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District Heating Energy Consumption of the Building Sector in the Jing-Jin-Ji urban Agglomeration: Decomposition and Decoupling Analysis

Linghui Zhang ^{1,*}, Xin Ma ² and Shushen Zhang ¹

¹ Key Laboratory of Industrial Ecology and Environmental Engineering (MOE), School of Environmental Science and Technology, Dalian University of Technology, Dalian 116024, China; zhangss@dlut.edu.cn

² Department of Engineering Technology, Purdue University, 610 Purdue Mall, West Lafayette, IN 47907, USA; ma633@purdue.edu

* Correspondence: huey Zhang@mail.dlut.edu.cn

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Abstract: China's rapid urbanization has caused dramatically increasing energy consumption in the district heating systems of the building sector in the Jing-Jin-Ji urban agglomeration, and this change has led to enormous air pollution issues in this region. However, the drivers and the sustainable development process of the district heating system of the building sector have not been investigated to understand the management of energy conservation and emissions reduction in the Jing-Jin-Ji urban agglomeration. This study investigates the drivers of the district heating energy consumption of the building sector (DHEB) in the Jing-Jin-Ji urban agglomeration between 2004 and 2016 by developing a decomposition framework. The decoupling status between the DHEB and gross domestic product (GDP) is then analyzed based on the Tapio decoupling index. The results show that a weak decoupling effect is mainly found between the DHEB and GDP in the Jing-Jin-Ji urban agglomeration from 2004 to 2016. The increase in the DHEB in 2004–2016 is largely driven by the growth of the district heating area and population, while the heating energy intensity negatively contributes to the increase. Significant differences in the effects of the share of the energy mix and share of heat production technology were found between subregions in response to government policy, which impacted levels in Beijing, Tianjin, and Hebei in decreasing order.

Keywords: Jing-Jin-Ji urban agglomeration; district heating; building sector; decomposition; decoupling

1. Introduction

The ambition to limit global warming to 2 °C requires transformation to an efficient energy system by national efforts. In particular, implementation of efficient and clean district heating systems can significantly mitigate climate change, air pollution, and improve thermal comfort and cost benefits in the building sector [1]. China has the largest district heating system in the world, whose networks have surpassed 200,000 kilometers and can provide heat to nearly 9 billion square meters of building space [1]. In 2015, China's district heating system accounted for 40% energy consumption of the building sector [2], the energy consumption of which experienced a significant increase of 96% from 2004 to 2015 (Table A1).

As one of the economic centers and the most urbanized areas in China, Jing-Jin-Ji urban agglomeration experienced a dramatic increase of 120% (Table A1) in the district heating energy consumption of the building sector (DHEB) from 2004 to 2016 due to the rapid urbanization in population and building area (Table A2). Moreover, the Jing-Jin-Ji urban agglomeration is challenged by severe air pollution, especially in winter, when the pollution is caused by huge heating fuel combustion [3]. To address this problem, the Ministry of Environmental Protection launched the

“Issuance of the Action Plan for the Control of Air Pollution” [4] and a series of actions that aimed to curb air pollution in the Jing-Jin-Ji urban agglomeration [5]. Efforts have vigorously been made to promote clean energy generation and efficiency heating systems in the Jing-Jin-Ji urban agglomeration [6], and their effectiveness has been proved [7]. Thus, it is necessary to quantitatively evaluate the influential drivers and sustainable development process of the DHEB in the Jing-Jin-Ji urban agglomeration, as this information is crucial for understanding the management of energy conservation and emissions reduction, as well as the influence of urbanization and mitigation policy.

A number of studies have analyzed the driving factors of the changes in heating energy consumption and carbon emissions. In particular, some studies have analyzed the impact factors from an econometric-based view. For example, Bissiri et al. [8] used a nonparametric model to analyze the impact factors of residential heating energy consumption in terms of price, income and heating degree days (HDD). The results indicated that domestic heating energy consumption was highly responsive to HDD and price elasticities. In contrast, there are also studies that used index decomposition analysis to provide a complete understanding of underlying drivers [9] of residential building energy consumption, which is the largest share of energy consumption represented by space heating in many regions, such as Europe [10] and China [11]. For instance, attempts have been made to investigate the effects of residential building energy consumption and carbon emissions in terms of price and expenditure [12], end-use structure [11], residential consumption structure [13], urbanization [14], standard of living [15] and climate [16], the results of which can provide energy-saving policy implications for space heating.

Moreover, previous studies utilized a combined decomposition and decoupling analysis to assess the drivers and the decoupling status between economic growth and energy consumption, which allows to provide key indicators for sustainable development [17]. For the building sector, Liang et al. [18] investigated the driving factors of the carbon dioxide intensity in China’s residential building sector by combining a decoupling approach, and further identified their relationships of weak decoupling with economic development. Ma et al. [19] assessed whether carbon dioxide intensity of the commercial building sector decoupled from the economic growth in China via a combined decomposition and decoupling approach. Results showed that the nationwide decoupling status shifted from weak decoupling in 2001–2005 to strong decoupling in 2006–2015, and the Jing-Jin-Ji urban agglomeration was found with larger decoupling effect due to the rapid development of energy conservation projects.

Although these studies contribute to understanding the drivers and decoupling effect of building energy consumption and revealing potential implications for the domestic heating sector, several limitations have been identified in these existing studies: (1) to the best of our knowledge, very few studies have studied driving factors of the changes in the DHEB in terms of the effect of the share of the energy mix, the effect of the share of heat production technology, and other effects from the results of urbanization with the logarithmic mean Divisia index (LMDI) decomposition framework, along with their decoupling effects on the economic growth; (2) most previous decomposition studies targeting China’s residential building sector used energy data from the “Energy Balance Table” in the China Energy Statistical Yearbook, and this dataset cannot directly provide energy consumption data for the district heating systems of the building sector; (3) the driving factors of the changes in the DHEB and their decoupling status for the Jing-Jin-Ji urban agglomeration have not been explored.

