# HOME ENERGY MANAGEMENT SYSTEM BASED ON IMPROVED GENETIC ALGORITHM AND MULTI-OBJECTIVE FUNCTION

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# ABSTRACT

In order to realize the economic operation of the home energy management, the electricity cost of user needs to be reduced and user-side satisfaction should be improved. A home energy scheduling optimization model including the home electricity load model and the photovoltaic power generation model was constructed, the objective function was established, which takes minimizing the electricity cost and the user satisfaction into consideration. With the constraints of power balance, load working time and temperature control load. Finally, combining genetic algorithm with local optimization method, a hybrid Pareto genetic algorithm for multi-objective optimization is proposed to obtain the approximate solution of multi-objective function. Based on the simulation results of MATLAB platform, the optimal solution which both minimizes electricity cost and maximizes user's satisfaction has been obtained.

**Keywords:** House energy management, Genetic algorithm, Multi-objective function, Optimization

# 1. INTRODUCTION

In recent years, society has put forward an increasingly high demand for power supply stability and power quality reliability, which has brought many new problems to the development of the power grid [1]. As the development of smart grids and the widespread application of renewable energy such as photovoltaics, the Home Energy Management System (HEMS) is constructed and its energy scheduling is optimized which make the electricity consumption of user-side more reasonable for the house and valid improve their electricity efficiency. The literature[2] classifies the electricity load in detail according to the residential electricity consumption cost and load peak-valley difference but does not consider the satisfaction of electricity users. In literature[3], the household load model established by using renewable energy and storage battery is relatively simple, and the relationship between user satisfaction and user power cost is not fully considered.

To sum up, this paper proposes a home energy management scheduling model that gives consideration to household load classification and photovoltaic power generation while fully considering the factors of minimizing electricity cost and maximizing customer satisfaction.

Since the genetic algorithm was first applied to multiobjective optimization in 1985, the research in this field has been developing continuously and various improved genetic multi-objective optimization algorithms have been proposed[4]. The sub-population co-evolution method is used to improve the efficiency of finding the most solution set, but this method has the disadvantage that the common search of multiple groups makes the search efficiency slow[5]. Genetic algorithm (GA) performs the evolutionary operation on the whole population which focuses on the collection of individuals and could obtain a large number of Pareto optimal solutions corresponding to different weight distribution situations at one time. Although it has stronger applicability and flexibility, it has disadvantages such as poor local search ability and precocious convergence. In order to enhance the local search ability of GA, a hybrid Pareto genetic algorithm (HPGA) for multi-objective optimization was proposed by combining GA with direct search method[6].

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# 2. BASIC PORTFOLIO MODEL OF HOME ENERGY MANAGEMENT

## 2.1 Photovoltaic cell model

The power generation of photovoltaic cells is affected by many factors. The output power equation of the photovoltaic panel is shown as follow[7]:

$$P_{PV}^{t} = n_{PV} \eta_{PV} S_{PV} I_{PV} \left[ 1 - 0.005 \left( T_{out}^{t} - T_{stc} \right) \right]$$
(1)

where  $p_{PV}^{t}$  is the photovoltaic panel output power;  $n_{PV}$ is the number of photovoltaic panels;  $\eta_{PV}$  is the photoelectric conversion efficiency of photovoltaic cells (%);  $S_{PV}$  is the area of the photovoltaic panel (m<sup>2</sup>);  $I_{PV}$  is the solar radiation intensity (kW/m<sup>2</sup>);  $T_{out}^{t}$  is outdoor temperature (°C), ambient temperature  $T_{str} = 25$  °C.

## 2.2 Load model

Due to a large number of household users, the types of the electricity load of each household are quite different. In order to make the home energy management system have wide applicability, it is necessary to classify the household electricity load according to the characteristics of the electricity load, the demand of the use of the load, the operation characteristics of the load, and other factors.

# 2.2.1 Necessary load

The necessary class load usually refers to the load that is very important to the user. In the home energy management system, the electricity demand has the highest priority which should be considered at first. The working model can be expressed as [3]:

$$\begin{cases} P_i^c = x_i^c P^{c, \text{ rated}} \\ E^c = P^{c, \text{ rated}} d^c \end{cases}$$
(2)

where  $c \in C$  is the necessary load,  $P_i^c$  is the working power of the necessary class load c at time  $i^{th}$ .  $X_i^c$  is the working state of the necessary class load c at time  $i^{th}$ .  $P^{c, rated}$  is rated power of the necessary class load c.  $E^c$  is the total power consumption of the necessary class load  $c, d^c$  is working time of necessary class load c

