



Optimal planning of Molten Carbonate Fuel Cell Power Plants at distribution networks considering Combined Heat, Power and Hydrogen production



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HIGHLIGHTS

- Propose a stochastic model for planning the location and operation of Molten Carbonate Fuel Cell Power Plants (MCF CPPs).
- Consider the effect of Combined Heat, Power, and Hydrogen (CHPH) simultaneously.
- Manage generation of thermal energy, and hydrogen, total emission of MCF CPPs and network.
- Consider uncertainties of the pressures of input hydrogen, oxygen, and carbon dioxide importing to MCF CPP.

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ABSTRACT

In this paper, a stochastic model is used for optimal planning of Molten Carbonate Fuel Cell Power Plants (MCF CPPs) in distribution network for producing Combined Heat, Power, and Hydrogen (CHPH). Total production costs of electrical energy, thermal energy, and hydrogen, emissions of MCF CPPs and network, and voltage deviation are considered in the objective function. In this paper, location and operation of MCF CPPs are taken into consideration while their investment cost is not taken into account. In this model, the uncertainties in forecasting the electrical and thermal loads, the pressures of hydrogen, oxygen, and carbon dioxide, and the indeterminacy of the nominal temperature of MCF CPP are considered using 2m + 1 Point Estimate Method (2m + 1 PEM). The problem of optimal planning of MCF CPPs as CHPH is of mixed integer nonlinear nature. So, a Self Adaptive Learning Bat-inspired Algorithm (SALBA) is employed for solving this problem. The problem is solved as a multi-objective one to achieve the best Pareto optimal set. A set of non-dominated solutions are saved in a repository and the proposed method is evaluated on a 69-bus distribution system.

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1. Introduction

Power system deregulation encourages electrical companies to employ renewable energy resources [1]. Renewable energies could properly substitute the fossil fuels and noticeably reduce green house gases [2]. Molten Carbonate Fuel Cell Power Plant (MCF CPP) is a profitable renewable energy resource in which electro-chemical reaction between hydrogen and oxygen produces

electricity, heat, and water [3], [4]. MCF CPPs have no rotating part, produce ignorable acoustic noise, and benefit from environment-friendly power production [5], [6]. They operate at high temperatures and are proper options for being used as CHPH [7]. In addition, as hydrogen is produced from methane, there is no need for direct sources of hydrogen [8].

Recently, many studies have addressed the operation of Distributed Generations (DGs) and renewable energy resources in distribution networks. Effects of DG locations on voltage stability margin in distribution networks are investigated in Ref. [9]. Moravej and Akhlaghi investigated the impacts of positioning of DGs on voltage profile improvements and power loss reductions using a novel approach based on cuckoo search [10]. Biswas et al. proposed a new formulation