The objectives of this study are to quantitatively evaluate the driving factors of the changes in the DHEB of the Jing-Jin-Ji urban agglomeration and its decoupling status from economic growth. Specifically, this study made several contributions. First, we concentrated on the effects of the share of energy mix, the share of heat production technology, heating energy intensity, population and district heating area by deploying decomposition analysis. Second, we acquired the DHEB data from the “China Urban–Rural Construction Statistical Yearbook” (CURCSY) [20], which can provide complete, transparent and consistent data for the district heating system in contrast to other existing data. Third, we conducted the study in the Jing-Jin-Ji urban agglomeration, which is a region experiencing rapid urbanization and clean energy actions in China.

The remainder of this paper is organized as follows. Section 2 describes the methodology and data. Section 3 provides the results of decomposition and decoupling. Section 4 provides the empirical discussions. Section 5 summarizes the main findings and conclusions.

2. Materials and Methods

2.1. Decomposition Analysis

Decomposition analysis methods have been extensively used to identify the contribution of a set of drivers on target variables. Two decomposition methods are commonly used: index decomposition analysis (IDA) and structure decomposition analysis (SDA). Based on the consideration of time-series sector detailed data, the IDA is more suitable for investigating the drivers behind decoupling and can be easily applied at any level of aggregation [21], while the SDA needs a complete input-out table [22,23]. Among various IDA methodologies, the logarithmic mean Divisia index (LMDI) method is the most popular method for studying energy and emission decomposition due to its path independence and consistency in aggregation [24].

In this study, we decomposed the DHEB in the Jing-Jin-Ji urban agglomeration by developing a decomposition framework based on LMDI approach as follows:

$$HE = \sum_i \sum_j HE_{ij} = \sum_i \sum_j \frac{HE_{ij}}{HE_j} \times \frac{HE_j}{HE} \times \frac{HE}{S} \times \frac{S}{P} \times P \quad (1)$$

$$HE = \sum_i \sum_j S_{ij} \times T_j \times I \times A \times P \quad (2)$$

where i and j represent, respectively, different fuel types (where $i = 1, 2, 3$ represents coal, natural gas and oil, respectively) and different heat production technology (where $j = 1, 2$ represents combined heat and power (CHP) and heating boilers, respectively); HE represents the DHEB; HE_{ij} represents the DHEB generated from heat production technology j by fuel type i ; HE_j represents the DHEB generated by heat production technology j ; and S and P represent the total district heating area and population, respectively.

$S_{ij} = HE_{ij}/HE_j$ is the proportion of the DHEB generated by fuel type i by heat production technology j in district heating systems, which represents the energy mix effect, and it is subdivided into the effect of the shares of coal, oil and gas in the entire DHEB; $T_j = HE_j/HE$ is the proportion of the DHEB generated by technology type j and represents the heat production technology structure effect, which is subdivided into the effect of the shares of CHP and heating boilers; $I = HE/S$ is the effect of the heating energy intensity and measures the district heating energy consumption per unit square meter of a building (kgce/m^2), which indicates the energy efficiency; $A = S/P$ represents the effect of the district heating area and measures floor area per person (m^2/person) equipped with a district heating system; and P represents the effect of population growth.

Thus, the changes of the DHEB can be further decomposed as follows:

$$\Delta HE = HE^t - HE^{t-1} \quad (3)$$

$$\Delta HE = S_{ij}^t \times T_j^t \times I^t \times A^t \times P^t - S_{ij}^{t-1} \times T_j^{t-1} \times I^{t-1} \times A^{t-1} \times P^{t-1} = \Delta HE_S + \Delta HE_T + \Delta HE_I + \Delta HE_A + \Delta HE_P \quad (4)$$

$$\Delta HE_S = \Delta HE_{\text{coal}} + \Delta HE_{\text{gas}} + \Delta HE_{\text{oil}} \quad (5)$$

$$\Delta HE_T = \Delta HE_{\text{CHP}} + \Delta HE_{\text{Boiler}} \quad (6)$$

$$\begin{aligned}
\Delta HE &= \Delta HE_{\text{coal}} + \Delta HE_{\text{gas}} + \Delta HE_{\text{oil}} + \Delta HE_{\text{CHP}} + \Delta HE_{\text{Boiler}} + \Delta HE_I + \Delta HE_A + \Delta HE_P \\
&= \sum_j^2 L(w_{1j}^t, w_{1j}^{t-1}) \ln\left(\frac{S_{1j}^t}{S_{1j}^{t-1}}\right) \\
&\quad + \sum_j^2 L(w_{2j}^t, w_{2j}^{t-1}) \ln\left(\frac{S_{2j}^t}{S_{2j}^{t-1}}\right) + \sum_j^2 L(w_{3j}^t, w_{3j}^{t-1}) \ln\left(\frac{S_{3j}^t}{S_{3j}^{t-1}}\right) \\
&\quad + \sum_i^3 L(w_{1i}^t, w_{1i}^{t-1}) \ln\left(\frac{T_1^t}{T_1^{t-1}}\right) + \sum_i^3 L(w_{2i}^t, w_{2i}^{t-1}) \ln\left(\frac{T_2^t}{T_2^{t-1}}\right) \\
&\quad + \sum_i^3 \sum_j^2 L(w_{ij}^t, w_{ij}^{t-1}) \ln\left(\frac{I_j^t}{I_j^{t-1}}\right) + \sum_i^3 \sum_j^2 L(w_{ij}^t, w_{ij}^{t-1}) \ln\left(\frac{A_j^t}{A_j^{t-1}}\right) \\
&\quad + \sum_i^3 \sum_j^2 L(w_{ij}^t, w_{ij}^{t-1}) \ln\left(\frac{P_j^t}{P_j^{t-1}}\right)
\end{aligned} \tag{7}$$

where $L(w_{ij}^t, w_{ij}^{t-1}) = (HE_{ij}^t - HE_{ij}^{t-1}) / (\ln(HE_{ij}^t) - \ln(HE_{ij}^{t-1}))$ is the logarithmic mean weight. ΔHE_{coal} , ΔHE_{gas} , ΔHE_{oil} are the changes of the DHEB owing to shifts in the proportion of coal, gas and coal; ΔHE_{CHP} and $\Delta HE_{\text{Boiler}}$ are the changes of the DHEB owing to shifts in the proportion of CHP and heating boilers, respectively; ΔHE_I , ΔHE_A and ΔHE_P are the changes of the DHEB owing to energy intensity changes, district heating area growth and population variation, respectively. t is the number of the specific year.

