Applied Energy 204 (2017) 143-162



Contents lists available at ScienceDirect

Applied Energy

journal homepage: www.elsevier.com/locate/apenergy

Study on day-ahead optimal economic operation of active distribution networks based on Kriging model assisted particle swarm optimization with constraint handling techniques



AppliedEnergy

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HIGHLIGHTS

• Presenting the optimal economic operation and schedule model of the ADN.

• Developing a Kriging model assisted modified fuzzy adaptive PSO algorithm.

• A constraint-handling method based on the basic principle of PSO is developed.

• A dynamic update method is developed in KMA-MFAPSO to rebuild the Kriging model.

ARTICLE INFO

Article history: Received 9 April 2017 Received in revised form 31 May 2017 Accepted 16 June 2017

Keywords: Active distribution networks Optimal operation Kriging model Particle swarm optimization Constraint handling technique

ABSTRACT

The distribution system is demanded to be more efficient, more flexible and more intelligent due to the continued growth of the electricity loads, the high efficiency of energy utilization and the environmental protection. The traditional distribution systems are facing the challenge of evolving from passive networks to the active distribution networks (ADN) with the integration of multiple controllable resources. This paper presents an optimal operation and schedule model of the ADN, considering a variety of controllable resources such as distributed generations, battery storages and interruptible loads. To solve the optimization problem, a modified fuzzy adaptive PSO assisted by Kriging model (KMA-MFAPSO) is developed in this paper. In KMA-MFPSO, a novel constraint handling technique (CHT) based on the PSO is proposed to handle the constraints effectively. In addition, under the premise of ensuring the accuracy of the calculation, the Kriging model is used in KMA-MFAPSO to calculate the power flow the ADN approximately, which greatly speeds up the solving process. Finally, the effectiveness of the proposed algorithm is tested on a modified IEEE-123 system. The optimal results obtained by the proposed method are compared with the results obtained by using other solving algorithms. The simulation results indicate that KMA-MFAPSO is very robust and fast to solve the optimization problem so that it can be used in practical systems.

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1. Introduction

Nowadays, the power system is demanded to be more efficient, more flexible and more intelligent due to the continued growth of electricity loads and the shortage of traditional energy sources. What is distinct about the modern distribution system is the

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http://dx.doi.org/10.1016/j.apenergy.2017.06.053 0306-2619/© 2017 Elsevier Ltd. All rights reserved. large-scale integration of distributed generations (DGs), especially the integration of wind turbines and photovoltaic power generations. However, high penetration rate of DGs will have a huge influence on traditional distribution system mainly in the following aspects: changing the voltage profiles of distribution system, increasing the difficulties of relay protection, deteriorating the power quality and influencing the reliability of the system and so on [1,2]. Although the maturing technology of micro-grid has provided a solution to the integration of DGs, due to the capacity Nomenclature

Nomenclature		
Т	total scheduling periods	tap ¹
Δt	time interval	-
$C_{i,t}^{\mathrm{buy}}$	electricity price of substation i at time interval t	MT
C_i^{dg}	generating cost of distributed generation <i>i</i>	MC
C_i^{batcha} , (C_i^{batdis} charging and discharging costs of battery storage <i>i</i>	
C_i^{IL}	cost of the interruptible load <i>i</i>	$P_{i,t}^{\mathrm{IL},\mathrm{U}}$
C_i^{tap}	operating cost of voltage regulator <i>i</i>	MIL
C_{i}^{IL} C_{i}^{tap} C_{i}^{cap} N_{sub}	operating cost of capacitor bank <i>i</i>	IVIIL
N _{sub}	number of substation	N _{sw}
N _{dg} N _{bat}	number of distributed generations number of battery storages	pbe
N _{IL}	number of interruptible loads	gbe. x ^k , t
N _{tap}	number of voltage regulators	ω
$N_{ ext{cap}} P_{i,t}^{ ext{buy}}$	number of capacitor banks active power of substation <i>i</i> buying from the transmis-	<i>c</i> ₁ , <i>c</i>
	sion system at time interval <i>t</i>	$v_{i,j}^{\min}$
$P_{i,t}^{\mathrm{dg}}$	generated active power of the <i>i</i> th DG at time interval <i>t</i>	$v_{i,j}^{\mathrm{upj}}$
$P_{i,t}^{bat}$	charging and discharging power of the <i>i</i> th battery stor-	
$P_{i,t}^{\mathrm{IL}}$	age at time interval <i>t</i> load curtailment of interruptible load <i>i</i> at time interval <i>t</i>	$x_{i,j}^{\mathrm{upp}}$
$tap_{i,t}$	actual tap position of the voltage regulator i at time	$x_{i,j}^{\max}$
	interval <i>t</i>	•i.j
$tap_{i,t}^{c}$	the mapped continuous tap position of the voltage reg- ulator <i>i</i> at time interval <i>t</i>	Δso
$cap_{i,t}$	actual switch status of the capacitor bank <i>i</i> at time inter- val <i>t</i>	Δso
$cap_{i,t}^{c}$	the mapped continuous switch status of the capacitor bank i at time interval t	$\Delta E_{i,j}^k$
$s_{i,t}^{\text{batcha}}$, $s_{i,t}$	at time interval <i>t</i>	MR_i^l
	active power and active load of node <i>i</i> at time interval <i>t</i> of phase <i>p</i>	ΔLE
	^{<i>p</i>} reactive power and reactive load of node <i>i</i> at time inter- val <i>t</i> of phase <i>p</i>	ΔIL_{i}^{I}
$V_{i,t}^p$	voltage magnitude node i at time interval t of phase p	I
$V_{j,t}^m$	voltage magnitude node <i>j</i> at time interval <i>t</i> of phase <i>m</i>	$\lambda_{i,j}^{k,m}$
$V^{\text{low}}, V^{\text{up}}$	^{pp} the minimum and the maximum allowable bus voltage magnitude	
G_{ij}^{pm} , B_{ij}^{pm}	^{<i>n</i>} real and imaginary parts of node admittance matrix	$t_{i,j,\mathrm{ir}}^{k,m}$
$\theta_{ij,t}^{pm}$	voltage phase angle difference	$\alpha_{i,j,t}^{k,m}$
Num	number of bus	
$S_{l,t}, S_l^{upp}$	limit of transmission capacity of line l	$\sigma_{i,j,i}^{k,n}$
N_l	number of transmission lines	
$pf_{i,t}^{dg}$	power factor of DG <i>i</i> at time interval <i>t</i>	$\varphi_{i,j}^{k,r}$
$pf_i^{ m dg,low}$, $P_i^{ m maxdis}$,	$pf_i^{dg,upp}$ lower and upper limits of power factor of DG <i>i</i>	
$P_i^{\text{individual}}$,	P_i^{maxcha} maximum discharging and charging power of battery storage <i>i</i>	KM
SOC _{i,t}	state of charge (SOC) of battery storage <i>i</i> at time interval <i>t</i>	CHT
SOC ^{low} ,	SOC_i^{upp} lower and upper limits of SOC of batter storage <i>i</i>	
$\eta_i^{ ext{cha}}$, $\eta_i^{ ext{dis}}$	^s charging and discharging efficiencies of battery storage <i>i</i>	FAP GA
B _i	battery capacity of battery storage <i>i</i>	ACC

