A Graph Theory-Based Method for Regional Integrated Energy Network Planning: A Case Study of a China–U.S. Low-Carbon Demonstration City

Yingying Chen and Jian Zhu *

School of Civil Engineering, Hefei University of Technology, Hefei 230009, Anhui, China; chenyingying_hfut@163.com
* Correspondence: zhujian@hfut.edu.cn; Tel.: +86-1395-517-6083

Received: 30 October 2019; Accepted: 21 November 2019; Published: 25 November 2019

Abstract: With the significant attention on global climate change, regional integrated energy systems (RIES) in low-carbon city planning has become one of the most important ways for global cities to achieve the goal of energy conservation and emission reduction. However, the planning strategy used in the primary stage of RIES establishment will greatly affect the system economy and environment. In view of the lack of planning guidance mechanisms for the large-scale RIES in China, a method for RIES preliminary overall planning, focused on energy types and use, is proposed in this paper. A graph theory-based mathematical optimal model was established with the lifetime costs of the whole system as the economic goal, and an improved Prim algorithm was put forward to solve the costs of the transmission and distribution network with the dynamic weight set of pipeline flow. The model was solved by an algorithm based on the idea of a dynamic minimum spanning tree and optimal path planning. The model and method were applied in a China–U.S. low-carbon demonstration city to verify feasibility and validity. The results could help us to comprehensively integrate regional energy and accurately plan future cities.

Keywords: regional integrated energy system; graph theory; Prim algorithm; transmission and distribution network; network layout

1. Introduction

1.1. Background of Environment and National Policy

Global climate change has become one of the most serious and complex issues worldwide. With the development of human society and city construction all over the world, energy consumption creates large quantities of human-produced greenhouse gas, contributing to global warming. Global primary energy consumption, especially natural gas and renewable energies, has increased rapidly in recent years. Meanwhile, the growth rate of carbon emissions has peaked in the last seven years [1]. The average temperature of the global land and ocean surface in 2018 was the fourth warmest since global records began in the mid-to-late 1800s [2]. Cities are estimated to contribute 76% to energy-related carbon emissions by 2030 according to the international energy agency (IEA) [3]. British Petroleum (BP) estimates that China accounts for 23.6% of global energy consumption and remains the world’s largest energy consumer [1]. Two general characteristics of China’s energy consumption have been summarized; the first is that it is densely packed with cities. China’s city energy consumption accounts for 85% of the total, 18 percentage points higher than the world average; the other is the high level of spatial agglomeration, as city energy consumption is concentrated in the developed eastern areas and several regional central cities [4]. Cities have their center vital point positions in the field of...
industry, construction, and transportation, and they have a significant role in the economic structure and strategic distribution of a country. With the rapid process of China’s rural urbanization, a great amount of the mobile population is flooding into the cities. China’s state council speculates that by 2020 the urbanization rate will reach 60%, and the overall city population will reach 810–840 million, making the cities spaces of rapid tempo, high density, and high demand [5]. Therefore, applying energy conservation and emission reduction to city planning is essential in order to realize low carbon and sustainable development.

The concept of “low carbon”, first appearing in the British energy white paper in 2003, has been widely extended to global city construction, and has been a recognized pattern of sustainable development. The low-carbon eco-city has been put forward as the way to achieve the low-carbon goal of energy structure and resident consumption, so that the city and nature can coexist harmoniously. Since 1988, the United Nations has paid close attention to the issue of global warming and emphasized urgent climate action. The intergovernmental panel on climate change (IPCC) was established to understand global climate change and its impacts [6]. To meet the climate challenge, UN plans to host the 2019 climate action summit and accelerate actions to implement the Paris agreement on climate change [7]. The European commission proposed to reduce carbon emissions by at least 40% compared to 1990 levels by 2030 as a measurable target in the Europe 2030 strategy [8]. China started to consider low-carbon cities as a new high standard of the urban brand as early as 2008 and launched the “China low-carbon city development project”, with eight cities as the pilot projects [9]. China’s 13th five-year plan emphasizes low-carbon pilot projects, including expanding the pilot projects to 100 low-carbon cities nationwide [10].

1.2. Regional Integrated Energy System in City Planning

City planning is the science of the rational organization and equitable allocation of city resources, combined with analysis and design [11]. Since the theories of neo-urbanism and smart growth were developed, cities have tended to be compact and intensive, with a high density of population and construction. Traditional energy use in buildings has the character of equipment decentralization and multipoint emissions. In this way, no energy is integrated, and it is difficult to realize centralization of management. Regional renewable energies have the characteristics of low energy density and discontinuity of capacity. A regional integrated energy system (RIES) has been suggested to comprehensively plan every kind of available energy in the design area [12,13]. A joint energy station is established to supply demand users with cold and heat through a transmission and distribution network. A RIES mainly consists of energy stations, transmission and distribution networks, and users. Unlike traditional distributed energy systems, a RIES distributes cold and heat mainly by using a flowing medium through a closed and insulated network [14], making better use of regional energies and resulting in less consumption.