# 2.2.2 Transferable load

The transferable class load has the characteristics of flexible use of electricity. The transfer of this kind of load type is realized on the basis of sacrificing certain user satisfaction. The working model can be expressed as<sup>[3]</sup>:

$$\begin{cases}
P_i^s = x_i^s P^{s, \text{ rated}} \\
E^s = P^{s, \text{ rated}} d^s \\
x_i^s = \begin{cases}
1, i \in H_s \\
0, i \notin H_s & i \in R_s \\
s \in S
\end{cases} (3)$$

where  $s \in S$  is the transferable load,  $P_i^s$  is the working power of the transferable class load s at time *i*,  $x_i^s$  is the working state of transferable class load s at time *i*,  $P^{s, rated}$  rated power of transferable class load *s*,  $E^s$  is the total power consumption of transferable class load *s*,  $d^s$  is the working hours of transferable class load *s*,  $H_s$ working time interval of transferable load *s*,  $R_s$ applicable time interval for transferable load *s* 

## 2.2.3 Air conditioning load

The air conditioning system model is mainly composed of two parts which include the air conditioning room heat transfer model and the air conditioning model[3].

(1) Model of heat transfer in the air-conditioned room

$$\frac{dT_i^{r}(t)}{dt} + \frac{T_i^{r}(t)}{M_{air} \cdot c \cdot R_{eq}} = \frac{Q_i^{h}}{M_{ar} \cdot c} + \frac{T_i^{out}(t)}{M_{air} \cdot c \cdot R_{eq}}$$
(4)

where  $T^{r}(t)$  is the temperature at room time t,  $M_{air}$  is the quality of air in the room, c is the specific heat capacity of indoor air,  $R_{eq}$  equivalent thermal resistance of the room,  $Q^{h}$  air conditioning cooling capacity at time t,  $T^{out}(t)$  is the outdoor temperature in t time.

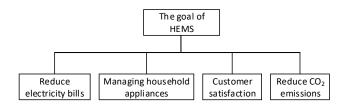
(2) Air conditioning model

When the set temperature is maintained, the relationship between the operating state of the compressor of the air conditioner and the set temperature can be expressed as:

$$G(t_{i}) = \begin{cases} 1 & T(t_{i}) \ge T_{set} + \varepsilon/2 \\ 0 & T(t_{i}) \le T_{set} - \varepsilon/2 \\ G(t_{i-1}) & T_{set} + \varepsilon/2 \ge T(t_{i}) \ge T_{set} - \varepsilon/2 \end{cases}$$
(5)

# 3. MULTI-OBJECTIVE OPTIMIZATION MODEL FOR HOME ENERGY MANAGEMENT

3.1 The goal of HEMS



#### Fig.1. The goal of HEMS

#### 3.1.1 Minimize user electricity costs

The electricity cost in the family is mainly composed of three aspects: the electricity fee to be paid from the power company, the income from selling electricity to the power company, and the state subsidy for photovoltaic power generation. The objective function is as follow[5]:

$$\min Cost = \sum_{i=1}^{24} \left[ C_1(t) - C_2(t) \right]$$
 (6)

$$C(t) = \begin{cases} P_b(t) \times P(t) \times T; & P(t) > 0\\ P_s(t) \times P(t) \times T; & P(t) \le 0 \end{cases}$$
(7)

$$C_2(t) = G(t) \times P_G \tag{8}$$

where  $C_1(t)$  is the electricity purchase cost of the user in  $t^{\text{th}}$  period,  $C_2(t)$  is the subsidy cost of the country for photovoltaic power generation during the  $t^{\text{th}}$  period, G(t) photovoltaic power generation for the  $t^{\text{th}}$  period,  $P_b(t)$  is the price of buying electricity for the  $t^{\text{th}}$  period,  $P_s(t)$  is the price of electricity sold for the  $t^{\text{th}}$  period,  $P_G$ subsidized price for photovoltaic power generation in the country.

#### 3.1.2 User satisfaction

When the HEMS is in operation, it is necessary to set the allowable operating time period  $[\alpha, \beta]$  and the impact level  $\delta$  of the power load for each power load in advance. The user satisfaction model can be expressed[5]:

$$f_{a}(t_{a}) = \begin{cases} \delta_{a} \times \frac{\beta_{a} - t_{a}}{\beta_{a} - (\alpha_{a} + l_{a})}, & \alpha_{a} + l_{a} \leq t_{a} \leq \beta_{a} \\ 0, & t_{a} < \alpha_{a} + l_{a} \text{ ort } _{a} > \beta_{a} \end{cases}$$
(9)

where  $f_a(t_a)$  is the satisfaction function of the electrical load a at the running completion time  $t_a$ ,  $\delta_a$  is the influence level of the electrical load a, when the electrical load a is finished at the end of time  $\alpha_a + I_a$ , the user satisfaction is the highest; when the running time is gradually moved backward  $\beta_a$ , the satisfaction will gradually decrease to zero. Therefore, the expression of user satisfaction:

$$S = \sum_{a=1}^{m+n+i} f_a(t_a)$$
(10)

3.1.3 Consider the objective function of user satisfaction and minimization of user electricity bills

User satisfaction is required in the model where the user's electricity bill is minimized, and the coefficient  $\sigma(\sigma \in [0,1])$  is used to adjust the proportion of user satisfaction in the objective function. The objective function expression for minimizing the user's electricity bill and taking into account user satisfaction is<sup>[5]</sup>:

$$\min C_{\text{cost}} - \sigma \cdot S \tag{11}$$

#### 3.2 Restrictions

According to the power flow characteristics of the power system, users' requirements for load work and load's own working characteristics, the power balance constraints, load working time constraints and temperature control load constraints are established.

