

EV-Connected DC Microgrid

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EV-Connected DC Microgrids (DCMGs) have been preferred due to their lack of frequency stability issues, current harmonics, phase imbalances, and reactive power. In addition, since the process of integrating DG sources into a DCMG system is more straightforward in comparison to ACMGs, DCMGs are constructed more effectively. As a result, recent research interest in MGs has been primarily focused on the DCMG system.

centralized control

constraint of electricity price

DC microgrid

EV-connected microgrid

multi-objective problems

power flow control scheme

1. Introduction

An microgrid (MG) is an electric power system interconnected with renewable energy sources, energy storage systems (ESSs), and loads in a specific area. Based on the bus voltage type, MGs can be categorized into three categories: AC microgrids (ACMGs), DC microgrids (DCMGs), and hybrid AC-DC microgrids (AC-DCMGs).

An MG can operate either in the grid-connected mode or the islanded mode as an autonomous system ^[1]. By injecting or absorbing power from the MG based on electricity pricing policies, such as the real-time price (RTP), time-of-use (TOU) price ^[2], and stepwise-power-tariff (SPT) method, the grid can support an MG if it is available. TOU pricing divides electricity prices into peak-valley and peak-normal-valley periods. In contrast, the RTP divides electricity costs into high-cost and low-cost periods and has higher fluctuations in price than the TOU ^[3]. Meanwhile, the SPT method is employed for energy saving in China. The SPT method divides the monthly electricity price into several levels based on the electric power consumption ^[4].

As grid outages become more frequent and serious due to the increasing frequency of extreme weather conditions, the reliability of main grids will continue to reduce ^[5]. To maintain the power quality under these conditions, adequate coordination between MG systems and main grids is required. MG control schemes are divided into three categories in terms of coordinating power agents: the centralized control scheme, decentralized control scheme, and distributed control scheme ^[6]. The decentralized control method is known to have the highest flexibility and scalability among the three schemes. All power agents decide on their operating modes independently without information from the other agents in this scheme. However, it is difficult to achieve an optimum solution in this scheme due to the lack of coordination between the power agents ^[7]. In the distributed control scheme, the local information of MG agents is shared among their neighbors, which choose their operating modes based on this shared, external information as well as other internal information. Thanks to the existence of these communication

links, this scheme has the possibility to reach the local optimum solution. On the other hand, in the centralized control scheme, all the power agents communicate with the central controller (CC), and therefore all agents determine their operating modes according to data obtained from the CC. Amongst those three control schemes, only the centralized control scheme could achieve a global optimum solution under multiple constraints by constructing an effective energy management system (EMS) alongside the CC in the MG system [8].

In general, the EMS uses the prime objective function of the MG to decide the MG agent operating modes. In [9], the home economy is maximized using the convex programming for the smart home system with ESS and photovoltaic (PV) power generation using a 1-day time period. Total cost minimization for annual operation cost and annual investment cost of a smart home with ESS and PV has been researched based on the stochastic and deterministic methods by three different pricing methods using a 1-year time period in [10]. An EMS can also solve a multi-objective problem of the MG such as the combination of minimizing the energy cost and the peak-to-average ratio (PAR) in total load [11]. However, integrating EVs into MGs makes the objective function more complex. The optimization of EV batteries' lifespan [12] and EV charging schedules [13] are common examples of selecting the objective function to provide the benefit for the EV owners in the MG system connected with EVs. To utilize the flexible potentials of EVs, several researchers have developed a parking lot MG with EV connections [14][15]. A study in [16] investigates the coordination of integrated energy systems and an EV charging station in multiple stakeholder scenarios. Another study in [17] implements the multiple model predictive control (MMPC) methods in EVs to improve the frequency stabilization in the MG. The minimization problem applied to the total energy consumption cost is also investigated for vehicle-to-grid (V2G) microgrid systems considering a multi-objective function [18]. The research in [19] describes the co-benefit of EV and V2G systems. However, a typical EMS is designed to achieve only one objective function for either the MG or EV system. Thus, particular consideration is required in the MG system to select a preferable objective function.

Although several researchers have investigated the MG behavior caused by the EV connections, none of them have considered the uncertainties of EVs, DG power, electricity price conditions, and the unreliable grid connection altogether. Moreover, a straightforward algorithm is preferred to reduce the computational burden in the CC and implement it easily in the real system.