RIES have become one of the important ways for global low-carbon eco-cities to achieve the goal of energy conservation and emission reduction. RIES include regional heating [14,15], regional cooling [16], regional cooling and heating [17,18], regional combined heat and power, regional combined cooling heating and power [19–21], and so on. At present, the research on the technology of RIES has been considerable. Most of these studies concentrate on specific technologies like the optimization of multi-energy storage [12,18], heat pumps with a solar assisted ground source [19,22,23], low temperature district heating networks [24,25], integration of renewable energy [26], and so on. The global spread of this kind of energy use is also considerable. RIES has developed rapidly in China in the past 60 years. China has a long history of heating. Heating in China’s northern part began in the 1950s [27], and regional heating accounted for about 2/3 of the heating area in the north in 2015 [28]. The total building area covered by heating pipeline networks in 2017 was estimated to be more than twice the area in 2009 [29]. Although regional cooling is far less common than heating in China, it is constructed widely in south China, such as Guangzhou, Shanghai, Chongqing, and so on. The regional cooling system in Guangzhou, with the service area of 6 million m², has become one of the world’s
largest systems [30]. The current situation of China’s rapid urbanization and the requirements of low-carbon cities planning are bound to increase the demand for RIES.

1.3. Graph Theory Application to Regional Energy Systems

Graph theory focuses on a graph composed of a number of given vertexes and the edges connecting two vertexes as a research object, where vertexes represent certain things and edges represent the specific relationship between these corresponding things. The application of graph theory to RIES has increased in recent years, especially for the number of energy stations and site selection, and the transmission and distribution network optimized layout. The transmission and distribution network system is an important part of RIES. Its initial costs and operation costs account for a large proportion of the total, consuming a lot of electricity every year [31]. Therefore, whether the transmission and distribution network layout is reasonable or not will greatly affect the initial investment and annual operating costs of RIES. There is a lot of research on RIES network layout. In China, the concept of the “Energy Internet” was earliest put forward by Liang H. et al. in the national HVAC academic conference in 2010, and the graph analysis method was put forward to solve the RIES energy station site selection and network layout problems, wherein the mixed integer linear programming optimization model was set up with the optimization target of lowest energy consumption, costs, and carbon emissions [32]. Chen J. et al. put forward the concept of energy distance and established a P-median model of ‘stations–pipelines’ layout, with the minimum energy distance as the economic goal [33]. Yi W.F. et al. also established a P-median based model for energy station location and the transmission and distribution network, and it was solved by an algorithm based on the enumeration method [34]. Wang Z.Q. proposed a network location algorithm based on the idea of the minimum spanning tree. The optimal solution was obtained on the basis of all feasible solutions in the ergodic solution set space [35].

The solution when addressing the network layout is to achieve the optimal goal for the energy station number and location, and the path of the energy transmission system. This problem is a P-median problem. The purpose of the P-median problem of the energy system is to find the optimal P stations from the set of alternative energy stations, making the sum product of the distances and the loads. The solutions mainly include the enumeration method, greedy algorithm, mathematical programming method, and so forth [36]. However, most of these algorithms rarely consider the optimal layout of the transmission and distribution network. Therefore, most studies have adopted the shortest path and minimum spanning tree theory to solve the optimization problem [33].

In general, applying RIES to low-carbon cities has become a recognized choice, and there are many large-scale systems in China. However, the planning guidance mechanism of RIES in China is always lacking and few literatures have proposed overall planning ideas for the large-scale RIES. In addition, an effective, economic, and reasonable planning is of great significance to establish RIES. The difference in economy and efficiency between different design planning can vary greatly. Moreover, there are very few algorithms that really fit the design idea of HVAC pipelines, and the study on complex network layout problems is less involved.

The aim of this paper is to provide a theoretical method for the primary stage of a RIES planning. First, the regional resource situation analysis and the forecast of cooling and heating load on the demand side provided date and background support to establish the structural model of RIES. Second, the mature graph theory was applied to the structural model of RIES. Because most of the algorithms used rarely accorded with the design idea of the transmission and distribution network layout, the improved Prim algorithm with the dynamic weight set of pipeline flow was put forward to solve the costs of the transmission and distribution network. Third, the mathematical optimal model of RIES was established, with the lifetime costs of the whole system as the economic goal. The model was solved by the idea of the dynamic minimum spanning tree and optimal path planning. Fourth, the theoretical method for the primary stage of a RIES planning was applied in a China–U.S. low-carbon demonstration city, and the mathematical optimal model and the improved Prim algorithm were
verified to provide effective support for primary planning of RIES and to comprehensively integrate regional energy and accurately plan cities.

2. Method for Regional Integrated Energy System Planning

2.1. Regional Energies Situation Analysis

The research on RIES has been focused on its energy types and use. Before designing a RIES, it is necessary to analyze and evaluate the existing energies. This determines the form of energy available to the system and the equipment selection, as well as the efficiency and energy saving potential of the whole system. However, different energy types and utilization forms have advantages and disadvantages. This relates to national energy policy, regional resource conditions, economic type, urban design characteristics, building formats, and energy system management modes. Yan J.L.; et al. evaluated RIES from three aspects: energy consumption, environmental influence, and economy, and determined the weight of the indexes at different levels [37]. And the evaluation indexes are shown in Figure 1.

Figure 1. Evaluation indexes of the regional energy system. Reprint with permission [37]; 2019, Thermal Power Generation.

