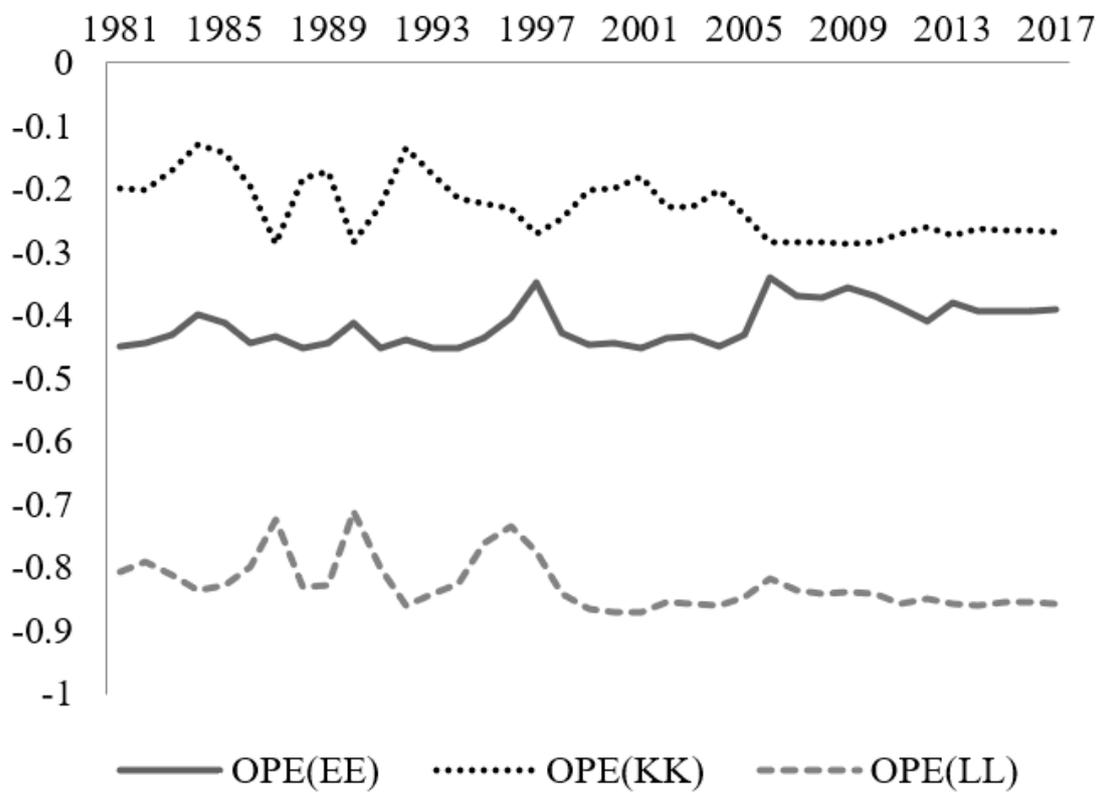


Figure 4. Own-price elasticities.

