Energy Conversion and Management 124 (2016) 231-246

Contents lists available at ScienceDirect



Energy Conversion and Management

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

Two-layer optimization methodology for wind distributed generation planning considering plug-in electric vehicles uncertainty: A flexible active-reactive power approach



Ali Ahmadian^{a,b,*}, Mahdi Sedghi^a, Masoud Aliakbar-Golkar^a, Michael Fowler^b, Ali Elkamel^b

^a Faculty of Electrical Engineering, K. N. Toosi University of Technology, P.O. Box 16315-1355, Tehran, Iran ^b Department of Chemical Engineering, University of Waterloo, 200 University Avenue West, Waterloo, ON N2L 3G1, Canada

ARTICLE INFO

Article history: Received 3 May 2016 Received in revised form 18 June 2016 Accepted 10 July 2016 Available online 18 July 2016

Keywords: Wind energy Flexible active-reactive power Distribution network Plug-in electric vehicles

ABSTRACT

With increasing the penetration of wind power, the voltage regulation becomes a more important problem in active distribution networks. In addition, as an uncertain load Plug-in Electric Vehicles (PEVs) will introduce a new concern in voltage adjustment of future distribution networks. Hence, this paper presents a flexible active-reactive power based Wind Distributed Generation (WDG) planning procedure to address the mentioned challenges. The uncertainties related to WDGs, load demand as well as PEVs load have been handled using the Point Estimate Method (PEM). The distribution network under study is equipped to on-load tap-changer and, as a conventional voltage control component, the Capacitor Banks (CBs) will be planned simultaneously with WDGs. The planning procedure has been considered as a two-loop optimization problem that is solved using Particle Swarm Optimization (PSO) and Tabu Search (TS) algorithms. The tap position and power factor of WDGs are taken into account as stochastic variables with practical limitations. The proposed methodology is applied to a typical distribution network and several scenarios are considered and analyzed. Simulation results show that the standard deviation of power factor depends on PEVs penetration that highlights the capability curve of WDGs. The optimal penetration of wind power increases nonlinearly versus increasing of PEVs connected to the distribution network, however the fixed CBs are required to increase the optimal penetration of WDGs. The proposed Modified PSO (MPSO) is compared with the conventional PSO in numerical studies that show MPSO is more efficient than the conventional algorithm for this analysis.

© 2016 Elsevier Ltd. All rights reserved.

1. Introduction

Due to the environmental and economic benefits, the renewable-based Distributed Generation (DG) sources have become more popular in today's distribution network. Loss reduction, flattering of peak, increasing reliability, modifying voltage profile are the strong motivation for increasing the penetration of DGs in distribution networks [1,2]. Among all renewable energy generation technologies, Wind-based Distributed Generation (WDG) units have emerged very rapidly in recent years. Reduction of investment costs, reliability improvement, and efficiency enhancement of WDGs have made them able to compete with the conventional power generation [3,4]. However, to release the maximum benefit from WDG there is a need for optimal planning, i.e. sizing and siting, as well as optimal operation [5,6]. The output

E-mail address: ali.ahmadian@uwaterloo.ca (A. Ahmadian).

power of WDGs is affected by natural intermittency and variably of the wind resource, so the related uncertainty should be considered in optimal planning and operation procedures [7,8].

In the literature some methods have been proposed to model the uncertainties in WDGs planning. Optimal combination of different renewable resources is proposed by Atwa et al. [9]. The uncertainty of load demand and renewable generation are often modeled with Probability Distribution Functions (PDF). In [10], Soroudi proposes a hybrid possibility–probabilistic method for technical assessment of DG impact on distribution network performance. In this work, the uncertainty related to load demand, DG operation and investments are taken into account, while the objective function considers active power losses and technical risk including the possibility of under/over voltage in load nodes. In [11], Soroudi et al. propose a possibilistic method to handle the uncertainties of load demands and energy prices considering different objective functions like cost and technical and economic risks. A stochastic dynamic model for integration of DGs in

^{*} Corresponding author at: Faculty of Electrical Engineering, K. N. Toosi University of Technology, P.O. Box 16315-1355, Tehran, Iran.