In this study, as there is no heating fuel information in the CURCSY data, we used heating fuel data from the China Energy Balance Table to estimate the heating fuel mix of the DHEB. This is because the heating fuel mix is determined at the heat production phase where the fuel source is completely transformed into heat before allocating to each end-use sector. In addition, we considered the heating fuel mix to be the same in the CHP and heating boilers. Therefore, the energy mix effect S_{ij} can be represented as follows:

$$S_{ij} = \frac{HE_{ij}}{HE_j} = \frac{HE_i^{\sim}}{HE^{\sim}} \tag{8}$$

where HE_i^{\sim} represents the total district heating energy consumption by fuel type i with the data from the China Energy Balance Table; HE^{\sim} represents the total district heating energy consumption with the data from China Energy Balance Table. The data for district heating fuel mix are obtained from "China Energy Balance Table" [25].

2.2. Decoupling Analysis

The Tapio decoupling model [26] is used to build the judgement standard for decoupling the DHEB from GDP in China. The decoupling indicator is expressed as Equation (9).

$$\varphi_{HE,GDP} = \frac{\Delta HE^{0 \rightarrow T} / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} \tag{9}$$

where $\varphi_{HE,GDP}$ is the decoupling level of the DHEB from GDP in China; HE^0 and GDP^0 represent the values of the DHEB and GDP in the base-year, respectively; and $\Delta HE^{0 \rightarrow T}$ and $\Delta GDP^{0 \rightarrow T}$ represent the changes of the DHEB and GDP over a period of ΔT , respectively. Table 1 shows the judgement standard of the decoupling status.

According to Equations (7) and (9), the extended decoupling model based on decomposition is expressed in Equation (10).

$$\begin{aligned}
 \varphi_{HE,GDP} &= \frac{\Delta HE^{0 \rightarrow T} / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} \\
 &= \frac{(\Delta HE_{coal}^{0 \rightarrow T} + \Delta HE_{gas}^{0 \rightarrow T} + \Delta HE_{oil}^{0 \rightarrow T} + \Delta HE_{CHP}^{0 \rightarrow T} + \Delta HE_{Boiler}^{0 \rightarrow T} + \Delta HE_I^{0 \rightarrow T} + \Delta HE_A^{0 \rightarrow T} + \Delta HE_P^{0 \rightarrow T}) / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} \\
 &= \frac{\Delta HE_{coal}^{0 \rightarrow T} / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} + \frac{\Delta HE_{gas}^{0 \rightarrow T} / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} + \frac{\Delta HE_{oil}^{0 \rightarrow T} / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} + \frac{\Delta HE_{CHP}^{0 \rightarrow T} / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} \\
 &\quad + \frac{\Delta HE_{Boiler}^{0 \rightarrow T} / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} + \frac{\Delta HE_I^{0 \rightarrow T} / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} + \frac{\Delta HE_A^{0 \rightarrow T} / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} + \frac{\Delta HE_P^{0 \rightarrow T} / HE^0}{\Delta GDP^{0 \rightarrow T} / GDP^0} \\
 &= \varphi_{coal} + \varphi_{gas} + \varphi_{oil} + \varphi_{CHP} + \varphi_{Boiler} + \varphi_I + \varphi_A + \varphi_P
 \end{aligned} \tag{10}$$

where the decoupling indicator $\varphi_{HE,GDP}$ of the DHEB from GDP can be decomposed into eight sub-indicators, which are the share of coal indicator (φ_{coal}), the share of gas indicator (φ_{gas}), the share of oil indicator (φ_{oil}), the share of CHP indicator (φ_{CHP}), the share of heating boilers indicator (φ_{Boiler}), the energy intensity indicator (φ_I), the heating area indicator (φ_A), and the population indicator (φ_P).

Table 1. Decoupling states and classification by the Tapio model.

State		$\Delta HE/HE$	$\Delta GDP/GDP$	$\varphi_{HE,GDP}$
Negative decoupling	Expansive negative decoupling	>0	>0	(1.2, +∞)
	Strong negative decoupling	>0	<0	(−∞, 0)
	Weak negative decoupling	<0	<0	[0, 0.8)
Decoupling	Weak decoupling	>0	>0	[0, 0.8)
	Strong decoupling	<0	>0	(−∞, 0)
	Recessive decoupling	<0	<0	(1.2, +∞)
Coupling	Expansive coupling	>0	>0	[0.8, 1.2]
	Recessive coupling	<0	<0	[0.8, 1.2]

2.3. Data

Jing-Jin-Ji urban agglomeration is the biggest urbanized region in Northern China, with two centrally and directly controlled municipalities (Beijing and Tianjin) and Hebei province (Figure A1). In this study, the accounting scope of the DHEB in the Jing-Jin-Ji urban agglomeration is the summed value of Beijing, Tianjin, and Hebei. Data describing the DHEB in 2004–2016 were obtained from the China Urban–Rural Construction Statistical Yearbook (CURCSY). In contrast to the “Energy Balance Table” of China Energy Statistical Yearbook and other statistics, the CURCSY database provides three advantages to account for the DHEB. (a) The district heating energy data from the CURCSY targets the building sector, including public buildings and residential buildings, which can provide an explicit and integral accounting scope for the building sector, rather than data classified by service activities (e.g., primary industry, secondary industry, tertiary industry, living sector). (b) The accounting scope for the district heating energy data in China Energy Balance Table is incomplete, and data only focus on large-scale heating enterprises (a main income larger than 20 million yuan or annual comprehensive energy consumption of more than 10,000 tons of standard coal). In contrast, CURCSY provides complete information for the district heating energy generators, which includes both the CHP and heating boilers (heating capacity both larger or less than 7 MW). (c) CURCSY simultaneously provides the district heating area of buildings, which parallels the DHEB data. In addition, the GDP values of Jing-Jin-Ji urban agglomeration are derived from “China Statistical Yearbooks 2005–2017” [27]. To eliminate the price inflation, the GDP values were completely adjusted to the constant price in 2004.