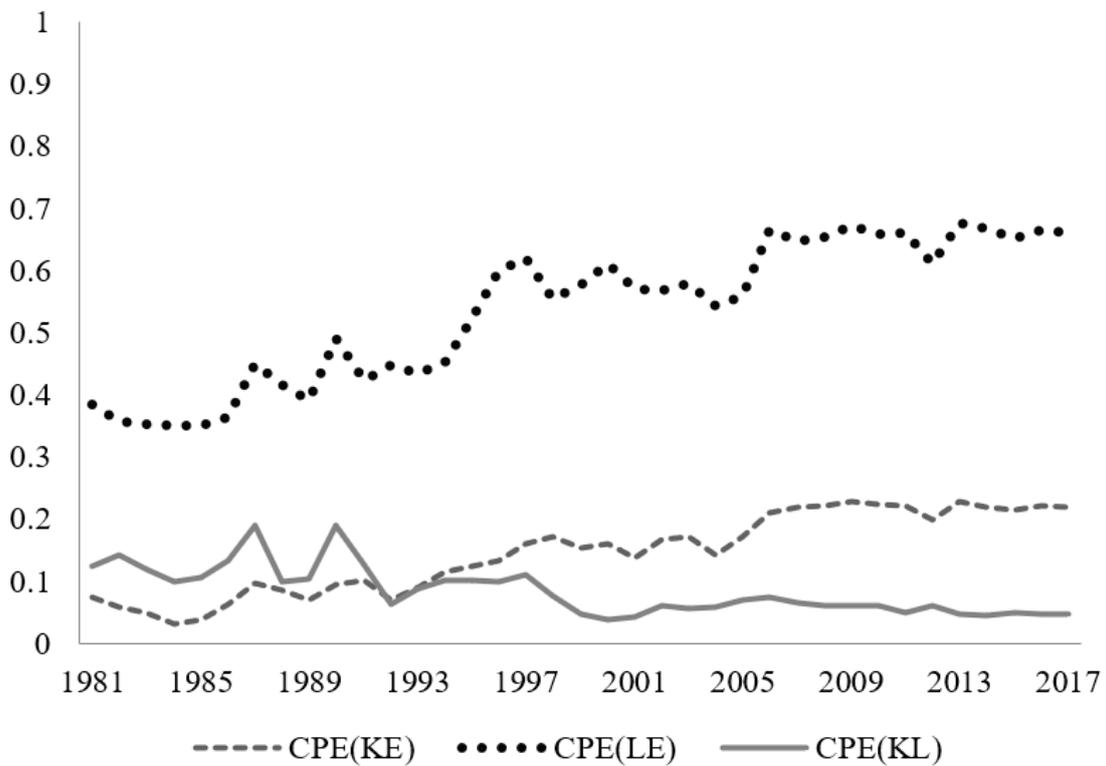


Figure 5. Cross-price elasticities.

The implied Morishima elasticities are depicted in Figure 6. The result provides several important conclusions, as shown in the following:

- (i) The estimated results imply that both capital and energy are substitutable, and similarly, labor is also a substitute for energy, indicating that both capital and labor demand are elastic to changes in energy price.
- (ii) The Morishima elasticities outnumber the corresponding cross-price elasticities in general, especially for capital and labor. The result is consistent with the findings of Koetse et al. [9], who found a distinction between Morishima elasticities and cross-price elasticities.
- (iii) Labor demand is more sensitive to energy-price change than capital, because  $MES(KE) = 0.5408$  and  $MES(LE) = 0.9426$ , inferring that holding other variables constant, energy prices increase 1%, and the demand ratio for both capital–energy and labor–energy will rise 0.5408% and 0.9426%, respectively.
- (iv) The average Morishima elasticities of the factor inputs are all less than one. This is similar to corresponding results in other Chinese industrial sectors [20], indicating that the substitutability of factors is limited in China [21].

