Determining the Maximum Allowable Current of an RBS using a Directed Graph Model and Greedy Algorithm

- $_3$ Binghui Xu $^{1\dagger},$ Guangbin Hua $^{1\dagger},$ Cheng Qian $^{1*},$ Quan Xia $^{1,2},$ Bo Sun 1, Yi Ren 1, and Zili Wang 1
- ⁵ School of Reliability and Systems Engineering, Beihang University, Beijing, 100191, China
- ⁷ School of Aeronautic Science and Engineering at Beihang University, Beijing, China *Address correspondence to: cqian@buaa.edu.cn
- †These authors contributed equally to this work.

10 Abstract

11

12

13

14

15

16

18

19

20

21

22

23

24

25

26

29

30

31

32

33

Reconfigurable battery systems (RBSs) present provide a promising alternative to traditional battery systems due to their flexible and dynamically changeable topological structure structures that can be adapted to different battery charging and discharging strategies. During RBS operation, a A critical system parameter known as the maximum allowable current (MAC) become pivotal is pivotal to RBS operation. This parameter is instrumental in maintaining the current of each individual battery within a safe range and serves as a guiding indicator for the system's reconfiguration, thereby ensuring its safety and reliability. This paperproposes a method to calculate In this paper, a method is proposed for calculating the MAC of arbitrary RBSs an arbitrary RBS using a greedy algorithm in conjunction with a directed graph model of the RBS. By introducing Using the shortest path (SP) of the battery, the greedy algorithm transforms the enumeration of exhaustion of the switch states in the brute-force algorithm into the or variable search without utilizing structures in the heuristic algorithms (simulated annealing and genetic algorithms) in the combination of the shortest paths, which greatly increases. This significantly enhances the efficiency with which the MAC is determined. The directed graph model, based on the an equivalent circuit, provides a specific method for calculating the MAC of a given structure. The proposed method is validated on two published four-battery-RBSs and using two previously published RBS structures and an additional one with a more complex structure. The results are the same as those of from the brute-force algorithm, but the proposed method significantly improves the computational efficiency (, being theoretically $N_s 2^{N_s - N_b} \log_{10} N_b$ times faster than the brute-force brute-force algorithm for an RBS with N_b batteries and N_s switches, theoretically). The main. Another advantage of the proposed method is its ability to calculate the MAC of RBSs with arbitrary structures and variable batteries, even in scenarios with random isolated batteries.

1 Introduction

47

51

52

54

62

63

64

65

67

68

71

Battery energy storage systems (BESSs) are extensively used widely utilized in various applications [1], such as wind power plants [2] and space power systems [3, 4], to store and release for the purpose of storing and releasing high-quality electrical energy [5]. Typically, a BESS consists of numerous batteries interconnected by series-parallel circuitry to provide the required capacity storage storage capacity. However, traditional conventional BESSs, in which the batteries are connected in a fixed 39 topology, suffer from a significant weakness in their worst battery batteries due to the so-called cask effect. MoreoverFurthermore, if the worst battery fails during operation, it is highly likely 41 to exacerbate accelerate the degradation of the other batteries, leading to resulting in reliability 42 and safety issues at the system level [6, 7, 8]. These problems have become significant technical barriers challenges have become major technical obstacles in many engineering projects requiring that demand high reliability, such as developing new-generation the development of next-generation 45 space vehicles [9].

Reconfigurable battery systems (RBSs), which can dynamically switch as required to between different circuit topologies as needed, are expected to solve this problem [10]. The switching circuit helps to isolate unhealthy batteries, thereby improving the safety and reliability of the battery system. To illustrate the working principle of an RBS, we consider be able to address these issues [10]. In a typical RBSstructure developed by Visairo [11] (Fig. ??), which is taken as an example to show the reconfiguration process. In this structure, the batteries can be connected not only in series when the switches S_1 , S_5 , S_6 , S_7 , S_8 , S_9 , and S_{13} are closed (see Fig. ??) but also in parallel when S_1 , S_2 , S_3 , S_4 , S_5 , S_9 , S_{10} , S_{11} , S_{12} , and S_{13} are closed (Fig. ??). Furthermore, when an unhealthy battery, for instance, the orange one B₃ in Fig. ??, appears in the RBS, it additional switches are introduced between the batteries to form a reconfigurable network, where the circuit's topology can be altered by opening or closing the switches. By opening the switches adjacent to the unhealthy batteries, they can be isolated by opening its two adjacent switches (i.e., S_4 and S₁₁), from the system, ensuring that the system remains in a reliable working mode. operational state [12]. Furthermore, an RBS can be reconfigured to adapt to different charging and discharging strategies, thereby enhancing the system's efficiency and prolonging the battery's lifespan [13]. These advantages make RBSs a promising alternative to traditional BESSs.

