



Review

# Distributed Control Methods and Impact of Communication Failure in AC Microgrids: A Comparative Review

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Abstract: The objectives of this paper are to review and compare the distributed control methods in AC microgrids and also to identify the impact of communication failure on this type of the controller. The current AC microgrids are distinguished from the traditional power system topologies because of the high penetration of advanced control methods, measurements, sensors, power electronic devices, and communication links. Also, because of the increasing integration of renewable energy sources, control strategy for congestion management, frequency control, and optimal dispatch of microgrids has become more complicated. This paper explains the characteristics and features of distributed control systems and discusses the challenges of these approaches. In addition, a comprehensive review of the advantages and disadvantages of these techniques are explained in detail. On the other hand, the possible challenges, related to communication failure, noise, delay, and packet dropout on the operation of the distributed controller are presented, and several techniques, which reduce the impact of communication failure of the distributed controller, are compared. This comprehensive study on distributed control systems reveals the challenges in and future possible studies on this issue.

Keywords: microgrid; distributed control; renewable energy; secondary control; communication link

# 1. Introduction

Traditional power systems are legacy systems with old technologies and components from as far back as 30–35 years ago. Fossil fuel-based energy sources, such as gas, coal, and diesel, are the main energy resources of traditional generators in the power system, and by enhancing the load demands, fossil fuel energy production leads to air pollution [1–3]. Thus, developing a renewable energy sources (RES)-based system is essential to provide reliable, low pollution, and costless energy production [4–6]. A microgrid is an interface between RES (such as wind turbine (WT), fuel cells, and photovoltaics (PV)) and grid, which provides electricity for loads directly [7]. The microgrid is a low voltage (LV) system, which includes RES, energy storage system, loads, and converters; also, this system can be operated as islanded and in grid-connected mode [8]. The nowadays generation of the power system is microgrids with a significant enhancement in the level of coordination, control, communication, and monitoring [9]. A general structure of an AC microgrid is depicted in Figure 1. As shown in Figure 1, WT and PV produce unidirectional power, but power of the grid and battery is bidirectional; this is because of that, during the low production of power by WT and PV, the grid and battery should inject power to load sides. In addition, the AC/AC convert of the grid has a switch, which determines the operation mode of the microgrid.

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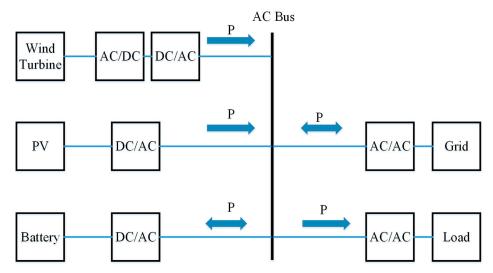


Figure 1. The general scheme of an AC microgrid. Abbreviations: AC; DC; P; PV, photovoltaics.

Microgrids provide many advantages for costumers and grid; however, there is a significant challenge in the control of microgrids since the number of RES is typically high. For instance, the islanded microgrid of [10] includes 47 WT and 5 combined heat and power plants, which shows that the number of controllable units is considerable. Therefore, centralized control methods are unable to operate microgrids with a high number of RES. In this regard, some reasons for controller failures are as follows:

- 1. The level of security and reliability of centralized controller schemes [11].
- 2. A difficulty in the redesign of a controller due to the change of even one RES.
- 3. High computational burdens due to the high number of under control units.
- 4. Unavailability for separate management units.

The task of a microgrid control system is divided into three different levels: (1) Primary control: Frequency, voltage, and current control; (2) secondary control: Optimal dispatch and frequency restoration; and (3) tertiary control: Coordination of grid and microgrid. Consequently, distributed or decentralized methods are the best option for providing required functionalities. In decentralized control schemes, interactions between subsystems are neglected; however, this assumption is invalid and results in poor system-wide performance. For example, in the report mentioned in [12], problems of this type of controlling scheme are presented. During instability cases, each controller tries to maintain its stability, which causes a cascading tripping event.

On the other hand, interactions between units are considered in distributed control methods. The concept of hierarchical controlling is based on spreading the task of controlling to different units, which are known as primary, secondary, and tertiary controllers. Due to requiring higher security and reliability, the requirement of distributed control schemes arises. Moreover, a microgrid is a multiagent system, in which every RES in this system is an agent, and each agent is connected to other agents through a communication link. The structure of agents can be determined by a graph. The graph of the agents makes a spanning tree, and the distributed control system controls the microgrid during topology changes until the graph of agents remains a spanning tree. Therefore, in a distributed system with multiagents, by removing or adding a plugin and plug out of the agents, the system can be restored.

One of the main parts of a distributed control system is the communication infrastructure. However, the majority of existing researches have assumed a perfect communication for these systems. In real situations, communication systems may face a failure, which causes a malfunction in the system [13]. Reference [14] uses storage for storing all excess energy during communication restoring, but it is only used for short-term interruptions. In [15], a damping controller is used for minimizing

the impact of the noise in the communication channels; however, it only reduces the noise related to oscillations and cannot reduce the effect of total communication losses.

The objective of this paper is to provide a comprehensive study on the existing distributed control schemes for microgrids. Also, this paper defines and presents future researches and works related to this issue. Moreover, in this paper, several methods, which reduce the effect of communication failure on distributed control approaches, are reviewed. There are several articles that provide a review of the control schemes of AC converters [16], microgrids [17], and smart-grid [18]. However, there is a difference between this paper and those mentioned previously. The current paper, specifically, studies distributed schemes and methods, which reduce the effect of communication loss on these types of controllers employed in a microgrid.

The rest of the paper is organized as follows: Section 2 discusses the components of microgrids and their differences from power systems. Section 3 explains the distributed control scheme, and different distributed control methods are reviewed and compared. Section 4 discusses the impact of communication failure on controller functionality. Furthermore, some techniques, which reduce the impact of communication failure on distributed controllers, are defined and compared. In the last section, the conclusion is discussed.