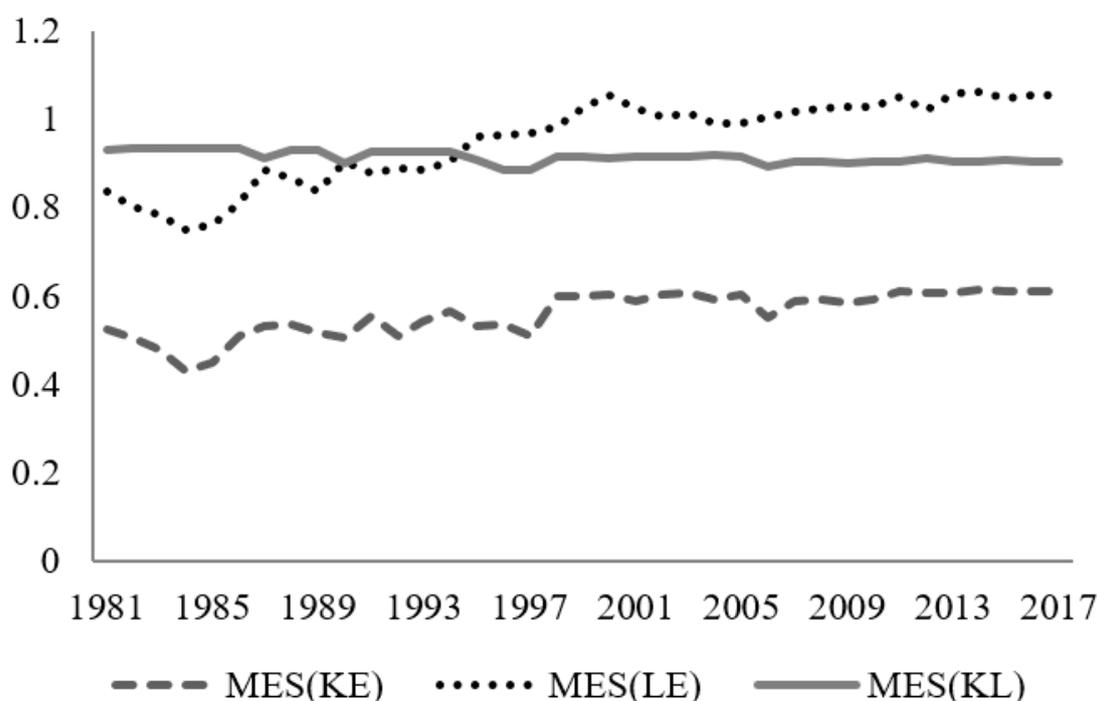


Figure 6. Morishima elasticities.

### 5.3. Scenario Analysis

In order to further explore the CO<sub>2</sub> emission reduction shock and the resultant potential energy conservation of the capital–energy substitution in China’s electricity industry, we computed both CO<sub>2</sub> emission reduction potentials and energy saving under different scenarios (displayed in Table 2). The results in Table 2 were obtained by adopting the Morishima elasticities of capital–energy substitution, energy use and CO<sub>2</sub> emission per unit of energy used in the electricity industry of China in 2014. The scenarios are the enhancement of capital input by both 5% and 10%. This underscores the potential of the Chinese electric power sector to realize emission reductions along with rising capital investment in energy conservation technology. To specify, when other factors remain constant, a 10% increase in the capital input results in a 17.96 Mtoe (million tons of oil equivalents) increase in the energy consumption savings in the electric power industry, bringing about a CO<sub>2</sub> emission reduction of 44.54 million tons.

**Table 2.** Energy saving and CO<sub>2</sub> emission reduction of China's electricity industry under different scenarios in 2014.

Year	Scenario 1: Capital Increased by 5%		Scenario 2: Capital Increased by 10%	
	Energy Savings (Mtoe)	CO <sub>2</sub> Emission Reduction (Million Metric Tons)	Energy Savings (Mtoe)	CO <sub>2</sub> Emission Reduction (Million Metric Tons)
2014	8.98	22.27	17.96	44.54

#### 5.4. The Role of Human Capital in Inter-Factor Substitution

Incorporating the human capital inputs in the trans-log model using Equation (8), we observe obvious changes in the Morishima elasticities. The estimated equations at this stage include one total factor cost equation and three factor share equations (aggregate energy, capital and labor shares; the human capital share equation is taken away due to adding-up the restriction). Compared with the corresponding original Morishima elasticities obtained in the three-factor model in the same term, we focus on the changes in the Morishima elasticities of both capital and labor, and examine the Morishima elasticities of human capital. The results of these calculations are displayed in Table 3. From the table, we observe the following:

- (i) Both capital and energy are substitutable, and similarly, labor is substantially substitutable with energy.
- (ii) In general, the Morishima elasticities of labor outnumber the corresponding Morishima elasticities of capital.
- (iii) Compared with the Morishima elasticities between the two cost-share equations, both labor and capital demand are more sensitive to energy-price change in the four-factor model, because both MES(KE) \* and MES(LE) \* increase by 23.96 and 12.31 points, respectively. This shows that human capital does contribute to both China's energy conservation and emissions reduction, and that it is useful to raise the factor elasticity of substitution, and optimize resource allocation and utilization.
- (iv) Human capital is a significant substitute to energy itself, with a median Morishima elasticity of 1.0522. This shows that human capital is more substitutable to energy than capital.

