



Article Operational Optimization of Regional Integrated Energy Systems with Heat Pumps and Hydrogen Renewable Energy under Integrated Demand Response

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Abstract: A regional integrated energy system (RIES), synergizing multiple energy forms, is pivotal for enhancing renewable energy use and mitigating the greenhouse effect. Considering that the equipment of the current regional comprehensive energy system is relatively simple, there is a coupling relationship linking power generation, refrigeration, and heating in the cogeneration system, which is complex and cannot directly meet various load demands. This article proposes a RIES optimization model for bottom-source heat pumps and hydrogen storage systems in the context of comprehensive demand response. First, P2G electric hydrogen production technology was introduced into RIES to give full play to the high efficiency advantages of hydrogen energy storage system, and the adjustable thermoelectric ratio of the HFC was considered. The HFC could adjust its own thermoelectric ratio according to the system load and unit output. Second, through the groundsource heat pump's cleaning efficiency function, further separation and cooling could be achieved. The heat and electrical output of RIES improved the operating efficiency of the system. Thirdly, a comprehensive demand response model for heating, cooling, and electricity was established to enable users to reasonably adjust their own energy use strategies to promote the rational distribution of energy in the system. The model integrates power-to-gas (P2G) technology, leveraging the tunable thermoelectric ratio of a hydrogen fuel cell (HFC) to optimize the generation of electricity and heat while maximizing the efficiency of the hydrogen storage system. Empirical analysis substantiated the proposed RIES model's effectiveness and economic benefits when integrating ground-source HP and electric hydrogen production with IDR. Compared with the original model, the daily operating cost of the proposed model was reduced by RMB 1884.16.

Keywords: regional integrated energy system; hydrogen energy storage; heat pump system; integrated demand response

1. Introduction

In the context of dwindling traditional energy reserves and escalating environmental challenges, the pursuit of clean energy and enhanced energy efficiency has become paramount for the future direction of the global energy industry, a priority underscored by the scholarly discourse [1]. Integrated energy systems (IES) are increasingly playing a role in promoting local sustainable development and integrating renewable energy use, a strength supported by policies for sustainable development and the promotion of renewable energy use [2,3]. The RIES, emblematic of user-side multi-energy coupling systems, incorporates an array of energy conversion technologies, including combined cooling, heating, and power (CCHP); power-to-gas (P2G); and energy storage systems (ESSs). These technologies are designed to enable a cohesive operation across electricity, heating (cooling), and gas modalities [4], thereby enhancing the efficiency of renewable energy consumption and reducing the environmental impact.



Citation: Duan, P.; Feng, M.; Zhao, B.; Xue, Q.; Li, K.; Chen, J. Operational Optimization of Regional Integrated Energy Systems with Heat Pumps and Hydrogen Renewable Energy under Integrated Demand Response. *Sustainability* 2024, *16*, 1217. https:// doi.org/10.3390/su16031217

Academic Editor: Antonio Caggiano

Received: 24 December 2023 Revised: 15 January 2024 Accepted: 29 January 2024 Published: 31 January 2024



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Hydrogen, as a versatile and clean secondary energy carrier, emerges as a key facilitator in bridging the gap between renewable energy and traditional fossil fuel systems [5]. Its ease of conversion into electricity and heat, coupled with its high conversion efficiency and various production pathways, positions hydrogen as a formidable candidate in the renewable energy landscape. Combining natural gas with P2G technology, the application of P2G technology to compensate for intermittency of the power grid on islands was studied, and the diversity of P2G technology applications was expanded [6]. Nazari-Heris [7] indicated that combination of P2G technology and demand response (DR) improved the high permeability of wind turbines. Thus, the operating cost of the system was lowest, social welfare was maximized, and carbon dioxide emissions were reduced. Analyses integrating P2G with carbon trading mechanisms into heating networks suggested economic and environmental benefits, transforming surplus electrical energy into hydrogen and then into natural gas [8]. Additionally, a new decision-making framework for electricity markets that incorporates photovoltaic (PV)-integrated hydrogen storage within RIESs has been proposed [9]. However, these studies generally limit their focus to the role of hydrogen storage without exploring its broader applications. Zhang et al. [10] presented an IES model that considered both carbon capture technology and P2G, where P2G captured surplus electricity from renewable sources, and carbon capture technology sequestered system-generated carbon dioxide (CO_2) , and the application of the P2G system to carbon dioxide was pointed out. Alessandra [11] combined P2G technology with the ESS of solid oxide coelectrolytic cells (EC) and high-temperature methanators and conducted a comprehensive technical and economic evaluation. Xin et al. [12] integrated P2G with carbon capture and storage (CCS) in interconnected energy systems and observed a 2.72% reduction in CO₂ emissions, enhancing the utilization of wind power. Alessandra and Xin collectively suggested substantial economic and environmental benefits from P2G implementation in RIESs, primarily focusing on conversion from electricity to natural gas. Nevertheless, there has been limited exploration of the electric hydrogen production phase, particularly the benefits of the tunable thermoelectric ratio of a hydrogen fuel cell (HFC) in the RIES.

