



Electricity-Water Consumption and Metropolitan Economic Growth: An Empirical Dual Sectors Dynamic Equilibrium Model

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This paper develops a dual sectors dynamic equilibrium model and introduces electricity consumption and water consumption in a growth model that tested by using a time series data set from 1950 to 2014 in Guangzhou, China. It presents a theoretical prediction on the interactions between electricity consumption, water consumption, and the metropolitan economic growth. Consistent with this prediction, electricity consumption and water consumption by themselves appear to have significant effects on metropolitan economic performance. The cointegration techniques show that electricity consumption, water consumption, and the metropolitan economic growth is positively correlated with electricity consumption. Also consistent with the theory, water consumption is positively associated with metropolitan economic performance. These results are generally stable and hold with alternative measures of unit roots, with alternative estimation strategies, and with or without controlling for trends, intercepts, and break points.

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1 INTRODUCTION

Casual empiricism suggests the presence of significant differences in metropolitan economic growth in China. For example, China's electricity consumption, heavy reliance on fossil fuel, and watering down environmental regulations have significantly affected the economic development of the country. Nevertheless, there has not been a systematic analysis of long-run differences in metropolitan economic growth with both electricity consumption and water consumption. Our primary motivation in this paper is to make a first attempt at such a systematic analysis and to investigate the relationship between electricity consumption, water consumption, and metropolitan economic growth in Guangzhou, China.

Electricity is one of the most studied energy classes—and for good reason. Growing demands for electricity and water resources and increasingly serious environmental challenges, particularly in metropolitan areas of developing countries like China, have elevated the urgency of studying the water-energy-growth nexus (Lai et al., 2011). China is the largest emitter of greenhouse gases (GHGs) in the world. In 2012, China was the largest contributor to carbon emissions, and with 8.50 Gt in CO_2 emissions from fossil burning and cement production in 2012, China was responsible for 25% of global carbon emissions, which was equivalent to the emissions of the U.S. and the E.U. combined

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(Liu, 2015). Between 2000 and 2010, the electricity production in China had increased threefold and accounted for 50% of domestic and 12% of global CO_2 emissions in 2010. For these reasons alone, researchers, policy-makers, consumers, and power plants all have a significant interest in accurately valuing the nexus between electricity consumption and economic development.

In addition to emitting large amounts of GHGs, electricity production also requires significant amounts of water, so water use is an important factor for electricity sector and for sustainable development in China. Continuous economic growth in China can have significant implications for water resources and environmental quality. Therefore, it is important to understand the interdependencies between electricity consumption, water use, and economic growth particularly when it comes to formulating policies targeted at environmental quality protection and sustainability.

Since the founding of China in 1949, electricity and water consumption has become an important issue for the economic development during the process of industrialization and urbanization (He et al., 2017). After 1978 China enforced the openness and reform policy, electricity consumption, water consumption, and economic performance increase dramatically (Kahrl and Roland-Holst, 2008; Jiang et al., 2014). Continuous economic growth of China will have significant implications for electricity and water. Therefore, understanding of the implications of the electricity and water use policies is an important factor for policy formulation targeted at driving economic growth. Moreover, the electricity production is related with the water use. Hence, water withdrawal is also an important factor for economy sustainable growth in China (Feng et al., 2014).

Electric grid and water infrastructure have been critical factors for the economic growth, industrialization, and urbanization of China since founding in 1949. In particular, energy consumption and water use increased dramatically after 1978 when China enacted the openness and reform policies that stimulated economic growth (Kahrl and Roland-Holst, 2008; Jiang et al., 2014). Over time, maintaining balance between electricity supply and demand in China required increase in power generation, which has been primarily based on thermal power generation from coal until recent years (Ohdoko et al., 2013). In 2001, thermal power generation accounted for 81.2% of total power generation, and 95% of thermal power was produced in coal power plants (Zhu et al., 2005). However, concerns have been mounting pertaining to emissions from coal power plant operations including carbon dioxide, nitrogen oxides, sulfur oxides, and particulate matter, including PM2.5 and PM10.

Water use in China rose by 12.47% from 2000 to 2013. On average, 64.06% of water use is attributed to agricultural industry, 22.66% to industry, 11.91% to household use, and 1.36% to biology use from 2000 to 2013 (Zhang et al., 2016). In Guangzhou, 18.43% of water use came from industry, 24.25% of water use came from public services, 55.19% of water use came from household, and others are occupied at 2.13% in 2014.

Guangzhou is the third largest metropolitan area in China, after Beijing and Shanghai, and the largest city in south central

China. Moreover, the Guangzhou Statistical Division provides the most complete and longest duration time series dataset (from 1950 to 2014) among the metropolitans in China. It helps us to observe and estimate the electricity consumption function with institutional reform. Throughout its 2100-year history, Guangzhou, the capital of Guangdong province, has been a major commercial center in south China. From 1949 to 1977, collectivist policies abolished much of the trade and commercial activities long associated with Guangzhou. The major breakthrough for repositioning of Guangzhou as a regional commercial hub came with implementation of the open-door policy in 1978, which allowed for international trade and investment introduced. In 1950, the GDP of Guangzhou reached \$43 million and per capita GDP was \$16. However, in 1978, the GDP reached \$0.72 billion and per capita was \$149. After economic reform, in 2014, the GDP reached \$278,448 million and per capita GDP was \$21419 (GSD, 2015). The 1950 census found the population of Guangzhou to be 2.5 million. As of 2014, it was estimated at 13 million. The industry sector accounted for 33.46% of its GDP in 2014, agriculture accounted for 1.32%, and the service sector represented 65.22%. Major industries in Guangzhou include automobiles, electronics, and petrochemicals.