**Table 3.** Morishima elasticities.

Year	MES(KE) *	Rate of Change (%)	MES(LE) *	Rate of Change (%)	MES(HE) *
1981	0.7235	37.9935	1.1828	41.7715	1.067
1982	0.7114	41.0668	1.1352	41.4403	1.053
1983	0.6984	45.5910	1.1459	46.1794	1.0303
1984	0.6725	56.2863	1.1531	53.9931	1
1985	0.6807	51.6711	1.1381	49.3962	1.0087
1986	0.7057	39.1364	1.1109	37.2837	1.048
1987	0.6955	31.0286	1.084	22.7216	1.1354
1988	0.723	34.7623	1.1789	35.8649	1.0641
1989	0.7095	37.9009	1.1409	36.2920	1.0445
1990	0.6573	29.7730	1.0635	17.7480	1.1529
1991	0.7172	29.3417	1.1086	26.6248	1.0838
1992	0.6988	37.2888	1.1419	28.7228	1.0242
1993	0.699	29.0382	1.0614	19.8374	1.0439
1994	0.7022	23.9541	1.049	16.0398	1.0713
1995	0.5617	127.2249	1.0245	6.8300	1.2144
1996	0.4211	25.4768	1.0321	6.9312	1.1995
1997	0.558	9.6698	1.0105	4.3043	1.1545
1998	0.6798	13.3189	1.0102	2.7984	1.0897
1999	0.7021	17.2120	1.0295	0.6649	1.0606
2000	0.7112	17.7094	1.0445	0.8919	1.0583
2001	0.7004	18.9740	1.0233	0.0685	1.0273
2002	0.6917	14.6337	1.0115	0.8073	1.059
2003	0.6967	15.0050	1.0176	0.4144	1.0533

Table 3. Cont.

Year	MES(KE) *	Rate of Change (%)	MES(LE) *	Rate of Change (%)	MES(HE) *
2004	0.688	16.2948	0.9957	0.4844	1.0068
2005	0.6824	13.4120	0.9988	0.8889	1.0332
2006	0.5997	8.8385	1.0089	0.2584	1.1162
2007	0.6513	10.5210	1.013	0.5107	1.0756
2008	0.6576	10.9312	1.0161	0.9359	1.0597
2009	0.6445	10.5299	1.0169	1.1471	1.0564
2010	0.6495	9.7314	1.0089	1.7241	1.0188
2011	0.6785	11.2660	1.0155	3.3041	0.9534
2012	0.6707	10.4215	0.9947	2.3943	0.9235
2013	0.6742	10.9064	1.0164	3.7409	0.9173
2014	0.6809	10.9138	1.0113	4.6573	0.8704
2015	0.6753	10.7472	1.0075	3.5975	0.9037
2016	0.6768	10.8558	1.0117	3.9986	0.8971
2017	0.6777	10.8389	1.0102	4.0845	0.8904
Mean	0.6709	25.4126	1.0547	14.3068	1.0396

Note: MES(KE) \*, MES(LE) \* and MES(HE) \* represent the Morishima elasticities of capital, labor and human capital in the four-factor trans-log model, respectively.

## 6. Conclusions and Implications

As far as we know, there has been no study in the existing literature on Chinese inter-factor substitution sustainability, especially in the electric power industry. This research fills in the essential gap in the literature by investigating the inter-factor elasticity of substitution between capital, labor, human capital, energy and the role of human capital in the substitution of the Chinese electricity generation industry by employing the trans-log cost function.

The empirical analyses document made several findings. First, we observe that the own-price elasticities of energy are statistically significant, and all are inelastic. There are many factors contributing to the inelasticity of energy demand, and the considerably low level of energy prices may be one of the prominent reasons. The current energy price level is not capable of producing incentives to spur industries to change their consumers' behavior. The government should make greater efforts to liberalize energy price in order to promote energy conservation and emissions reduction. Second, substitutability was detected between capital and energy, and between human capital and energy. Capital–energy substitutability could be mostly ascribed to applications of energy conservation technologies via capital purchases, which were subsequently supported using the scenario analysis, offering an important channel of GHGs mitigation. They indicate that the Chinese electric power industry can switch from carbon-emitting energy to capital, thus achieving the goal of gradually boosting its industrial economy, as well as relieving the negative impact on the environment. Third, by highlighting that labor and the energy inputs are substitutes, we conclude that the development of technology will likely reduce labor intensiveness. This shows that the electric industry has incentives for releasing the surplus labor force by obtaining more energy consumption in the context of advancing electricity generation technology. Lastly, our results provide evidence to support our conjecture that human capital indeed plays a crucial role in the course of inter-factor substitution. It contributes to shortening the industrial transition time and optimizing the allocation of the inputs. In other words, human capital should not only increase the substitutability between capital–energy and labor–energy, but also reduce the Chinese electric power industry's reliance on energy, since there is significant substitution sustainability between human capital and energy (1.0522). This would alleviate the situation under which China is suffering severe energy constraints and environmental pollution. Therefore, a major increase in the share of human capital as a portion of the production inputs will contribute to the transformation towards a low-carbon and energy-saving electric power system.

