Resolution of the unit commitment problems by using the hybrid Taguchi-ant colony system algorithm

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1. Introduction

Unit commitment (UC) profoundly impacts power system operations under both centralized and competitive frameworks. Unit commitment (UC) attempts to schedule the operation of the generating units in order to fulfill the load demand at minimum operating costs while examining all plant and system constraints over a given scheduling period, ranging from several hours to days ahead. In the UC problem, the decisions are the selection of the time for each unit to be on and/or offline as well as economic generation level of the unit in every scheduling sub-period.

As a large-scale, non-convex, nonlinear, mixed-integer optimization problem [1], UC is normally too complex to solve owing to its enormous dimensions, nonlinear objective function and coupling constraints. This complex problem has received considerable attention for decades, with many conventional and heuristic techniques developed. Additionally, several hybrid techniques combining different optimization techniques have been developed to solve this power-system-optimization problem. The need for feasible, cost-effective UC solutions led to the development of various UC algorithms that provide suboptimal yet efficient scheduling for real sized power systems comprising hundreds of generators.

Many efficient and robust UC methods have been developed. The algorithms so far developed can be classified into two main categories. The first category represents numerical optimization techniques such as priority list methods [2], dynamic programming [3], Lagrangian relaxation methods [4,5], branch-and-bound methods [6], integer and mixed-integer programming method, and mixed-integer programming [7]. The second category consists of stochastic search methods such as genetic algorithms [8,9], evolutionary programming [10], simulated annealing [11,12] and particle swarm optimization [13]. Although simple and fast, the priority list approach usually incurs high production costs. The dynamic programming is flexible yet incurs the problem of high dimensionality. The branch-and-bound method uses a linear function to represent the fuel consumption and time-dependent startup costs, as well as obtains the required lower and upper bounds. However, its computational time increases exponentially with the increment of the dimensions of the UC problem. Despite using linear programming to obtain and verify an integer solution, the mixed integer programming method is limited by an excessive computational time. While offering a fast solution, the Lagrangian relaxation method may incur numerical convergence problems. Despite its development, an artificial intelligence method based on heuristic depth-first search method [14] may be limited in its application to large-scale UC problems.

Additionally, related works have attempted to reduce the search space in large-scale problems by using hybrid methods such as the Lagrangian relaxation and genetic algorithm (LRGA) [15] as well as the priority list-based evolutionary algorithm (PLEA) [16]. These methods are more efficient than single ones due to lower production costs and a faster computational time.

The ant system has been applied recently to combinatorial optimization problems such as the UC problem. Owing to their feasibility for combinatorial optimization problems and their potential in locating a nearly global optimization solution, ant algorithms are appropriate for the UC problem. Huang [17] solved a hydroelectric
generation scheduling problem using ant colony system-based optimization approaches. Simon and Yu [18,19] introduced the ant colony search algorithm for solving UC problems. Shi [20] developed an approach by combining an ant colony optimization method and random search schemes for the UC problem. Christopher [21] proposed a nodal ant colony optimization scheme to solve a profit based UC problem in a deregulated power system environment. Vaisakh [22] developed an evolving ant colony optimization method for solving UC problem. In this method, an optimal set of ant colony optimization (ACO) parameters is found, while ACO solves the UC problem. Owing to the nature of the ACS algorithm, the application of ACS on the UC problem of a large-scale power system would require an appropriately simplified procedure. Jang [23] proposed a useful and efficient method to enhance the ACS performance on the UC problem.

Despite the satisfactory performance of conventional ant systems for small problems with a moderate dimension and searching space, ant systems require an excessive computational time for large-scale unit commitment. In addition, these systems may be limited in convergence, and local optimal solutions may be generated. To overcome these limitations, this work presents a novel method, HT-ACS, for solving the UC problem for a 24 h period. HT-ACS combines ACS [17–20] with the Taguchi method [24]. HTACS is inserted into the ACS procedure before its global pheromone update. The systematic reasoning ability of TM is used to select an improved UC solution quickly for representing the potential of parameters with a high degree of freedom. For parameter optimization, a limited in convergence, and local optimal solutions may be generated. Hence, the ACS algorithm, the application of ACS or HTACS on the UC problem of a large-scale power system would require an appropriate simplified procedure. The main goal of this paper is to enhance ACS, explaining why HTACS is more robust, statistically sound, and quickly convergent. Moreover, effectiveness of the proposed method is demonstrated by testing it on an actual system.