Furthermore, as systems' operators and end users pay more and more attention to energy conservation, there have been more studies devoted to increasing the resilience of independent IES [13], which lays the foundation for the following RIES DR, and demand response (DR) technology is becoming increasingly relevant. DR contributes to peak shaving and load shifting, smoothing the load curve, and reducing both the system's costs and the retailer's risks, as widely documented in power system applications [14]. The DR mechanism is typically bifurcated into price-based and incentive-based responses [15,16], each playing a strategic role in energy management and the system's efficiency.

Simulations across various scenarios in Ali et al. [17] examined the impact of differing DR levels, the integration of renewable energy, and capacity payments, confirming that DR can enhance renewable energy uptake and reduce CO_2 emissions. Rezaee et al. [18] built the heating network model and a mathematical model of P2G, and introduced the carbon trading mechanism into it, taking the minimum operating cost as the objective function to verify the economy of P2G. Jeseok and Jinho [19] outlined a DR participation strategy that offered high profit potential, albeit with considerable uncertainty, and presented a non-cooperative game model to mitigate this uncertainty through trading surplus power from ESS. However, these investigations largely focused on the traditional electrical DR and did not account for the subjective nature of human perceptions of temperature changes. Adjusting temperatures within a certain threshold generally does not significantly impact human comfort. Zhang et al. [20] utilized Monte Carlo simulation to evaluate the reliability of an electric heat pump (HP) heating system, illustrating its significant impact on the power grid's safety and the system's overall reliability. Linfei and Min [21] considered an IES with a distributed ground-source HP for heat storage and carbon recapture devices, showing that such a system could reduce carbon emissions by at least 82.02% from conventional levels. Linfei et al. [22] developed an optimal scheduling model for a distributed ground-source

HP heat storage system, indicating reductions in waste and economic costs over traditional systems. However, they did not explore the potential advantages of combining HP with P2G in a RIES to enhance cold thermoelectric production in the system, leaving the energy-saving potential of HP with P2G in this configuration largely untested. A comparative analysis of appellate literature and this paper is shown in Table 1. Abbreviations of all devices and parameters used in this paper are shown in Table 2.

				P2G		HP	D	DR
Ref.	System	Objective Function	Electric Natural Gas	Electric Hydrogen Production	Thermoelectric Ratio Adjustability		CDR and HDR	EDR
[6]	P2G system	Minimal operating cost	\checkmark	\checkmark				
[7]	Electric power system	Minimal operating costs and minimal environmental costs	\checkmark					\checkmark
[8]	P2G system, heating system	Minimal operating costs and minimal carbon trading costs	\checkmark					
[9]	RIES	Minimal operating costs		\checkmark				\checkmark
[10]	IES	Minimal operating costs and minimal environmental costs						
[11]	P2G system	Minimal operating costs and reductions in CO ₂	\checkmark					
[12]	P2G system, CHP	Minimal operating costs and CO ₂ reduction	\checkmark					
[18]	P2G system, electric power system	Minimal operating costs		\checkmark				
[20]	CHP	Minimal operating costs				\checkmark		
[21]	IES	Minimal operating costs				\checkmark		
[22]	IES	Minimal operating cost and stability				\checkmark		
Proposed method	RIES	Minimal operating costs		\checkmark	\checkmark	\checkmark	\checkmark	\checkmark

Table 1. Summary of the literature review.

This study sought to address this gap by examining the inclusion of ground-source HP and hydrogen ESS, coupled with a cold, heat, and electricity IDR optimization model that accounted for dynamic electricity pricing and the human body's variable temperature perception. The research pivoted from a single-user perspective to a multi-agent system at the source side, with the aim of meeting the users' energy needs while reducing costs within the RIES. An example is provided to demonstrate the model's effectiveness. The study's main contributions are as follows:

- Introducing an IDR into RIES that considers the human body's delayed and subjective temperature perception, proposing an optimization method for the operation of the RIES that includes IDR.
- (2) Incorporating a refined P2G electric hydrogen production model with an adjustable thermoelectric ratio to support the RIES's transition from traditional fossil fuels to RESs.

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- (3) Including the ground-source HP as conversion equipment in the traditional RIES and conducting a comparative analysis to confirm the model's economic and energy-saving benefits.

Table 2. Abbreviations and variable parameters.

Abbreviation		P_{pv}^t	Photovoltaic Output Value
AR	Absorption refrigerator	P ^t _{pv}	Photovoltaic output value
CS	Cooling storage	L [†] _p	Solar irradiance
DR	Demand response	Q ^t _{et}	GT's remaining heat at time t
EC	Electrolytic cell	$Q_{c.ar}^{f}$	Cooling capacity of the AR at time t
E/H/CL	Electrical/heating/cooling load	Q ^t _{ar.in}	Input power of the AR at time t
GT	Gas turbine	$Q_{h,whb}^{t}$	Heat production of the WHB at time t
HFC	Hydrogen fuel cell	P ^t _{e,gt}	GT output power at time t
HP	Geothermal heat pump	Q _{whb.in}	Input power of the WHB at time t
HST	Hydrogen storage tank	L ^t _{felehc}	Fixed load at time t
HS	Heating storage	L ^t selehc	The load can be transferred at time t
IDR	Integrated demand response	$P_{h_{2el}}$	Hydrogen energy output by the EC
PV	Photovoltaic	P _{e,hfc} , P _{h,hfc}	HFC's output of electricity and heat
RIES	Regional integrated energy system	$Q_{hp,c'}^{t}Q_{hp,h}^{t}$	HP's cooling, heat production at
WHB	Waste heat boiler		time t
WT	Gas turbine	P ^t _{hp,c} , P ^t _{hp,h}	HP's cooling and heat consumption at
Variables		1,- 1,-	time t
Pwt	Wind field's output value	Indices	
v _m ^t	Field speed	Т	Index for a typical day
P _{r,m}	Rated power of the wind farm	t	Index for time periods in a typical day
v_c^t	Wind field's cut wind speed	i, j	Index of scenarios
v_r^t	Wind field's cutting wind speed		