A principal issue for energy policy planning is the possibility of non-energy factors to substitute for energy, and the influences of substitution on future economic development. The results of this research offer essential policy implications for China. First, there are many factors which contribute

to the inelasticity of demand for energy, where the level of energy prices could be one of the most essential reasons. Competition is the foundation in any market, and the electricity sector is no exception. Since there is a lot of imperfection in the Chinese market, particularly in the electricity industry, it is a challenge for the Chinese government to boost electricity price reform and deregulate rules to release more potential energy conservation. The current electric price mechanism is unable to truly reveal the resource scarcity and the environmental costs, lagging far behind the market demand [28]. The resource tax on fossil energy is too low to fully reflect their un-renewability, and energy subsidies are the main cause of China's high energy intensity. Hence, they should also be gradually reduced until they are abolished. Furthermore, subject to regulated coal market prices, the electricity price mechanism cannot steer and accelerate energy conservation and emissions reduction. Efforts should include raising fossil fuel prices by using economic methods like energy or environmental taxes, removing subsidies for non-renewable energy, and enhancing the linkage between coal price and electricity price. Second, since all inputs are estimated to be substitutes, there appears to be a possibility for the government to promote an increase in the use of alternative factors. The substitutions between capital and labor, and between labor and energy show that the electric industry has the potential to realize a transition from being a labor-intensive to a capital-intensive sector by using the consistent accumulation of capital. With capital being substitutable with energy, it will reduce the energy demand and promote energy conservation technology through the capital investment in each enterprise, thereby booming the tempo of the capital–energy substitution. More capital input could help to improve energy efficiency, and thus accomplish the goal of energy conservation in China's heavy industry [35]. Therefore, policies that aim at deepening financial mechanism reform should be prioritized. Energy supply shortages and GHG emissions may be mitigated by carrying out effective economic strategies and regulatory policies to decrease instituting rates. In addition, capital subsidies and favorable taxes targeted at lessening capital expenditure should also be implemented. Third, given that human capital serves as not only a catalyst but also an actor in inter-factor substitution, it becomes even more important to stimulate the factor switching process and reduce energy use by utilizing more human capital. There is immense potential for energy conservation to be liberated by obtaining more human capital accumulation, as China lags far behind the global average level in governmental expenditure on human capital investment. Policymakers' attention should be focused on enhancing direct financial support or subsidies for education and training, boosting the legislative progress, freeing up the labor market to spur private incentives and establishing a supervision mechanism to monitor the movement of government funds.

Although China has made considerable efforts to impel energy conservation and emissions reduction in the electricity sector, they still have a long way to go. To reconstruct a decarbonized electric power sector, the government's future policies should concentrate on: (1) accomplishing the reestablishment of an energy price system and the consummation of relevant regulations and laws, (2) stimulating human capital investment and utilization, (3) improving the appraisal system and supervision mechanism for the power sector in order to motivate the enthusiasm of governments and enterprises in energy conservation and emissions reduction, and (4) encouraging domestic fundamental research and sustainable development abilities.

This paper studied the sustainability of energy substitution in the Chinese electric power sector. Extensions of our paper could include applying our approach to study other energy issues (Lv, et al., 2019 [35]; Lean, et al., 2010, 2012, 2015 [24,30,36]), other resource issues (Woo, et al., 2012, [37]) and other industries (Guo, et al., 2010, 2015, 2018, 2019, 2020 [16,38–41]; Moslehpour, et al., 2018, 2019 [42,43]; Mou, et al., 2018 [44]; Nguyen, et al., 2020 [45]). Academics could also apply our approach to study other markets; for example, stock markets (Batmunkh, et al., 2018 [5]; Cheng, et al., 2019 [32]; Demirer, et al., 2019 [46]), funds (Gupta, et al., 2019 [47]; Abid, et al., 2014 [1]; Chiang, et al., 2008 [23]; Lean, et al., 2012 [36]; Levy, 2020 [48]; Li, et al., 2018 [29]), commodity (Yuan, et al., 2020) [49], warrant (Wong, et al., 2018 [50]) and futures (Clark, et al., 2016 [51]; Qiao, et al., 2012, 2013 [52,53]). There are many other issues to which academics can apply our approach; see, for example, Woo, et al., (2020) [54] and Chang et al., (2018) [55] for more information.

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