

Article

# Application of Demand Response Strategy in Optimal Configuration of a Standalone Wind-Solar-Battery System

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**Abstract:** This study presents application of demand response strategy in a standalone wind-solar-battery hybrid energy system (HES). Inputs for the designed HES are wind speed, solar radiation, temperature and load demand which is variable with time. In this study, hourly values of meteorological data and hourly load demand are considered in one year. An improved gravitational search algorithm (IGSA) is used to optimize the configuration of the standalone wind-solar-battery hybrid power system. The optimal objectives of the system are cost of the system in life cycle, the loss of power supply probability (LPSP) and the energy excess percentage (EXC). The effect of demand response on economic benefit and energy storage allocation of the standalone wind-solar-battery system is studied. The obtained optimal configuration of the proposed HES can provide minimal energy cost with excellent performance and reduced waste and unmet load.

**Keywords:** multi-objective optimization; optimal configuration; improved gravitational search algorithm (IGSA); wind-solar-battery system; demand response

## 1. Introduction

Energy is a vital factor for social and economic development of any country. Nowadays 80% of worldwide energy demand is met by means of fossil fuels. Therefore, researches on the development of renewable energy have generated wide spread attention recently [1-2]. Renewable energy resources have enormous potential and can meet the present world energy. They can enhance diversity in energy supply markets, secure long-term sustainable energy supply, and reduce local and global atmospheric emissions [3]. Wind and solar energy is considered the most preferred renewable energy source, and they are widely used to supply load in remote areas and non-grid areas such as independent islands [4-5]. Combining these two types of energies effectively to create a hybrid energy system can not only improve the power supply reliability but can also make full use of a variety of clean energies, while reducing energy wastage [6]. In general, multi-source hybrid energy systems have a higher quality and greater reliability than a system comprising of a single resource [7-8]. Koutroulis et al. [5] proposed a genetic algorithm (GA) based on cost minimization of a stand-alone hybrid solar-wind system in order to determine the optimal number and type of each energy source. B. Tudu et al. [9] presented a novel methodology for strategic selection of different components of PV-wind hybrid system.

In recent years, a battery energy storage system (BESS) is used in various aspects of hybrid power system. The main challenge in integrating a BESS into a hybrid power system is to evaluate an

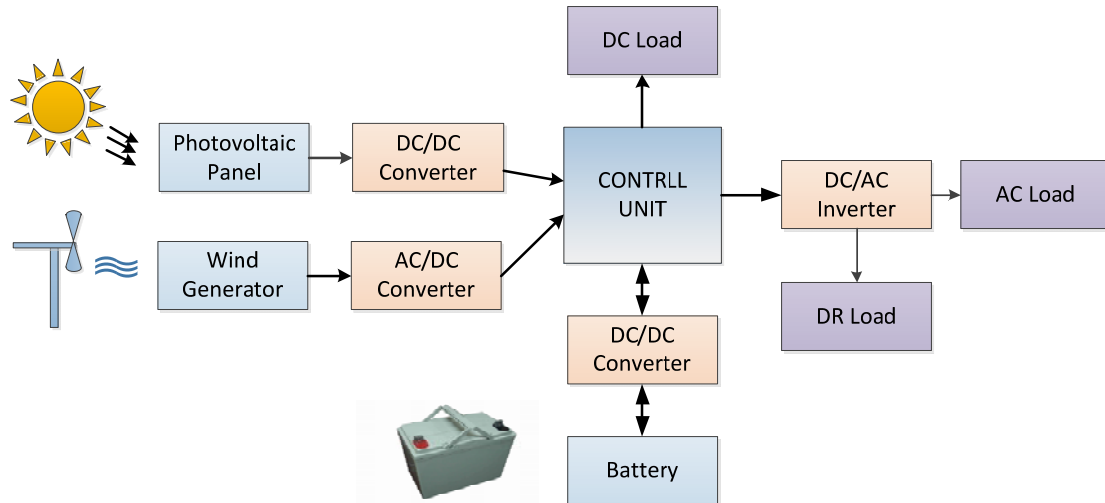
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optimum size of BESS to prevent the hybrid power system from instability and system collapse [10]. Daniel Akinyele et al. [11] presented an overview of batteries and compared their technical properties such as the cycle life, power and energy densities, and the costs. They proposed an optimal battery technology sizing and selection strategy, then assessed the environmental impact of batteries in efficiencies a typical renewable energy application by using a stand-alone PV system as a case study. In [12], a method for optimal sizing and operation of a BESS is proposed for spinning reserve in a small isolated power system. Obviously, considering BESS into a hybrid power system can overcome the stochastic availability and the unpredictable availability of renewable energy especially in the state of off-grid model. In standalone wind-solar-storage hybrid power system, Wang and Nehrir [13] proposed a standalone hybrid alternative energy system consisting of wind, PV, FC (fuel cell), electrolyzer and BAT. Wind and PV were the primary power sources of the system to take full advantage of renewable energy, and the FC-electrolyzer combination is used as a backup and a long-term storage system. A BAT bank was also used in the system for short-time backup to supply transient power. D.Suchitra et al. [8] proposed an optimally designed multi-source standalone hybrid generation system comprising of photovoltaic panels, wind turbine generators, batteries and diesel generator. Rodolfo Dufo-Lopez et al. [14] showed a multi-objective evolutionary algorithm (MOEA) for the optimization of standalone (off-grid) hybrid systems (photovoltaic-wind-diesel-battery) to minimize total net present cost (NPC) and maximize human development index (HDI) and job creation (JC). However, there are several factors that can give more consideration, the factors of storage capacity allocation, environmental problems, the full use of the complementary characteristics of wind and solar, the total investment of the system.