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Nomenclature

V_{MFCFPP}	output voltage of MFCFPP	$e_{\text{grid,CO}_2}$	emission coefficient of grid
E_{eq}	equivalent cell potential	$e_{\text{MFCFPP,SO}_2}$	emission coefficient of MFCFPP units
R	universal gas constant ($8.3145 \text{ kJ kmol}^{-1} \text{ K}^{-1}$)	$e_{\text{MFCFPP,NO}_x}$	emission coefficient of MFCFPP units
z	number of electrons transferred per molecule of fuel	$e_{\text{MFCFPP,CO}_2}$	emission coefficient of MFCFPP units
P_{H_2}	pressure of hydrogen gas	V_{ref}	nominal voltage
P_{O_2}	pressure of oxygen gas	V_i^t	voltage magnitude of the bus i during time t
$P_{\text{H}_2\text{O}}$	pressure of water vapor	V_{Min}	lowest voltages of each bus
$P_{\text{CO}_2,c}$	pressure of carbon dioxide gas in cathode	V_{Max}	highest voltages of each bus
$P_{\text{CO}_2,a}$	pressure of carbon dioxide gas in anode	$P_{\text{MFCFPP},j}^{\text{t,min}}$	the minimum of active power produced by MFCFPP j during time t
Temp	MFCFPP temperature	$P_{\text{MFCFPP},j}^{\text{t,max}}$	the maximum of active power produced by MFCFPP j during time t
i_{MFCFPP}	output current of MFCFPP	Δt	step time
$-\Delta\bar{g}_f$	change in Gibbs free energy of the hydrogen reaction	$P_{\text{Max,MFCFPP}}$	the maximum power of MFCFPP
F	Faraday constant	$r_{\text{TE},j}$	the thermal energy to electrical energy ratio
R_{ohmic}	ohmic cell resistance	f_j^{min}	the lower bound of each objective function
η_{cathode}	the re-scaled cathodic over-potential	f_j^{max}	the upper bound of each objective function
η_{anode}	the re-scaled anodic over-potential	$\mu_{f_i(x)}$	membership function of each objective function
A	active area cell	$l_{\text{iter,max}}$	maximum number of iterations
I_{max}	maximum current density	l_{iter}	current iteration
P_{MFCFPP}	output active power of MFCFPP	A_m^k, r_m^k	Pulse loudness and emission rate for the m th bat in iteration k , respectively.
X	decision variable	A_m^k	Mean of the pulse loudness for all bats in iteration k .
t_{max}	total time	\mathbf{Gbest}^k	Best compromise solution in iteration k .
P_{Sub}^t	active power produced by the substation of network	f_m	Pulse frequency of the m th bat.
C_{Sub}	cost of substation active power	$f_m^{\text{min}}, f_m^{\text{max}}$	Minimum and maximum pulse frequency for bat m , respectively.
C_{n1}	price of purchasing natural gas for FCPPs	\mathbf{Mean}^k	Mean of the population positions.
N_{FCPP}	total number of FCPPs	\mathbf{Pbest}_m^k	Best position for bat m in iteration k .
$P_{\text{FCPP},j}^t$	active power generated by MFCFPP j during time t	\mathbf{Worst}^k	The worst solution among all bats in iteration k .
$P_{\text{HFCPP},j}^t$	equivalent electric power for hydrogen production during time t	N_{Bat}	Number of bats
$\eta_{\text{MFCFPP},j}^t$	efficiency of MFCFPP j during time t	$\mathbf{Vel}_{\text{new},j}^k, \mathbf{Vel}_j^k$	new and old velocities of the j th bat, respectively
C_{n2}	fuel price for thermal loads	$\mathbf{X}_{\text{new},j}^k, \mathbf{X}_j^k$	new and old positions of the j th bat, respectively.
N_{bus}	total number of buses	$\text{rand}(\cdot)$	random number between 0 and 1
L_{thi}	thermal load demand of bus i	w_k	weight of k th objective function
P_{thi}^t	heat produced by MFCFPP in bus i if there is a FCPP in this bus during time t	η_{st}	hydrogen storage efficiency
OM	operation and maintenance costs of FCPPs	η_{overall}^t	overall efficiency of MFCFPP
C_{pump}	hydrogen pumping cost		
$P_{\text{HFCPPUsage}}$	equivalent electrical energy of used hydrogen		
C_{HS}	hydrogen selling price		
$P_{\text{Hsave,FCPP}}$	equivalent electrical energy of saved hydrogen		
H_{factor}	a conversion factor (kg of hydrogen/kW of electric power), where $H_{\text{factor}} = 1.05 \times 10^{-8}/v_{\text{cell}}$ and v_{cell} is the cell operating voltage, $v_{\text{cell}} = 0.6 \text{ V}$.		
h	number of objective function		
locat_n	the location of MFCFPP $_n$		
$N_{\mu}(X_i)$	normalized membership value of each particle in the repository		
d	number of non-dominated solutions in the repository		
T	total time		
E_{grid}^t	emission produced by grid during time t		
$E_{\text{FCPP},j}^t$	emission produced by MFCFPP j units during time t		
$e_{\text{grid,SO}_2}$	emission coefficient of grid		
$e_{\text{grid,NO}_x}$	emission coefficient of grid		

List of abbreviations

MFCFPP	Molten Carbonate Fuel Cell Power Plants
CHPH	Combined Heat, Power, and Hydrogen
SALBA	Self Adaptive Learning Bat-inspired Algorithm
DG	Distributed Generation
RWM	Roulette Wheel Mechanism
MCS	Monte Carlo Simulation
SD	Standard Deviation
CF	Constant Frequency
FM	Frequently Modulated
SALM	Self Adaptive Learning Mechanism
PEM	2m + 1 Point Estimate Method
PDF	Probability Distribution Function
DLF	Deterministic Load Flow
MOP	Multi-objective Optimization Problem

for placement of DGs using a combination of technical factors such as power loss minimization and voltage sag reduction and economical factors such as costs of maintenance and amount of DGs in the system [11]. Optimal placement and sizing of DGs using a multi-objective methodology are offered in Ref. [12]. El-Zonkoly studied the optimal positioning of new resources of energy in distribution networks considering different load models [13]. In Refs. [14], optimal placement and sizing of DGs are addressed based on fuel cost

minimization, power loss reduction, and voltage profile improvement using a novel efficient population-based heuristic approach. Moradi and Abedini studied the problem of optimal placement and sizing of FCPPs using a combination of PSO and GA algorithms [15]. Niknam et al. formulated the optimal positioning of renewable energy sources in distribution networks in a Multi-objective Optimization Problem and used a modified honey bee mating optimization algorithm to obtain the solutions [16]. For location and size

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