Electricity in Guangzhou comes from the power stations in Guangdong province. There are 33 coal-, gas-, and fuel oil-based power stations, five nuclear power stations, three hydro power stations, three wind power stations, and four pumped-storage power stations in Guangdong province. Except for Zhujiang, Huangpu, Guangzhou (coal-, gas-, and fuel oil-based), and Guangzhou Pumped Storage Power Station, all other power stations are outside Guangzhou. The grid that supplies electricity to the city is Guangzhou Power Supply Co., Ltd. (GZPS). GZPS is one of the enterprises owned by China Southern Power Grid Co., Ltd. (CSG), which is state owned. In 1996, the first "People's Republic of China Electricity Law" was enforced and followed by a series of supporting laws, marking the electricity industry of Guangzhou into legal constraint. This "Electricity Law" makes it clear on electricity infrastructure construction, electricity production and management, electricity supply and use, supervision, and legal responsibility. On the other hand, in 2005, the Chinese Council promulgated the "Electricity Control System", which marked the electricity industry of Guangzhou has established a clear monitoring framework. Meanwhile, the "Electricity Business License", "Electricity Markets Operation Basic Management", and "Electricity Markets Supervision Measures" are the three regulations that came into effect. As a result, the electricity consumption in Guangzhou grows following the economic growth.

The water in Guangzhou comes from Nanzhou Water Supply Plant (NWSP) in Guangzhou that puts into production at fullscale with the capacity of one million cubic meters per day and from Guangzhou Water Supply Co. (GWSC) that is a large stateowned water supply enterprise and provides all water supply services, including water treatment and and diversified business development. The electricity of Guangzhou comes from the power stations in Guangdong province. There are 33 coal-,





gas, and fuel oil-based power stations, five nuclear power stations, three hydro power stations, three wind power stations, and four pumped-storage power stations in Guangdong province. Except for Zhujiang Power Station, Guangzhou Huarun Thermal Power Station, and Guangzhou Pumped Storage Power Station, other power stations are outside Guangzhou that is the capital of Guangdong province. The water in Guangzhou comes from Nanzhou Water Supply Plant (NWSP) in Guangzhou that puts into production at full-scale with the capacity of one million cubic meters per day and from Guangzhou Water Supply Co. (GWSC) that is a large state-owned water supply enterprise and provides all water supply services, including water treatment and diversified business development. Figure 1 shows the increase in water consumption for Guangzhou, China, from 1950 to 2014. Before 2004, the water in Guangzhou just came from state-owned NWSP and GWSC. NWSP was built in 2004

and put into production at full-scale with the capacity of one million cubic meters per day. It is the largest advanced water treatment plant in the nation. GWSC is a large state-owned water supply enterprise and provides all water supply services, including water treatment.

From **Figure 2**, it is found that the metropolitan economy grow after 1980s. Actually, China is a marketization transition country, and the Chinese government enforced the Open and Reform Policy at 1978. Since then, the economic institution in China gradually has been transferred from planning system into market system (Xu, 2011). In particular, being the capital of Guangdong province and the third largest Chinese city, Guangzhou was the earliest metropolitan in China to enforce that marketization policy. Furthermore, 1994 was the other important year for Chinese metropolitan economic development. Because the Chinese tax institution was



transferred from the tax contracting system into the tax sharing system at 1994 (Li and Kung, 2015), the metropolitan economy grows dramatically after 1994.

GDP in Guangzhou is divided into three industrial sectors product values: primary industry (agriculture), secondary industry (mainly manufacture and construction), and tertiary industry (service, business, tourism, etc.). In 2014, the ratios of agriculture, manufacture, and service respect to GDP are 1.31%, 33.47%, and 65.22%, respectively. The total electricity consumption of agriculture, manufacture, and service are 575,520,000 (kWh), 39,527,160,000 (kWh), and 20,421,400,000 (kWh) for Guangzhou in 2014, respectively. The water for production use, public services, household use, and others are 296,263,400 (cu.m), 389,842,300 (cu.m), 8,87,217,300 (cu.m), and 34,117,600 (cu.m) for Guangzhou in 2014, respectively. The electricity of Guangzhou comes from the power stations in Guangdong province. There are 33 coal-, gas, and fuel oil-based power stations, five nuclear power stations, three hydro power stations, three wind power stations, and four pumped-storage power stations in Guangdong province. Except for Zhujiang Power Station, Guangzhou Huarun Thermal Power Station, and Guangzhou Pumped Storage Power Station, other power stations are outside Guangzhou that is the capital of Guangdong province.

Therefore, the time series data with 65 observations from Guangzhou to investigate the interactions between electricity consumption, water consumption, and metropolitan economic growth were carried out. This dataset from Guangzhou can provide us a long-run data sample, and the data are all historic and realistic, which is the only metropolitan level statistical dataset including the data from 1950 to 2014 in China.

The rest of the paper is organized as follows. **Section 2** reviews related literature. **Section 3** introduces the theoretical framework, characterizes the equilibrium, and derives the major hypotheses and results. **Section 4** provides the data description and econometric strategy. **Section 5** shows empirical results. **Section 6** concludes.

2 LITERATURE REVIEW

Actually, limited attention has been devoted in academic literature to the nexus between water consumption and economic growth, and studies have provided evidence that national per capita water consumptive use seem to follow an inverted U-shaped path, with respect to per capita income, consistent with Kuznets type of relation. The research finds some support for the existence of an Environmental Kuznets Curve, but results are highly dependent on choice of datasets and statistical technique. For example, Katz (2015) presented the results of the analysis of the relationship between national per capita water use and per capita income using international cross-sectional data and panel data for OECD nations and U.S. states. He found that the water consumption initially rise and then decline with respect to income.