A city is generally equipped with basic energy facilities and natural renewable energies. Facilities include the city heat supply, electric power resources, and natural gas pipeline network, and renewable energies include solar power, renewable water resources, shallow geothermal energy, and so on. Combined heat and power are energy uses encouraged by national policy, which is more energy-saving and environmentally friendly than boiler heating. Natural gas is a clean energy and one of the key energy forms in the construction of low-carbon cities with significant emission reductions. It can realize energy cascade utilization and reasonable distribution of energy according to the different energy grades and energy demands. Unless the technology and economy are unreasonable, the application of cogeneration heating and natural gas cascade utilization should be considered as far as possible, with related national policies. Public subsidies should be added to the life cycle costs of the energy system. The hourly distribution of the heating and cooling load means a shortage of electricity during peak hours. However, energy–storage air conditioning can realize peak load shaving for power demand, which reflects the concept of power demand side management. Therefore, it is necessary to consider ice storage and other energy storage applications. Renewable water sources and shallow geothermal energy are excellent energies due to their stability.
All these energy types and their forms of utilization, recommended for RIES, can be roughly divided into four levels according to the above evaluation indexes. As shown in the table below (Table 1), a higher level means a lower recommendation.

| Level | Energy                          | Utilization                                      |
|-------|---------------------------------|--------------------------------------------------|
| I     | Renewable water resources       | Cooling in summer; Heating in winter              |
|       | Shallow geothermal energy       | Cooling in summer; Heating in winter              |
|       | Solar energy                    | Heating in winter; Electricity generation         |
|       | Municipal heat                  | Heating in winter                                |
| II    | Natural gas                     | Combined cooling, heat and power                 |
| III   | Combined municipal heat and electricity | Cooling by electricity and ice storage in summer; Heating by municipal heat in winter |
| IV    | Combined natural gas and electricity | Cooling by electricity and ice storage in summer; Heating by natural gas in winter |

2.2. Forecast of Cooling and Heating Load on the Demand Side

Forecasting and analysis of cooling and heating load on the demand side provides an important design basis for the selection of cooling and heating primary equipment capacity in energy stations. There exists the unit area index method, software simulation prediction method, and data statistical analysis method for load forecasting [38]. The algorithm used by the unit area index method to calculate the total load of regional buildings simply accumulates the loads of individual buildings in the region, which does not reflect the time variation characteristics of the load. The software simulation prediction method’s workload is exponentially increased compared with the unit area index method. The data statistical analysis method is based on a large number of historical data, to judge the regional building load trend by using mathematical statistics or other methods. Therefore, combining the unit area index method with the software simulation prediction method to calculate the regional building load based on the heat transfer coefficient method is preferred.

The hourly load of the block on the energy demand side of the region is:

\[ Q_{DZS}^j = \sum_{m \in M} \rho_m q_x S_m \]  

\( Q_{DZS}^j \) denotes the hourly load of block \( j \) (kW); \( M = \{1, 2, \ldots, |M|\} \) denotes the code set of individual buildings in block \( j \); \( \rho_m, S_m \) respectively denote the plot ratio and area of building \( m \) (m²); \( q_x \) denotes the design of daily cooling or heating load indexes of specific types of building (kW/m²) from the simulation software; and \( x \) denotes the building type, with \( x = 1 \) for residential building, \( x = 2 \) for department store, \( x = 3 \) for office building, \( x = 4 \) for hotel building, and \( x = 5 \) for school building.

The design load of the block is:

\[ Q_j^D = \max(Q_{DZS}^j)f \]  

\( Q_j^D \) denotes the maximum of the hourly load (kW); and \( f \) denotes the coincidence factor, in consideration of different simultaneous uses in different building.

The total capacity of the energy station is:

\[ Q_{ES}^k = \sum_{j \in V} Q_j^D Z_{jk}, \]  

\( Z_{jk} \) is a 0–1 variable, denoting the distribution relationship of blocks and stations; \( Z_{jk} = 1 \) for station \( k \) provides the energy for block \( j \), or \( Z_{jk} = 0 \).
2.3. Graph Theory-Based Structural Model

The regional integrated energy network could be described as an undirected graph with weight, $G = (V, E, Q, D)$, with vertex set $V$, edge set $E$, vertex weight set $Q$, and edge weight set $D$. The diagrammatic structure graph of RIES is shown in Figure 2a.

![Diagram](image)

**Figure 2.** Structure graph. (a) City network layout; (b) Tree-shaped pipeline network layout.

$V = \{V_a | a = 0, 1, 2, 3\}$, $V$ can be divided into four subsets, $V_0 = \{v_i | i = 1, 2, \ldots, |V_0|\}$ denotes the energy sources set, represented by triangles in the graph, such as water resources, ground sources, and where the energy station absorbs energy. The position of the energy source has a great influence on the selection of energy stations. They would be closer with better energy efficiency and economy.

$V_1 = \{v_k | k = 1, 2, \ldots, |V_1|\}$ denotes the set of energy stations, represented by small squares in the graph. $V_2 = \{v_n | n = 1, 2, \ldots, |V_2|\}$ denotes traffic intersection sets, represented by small dots on every intersection of two lines. $V_3 = \{v_j | j = 1, 2, \ldots, |V_3|\}$ denotes blocks to be served in the design region set, represented by big dots. $E = \{e(v, v') | v, v' \in V\}$ denotes city road sets, along which the pipeline should lay according to specification. $e$ is a 0–1 variable, reflecting the adjacency relationship between two vertexes. $Q = \{q(v) | v \in V_3\}$ denotes the load set of vertexes $V_3$. $D = \{c(e) | e \in E\}$ denotes the weight set of edges, representing the costs of the pipeline.

Spanning tree, which is characterized by no-loop and ergodicity, is a very important concept in graph theory. The pipeline networks of the system can be described as the spanning tree $T = (X, L, W)$, $T \subset G$, $T \neq \emptyset$. The diagrammatic structure graph of the tree-shaped pipeline network is shown in Figure 2b. Tree $T$ is the nonvoid proper subset of graph $G$, $X = \{X_a | a = 0, 1, 2, 3\}$, $X \subset V$, $X_{V_a} = X \cap V_a$, denotes the vertex set of the tree, the constraint condition is $|X_{V_1}| = 1$, tree $T$ contains only one energy station $v_{V_1}$. $L = \{|l| = 1, 2, \ldots, |L|\}$ denotes the pipeline set. In this model, water supply and return pipes are simplified to a single line $l$. $W = \{w(l) | l \in L\}$ denotes the weight set of edges. $w(l) = |l| \times Q^{0.5}$ denotes loads along the pipes, $Q_l = \sum_{i=1}^{l} Q_i$ is the accumulated load of all downstream points to be served. $W$ changes with the connected relationship of the pipeline network layout, so it is the dynamic weight set of the edges of the tree.