Recently, various types of RBSs with different. The early research on RBSs mainly focused on the topological design of their structures, incorporating different levels of flexibility and reconfigurability have been designed to meet application requirements. For example, Ci et al. [14] proposed an RBS structure that dynamically adjusts the battery discharge rate to fully exploit the available capacity of each battery. Jan 's [15, 16] structures reconfigure structures et al. [15, 16] designed structures that reconfigure circuits with variant batteries in series to reach the (constantly changing)—accommodate the constantly changing voltage requirements during electric vehicle charging. As shown in Fig. ??, the A structure proposed by Visairo et al. [11] changes and Kumar [11] alters the system's output voltage based on the load conditions, thereby reducing the power loss of power

loss in the voltage regulator during the power supply process and improving the efficiency of energy use. Also, to enhance the energy efficiencyof the system, Lawson et al. enhancing energy efficiency. Lawson [17] and He et al. [18] also focused on enhancing energy efficiency, and proposed simplified structures that have fewer switches than Visairo's design the design of Visairo and Kumar. Kim et al. [19] improved the system's ability ability of an RBS structure to recover from battery failures by introducing multiple portsinto the structure.

76

77

78

79

80

82

83

84

88

91

93

94

97

99

100

101

102

103

105

106

107

108

109

111

112

114

The complex structure. These complex structures between batteries and switches gives RBSs flexibility but also creates challenges in the designand control of the system. Thus, several approaches to analyze the RBS structure and performance have been proposed to tackle these challenges. For instance, provide flexibility to RBSs but also pose challenges in hardware design. During the reconfiguration process, current deviation and fluctuation may occur. Specifically, when the system switches from series to parallel connection, a circulating current between parallel cells can be triggered due to a voltage imbalance [20]. Failure to fully consider this issue during the design of RBSs can result in damage to the batteries, switches, and wires. For example, Engelhardt et al. [21] applied an RBS to a fast-charging scenario with adaptive cell switching to balance cell states while adhering to voltage requests. However, the switching of batteries leads to intolerable current variations. To address this problem, Han et al. [22] derived an analytical expression for the maximum switch current during battery system reconfiguration of a specific RBS structure. This helps guide. This analytical expression aids in the selection of switches and supports the designof RBS hardware, general hardware design.

Recently, increasing attention has been paid to the estimation and control of RBS system states, and several approaches have been proposed to optimize the performance of these systems. State estimation, which is an essential technology in traditional battery management systems, serves as the foundation for system control and holds great potential in the context of RBSs [23]. Couto et al. [24] introduced a partition-based unscented Kalman filter to estimate the state of a large-scale RBS, utilizing an enhanced reduced-order electrochemical model. Kersten et al. [25] utilized the balancing current of neighboring cells in parallel operation to determine the battery impedance, thereby obtaining information about the state of health and power capability of the RBS. Schmid et al. [26] further leveraged the reconfigurable nature of the system to actively diagnose faults, employing an algorithm that changes the system structure to enhance the fault isolability. Another active research area is the development of effective control strategies for RBSs to achieve optimal performance, including improved stability [27] and efficiency [28]. Han et al. [29] proposed a near-fastest battery balancing algorithm to minimize the time required for battery charge equalization. Liu et al. [30] also proposed a scheme for maximizing capacity utilization based on a path planning algorithm, aiming to enhance the battery consistency within the system. To break through the bottleneck of the potential short-circuit paths increasing exponentially with the RBS scale, Chen et al. [31] proposed a systematic approach based on sneak circuit theory to fundamentally avoid the short-circuit problem of RBSs: They thoroughly analyzed. They conducted a comprehensive analysis of all paths between the cathode and anode of each battery in the RBSand identified paths that only contain, identifying paths that consist only of switches as short-circuit paths for pre-checking before system reconfiguration. Artificial intelligence has also appeared in RBS management [32]. The effectiveness of the deep reinforcement learning method has been validated in real-world RBSs [28].