2. Configuration of EV-Connected DCMG System

Figure 1 shows a configuration of the DCMG system with EV and grid connections. **Figure 1a** shows a structure of a DCMG, in which the DCMG system is composed of different sources such as grid, battery, wind, load, and EV agents. The DCMG system in **Figure 1** can be extended into a larger system with multiple renewable energy sources, batteries, loads, and EVs. The reference direction of current is also shown in this figure, in which the current that flows into the DC-link is denoted as negative (–) current. In contrast, the current that flows out of the DC-link is denoted as positive (+) current. In **Figure 1a**, P_G denotes the power exchange between the DC-link and grid agent, P_B denotes the power exchange between the DC-link and battery agent, P_{EV} denotes the exchange power between the DC-link and EV agent, P_L is the power absorbed in the load agent, and P_W is the power generated by the wind power agent. A grid-connected inverter and transformer are installed to exchange the power

between the grid agent and the DCMG. A bidirectional DC-DC converter is employed in the battery and EV agent systems in order to connect those systems to the DC-link. The wind turbine is connected with a unidirectional AC-DC converter and a permanent magnet synchronous generator (PMSG) to inject the generated power into the DCMG system. Load shedding and reconnection are carried out in the load agent by switching the electronic switches.

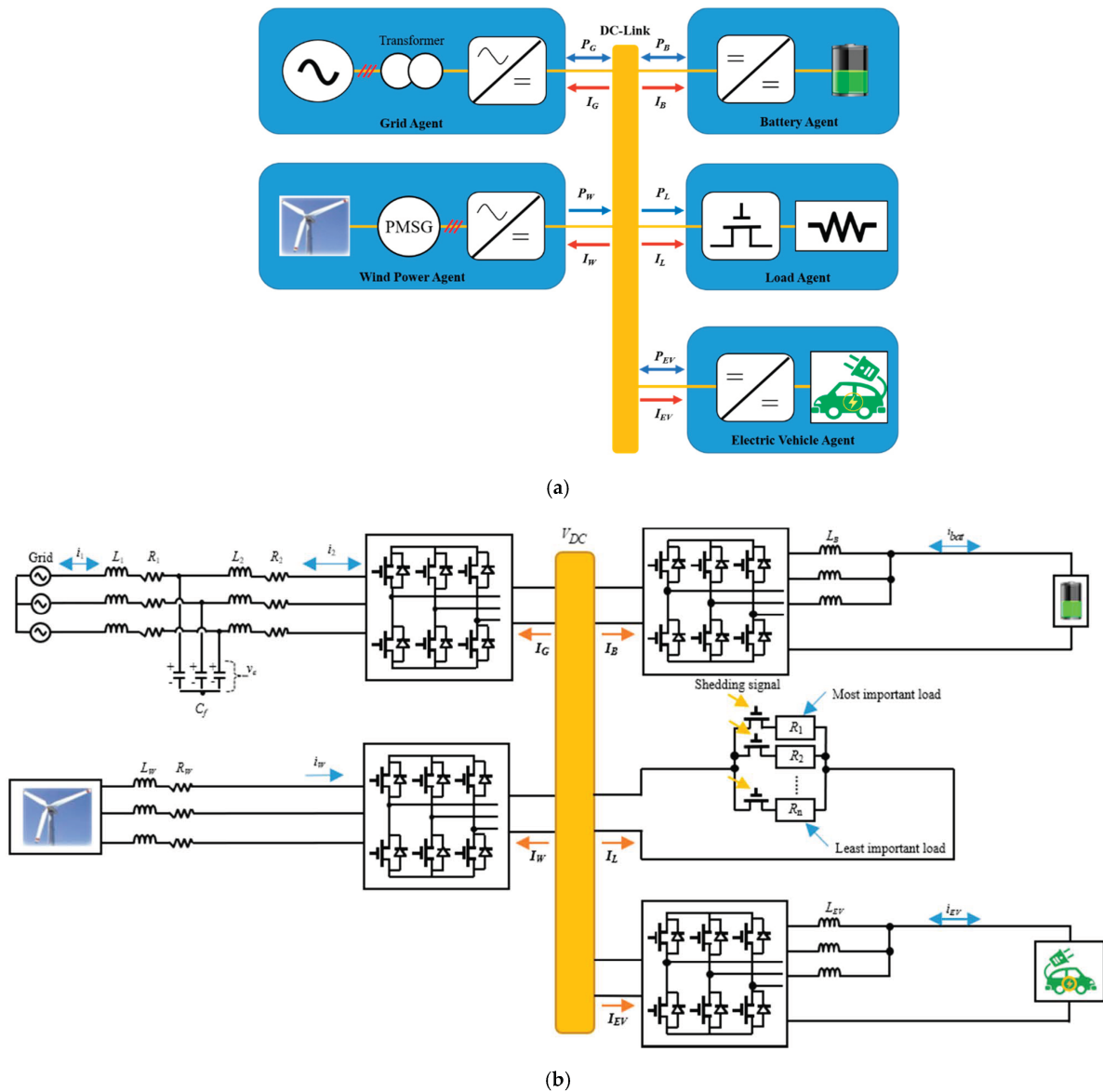


Figure 1. Configuration of EV-connected DCMG. (a) Simplified configuration of EV-connected DCMG; (b) detailed configuration of EV-connected DCMG.