Further, demand response (DR) plays a key role for optimum asset utilization and to avoid or delay the need of new infrastructure investment [15]. DR programs are increasingly being deployed for various grid support functionalities, including demand shifting, power balancing, voltage/frequency control, and ancillary services [16]. M.A.Zehir et al. [17] proposed a novel retail demand response program in which the incentives are dependent on the aggregated performance of participants. Ugonna R. et al. [18] presented a method of evaluating consumer energy consumption behaviors whereby such information can be useful in designing grid network with improved active user participation in DR programs. However, the existing demand response is mainly based on the grid-connected, and less to explore the optimal configuration of stand-alone hybrid power system. This paper combines the DR with a stand-alone wind-solar-battery hybrid power system, and considers the cost of the system, the loss of power supply probability (LPSP) and the energy excess percentage (EXC), presents a method for a multi-objective optimal configuration to balance the power generation and load consumption using an improved gravitational search algorithm (IGSA). Simultaneously, this paper combines BESS and DR to carry out the comprehensive resource planning and adopts a case study to demonstrate the effectiveness of the optimization algorithm.

## 2. Architecture of the System

The proposed standalone wind-solar-battery hybrid power system consists of wind turbine generator, photovoltaic panel and battery as energy storage, shown in Fig1. The wind generator and photovoltaic cell are the generators of the system, the batteries are used for energy storage and power supply. The control unit decides which energy source will supply the performance of the hybrid energy system. In order to improve the reliability of the power supply, this paper considers the demand response. The effective modeling of each source and component is essential for overall effectiveness of this system in terms of power generation, utilization and power consumption.



**Fig.1** Structure of the standalone wind-solar-battery system

## 2.1 Modeling of wind turbine generator

Power output of wind turbine is calculated according to local weather conditions. At the particular wind speed, power extracted at time  $t$  is calculated as follows [19].

$$P_w = \begin{cases} 0 & v_w < v_{ci} \\ P_{Rw} \left( \frac{v_w - v_{ci}}{v_r - v_{ci}} \right)^n N_w v_{ci} & v_{ci} \leq v_w < v_r \\ P_{Rw} N_w & v_r \leq v_w < v_r \\ 0 & v_w \geq v_{co} \end{cases} \quad (1)$$

Where  $P_{Rw}$  is rated power of wind turbine (kW),  $V_w$  is wind speed (m/s),  $V_{ci}$  is cut-in speed of wind turbine (m/s),  $V_r$  is rated speed of wind turbine (m/s),  $V_{co}$  is cut-off speed of wind turbine (m/s) and  $N_w$  is the number of wind turbines. The value of  $n$  varies from 1 to 3 and in this paper is 1[20].

The rated power of wind turbine is calculated by rated wind speed ( $V_r$ ), swept area of rotor ( $A$ ), air density ( $\rho$ ), power coefficient ( $C_p$ ) and the efficiency ( $\eta_w$ ) associated with the energy conservation system and is given by (2).

$$P_{Rw} = \frac{1}{2} \rho A C_p \eta_w v_r^3 \quad (2)$$

The value of  $C_p$  for rated power output is determined by wind turbine curve.

## 2.2 Modeling of PV panel

The mathematical model of PV cell is approximately calculated, and the estimated solar radiation intensity in a certain time period (1h) can be simulated by a beta distribution [21]. Its expression is:

$$P_{pv}(t) = P_{STC} \frac{G_e(t)}{G_{STC}} [1 + k(T_c(t) - T_{STC})] \quad (3)$$

It is difficult to measure the actual temperature of PV panel. But it can be estimated by empirical formula testing the ambient temperature, expressed as:

$$T_c(t) = T_E(t) + 30 \times \frac{G(t)}{1000} \quad (4)$$

Where  $P_{pv}(t)$  is power output of photovoltaic cell at time  $t$ ,  $P_{STC}$  is power output of PV panels in standard test condition (STC);  $G_{STC}$  is intensity of illumination in STC,  $1000 \text{ W/m}^2$ ;  $T_{STC}$  is reference temperature in STC,  $25^\circ\text{C}$ ;  $k$  is temperature correction coefficient,  $-0.0047/^\circ\text{C}$ ;  $G_e(t)$  is actual radiation intensity;  $T_c(t)$  is actual ambient temperature;  $T_E(t)$  is ambient temperature.

### 2.3 Modeling of Battery

State of charge (SOC) of battery (indicates the amount of electricity remaining after the battery is used [22]). SOC of battery can be obtained by the method of electricity accumulation.

When the battery is discharged, the SOC at time  $t$  is:

$$SOC(t) = (1 - \eta)SOC(t - 1) - \frac{P_{bd}(t)\Delta t}{\eta_{bd}E_r} \quad (5)$$

When the battery is charged, the SOC at time  $t$  is:

$$SOC(t) = (1 - \eta)SOC(t - 1) - \frac{P_{bc}(t)\eta_{bc}\Delta t}{E_r} \quad (6)$$

Where  $\eta$ ,  $\eta_{bc}$ ,  $\eta_{bd}$ ,  $P_{bc}(t)$ ,  $P_{bd}(t)$  and  $E_r$  are self discharge rate of storage battery, charging efficiency, discharging efficiency, charging power, discharge power and rated capacity respectively;  $\Delta t$  is sampling step, here's 1.