115

116

117

118

119

120

121

123

124

125

126

127

129

130

131

132

133

134

135

136

137

138

139

140

141

142

143

144

145

146

147

148

149

150

152

153

In spite of the maximum switch current mentioned above, the maximum The maximum allowable current (MAC), which is defined as the maximum allowed current under current allowed within the constraints of the a battery cell, is another critical a crucial indicator of RBSs that needs need to be evaluated during the design or and control of the system. The MAC helps the designers assess assists designers in assessing whether the RBS meets the output current requirements and contributes to the formulation-development of appropriate and safe management strategies for the battery management system. Unfortunately However, few studies have analyzed the RBS structure to determine the RBS MAC. An intuitive and straightforward method is to enumerate all possible switch states and calculate the output current of directly determined the MACs of RBSs, primarily due to the system under each reconfigurated structure complexity arising from reconfiguration. In the field of computer science, there is a similar problem with scheduling tasks on dynamically reconfigurable hardware with limited resources and task interdependencies. This problem is analogous to the determination of the MAC and a corresponding solution has been proposed [33, 34]. However, dealing with the structural characteristics and circuit equations of RBSs is challenging for this method. From the perspective of RBS structure analysis, the MAC problem can be transformed into a problem of finding the maximum output current among all possible reconfigurations of the RBS. However, this method is inefficient and time-consuming, especially for RBSs with a large number of switches. may be an NP-hard problem [35]. Common methods such as brute-force algorithms, simulated annealing (SA) algorithms, and genetic algorithms (GA) have the drawbacks of inefficiency, excessive time consumption, and an inability to guarantee the globally optimal solution.

To solve this issue, this paper proposes an efficient method to evaluate the MAC-MACs of RBSs. In this method, a greedy algorithm is designed to efficiently search the possible circuit topology of RBSswith MACtopologies of RBSs. This algorithm transforms the enumeration of switch states in the brute-force algorithm into the combination of the batteries' shortest paths. An improved direct graph model that considers the inefficient search for reconfigurations into a proactive combining of the shortest paths of the batteries. Furthermore, an improved directed graph model is introduced to analyze the current of the RBS, taking into account factors such as the voltage, the internal resistance, the MAC of the battery, and the external loadis also introduced to analyze the current of the RBS, external load. The main contributions of this paper study can be summarized as follows:

- An efficient method is proposed to determine the MAC MACs of RBSs with arbitrary structures, including scenarios with isolated batteries.
- A greedy algorithm is applied to solve the MAC problem, the computational complexity of which is greatly reduced compared with the brute-force algorithm.
- An improved directed graph model is introduced to provide a specific method for calculating the MAC of a given structure.

The remainder of this paper is organized as follows: Section II presents the framework and details of the proposed directed graph model and the greedy algorithm. Section III discusses a case study that uses applies the proposed method to determine the MACs of two published four-battery

RBSs and RBS structures and a new one with a more complex structure. The calculation results, the algorithm's computational complexity computational complexity of the algorithm, and scenarios such as battery random isolation are also discussed. Finally, the concluding remarks are presented in Section IV.

¹⁵⁸ 2 Methodology

159

160

162

163

164

165

166

168

169

The central principle of this the proposed method is to connect the batteries in an RBS in parallel to the extent possible maximum possible extent, thereby maximizing the output current the RBS. To achieve this universally and automatically, the overall process is divided into the four steps shown in Fig. 1. First, a directed graph model is established for the subsequent computations. The model not only contains the connected relationships between batteries and switches but also retains the performance parameters of the batteries. Subsequently, based on the equivalent circuit, the MAC problem is transformed into specific objective functions and constraints. The shortest paths (SPs, where additional batteries and switches on the path are penalized as distance) for for distance) of the batteries are then obtained by using the Dijkstra algorithm to connect the batteries in the RBS in parallel. Finally, a greedy algorithm is used to organize the switches, allowing the batteries to connect via their SPs while satisfying the constraints, resulting in the MAC of the RBS.

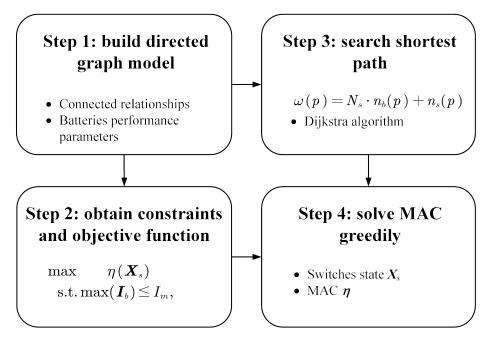


Figure 1: Diagram A diagram of this the proposed method, which contains four main steps.