Barbier (2004) provided strong support for the inverted U relationship between economic growth and the rate of water use across countries. Barbier and Chaudhry (2014) showed that, in urban counties in the U.S., higher water use and population growth are associated with increase in income per capita. Cazcarro et al. (2013) found that a significant growth of per capita income has been the main factor driving the increase in water consumption growth. Ngoran et al. (2016) suggested that economic growth in 38 Sub-Saharan African countries for the period 1980–2011 is driven mainly by water and labor. Hence, their study supports the growth hypothesis for water consumption.

However, Gleick (2003) even found no relationship between per capita national water consumption and income, which is consistent with neutrality hypothesis. Obviously, previous empirical studies above have come under scrutiny in distinct literatures, and the literatures remain disjointed. Although some empirical researches have calculated empirically the links between water consumption GDP, they do not examine the causality among GDP and water consumption, comparing empirical results among different metropolitans. According the literature above, the nexus between electricity consumption and economic output has been extensively studied, but the evidence so far is contradictory and inconclusive (Stern et al., 2018).

On the other hand, Magazzino (2014) applied time series methodologies to examine the causal relationship among electricity demand, real per capita GDP, and total labor force for Italy from 1970 to 2009. The results of estimation indicate that one cointegrating relationship exists among these variables. This equilibrium relation implies that, in the long-run, GDP and labor force are correlated negatively, as well as GDP and electricity. Moreover, there is a bidirectional Granger causality flow between real per capita GDP and electricity demand; whereas labor force does not Granger cause neither real per capita GDP nor electricity demand. This implies that electricity demand and economic growth are jointly determined at the same time for the Italian case. The forecast error variance decomposition shows that forecast errors in real per capita GDP are mainly caused by the uncertainty in GDP itself, whereas forecast errors in labor force are mainly resulted from the labor force itself, although aggregate income and electricity are important, too.

Magazzino (2017) investigated the stationary properties of electric power consumption in 18 countries in the Middle East and North Africa by using yearly data over the period 1971–2013. After having controlled for the presence of cross-sectional dependence, the "second-generation" panel unit root tests reveal mixed results.

However, concerning the academic literature, there are limited studies about the nexus among economic growth, water consumption, and electricity consumption. For example, Wang et al. (2016) only discussed the electricity consumption in Beijing. Because it is the third largest urban economy in China, that is a surprising omission from the resource economics and China metropolitan economy issue literature (Mele and Magazzino, 2020). Udemba et al. (2020) conducted an empirical research to analyze the relationship between pollutant emission, energy consumption, foreign direct investment, and economic growth.

Indeed, there is a large literature on related problems, including Rehman and Deyuan, (2018), Rehman, (2020), Rehman et al. (2020), Rehman et al. (2021a), Rehman et al. (2021b), Rehman et al. (2021c), Rehman et al. (2021d), Muhammad Awais Baloch et al. (2021), Andrew Adewale Alola et al. (2021), Bright Akwasi Gyamfi et al., 2021, Festus Victor Bekun et al. (2021a), Festus Victor Bekun et al. (2021b), and Festus Victor Bekun et al. (2021a). For example, He (2020) identified the causal effects of urbanization and metropolitan economic performance on electricity consumption by a dynamic general equilibrium theoretical framework and time series econometric models using data from 1949 to 2016 in Guangzhou, China. Both autoregressive distributed lag (ARDL) and Johansen cointegration techniques show that electricity consumption, urbanization, and metropolitan economic performance are cointegrated. The results of twostage least squires demonstrate that the marginal propensity to electricity consumption in Guangzhou is about 1.28.

Meanwhile, He and Fullerton (2020) examined the nexus between water consumption and economic growth. Water consumption function is derived using an optimal dynamic equilibrium model. Two instrument variable models are proposed with real per capita economic output specified as a function of institutional reform and urbanization, which are used to examine the nexus among water consumption, reform, urbanization, and economic growth in Guangzhou, China.

On the basis of the literature above, this study builds a dual sectors dynamic equilibrium model where the economic performance is expressed as a linear combination of electricity consumption and water consumption. The objectives and contributions of this paper are both theoretical and empirical. The theory described in this paper 1) applied dynamic general equilibrium models to metropolitan economic growth using a social optimal method and 2) describes an estimation procedure that connects electricity consumption and water consumption and also yields interpretable implications. Empirically, this study finds the following: 1) electricity consumption, water consumption, and the metropolitan economic performance have long-run equilibrium relationship; 2) the kernel-based regularized least squares (KRLS) approach reveals that metropolitan economic growth is positively correlated with electricity consumption and water consumption is positively associated with metropolitan economic performance; and 3) results of unit roots are generally robust with alternative estimation strategies and with or without controlling for trends, intercepts, and break points.