### 2.4 Modeling of demand response load

When electricity wholesale market prices or system reliability is threatened, power users receive the power supply party leading to reduce load of direct compensation notice or signal after the electricity price rise, change the habits of its inherent power mode, to reduce or goes on a certain period of time of power load and power supply response and to ensure the stable, and curb short-term behavior of electricity price rise. With the gradual improvement of active distribution network construction, the demand response will gradually become a normal in the future.

This paper presents a strategy for a multi-objective optimal configuration according to demand response of electricity compensation. The calculation demand response compensation cost is shown in formula (7).

$$C_c = \sum_{i=1}^{N_c} m_c \cdot P_{c,i} \cdot T_{c,i} \quad (7)$$

Where  $C_c$  is demand response compensation cost in one day;  $m_c$  is compensation price of unit electricity;  $P_{c,i}$  and  $T_{c,i}$  are load size and response time of the  $i$  response load respectively;  $N_c$  is the number of response users.

Analyzing the time behavior of demand response load, adding the demand response capacity, response times in the period and continuous response time to the contracts can improve the binding force and execution of the contract, further explore the potential of demand response.

(1) Demand response capacity constraint

$$0 \leq P_{d,i} \leq P_{d,i,max} \quad (8)$$

Where  $P_{d,i,max}$  is the maximum capacity of response users.

(2) Demand response times constraint.

$$0 \leq n_{d,i} \leq n_{d,i,max} \quad (9)$$

Where  $n_{d,i}$  and  $n_{d,i,max}$  are response times and the maximum response times.

(3) Response time constraint of continuous demand

$$T_{d,i,min} \leq T_{d,i,t} \leq T_{d,i,max} \quad (10)$$

Where  $T_{d,i,t}$ ,  $T_{d,i,min}$  and  $T_{d,i,max}$  are response time, the minimum response time and the maximum response time respectively.

### 3. Planning model

#### 3.1 Objective function

(1) Total system investment

(a) Equipment cost

The cost of purchasing wind turbines, photovoltaic panels, and energy storage batteries is called the initial cost, it's also the biggest, simplest, direct part of the cost, expressed as:

$$C_0 = \sum_{i=1}^3 N_i C_i \quad (11)$$

Where  $i$  represents different power types, that is, wind turbines, photovoltaic panels and batteries;  $N_i$  and  $C_i$  represent the number and unit price of each power supply.

(b) Maintenance cost

In the whole process of wind-solar-battery hybrid power system, it is necessary to maintain the power generation unit so as to ensure the normal operation of wind turbine/photovoltaic panel and the energy storage battery, the expression is:

$$C_m = \sum_{t=1}^{8760} \sum_{i=1}^3 K_i N_i P_i(t) \Delta t \quad (12)$$

Where  $t$  represents sampling time;  $K_i$  is maintenance cost factor for each power supply, rmb/kWh;  $P_i(t)$  is operating power of each supply at time  $t$ , kW;  $\Delta t$  represents sampling time period.

(c) Generation subsidy

Due to the wind-solar-batter hybrid power system for new energy generation, it has not become popular as conventional power generation such as firepower, and the construction cost is higher, resulting in the sale price is higher than the conventional power generation price. In order to encourage clean energy generation, the government subsidies for wind-solar power generation, generation subsidy is expressed as:

$$C_s = \sum_{t=1}^{8760} \sum_{j=1}^2 K_{s_j} N_j P_j(t) \Delta t \quad (13)$$

Where,  $K_{s_j}$  is subsidy price for power generation of renewable energy.

(d) Demand response compensation

$$C_d = K_{l,y} C_c \quad (14)$$

Where,  $K_{l,y}$  is conversion coefficient.

The total system investment of wind-solar-battery hybrid power generation system is:

$$C = C_0 + C_m - C_s - C_d \quad (15)$$

(2) LPSP

It often happens that wind-solar power generation is less than the power required by the load in wind- solar-battery hybrid power generation system, moreover, the SOC of the battery is less than the limit value (usually 0.1) and cannot be discharged. In this period, the proportion of the system cannot meet the load power demand is called the loss of power supply probability, expressed in LPSP [23], the formula is (16).

$$LPSP = \frac{\sum_{i=1}^m (P_l(t) - \sum_{i=1}^4 P_i(t))}{\sum_{t=1}^{8760} P_l(t)} \quad (16)$$

Where,  $P_l(t)$  is demand load at time  $t$ .

(3) EXC

If wind-solar power generation is more than the use of all power loads and the battery can no longer continue to be charged, the excess energy will be wasted. In a given time, the ratio of the energy wasted in the system to the total load is called the energy wastage rate, expressed in EXC, which is usually between 5% and 30% in the project.

$$EXC = \frac{\sum_{l=1}^m (\sum_{t=1}^4 P_l(t) - P_l(t))}{\sum_{t=1}^{8760} P_l(t)} \quad (17)$$

The final optimization objective function is Eq.(18).

$$\begin{cases} \min C = C_0 + C_m - C_s - C_d \\ \min LE = \theta_1 LPSP + \theta_2 EXC \end{cases} \quad (18)$$

Where  $\theta_1$  and  $\theta_2$  are weight vector, the values depend on the importance of the target, it is generally decided by decision makers.