2. The RIES' Structure and IDR Model

2.1. The RIES Framework

The architecture of the cogeneration microgrid, as proposed in this study, integrates HP and electric hydrogen production technologies. This model, depicted in Figure 1, delineates four principal energy streams: electricity, heat, cold, and gas [23]. The framework comprises the following:

- (1) Renewable energy units: these include wind turbines (WT) and PV panels, which harness wind and solar energy for generating power.
- (2) Energy conversion devices: this category encompasses a suite of technologies such as gas turbines (GT), ground-source HP units, waste heat boilers (WHB), absorption refrigerators (AR), EC, and HFC, each contributing to the system's ability to convert energy from one form to another.
- (3) Energy storage devices: the system incorporates a hydrogen storage tank (HST) for hydrogen retention and a combined heat and cold storage tank (HS/CS) to maintain thermal energy reserves.
- (4) Load side components: the demand side includes the electrical load (EL), the heat load (HL), and the cold load (CL), all of which are integral to the IDR strategy.

Priority should be given to the use of renewable WT PV in regional cooperation and supply. The gas turbine (GT) can purchase natural gas from the natural gas network for combustion and work, and the high-grade energy released is used for power generation. The high-temperature smoke discharged is reused by the waste heat, cooled by absorption chillers, and utilized by the thermal production capacity of the waste heat boilers. Ground-source heat pumps convert electrical and surface geothermal energy into cold and heat energy. The source side energy in this study is mainly composed of renewable energy, electric energy interactions, and natural gas. It uses the energy source side through a coupling device to meet the load demand of energy conversion; at the same time, the



remaining energy is stored in the storage unit of the HST and HS/CS. The energy structure of the entire system is shown in Figure 1.

Figure 1. Framework of the RIES.

2.2. IDR Model

This study expanded upon traditional DR techniques by developing an integrated electric-hot-cold IDR model for the coordinated management of multiple loads within an energy hub. The users' perceptions of temperature are somewhat imprecise and delayed, allowing for a greater range of adjustment in the load's flexibility. Consequently, implementing IDR in a RIES can effectively smooth out fluctuations in the load, facilitate the integration and complementarity of multiple forms of energy, and enhance the utilization of clean energy sources.

2.2.1. Power Load IDR

This study introduced a price-based electricity DR, which uses time-of-use pricing to guide users in adjusting their energy consumption. Users can strategically shift their usage to times when electricity costs are lower [24].

$$n = \frac{\Delta p}{p} \frac{q}{\Delta q} \tag{1}$$

The elasticity matrix *N*, representing the responsiveness of the demand side to changes in electricity prices, is defined as

$$N = \begin{bmatrix} \varepsilon_{11} & \varepsilon_{12} & \cdots & \varepsilon_{1m} \\ \varepsilon_{21} & \varepsilon_{22} & \cdots & \varepsilon_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ \varepsilon_{n1} & \varepsilon_{n2} & \cdots & \varepsilon_{nm} \end{bmatrix}$$
(2)

$$\varepsilon_{ii} = \frac{\Delta p_i}{p_i} \frac{q_i}{\Delta q_i} \tag{3}$$

$$\varepsilon_{ij} = \frac{\Delta p_i}{p_i} \frac{q_j}{\Delta q_j} \tag{4}$$

where ε_{ii} and ε_{ij} are the self-elastic coefficients and cross-elastic coefficients, respectively. In this context, customer loads are categorized into fixed electrical loads (FELs) and transferable electrical load (TELs), with the latter being adjustable [25].

$$L_e^t = L_{fel,e}^t + L_{sel,e}^t = p_i + p_i N \frac{\Delta q_i}{q_i}$$
(5)

The two types of loads are subject to the following constraint condition

$$\begin{cases} 0 < L_{sel,e}^{t} \leq L_{sel,e,t}^{max} \\ \sum_{t=1}^{T} L_{sel}^{t} \Delta t = M_{sel,e} \end{cases}$$
(6)

where $M_{sel,e}$ is the total amount of power load that can be transferred. The cumulative EL remains constant pre- and post-DR, with the TEL designated as 10% of the total load in this study.

2.2.2. HL Demand Response

Users have a degree of tolerance for temperature fluctuations within a comfortable range, as their perception of changes in temperature is not immediate or precise [26]. This study focused on the hot water load for DR, which can vary within an acceptable range without significantly affecting the users' comfort.

$$\begin{cases} L_{h,min}^{t} \leq L_{h}^{t} \leq L_{h,max}^{t} \\ L_{h,min}^{t} = \gamma \rho_{w} V_{t} (T_{h,min} - T_{h,in}) \Delta t \\ L_{h,max}^{t} = \gamma \rho_{w} V_{t} (T_{h,max} - T_{h,min}) \Delta t \end{cases}$$
(7)

where γ and ρ_w refer to the specific heat capacity and density of water, respectively, which can be set to a fixed value of 1.1667×10^{-3} kwh/(kg °C) and 1000 kg/m³ [27], regardless of changes in temperature; $T_{h,in}$ is the initial water temperature, say 15 °C.