3 THEORETIC MODEL

3.1 Electricity Sector

In the model of this study, metropolitan economy is composed of only electricity and water sectors. There are many companies (utilities) in both sectors. Consider that the problem of a representative company i in electricity sector cares about the net benefit and production function per capita is supposed as the Cobb-Douglas form:

 $y_i^e(k_i^e, e_i^e) = A_i^e(k_i^e)^a(e_i)^b (0 < a < 1, 0 < b < 1).$ The maximization problem can be written as follows:

$$\max_{k_i^e, e_i^e} NB_i^e = A_i^e \left(k_i^e\right)^a \left(e_i\right)^b - \left(pe_i + \theta k_i^e\right),\tag{1}$$

where A_i^e denotes the technology level in the *i*th electricity company, k_i^e is capital per capita in the *i*th electricity company, e_i is electricity per capita in the *i*th electricity company, p is price of electricity, and θ is interest rate. The first-order conditions of **Equation 1** are as follows:

$$\frac{\partial NB_i^e}{\partial k_i^e} = aA_i^e \left(k_i^e\right)^{a-1} \left(e_i\right)^b - \theta = 0, \tag{2}$$

$$\frac{\partial NB_i^e}{\partial e_i^e} = bA_i^e \left(k_i^e\right)^a \left(e_i\right)^{b-1} - \mathbf{p} = 0.$$
(3)

Combining Equations 2 and 3, the following was obtained:

$$k_i^e = \frac{a}{b} \frac{p}{\theta} e_i. \tag{4}$$

Suppose there are n homogeneous companies in electricity sector, so the electricity consumption per capita in the whole sector $E = ne_i$ and capital per capita in the whole sector

 $K^e = nk_i^e = n\frac{a}{b}\frac{p}{\theta}e_i = \frac{a}{b}\frac{p}{\theta}ne_i = \frac{a}{b}\frac{p}{\theta}E$. Hence, given technology level in the whole sector is identical $A_i^e = A^e$, the output per capita of electricity sector is as follows:

$$Y^{e} = ny_{i}^{e} = nA_{i}^{e} \left(k_{i}^{e}\right)^{a} \left(e_{i}\right)^{b} = nA^{e} \left(\frac{K^{e}}{n}\right)^{a} \left(\frac{E}{n}\right)^{b},$$

$$= n^{1-a-b}A^{e} \left(K^{e}\right)^{a} \left(E\right)^{b} = n^{1-a-b}A^{e} \left(\frac{a}{b} \frac{P}{\theta}E\right)^{a} \left(E\right)^{b}, \qquad (5)$$

$$= n^{1-a-b}A^{e} \left(\frac{a}{b} \frac{P}{\theta}\right)^{a} \left(E^{e}\right)^{a+b}.$$

3.2 Water Sector

Furthermore, the economy activities in water sector are considered. Suppose the production function per capita in the *j*th water company also takes C-D form with the price of water (r):

$$y_j^w \left(k_j^w, w_j \right) = \mathcal{A}_j^w \left(k_j^w \right)^a \left(w_j \right)^b.$$
(6)

The maximization problem can be written as follows:

$$\max_{k_j^w, e_j^w} NB_j^w = A_j^w (k_j^w)^a (w_j)^b - (rw_j + \theta k_j^w).$$
(7)

The first-order conditions of Equation 7 are as follows:

$$\frac{\partial NB_i^{\omega}}{\partial k_j^{\omega}} = a \mathcal{A}_j^{\omega} \left(k_j^{\omega} \right)^{a-1} \left(w_j \right)^b - \theta = 0, \tag{8}$$

$$\frac{\partial NB_i^w}{\partial k_j^w} = aA_j^w \left(k_j^w\right)^{a-1} \left(w_j\right)^b - \theta = 0.$$
(9)

Combining Equations 8 and 9, the following was obtained:

$$k_i^w = \frac{a}{b} \frac{r}{\theta} w_j. \tag{10}$$

Suppose there are m homogeneous water companies in this sector, water consumption per capita in the whole sector $W = mw_i$ and capital per capita in the whole sector.

$$K^{w} = mk_{j}^{w} = m\frac{a}{b}\frac{r}{\theta}w_{j} = \frac{a}{b}\frac{r}{\theta}mw_{j} = \frac{a}{b}\frac{r}{\theta}W.$$
 (11)

Given technology level in water sector is identical $A_j^w = A^w$, the output per capita in water sector is as follows:

$$Y^{w} = my_{j}^{w} = mA_{j}^{w} \left(k_{j}^{w}\right)^{a} \left(w_{j}\right)^{b} = mA^{w} \left(\frac{K^{w}}{m}\right)^{a} \left(\frac{W}{m}\right)^{b},$$

$$= m^{1-a-b}A^{w} \left(K^{w}\right)^{a} \left(W\right)^{b} = m^{1-a-b}A^{w} \left(\frac{a}{b}\frac{r}{\theta}W\right)^{a} \left(W\right)^{b}, \qquad (12)$$

$$= m^{1-a-b}A^{w} \left(\frac{a}{b}\frac{r}{\theta}\right)^{a} \left(W\right)^{a+b}.$$

3.3 Dynamic General Equilibrium

Now, the optimal electricity and water consumption function to find the dynamic paths for electricity sector and water sector over time are derived. Hence, the problem that this study aims to solve can be expressed as follows:

$$W_m = \int_0^T U_m(c_{m,t})e^{-\theta t} \mathrm{dt}, \qquad (13)$$

where c_m denotes private *numeraire* consumption per capita in metropolitan so as to maximize the social welfare (W_m) in metropolitan. It is assumed that the metropolitan utility function is an iso-elastic utility function with the coefficient of relative risk aversion σ :

$$U_m(c_m) = \frac{c_m^{1-\sigma}}{1-\sigma}.$$
 (14)

Hence, the social welfare maximization problem with utility discount rate θ (assume being identical to interest rate) in the metropolitan is expressed as follows:

$$\max_{c_m} \int_0^T \frac{c_m^{1-\sigma}}{1-\sigma} e^{-\theta t} dt.$$
(15)