# 2. Structure of a Microgrid and Modeling of RES

An AC microgrid consists of RES, loads, converters, and energy storage devices. This type of electrical system is part of the distribution system, which can be operated as grid-connected and in islanded mode. In this section, the main components of an AC microgrid and the modeling equations of these components are accurately discussed and determined.

#### 2.1. RES

The most common RES in microgrids are PV, WT, and fuel cells with low power level, less than 200 kW, which use power converters for connecting to the main bus [19]. Because these RES are installed near the loads, they should have low emission and noise performance, and they should also provide highly reliable and low-cost power. The selection of the appropriate RES for a microgrid is a complex issue. The availability of a main source of power, cost, and output power are key factors which should be considered [20].

In this paper, the modeling of PV and WT have been explained accurately, because the accurate modeling of RES is essential for designing a controller. Performance of a controller in a microgrid should be designed under different conditions. One of these conditions is fault, in which, during the fault, the controller should limit and control the output of each RES. Therefore, in the first step, the transient behavior of RES should be taken into account, but a suitable model should ensure high accuracy and computational speed. Therefore, the modeling of PV and WT based on this purpose is described as follows.

The PV system includes an output (Inductance (L) – Capacitance (C) – Inductance (L)) LCL filter, converter, and PV panels. The accurate two-diode model of a PV panel is shown in Figure 2, and the related equations are as follows [21]:

$$I_{out} = I_{PV} - I_{01} \left[ \exp\left(\frac{q(V_{out} + IR_S)}{a_1kT}\right) - 1 \right] - I_{02} \left[ \exp\left(\frac{q(V_{out} + IR_S)}{a_2kT}\right) - 1 \right] - \frac{V_{out} + IR_S}{R_P}$$
(1)

where  $I_{out}$  is output current,  $I_{PV}$  is photocurrent,  $I_D$  is diode current,  $I_{01}$  and  $I_{02}$  is saturation current of first and second diodes, respectively, k is Boltzmann's constant, T is the junction temperature,  $\alpha_1$  and  $\alpha_2$  is ideality factor of the first and second diodes, respectively,  $V_{out}$  is voltage of PV,  $R_s$  is series resistance,  $R_P$  is parallel resistance, and q is charge of electrons. Therefore, because the aim of the majority of controllers is controlling the output current, the mentioned equation is used for this purpose. On the other hand, some controllers use dq environment to design the controllers; thus, the modeling of a three-phase PV panel in term of dq can be represented by [22]:

$$\dot{L}_{1i}\dot{I}_{1di} = -R_iI_{1di} - \omega_iL_{1i}I_{1qi} + R_iI_{2di} - V_{cfdi} + v_{pv}K_{di}$$
(2)

$$L_{1i}\dot{I}_{1qi} = -R_{i}I_{1qi} - \omega_{i}L_{1i}I_{1di} + I_{2qi} - V_{cfqi} + V_{out}K_{qi}$$
(3)

$$\dot{L_{2i}I_{2di}} = -R_iI_{1di} - R_iI_{2di} - \omega_iL_{2i}I_{2qi} + V_{cfdi} - V_{di}$$
(4)

$$\dot{L_{2i}I_{2qi}} = -R_iI_{1qi} - R_iI_{2qi} + \omega_iL_{2i}I_{2di} + V_{cfqi} - V_{qi}$$
(5)

$$C_{fi}\dot{V}_{cfdi} = -\omega_i C_{fi} V_{cfqi} + C_{fi} (I_{1di} - I_{2di})$$
(6)

$$C_{fi}\dot{V}_{cfqi} = \omega_i C_{fi} V_{cfdi} + C_{fi} (I_{1qi} - I_{2qi}) \tag{7}$$

$$C_{fi}\dot{V}_{out} = I_{PVi} - I_{1di}K_{di} - I_{1qi}K_{qi}$$
 (8)

where the subscripts of d and q are for direct and quadrature components,  $K_{di}$  and  $K_{qi}$  are the inputs of the switching signals,  $\omega$  is the angular frequency,  $I_{1i}$  and  $I_{2i}$  is the output current of inverter and filter, respectively,  $V_{cf}$  is the voltage of the capacitor of filter,  $C_{fi}$  is the capacitance of the filter,  $C_{pi}$  is the DC capacitance, L is the filter inductance, and  $R_i$  is the resistance of the filter.

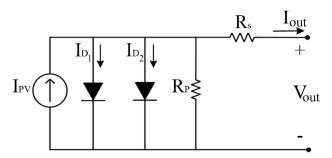


Figure 2. Two-diode model of a PV array.

WT is the one of the main RES in microgrids; therefore, modeling of WT is an essential part of designing a controller. The transient impedance of WT is calculated by [23]:

$$Z' = r_s + jx' = r_s + j\omega_0(L_h + L_S - \frac{L_h^2}{L_h + i_R})$$
(9)

where the subscripts of R and S represent the rotor and stator, and x, r, L are reactance, resistance, and inductance, respectively. The subscript of h shows the main field,  $\omega$  is the angular speed, and Z' is the transient impedance. The value of the source of voltage is obtained as follows:

$$u' = j\omega_0 k_R \psi_R \tag{10}$$

where  $\psi$  is the complex flux leakage, and k is frame of arbitrary reference. The unknown parameters are calculated by the following differential equations:

$$\frac{d\psi_{Rd}}{dt} = -\frac{r_R}{L_R}\psi_{Rd} - (\omega_R - \omega_0)\psi_{Rq} + k_R r_R i_{Sd} + u_{Rd}$$
(11)

$$\frac{d\psi_{Rq}}{dt} = (\omega_R - \omega_0)\psi_{Rd} - \frac{r_R}{L_R}\psi_{Rq} + k_R r_R i_{Sd} + u_{Rq}$$
(12)

$$\frac{d\omega_R}{dt} = \frac{1}{\theta_m} (k_R (\psi_{Rd} i_{Sq} - \psi_{Rq} i_{Sd}) + t_m)$$
(13)

where  $\theta$  is complete rotor shaft inertia, and the subscripts of q and d is quadrature and direct axis component, respectively.

