

Contents lists available at ScienceDirect

Electric Power Systems Research



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

A mixed integer linear programming model for the energy management problem of microgrids

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ARTICLE INFO

Article history: Received 30 June 2014 Received in revised form 12 December 2014 Accepted 19 December 2014

Keywords: Energy management Microgrids Microturbines Fuel cells Controllable load demand Mixed linear integer programming

ABSTRACT

This paper presents a mathematical model for the energy management (EM) problem of a microgrid (MG) by means of a mixed integer linear programming approach. In the EM problem, the objective is to determine a generation and a controllable load demand policy that minimises, over a planning horizon, the operation cost subject to economical and technical constraints. We propose a detail modelling for microturbines (MTs) and fuel cells (FCs), where the constraints associated with such factors as the ramps, minimum up and downtime, and generation limits, represent various peculiarities that have not been adequately considered in literature. The proposed model also considers a detailed representation of critical, reschedulable and curtailable loads, which are important aspects in the MG concept. To analyse the proposed modelling, a MG is used along with a MT, a FC, a battery bank, wind and photovoltaic generators connected to the main grid. The results indicate that the model is adequate for the MG EM.

1. Introduction

The modern electrical energy industry addresses more affordable electronic technologies. The integration of small Distributed Energy Resources (DERs), such as Microturbines (MTs), Fuel Cells (FCs), batteries, wind and photovoltaic generators, is a trend that is currently in progress. The presence of the DERs and the demand management can reduce fossil fuel consumption, load peak shaving, as well as postpone investments in new transmission and distribution lines [1,2]. In this new paradigm, it is important to highlight the Microgrids (MGs), which are emerging as an additional element to maintain the growth and sustainability of the modern electric energy industry. Roughly speaking, a MG consists of a group of DERs and controllable and uncontrollable loads that operate either synchronised with the main grid or autonomously.

Despite several advantages of MGs, the new challenges are inherent, such as those related to DERs. In this context, a methodological challenge that supports the economic and technical operational issues of MGs is the energy management (EM) problem [3,4]. In general, solving this problem requires determining a generation and a controllable load demand policy that minimises, over a planning horizon, an objective function subject to economical and technical constraints. The policy is given by the on/off status, the respective output active power of each DER, the on/off status of the curtailable load demand (CLD) and the schedule of the reschedulable load demand (RLD). This strategy is used for the voltage and frequency control in MG real-time operation. Therefore, because it is necessary to minimise an objective function subject to constraints, the EM is usually performed based on the solution of an optimisation problem, although there are other possibilities, such as fuzzy logic and expert systems [5,6] and hierarchical and decentralised control [7,8].

The EM in [9] is performed for a MG with wind and thermal generations, aiming for the minimisation of the total operation cost and considering the stochastic behaviour of the wind. In [10], the EM addresses a MG composed of batteries and photovoltaic generation with connected operation with the main grid, which allows for the purchasing and selling of energy. In [11], the EM is performed for a MG composed of FC, wind, photovoltaic generation and batteries, considering a quadratic objective function for the cost. In [12], the EM is performed for a Microgrid with a MT, a FC, a diesel, a wind, and a photovoltaic generator and a battery, with a multiobjective approach to minimise the cost of operation and reduce

Abbreviations: CLD, curtailable load demand; DER, distributed energy resource; FC, Fuel cells; MG, microgrid; MILP, mixed-integer linear programing; MT, micro-turbine; RLD, reschedulable load demand; SOFC, solid oxide fuel cell.