### 3.2 Constraint condition

(1) Constraint of power balance

$$P_w + P_{pv} + P_b \geq P_l - P_d \quad (19)$$

Where  $P_w$ ,  $P_{pv}$ ,  $P_b$  represent power output of wind generation, photovoltaic generation and battery,  $P_l$  is load power,  $P_d$  is DR load.

(2) Constraint of generator power output

$$P_{min} \leq P(t) \leq P_{max} \quad (20)$$

The constraint of generator power output is mainly considered to prolong the working life of power supply,  $P_{min}$ ,  $P_{max}$  are upper and lower limits of generator power output.

(3) Capacity constraint of energy storage battery

$$SOC_{min} \leq SOC(t) \leq SOC_{max} \quad (21)$$

## 4. Improved Gravitational Search Algorithm

Gravitational search algorithm (GSA) is proposed by Professor Esmat Rashedi of Iran [24]. The algorithm is a meta heuristic intelligent optimization algorithm based on Newton's law of universal gravitation. The principle of the algorithm is that the search particles are considered as a group of objects running in space. Objects are attracted to each other by gravitational attraction, and the motion of objects follows the law of dynamics. The greater the fitness value is, the larger inertial mass of the particle is, the gravity causes objects to move toward the greatest mass, gradually approaching the optimal solution of the optimization problem.

Supposes a D-dimensional search space which includes N objects, the position of the  $i$  object is expressed as:

$$X_i = (x_i^1, x_i^2, \dots, x_i^k, \dots, x_i^d) \quad (22)$$

At time  $t$ , the gravitational force of object  $j$  from object  $i$  in the  $k$ -dimension is expressed as:

$$F_{i,j}^k(t) = G(t) \frac{M_{p,i}(t) \times M_{a,j}(t)}{R_{i,j}(t) + \varepsilon} (x_j^k(t) - x_i^k(t)) \quad (23)$$

Where  $\varepsilon$  is a very small constant;  $M_{a,j}(t)$  is inertial mass of object  $j$ ;  $G(t)$  is gravitational constant.

$$G(t) = G_0 \times e^{-\alpha t/T} \quad (24)$$

$T$  is maximum number of iterations.

$R_{i,j}(t)$  is Euclidean distance between objects  $i$  and objects  $j$ , expressed as:

$$R_{i,j}(t) = \|X_i(t), X_j(t)\|_2 \quad (25)$$

Therefore, the sum of the forces acting on the  $X_i$  is equal to the sum of all the other objects forces act on it, the force is calculated as:

$$F_i^k(t) = \sum_{j=1, j \neq i}^N \text{rand}_j F_{i,j}^k(t) \quad (26)$$

In GSA, each particle updates its speed and position at each iteration as Eq.(27).

$$\begin{cases} v_i^d(t+1) = rand \times v_i^d(t) + a_i^d(t) \\ x_i^d(t+1) = x_i^d(t) + v_i^d(t+1) \\ a_i^d = F_i^d(t)/M_i(t) \end{cases} \quad (27)$$

Although compared with PSO and GA algorithm, GSA has fast convergence speed and strong global search ability, it also may fall into local solution. In this paper, parameters are changed in different stages of the optimization to improve convergence of GSA. The change of parameters not only enhances the global searching ability of the particle, but also accelerates the convergence of the algorithm. It also enhances the local search capability of the particle, improves the accuracy of the solution, and prevents the algorithm from falling into the local solution. At the same time, the gravity search algorithm also introduces the memory and group communication of particles in particle swarm algorithm to improve the effectiveness of the algorithm. The new speed and position update formula are defined as Eqs.(28)(29).

$$v_i^d(t+1) = rand_i v_i^d(t) + a_i^d(t) + c_1 rand_j \times (p_{best}^d - x_i^d(t)) + c_2 rand_k \times (g_{best}^d - x_i^d(t)) \quad (28)$$

$$X_i^{t+1} = X_i^t + rand \times V_i^t + G_0 \times e^{-at/T} \times \sum_{j=1}^N \left( \frac{M_j^t}{R_{ij}} (rand \times (X_j^t - X_i^t)) \right) \quad (29)$$

$rand_i$ 、 $rand_j$ 、 $rand_k$ 、 and  $rand$  represent random variable between 0 and 1.

## 5. Demand response strategy

Demand response programs coordinating large numbers of consumers can play an important role in electricity system. Demand shifting is to make the load demand maximize close to the wind-solar power curve in time sequence. Intelligent system installed on the user side can be used to receive the incentive or price signal, trigger preprogrammed specified response strategy according to different users' location, energy power demand and price level, and automatically adjust the load time of electricity, for example, the electricity time will shift from peak to nonpeak period. The expression as follows:

$$\begin{cases} \min \sum_{i=1}^T |L(t) - P_{pv}(t) - P_w(t)| \\ L(t) = L_{befo}(t) + L_{SLin}(t) - L_{SLout}(t) \end{cases} \quad (30)$$

Where  $T$  is scheduling period;  $P_{pv}(t)$  is photovoltaic power generation at time  $t$ ,  $P_w(t)$  is wind power generation at time  $t$ ;  $L(t)$ 、 $L_{befo}(t)$ 、 $L_{SLin}(t)$  and  $L_{SLout}(t)$  are post-demand response capacity、pre-demand response capacity、transferred to the load and roll out load.