2.1 Directed graph model

He et al. [36] proposed an abstracted directed graph model for an RBS, where the nodes represent the batteries, the edges represent the configuration flexibility, and the weight of each vertex

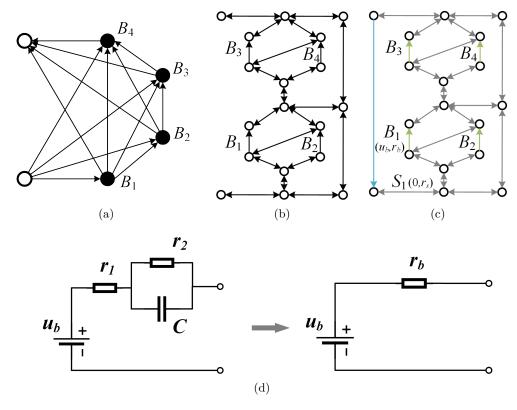


Figure 2: Directed The directed graph models used in (a) He's the work of He et al. [36], (b) our previous work, and (c) the improved model in this paper. (d) The equivalent circuit of a battery in this method.

corresponds to the battery voltage (Fig. 2a). The model captures all potential system configurations and offers a direct metric for configuration flexibility, but it does not specify the physical implementation of the connectivity between batteries, meaning that one graph might correspond to multiple RBS structures. We previously proposed a directed graph model that differs completely from He significantly from He et al.'s model by using nodes to represent the connections between batteries and switches and directed edges to represent batteries and switches (Fig. 2b), allowing for a one-to-one correspondence between the an RBS structure and the its directed graph model. This model accurately and comprehensively represents the RBS topological structure but cannot be used for quantitative MAC calculations because it does not consider the voltage, internal resistance, and or MAC of the battery. To address this issue, we improve our previous model by adding electromotive force and resistance attributes on the edges based on its the corresponding equivalent circuits. The model also considers the external load as an equivalent resistance and integrates it into the analysis, making it a complete circuit model for later circuit analyses analysis. Fig. 2c shows the improved directed graph model used in this paper. The following provides a detailed explanation of the method used for equating components in RBSs and constructing the directed graph model.

To use circuit analysis methods to solve the MAC of the RBS, the components in the RBS are equated to ideal circuit elements. For instance, as shown in Fig. 2d, the battery in the RBS is represented as a black-box circuit consisting of two resistors r_1 and r_2 and a capacitor C, in what is known as the Thevenin model [37, 38]. With an emphasis on the stable output of the RBS, the capacitor in the Thevenin model can be considered as an open circuit without affecting the steady-state current. Therefore, battery B_i in the RBS can be simplified as a series connection between a constant voltage source u_i and a resistor r_i . Furthermore, the state of the switch S_j in the RBS is represented by a binary variable x_j , where 0 is ON and 1 is OFF. When the switch is closed, the circuit can be regarded as a resistor with a very small resistance r_j . Finally, the external load is considered as a resistor with resistance R_o .

For a given RBS structure, its directed graph model G(V, E) is constructed as follows:

- 1. Nodes: The nodes in the directed graph correspond to the connection points of components in the actual RBS. Assuming there are a total of N nodes in the RBS, for the sake of convenience, the anode of the RBS is denoted as v_1 and the cathode as v_N .
- 2. Edges: The edges in the directed graph correspond to the batteries, switches, and external electrical loads in the actual RBS. Therefore, there are three types of directed edges. For battery B_i , its directed edge e_i is drawn from the cathode to the anode because the battery in operation only allows current to flow in one direction. For switch S_j , since it is allowed to work under bidirectional currents, it is represented by a pair of directed edges with two-way directions. Regarding the external electronic For the external electrical load, because it is connected to the anode and cathode of the RBS, a directed edge from v_N to v_1 represents is used to represent it. In conclusion, for a given RBS structure with N_b batteries and N_s switches, the number of directed edges is $N_b + 2N_s + 1$, where 1 refers to represents the external electrical load.
- 3. Attributes of edges: Each edge is assigned two attributes, a voltage difference and a resistance,

based on the equivalent method mentioned above. The values for battery B_i , switch S_j , and the external loads correspond to (u_i, r_i) , $(0, r_j)$, and $(0, R_o)$, respectively.