The user's heat demand is divided into two components: a fixed heat load (FHL) and a transferable heat load (THL), defined as follows:

$$\begin{cases} L_h^t = L_{fel,h}^t + L_{sel,h}^t \\ 0 < L_{sel,h}^t \le L_{sel,h,t}^m \\ \sum_{t=1}^T L_{sel,h}^t \Delta t = M_{sel,h} \end{cases}$$
(8)

2.2.3. CL Demand Response

$$\begin{cases} L_{c,min}^{t} \leq L_{c}^{t} \leq L_{c,max}^{t} \\ L_{c,min}^{t} = (T_{out}^{t} - T_{c,min})\Delta t / R_{d} \\ L_{c,max}^{t} = (T_{out}^{t} - T_{c,max})\Delta t / R_{d} \end{cases}$$

$$\tag{9}$$

$$\begin{cases} L_c^t = L_{fel,c}^t + L_{sel,c}^t \\ 0 < L_{sel,c}^t \le L_{sel,c,t}^{max} \\ \sum_{t=1}^T L_{sel,c}^t \Delta t = M_{sel,c} \end{cases}$$
(10)

where L_c^t is the cooling load power at time t; $M_{sel,c}$ is the total amount of transferable heat load; T_{out}^t is the outdoor temperature at time t; and R_d is the thermal resistance at 18 °C/kW.

The flexible cooling load considered in this study was the transferable load, and the user's cooling load consisted of the fixed cooling load (FCL) and the transferable cooling load (TCL). The total amount of cooling load transmitted before and after the demand response was equal.

3. Output and Constraint Modeling of Equipment in the RIES

3.1. Scenery Output Model

The generation of wind power is directly related to wind speed, and the formula for calculating wind power is widely accepted and utilized both nationally and internationally.

$$v_{wt}^{t} = \begin{cases} 0, v_{m}^{t} \leq v_{c}^{t}, v_{m}^{t} > v_{f}^{t} \\ P_{r,m} \frac{v_{m}^{t} - v_{c}^{t}}{v_{r}^{t} - v_{c}^{t}}, v_{c}^{t} < v_{m}^{t} < v_{r}^{t} \\ P_{r,m}, v_{r}^{t} \leq v_{m}^{t} \leq v_{f}^{t} \end{cases}$$
(11)

$$P_{pv}^{t} = L_{p}^{t} M_{p} \theta_{p} \tag{12}$$

$$\theta_p = \theta_{test} \left[1 - \varphi \left(T_p^t - T_{test} \right) \right] \tag{13}$$

where, θ_p is the photovoltaic power generation efficiency and φ is the power temperature coefficient. Equation (11) quantifies the wind power plant's output as a function of wind speed. Equation (12) denotes the theoretical output of a solar power plant at time *t*, while Equation (13) specifies the efficiency of solar power generation at that same time.

3.2. Mathematical Model of CCHP

$$\begin{pmatrix}
Q_{gt}^{t} = P_{e,gt}^{t} \left(\frac{1-\eta_{e}-\eta_{l}}{\eta_{e}}\right) \\
V_{gt,t} = \frac{P_{e,gt}^{t}}{\eta_{e} \times L_{p}} \\
P_{e,st}^{min} \leq P_{e,gt}^{t} \leq P_{e,st}^{max}
\end{cases}$$
(14)

$$\begin{cases} Q_{c,ar}^{t} = \eta_{ar} Q_{ar,in}^{t} \\ Q_{c,ar}^{min} \le Q_{c,ar}^{t} \le Q_{c,ar}^{max} \end{cases}$$
(15)

$$\begin{cases}
Q_{h,whb}^{t} = \eta_{whb} Q_{whb,in}^{t} \\
Q_{h,whb}^{min} \leq Q_{h,whb}^{t} \leq Q_{h,whb}^{max}
\end{cases}$$
(16)

where η_e and η_l are the power generation efficiency and heat loss coefficient of the GT respectively; η_{ar} is the conversion efficiency of the AR; and η_{whb} is the conversion efficiency of the WHB. Formulas (14)–(16) are mathematical models for the power generation efficiency of gas turbines, absorption coolers, and waste heat boilers, respectively, including the cooling power of the absorption cooler and the heating power of the waste heat boiler.

3.3. Mathematical Model of Ground-Source HPs

Ground-source HP systems utilize soil, groundwater, surface water, and low-temperature geothermal tail water as sources of geothermal energy [28]. Their advantages are that they can be widely used, including summer cooling and winter heating for users, and they can also be used for low-temperature heat source refrigeration, heating, and refrigeration. Water coolers and boilers are effective ways to improve the urban atmospheric environment and save energy, and are a new development direction for local energy use. The mathematical model for this system is as follows

$$\begin{cases}
Q_{hp,c}^{t} = C_{cop,c} P_{hp,c}^{t} \\
Q_{hp,h}^{t} = C_{cop,h} Q_{hp,h}^{t} \\
Q_{hp} \leq Q_{hp,max}
\end{cases}$$
(17)

where $Q_{hp,max}$ is the maximum installed capacity of the HP. The coefficient of performance (COP) is an important indicator of a heat pump's energy efficiency, with an HP energy efficiency coefficient of 4.