Based on the analysis of wind generation power output characteristics, load characteristics and energy storage system characteristics, the typical daily charge and discharge strategy of energy storage system in wind-solar hybrid power system can be summarized as follows:1) If the wind-solar power output is greater than the required load, the remaining energy is charged to battery until the battery is filled and the SOC reaches the upper limit, the charging is stopped; 2) If wind-solar power output is less than the required load, the demand response start and the discharge stop when the SOC reaches the lower limit .

## 6. Case study

### 6.1 Raw data

In order to verify the rationality and superiority of the proposed optimal configuration scheme, this paper researches on a systems through one year with 1-h time steps. In the paper, the system life is

regarded as 20 years, the compensation price of DR is 0.1 rmb/kWh, the initial state of charge is 0.5 and the range of SOC is [0.1, 0.9]. The power output curve of the photovoltaic panel and wind turbine generation are shown in Fig.2-3. The distribution of the consumer power requirements is shown in Fig.4. Single unit specification of components wind turbine, PV panel and battery are presented in Table 1. In optimization algorithm, the number of population is 50, the number of generation is 200,  $G_0$  is 100 and  $\alpha$  is 20.

Table 1.Component specifications

Wind turbine generator	PV panel	Battery
model:FD21-100	model: DJB-36V250WK	model:BT-MSE-2000
rated power:100kW	rated power:0.25kW	rated capacity:2000Ah
cut-in speed:3m/s	open-circuit voltage:37.7V	rated power:2V
rated power wind speed:13m/s	open-circuit current:8.79A	minimum SOC:10%
cut-out speed:25m/s	size: 1650*992*45mm	maximum SOC:90%
extreme wind speed:59.9m/s	conversion efficiency:18%	charge discharge efficiency:90%
investment cost: 80000rmb/unit	investment cost: 1120rmb/unit	investment cost:3200rmb/unit
yearly O&M cost: 1.87rmb/kW	yearly O&M cost: 5.43rmb/kW	yearly O&M cost: 4rmb/unit

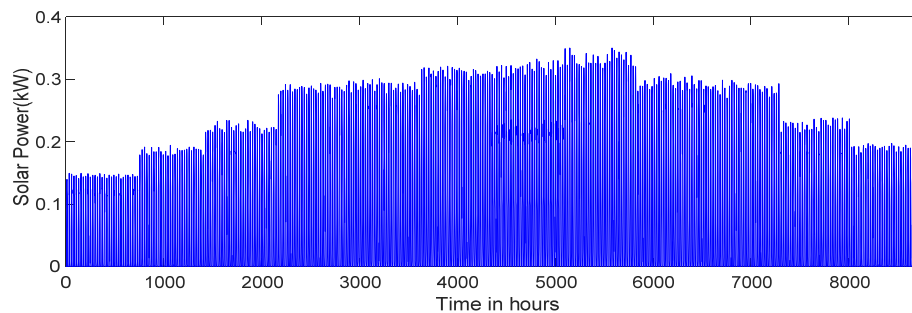


Fig2. Photovoltaic panel power output curve in one year

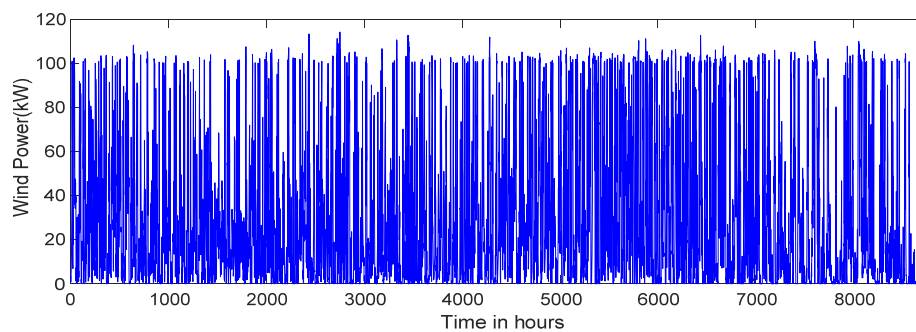


Fig3. Wind turbine generation power output curve in one year

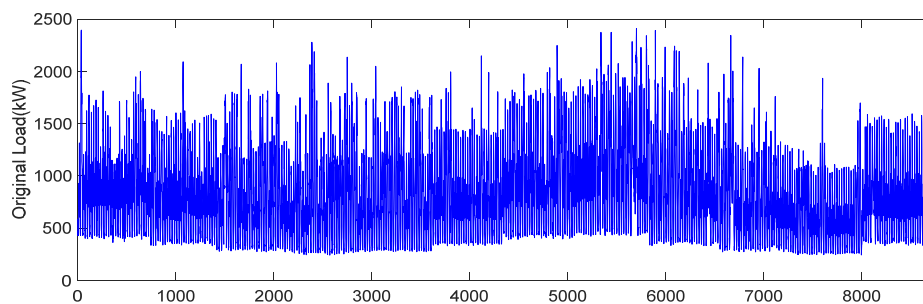


Fig4. Load curve in one year



## 6.2 Economic analysis of configuration results

IGSA is used to optimize the multi-objective system according to three schemes. Scheme 1: Planning the system without DR. Scheme 2: Based on the optimized configuration of the scheme 1, considering the demand response with different participation levels of the total load demand. Scheme 3: Optimal configuration of the system with different participation levels of the total load demand. Comparison of the optimized results and related indicators in three different schemes are shown in Table 2. The optimal configuration annual cost is shown in Table 3.