3.2.1 Power balance constraint

The electrical energy produced in the system must be balanced with the electrical energy consumed at any time. The established power balance constraints are as follows[4]:

$$\sum_{i} \left( \sum_{j} W_{ij} \times A_{ij}(t) \right) + M_{i}(t) = G(t) + P(t)$$
 (12)

where  $W_{ij}$  is the *j*<sup>th</sup> load power on the *i*<sup>th</sup> line,  $A_{ij}(t)$  is the working state of the *j*<sup>th</sup> load on the ith line in the *t*<sup>th</sup> period,  $M_i(t)$  is the unschedulable load power on the *i*<sup>th</sup> line of the *t*<sup>th</sup> period, G(t) is photovoltaic power generation for the *t*<sup>th</sup> period, P(t) is the amount of electricity traded for the family and the grid during the *t*<sup>th</sup> period

# 3.2.2 Necessary load

The load working time includes two aspects. On the one hand, the user has requirements for the total duration of the load work, and on the other hand, the load's own working characteristics determine the shortest continuous working time.

(1) The constraints on the total duration of load work are as follows:

$$\sum_{t=1}^{24} A_{ij}(t) = N_{ij}$$
(13)

where  $N_{ij}$  is the total length of time required in the day of load  $j^{th}$  on line  $i^{th}$  set for user

(2) The shortest continuous working time constraint is expressed as follows:

$$A_{ij}(t+x) = 1, \forall x \in [0, L_{ij}], \quad B_{ij}(t) = 1$$
(14)

 $L_{ij}$  is the shortest continuous working time of the  $j^{th}$  load on the  $i^{th}$  line.  $B_{ij}(t)$  is the working state of the  $j^{th}$  load in the time period t on the  $i^{th}$  line

3.2.3 Temperature controlled load constraints

There is no working time constraint for temperaturecontrolled loads, only temperature constraints, which consider users satisfaction. The temperature constraints are as follows:

$$T_{\rm k}(t) = T_{\rm s}(t) \pm \Delta T \tag{15}$$

Where  $T_k(t)$  is the actual temperature value in the  $t^{th}$  period.  $T_s(t)$  is the optimal temperature for the  $t^{th}$  period set for the user,  $\Delta T$  is the temperature deviation allowed by the user.

# 4. SCHEDULING STRATEGY OF HOME ENERGY MANAGEMENT SYSTEM

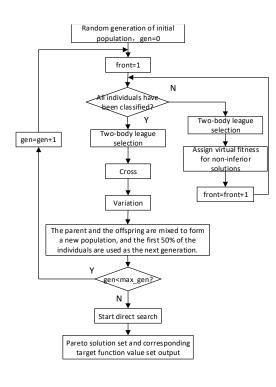
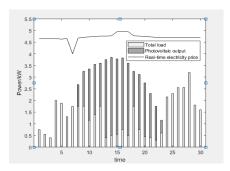


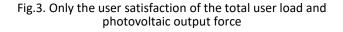
Fig.2. HPGA flow chart

4.2 Case studies

4.1 HPGA flow chart

This case studies takes the family in a certain area of Tianjin as an example to construct the family energy management model, considering the user's satisfaction and the minimum cost of the user, and using the improved genetic algorithm to solve. Scenario 1 only considers user satisfaction and simulates the electricity habits of most users; Scenario 2 only considers the minimum cost of electricity; Scenario 3 considers the multi-objective function that considers the minimum cost of electricity and considers user satisfaction. The results obtained are as follows:





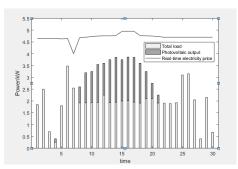


Fig.4 Only consider the total load and photovoltaic output with the lowest electricity cost

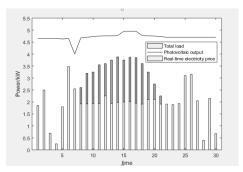


Fig.5. Total user load and Photovoltaic output based on user satisfaction and cost minimization

Table 1. Partition scheme and objective function value

Situation	customer satisfaction	Costs
Situation 1	100	39.6
Situation 2	20.14	34.3
Situation 3	95.29	35.2

## 5. CONCLUSIONS

In this paper, a multi-objective optimization model of cost minimization and user satisfaction is established which can effectively optimize the household load and ensure the satisfaction while reducing the cost. The improved genetic algorithm HPGA, combining Pareto genetic algorithm with local optimization method, realizes the complementarity of optimization mechanism and optimization behavior, and has not only good global optimization performance, but also strong local search ability.

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