3. Results

3.1. Decomposition Analysis Results

Figure 1 illustrates the decomposition results for the changes in the DHEB in the Jing-Jin-Ji urban agglomeration from 2004 to 2016. The total DHEB of the Jing-Jin-Ji urban agglomeration experienced significant growth of 120% from 2004 to 2016, and it reached 52.3%, 18.3% and 22.0% during 2004–2008, 2008–2012 and 2012–2016, respectively. An increasing trend was also found in each subregion in 2004–2016, with growth rates of 184%, 80% and 85% in Beijing, Tianjin and Hebei, respectively (Figure 2). These increases reflect rapid development of district energy systems during urbanization in the Jing-Jin-Ji urban agglomeration in the past decade.

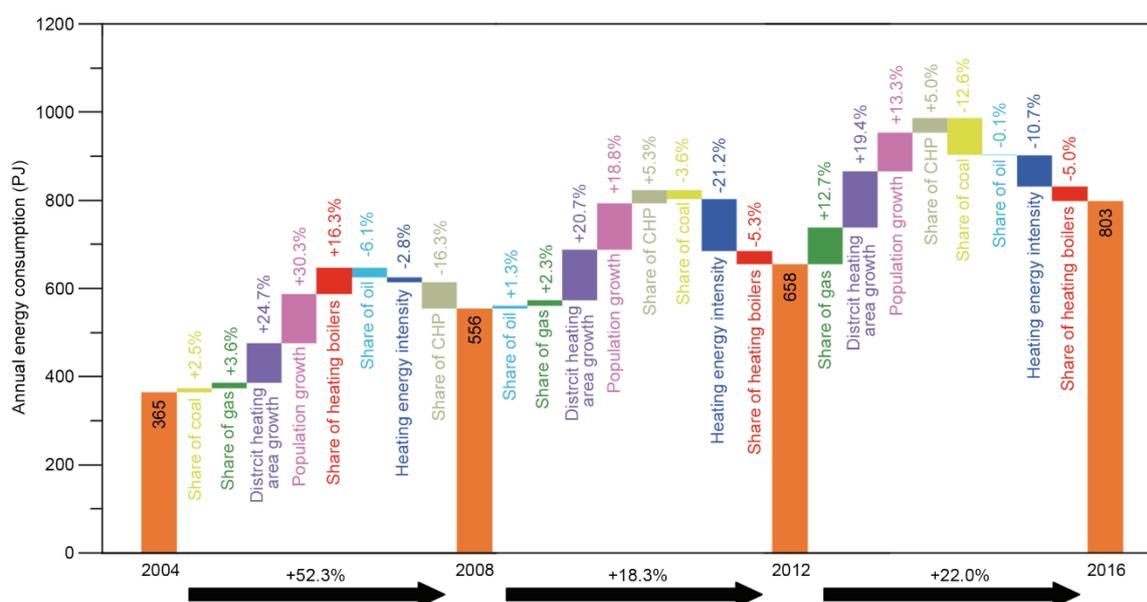


Figure 1. Contribution of drivers to the changes in district heating energy consumption of the building sector in the Jing-Jin-Ji urban agglomeration. Abbreviations: CHP (combined heat and power).

Regarding the indicators with positive contributions to the changes in the DHEB, the effects of district heating area and population always positively contributed to the increase in the DHEB of the Jing-Jin-Ji urban agglomeration during 2004–2016, as well as for each subperiod and subregion. The contribution of the heating area effect to the changes in the DHEB of the Jing-Jin-Ji urban agglomeration reached 24.7%, 20.7% and 19.4% during 2004–2008, 2008–2012 and 2012–2016, respectively, while that for the population effect was 30.3%, 18.8% and 13.3% during the same horizon, respectively (Figure 1). The effects of district heating area and population of Beijing, Tianjin and Hebei are shown as Figure 2.

Regarding the indicators with negative contributions to the changes in the DHEB, the effect of heating energy intensity always decreased the total DHEB of the Jing-Jin-Ji urban agglomeration during 2004–2016, which contributions reached -2.8% , -21.2% and -10.7% during 2004–2008, 2008–2012 and 2012–2016, respectively (Figure 1). Nonetheless, the effect of heating energy intensity appeared to positively contribute to the increase in the DHEB at the sub-regional level in Beijing (66.7% during 2004–2008) (Figure 2a) and Hebei (5.1% during 2012–2016) (Figure 2c).