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Nomenclature				
Index/sets				
а	index related to MTs ($a \in A$)			
b	index related to SOFCs $(b \in B)$			
C	index related to CLDs ($c \in C$)			
a	index related to KLDS $(a \in D)$			
e ;	index related to datteries $(e \in E)$			
l	and ECs ramps, PLD and CLD			
t	index for the time step ($t \in ND$).			
Vaniable	-			
variable.	s			
upj _{bt}	absolute power output difference between $t = 1$ and t stages of SOEC $h(kM)$			
eh .	t stages of sore $b(KW)energy of battery e(kWh) in step t$			
nhCat	nower charge of battery $e(kW)$ in stage t			
nhdat	power discharge of battery $e(kW)$ in stage t			
ndC _{ct}	power of CLD c (kW) in stage t			
pdd_{dt}	RLD d in stage t (kW)			
pde_t	system deficit (kW) in stage t			
pex _t	excess generation (kW) in stage t			
pf_{bt}	power output of SOFC b (kW) in stage t			
pgb_t	power purchased from the grid (kW) in stage t			
pgs _t	power sold to the grid (kW) in stage t			
pt _{at}	power output of MT a (kW) in stage t			
rb _{et}	reserve of battery <i>e</i> (kW) in stage <i>t</i>			
ub _{et}	binary variable that indicates whether battery e is			
	discharging (<i>ub_{et}</i> = 1) in stage <i>t</i>			
<i>uc_{ct}</i>	binary variable that indicates whether CLD <i>c</i> is on			
	$(uc_{ct} = 1)$ or off $(uc_{ct} = 0)$ in stage t			
ud _{dt}	binary variable that indicates whether RLD <i>d</i> starts			
	$(ua_{ct} = 1)$ in stage t him emutation is the time director if an SOFC is an (uf = 1)			
uj _{bt}	Difference of the state of the			
110	of off $(U_{j_t} = 0)$ in stage t binary variable that indicates whether the MC is			
ugt	importing energy $(ug = 1)$ in stage t			
111-4	hippering energy $(ugr = 1)$ in stage t hipperiod binary variable that indicates whether MT a is on			
utat	$u_{tat} = 1$ or off $(u_{tat} = 0)$ in stage t			
VCct	auxiliary binary variable for indicating the start of			
9000	the load shedding in stage t of CLD c			
vf _{bt}	SOFC <i>b</i> auxiliary binary variable for the start-up			
5500	ramp rate in stage t			
yt _{at}	MT <i>a</i> auxiliary binary variable of start-up ramp rate			
-	in stage t			
$z f_{bt}$	SOFC <i>b</i> auxiliary binary variable for the shutdown			
	ramp rate in stage t			
<i>zt_{at}</i>	MT a auxiliary binary variable of shutdown ramp			
	rate in stage <i>t</i> .			
Paramet	ers			
AT _{at} BT	at constants associated with the fuel consumption			
	function of MT <i>a</i> operating with a fixed ambient			
	temperature, in (R\$/h) and (R\$/kWh), respectively			
BP_t	energy purchase price in stage <i>t</i> (R\$/kWh)			
CB	battery <i>e</i> charge step ramp (kW)			
CCc	incremental cost during one hour of load shedding			
-	(R\$/kWh) of CLD c			
CD	load deficit incremental cost (R\$/kWh)			
CE	system excess energy incremental cost (R\$/kWh)			
CF_b	incremental operating cost of SOFC b (R\$/kWh)			
DB	battery e discharge ramp (kW)			

		UDD _d UDF _b	number of number of
sociated with the fuel consumption T <i>a</i> operating with a fixed ambient n (R\$/h) and (R\$/kWh), respectively se price in stage <i>t</i> (R\$/kWh) ge step ramp (kW) ost during one hour of load shedding <i>LD c</i> remental cost (R\$/kWh) energy incremental cost (R\$/kWh) operating cost of SOFC <i>b</i> (R\$/kWh) arge ramp (kW) in stage <i>t</i> (kW)		$UDT_a \\ \alpha_a \\ \eta_e^{bc} \\ \eta_e^{bd} \\ \xi_b$	number of constant fo charge effic discharge e incrementa tion of SOF
		the greenhouse gas em the EM is performed fo and a battery, with a r	