Table 2 Results of the system configuration

scheme	Different percentage DR	Number of wind turbine	Number of PV panel	Number of Battery	DR capacity (MWh)	LPSP/%	EXC/%
1	0%	15	5220	15417	0	5.74	6.42
2	5%	15	5220	15417	364.536	5.24	5.90
	10%	15	5220	15417	729.071	4.74	5.37
	15%	15	5220	15417	1093.607	4.24	4.84
	20%	15	5220	15417	1458.143	3.74	4.32
	3	5%	15	5220	13000	364.536	5.38
3	10%	15	5220	11076	729.071	5.00	5.54
	15%	15	5220	9867	1093.607	4.57	5.06
	20%	15	5220	8793	1458.143	4.13	4.58

Table 3. Yearly cost of the system

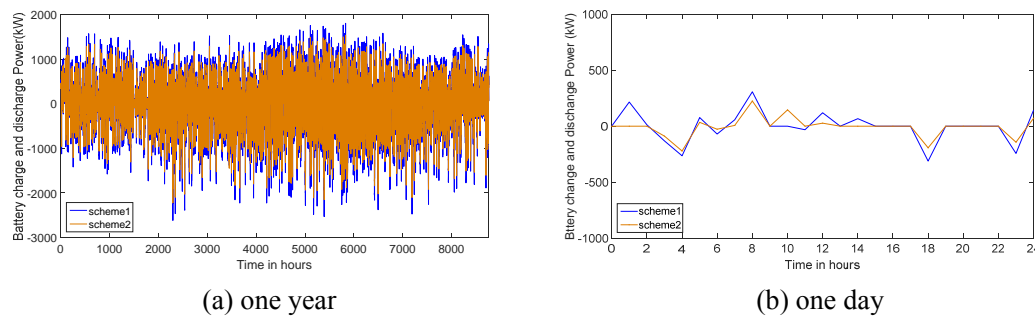
(unit: Millions rmb)

scheme	Different percentage DR	Annual investment cost	Annual maintenance cost	DR compensation cost	total investment
1	0%	8.2925	25.7987	0	30.6810
2	5%	8.2925	25.7987	0.03645	30.6445
	10%	8.2925	25.7987	0.07291	30.6080
	15%	8.2925	25.7987	0.1094	30.5716
	20%	8.2925	25.7987	0.1458	30.5351
	3	5%	7.1323	25.7890	0.03645
3	10%	6.2088	25.7813	0.07291	28.5070
	15%	5.6285	25.7765	0.1094	27.8854
	20%	5.1130	25.7722	0.1458	27.3292

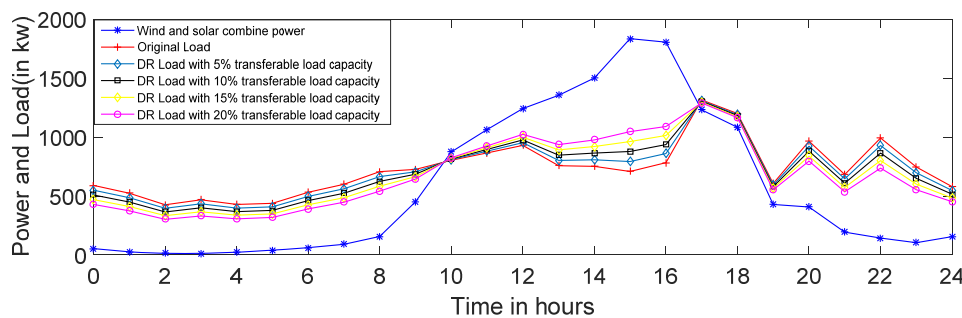
According to the configuration results in Table 2 and Table 3, the consequence is effective. Comparison the Scheme 1 and the Scheme 2, in the same capacity configuration, with the increasing of participation levels of the total load demand, the user's electrical behavior follows the power generation behavior, in the case that the number of batteries remains unchanged, the battery charge and discharge depth are reduced, and the charge and discharge times are shifted as shown in Fig.5. It is beneficial to prolong battery life. Further, the demand response improves the power grid operation effect obviously, and the load shifting effectively improved the load characteristics and fully responds to renewable power generation, load demand curve after DR with different participation levels is shown in Fig.6, it is helpful for the safe and reliable operation of the whole system. In addition, according to the price compensation mechanism established in this paper, with the increase of participation levels of the total load demand, the DR capacity and DR compensation cost are increased, configuring the demand

response is cost-effective for reducing the overall cost of the system.

Comparison the results of Scheme 1 and Scheme 3, with the increase of participation levels of the total load demand, the DR load is much closer to the power generation, and there is less demand for batteries, which directly leads to lower total costs. In addition, the EXC and LPSP in Scheme 3 are lower than that in Scheme 1. Due to the decrease in the number of batteries, the EXC and LPSP in Scheme 3 are higher than that in Scheme 2. In summary, demand response can effectively reduce the storage capacity of the standalone wind-solar-battery system and reduce the energy storage system maintenance, at the same time, economy and reliability of the system are improved.