There are two constraints that must be satisfied by optimal solution. First, the average stock of electricity s is to be used. Electricity consumption per capita and water consumption per capita are E^u and E^r . Since electricity and water in Metropolitan of China with limited observation period both are considered as renewable resources. Assume the net generating growth of them is the function of their capacity s_c , $G = G(s_c)$. Hence, the average renewable resources stock constraint is expressed as follows:

$$\dot{s} = \mathbf{G}(\mathbf{s}_c) - E - \mathbf{W}. \tag{16}$$

A second constraint on the change of capital stock per capita derives from the accounting identity relating private *numeraire* consumption per capita, output per capita in electricity sector and water sector:

$$\dot{\mathbf{K}} = Y^e + Y^w - c_m = Y - c_m.$$
 (17)

According to **Equations 5**, **12**, and **17**, the following was obtained:

$$\dot{\mathbf{K}} = n^{1-a-b} A^e \left(\frac{a}{b} \frac{p}{\theta}\right)^a (\mathbf{E})^{a+b} + m^{1-a-b} A^w \left(\frac{a}{b} \frac{r}{\theta}\right)^a (\mathbf{W})^{a+b} - \mathbf{c}_m.$$
(18)

Then, the optimal social welfare problem is as follows:

$$\max_{c_m} \int_0^T \frac{c_m^{1-\sigma}}{1-\sigma} e^{-\theta t} dt$$

s.t. $\dot{K} = n^{1-a-b} A^e \left(\frac{a}{b} \frac{P}{\theta}\right)^a (E)^{a+b} + m^{1-a-b} A^w \left(\frac{a}{b} \frac{r}{\theta}\right)^a (W)^{a+b} - c_m$
 $\dot{s} = G(s_c) - E - W.$
 $E + W \ge 0$ and $\int_0^T (E + W) dt \le s$

Set up the current Hamiltonian function:

$$\mathcal{H}_{C} = \frac{c_{m}^{1-\sigma}}{1-\sigma} + \varphi \Big[n^{1-a-b} A^{e} \Big(\frac{a}{b} \frac{p}{\theta} \Big)^{a} (\mathbf{E})^{a+b} + m^{1-a-b} A^{w} \Big(\frac{a}{b} \frac{r}{\theta} \Big)^{a} (\mathbf{W})^{a+b} - c_{m} \Big] + \tau [\mathbf{G}(\mathbf{s}_{c}) - \mathbf{E} - \mathbf{W}].$$
(19)

Resources Consumption and Metropolitan Growth

TABLE 1 | Variable definitions.

Variable	Notation	Definition	Unit
Metropolitan economic performance	Y	Annual GDP in Guangzhou	10,000 Yuan RMB
Electricity consumption Water consumption	E W	Total annual consumption of electricity in Guangzhou Total annual consumption of water in Guangzhou	10,000 kWh (kilowatt hours) 10,000 cu.m

The maximum principle conditions are as follows:

$$\frac{\partial \mathcal{H}_C}{\partial c_m} = c_m^{-\sigma} - \varphi = 0, \qquad (20)$$

$$\frac{\partial \mathcal{H}_C}{\partial \mathbf{E}} = \varphi \left[(a+b)n^{1-a-b}A^e \left(\frac{a}{b}\frac{p}{\theta}\right)^a (\mathbf{E})^{a+b-1} \right] - \tau = 0, \qquad (21)$$

$$\frac{\partial \mathcal{H}_C}{\partial W} = \varphi \left[(a+b)m^{1-a-b}A^w \left(\frac{a}{b}\frac{r}{\theta}\right)^a (W)^{a+b-1} \right] - \tau = 0, \quad (22)$$

$$\frac{\partial \mathcal{H}_C}{\partial s} = -(\dot{\tau} - \theta \tau) = 0, \qquad (23)$$

$$\frac{\partial \mathcal{H}_C}{\partial K} = -\left(\dot{\phi} - \theta\varphi\right) = 0. \tag{24}$$

The solution of the optimal control model above is as follows:

$$\varphi^*(t) = \varphi(0)e^{\theta t},\tag{25}$$

$$\tau^*(t) = \tau(0)e^{\theta t} \tag{26}$$

$$E = \frac{a+b}{\frac{\tau^{*}(t)}{\varphi^{*}(t)}}Y^{e} = \frac{a+b}{\frac{\tau(0)}{\varphi(0)}}Y^{e}$$
(27)

$$W = \frac{a+b}{\frac{\tau^{*}(t)}{\varphi^{*}(t)}} Y^{w} = \frac{a+b}{\frac{\tau(0)}{\varphi(0)}} Y^{w}.$$
 (28)

Actually, let $\beta_e = \frac{a+b}{\tau(0)}$, and then, the electricity consumption function is as follows:

$$E_t = \beta_e Y_t^e. \tag{29}$$

Similarly, the water consumption function is as follows:

$$W_t = \beta_w Y_t^w. \tag{30}$$

Finally, presume that $Y_t = Y_t^e + Y_t^w + v_t$, where v_t represents other factors. Combining **Equations 29** and **30**, $Y_t = \frac{\beta_w}{\beta_e \beta_w} E_t + \frac{\beta_w}{\beta_e \beta_w} W_t + v_t$ is obtained, which is the expression about the nexus among water consumption, electricity consumption, and economic output for long run.

4 DATA AND ECONOMETRIC METHODOLOGY

4.1 Data and Statistical Description

Annual data from 1950 to 2014 was obtained from the "Guangzhou Statistical Yearbook", within the Guangzhou Municipal Statistics Bureau. The empirical research by Stata 15 was conducted.

 Table 1 lists all variables and their definitions used in the empirical analysis.
 Table 2 lists the summary statistics for the sample. The sample data exhibit good variability.