2.4. The Improved Prim Algorithm

The Prim algorithm is one of the most common greedy algorithms for a weighted undirected graph to find its minimum spanning tree, including every vertex in it, where the total weight of the edges in the tree is minimized. The basic idea of this algorithm is that it operates from an arbitrary starting...
vertex, one vertex at a time, and at each step it adds the cheapest possible connection, until vertices are added to the spanning tree. The transmission and distribution network is characterized by the network topology, and the best layout can be calculated [32–35,39,40]. However, the calculations of most of the studies directly use the Prim or Kruskal algorithms with a distance weight set, and not enough details. This paper puts forward an improved Prim algorithm in accordance with the design idea of cold and heat pipelines of HVAC. It operates from the vertex representing the energy station. The weight set represents the costs of the edges, and the weight of every added edge is updated at every step. For example, the small blockage denotes an energy station for hot water supply, and the dots denote several load points. The best pipeline network layout is calculated by using the Prim and improved Prim algorithm, respectively. The calculation result of contrast is shown in Figure 3. The layout (a) of pipelines could be more uniform and stable by using the improved algorithm. The total pipeline length of (b) could be shorter, but the most unfavorable pipe section is much longer than in (a), which means the on-way resistance is obviously and unreasonably large.

![Figure 3. Results comparison. (a) Result with the improved Prim algorithm; (b) Result with the Prim algorithm.](image)

3. Establishment of the Mathematical Model

3.1. Objective Function

On the basis of the graph $G$ of the regional energy system, the set $T^k$, including all lode vertexes of $V_3$, is found to minimize the lifetime costs of this energy system in the region. The model is called the best economic model, which can be described to find the set $T^k = \{T^k_i| i = 1, 2, \ldots, k\}$, making the object function true.

$$f(T^k) = \min \sum_{i=1}^{k} LCC(T^k_i), \quad T^k_i \subset P(T)$$ (4)

The objective function is a discrete optimal control problem. In this formula, $P(T)$ denotes the collection of all energy pipeline network layout diagrams $T$; $T^k$ denotes the number of its elements is $k$, for the number of energy stations; and $T^k_i$ is the $i$th element of $T^k$.

The lifetime costs of this energy system in the region are:

$$LCC(T^k_i) = C_{in} + \frac{(1 + i_0)^m - 1}{i_0(1 + i_0)^m} C_{op}$$ (5)

$C_{in}$ denotes the initial investment of the system; $C_{op}$ denotes the annual operating costs; $i_0$ denotes the benchmark discount rate; and $m$ denotes the life cycle.
The initial investment is:

$$C_{in} = C_{in,fw} + C_{in,sh} + C_{in,gw}$$  (6)

The initial investment of the regional energy system mainly contains the costs of the energy station room, equipment unit, and the pipeline network. The civil construction costs of the energy station room are:

$$C_{in,fw} = p_{fw} \times S_k, \quad S_k = 2605 + 0.1 Q^{ES}_k$$  (7)

This paper combines engineering examples, fitting the linear regression equation for load $Q^{ES}_k$ and the station room costs $C_{in,fw}$ $p_{fw}$ denotes the unit price of civil construction (RMB/m$^2$). $S_k$ denotes the room area of the energy station $m^2$; $Q^{ES}_k$ denotes the design capacity of the equipment unit (kW).

The costs of equipment are:

$$C_{in,sh} = \sum_{n \in N} q_{k,n} \times Q^{ES}_{k,n}, N = \{1, 2, \ldots, n\}$$  (8)

$q_{k,n}$ denotes the $n$th equipment unit of the $k$th energy station (RMB/kW); and $Q^{ES}_{k,n}$ denotes the design capacity of the equipment unit (kW).

The costs of the cooling and heating pipeline network are:

$$C_{in,gw} = \sum_{l \in L} C(d_l) \times l_i$$  (9)

$d_l$ denotes the pipe diameter of line $l$ (m); $C(d)$ denotes the unit costs of the pipes (RMB/m), fitted to the linear regression equation of pipe diameter according to market price, $C(d) = 4147d - 80$. The relationship between pipe diameter and the load carried under ideal conditions, according to heat transfer theory, is:

$$d_l = \sqrt{\frac{4Q_l}{3.14cv\Delta T}}$$  (10)

$Q_l$ denotes the load of pipeline $l$ (kW); $c$ denotes the specific heat of the fluid in pipe (kJ/kg·ºC); $v$ denotes the flow velocity (m/s); $\rho$ denotes the fluid density (kg/m$^3$); $\Delta T$ denotes the temperature difference between supply and return of the fluid. These two formulas give the costs of pipeline changes with the layout of the network.

The annual operating costs are:

$$C_{op} = C_{op,wh} + C_{op,sh} + C_{op,pump}$$  (11)

The annual operating costs of the energy system generally include management costs, energy consumption costs, and operating costs of the transmission and distribution network. The annual management costs are:

$$C_{op,wh} = gQ^{ES}$$  (12)

$g$ denotes the annual management costs index (RMB/kW); the management costs can be estimated by loads.