DD ig namber of blageb of mit a b bilatao mit famp fate	DDT_a	number of stages of MT <i>a</i> 's shutdown ramp rate
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forecast CLD c

 DC_{ct}

_		
	D_t	forecast critical load demand in stage <i>t</i> (kW)
	DT_a	start-up cost of MT <i>a</i> (R\$)
	EB_e^F	final energy of battery <i>e</i> (kWh)
	EB_e^{max}	battery e maximum energy (kWh)
	EB _e ^{min}	battery e minimum energy (kWh)
	EB_e^I	initial energy of battery e (kWh)
	EB_e^L	energy lost in one time step of battery <i>e</i> (kWh)
	ED	error associated with the demand (%)
	EFb	start-up cost of FC b (R\$)
	EPV	error associated with the photovoltaic generation
		(%)
	EPW	error associated with the wind generation (%)
	FDD_d	final stage where the RLD <i>d</i> load has to be fully sup-
	-	plied
	FTa	shutdown cost of MT <i>a</i> (R\$)
	GF_b	shutdown cost of FC b (R\$)
	Η	horizon time (<i>h</i>)
	IDD _d	initial stage where the RLD <i>d</i> could be turned on
	MFC _b	maintenance incremental cost of FC b (R\$/kWh)
	<i>MTC</i> _a	maintenance incremental cost of MT a (R\$/kWh)
	NC _c ^{max}	maximum number of stages of load shedding for
		CLD c
	ND	number of stages in the planning horizon
	NDC_c^{st}	maximum number of load shedding for CLD c
	NF _b st	maximum number of start-ups of SOFC b
	NT _a st	maximum number of start-ups allowed of MT a
	P_a^{\max}	nominal maximum output power of MT a (kW)
	PDD _{id}	forecast RLD d in stage i (kW)
	PF_b^{\max}	maximum output power of SOFC b (kW)
	PF_b^{min}	minimum output power of SOFC b (kW)
	PFU _{bi}	output power in stage <i>i</i> of the SOFC <i>b</i> start-up ramp
		rate (kW)
	PGB_t^{max}	grid maximum power purchase in step t (KW)
	PGB_t	grid minimum power purchase in step t (KW)
	PGS_t^{max}	grid maximum power sell in step t (kw)
	PGSt ^{mm}	grid minimum power sell in step t (KW)
	PI_{at}^{max}	maximum output power of M1 <i>a</i> operating with an ambient temperature of T_{i} (kW)
	PT min	minimum output power of MT $a(kW)$
		output power in stage <i>i</i> of MT <i>a</i> 's shutdown ramp
	IID_{al}	rate (kW)
	PTU _{ai}	output power in stage <i>i</i> of MT a 's start-up ramp rate
	110 01	(kW)
	PV_t	forecast photovoltaic power in step t (kW)
	PW _t	forecast wind power in step t (kW)
	RB	number of time steps to the reserve of the system
	SPt	energy selling price in stage t (R\$/kWh)
	TCa	ambient temperature where the maximum output
	u	power of MT a decreases (°C)
	UDDa	number of stages in which RLD <i>d</i> is on
	UDFh	number of stages in the SOFC <i>b</i> start-up ramp rate
	UDTa	number of stages of MT <i>a</i> start-up ramp rate
	α_a	constant for MT <i>a</i> (slope of the line, in $kW/^{\circ}C$)
	η_e^{bc}	charge efficiency of battery <i>e</i> (%)
	η_e^{bd}	discharge efficiency of battery e (%)
	ξ _b	incremental cost associated with the cycling opera-
		tion of SOFC <i>b</i> (R\$/kWh).

nissions, such as NO_x, SO₂ and CO₂. In [13], or a MG with thermal and wind generators multi-objective approach to minimise the cost of operation and reduce the greenhouse gas emissions and energy capacity reserve. In [14], the EM is performed for a Microgrid with thermal, wind and photovoltaic generators with batteries, Download English Version:

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