TABLE 2 Descriptive statistics.					
Variable	Е	Y	w		
Mean	1,346,930	20,713,531	65,740.22		
Median	322,323	721,533.0	48,022.00		
Max	7,658,542	1.67E+08	167,314.6		
Min	4,872	25,620.00	1,140		
Std. Deviation	2,094,516	40,189,753	55,943.97		
Skewness	1.77870	2.249,855	0.418,239		
Kurtosis	4.90564	7.171,236	1.698,786		
Observation	65	65	65		

4.2 Econometric Methodologies 4.2.1 Stationarity Tests

Standard Granger causality tests have to be conducted on stationary time series. Following this line, the unit roots of Xt to confirm the stationary properties of each variable were first tested. This is achieved by using the augmented Dickey–Fuller test. The augmented Dickey–Fuller test (Granger, 1969) can be revised and derived from the Perron (1989) test:

$$\Delta Z_{t} = \hat{\mu}^{1} + \hat{\theta}^{1} D U_{t} + \hat{\beta}^{1} t + \hat{d}^{1} D (T_{B})_{t} + \hat{\alpha}^{1} Z_{t-1} + \sum_{i=1}^{k} \hat{c}_{i}^{1} \Delta Z_{t-i} + \hat{\varepsilon}_{t},$$
(31)

$$\Delta Z_{t} = \hat{\mu}^{2} + \hat{\gamma}^{2} D T_{t}^{*} + \hat{\beta}^{2} t + \hat{\alpha}^{2} Z_{t-1} + \sum_{i=1}^{k} \hat{c}_{i}^{2} \Delta Z_{t-i} + \hat{\varepsilon}_{t}, \qquad (32)$$

$$\Delta Z_{t} = \hat{\mu}^{3} + \hat{\theta}^{3} D U_{t} + \hat{\beta}^{3} t + \hat{\gamma}^{3} D T_{t}^{*} + \hat{d}^{3} D (T_{B})_{t} + \hat{\alpha}^{3} Z_{t-1} + \sum_{i=1}^{k} \hat{c}_{i}^{3} \Delta Z_{t-i} + \hat{\varepsilon}_{t},$$
(33)

where $Z_t = (Y_t, E_t, W_t)^T$.

Thus, the choice of the break point is correlated with the data in hand and the choice of break point cannot be considered as independent of the data. Zivot and Andrews (1992) addressed this issue by estimating the structural break data endogenously instead of considering an exogenous break date. The following equations for the Zivot and Andrews (ZA) test are estimated, when $Z_t = (Y_t, E_t, W_t)^T$:

$$\Delta Z_t = \hat{\mu}^4 + \hat{\theta}^4 D U_t \left(\hat{\lambda} \right) + \hat{\beta}^4 t + \hat{\alpha}^4 Z_{t-1} + \sum_{i=1}^k \hat{c}_i^4 \Delta Z_{t-i} + \hat{\varepsilon}_t, \quad (34)$$

$$\Delta Z_{t} = \hat{\mu}^{5} + \hat{\gamma}^{5} DT_{t}^{*}(\hat{\lambda}) + \hat{\beta}^{5} t + \hat{\alpha}^{5} Z_{t-1} + \sum_{i=1}^{k} \hat{c}_{i}^{5} \Delta Z_{t-i} + \hat{\varepsilon}_{t}, \quad (35)$$

$$\Delta Z_{t} = \hat{\mu}^{6} + \hat{\theta}^{6} D U_{t} (\hat{\lambda}) + \hat{\beta}^{6} t + \hat{\gamma}^{6} D T_{t}^{*} (\hat{\lambda}) + \hat{\alpha}^{6} Z_{t-1} + \sum_{i=1}^{k} \hat{c}_{i}^{6} \Delta Z_{t-i} + \hat{\varepsilon}_{t}.$$
(36)

4.2.2 Cointegration Analyses

Following the application of cointegration approach in electricity consumption and GDP by Shiu and Lam (2004), and Lai et al. (2011), the Johansen cointegration test (Johansen, 1991) was used:

$$\Delta Z_t = \mu + \Pi Z_{t-1} + \sum_{i=1}^k \Gamma_i \Delta Z_{t-i} + \varepsilon_t.$$
(37)

In addition, Tang et al. (2013) also used the ARDL model to test the cointegration relationship. Hence, the ARDL model in this study can be expressed as follows:

$$D(Y_{t}) = a_{01} + b_{11}Y_{t-1} + b_{21}E_{t-1} + b_{31}W_{t-1} + \sum_{i=1}^{p} a_{1i}^{1}D(Y_{t-i})$$

+
$$\sum_{i=1}^{q} a_{2i}^{1}D(E_{t-i}) + \sum_{i=1}^{q} a_{3i}^{1}D(W_{t-i}) + \varepsilon_{1t},$$
(38)

$$D(E_t) = a_{02} + b_{12}Y_{t-1} + b_{22}E_{t-1} + b_{32}W_{t-1} + \sum_{i=1}^p a_{1i}^2 D(E_{t-i}) + \sum_{i=1}^q a_{2i}^2 D(Y_{t-i}) + \sum_{i=1}^q a_{3i}^2 D(W_{t-i}) + \varepsilon_{2t},$$
(39)

$$D(W_{t}) = a_{3j} + b_{13}Y_{t-1} + b_{23}E_{t-1} + b_{33}W_{t-1} + \sum_{i=1}^{p} a_{1i}^{3}D(W_{t-i}) + \sum_{i=1}^{q} a_{2i}^{3}D(Y_{t-i}) + \sum_{i=1}^{q} a_{3i}^{3}D(E_{t-i}) + \varepsilon_{3t},$$
(40)

where D () represents the change of Z_t , $Z_t = (Y_t, E_t, W_t, K_t, L_t)^T$.