Several up-to-date reviews of the literature on UC problems are listed in [25–30]. Furthermore, recent research on the UC problem concerning the IEEE 118-bus system is presented in [31–33]. Indeed, owing to the nature of the ACS algorithm, the application of ACS or HTACS on the UC problem of a large-scale power system would require an appropriate simplified procedure. The method proposed by Se-Hwan Jang et al. is one of the useful and efficient methods. In our study, the simplified procedure for the UC problem includes:

2. Problem formulation

The UC problem attempts to determine the start-up and shut-down schedules as well as generation outputs of thermal units to satisfy the fluctuating load requirements over a scheduling period at minimal cost. With the total scheduling time divided into a set of shorter time intervals, the UC problem can be mathematically formulated as a constrained nonlinear optimization problem.

Fuel cost is the major component of the operating costs. For all committed generating units, the total fuel cost is minimized by economically dispatching the units. The fuel cost function of a unit can be expressed as a second-order polynomial. The UC problem is described as follows:

Minimize $F = \sum_{i=1}^{T} \sum_{t=1}^{N} P_{it} FC_{it}(P_{it}) + ST_{it} + SD_{it}$

(1)

where $F$ denotes the total cost over the entire scheduling periods including operating costs; $T$ represents the total scheduling period; $N$ refers to the number of a generator; $FC_{it}(P_{it})$ is the full cost at the $t$th hour (\$/h); $P_{it}$ denotes the on/off status of unit $i$ at the $t$th hour (on = 1 and off = 0); $ST_{it}$ (\$/h) is start-up cost of unit $i$ at the $t$th hour; and $SD_{it}$ (\$/h) refers to the shut down cost of unit $i$ at the $t$th hour.

Fuel cost is generally the I/O curve that is modeled with a quadratic function

$$FC_{it}(P_{it}) = a_{it} P_{it}^2 + b_{it} P_{it} + c_{it} \quad (\$/h)$$

(2)

where $P_{it}$ (MW) denotes the power level at the $t$th hour; and $a_{it}$, $b_{it}$, and $c_{it}$ are the unit cost coefficients.

Start-up cost is a function of hours that the unit has been down. A simplified time-dependent start-up cost is also adopted:

$$ST_{it} = \begin{cases} \text{hot start cost} & t_{i} \leq t_{i,off} < t_{i,off} \\ \text{cold start cost} & t_{i,off} \geq t_{i,off} \end{cases}$$

(3)

$$H_{i,off} = t_{i,down} + t_{i,cold}$$

where $T_{i,off}$ is the duration for which the $i$th unit is continuously off-line at the $t$th hour; $T_{i,down}$ is the $i$th unit minimum down time; and $T_{i,cold}$ is the $i$th unit cold start time.

Minimization of the cost $F$ is subject to the following constraints:

(a) Power balance

$$\sum_{i=1}^{N} P_{it} = P_{Dt}$$

(4)

where $P_{Dt}$ is system load demand at the $t$th hour.

(b) Minimum up-time

$$T_{g,on} \geq T_{ip}$$

(5)

where $T_{g,on}$ is duration for which the $i$th unit is continuously on-line at the $t$th hour, and $T_{ip}$ refers to the $i$th unit minimum up time.

(c) Minimum down-time

$$T_{g,off} \geq T_{idown}$$

(6)

where $T_{g,off}$ is duration for which the $i$th unit is continuously off-line at the $t$th hour, and $T_{idown}$ is the $i$th unit minimum down time.

(d) Maximum and minimum output limits on generators

$$P_{i,off} \leq P_{i}\leq P_{i,max}$$

(7)

where $P_{i,off}$ and $P_{i,max}$ are the minimum and maximum limits of unit $i$ respectively.

(e) Ramp rate limits for unit generation changes

$$P_{i} \leq P_{i}^{(t-1)} + UR_{i} \quad \text{as generation increases}$$

(8)

$$P_{i} \geq P_{i}^{(t-1)} - DR_{i} \quad \text{as generation decreases}$$

(9)

where $UR_{i}$ (MW/h) denotes the ramp-up rate limit of unit $i$, and $DR_{i}$ (MW/h) is the ramp-down rate limit of unit $i$.

(f) Spinning reserve

$$\sum_{i=1}^{N} v_{it} \times P_{i,max} \geq P_{Dt} + R_{t}$$

(10)

where $R_{t}$ refers to the spinning reserve at the $t$th hour.

The system and unit capacity related constraints (4), (7), and (10) can be handled by solving the economic dispatch problem at each state. After the searching space of multi-stage scheduling is determined, the unit transition-related constraints, i.e. (3), (5), and (6), are considered during state transition by the ant colony system algorithm.

3. Brief introduction of Taguchi method

Pioneered by G. Taguchi in 1985, the Taguchi method(TM) is based on an appropriate orthogonal array (OA), and is normally used to analyze experimental results and derive the optimum value of parameters with a high degree of freedom. For parameter optimization problems with a given computable objective function, use of
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