The energy consumption costs are:

$$C_{op,sh} = \sum_{i=1}^{n} Q_i P_i Z_{0-1}$$  (13)

$Q_i$ denotes the usage amount of energy; $P_i$ denotes the unit price of $i$th energy; $Z_{0-1}$ is a 0–1 variable, and $Z_{0-1} = 1$ for the energy station uses the energy, or $Z_{0-1} = 0$. 

The costs of the transmission and distribution network are:

\[ C_{op,pump} = c_c \sum_{r=0.2}^{1} \frac{Q_{ES}}{COP_{ES}} r \tau_r \]  

(COP\textsubscript{ES} denotes the performance coefficient of the system; \(c_c\) denotes electricity price (RMB/kWh); \(r\) denotes the loading rate of the pumps; \(\tau_r\) denotes the unit operation time at a certain loading rate.

3.2. Constraints

The constraints limit the maximum distance of the transmission and distribution network and the correspondence between the energy stations and the load blocks. All load blocks in set \(V_3\) should be divided into \(k\) parts and one energy station serves one part. The network layout \(T_i^k\) denotes a subsystem of the network layout; \(X_i,V_3\) denotes the set of all load blocks in \(T_i^k\); \(X_i,V_1\) denotes the energy station in \(T_i^k\). 

\[ T_i^k = (X_i,L_i), \quad X_i,V_3 = X_i \cap V_3, \quad \bigcap_{i=1}^{k} X_i,V_3 = \emptyset, \quad \bigcup_{i=1}^{k} X_i,V_3 = V_3 \]  

This constraint assures one load block can only map to one energy station and every load block in set \(V_3\) is served. One load block cannot be served by two or more of the energy stations.

\[ X_i,V_1 = X_i \cap V_1 = z_i, \quad \forall i = 1,2,\ldots,k, \quad z_i \in V_1 \]

This constraint assures that there must be only one energy station in a subsystem of the network layout.

\[ \max W_{DIS}(z_i,f_{ij}) \leq 3, \quad \forall z_i \in T^k(X_i,V_1), \quad \forall f_{ij} \in T^k_i,V_3 \]

The initial investment and the operating costs of the transmission and distribution network increase with the service range. This constraint assures that the longest distance from an energy station to its load blocks cannot be more than 3.0 km [41].

3.3. Algorithm Realization

In the flow chart of the regional energy system’s optimum economic distribution algorithm, \(T_K\) is assumed as the set of dynamic minimum spanning trees, and \(K\) is a cyclic variable that represents the number of elements of the tree in the set of \(T_K\). \(P(V_1,K)\) is assumed as the set of selected energy stations, \(N\) is a cyclic variable that represents the number of elements in the set of \(P(V_1,K)\), and \(P_N(V_1,K)\) represents the nth element of the set \(P(V_1,K)\).

The specific process of the algorithm is as follows:

Initially, establish the structure graph \(G = (V,E,Q,D)\), forecast the load of all points in set \(V_3\), set all the initial parameters, and let variables \(K = 1, SSmin = \infty, N = 1\), flag = 0.

Step 1: get minimum path matrix \(W_{min} = (d_{v,v'})_{N \times N}, N = |V_0| + |V_1| + |V_2| + |V_3|\) based on the Floyd algorithm.

\[ d'_{v,v'} = \begin{cases} 0, & v = v' \\ d(e), & e \in E \\ \min_{e \in E} D(e), & \text{other} \end{cases} \]

Step 2: if \(K > |V_1|\), go to step 13, otherwise, execute step 3;

Step 3: get set \(P(V_1,K)\), the set of all subsets of set \(V_1\), and the number of elements of every subset is \(K\);

Step 4: if \(N > |P(V_1,K)|\), go to step 12, otherwise, execute step 5;
Step 5: if the path between \( v_i^j \) and \( v_j^1 \) is the shortest compared to \( v_1^1 \), then distribute \( v_i^j \) into \( X^j \), until all the elements of set \( V_3 \) are distributed, to finally get the set of \( X_K \) in \( T_K \);

Step 6: if the longest distance between every load point and its corresponding energy station is more than 3 km, go to step 7, otherwise, go to step 8;

Step 7: flag = 1, go to step 10;

Step 8: improve the Prim algorithm, and get the set of dynamic minimum spanning trees by using it;

Step 9: calculate objective function LCC, then sum up and get SS;

Step 10: if SS is the least in the loop and flag = 0, update SSmin, and save \( P_N(V_1, K) \), and \( T_K \), otherwise, go to step 11;

Step 11: \( N = N + 1, SS = 0, \) flag = 0, go to step 4;

Step 12: \( K = K + 1, \) go to step 2;

Step 13: output SSmin, \( P_N(V_1, K) \) and \( T_K \).

At this point, the algorithm ends. And the flow chart is as follows: (Figure 4)

![Flow chart of RIES’s optimum economic distribution algorithm.](image)

**Figure 4.** Flow chart of RIES’s optimum economic distribution algorithm.

### 4. Case Study

This paper analyzes a large-scale RIES of a China–U.S. low-carbon demonstration city. The city is designed to serve about 150,000 local residents. This project adopts integrated energy for the municipal heat source, ice storage, natural gas distributed energy, ground source heat pump, and renewable water source heat pump. This region is equipped with 19 sub-section posts and natural gas pipeline networks, generally uniformly distributed, and there is a renewable water plant by the river, northwest
of the region. Its processing capacity is 20,000 tons of water, 12 °C in winter, 25 °C in summer, per day. A large area of green space is planned, which will provide congenital advantages for the use of the soil source heat pump system. There is a local heating power company, whose heat supply network covers this design area. The access point of the municipal heat source network is located at the intersection of two main roads. When any energy station fails, the municipal heat source will be used as a backup heat source to directly heat the buildings in the area. The regional energy distribution map is shown in Figure 5a.