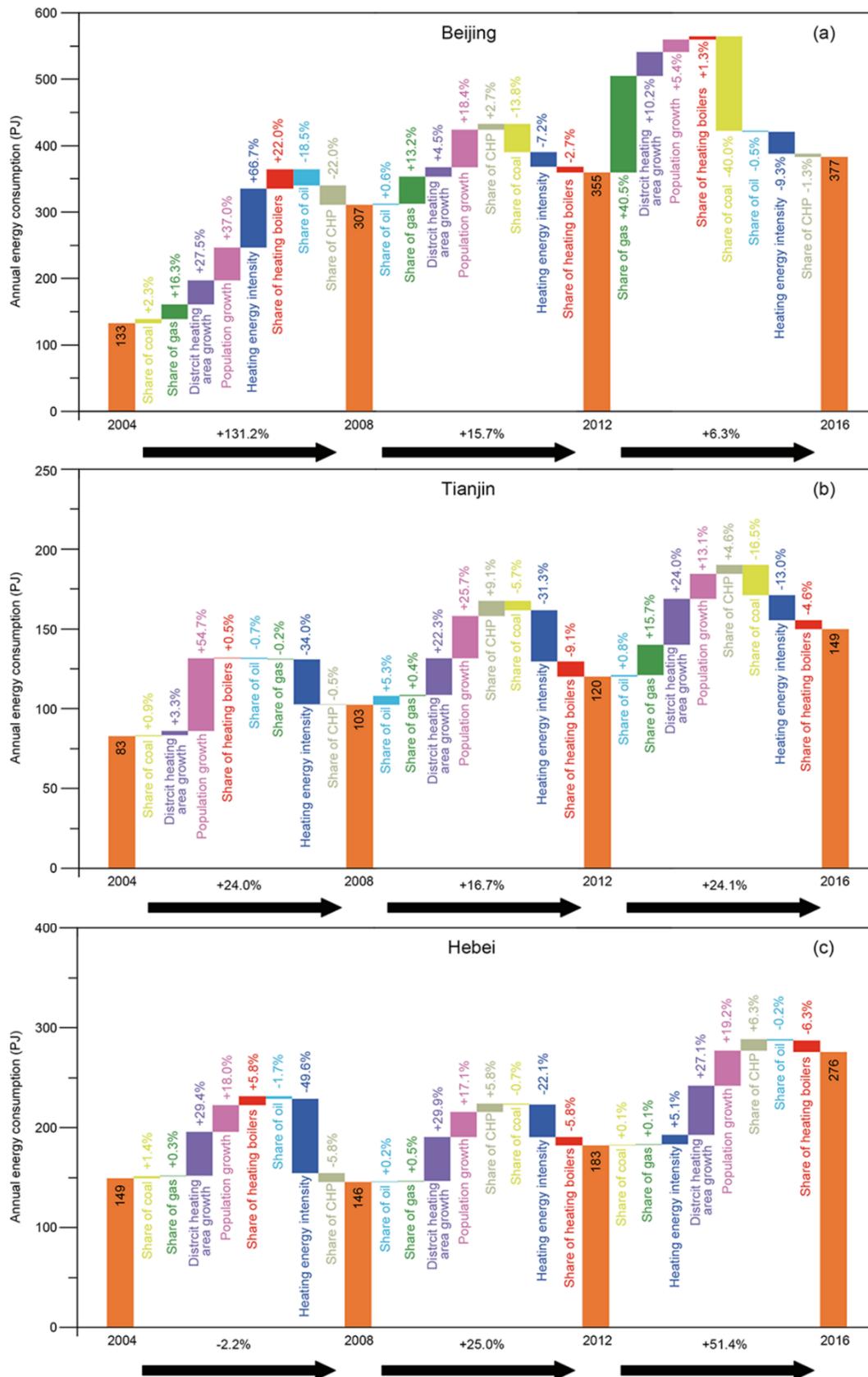


Figure 2. Contribution of drivers to the changes in district heating energy consumption of the building sector in three subregions of the Jing-Jin-Ji urban agglomeration. Abbreviations: CHP (combined heat and power).

The energy mix effect is described as a subdivision of the effects of the share of coal, oil and gas, which aims to illustrate the effect originating from the proportion of energy mix changed in the district heating system. Regarding the 2004–2008 period, the effect of the share of coal and the share of gas contributed 2.5% and 3.6% to the increase of the DHEB of Jing-Jin-Ji urban agglomeration, respectively, while the effect of the share of oil played a negative role with -6.1% (Figure 1). For the three subregions, effects of the share of coal, oil and gas of Beijing and Hebei had the same effect trend as that of the Jing-Jin-Ji urban agglomeration in 2004–2008, while the effect of share of gas had a negative effect on the increase in the DHEB in Tianjin. For the 2008–2012 period, the effect of the share of coal shifted to significantly decrease the DHEB of the Jing-Jin-Ji urban agglomeration, while the effect of the shares of oil and gas caused an increase. Regarding the 2012–2016 period, the effect of the share of coal played a more negative role in the increase of the DHEB of the Jing-Jin-Ji urban agglomeration (-12.6%), while the effect of the share of gas accelerated the increase in the DHEB (12.7%) (Figure 1). A similar effect trend was found in Beijing and Tianjin in 2012–2016, but it did not have a significant effect on Hebei's DHEB.

The heat production technology structure effect is subdivided into the effect of the share of CHP and the effect of the share of heating boilers. Regarding the 2004–2008 period, the effect of the share of boilers caused 16.3% of the increase in the DHEB of the Jing-Jin-Ji urban agglomeration, while the effect of the share of CHP contributed a negative figure (-16.3%). Regarding the 2008–2012 and 2012–2016 periods, the effect of the share of CHP had a positive effect, with values of 5.3% and 5.0%, respectively, while the effect of the share of heating boilers had a negative effect on the DHEB changes (Figure 1). In addition, both Tianjin and Hebei shared similar effect trends as that of the Jing-Jin-Ji region. However, the effect of the share of CHP negatively contributed to the increase in the DHEB in Beijing during the 2012–2016 period, which appeared to express a balanced state of installed capacity of CHP in the regional scale.

3.2. Decoupling Analysis Results

Table 2 indicates the decoupling status between the DHEB and GDP in the Jing-Jin-Ji urban agglomeration, and that for each subregion in 2004–2016, as well as the contribution of each subfactor. The total decoupling status between the DHEB and GDP in the Jing-Jin-Ji urban agglomeration mainly represents weak decoupling. Regarding each subperiod, the total decoupling status was gradually degraded from 2004–2008 ($\varphi_{HE,GDP} = 0.852$) to 2012–2016 ($\varphi_{HE,GDP} = 0.454$). The decoupling statuses of each subregion in three subperiods are shown as Table 2.

The total decoupling elasticity can be decomposed into a series of subfactors (φ_{coal} , φ_{oil} , φ_{gas} , φ_{CHP} , φ_{Boiler} , φ_I , φ_A , and φ_P). The sum of the values of three decoupling elasticity traces (φ_{coal} , φ_{oil} , φ_{gas}) nearly equaled zero between 2004 and 2016 (Table 2). Specifically, the decoupling status of φ_{coal} experienced a transition from weak decoupling in 2004–2008 to strong decoupling in 2008–2016. Moreover, the decoupling status of φ_{gas} showed weak decoupling between 2004 and 2016, and the decoupling status of φ_{oil} showed strong decoupling (2004–2008 and 2012–2016) and weak decoupling (2008–2012).

The sum of the values of φ_{CHP} and φ_{Boiler} nearly equaled zero during the time period of 2004–2016 (Table 2). The decoupling status of φ_{Boiler} was found to shift from weak decoupling (2004–2008) to strong decoupling (2008–2012 and 2012–2016), while the decoupling status of φ_{CHP} shifted from strong decoupling (2004–2008) to weak decoupling (2008–2012 and 2012–2016).