2.2 Constraints and objective function

For a given RBS, determining its the MAC involves maximizing the RBS output current while ensuring that all battery currents do not exceed the batteries' MAC. This subsection establishes the constraints and objective function used to determine the RBS's MAC through circuit analysis based on the directed graph model provided in the previous section.

First, the topology in the directed graph model is represented in matrix form the form of a matrix A, which is known as the incidence matrix and is defined as follows:

$$a_{kl} = \begin{cases} 1, & \text{edge } l \text{ leaves node } k, \\ -1, & \text{edge } l \text{ enters node } k, \\ 0, & \text{otherwise.} \end{cases}$$
 (1)

For a directed graph consisting of N nodes and $N_b + 2N_s + 1$ directed edges, its the incidence matrix \boldsymbol{A} is an $N \times (N_b + 2N_s + 1)$ matrix. In this matrix, the rows and columns represent the nodes and edges of the directed graph, respectively. By distinguishing the components in the RBS corresponding to each column, \boldsymbol{A} can be rewritten as

$$\boldsymbol{A} = \begin{bmatrix} \boldsymbol{A}_b & \boldsymbol{A}_s & \boldsymbol{A}_o \end{bmatrix}, \tag{2}$$

where A_b , A_s , and A_o are the submatrices corresponding to the batteries, switches, and external electrical load, respectively. To reduce the computational complexity, the dimensions of matrix A are reduced. Since each directed edge has one node to leave and one to enter, the values in every column of A sum to zero. Therefore, removing the last row will not result in a loss of information. Conversely, since each switch in the RBS is represented by a pair of directed edges with two-way directions, the two columns corresponding to the switch are mutually opposite. Thus, for the submatrix A_s , only one column is retained for each pair of columns representing the same switch. As a result, A can be reduced to an $(N-1) \times (N_b + N_s + 1)$ matrix, denoted \tilde{A} , for further calculation of the current and voltage. Similar to Eq. (2), \tilde{A} can be rewritten as

$$\tilde{\boldsymbol{A}} = \begin{bmatrix} \tilde{\boldsymbol{A}}_b & \tilde{\boldsymbol{A}}_s & \tilde{\boldsymbol{A}}_o \end{bmatrix}. \tag{3}$$

After obtaining the incidence matrix, the currents of all batteries and output in the RBS are determined by solving the circuit equations. According to Kirchhoff's laws, we have

$$\begin{cases} \tilde{A}I = 0, \\ U = \tilde{A}^{\mathrm{T}}U_n, \end{cases}$$
 (4)

where I and U indicate the current and voltage difference arrays of the N_b+N_s+1 edges, respectively,

and U_n is the voltage array of the N-1 nodes. These directed edges are treated as generalized branches and expressed in matrix form as follows:

$$I = YXU - YXU_s + I_s, (5)$$

where U_s and I_s denote the source voltage and source current of the generalized branches, respectively. Because all batteries have been made equivalent to voltage sources rather than current sources in the previous subsection, all elements of the array I_s are zero, whereas the elements of the array U_s are equal to the first attribute of the corresponding edges in the directed graph. The matrix Y in Eq. (5) is the admittance matrix of the circuit and is defined as the inverse of the impedance matrix. The elements on the diagonal of matrix Y are equal to the reciprocal of the resistance, which is the second attribute of the corresponding edges in the directed graph. The off-diagonal elements of Y are zero. X is the state matrix that determines whether the RBS batteries and switches can pass current. It is defined as

$$\boldsymbol{X} = \operatorname{diag}(\underbrace{1, 0, \dots, 1}_{N_b \text{ of } 0/1}, \underbrace{1, 0, \dots, 1}_{N_s \text{ of } 0/1}, 1) = \begin{bmatrix} \boldsymbol{X}_b \\ & \boldsymbol{X}_s \\ & 1 \end{bmatrix}, \tag{6}$$

where element x_i of matrix X_b indicates whether battery B_i has been removed from the circuit, with $x_i = 1$ indicating removal and $x_i = 0$ indicating that battery B_i is still available to supply power. When all batteries are healthy and capable of providing current to the external load, X_b is the identity matrix. The elements x_j of matrix X_s determine whether switch S_j is closed, with $x_j = 1$ indicating a closed switch and $x_j = 0$ indicating an open switch, which is consistent consistently with the previous subsection.