Nomenclature

Parameters and variables

C_{DG}^{INS}	installation cost related to the capacity of WDG (\$)	S_t^{LOSS}	t
C_{CAP}^{INS}	installation cost related to the capacity of capacitor	$S_{n,st}$	S
CCAP	banks (\$)	T_i	d
C_c^{OM}	operation and maintenance cost of WDC units in <i>i</i> th	Ŭ	S
C _{DG,j}	season (\$)	$V_{j,t}$	v v
$C_{s,i}^{SS}(t)$	operation cost of the s-th HV/MV substation at time t of	V min V max	a
anr (c)	<i>j</i> -th season (\$)	$V_{n,st}$	ν
$C_{p,j}^{pr}(t)$	active power price at time t of j-th season (\$)	v_{ij}^k	t
$C_{q,j}^{pr}(t)$	reactive power price at time <i>t</i> of <i>j</i> -th season (\$)	41	11
c_1, c_2	constant controller parameters of PSO	$\frac{\nu_{\min}}{V_{ii}}$	a a
EDG	capacity of the k th WDC upit $(100A)$	• y	S
E_k E(·)	the expected value	x_j^{CB}	t
F	cost objective function (\$)	x_j^{DG}	t
h (.)	nonlinear function of deterministic load flow	VS	(
IC Infi	total investment cost (\$)	X_n	S
iiiji Intr	interest rate	X_n^{m}	e
I ^{scwd}	maximum allowed rotor current (p.u.)	X_{ij}^{κ}	t
M	a fixed large number	χ^{k}	t
n _{DG}	number of all WDG units	pD,IJ	i-
n_{CB}	number of all HV/MV substations	$x_{gb,j}^k$	t
n_{SS} n_{FO}	number of all the equipment	141	11
n_N	number of all nodes in distribution network	w Z ^m	e
n_{LD}	number of all load nodes	Z_n^s	s
00	total operation cost (\$)	γ _{n,st}	р
DC _j PF	penalty factor (\$)	φ_c	C
$P_{p,i}^{SS}(t)$	absorbed/injected active power in s-th HV/MV substa-	$(o^{SS}(t))$	e r
- DEV	tion at time <i>t</i> of <i>j</i> -th season (kW)	$\varphi_{s,j}(t)$	j.
$P_{l,t}^{rLv}$	PEVs load demand in <i>l</i> -th load node at time <i>t</i> (kW)	$\sigma^{\scriptscriptstyle H}_{ij}$	n
$p_{j,t}^u$	probability of capacity constraint violation in device <i>u</i> at	_D	n
n_{\cdot}^{n}	probability of voltage constraint violation in <i>n</i> -th node	o_{ij}^{D}	n n
Pj,t	at time <i>t</i> of <i>j</i> -th season	3	S
$P_{n,st}^{wd}$	active power injection of Doubly Fed Induction Genera-		
a	tor (DFIG) (p.u.)	Abbrevia	tio
q	intensification	CBs	C
$Q_{si}^{SS}(t)$	absorbed/injected reactive power in <i>s</i> -th HV/MV substa-	DFIG	L T
CAD	tion at time t of j-th season (kVAR)	HV/MV	ŀ
Q_c^{CAP}	injected reactive power of <i>c</i> -th capacitor bank to the	MCS	N
O^{wd21}	grid (KVAR) maximum allowed reactive power absorption (p.u.)	MINLP	N
$\sim_{n,st}$ o^{wdlu}	maximum allowed reactive power injection (p.u.)	MPSO NHTS	n N
$Q_{n,st}$	maximum anowed reactive power injection (p.u.)	O&M	0
rand R ^m	a random number within $(0, 1)$	PAROPF	P
R _n D ^S	stator resistance of a DEIC (p.u.)	PDF	P
α _n ς ^u	nower of equipment u at time t (EVA)	PEM	P
S _t C ^u	power of equipment u at time t (KVA)	PLF	P
S _{max}	anowed maximum power of equipment u (KVA)	PSO	P
$S_{s,t}^{}$	power of s-th HV/MV substation at time t (KVA)	TS	T
$S_{k,t}^{}$	generated power of κ -th DG unit at time t (KVA)	WDGs	V

distribution networks is presented in [12]. The proposed multiobjective model optimizes three objectives including technical constraint dissatisfaction, costs and environmental emissions. A scenario-based uncertainty modeling is applied to load demands, electricity price and wind power generators.