 tap_i^{low} , tap_i^{upp} actual lower and upper limit of tap positions of voltage regulator *i*

- MT_i maximum tap operating number of voltage regulator *i* during the scheduling periods
- MC_i maximum switch operating number of capacitor bank *i* during the scheduling periods
- $P_{i,t}^{\text{IL,upp}}$ upper limit of load curtailment of interruptible load *i* at time interval *t*
- *MIL*_{*i*} maximum load curtailment amount of interruptible load *i* during the scheduling periods
- *N*_{swarm} the population size of particles
- $bbest_i$, the best positon recorded by particle *i*
- best the best position among all particles
- $\mathbf{x}_{i}^{k}, \mathbf{v}_{i}^{k}$ the position and the velocity of particle *i* at iteration *k* ω inertia weight
- 1,*c*₂ the individual and social acceleration coefficients
- $v_{ij}^{\min,k}$, $v_{ij}^{\max,k}$ variable range of the velocity of the *j*th decision variable of particle *i* at iteration number *k*
- $v_{ij}^{\text{upp},k}, v_{ij}^{\text{low},k}$ the upper and lower boundaries of particle's velocity at iteration k.
- x_{ij}^{upp} , x_{ij}^{low} the lower and upper boundaries of the *j*th dimension of particle *i*
- $x_{ij}^{\max,k}$, $x_{ij}^{\min,k}$ the flying range of the *j*th decision variable of particle *i* at iteration *k*
- As $oc_{i,j,t}^k$ variation of SOC of the *j*th battery storage of particle *i* at iteration number *k*
- Asoc^{low}, Δsoc^{upp}_{j} the lower and upper boundaries of the *j*th dimension of variation of SOC
- $\Delta E_{ij,t}^k$ the excessive amount of SOC constraint for the *j*th battery storage of particle *i* at iteration number *k*
- $MR_{i,j,t}^k$ the maximum reduction amount of SOC without interfering the natural process of PSO
- $\Delta LE_{i,j}^k$ the excess amount of load reduction of the *j*th interruptible load of particle *i* at iteration number *k*
- $\Delta IL_{i,j}^{\max,k}$ the maximum reduction amounts for all time intervals in the flying range of the *j*th interruptible load of particle *i* at iteration number *k*
- $a_{ij}^{k,m}$ the tap positions adjustment coefficient of time period m of the *j*th voltage regulator of particle *i* at iteration number *k*
- $k,m_{i,j,ini},t_{i,j,end}^{k,m}$ the firs and the last time intervals of period m
- $\alpha_{i,j,t}^{k,m}, \beta_{i,j,t}^{k,m}$ forward-adjustment coefficient and backwardadjustment coefficient of the tap position at time interval *t* in the period *m*
- $\sigma_{ij,t}^{k,m}$ the adjustment coefficient of switch state of the time interval *t* in time period *m* of the *j*th capacitor bank of particle *i* at iteration number *k*
- $\varphi_{ij}^{k,m}$ the adjustment coefficient of switch state of time period m of the *j*th capacitor bank of particle *i* at iteration number k
- KMA-MFAPSO Kriging model assisted modified fuzzy adaptive PSO
- CHT-MFAPSO constraint handling technique based modified fuzzy adaptive PSO
- FAPSO fuzzy adaptive particle swarm optimization
- GA genetic algorithm
- ACO ant colony optimization

limitation of it, the large scale DGs' integration problem still cannot be settled effectively [3]. With the great development of the distributed generation technologies, energy storage technologies and power electronics, the grid-connection problem of DGs has been settled to a large extent [4]. However, owing to the lack of efficient optimization strategy, the low degree of automation and the lack of demand response in the traditional distribution system, the renewable energy consumption capability of the system is greatly limited, which restricts the development and the application of the clean and renewable energy sources and is not conducive to the energy structure adjustment [5]. In this situation, the active distribution network (ADN) technology emerges and is

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