![Figure 5a](image)

**Figure 5.** Regional distribution. (a) Energy resources distribution; (b) Load blocks and preparatory energy station distribution. Adapted from China Shanghai architectural design and research institute Co., Ltd. [42].

The system covers a total planned area of 18.1 million square meters. The construction area for the efficient cooling and heating is planned to be 6 million square meters. The area distribution table of each type of building is shown below (Table 2).

| Building Types | Residential Building | Department Store | Office Building | Hotel Building | School Building |
|----------------|----------------------|------------------|----------------|----------------|----------------|
| construction area $[10^4 \text{ m}^2]$ | 195.5 | 217.2 | 161.0 | 20.0 | 7.3 |

According to the principle of energy station selection, energy stations should be located close to energy sources and the load position, but far from prosperous central parts of the city or areas requiring lower noise. There are eight probationary energy stations, from which 85 blocks should be served, as shown in Figure 5b. The type of block is defined as the type of building of the largest area. This chapter aims to choose the best part of these eight energy stations and to design an optimal pipeline network layout, so that this system can obtain the optimum economy using the mathematical model put forward above. According to the energy type recommended level based on evaluation indexes, these eight energy stations connect properties of recommended energy types and their forms of utilization. The system adopts a multi-source tree-shaped network, and the temperature of the chilled water supply and return water is 4/11 °C in summer, while that of the hot water supply and return water is 55/45 °C in winter.
5. Results Analysis

5.1. Forecast of the Region Load

The load of different blocks cannot be simply superimposed, considering the asynchronism of energy use and overall energy planning of the region. Software DeST is used to simulate different types of building and to obtain each type’s hourly load index. The load index of a residential building is shown in Table 3.

| Time | Index [W/m²] |
|------|--------------|
| 1:00 | 21.29        |
| 2:00 | 21.55        |
| 3:00 | 21.89        |
| 4:00 | 22.14        |
| 5:00 | 22.40        |
| 6:00 | 25.20        |
| 7:00 | 27.12        |
| 8:00 | 27.89        |
| 9:00 | 30.35        |
| 10:00| 32.82        |
| 11:00| 34.69        |
| 12:00| 35.79        |
| 13:00| 36.86        |
| 14:00| 35.67        |
| 15:00| 37.96        |
| 16:00| 36.00        |
| 17:00| 34.73        |
| 18:00| 35.04        |
| 19:00| 36.08        |
| 20:00| 35.66        |
| 21:00| 36.68        |
| 22:00| 35.64        |
| 23:00| 35.64        |
| 24:00| 24.97        |

Finally, the design load of every block and the total load of the region were obtained. The total cooling load is 329.97 MW, and total heat load is 130.88 MW. Part of some blocks’ cooling load results are shown in Table 4. The biggest load is at 5 pm, and the department store accounts for the largest portion of the region’s total load, as shown in Figure 6.

| Block No. | Cooling Load [MW] |
|-----------|-------------------|
| 408       | 0.82              |
| 413       | 1.07              |
| 414       | 1.07              |
| 415       | 3.22              |
| 416       | 3.05              |
| 417       | 1.03              |
| 445       | 0.63              |
| 446       | 3.02              |

Figure 6. Hourly cooling load of each type of building and the hourly total load.

5.2. Model Calculation

The selection of the parameters was as follows: (Table 5)
Table 5. Simulation parameters of the case study.

| Parameters       | Value  | Properties                               |
|------------------|--------|------------------------------------------|
| $i_0$ [%]        | 10     | Benchmark discount rate                  |
| $c$ [kJ/(kg·°C)] | 4.2    | Specific heat                            |
| $\Delta T$ [°C] | 8      | Supply and return water temperature diff  |
| $c_e$ [RMB/kWh]  | 0.9    | Electricity price                        |
| $\rho$ [kg/m³]   | 1000   | Fluid density                            |
| $m$ [a]          | 20     | Life cycle of the energy system          |
| $v$ [m/s]        | 0.8    | Flow velocity                            |
| $g$ [RMB/(kW·a)] | 50     | Annual management costs index            |
| $p_{fw}$ [RMB/m²] | 7050  | Unit price of civil construction         |
| $COP_{ES}$       | 3.88   | Performance coefficient of the system    |

The structure graph of this region is drawn based on graph theory, and its prototype comes from the original city planning diagram. As is shown in Figure 7, the energy resources, like the water resources, the access point of the municipal heat source, and so on, are represented by triangles in the graph. The numbers of the load blocks and city traffic intersections are 85 and 409, respectively.

![Figure 7. Structure distribution graph of the case study.](image)

In the simulation by MATLAB R2014a, the eight preliminary energy stations provided 255 options in total, and under the constraint conditions, 87 options were still available. Figure 8 shows that there is a great difference in the costs of different options and network layouts. The maximum of life cycle costs reached 2.45 billion RMB, 9.8% more than the lowest. This result provides us with an optimal choice, represented by bigger dots in Figure 8, with the lowest life cycle costs.
6. Discussion