3.4. Mathematical Model of Electric Hydrogen Production Equipment

In this study, part of the hydrogen produced by EC electrolysis was supplied to a hydrogen fuel cell (HFC) for energy use, and part is stored in a hydrogen storage tank for backup. The HFC converts the hydrogen energy obtained from the electrolytic cells and hydrogen storage tanks into electricity and heat for the users to use [29].

The mathematical model of EC is as follows:

$$\begin{cases}
P_{h_{2,ec}}^{t} = \eta_{el} P_{e,ec}^{t} \\
P_{e,ec}^{min} \leq P_{e,ec}^{t} \leq P_{e,ec}^{max} \\
\Delta P_{e,ec}^{min} \leq P_{e,ec}(t+1) - P_{e,ec}^{t} \leq \Delta P_{e,ec}^{max}
\end{cases}$$
(18)

In the formula, $\Delta P_{e,ec}^{max}$ and $\Delta P_{e,ec}^{min}$ are the upper and lower limits of EC, and $P_{e,ec}^{max}$ and $P_{e,ec}^{min}$ are the upper and lower limits of EC power.

The mathematical model of HFC is as follows:

$$\begin{cases}
P_{e,hfc}^{r} = \eta_{hfc}^{e} P_{h_{2,hfc}}^{t} \\
P_{h,hfc}^{t} = \eta_{hfc}^{h} P_{h_{2,hfc}}^{t} \\
P_{h_{2,hfc}}^{min} \leq P_{h_{2,hfc}}^{t} \leq P_{h_{2,hfc}}^{max} \\
\Delta P_{h_{2,hfc}}^{min} \leq P_{h_{2,hfc}}^{t+1} - P_{h_{2,hfc}}^{t} \leq \Delta P_{h_{2,hfc}}^{max} \\
\chi_{hfc}^{min} \leq P_{h,hfc}^{t} / P_{e,hfc}^{t} \leq \chi_{hfc}^{max}
\end{cases}$$
(19)

In the formula, $\Delta P_{h_{2,h_{fc}}}^{max}$ and $\Delta P_{h_{2,h_{fc}}}^{min}$ are the upper and lower limits of the HFC's climb, $\chi_{h_{fc}}^{max}$ and $\chi_{h_{fc}}^{min}$ are the upper and lower limits of the HFC's thermoelectric ratio, and $P_{h_{2,h_{fc}}}^{max}$ and $P_{h_{2,h_{fc}}}^{min}$ are the upper and lower limits of the HFC's power, respectively. Equations (18) and (19) are the mathematical models of the electrolyzer's EC and the HFC, respectively, which contain the climbing constraints of the EC and HFC equipment and the range constraints of the HFC's adjustable thermoelectric ratio. $P_{h_{2,ec}}^t$ is th e hydrogen energy output of the EC at time *t*, and is the electricity and heat energy output of the HFC at time *t*, respectively.

3.5. Mathematical Model of the Energy Storage Device

The system incorporates energy storage devices such as hydrogen storage tanks (HST) and combined HS/CS, with their operations governed by the following mathematical models.

$$\begin{cases} 0 \leq P_{x,t}^{cha} \leq C_{x,t}^{cha} P_{max}^{cha} \\ 0 \leq P_{x,t}^{dis} \leq C_{x,t}^{dis} P_{max}^{dis} \\ W_{x}^{t} = W_{x}^{t-1}(1-\eta_{x}^{s}) + \left(\eta_{x}^{cha} P_{x,t}^{cha} - \frac{P_{x,t}^{dis}}{\eta_{x}^{dis}}\right) \\ W_{x}^{min} \leq W_{x}^{t} \leq W_{x}^{max} \\ W_{x}^{t=1} = W_{x}^{t=T} \\ C_{x,t}^{cha} + C_{x,t}^{dis} = 1 \end{cases}$$

$$(20)$$

where, respectively, $P_{x,t}^{cha}$ and $P_{x,t}^{dis}$ are the charging and discharging power of energy storage device x during the period t; $C_{x,t}^{cha}$ and $C_{x,t}^{dis}$ are binary variables; η_x^{cha} and η_x^{dis} are the of charge and discharge, respectively; W_x^{max} and W_x^{min} are the upper and lower efficiency limits of energy storage devices of type x, respectively; and P_{max}^{cha} and P_{max}^{dis} are the maximum charge and discharge capacity of energy storage devices of type x, respectively. The constraints of the energy storage equipment are shown in Equation (20).