#### 2.2. Power Converter

The performance and operation of the power converters in microgrids during fault cause an uncontrolled current by RES and discharge current by capacitors. The modular multilevel converter (MMC) is the most applicable type of converter in microgrids. This type of converter consists of two-level half-bridge converters as controllable submodules. In MMCs, the fault current characteristics are affected by two different resources. Capacitors are distributed as series submodules, and submodules block (Insulated Gate Bipolar Transistor) IGBT discharges through the capacitors. Second, converters include series reactors, which, during the capacitor discharge, restrict the fault current increasing rate.

During controller design, a simplified model of converters reduces the computational time. Therefore, an equivalent (Resistant-Inductance (L)-Capacitance) RLC model is suggested in [24], as depicted in Figure 3. A technique to model the effect of converter capacitors is to consider the reducing trend of the converter voltage. An MMC has several arms and submodules. Thus, the maximum amplitude of the generated voltage by one converter's arm is the total of the available arm voltage. The arm voltage variations of a converter in terms of the current can be obtained by:

$$\begin{cases}
\frac{du_u}{dt} = \frac{n_u i_u}{C_{arm}} \\
\frac{du_L}{dt} = \frac{n_L i_L}{C_{arm}}
\end{cases}$$
(14)

where  $u_L$  and  $u_u$  are the bottom and top voltage of the arm, respectively,  $n_u$  and  $n_L$  are the numbers of submodules in the upper and lower of the arm,  $i_L$  and  $i_u$  are the bottom and top current of the arm, respectively, and  $C_{arm}$  is equivalent series capacitance. The voltage of the one module is obtained by summation of the voltage of the lower and upper arm. Moreover, the upper and lower arm current are equal. Thus, the capacitor is calculated by:

$$\frac{du_{leg}}{dt} = \frac{i_{leg}}{C_{arm}} \tag{15}$$

where  $i_{leg}$  and  $u_{leg}$  are the arm voltage and current after the fault, respectively. The value of model resistance is obtained by:

$$R = n_u \cdot R_{IGBT} + (N - n_u) \cdot R_d + n_L \cdot R_{IGBT} + (N - n_L) \cdot R_d \tag{16}$$

where N is the number of legs. The resistance of the lower diode is  $R_d$ , and resistance of the IGBT is  $R_{IGBT}$ .

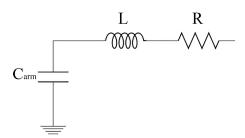


Figure 3. RLC model of converters.

# 2.3. Energy Storage

In microgrids, due to the uncertainties of RES, the battery is one of the main components of each microgrid. The equivalent circuit of batteries includes capacitors, resistors, and a DC voltage source for providing accurate dynamic model. This model is depicted in Figure 4 by suitable complexity and accuracy for electrical studies.

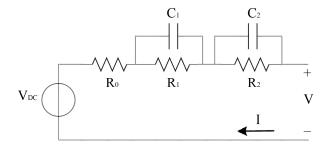


Figure 4. Equivalent model of battery.

The paralleled (Resistant Capacitance) RC is the response of battery nonlinear polarization. The value of  $V_{DC}$  is the open-circuit voltage of the battery,  $C_1$  is polarization capacitance,  $R_1$  and  $R_2$  are polarization resistances, and  $R_0$  is the battery connection resistance.

## 3. Distributed Control Scheme for AC Microgrids

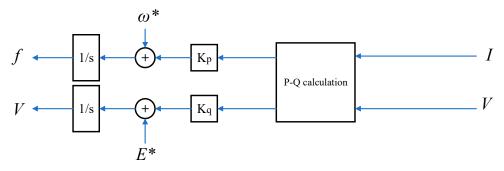
The main challenge of distributed control methods is to solve the optimization problem by a communication-based distributed procedure. Thus, due to the different problem definitions, the diagnosis between different distributed methods is not clear. In this section, different distributed control techniques are summarized.

### 3.1. Droop-Based Methods

Droop based methods are a widely used approach [25–27]. Droop principle is emanated by the concept of synchronous generator power balances. An unbalance between the output electrical and input mechanical power makes a change in the frequency of system and rotor speed. Moreover, variations in the reactive power of output cause a deviation in the voltage amplitude. Similarly a characteristic can be artificially made for electronically interfaced RES. In this technique, the equations of active power and frequency and reactive power and voltage are obtained by:

$$\begin{cases}
f = f^* - K_P(P - P^*) \\
V = V^* - K_Q(Q - Q^*)
\end{cases}$$
(17)

In [28], droop control-based distributed control for a microgrid is suggested by using the active and reactive relationships of Equation (17) for an LV AC microgrid. The main benefits of this controller are using localized measurements instead of communication links. However, it causes several disadvantages for the microgrid, such as inability to following a fixed frequency, unsuitability for nonlinear loads, inability to control accurately the power-sharing during uncertainties, poor performance in high penetration of RES, inability to black startup, and poor transient characteristic, which restrict the application of it on modern microgrids [28–30]. Recently, in [31], a washout filter-based technique is proposed for improving the steady-state characteristic of this type of controller. The general structure of droop control-based methods is shown in Figure 5.



**Figure 5.** The general structure of droop control-based methods.

#### 3.2. Distributed Model Predictive Control (MPC)-Based Methods

The MPC is a standard for controlling large systems [32–34], which provides explicit consideration of constraints, is easy to tune, and handles multivariable control problems. An MPC control technique is shown in Figure 6, which shows a discrete-time control scheme by using the minimizing of a fitness function associated with the characteristics of a system over a finite number of steps by using the model of the system [35]. The fitness function of this method is a combination of minimization terms and the system setpoints. In a linear MPC, the linearized model of a discrete-time system is occupied. Each step time consists of a calculation of the control sequence for a prediction horizon future steps for minimizing the errors.