The decoupling status of φ_I showed strong decoupling (average value of -0.214) in 2004–2016, which was the most significant factor promoting the overall decoupling. The strongest decoupling status of φ_I was found during the period of 2008–2012, with value of -0.374 . Furthermore, the decoupling statuses of φ_A (average value of 0.389) and φ_P (average value of 0.367) consistently showed weak decoupling in 2004–2016 (Table 2). The decoupling statuses of each subfactor in each subregion in three subperiods are shown as Table 2.

Table 2. Decoupling statuses between district heating energy consumption of the building sector and economic growth in the Jing-Jin-Ji urban agglomeration in three subperiods.

Period	Indicator	Jing-Jin-Ji	Beijing	Tianjin	Hebei
2004–2008	$\varphi_{HE,GDP}$	0.852	2.354	0.315	−0.038
	φ_{coal}	0.041	0.041	0.012	0.024
	φ_{oil}	−0.099	−0.332	−0.009	−0.029
	φ_{gas}	0.059	0.292	−0.003	0.005
	φ_I	−0.046	1.197	−0.446	−0.831
	φ_A	0.403	0.493	0.044	0.492
	φ_P	0.495	0.664	0.717	0.302
	φ_{CHP}	−0.267	−0.395	−0.007	−0.097
	φ_{Boiler}	0.267	0.395	0.007	0.097
	$\varphi_{HE,GDP}$	0.323	0.560	0.145	0.303
2008–2012	φ_{coal}	−0.064	−0.495	−0.049	−0.008
	φ_{oil}	0.022	0.022	0.046	0.002
	φ_{gas}	0.041	0.473	0.003	0.006
	φ_I	−0.374	−0.256	−0.271	−0.267
	φ_A	0.365	0.160	0.193	0.363
	φ_P	0.332	0.657	0.223	0.207
	φ_{CHP}	0.094	0.098	0.079	0.070
	φ_{Boiler}	−0.094	−0.098	−0.079	−0.070
	$\varphi_{HE,GDP}$	0.454	0.238	0.229	0.831
	φ_{coal}	−0.259	−1.521	−0.158	0.001
2012–2016	φ_{oil}	−0.002	−0.019	0.008	−0.002
	φ_{gas}	0.261	1.540	0.150	0.001
	φ_I	−0.221	−0.355	−0.124	0.082
	φ_A	0.400	0.388	0.229	0.438
	φ_P	0.274	0.205	0.125	0.311
	φ_{CHP}	0.103	−0.051	0.044	0.102
	φ_{Boiler}	−0.103	0.051	−0.044	−0.102

4. Discussion

Section 3 presents the decomposition and decoupling results of the Jing-Jin-Ji urban agglomeration and each subregion in 2004–2016, and the major findings are then discussed below. The results showed that the increase in the DHEB of the Jing-Jin-Ji urban agglomeration was consistently driven by the growth of the district heating area and population in 2004–2016, while the decrease in heating energy intensity negatively contributed to the increase during this period. Similar effect trends were identified in each subregion and subperiod, excluding Beijing in 2004–2008 and Hebei in 2012–2016 (Figure 2). Regarding the factor of the effect of population growth, its positive contribution was consistent with IDA studies that targeted industrial and residential sectors [16,28,29]. Moreover, the significant, negative effect of the heating energy intensity acted as the key driver in reducing the DHEB. A significant, negative effect of energy intensity has also been found in other sectors by IDA studies, such as the productive sector [28] and electricity power sector [30].

The novel features for analyzing the effects of the share of the energy mix and share of heat production technology used in this paper have produced more detailed insights into the underlying drivers of the DHEB in the Jing-Jin-Ji urban agglomeration. The results showed that the driving effect of share of energy mix had significantly different features between the periods of 2004–2008 and 2008–2016 (Figure 1). This difference mainly was due to the transition of the heating energy mix in the Jing-Jin-Ji urban agglomeration put in place by Chinese government policy. For example, the Ministry of Environmental Protection launched the milestone policy of the “Issuance of the Action Plan for the Control of Air Pollution” (referred to as “Atmosphere Ten Plans”) in 2013, which aims to curb coal combustion and improve air quality [4]. In the same year, the “Detailed Rules for the Implementation of the Air Pollution Prevention and Control Action Plan in Jing-Jin-Ji Region and Surrounding Areas” [5] was issued in response to the “Atmosphere Ten Plans”, which depended on a joint prevention and

control method to mitigate air pollution in the Jing-Jin-Ji region. Such related policies led to significant changes in the heating energy mix and, thus, generated a great driving effect on the share of coal and share of gas to the changes of the DHEB in 2008–2016. Moreover, shifts between the positive and negative occurred regarding the effects of the share of CHP and the share of boilers in different subperiods. The significant shifts between positive and negative effects were largely associated with the implementation of the “elimination of old boilers” pushed by the Chinese government.

However, via the analysis in Section 3, at the subregion scale, we observed that the effects of the share of the energy mix and the share of the heat production technology were significantly different among regions. Thus, when interpreting the results, we must consider the different features in terms of policy and resource orientation between subregions. The different impact levels of the driving factors of the effects of the share of the energy mix and the share of the heat production technology of each subregion are further discussed in the following sections. Furthermore, considering the weak decoupling effect mainly found in the Jing-Jin-Ji urban agglomeration, it is suggested that the development of the district heating system of the building sector is no longer occurring at the expense of fast growth in energy consumption. The different impact levels of driving factors and decoupling modes of each subregion are also discussed in the following sections.

4.1. Beijing

For Beijing, the effects of the share of coal and the share of gas significantly affected the changes in the DHEB. For example, in 2012–2016, the effect of the share of coal contributed to -40.0% in the DHEB of Beijing (-0.1% of Hebei and -16.5% of Tianjin), while the effect of the share of gas caused a 40.5% increase in Beijing (0.1% of Hebei and 15.7% of Tianjin) (Figure 2). This result suggests that Beijing experienced the most advanced upgrading of the heating energy mix within the Jing-Jin-Ji urban agglomeration. In particular, the start of the transition of the heating energy mix in Beijing could be traced back to 2003 [31]. Coal combustion was gradually abandoned in Beijing’s core urban area and was replaced by clean energy, such as gas and electricity. Nonetheless, an acceleration of “coal to gas” has been seen since 2013, as the Beijing government started to implement the detailed rules of the “Atmosphere Ten Plans”. In 2014, the Beijing Environmental Protection Bureau designated a “coal abandoned area” in the whole urban area, and by the end of 2020, six districts in the city will be classified as being fully “coal abandoned” [32]. Our decomposition results, that the effect of share of gas contributes to the increase in the DHEB in Beijing in 2012–2016, proves the effect of the policy orientation.