4.2.3 Kernel-Based Regularized Least Squares

After cointegration, the KRLS method to estimate the cointegration relationship among metropolitan economic growth, electricity consumption, and water consumption was used:

$$k(\mathbf{X}, X_t) = e^{-\frac{||\mathbf{X} - X_t||^2}{\sigma^2}},$$
(41)

where $X_t = W_t$, or E_t . The target function Y = f(X) is as follows:

$$Y = f(X) = \sum_{t=1}^{T} c_{t}k(X, X_{t}) = \begin{bmatrix} k(X_{1}, X_{1}) & \cdots & k(X_{1}, X_{T}) \\ \vdots & \ddots & \vdots \\ k(X_{T}, X_{1}) & \cdots & k(X_{T}, X_{T}) \end{bmatrix} \begin{bmatrix} c_{1} \\ \vdots \\ c_{T} \end{bmatrix} = KC,$$
(42)

where c_i is a weight for each covariate vector. Hence, the solution is follows: $C^* = argmin \sum_{t} V(Y_t, f(X_t)) + \pi R(f(X_t)) = argmin(Y - KC)^T (Y - KC) + \pi C^T KC.$

5 EMPIRICAL RESULTS

The data of electricity consumption, water consumption, and GDP may have structural change during 1950–2014. Thus, the stability with break point to complement to the ADF approach that does not consider the data structural change has to be

examined. **Table 3** shows the unit root test results from the ADF test. **Table 4** shows the unit root test results from the Perron test, and **Table 5** shows the unit root test results from the Zivot-Andrews test. All of them confirm that Y, E, and W are integrated at I (1).

In terms of cointegration tests, the Jonansen test results in **Table 6** and the ARDL bounds test results in **Table 7** both show that the economic performance, water, and electricity consumption have long-run equilibrium relationship.

5.1 Unit Root Tests

5.1.1 ADF Test and PP Test

The ADF test is applied to detect the possible presence of unit roots in Y_t , W_t , and E_t . The null hypothesis of unit root can be rejected in favor of the alternative hypothesis of no unit root when the *p*-value large. **Table 4** represents that no variable is stationary in their levels. On the other hand, Y_t , W_t , and E_t are the stationary process in their first.

5.1.2 Perron Modified ADF Test With Exogenous Break Point

The framework that follows the work of Perron (1989), Perron and Vogelsang (1992), Vogelsang and Perron (1998), and Banerjee et al. (1992) supports the computation of the modified Dickey–Fuller tests that allow for levels and trends that differ across a single break date. The results of the Perron's modified ADF test in **Table 5** and the Zivot–Andrews test are detailed in **Table 6**, which show that non-stationary process is found in all series at level with intercept and trend, but variables are found to be stationary at first difference. This confirms that Y_{tb} W_{tb} and E_{t} are integrated at I (1).

5.2 Cointegration Tests

According to the unit root test results, the integration of the variables is of the same order, and testing whether these variables are cointegrated over the sample period was continued.

5.2.1 Johansen Cointergration Test

Table 7 shows the results of the Johansen test. Because the trace statistic of non-cointegrating equation and at most one cointegrating equation are greater than the 5% critical values, respectively, the test rejects the hypothesis of no cointegration and indicates that there is one cointegrating equation at the 5% significance level, so there is a long-run relationship between Y_t , W_t , and E_t for Guangzhou.

5.2.2 ARDL Bounds Test Approach to Cointegration

Pesaran et al. (2001) critical values are based on the assumption that the variables are integrated of order I (0) or I (1). Unit root tests insure that none of the series is integrated of I (2) or higher. Armed with information about stationarity, the ARDL bounds testing approach to cointegration was applied. The results of the bound test are given in **Table 8**. From these results, it is clear that there is a long-run relationship between Y_t , W_t , and E_t , because their F-statistic are higher than the upper-bound critical value at the 1% level. This implies that the null

TABLE 3 | ADF unit root test results.

Levels			First differences				
Variables	ADF test statistics	Lag length	Critical values	Variables	ADF test statistics	Lag length	Critical values
Y	-1.613,673	10	-2.916,566*	ΔΥ	1.721,083	9	-1.612,934*
E	2.853,394	1	-3.482,763***	ΔE	2.822,923 6	15	-2.614,029***
W	1.891,674	0	-2.907,660***	ΔW	-6.004387	2	-3.538,362***

Note: * shows significance at 10% level; ** shows significance at 5% level; and *** shows significance at 1% level.

TABLE 4 | Perron's modified ADF unit root test results.

Break Type: Innovative outlier				Break Type	r		
Variables	T-statistic	Break data	5% critical values	Variables	T-statistic	Break data	5% critical values
Y	-4.283,895	2006	-4.443,649	ΔΥ	-5.743,232**	2003	-4.443,649
E	-0.960,870	1984	-4.443,649	ΔE	-7.42739 ***	2001	-4.949,133
W	-0.960,871	1984	-4.443,649	ΔW	-7.427,391**	2001	-4.193,627

Note: 1) ** indicates significance at 5% level; 2) innovative outlier affecting every member of a set of autoregressive time series at the same time point are represented as independent random effects.

TABLE 5 | Zivot-Andrews structural break trended unit root test results.

Variables At level		Variables	At first difference				
	T-statistic	Break data	5% critical values		T-statistic	Break data	5% critical values
Y	-2.293,951	1999	-4.93	ΔΥ	-5.264,743**	1998	-5.08
E	-0.944,716	2003	-4.93	ΔΕ	-4.352,876*	1988	-4.11
W	-3.251,608	2001	-4.93	ΔW	-5.805,366**	1995	-4.93

Note: **indicates significance at 5% level.