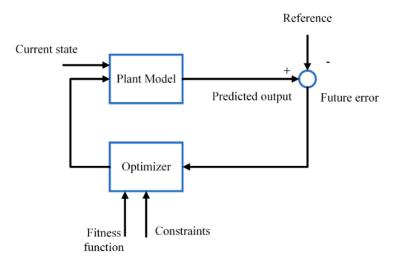


Figure 6. Model predictive control strategy.

#### 3.3. Consensus-Based Methods

Recently, the literature has studied the distributed controller by using consensus, which is a method for solving the mentioned problem and provides a flexible formulation that offers scalability and extendibility [36]. The main aim of the consensus-based methods is to provide a convergence for different RES in microgrid to a single value. A consensus-based technique attains to a global optimal value by using a communication link between adjacent RES, without requiring any proprietary units. In [37–39], a study on unconstrained consensus strategy for a distributed control of microgrid is performed. The principles for defining the consensus method is based on the following equation and two theories:

$$\dot{x}_i = u_i = \sum_{j=N} a_{ij} (x_j - x_i)$$
 (18)

where  $a_{ij}$  is the element of adjacency matrix of the graph, and x is the graph nodes. Also, any scalar value of graph nodes should satisfy the principle of distributed consensus during a continuous range of time. The first theory explains that a communication graph consisting of a spanning tree causes a consensus control and all eigenvalues have zero or positive parts. The second theory claims that if a communication graph consists of a spanning tree, all agent states will converge to the external control signal. These two theories are often used in the consensus control for analyzing microgrid stability.

## 3.4. Agent-Based Methods

Agent-based or multiagent methods are another distributed control technique. Agents can have communication ability and limited knowledge of the system [40]. An intelligent agent tries to modify the environment, relies on communication, and requests initiatives. An agent-based system is especially appropriate for microgrids, in which a large number of agents of different types of RES and local information is available. In addition, agents are categorized as follows:

- 1. Hierarchical agents include different agent layers.
- 2. Decentralized agent structure includes several localized agents.
- 3. Centralized agent structure includes a single agent.

The data flows from lower to higher layers, and the demand flows from higher to lower layers. Moreover, the flowed data in layers cause communication of each agent [41]. Because microgrids require to operate separate to the grid, the multiagent system can be used for control of them. A review of multiagent systems and applications of them are reviewed in [42]. Simulation of agent-based microgrids requires combining a communication backbone with a microgrid simulator. Recent research using PSCAD [43] and C language [44] for communication is proposed in [45]. A neighbor-to-neighbor communication strategy in [46] is presented for implementing a multiagent system control in a microgrid, and the aims of the controller are achieving power balance and restoring voltage to keep the stability of the microgrid. A multiagent hierarchical controller is developed in [47] to provide both hybrid dynamic behavior and hierarchical hybrid control for a microgrid. This control strategy minimizes costs and maintains the voltage stability of a microgrid. Figure 7 depicts an example of the general structure of a hierarchical agent controller.

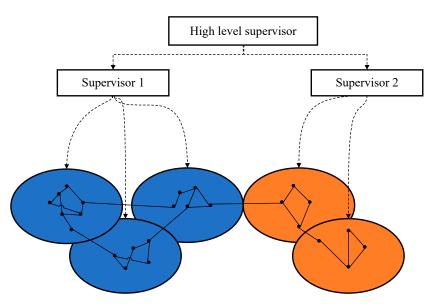


Figure 7. General structure of a hierarchical agent architecture.

## 3.5. Decomposition-Based Methods

In [48–55], several decomposition techniques, such as the alternating direction, predictor-corrector proximal multiplier, and auxiliary problem principle technique, are proposed. These methods divide the optimization problems into subproblems, which are solved interactively. These techniques decompose the optimization problems into different defined areas, such as based on the controllability and sensitivity factors of different locations. The concept of the decomposition-based approach is based on the decomposition of an optimization problem into different subproblems. This method assigns each variable and constraint to one subproblem, and sets the not decision variables into the considered subproblem to fixed values. Therefore, the overall structure of problem is as follows:

$$\begin{cases}
\min f(x_{1}, \dots, x_{m}) \\
s.t.g_{p}(x_{p}) = 0 \\
h_{p}(x_{p}) = 0 \\
g_{p,int}(x_{1}, \dots, x_{m}) = 0 \\
h_{p,int}(x_{1}, \dots, x_{m}) = 0
\end{cases}$$
(19)

where *g* and *h* depend on the decision variables from multiple coupling constraints.

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#### 3.6. Finite-Time Convergence Control Methods

The finite-time convergence control techniques are researched for achieving a finite settling time for secondary control of microgrids [56–59]. Reference [56] suggests a finite-time controller method for voltage and frequency restoration with accurate power-sharing. Reference [60] proposes a finite-time convergence control method for synchronizing the frequency and voltage during the islanded mode of a microgrid, and changing of communication topologies. Moreover, Reference [61] suggests a robust finite-time control method for voltage and frequency control and regulation in an autonomous AC microgrid with high penetration of RES. This controller is robust under various types of disturbances and faults. However, there is a lack of enough study on the finite-time convergence controllers with bounded control input constraint. A finite-time convergence controller with bounded control input is presented in [62] to restore the frequency. In this method, a saturation function is pursued, avoiding the control input bound exceeding. By using this controller, the frequency of RES is restored to the rated frequency during the islanded mode, and also transient peak shaving and accurate power-sharing are achieved. Studying the controller with bounded input is valuable due to the practical constraint and applications.

## 3.7. Aperiodic Sampling Data-Based Control Methods

Aperiodic sampling data-based controllers are divided into two different sections: Extrinsic and intrinsic. In the extrinsic aperiodic method, the sampling period modifies randomly, because of communication links and hardware impact on it [63]. In contrast, in the intrinsic aperiodic technique, the sampling time interval includes self-triggered and event-triggered methods, which is created by the controller. In the self-triggered approach, based on the last triggered data and dynamic of system, the next sampling time is predicted at the control updates [64]. On the other hand, the event-triggered technique requires some specific condition to trigger the sampling, and also continuous monitoring [65]. In [66–69], the application of an intrinsic aperiodic sampled data-based controller in power systems is suggested; however, there is lack of enough study on the application of this type of controller in microgrids. In [70], a time-delayed and aperiodic sampled data-based distributed controller in a microgrid is proposed to restore voltage and frequency.