This paper discusses the optimal choice further below. The differences in costs between the three lowest-cost options are small, so the bottom three all can be listed into one selection. But from a station management perspective, the selection of four energy stations is more efficient than five stations. Therefore, the optimal network layout result, with the lowest life cycle costs, is shown in Figure 9. Four energy stations, represented with bigger squares, have been selected, and the region energy system has been divided into four subsystems. All load points have found their corresponding station from which they absorb energy. The lowest life cycle costs and the lowest costs for the transmission and distribution network are finally reached. The lowest life cycle costs of this case study are calculated to be 2.233 billion RMB. The layout of the transmission and distribution network has the shortest trunk line with many branches, making it uniform and stable. The result verifies the feasibility and validity of the model and method and helped us to comprehensively integrate regional energy and accurately plan the city.
7. Conclusions

This paper studied the preliminary planning of RIES. A method of energy system economic planning with graph theory-based model was applied to RIES in a China–U.S. low-carbon demonstration city. The civil construction costs of the energy station room, the equipment costs, management costs, energy consumption costs, and operating costs and initial investment of the pipeline network were considered. The energy types and use are closely related to the system economy. National energy policy affects largely the energy types and utilization forms, and public subsidies should be added to the life cycle costs of the energy system. Specifically, some conclusions were obtained, as follows.

The regional energies situation was analyzed from the aspects of energy consumption, environmental influence, and economy in the primary stage of RIES establishment. The different recommendation rates of different energy types and use are influenced by national energy policy, regional resource conditions, building formats, and so on.

The improved Prim algorithm was proved to be in better accordance with the design idea of the tree-shaped pipeline network. The improved Prim algorithm considered the accumulated load of all downstream points to be served. The calculated results of this algorithm showed the reliability and stability of the pipeline network layout.

The graph theory-based model solved the location problem of energy stations and calculated the optimal number of energy stations to be established. The optimal layout of the pipeline network was then determined. The calculation results of the study case verified the feasibility of the proposed model.

Author Contributions: Conceptualization, Y.C. and J.Z.; Investigation, Y.C.; Methodology, Y.C.; Software, Y.C.; Supervision, J.Z.; Validation, J.Z.; Writing—original draft, Y.C.; Writing—review and editing, J.Z.

Funding: This research received no external funding.

Conflicts of Interest: The authors declare no conflict of interest.