3.6. Constraints on the Equipment's Start-Up and Shutdown

Frequent activation and deactivation of the system's equipment can have detrimental effects; it not only diminishes the equipment's lifespan but also escalates the likelihood of operational failures [30]. Consequently, it is imperative to enforce constraints on the starting and stopping of the equipment.

$$\begin{cases} I_{pi}^{t} = 1, t = 1, 2, \cdots, U_{i} \\ \sum_{n=t}^{t+T_{pi}^{on}-1} I_{pi}(n) \geq T_{pi}^{on} \left(I_{pi}^{t} - I_{pi}^{t-1} \right), t = U_{i} + 1, U_{i} + 2, \cdots, N_{t} - T_{pi}^{on} + 1 \\ \sum_{n=t}^{N_{t}} \left[I_{pi}(n) - \left(I_{pi}^{t} - I_{pi}^{t-1} \right) \right] \geq 0, t = N_{t} - T_{pi}^{on} + 2, N_{t} - T_{pi}^{on} + 3, \cdots, N_{t} \\ U_{i} = min \left\{ N_{t}, \left(T_{pi}^{on} - K_{pi}^{on} \right) I_{pi}(0) \right\} \end{cases}$$

$$I_{pi}^{t} = 0, t = 1, 2, \cdots, V_{i}$$

$$I_{pi}^{t+T_{pi}^{off}-1} \left(1 - I_{pi}(n) \right) \geq T_{pi}^{off} \left(I_{pi}^{t-1} - I_{pi}^{t} \right), t = V_{i} + 1, V_{i} + 2, \cdots, N_{t} - T_{pi}^{off} + 1 \\ \sum_{n=t}^{N_{t}} \left[1 - I_{pi}(n) - \left(I_{pi}^{t-1} - I_{pi}^{t} \right) \right] \geq 0, t = N_{t} - T_{pi}^{off} + 2, N_{t} - T_{pi}^{off} + 3, \cdots, N_{t} \\ V_{i} = min \left\{ N_{t}, \left(T_{pi}^{off} - K_{pi}^{off} \right) \left(1 - I_{pi}(0) \right) \right\}$$

$$(22)$$

where $I_{pi}(t)$ is the state of unit *i* at time period t; $I_{pi}(0)$ is the state of unit *i* at the initial time of scheduling; T_{pi}^{on} , T_{pi}^{off} are the minimum opening and downtime of unit *i*, respectively; U_i , V_i are, respectively, the time that unit *i* needs to be in the starting and stopping state after the start of scheduling; and K_{pi}^{on} , K_{pi}^{off} refer to the continuous on and off times of unit *I* at the initial moment, respectively.

4. The RIES Optimization Model under IDR

4.1. Objective Function

This study introduced the energy conversion device of a ground-source heat pump (HP) and thermocouple, adjustable hydrogen P2G technology under the traditional cogeneration mode. Taking comprehensive demand response into account, the optimal output model of variable cogeneration of cold, heat, and electricity is formulated from both the supply and demand sides by considering various energy storage methods, energy conversion models, and complementary characteristics of multi-energy [31]. To demonstrate the model's effectiveness, the study examined five different operational scenarios, each designed to minimize the daily operating costs.

For a grid-connected RIES, the economic scheduling model's objective function is outlined as follows

$$C = min(C_{grid} + C_{gas} + C_{im} + C_{om})$$
⁽²³⁾

(1) Cost of gas of the unit:

$$C_{gas} = \sum_{t=1}^{T} c_{gas}^{t} \frac{P_{e,gt}^{t}}{\eta_e L_p}$$
(24)

(2) Operation and maintenance cost of the unit:

$$C_{im} = \sum_{t=1}^{T} K_i P_i^t + \sum_{t=1}^{T} K_j Q_i^t + \sum_{t=1}^{T} K_x \mid P_x^t \mid$$
(25)

(3) Equipment start-up and shutdown costs:

$$C_{om} = \sum_{t=1}^{T} max \left\{ 0, I_{pi}^{t} - I_{pi}^{t-1} \right\} C_{st,i}$$
(26)

(4) Electric energy interaction cost:

$$C_{grid} = \sum_{t=1}^{T} P_{buy}^t C_b^t - P_{sell} C_s^t$$
⁽²⁷⁾

- 4.2. Constraints of the Balance of the RIES
- (1) Electrical power balance constraints:

$$P_{buy}^{t} + P_{wt}^{t} + P_{pv}^{t} + P_{e,gt}^{t} + P_{e,hfc}^{t} = L_{e}^{t} + P_{e,el}^{t} + P_{sell}^{t} + P_{hp}^{t}$$
(28)

(2) Thermal power balance constraints:

$$Q_{h,whb}^t + P_{h,hfc}^t + P_{hs,t}^{dis} = L_h^t + P_{hs,t}^{cha}$$
⁽²⁹⁾

(3) Cold power balance constraints:

$$Q_{c,ar}^{t} + Q_{hp,c}^{t} + P_{cs,t}^{dis} = L_{c}^{t} + P_{cs,t}^{cha}$$
(30)

(4) Hydrogen power balance constraints:

$$P_{h_{2,el}}^t + P_{hst,t}^{dis} = P_{h_{2,hfc}}^t + P_{hst,t}^{cha}$$
(31)

4.3. Solution Method

This model is an MILP problem. The Yalmip toolbox on was is used for modeling the RIES due to its effectiveness, and the commercial solver CPLEX was used for quick solutions via this platform.

5. Results and Discussion

5.1. Baseline Data

Due to the unpredictable nature of the weather and load data, the analysis focuses on typical winter and summer days to reflect the cyclical energy consumption patterns of a building's energy systems. The forecasted wind and cooling/heating load profiles are illustrated in Figures 2 and 3, while the ToU electricity pricing for the peak and off-peak periods is shown in Figure 4. The model operates with a time interval Δt of 1 h over a scheduling period T of 24 h.