## 3.8. (Radial Basis Function) RBF Neural Network Sliding Mode Control Methods

The RBF neural network sliding mode distributed controller in a microgrid is suggested in [71]. The RBF neural network sliding mode distributed controller incorporates the advantages of the sliding mode and neural network into a distributed controller. The target of the controller is converted from error to function of sliding mode [72]. Instead of using the error for the input of the controller, its input is sliding mode function. When the sliding mode function reaches zero, the tracking error will reach zero. Also, the RBF neural network sliding mode distributed controller does not require accurate mathematical data of the system, and it softens the signal of control for reducing the chattering phenomenon which is made by general sliding mode control. Thus, the RBF neural network sliding mode distributed controller is designed as depicted in Figure 8.

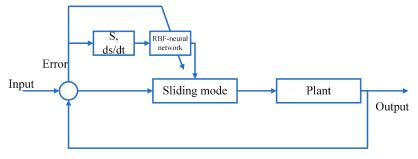


Figure 8. RBF neural network sliding mode controller.

#### 3.9. Discussion on the Distributed Control Methods

In this paper, the main strategies for distributed control are reviewed and discussed. The main challenges of these methods are as follows.

- 1. Results convergence: In distributed control strategies, each controller optimizes the fitness function based on the system for only its own control inputs.
- 2. Stability analysis: Analyzing the stability increases the safety of the design. It is important to implement the results from the control area to microgrid applications.
- 3. Cybersecurity: One of the main reasons for increasing the use of distributed strategies is the unwillingness of stakeholders to share data with other systems against cyberattacks. The communication system can be faced with cyberattacks; thus, increasing the security of these systems against cyberattacks is an important challenge.
- 4. Requirements: The slow and minimal communication requirements for distributed controllers are in contrast with the existing standards in microgrids for avoiding inherent uncertainties. Future researches should focus more on communication-less systems.
- 5. Regulatory consideration: Exchanging data in the distributed control systems is not standard protection in the microgrids.
- 6. Methods independent of the model: Distributed control methods are independent of the detailed system model, and robust during topology changes, in which the status of RES can be changed frequently and impact on the performance of the system.

The comparison between the presented methods is summarized in Table 1. Based on this comparison, some methods, such as consensus-based or agent-based methods, are dependent on communication channels, which increases the sensitivity of the system to communication failure. Therefore, aperiodic sampling data-based control methods, which are immune to disturbances, are suitable for distributed control systems. On the other hand, in systems with high reliable communication channels, due to the high number of advantages of agent-based methods, these methods will make a effective control system for a microgrid.

**Table 1.** The summary of the distributed controllers.

Distributed Method	Advantages	Disadvantages
Droop-Based Methods	<ul> <li>Localized measurements</li> <li>Easy to use</li> <li>Low cost</li> <li>Applicable for both grid-connected and islanded modes</li> </ul>	<ul> <li>Inability to following a fixed frequency</li> <li>Unsuitable for nonlinear loads</li> <li>Inability to control accurately the power-sharing during uncertainties</li> <li>Poor performance in high penetration of RES</li> <li>Inability to black startup</li> <li>Poor transient characteristic</li> </ul>
MPC Control-Based Methods	<ul> <li>Explicit consideration of constraints</li> <li>Easy to tune</li> <li>Handling of multivariable control problems</li> <li>Applicable for both grid-connected and islanded modes</li> </ul>	<ul><li>Cannot tune offline</li><li>Cannot be used in uncertain systems</li></ul>
Consensus-Based Methods	<ul><li>Scalability</li><li>Extendibility</li><li>Applicable for both grid-connected and islanded modes</li></ul>	<ul> <li>Communication-based</li> <li>Costly</li> <li>Sensitive to communication failure</li> <li>Time consuming</li> </ul>

Table 1. Cont.

Distributed Method	Advantages	Disadvantages
Agent-Based Methods	<ul> <li>Minimize cost</li> <li>Effective in a microgrid with a large number of RES</li> <li>Effective in autonomous microgrids</li> <li>Applicable for both grid-connected and islanded modes</li> </ul>	<ul> <li>Rely on communication</li> <li>Sensitive to communication failure</li> </ul>
Decomposition-Based Methods	<ul> <li>Increase the convergence of the optimization problem</li> <li>Improve the value of error</li> <li>Improve the controllability of system</li> <li>Applicable for both grid-connected and islanded modes</li> </ul>	<ul><li>Slow</li><li>Require high number of data</li></ul>
Finite-Time Convergence Control Methods	<ul><li> Effective during disturbances</li><li> Consider the system transients</li></ul>	<ul><li>Ineffective in bounded control input</li><li>Only for islanded mode</li></ul>
Aperiodic Sampling Data-Based Control Methods	<ul> <li>Immune against disturbances</li> <li>Consider the system dynamics</li> <li>Applicable for both grid-connected and islanded modes</li> </ul>	<ul><li>Require continuous monitoring</li><li>Require continuous sampling</li></ul>
RBF Neural Network Sliding Mode Control Methods	<ul> <li>Lowest error</li> <li>Do not require mathematical data</li> <li>Soften the signal of control</li> <li>Applicable for both grid-connected and islanded modes</li> </ul>	<ul><li>Complex</li><li>Low speed</li></ul>

## 4. Communication Degradation Processes

In the majority of presented distributed controllers, the communication infrastructures are assumed perfect; however, this is not valid in practical cases. In real cases, communication failure is the most critical issue which can cause malfunction to operate secondary controllers. In this section, the impact of communication link degradation processes, such as noise, packet dropout, delay, and communication failure, are discussed. In addition, the techniques for reducing the impact of communication degradation processes on the controllers are reviewed.