However, regarding the effects of the share of CHP and the share of boilers, their effects on the changes in the DHEB tended to gradually decrease from 2004 to 2016 in Beijing. For example, the effect of the share of boilers reached 22.0% in 2004–2008, while it was only 1.3% in 2012–2016 (Figure 2a). This result suggests that the change in the DHEB in Beijing has become increasingly irrelevant to the transition in the share of heat production technology over time. This mainly is due to a large proportion that was represented by the coal-fired CHP before 2010 but was replaced by gas-fired units after 2010 in Beijing’s district heating energy system. Unexpectedly, the traces of the effects of the share of CHP and the share of boilers in Beijing are different from that of Tianjin and Hebei, where the effects of the share of CHP and the share of boilers in 2008–2012 and 2012–2016 are significantly larger than those in 2004–2008 (Figure 2a).

Based on the decoupling results, we found that Beijing experienced expansive negative decoupling in 2004–2008 and then shifted to weak decoupling in 2008–2016. The expansive negative decoupling was due to the unexpected increase in the statistical data of the DHEB in the year of 2008. The decoupling status of the share of the energy mix (φ_{coal} , φ_{oil} , and φ_{gas}) and the share of heat production technology (φ_{CHP} and φ_{Boiler}) fluctuated over time, and the sum of the values nearly equaled zero during the time period of 2004–2016. Specifically, the decoupling status of φ_{gas} showed an expansive negative decoupling in 2012–2016 (Table 2), which suggested that the development of Beijing’s district heating

system of the building sector was highly dependent on the use of gas in this period. In contrast, the role of coal was strongly decoupled in 2012–2016 ($\varphi_{\text{coal}} = -1.521$).

4.2. Tianjin

For Tianjin, the effects of the share of coal and the share of gas played an important role in the changes in the DHEB of Tianjin; however, their impact levels were much smaller than those of Beijing. Regarding the 2012–2016 period, the effects of the share of coal and share of gas were -16.5% and 15.7% in Tianjin (Figure 2b), and -40.0% and 40.5% for that in Beijing, respectively. This finding suggests that although a transition from coal to gas in the energy mix was experienced in Tianjin in 2012–2016 in response to the “Atmosphere Ten Plans”, there is still enough space for further implementation of gas in the urban district heating systems compared to the upgraded level of Beijing. Furthermore, the effect of the share of CHP in Tianjin appeared to positively contribute to the increase in the DHEB in 2008–2016. This finding suggested that the newly built CHP represented a greater proportion of energy in the district heating systems of Tianjin in 2008–2016.

The decoupling status of Tianjin mainly showed a weak decoupling during 2004–2016. In contrast to Beijing, the fluctuation of the decoupling status of Tianjin was slight. For example, regarding the period of 2012–2016, the decoupling status of Beijing experienced a shift from strong decoupling to weak decoupling and was then gradually reinforced, while that of Tianjin remained at the shallow levels of weak decoupling (0.304 – 0.387), excluding 2013–2014 (Table A3). This finding suggests that although the district heating system of Tianjin appears to represent weak decoupling, it still lacks sufficient drivers for decoupling compared to that of Beijing.

4.3. Hebei

For Hebei, lower impact levels were identified for the effects of the share of coal, share of oil, and share of gas on the changes in the DHEB in 2004–2016 compared to those in Beijing and Tianjin. This finding suggests that limited promotion has been achieved for the energy mix in Hebei’s district heating systems. As an important part of Jing-Jin-Ji, both the energy mix and the heat production technology structure need to be upgraded in response to the “Atmosphere Ten Plans”. However, the upgrading of the energy mix is not significant in the urban district heating systems of Hebei (Figure 2c). This result could mainly be due to the urgent priority to carry out “coal replacement” in rural households to mitigate air pollution in Hebei during this period [33], rather than for the urban district heating system. Thus, it is necessary to accelerate the optimization of the primary fuel mix in Hebei’s urban district heating system in the future. Furthermore, the effect of the share of CHP of Hebei appears to represent a similar development trend as that of Tianjin, which showed a positive contribution to the increase in the DHEB in 2008–2016. The decoupling status of Hebei also had a similar trend as that of the Jing-Jin-Ji urban agglomeration, which mainly showed weak decoupling during 2004–2016, although the decoupling status of each subsector fluctuated between strong decoupling and weak decoupling.

5. Conclusions

In this study, a novel decomposition framework was developed to analyze the driving factors of the district heating energy consumption of the building sector (DHEB) in the Jing-Jin-Ji urban agglomeration in 2004–2016. The drivers include the effects of the shares of coal, gas and oil, the effects of the shares of combined heat and power (CHP) and heating boilers, the heating area effect, the heating energy intensity effect, and the population effect. Then, the Tapio decoupling model was used to investigate the decoupling states between the DHEB and GDP of the Jing-Jin-Ji urban agglomeration, along with the decoupling effect of each subsector. A significant effect of the shares of the energy mix and the shares of heating production technology was identified in the Jing-Jin-Ji urban agglomeration in 2004–2016 in response to government policy. However, different features in terms of policy and

resource orientation were also found between Beijing, Tianjin, and Hebei in the Jing-Jin-Ji urban agglomeration in 2004–2016 in response to government policy. The main conclusions are as follows:

(1) The main decoupling status between the DHEB and economic growth of the Jing-Jin-Ji urban agglomeration showed weak decoupling in 2004–2016. The increase in the DHEB of the Jing-Jin-Ji urban agglomeration was consistently driven by the growth of the district heating area and population in 2004–2016, while the heating energy intensity played a negative role. The effect of the share of coal experienced a transition from a positive effect in increasing DHEB in 2004–2008 to a negative effect in 2008–2016, while the effect of the share of gas consistently played a positive role in 2004–2016. The effect of the share of CHP negatively contributed to the increase in the DHEB in 2004–2008 and had a significant, positive effect in 2008–2016. In contrast, the effect of the share of heating boilers had an adverse effect on the DHEB changes during the same period.