Theoretically, the output current I_o and the currents of each battery I_b in the RBS can be determined by solving Eqs. (4)–(6) under any given state X. To further simplify the problem, it is assumed that all batteries have the same electromotive force and internal resistance, which are denoted u_b and r_b , respectively. This allows us to derive explicit expressions for I_o and I_b . After derivation and simplification, the output current I_o and the currents of each battery I_b are ultimately represented as in Eqs. (7) and (8), respectively:

$$I_o = \frac{1}{R_o r_b} \tilde{\mathbf{A}}_o^{\mathrm{T}} \mathbf{Y}_n^{-1}(\mathbf{X}) \tilde{\mathbf{A}}_b \mathbf{U}_b, \tag{7}$$

$$I_b = \frac{1}{r_b^2} [\tilde{\boldsymbol{A}}_b^{\mathrm{T}} \boldsymbol{Y}_n^{-1} (\boldsymbol{X}) \tilde{\boldsymbol{A}}_b \boldsymbol{U}_b - r_b \boldsymbol{U}_b], \tag{8}$$

where U_b is an $N_b \times 1$ array with all elements equal to u_b , and Y_n is the equivalent admittance matrix of the circuit and is defined as

$$Y_n(X) = \frac{1}{R_o} \tilde{A}_o \tilde{A}_o^{\mathrm{T}} + \frac{1}{r_b} \tilde{A}_b X_b \tilde{A}_b^{\mathrm{T}} + \frac{1}{r_s} \tilde{A}_s X_s \tilde{A}_s^{\mathrm{T}}.$$
 (9)

To characterize the current output capacity of the RBS structure under different switching states,

an indicator η is defined by the ratio of I_o to $\max(I_b)$:

$$\eta = \frac{I_o}{\max(I_b)}. (10)$$

Finally, the problem of finding the MAC can be formulated as

$$\max \eta(\boldsymbol{X}_s) \tag{11}$$

s.t.
$$\max(\mathbf{I}_b) \le I_m$$
, (12)

where I_m is the MAC of the battery.

However, it remains computationally difficult to solve Eq. (11) because of Y_n^{-1} . On one handFirstly, the introduction of nonlinear terms by through Y_n^{-1} renders many methods in linear optimization unsuitable for this problem. On the other handSecondly, the rank of Y_n is proportional to the number of batteries and switches, which can be very large for a large RBS, leading to a significant computational burden. As a result, intelligent algorithms that rely on evolution by iteration may face efficiency problems when dealing with a large RBS. To address this issue, the problem should be considered from the perspective of guiding the RBS to reconstruct as many parallel structures as possible. Consequently, a greedy algorithm based on the shortest path is proposed. The detailed implementation of this algorithm is presented in the following two subsections.

2.3 Shortest path

The path p used in this method is defined as the complete route that passes through one battery (or a consecutive series of batteries) and closed switches, connecting the anode v_1 to the cathode v_N of the RBS. By applying a penalty to the series-connected batteries on the path, where additional batteries imply a greater distance, the algorithm encourages the RBS to form parallel structures to the maximum extent possible. In addition, to reduce the number of switches controlled during the reconstruction process, a penalty is also applied to the total number of switches on the path while ensuring the minimum number of batteries. Therefore, the distance ω of path p is

$$\omega(p) = N_s n_b(p) + n_s(p), \tag{13}$$

where N_s is the total number of switches in the system, and $n_b(p)$ and $n_s(p)$ are the number of batteries and switches in along path p, respectively. Moreover, the shortest path SP_i is defined as the path with the minimum ω for battery B_i :

$$SP_i = \underset{p \in P_i}{\operatorname{arg\,min}} \omega(p),$$
 (14)

where P_i is the set of all paths from v_1 to v_N that pass through directed edge i.

 SP_i can be solved by using the Dijkstra algorithm. The Dijkstra algorithm is a graph-search method that finds the shortest path between two given nodes in a weighted graph, efficiently solving the single-source shortest-path problem. Denoting the cathode and anode of battery B_i as v_i^- and

 v_i^+ respectively, then path p of battery B_i can be divided into three segments: $v_1 \to v_i^-$, $v_i^+ \to v_N$, and $v_i^- \to v_i^+$. $v_i^- \to v_i^+$ is the directed edge corresponding to battery B_i . With the Dijkstra algorithm, the shortest paths for $v_1 \to v_i^-$ and $v_i^+ \to v_N$ can be calculated under the weights given in Eq. (13) and denoted $SP(v_i^- \to v_i^+)$ and $SP(v_i^+ \to v_N)$, respectively. Finally, SP_i for battery B_i is formed by the complete path, which consists of $SP(v_1 \to v_i^-)$, $v_i^- \to v_i^+$, and $SP(v_i^+ \to v_N)$.