# 4.1. Delay

Most studies assume the communication delays between the controllers and agents are negligible. However, due to the vast application of communication links consisting of WiFi, WiMax, Internet, and Ethernet in the smart grid, microgrid, and power controllers, the signals between these units may be delayed during their transmission [73–77]. The characteristics of network-induced time delays are random, bounded, and constant, which depend on the internal performance of the communication system [78]. In addition to time delay, the operation of controllers in microgrids can be affected by the sampling rate of these communication systems. However, the time delay caused by the sampling rate is much less than other delays; the sampling rate of the power line carrier is between 9600 to 19200 b/s, WiFi Ethernet has Gigabit bit rates, and Ethernet has Megabit data rates. Moreover, the required data of the secondary controller are around 100 bytes [79]. Therefore, the delay of communication infrastructure is an essential part of the operation of controllers. In fact, one of the inherent characteristics of communication infrastructures is time delays. Communication delays can be divided into two groups: Input and communication delay. Communication delay includes the transmitted data from an agent to another agent, whereas the information is received from neighbor agents [80]. Communication delays cause poor performance and instability for the dynamic of microgrids. This determines that it is essential to eliminate the impact of communication delays in distributed controllers.

The characteristic equation of a system can be defined as follows:

$$\begin{cases} \det(\lambda E - \Delta(\lambda, \tau)) = 0\\ \Delta(\lambda, \tau) = A + A_{\tau} e^{-\lambda \tau} \end{cases}$$
 (20)

where  $\tau$  is time delay, A is the system matrix, and E is the singular system matrix. Also, if all generalized roots of Equation (20) are in the open left-hand plane, for given value of  $\tau$ , the delayed description of the system in the system model will be stable. In [81], a method for determining the margins of delay for microgrids is presented. In this paper, the extended eigenvalue method in [82] is discussed. In the following equations,  $\lambda_{img} = \pm j\omega$  donates generalized conjugate eigenvalue pairs on the imaginary axis.

$$j\omega = eig(\Delta(\omega, \tau)) \tag{21}$$

where the eig(f) is the equivalent of f. By defining  $\eta = \omega \tau$ , Equation (20) can be rewritten as follows:

$$\Delta(\eta) = A + A_{\tau} e^{-j\eta} \tag{22}$$

The value of  $e^{-j\eta}$  changes by the period of  $2\pi$ , and the value of  $\eta$ . Therefore,  $\Delta(\eta)$  modifies by period of  $2\pi$ . By changing  $\eta$  between 0 to  $2\pi$ , the root locus of the  $\Delta(\eta)$  can be obtained. Therefore, when the existing eigenvalues of  $\pm j\omega_c$  stand on the imaginary axis, the critical time delay or  $\tau_c$  can be calculated by:

$$\tau_c = \frac{\eta_c}{\omega_c} \tag{23}$$

The margins of delay by considering the different values of gains are evaluated in [83] by using the explained method. The results show that the delay margin enhances by increasing the proportional gain and decreasing the integral gain.

Consequently, in some literature, methods for solving the aforementioned problems are presented. In [83], the effect of communication delay on the operation of the secondary controller in islanded microgrids is evaluated. The relationships between the secondary frequency control gain and the communication delay margins are investigated by using small-signal analysis. For eliminating the impact of communication delay on the microgrid, a gain-scheduling frequency control technique is proposed for islanded microgrids. The conditions of delay stability are derived in [84] by using Razumikhin stability theory to design a delay independent distributed controller. Also, the stability of the delay dependency on the microgrid is studied by linear matrix inequality. In [85], the stability of the multiagent distributed controller is analyzed by absolute damping under communication delay, and a frequency domain analysis technique is applied for ensuring the achievement of formation control.

# 4.2. Noise

The communication links between inverters are assumed to be noise-free in the existing distributed control methods in microgrids. However, in practical situations, noise affects the communication link. Also, some environmental events can cause noise in the communication links. In the wireless-based communication links, during the generation of noise in the receiver front end, the antenna picks up the surrounding noise [86]. In the presence of the zero-mean white noise model, large deviations with a nominal threshold for noise mode cause circulating current and destabilize the microgrids, and small deviations affect sensitive electronic devices [87]. Due to using small-signal modeling for distributed control methods, this controller cannot respond during large-signal disturbances [88–90]. In addition, microgrids are structurally unknown and parametrically uncertain. In [91], the system uncertainties and unknown disturbances are considered in the design of the distributed control system. The impact of communication noise on the distributed estimators in a multiagent system is discussed in [92]. In [93], it is assumed that the data transmitted by communication channel are corrupted with generalized Gaussian noise. Moreover, in this paper, the complete nonlinear model of a microgrid is considered to

design a robust distributed control method with an additional control input to counterbalance and correct the noise impact.

## 4.3. Packet Dropout

Packet dropout means lost data or information in the communication channel, which occupy the bandwidth of the network but cannot transmit to the destination. Thus, It effects on the operation of microgrids and RES, and the reduction on fluctuations of frequency. In [94], packet dropout is explained by using Bernoulli-distributed variables. Package dropout occurs for three different reasons, including time-out retransmission, time-out transmission, and network disconnection [95]. The Phasor Measurement Unit (PMU) is a time trigger sensor, and it measures during every sampling interval  $T_s$ . The relationship between the frequency received by controller and measured by PMU can be obtained by:

$$\widetilde{\Delta f}(t_n) = y_k \Delta f(kT_s) \tag{24}$$

where  $t_n$  is the time instant when the nth packet data is received by the controller, and  $y_k = 1$  defines the success of the data transmission. Also, for this situation,  $\tau_{sc} = (t_n - kT_s)$ , where  $\tau_{sc}$  is the time needed for protocol discard of the packet and report of an error [95]. Thus, if the density of the data is increased, the time delay also increases and therefore, the number of packet dropouts will increase. Therefore, in [96], the packet dropout is modeled by a stochastic process. Using binary switching sequences increases the probability of packet dropout [97]. The value of  $y_k$  of the binary switching sequence takes the value of 0 or 1 with:

$$P\{y_k = 0\} = L_d \tag{25}$$

where  $L_d$  is the expected packet loss probability. The measurement of frequency will be updated after receiving the packet, and the evolution of it calculated by:

$$\widetilde{\Delta f}(t) = \widetilde{\Delta f}(t_n) \tag{26}$$

In addition, the controller updates the signal of control and sends it to RES after the controller receives the update of frequency measurement. Thus, if the nth packet is received, the control signal is stored in the buffer of the RES, and update as:

$$\Delta \widetilde{u}_i(t) = \Delta \widetilde{u}_i(t_m) 
\Delta \widetilde{u}_i(t_m) = y_n u(t_n)$$
(27)

where  $t_m$  is the time of receiving the mth data in RES, and if the signal of control,  $u(t_n)$ , calculated based on the frequency measurement is not dropped, then  $y_n = 1$ . If  $y_n = 1$ ,  $\tau^m_{ca} = (t_m - t_n)$ , and  $\tau^m_{ca}$  is the time of transmission of packet m from the controller to RES i. The controller sends signals of control to the energy storage, and finally to RES.

# 4.4. Communication Failure

The secondary controller is responsible for the dynamics of the microgrid and brings it to a set point, and also compensates the deviations of the frequency and voltage during the changes in the system. However, failure in communication is the most critical event that can eliminate the performance of the secondary control [98]. In [99,100], separating the faulty RES is proposed as a method for supporting the secondary control in a microgrid during communication failures. These methods provide an enhancement by reducing the disturbance associated with communication failure. Reference [101] proposes a robust secondary controller for restoring the frequency and voltage of a microgrid to the setpoints with an associated feature which ensures the perfect performance during the communication failure. This method uses the predictive values of voltage in the dq-rotating frame by two dynamic

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lookup tables during communication failure. In [102], a technique is proposed for ensuring the stability of a distributed model controller method in case of communication failure. The presented approach replaces the affected coupling constraints to decrease the cost and increase the stability during communication failure.

The summary of the reviewed solutions for solving communication degradation processes are presented in Table 2.

Type of Communication Degradation Process	Method	Advantages	Disadvantages  Depends on topology variations Only works on inverter-based RES Lack of considering the characteristics of source Only implemented on islanded microgrids Reduced accuracy due to the linearization of system Slow	
Delay	[83]	<ul> <li>Online method</li> <li>Increases the stability</li> <li>Wide delay boundary</li> <li>The relation between control parameters and delay is considered</li> <li>Lacks requirement of any additional equipment</li> <li>Low cost</li> </ul>		
Noise	<ul> <li>Considers uncertainty</li> <li>Applicable during unknown disturbances</li> <li>Improves the stability</li> <li>Considers the dynamic of each agen</li> <li>Considers the nonlinearity of the system</li> <li>Robust</li> </ul>		Lack of consideration of relation between power and frequency     Reduced accuracy due to the linearization of system     Slow     Inaccurate reactive power sharing     Does not consider the line impedance effect	
Packet Dropout	[96]	<ul> <li>Fast</li> <li>Considers the nonlinearity</li> <li>Fast</li> <li>Does not require additional requirements</li> </ul>	<ul> <li>Inaccurate during uncertainties</li> <li>Costly</li> <li>Does not consider different RES types</li> </ul>	
Communication Failure	[100]	<ul> <li>Detects the failure</li> <li>Stabilizes system after disconnection</li> <li>Fast</li> <li>Accurate power sharing</li> <li>Considers nonlinearity</li> </ul>	<ul> <li>Costly</li> <li>Only considers PV-based systems</li> <li>Only works on radial systems</li> <li>Considers only temporary failures</li> <li>High conservatism</li> </ul>	

**Table 2.** The summary of the communication degradation processes solutions.

# 4.5. Network Requirements

In addition to the theoretical methods and works in the literature, several technologies have been developed and designed in recent years to improve network and communication facilities. Microgrids involve different types of communication services, such as power consumption information collection, power quality monitoring, and microgrid automation. Corresponding network technologies include GPRS, wireless narrowband private network, wireless broadband private network (WIMAX), Ethernet passive optical network (EPON), and power line communication (PLC), and are presented in [103,104]. The summarized comparison between these technologies is summarized in Table 3.

In addition, for improving the perfect performance of a communication network in a microgrid, different requirements must be implemented to ensure the stable operation of the microgrid [105,106]. In Table 4, the general characteristics of these requirements are mentioned. Moreover, the communication protocols for application in microgrids are represented in Table 5 [107]. Also, in terms of energy consumption of communication mechanism, Reference [108] presents that among the technologies of Table 5, EPON has the most energy consumption per transmitted bit. Moreover, in [109], the energy consumption of WiFi-based systems and GPRS systems are compared, which shows the higher efficiency of wireless systems.

**Table 3.** Different network technologies. Abbreviations: GPRS; WIMAX, wireless broadband private network; PLC, power line communication; EPON, Ethernet passive optical network.