References
1. BP British Petroleum. BP Statistical Review of World Energy 2019, 68th ed.; BP British Petroleum: London, UK, 2019.
2. Sánchez-Lugo, A.; Berrisford, P.; Morice, C.; Nicolas, J.P. Global surface temperature. In State of the Climate in 2018; Blunden, J., Arndt, D.S., Eds.; B American Meteorological Society: Washington, DC, USA, 2019; Volume 100, pp. S11–S22.
3. IEA. World Energy Outlook 2008; International Energy Agency/OECD: Paris, France, 2008.
4. Li, W.Y.; Wang, L.Y. The first “sketch” of China’s urban energy development has been unveiled. China Energy News, 22 October 2018; p. 011.
5. China’s State Council. National New Urbanization Plan (2014–2020). Available online: http://www.gov.cn/zhengce/2014-03/16/content_2640075.htm (accessed on 16 March 2014).
6. IPCC. History of the IPCC. Available online: https://www.ipcc.ch/ (accessed on 1 October 2017).
7. UN. Climate Action Summit: UN Secretary General Calls for The Climate Crisis to Stop before It Ends Humanity. Available online: https://news.un.org/zh/story/2019/09/1042012 (accessed on 23 September 2019).
8. Rabe, M.; Streimikiene, D.; Bilan, Y. EU Carbon Emissions Market Development and Its Impact on Penetration of Renewables in the Power Sector. Energies 2019, 12, 2961. [CrossRef]
9. National Development and Reform Commission. Circular of the National Development and Reform Commission on the Pilot Work of Low-Carbon Provinces and Cities. Available online: http://www.ndrc.gov.cn/zcfb/zcfbtz/201008/t20100810_365264.html (accessed on 19 July 2010).
10. National Development and Reform Commission. Circular of the National Development and Reform Commission on Launching the Third Batch of National Low-Carbon City Pilot Projects. Available online: http://www.ndrc.gov.cn/zcfb/zcfbtz/201701/t20170124_836394.html (accessed on 7 January 2017).
11. Long, W. Community Energy Planning for Built Environment in Low Carbon Cities, 1st ed.; China Architecture & Building Press: Beijing, China, 2011; pp. 15–25.
12. Zhao, D.M.; Xia, X.; Tao, R. Optimal Configuration of Electric-Gas-Thermal Multi-Energy Storage System for Regional Integrated Energy System. Energies 2019, 12, 2586. [CrossRef]
13. Cesena, E.A.M.; Mancarella, P. Energy Systems Integration in Smart Districts: Robust Optimisation of Multi-Energy Flows in Integrated Electricity, Heat and Gas Networks. *IEEE Trans. Smart Grid* 2019, 10, 1122–1131. [CrossRef]
14. Bordin, C.; Gordini, A.; Vigo, D. An optimization approach for district heating strategic network design. *Eur. J. Oper. Res.* 2016, 252, 296–307. [CrossRef]
15. Lygnerud, K. Business Model Changes in District Heating: The Impact of the Technology Shift from the Third to the Fourth Generation. *Energies* 2019, 12, 1778. [CrossRef]
16. Eveloy, V.; Ayou, D.S. Sustainable District Cooling Systems: Status, Challenges, and Future Opportunities, with Emphasis on Cooling-Dominated Regions. *Energies* 2019, 12, 235. [CrossRef]
17. Rezaie, B.; Rosen, M.A. District heating and cooling: Review of technology and potential enhancements. *Appl. Energy* 2012, 93, 2–10. [CrossRef]
18. Li, Y.; Rezgui, Y.; Zhu, H.X. District heating and cooling optimization and enhancement—Towards integration of renewables, storage and smart grid. *Renew. Sustain. Energy Rev.* 2017, 72, 281–294. [CrossRef]
19. Ma, W.W.; Fang, S.; Liu, G. Hybrid optimization method and seasonal operation strategy for distributed energy system integrating CCHP, photovoltaic and ground source heat pump. *Energy* 2017, 141, 1439–1455. [CrossRef]
20. Feng, L.J.; Dai, X.Y.; Mo, J.R.; Shi, L. Performance assessment of CCHP systems with different cooling supply modes and operation strategies. *Energy Convers. Manag.* 2019, 192, 188–201. [CrossRef]
21. Zhu, S.J.; Liu, H.M.; Xu, J.Y.; Chen, Z.; Niu, M. Study on the day-ahead co-operation strategy of regional integrated energy system including CCHP. *J. Eng. JOE* 2019, 18, 5219–5223. [CrossRef]
22. Nouri, G.; Noorollahi, Y.; Yousefi, H. Designing and optimization of solar assisted ground source heat pump system to supply heating, cooling and hot water demands. *Geothermics* 2019, 82, 212–231. [CrossRef]
23. Blazquez, C.S.; Borge-Diez, D.; Nieto, I.M.; Martin, A.F.; Gonzalez-Aguilera, D. Technical optimization of the energy supply in geothermal heat pumps. *Geothermics* 2019, 81, 133–142. [CrossRef]
24. Schmidt, D. Low Temperature District Heating for Future Energy Systems. *Energy Procedia* 2018, 149, 595–604. [CrossRef]
25. Kauko, H.; Kvalsvik, K.H.; Rohde, D.; Nord, N.; Utne, A. Dynamic modeling of local district heating grids with prosumers: A case study for Norway. *Energy* 2018, 151, 261–271. [CrossRef]
26. Tran, T.T.D.; Smith, A.D. Stochastic Optimization for Integration of Renewable Energy Technologies in District Energy Systems for Cost-Effective Use. *Energies* 2019, 12, 533. [CrossRef]
27. BERC. *China Building Energy Use 2018*; Building Energy Research Center of Tsinghua University: Beijing, China, 2018.
28. IEA-BERC. *Building Energy Use in China: Transforming Construction and Influencing Consumption to 2050*; International Energy Agency/OECD: Paris, France, 2015.
29. MOHURD. *Statistical Yearbook on Urban-Rural Construction 2017*; Ministry of Housing and Urban-Rural Development of the People’s Republic of China: Beijing, China, 2017.
30. IEA-BERC. *District Energy Systems in China*; International Energy Agency/OECD: Paris, France, 2018.
31. Wang, M.M. Study on Optimal Design of District Energy Base Station and Pipe Network. Master’s Thesis, Hunan University of Technology, Hunan, China, 2015.
32. Liang, H.; Long, W. Application of regional energy Internet model based on multi-energy complementarity in the construction of low-carbon eco-city. In Proceedings of the National HVAC Academic Conference 2010, Hangzhou, Zhejiang, China, 9 November 2010; Chinese Association of Refrigeration: Beijing, China, 2010; pp. 268–278.
33. Chen, J.; Huang, Y.S.; Lu, B. Research on “Stations-Pipelines” Layout and Optimization of Regional Energy Internet. *Proc. CSEE* 2019, 38, 0675–0685.
34. Yi, W.F.; Yu, Y.Z.; Zhang, Y.W.; Li, Z.C.; Huang, Y.Z. P-media Model Based Optimal Planning of Energy Station for Regional Integrated Energy System. *Autom. Electr. Power Syst.* 2019, 43, 107–114.
35. Wang, Z.Q.; Liu, Q.R.; Ruan, Y.J.; Wang, H. Research and Application of Optimal Layout Method for Station Network Layout in Regional Energy Planning. *J. Eng. Therm. Energy Power* 2019, 08, 25–31.
36. Dan, Z.G. Investigation and evaluation on large scale p-median problem—a case study of Dalarna, Sweden. Master’s Thesis, Tianjin University of Finance and Economics, Tianjin, China, 2013.
37. Yan, J.L.; Lin, J.G.; Lou, K.W.; Zhang, X.; Sheng, D.R. Evaluation system for building integrated energy system based on AHP-CV method. *Therm. Power Gener.* 2019, 49, 1–7. [CrossRef]
38. Fan, J.J. Analysis of energy consumption characteristics and energy consumption prediction method. Master’s Thesis, Beijing University of Civil Engineering and Architecture, Beijing, China, 2017.

39. Zhu, H.G.; Wang, Z.; Shi, H.X.; Pei, X.P. Optimization of the Centralized Biogas Supply Pipeline Transmission and Distribution Network based on Graphic Theory. In Proceedings of the International Symposium on Biogas Technology and Environmental Engineering 2011; Chengdu, Sichuan, China, 26 October 2011, Editorial Department of China Biogas: Chengdu, China, 2011; Volume 18, pp. 135–141.

40. Xiao, Y.M.; Fu, X.Z. Primary Research about the Analysis of Pipe-network for Fluid Supply by MATLAB. J. Chongqing Univ. (Nat. Sci. Ed.) 2002, 08, 14–18.

41. Kang, Y.Z.; Wu, H.; Hua, B. Research on economical cooling supply distance of district cooling systems. HVAC 2010, 40, 135–140.

42. China Shanghai Architectural Design & Research Institute Co., Ltd. Feasibility Report of Regional Energy Project in Core Srea of Binhu; China Shanghai Architectural Design & Research Institute Co., Ltd.: Hefei, China, 2017.

© 2019 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (http://creativecommons.org/licenses/by/4.0/).