2.4 Greedy algorithm

From the perspective of series vs. parallel connections, integrating more batteries into the circuit through their shortest paths (SPs) results in more batteries connected in parallel, thereby increasing the total output current of the RBS. However, conflicts may arise between the SPs of different batteries. For instance, the SPs of two batteries might form a short-circuit RBS structure, which is not allowed. To address this issue, a greedy algorithm incorporates as many SPs as possible while satisfying the reconstruction requirements.

The algorithm (see pseudo-code the pseudocode in Algorithm 1) is illustrated in Fig. 3 and is summarized as follows: First, the SPs are obtained by using Eqs. (13) and (14) in conjunction with the Dijkstra search. Next, the matrix \boldsymbol{A} is calculated using Eq. (1), and the initial $N_{\rm set}$ is set to N_b . The algorithm uses a dichotomy method to iteratively check until convergence different combinations of c_b batteries from N_b and updates $N_{\rm set}$. For each combination, the algorithm constructs an effective solution if possible, and calculates the currents I_o and I_b by using Eqs. (7) and (8). If the maximum current I_b is less than or equal to I_m , η is calculated by using Eq. (10), and the maximum η is updated accordingly. Finally, the algorithm outputs the maximum η once $N_{\rm set}$ converges.

3 Case Study

3.1 Structures and details

Currently, there are two types of RBS structures have been proposed by Visairo et al. in the existing literature — those of Visairo and Kumar [11] and Lawson et al.—[17], both of which have seen real use. The primary goal of Visairo and Kumar's structure (Fig. 4b) is to dynamically adjust the RBS output power. However, the isolation of unhealthy batteries is not sufficiently addressed in their work. Lawson et al.—designed the RBS structure shown in Fig. 4a to isolate batteries. Although this structure easily isolates batteries, it cannot dynamically adjust the output current of the RBS. Based on the structures structure of Visairo and Kumar and that of Lawson, this paper proposes the structure shown in Fig. 4c. By integrating the Visairo and Kumar RBS structure into the Lawson RBS structure, the proposed structure not only has the flexibility to switch the batteries between series, parallel, and mixed series-parallel modes, but also allows the isolation of highly degraded batteries from the RBS. These four-battery RBS structures are investigated in

3.2 Result

In the case study, including the scenarios with random isolated batteries.

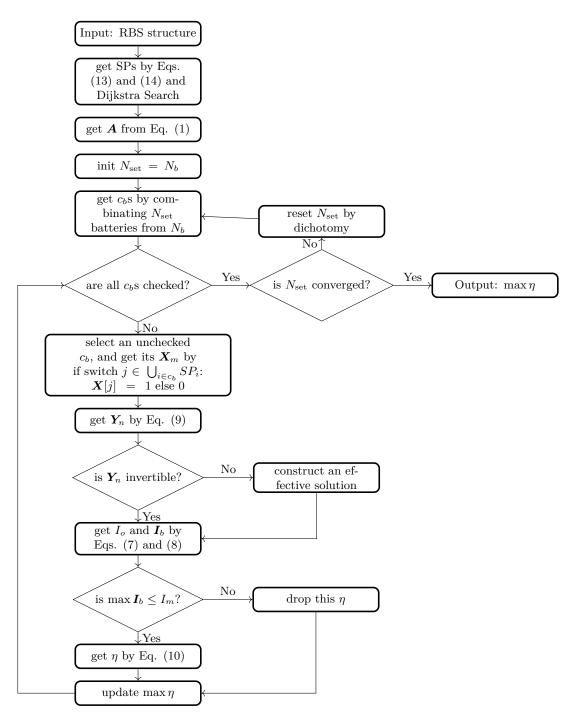


Figure 3: The computational flowchart of the MAC for a given RBS.