Technology	Application	Advantages	Disadvantages
GPRS	<ul><li>Large systems</li><li>Central controllers</li></ul>	<ul><li>Low implementation cost</li><li>Simple</li><li>Wireless</li><li>Global range</li></ul>	<ul> <li>High operation cost</li> <li>Low reliability</li> <li>High energy consumption</li> <li>Low bandwidth (57.6 kbit/s)</li> </ul>
Wireless Narrowband Private Network	<ul><li>Local controllers</li><li>High speed management systems</li></ul>	<ul><li>Low implementation cost</li><li>Simple</li><li>Wireless</li><li>Long range</li></ul>	<ul> <li>Low bandwidth</li> <li>Communication effect is limited by terrain</li> <li>Low bandwidth (144 kbit/s)</li> </ul>
WIMAX	Communication with requiring bidirectional data flow	<ul> <li>High bandwidth</li> <li>Good scalability</li> <li>Wireless</li> <li>High bandwidth (70 Mbit/s)</li> <li>Long range</li> </ul>	<ul> <li>Frequency problem</li> <li>Still needs some technological improvement</li> </ul>
PLC	<ul><li>Large systems</li><li>Control of distributed systems</li></ul>	<ul> <li>High security</li> <li>Special network operation</li> <li>Low construction cost</li> <li>High bandwidth (100 kbit/s)</li> </ul>	<ul> <li>Affected by the operation of line</li> <li>Low bandwidth</li> <li>Limited range due to the requiring wiring</li> </ul>
EPON	<ul><li>Small systems</li><li>High speed controllers</li></ul>	<ul> <li>Good scalability</li> <li>Large system coverage</li> <li>High bandwidth</li> <li>Low energy consumption</li> <li>High bandwidth (1000/1000 Mbit/s)</li> </ul>	<ul><li>High cost</li><li>Limited range due to the requiring wiring</li></ul>

**Table 4.** Communication requirements.

Data Type	Data Flow	Delay	Safety Requirements
Telecommunication	>128 bit/s	Delay <100 ms	Allow public network transmission
Telemetry	>1024 bit/s	Delay <100 ms	Allow public network transmission
Telecontrol	>128 bit/s	Delay <100 ms	Control commands require secure channel guarantee
Monitoring	>2.4 kbit/s	Delay <100 ms	Allow public network transmission

The integration of communication links in control systems increases the vulnerability of microgrids to the cyberattacks, which is one of the challenges nowadays in reliable microgrids [110]. Therefore, this problem should be considered in the designing of microgrid control systems. One of the cyberattacks is false data injection (FDI), which causes bad data detection in SCADA-based systems and makes an error in the state estimation of system [111]. Moreover, during FDI in a microgrid, it removes some components of the microgrid and causes harmful commands and even a blackout in the system. Recently, several studies have been conducted to increase the security of the system against FDI attacks. In [112], a method is proposed to secure some state variables and measurements to secure the system against FDI attacks. Moreover, the effect of FDI on electricity market operations is investigated in [113] by manipulating the real-time locational marginal price.

**Table 5.** Communication protocols [107].

Application	<b>Communication Protocol</b>
Distributed Energy Resources (DER) control	IEC 61850
Demand response	IEC 61968
(Advanced Metering Infrastructure) AMI	IEC 61968
Synchro phasors	IEEE C37.118.1
Transmission automation	IEC 60870-5 or DNP3
Substation automation	IEC 61850
Tele-protection relays	IEC 60834 or IEEE C37.236

On the other hand, in a microgrid, several components, such as smart measurements, switches, communication lines, and control systems, are installed. This high penetration of communication channels and processors into the microgrids enhances the vulnerability of the system to cyberattack. In this case, the hacker disrupts the communication network to disable the control system. In this case, considering cyberattacks during the designing of the system is essential. In this case, several approaches are presented to defend and detect cyberattacks based on communication facilities [114].

In addition, the related metrics of the bandwidth of communication channels are link capacity, bulk transfer capacity, and available bandwidth. Bulk transfer capacity is determined for an end-to-end path, while available bandwidth and link capacity are indicated for end-to-end paths of individual channels. The measured link capacity of wireless links depends on the cross traffic intensity and packet size [115]. In WIMAX networks, the available bandwidth undergoes fast variations, because of channel fading and error from the physical problems. Increasing the required bandwidth causes the use of EPON in the control systems. Due to the increasing demand of bandwidth in the future, a next generation EPON (NG-EPON) is developed to provide huge data rate capacity for future usages [116]. A comparison of the quantity value of GPRS [117], wireless narrowband private network [118], WIMAX [119], PLC [120], and EPON [121] is presented in Table 3.

Consequently, communication channels are the main part of control systems, and the presence of communication failures requires the designing of several structures and methods for solving these problems. For example, multilayer communication systems consist of an input interface connected to input communication lines, a switch circuit, and an output interface connected to output communication lines. A class identifier is indicative of one of the classes allocated by input interface, and an IP packet is received via one of the input communication lines, and allocates an internet protocol quality of service code to IP packets. The fault diagnosis and tolerance of this type of communication link are presented in [122].

#### 5. Conclusions

The high penetration of RES into the power system leads to the rise of the concept of microgrids as a controllable system. Microgrids are based on control and communication to provide perfect operation and requirements. In recent years, the main researches have been focused on the distributed control of microgrids to decrease the communication and computational burden made by high penetration of RES. In this paper, a comprehensive review of the distributed control of microgrids and different distributed control schemes was presented. The main aim of this paper was to determine the various existing methods and classify these methods based on their formulation. Moreover, in this paper, the impact and importance of the communication degradation processes, such as communication failure, noise, delay, and packet dropout on the distributed control of microgrids and the modeling techniques of them weere comprehensively reviewed. This paper also discussed the relation between distributed control strategy and communication degradation processes. It revealed the importance of considering communication failure, noise, and delay in the designing of a distributed control system. Therefore, strategies for improving the control system during communication degradation processes are compared and reviewed.

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#### **Abbreviations**

RES Renewable energy source

WT Wind turbine PV Photovoltaics LV Low voltage

MMC Modular multilevel converter
MPC Model predictive control
PMU Phasor measurement unit

WIMAX Wireless broadband private network EPON Ethernet passive optical network PLC Power line communication

FDI False data injection

NG-EPON Next generation ethernet passive optical network

AMI Advanced Metering Infrastructure
DER Distributed Energy Resources
GPRS General Packet Radio Services

RBF Radial Basis Function RC Resistant Capacitance

RLC Resistant-Inductance (L)-Capacitance IGBT Insulated Gate Bipolar Transistor

LCL Inductance (L) – Capacitance (C) – Inductance (L)

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