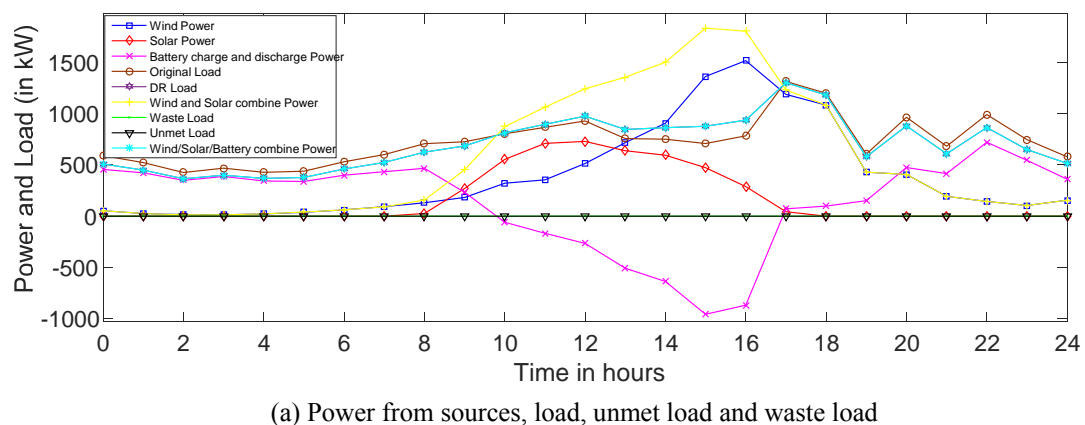
**Fig5.** Battery charge and discharge power as the DR percentage is 10%



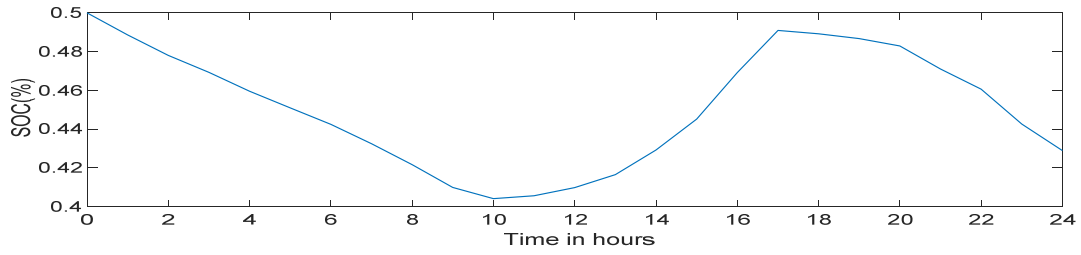
**Fig6.** Load demand curve after DR with different participation level

### 6.3 Analysis of battery state

Take 10% of the participation level of the total load demand of Scheme 3 as an example, the convergence characteristic of the proposed project based on IGSA for the solution of objective function is effective. The performance analysis of this standalone wind-solar-battery hybrid power system for obtained optimal configuration is done for one year and the results are shown in Fig.7-9. Fig.7 shows a typical day without waste and lack of electricity. Fig. 8 shows the curve of a typical day with lack of electricity and Fig. 9 is a typical day with waste of electricity.