5.2. Analysis of the Influence of Different Scheduling Models on the Simulation's Results

To test the effectiveness of the integrated multi-energy system model that includes a ground-source HP, hydrogen energy storage, and comprehensive DR, five optimization scenarios were created for simulation and comparison, as shown in Table 3. The specific operational parameters for each piece of equipment are listed in Table 4, and to reduce the wear and costs associated with frequent turning on and off of the devices, minimum operation times were set, as shown in Table 4. Additionally, Table 5 details the capacity and parameters of the energy storage tanks and hydrogen storage systems. The operation and maintenance costs of each equipment are shown in Table 6.



Figure 2. Forecast curve of WT, PV, and loads in winter.



Figure 3. Forecast curve of WT, PV, and loads in summer.



Figure 4. Electricity tariffs.

Multi-Energy Coupling	Heat Pump System	Energy Storage System	IDR
	×	×	
v V	×		Ň
, V	\checkmark	×	, V
			×
			\checkmark
	Multi-Energy Coupling √ √ √ √ √ √	Multi-Energy CouplingHeat Pump System	Multi-Energy CouplingHeat Pump SystemEnergy Storage System×××××××××××

Table 3. Study cases.

Table 4. System parameters of the RIES's operation.

Microsource Type	Lower Power Limit (kW)	Upper Power Limit (kW)	Lower Limit of Climbing Speed (kW/h)	Upper Limit of Climbing Speed (kW/h)	Minimum Boot Time (h)	Minimum Shutdown Time (h)
GT	20	220	7	14	3	2
AR	30	280	6	12	3	2
WHB	30	280	6	12	3	2
HP	10	150	10	30	2	2
EC	10	100	8	20	2	2
HFC	10	100	8	20	2	2
Power grid	-300	300	-	-	-	-

Table 5. Parameters of the energy storage devices.

Type of Energy Storage	Charge and Discharge Rate	Consumption Rate	Minimum State	Maximum State	Capacity (kW·h)
Hydrogen energy storage	0.92	0.009	0.2	0.9	200
Thermal energy storage	0.90	0.01	0.2	0.9	200
Cold energy storage	0.90	0.01	0.2	0.9	200

Table 6. Unit's maintenance costs.

Micro	osource Type	Unit Price (RMB/kW)	Microsource Type	Unit Price (RMB/kW)
	WT	0.029	HP	0.023
	PV	0.025	WHB	0.021
	GT	0.025	EC	0.012
	AR	0.021	HFC	0.028
	HST	0.0018	HS/CS	0.0014

5.2.1. Influence of IDR on the RIES's Operation

Figure 5 incorporates IDR into the system without changing the total equipment. It encourages users to actively engage in shifting their energy use, moving demand from peak to off-peak hours based on personal preferences and making sensible adjustments during normal periods, as illustrated in Figure 5. For instance, during winter, the electricity load is highest between 10:00–13:00 and 18:00–21:00 when prices are also at their peak. Users can reduce their consumption during these times, which helps to dampen fluctuations in the load and shave peaks, especially during the off-peak hours of 22:00 to 6:00 when electricity is cheaper. Figure 6 shows the changes after IDR in summer. This results in a 5.01% and 5.2% reduction in the peak-to-valley difference in winter, while in summer, the reduction is 5.6% and 4.4%, respectively.



Figure 5. IDR load curve in winter.



Figure 6. IDR load curve in summer.

5.2.2. Analysis of Supply–Demand Balance Results of Optimization of Typical Winter Days

In Option 5, the heating needs of the RIES are predominantly met by geothermal HP, HFC, cogeneration systems, and HS. The use of power for geothermal HP, EC, and HFC is managed in conjunction with the grid, the cogeneration systems, and renewable sources such as solar power and WT.

As shown in Figure 7a, the internal power demand of the RIES is lower during the early morning hours (1:00–6:00) and late at night (22:00–24:00), while there is a high demand for heat during winter nights. After evaluating the costs of natural gas and grid electricity, we determined that grid electricity is more cost-effective. According to Figure 8a,b, the ground-source HP and HFC are mainly responsible for meeting the heating demand, while the HST continues to store energy.



-100 0 2 4 6 8 10 12 14 16 18 20 22 24 Time(h)



(c) Optimal results of power supply scheduling in summer (d) Summer cooling scheduling optimization results

Figure 7. Optimal intraday dispatch results of Scheme 5.



Figure 8. Balance of the energy storage system.

During the hours of 7:00–9:00 and 14:00–17:00, when electricity prices are normal, a cost comparison was made between producing electricity using the GT and drawing electricity from the grid. The GT, being the more cost-effective option, was chosen for power generation. The waste heat from the GT is utilized by a WHB to contribute to HL, with the remaining demand for heating being met by the ground-source HP and HFC. During this period, the HST is in a standby mode and does did not actively store energy, as depicted in Figure 7a,b.