Figure 1b shows a detailed configuration of the DCMG system with EV and grid connections. The grid-connected inverter uses an inductive-capacitive-inductive (LCL) filter to reduce the switching harmonics in the grid-side currents in the grid agent. The filter resistances, filter inductances, and filter capacitance are denoted by R_1, R_2, L_1, L_2 , and C_f , respectively. In addition, i_1 and i_2 represent the grid-side current and inverter-side current, respectively. In the grid-connected mode of the DCMG, the grid agent operates with three operating modes. When

the grid agent supplies the power from the main grid into the DCMG system to support the DC-link voltage, the grid agent operates in the DC-link voltage control converter (VDC CON) mode. The grid agent also works in the inverter mode (VDC INV), in which the grid agent regulates the DC-link voltage by injecting the surplus power from the DCMG system into the main grid. The remaining operation of the grid agent is the IDLE mode in which the grid agent and DCMG system are interconnected without any power exchange.

Two bidirectional interleaved buck-boost DC-DC converters are installed to connect the DCMG system with the battery agent and the EV agent through an inductive (L) filter. The battery and EV agents have five operating modes: two voltage control modes, two current control modes, and the IDLE mode. The DC-link voltage control by discharging (VDC D) mode is the first one, in which the battery and EV agents control the DC-link voltage by supplying the discharging power into the DC-link. On the contrary, the agent operating mode is the DC-link voltage control by charging (VDC C) mode, if the DC-link voltage is maintained by charging the battery or EV agent with the surplus power. Two current control modes are charging with the maximum allowable power (Max C) and discharging with the maximum allowable power (Max D). The last operating mode of the battery and EV agents is the IDLE mode in which the power does not exchange between these agents and the DC-link. In **Figure 1b**, L_B and L_{EV} represent the filter inductances of the battery and EV agents, respectively, and i_{bat} and i_{EV} denote the battery and EV currents, respectively.

An L filter is also employed in the wind power agent to connect the PMSG and unidirectional AC-DC converter. In **Figure 1b**, L_W and R_W are the filter inductance and resistance parameters, while i_W denotes the phase current of the PMSG. There are two operating modes in the wind power agent. The first operating mode is the maximum power point tracking (MPPT) mode, in which the wind power agent maximally extracts power from the wind turbine. The next operating mode is the DC-link voltage control (VDC) mode. This mode occurs in the condition that the generated power from the wind power agent P_W is greater than the demand load power P_L , and any other agents cannot absorb the surplus power of the DCMG system. In this situation, the nominal DC-link voltage should be maintained by reducing P_W .

Whenever the load power is higher than the available power in DCMG, the load agent commences load shedding (SHD) mode to prevent the DCMG system from collapsing by eliminating the unnecessary load. In this operation, the least important loads are first disconnected until the power balance in the system is guaranteed. When the grid is reconnected to the DCMG or the battery/EV agents have sufficient SOC level, shedded loads are reconnected by the load reconnection (RECON) algorithm. **Table 1** shows the summary of the operating modes of each power agent for the EV-connected DCMG. To ensure a DCMG operation, the reliable control of power electronics converters is an indispensable part. To effectively realize all the operating modes by each power agent in **Table 1**, power converter controllers are implemented by modifying the configuration in [8].

Table 1. Operating modes of power agents for EV-connected DCMG.

Power Agent	Operating Modes	Abbreviation
Grid agent	Idle	IDLE

Power Agent	Operating Modes	Abbreviation
	DC-link voltage control by converter operation	VDC CON
	DC-link voltage control by inverter operation	VDC INV
Battery agent	Idle	IDLE
	DC-link voltage control by charging	VDC C
	DC-link voltage control by discharging	VDC D
	Charging with maximum allowable power	Max C
	Discharging with maximum allowable power	Max D
EV agent	Idle	IDLE
	DC-link voltage control by charging	VDC C
	DC-link voltage control by discharging	VDC D
	Charging with maximum allowable power	Max C
	Discharging with maximum allowable power	Max D
Wind power agent	Maximum power point tracking	MPPT
	DC-link voltage control	VDC
Load agent	Load shedding	SHD
	Load reconnection	RECON

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