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Reliability constrained decision model for energy service provider incorporating demand response programs



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HIGHLIGHTS

• The operation of Energy Service Providers (ESPs) in electricity markets

- is modeled.
 Demand response as the costeffective solution is used for energy service provider.
- The market price uncertainty is modeled using the robust optimization technique.
- The reliability of the distribution network is embedded into the framework.
- The simulation results demonstrate the benefits of robust framework for ESPs.

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ABSTRACT

Demand response (DR) programs are becoming a critical concept for the efficiency of current electric power industries. Therefore, its various capabilities and barriers have to be investigated. In this paper, an effective decision model is presented for the strategic behavior of energy service providers (ESPs) to demonstrate how to participate in the day-ahead electricity market and how to allocate demand in the smart distribution network. Since market price affects DR and vice versa, a new two-step sequential framework is proposed, in which unit commitment problem (UC) is solved to forecast the expected locational marginal prices (LMPs), and successively DR program is applied to optimize the total cost of providing energy for the distributed generation (DG) units, incentive cost paid to the customers, and compensation cost of power interruptions. To obtain compensation cost, the reliability evaluation of the distribution network is embedded into the framework using some innovative constraints. Furthermore, to consider the unexpected behaviors of the other market participants, the LMP prices are modeled as the uncertainty parameters using the robust optimization technique, which is more practical compared to the conventional stochastic approach. The simulation results demonstrate the significant benefits of the presented framework for the strategic performance of ESPs.

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Nomenclature

Indices		$\lambda_f^{failure/T}$	occurrence rate of fault <i>f</i> (failure rate) in the total time
d	DG units	J	periods
b	distribution network buses	PL_b	penetration level of active consumers at bus b
f	fault occurrences	ρ_{dt}^{DG}	power price of DG unit d at hour t
g	generator units	C_g	production cost of unit g
r	groups of the restoration load points for each fault	t ^{repair}	repair time of distribution faulty elements
	occurrence	C_{g}^{SD}, C_{g}^{SU}	shutdown/startup cost of unit g
а	responsive appliances	ρ_{h}^{ENS}	value of lost load (VOLL) at bus b
B_f^{dam}	set of buses which are the damaged points for fault f		
$B_{f,r}^{restore}$	set of buses which are the restoration points for fault <i>f</i> ,	Variables	
	and group <i>r</i>	$PD_{b,t}$	APL at bus <i>b</i> and hour <i>t</i>
$D_{f,r}^{restore}$	set of DG units which can supply the restoration load	$PD_{f \ b \ t}^{fault}$	APL at bus <i>b</i> and hour <i>t</i> during fault <i>f</i>
	points for fault <i>f</i> , and group <i>r</i>	$PS_{a,b,tt,t}$	APL of appliance a at bus <i>b</i> that is shifted from hour <i>t</i> to
g(n)	set of generator units at node <i>n</i>		hour <i>tt</i>
k(n, .), k	(., n) set of transmission assets with n as the "to" and	PSUB _t	APL which is fed through the substation bus at hour t
	"from" node, respectively	$PDG_{d,t}$	active power of DG unit d at hour t
t, tt	time (h)	$P_{g,t}$	active power of unit g at node n and hour t
k	transmission elements (line or transformer)	z_t, y_t	auxiliary variables
п	transmission node	S _{d,t}	commitment state of DG unit <i>d</i> at hour <i>t</i>
		$PC_{f,b,t}$	curtailed APL at bus <i>b</i> and hour <i>t</i> during fault <i>f</i>
Paramete	ers	Ζ, ζ _t	dual variables
AST_a	acceptable shift time for appliance <i>a</i>	ENS_b	ENS at bus b
$d_{n,t}$	active power load (APL) at node n and hour t	$P_{k,t}$	power flow for transmission element k, at hour t
$PNR_{b,t}$	APL of nonresponsive loads at bus <i>b</i> and hour <i>t</i>	$\rho_t^{g'rid}$	power price of the substation bus at hour <i>t</i> (LMP value
$PR_{a,b,t}$	APL of responsive appliance <i>a</i> , at bus <i>b</i> and hour <i>t</i>	, .	at node <i>n</i> *)
ρ^{inc}	incentive price	$W_{g,t}$	shutdown binary variable of unit g at hour t (1 for shut-
DTg, UTg	min down/up time of unit g	0.	down, 0 otherwise)
PDG_d^{\min} ,	PDG_d^{max} min/max capacity of DG unit d	$v_{g,t}$	startup binary variable of unit g at hour t (1 for startup,
Г	number of hours in which the uncertainty of electricity	<u>.</u>	0 otherwise)
	prices is considered	PS_{ht}^{tot}	total APL of bus b that is shifted to hour t
Т	number of time periods	5,0	
	-		

1. Introduction

1.1. Motivation and approach

Demand response (DR), as a key characteristic of the future smart grid, can provide several financial and technical benefits for electric power industries, such as, deferring capital intensive reinforcements [1], alleviating the need to use high-emission high-cost generating units [2], utilizing as ancillary services [3], tempering price spikes in the electricity market, mitigating the potential of market power [4], improving the operation of renewable energies [5,6], and cost saving in the electric bills of customers [7]. Moreover, it can be applied as a tool by energy service providers (ESPs) to manage the volatility of the electricity market prices [8].

In this work, DR is used to enhance the performance of ESPs in the smart distribution network. Proper applying DR problem provides special abilities for these entities, and could lead to more profit. On the other hand, participating in the upstream market and demand allocation in the downstream network are two main tasks of ESPs. These two tasks affect each other, and a simultaneous attention to them is needed for more efficiency.

To better understand, suppose that an ESP wants to reduce its energy purchase cost with applying DR. This entity initially forecast its load consumption, and participates in the electricity market. After market clearing, the values of locational marginal price (LMP) will be determined for the next 24 h. Now, applying DR and moving the load consumption to the less expensive hours will reduce the final purchase cost. However, moving the load consumption changes the LMP values in the substation bus of the distribution network. It is due to the dependencies between the load consumption and the prices. In this condition, the interactions between the load curve and price changes should also be considered. Disregarding the mentioned dependencies will limit DR capabilities.

Furthermore, delivering a high quality power with the least interruption to the customers is another challenge for ESPs, and ignoring this matter can increase the compensation costs. The cost of load shedding is the most important compensation cost which should be considered. DR implementation with changing the load curve can affect the service reliability of the distribution networks. Considering the reliability constraints in optimizing demand allocation can improve network efficiency, increase customer satisfaction, and decrease the final cost of ESPs.

Hence, a new two-step sequential framework is proposed in this paper, to enhance the performance of ESPs in the smart distribution network. The main problem includes a reliability constrained model to optimize the total cost of providing energy for the downstream network using DR. The subsidiary problem includes electricity market modeling. The load curve is determined in the main problem, and the amounts of the energy price under different conditions will be determined in the subsidiary problem, in a recursive manner. Such a framework guides ESPs to analyze how market clearing affects DR and vice versa.

Moreover, since the behaviors of other market participants are not fully predictable, the energy prices in each iteration are modeled as the uncertainty parameters. To this end, a robust approach as a recently gained substantial popularity method is applied. This approach has more advantages in practical applications, with respect to the stochastic methods. Robust model only needs moderate information about the uncertainty parameters, e.g., the range and the mean of the uncertain data (not an exact probability Download English Version:

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