During peak hours, from 10:00 to 13:00 and after 18:00, there is a high demand for electricity. A cost comparison indicated that it cheaper to generate electricity using the GT rather than buying from the grid. Thus, GTs are primarily used to supply electricity during these times, with any excess electricity potentially being sold back to the grid. The heat demand is covered by the cogeneration system. At the same time, hydrogen is released from storage to be used, and the HS reduces its absorption of excess heat, as illustrated in Figure 8a,b.

Results of Optimization of Typical Summer Days

To address the cooling needs in summer, the cogeneration system's operational mode is switched to produce both electricity and cooling, and the functioning of the groundsource HP compressor is adjusted accordingly. Figure 7 displays typical results for summer operation and energy scheduling of the HP.

Unlike in winter, the summer cooling is mostly provided by the cogeneration system and the ground-source HP. During times with lower electricity prices, the ground-source HP primarily handles the cooling load. When electricity prices are high, the GT takes over most of the cooling production, with the ground-source HP covering any additional requirements. In periods of low electricity costs, any excess power is directed to the electrolyzer for further use, while in times of high electricity prices, it is sold to the grid. The configuration of the hydrogen storage system related to this process is also depicted in Figure 8.

5.2.3. Simulation and Comparative Analysis of the Optimization Schemes

As shown in Table 7, Option 2 reduced the fuel costs and thus the overall operating expenses. Scheme 3 added a ground-source HP to Scheme 1, enhancing the system's ability to convert low-grade to high-grade energy and providing flexibility in energy supply. Scheme 5, which included both a ground-source HP and an ESS, showed even lower operating costs than Schemes 2 and 3, as depicted in Table 7, confirming the benefits of integrating these technologies. Comparing Schemes 4 and 5, Scheme 5 achieved savings of RMB 123.4 in summer and RMB 202.4 in winter. Additionally, it reduced the equipment maintenance costs, costs associated with electrical interactions, and fuel costs compared with scenarios without DR, highlighting the economic benefits of incorporating IDR into the system.

Table 7. Operating costs of different RIES cases.

Case	RIES's Running Cost (RMB)	Interaction Cost (RMB)	Start–Stop Cost (RMB)	Fuel Cost (RMB)	RIES's Total Cost (RMB)
1	377.18	-422.38	5.68	10,242.15	10,202.63
2	401.47	183.99	23.74	8907.47	9516.67
3	275.26	3868.00	7.7	4701.18	8852.15
4	330.23	3650.29	31.54	4635.87	8647.94
5	321.75	3526.33	28.01	4442.3	8318.47

5.2.4. Influence of the HFC's Adjustable Thermoelectric Ratio System

During midday, the demand for heat drops but the demand for electricity rises, resulting in a decreased thermoelectric ratio. In the afternoon, even though EC reaches its highest and the GT is active, its heat output alone cannot satisfy the heating requirements;

hence, the HFC continues to operate at an efficient thermoelectric ratio. The adjustable thermoelectric ratio of the HFC allows it to adapt to varying loads, enhancing the system's energy efficiency and reducing the operating costs of the RIES. The HFC thermoelectric ratio is shown in Figure 9.



Figure 9. Diagram of the thermoelectric ratio.

6. Conclusions

An optimal scheduling model for an RIES based on IDR was developed, incorporating a ground-source HP and a hydrogen storage system. After we had compared five scenarios, the following conclusions were drawn.

(1) Introducing P2G technology into the RIES not only enhanced the usage of wind power but also took full advantage of the efficiency of hydrogen energy. The HFC can adjust its thermoelectric ratio to match the system's energy needs, ToU electricity prices, and the output from each unit. In winter, the HFC can provide heating, reducing the demand on the GT and HP, thus lowering the RIES's operating costs. It can be seen from the operation results that the operation cost reduced by RMB 685.96.

(2) The ground-source HP's ability to upgrade low-grade to high-grade energy improved the primary energy use efficiency. According to the results of the system's operation, the RIES containing a HP can reduce the operating cost by RMB 1198.2. The integration of the HP and the hydrogen storage system decreased the dependence of cooling, heating, and electricity, allowing for better energy management and cost-effectiveness.

(3) Implementing IDR in RIES significantly narrowed the peak-to-valley difference in energy use, leading to a reduction in the daily operating costs of RMB 329.47. Users can adjust their energy use within a comfortable range. This smoothed out the energy demand, created complementarity between different energy types, and reduced the microgrid's costs while improving energy efficiency.

This study validated the advantages of introducing HP and hydrogen energy storage systems in a RIES under integrated demand response (IDR). Especially in the current environment, industry manufacturers urgently need to find a renewable energy source that can replace traditional fossil energy sources. This study presents an electric hydrogen production technology with an adjustable thermoelectric ratio for energy policy makers and industry practitioners. However, this study was a simulation under ideal conditions. In order to facilitate calculation, the mathematical models of some devices have been simplified, and the accuracy needs to be improved. Moreover, this research did not study the impact of this model on the environment, and did not consider the problems existing in practical applications of the model. In the future, it is recommended that researchers take into account the problems that arise during the implementation of the model and constantly optimize it.

Author Contributions: Methodology, P.D.; Formal analysis, B.Z.; Investigation, K.L.; Resources, Q.X.; Writing—original draft, M.F.; Visualization, J.C. All authors have read and agreed to the published version of the manuscript.

Funding: This research received no external funding.

Data Availability Statement: Data are contained within the article.

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

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