$S_{l,t}^{LD}$ S_{t}^{LOSS} Sn st	demand of the <i>l</i> -th load node at time <i>t</i> (kVA) total power loss in distribution network at time <i>t</i> (kVA) slip associated with steady state operation of DEIG	
T	period of the project (years)	
T:	duration of the <i>i</i> -th season in one year (days)	
U	standard step function	
V	voltage magnitude in <i>i</i> -th node at time $t(n u)$	
V_{min}	allowed minimum voltage magnitude (p.u.)	
Vmax	allowed maximum voltage magnitude (p.u.)	
Vn st	voltage magnitude at node <i>n</i> over a system state <i>st</i> (p.u.)	
v_{ij}^{k}	the <i>j</i> -th component of velocity of the <i>i</i> -th particle in k -th iteration of PSO	
$v_{\rm min}$	minimum allowed velocity	
V _{ij}	average voltage level in <i>j</i> -th node considering the <i>i</i> -th	
x_i^{CB}	total capacity of the installed CBs in <i>j</i> -th node (kVAR)	
χ_{i}^{DG}	total capacity of the installed WDG units in <i>i</i> -th node	
	(kVA)	
X_n^s	stator reactance of a DFIG (p.u.)	
X_n^m	equivalent main reactance of a DFIG (p.u.)	
x_{ii}^k	the <i>j</i> -th component of position of the <i>i</i> -th particle in <i>k</i> -th	
	iteration of PSO	
$x_{pb,ij}^{k}$	the <i>j</i> -th component of the previous best position of the	
k	<i>i</i> -th particle in <i>k</i> -th iteration of PSO	
$x_{gb,j}^{n}$	in $k_{\rm t}$ th iteration of PSO	
147	inertia weight of PSO	
7 ^m	equivalent main impedance of a DFIC $(n \mu)$	
Z_n Z^S	stator impedance of a DFIG (p.u.)	
2n v.	power factor angle	
n,st (D	critical power factor angle for free reactive power in	
τc	electricity market (°)	
$\varphi_{s,j}^{SS}(t)$	power factor angle in s-th HV/MV substation at time t of	
_H	<i>j</i> -th season (°)	
o_{ij}	normalized nourly variance of voltage magnitude in <i>j</i> -th	
σ^{D}	normalized daily variance of voltage magnitude in <i>i</i> -th	
⁰ ij	note considering the <i>i</i> -th solution	
3	small constant controller parameter in MPSO	
0		
Abbreviat	ions	
CBs	Capacitor Banks	
DFIG	Double Fed Induction Generator	
DG	Distributed Generation	
HV/MV	High Voltage/Medium Voltage	
MCS	Monte Carlo Simulation	
MINLP	Mixed Integer Nonlinear Programing	
MPSO	modified particle swarm optimization	
NHTS	National Household Travel Survey	
0&M	operation and maintenance	
PAROPF	Probabilistic Active/Reactive Optimal Power Flow	
PDF	Probability Distributed Function	
PEM	Point Estimate Method	
PEVS	Plug-in Electric Vehicles	
PLF	Probabilistic Load Flow	
PSU	Particle Swarm Optimization	
15	IdDU SedfCD Wind Distributed Conceptions	
VVDGS		

In addition to the uncertainty modeling, if the WDGs are planned without considering optimal operation, the WDGs ability on improving branch power flow and node voltage may be reduced. This concern is highlighted with high penetration of WDGs. Therefore, they should be planned under active operation Download English Version:

https://daneshyari.com/en/article/764975

Download Persian Version:

https://daneshyari.com/article/764975

Daneshyari.com