(2) Beijing experienced expansive negative decoupling 2004–2008 and then shifted to weak decoupling in 2008–2016. The effects of the share of coal and the share of gas significantly affected the changes in the DHEB. Beijing has the most advanced upgrades of the heating energy mix within the Jing-Jin-Ji urban agglomeration. However, the effects of the share of CHP and the share of boilers tended to degrade over time due to the large share of CHP in history.

(3) The fluctuation of the decoupling status of Tianjin was very slight and showed weak decoupling during 2004–2016. A significant transition from coal to gas in the energy mix was experienced in Tianjin in 2012–2016 in response to the “Atmosphere Ten Plans”, but the level was much smaller than that of Beijing. CHP appeared to represent a greater proportion of energy in the district heating systems of Tianjin over time.

(4) The decoupling status of Hebei mainly showed weak decoupling during 2004–2016. Limited promotion has been achieved for transitioning from coal to gas in the energy mix in Hebei’s district heating systems compared to that seen in Beijing and Tianjin, which occurs in response to the “Atmosphere Ten Plans”. Furthermore, the effect of the share of CHP in Hebei positively contributed to the increase in the DHEB in 2008–2016.

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Appendix A

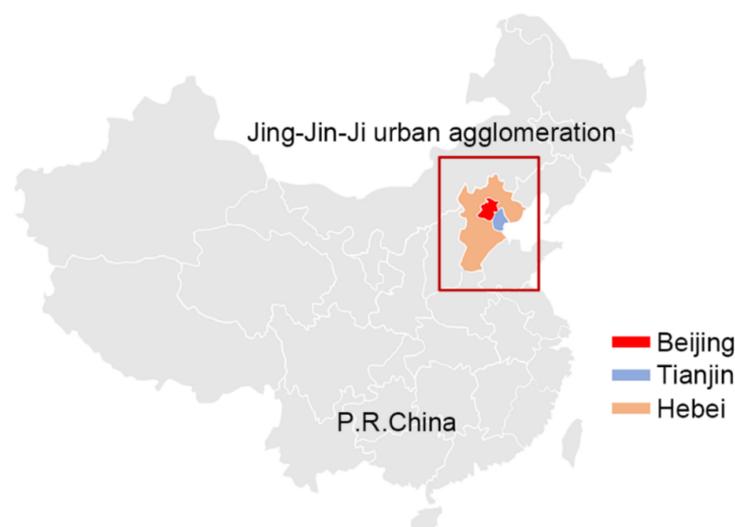


Figure A1. Map of Jing-Jin-Ji urban agglomeration.

Table A1. The energy consumption and growth rate of district heating systems of the building sector in China and Jing-Jin-Ji urban agglomeration from 2004 to 2016.

	2004	2005	2006	2007	2008	2009	2010	2011	2012	2013	2014	2015	2016
DHEB (PJ) (China)	1738	1873	1979	2070	2336	2429	2685	2638	2786	3020	3129	3400	3545
GR (%) (C)	-	8	14	19	34	40	54	52	60	74	80	96	104
DHEB (PJ) (JJJ)	365	408	428	406	556	613	640	645	658	702	728	756	803
GR (%) (JJJ)	-	12	17	11	52	68	75	77	80	92	99	107	120

Note: Data of DHEB is originated from the China Urban–Rural Construction Statistical Yearbook [20]. The DHEB represents the district heating energy consumption of the building sector; GR represents the growth rate of DHEB in a specific year compared to 2004; JJJ is the abbreviation of Jing-Jin-Ji urban agglomeration.

Table A2. The district heating area and population in the Jing-Jin-Ji urban agglomeration from 2004 to 2016.

	2004	2005	2006	2007	2008	2009	2010	2011	2012	2013	2014	2015	2016
Area (million m ²) (JJJ)	568	643	691	760	885	954	1094	1200	1272	1377	1433	1549	1710
Area (million m ²) (BJ)	282	317	350	372	425	442	467	508	526	546	568	585	611
Area (million m ²) (TJ)	114	140	151	169	192	206	240	272	300	329	342	377	418
Area (million m ²) (HB)	172	186	189	219	267	306	387	420	447	502	523	588	681
Pop (million) (JJJ)	42	47	48	51	53	56	59	61	63	66	67	70	72
Pop (million) (BJ)	12	13	14	14	15	16	17	17	18	18	19	19	19
Pop (million) (TJ)	6	8	8	9	9	10	10	11	12	12	12	13	13
Pop (million) (HB)	24	26	27	28	29	31	32	33	34	35	36	38	40

Note: Data of district heating area originated from the China Urban–Rural Construction Statistical Yearbook [20]. Data of population originated from the China Statistical Yearbook [27]. JJJ is the abbreviation of Jing-Jin-Ji urban agglomeration; BJ is the abbreviation of Beijing; TJ is the abbreviation of Tianjin; HB is the abbreviation of Hebei; Pop is the abbreviation of population.

Table A3. Total decoupling status between district heating energy consumption of the building sector and economic growth in the Jing-Jin-Ji urban agglomeration.

Period	$\Phi_{HE,GDP}$ (Jing-Jin-Ji)	$\Phi_{HE,GDP}$ (Beijing)	$\Phi_{HE,GDP}$ (Tianjin)	$\Phi_{HE,GDP}$ (Hebei)
2004–2005	0.907	1.615	1.021	0.281
2005–2006	0.366	0.007	0.297	0.715
2006–2007	−0.368	−0.435	−0.016	−0.531
2007–2008	2.590	9.288	0.174	0.553
2008–2009	0.849	2.682	0.245	−0.236
2009–2010	0.333	−0.017	0.232	0.658
2010–2011	0.063	−0.131	−0.241	0.480
2011–2012	0.150	−0.247	0.385	0.161
2012–2013	0.547	−0.614	0.304	1.648
2013–2014	0.346	0.581	−0.096	0.606
2014–2015	0.357	0.238	0.298	0.480
2015–2016	0.549	0.702	0.387	0.546

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