TABLE 6 | Johansen cointegration test results.

Hypothesized number of cointegrating equation	Maximum eigenvalues statistics	Trace statistic	5% Critical value
None**	100.3432	151.1318	69.81889
At most 1**	78.4543	77.37808	47.85613
At most 2	34.1234	35.69042	29.79707
At most 3	14.8879	15.78851	15.49471
At most 4	1.2333	1.516,098	3.841,466

Note: **indicates significance at 5% level.

TABLE 7 | Bounds test results.

Estimated model	Lag length	F-statistic
f (Y/E,W)	(4,4,3,4,3)	9.976,565**
f (E/W,Y)	(4,3,4,4,4)	11.85424**
f (W/E,Y)	(2,1,3,2,3)	8.490,854**
5% critical values	I (O)	l (1)
	3.29	4.37

Note: ***indicates significance at 5% level. The estimated models just show which is the dependent variable.

TABLE 8 Results of KRLS (dependent variable = Y_t).				
W _t	130.899 ***(33.0933)			
Et	10.1467***(0.801,614)			
R-squared	0.9335			
Number of observations	67			
Lambda	0.07082			
Tolerance	0.067			
Sigma	2			
Looloss	52,180			

Notes: 1)*** indicates that the variable is significant at the 1% level, ** indicates significant at the 5% level, and * indicates significant at the 10% level; 2) robust standard errors in parentheses.

hypothesis of no cointegration between $Y_t, \ W_t, \ and \ E_t$ is rejected.

5.3 Further Discussion: Results From Kernel-Based Regularized Least Squares

At the end, to estimate the relation in (2.31) above, KRLS to tackle regression problem without a specification search was utilized. The results from **Table 8** suggest a statistically significant relationship between Y, E, and W. The KRLS results also suggest that the R^2 from KRLS is high enough, which reveals that this regression model fit results. The empirical results indicate that the metropolitan economic growth is positively correlated with the electricity consumption and the water consumption is positively associated with the metropolitan economic performance.

The empirical results are consistent with other studies in developing countries (Ghosh, 2002; Jumbe, 2004; Mehrara, 2007; Narayan and Prasad, 2008; Jamil and Ahmad, 2010; Shahbaz et al., 2011). However, the empirical results in this study contradict with the empirical results of He and Huang (2020). This result is also helpful for metropolitan policymaking in balancing the relationship between electricity use, water consumption, and economic growth. In particular, the experience of economic growth in Guangzhou provides evidence that the resources consumption generates more GDP.

Indeed, both ARDL and Johansen cointegration techniques electricity consumption, urbanization, show that and metropolitan economic performance are cointegrated (He, 2020). Hence, the empirical results regarding to the relationship between electricity consumption and metropolitan economic growth are consistent with his findings, but our work considers the interaction effect of water consumption. Actually, the water consumption is determined by the intersection of endogenous growth function and water consumption function, neither function can be consistently identified by comparing average quantities of water consumed at different values of observed real per capita output (He and Fullerton, 2020). Obviously, our results confirm their arguments and provide new evidence for the long-run relationship between water consumption, electricity consumption, and endogenous economic growth.

6 CONCLUSION AND POLICY RECOMMENDATION

The electricity consumption and water consumption are vitally important for an economic developing region. This study builds a dynamic general equilibrium model that connects electricity consumption and water consumption with metropolitan economic performance. It also conducts econometric research on the interconnection among them for the third largest metropolitan in China (Guangzhou) from 1950 to 2014. The cointegration techniques show that electricity consumption, water consumption, and the metropolitan economic performance have long-run equilibrium relationship. The results of KRLS reveal that the metropolitan economic growth is positively correlated with electricity consumption. Also consistent with the theory, the water consumption is positively associated with the metropolitan economic performance. These results are generally stable and hold with alternative measures of unit roots, with alternative estimation strategies, and with or without controlling for trends, intercepts, and break points.

Thus, the policy implication of that is, if to improve the electricity and water consumption efficiency, then the resources consumption pattern should be changed into a pattern of low resource intensity. The policy decision-makers should 1) set up R&D funds to drive up distributed energy resources technology, 2) implement subsidy policies to develop resource-saving business models, 3) design an efficient system of prices and regulate charges for electricity services, and 4) enlarge the scale of investment on water utilities and development of water plants. Furthermore, on the basis of the analytical findings of this study, it is proposed that policy-makers and officials continue to enhance their interventions aimed at promoting successful economic development, water use, and increasing electricity consumption. In China, energy, water, and economic developments have had a severe effect. China has faced both the beneficial and detrimental impacts of economic development, as have many other developed countries.

However, this study has limitations that include the following: 1) the theoretical model assumes utility optimization and dynamic general equilibrium; 2) the specification of econometric model does not consider macroeconomic factors such as interest rates; and 3) important uncertainties referring to the export and import between districts, electricity market structures, and decarbonizing technology are not included in the empirical models. It is supported that a more systematic investigation and measurement of joint policy of energy and environment and how they interact with economic growth factors are promising areas for future research (Henryk and Łukasz, 2012; Han and Kung, 2015; Alola et al., 2021; Gyamfi et al., 2021; Dagar, 2021; Gyamfi et al., 2021).

DATA AVAILABILITY STATEMENT

The original contributions presented in the study are included in the article/Supplementary Material; further inquiries can be directed to the corresponding authors.

AUTHOR CONTRIBUTIONS

YH collected and analyzed data, and built theoretical models. SG wrote and revised the manuscript.

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