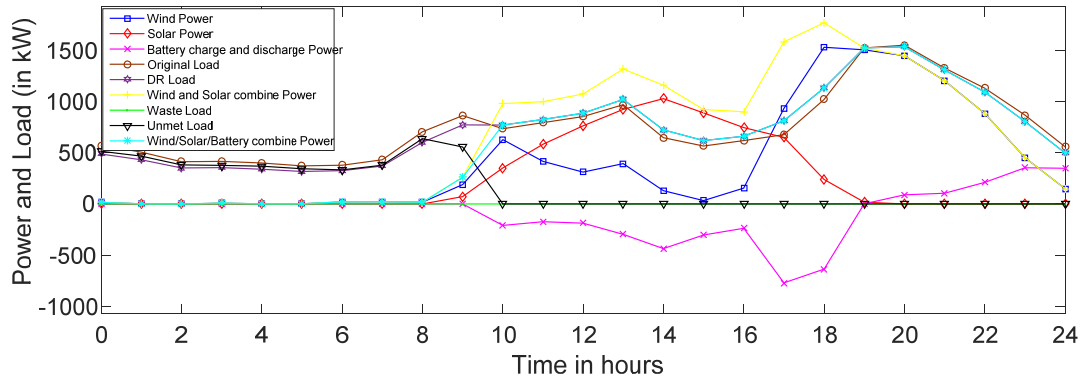


(a) Power from sources, load, unmet load and waste load

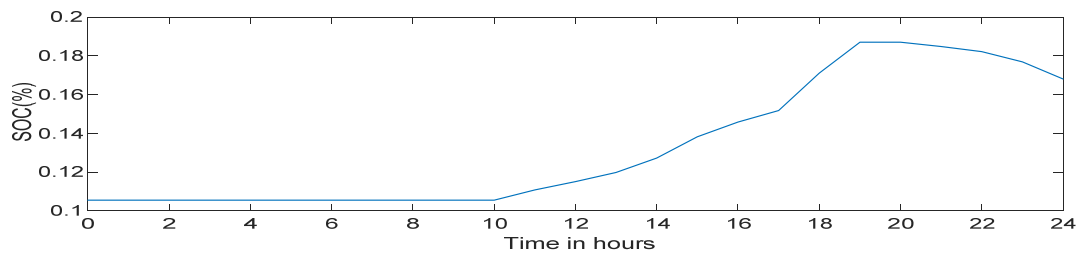


(b) SOC of battery

**Fig7.** Without power waste and lack of electricity in a typical day

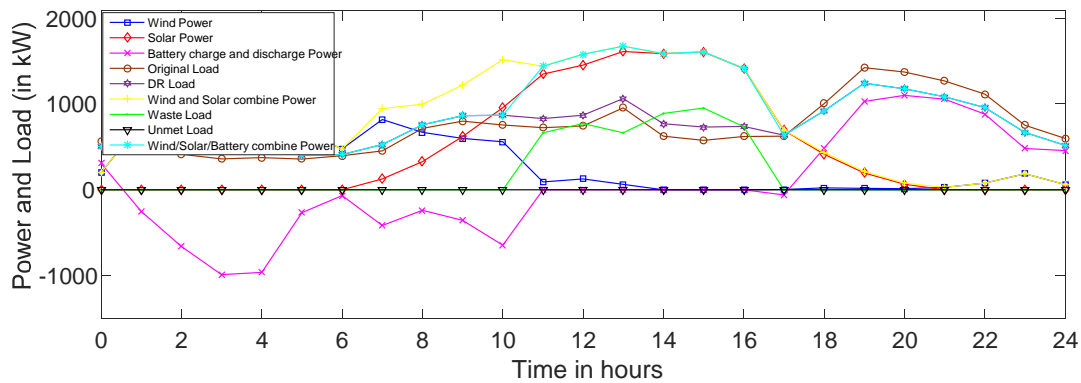


(a) Power from sources, load, unmet load and waste load

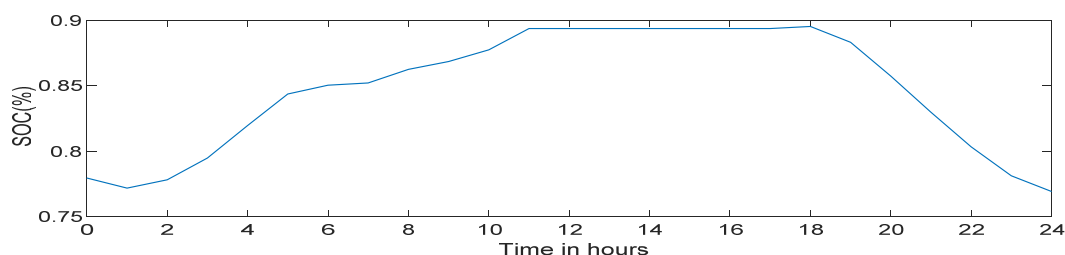


(b)SOC of battery

**Fig8.** Lack of electricity in a typical day



(a) Power from sources, load, unmet load and waste load



(b) SOC of battery

**Fig9.** Waste of electricity in a typical day

It can be depicted that the standalone wind-solar-battery system are able to meet the user load completely as shown in Fig.7 (a). The load demand is higher than wind-solar power during 0:00-10:00 and the battery discharge power is sufficient to meet the load at the time as shown in Fig.7 (b). During 10:00-16:00, the wind-solar power output is higher than the load, excess power is charged to the battery to avoid waste. After 18:00, the wind-solar power output reduces and cannot meet the load demand, the battery keeps discharging to maintain the system stability.

In the typical day with lack of electricity, the wind-solar power cannot meet the daily load, and the battery's discharge cannot meet the daily lack of electricity, the SOC is in a low state as shown in Fig.8 (b). For the period from 0:00 to 10:00, the battery stop charging and there is an unmet load as shown in Fig.8 (a). During 10:00-24:00, the system began to resume stable operation. In the typical day with waste of electricity, the renewable energy is plentiful and the daily load is smooth as shown in Fig.9 (a). For the period from 1:00 to 11:00, the wind-solar power is higher than the load and the battery keeps charging to balance the system. After 11:00, the battery is fully charged as shown in Fig.9 (b) and the excess power cannot be charged to the battery, this causes waste during 11:00-18:00. After 18:00, the wind-solar power output is lower than the load and the battery discharge to maintain the continuity of power supply.

The total wind-solar-battery power generation with DR can basically maintain stable operation of the system, except in some periods at noon time and night time when wind-solar power is not available. Analysis of the results shows that the configured system can meet the normal operation of remote area or isolated island.

## 7. Conclusions

This paper proposed an application of demand response strategy to exploit the benefits of DR in the optimal configuration of a standalone wind-solar-battery system. A case study based on IGSA has been developed on MATLAB platform for costing, sizing and performance analysis of a standalone HES. Based on the optimal configuration results, the following can be obtained: (1) Demand response strategy proposed in this paper improves the load characteristic effectively, which can fully response wind and solar power generation characteristics, improve the system reliability and economy, make the most of the energy storage. (2) Considering the demand response strategy, it can effectively reduce the storage capacity allocation, reduce the initial investment cost of energy storage, improve the operating reliability of the system, and the system economic benefits can be further improved. (3) With the four different participation levels of the total load demand were considered, the overall performance of the system has been improved. From the overall result analysis, it is clear that the proposed strategy is able to bring the benefits to both utility and generation sides. With the higher of participation level, the economy of the system is higher.

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