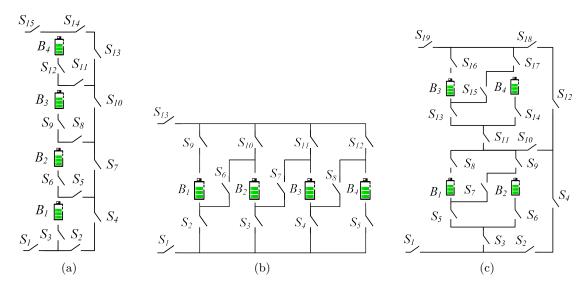


Figure 4: The four-battery RBS structures proposed by (a) Lawson [17], (b) Visairo and Kumar [11], and (c) this paper.

As shown the following RBS systems are investigated and compared: (a) three different structures (Figs. 4a–4c) with the same four batteries; (b) the same structure as in Fig. 4c, the new RBS structure consists of fourbatteries with two/four/six batteries; and (c) the four-battery structure in Fig. 4c with random isolated batteries. The greedy algorithm proposed in this work is also compared with the brute-force algorithm, SA, and 19 switches. Figure ?? shows the corresponding directed graph, which is composed of 18 nodes and 43 edges. Batteries B_1 , B_2 , B_3 , and B_4 are denoted by green directed edges in the graph, and the 19 switches are represented by gray directed edges with bidirectional arrows. The external electrical load is treated as a directed edge from the cathode of the RBS (i.e., node 18) to the anode (i.e., node GA to validate its effectiveness and efficiency. In order to adapt the two heuristic algorithms to the system's structure and scale, the number of state neighbors of SA and the population size of GA are both set to $N_b \cdot N_s$, which increases with the number of batteries and switches in the system. The parameters of the other algorithms are shown in Tab. 1.

Table 1: The SA and GA algorithm parameters.

Algorithm/parameter	Value
SA/initial temperature SA/final temperature SA/cooling rate	1), as indicated by the blue directed edge in the graph. 0.95
GA/crossover probability GA/mutation probability	$ \begin{array}{c} 100 \\ 0.8 \\ 0.02 \end{array} $

3.2 Result

3.2.1 The shortest path

Using Eq. (13) and the Dijkstra algorithm, the SPs of the four batteries in the RBS structure of Fig. 4c are highlighted in red in Figs. ?? and ??. Finally, the calculated structures of Figs. 4a, 4b, and 4c are calculated and highlighted with different colors in Figs. 5a, 5b, and 5c, respectively.

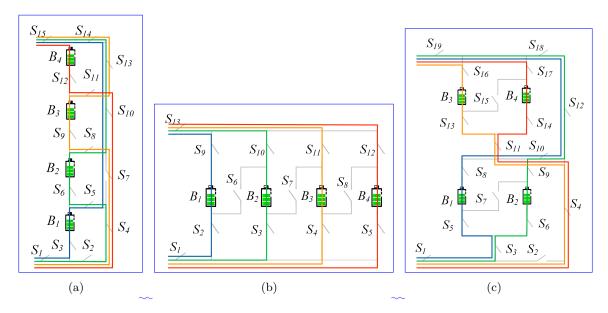


Figure 5: The SPs of the four batteries in the RBS structures of (a) Fig. 4a, (b) Fig. 4b, and (c) Fig. 4c.

3.2.2 Three structures with four batteries

After obtaining the SPs, the MACs of the structure in Fig. 4e are listed in Tab. 4 and shown in Fig. ??, as obtained by the greedy algorithm 1. Tab. 4three RBS structures with four batteries are calculated using the proposed greedy algorithm, and the results are shown in Tabs. 2, 3, and 4, each of which contains the states of the switches, the output current I_o , the battery current I_b , and the ratio η of the RBS structure with all batteries in good health when the RBS when the system output reaches the MAC. Fig. ?? presents the corresponding circuit, with the red highlight indicating that the current is flowing through the respective branches.

— — For the RBS structure in Fig. 4c, (a) its directed graph and the SPs (highlighted in red) of battery (b) B_1 , (c) B_2 , (d) B_3 , and (e) B_4 . (f) Circuit of RBS with its output reaching the MAC.

Calculated MAC for four-battery RBS structure in Fig. 4c. Structure Figure 4c with four batteries and 19 switches Switch on $S_1, S_3, S_5, S_6, S_8, S_9, S_{10}, S_{12}, S_{18}, S_{19}$ I_o $2u_b/(2R_o + r_b)$ I_b $[u_b/(2R_o + r_b), u_b/(2R_o + r_b), u_b/(2$

Similarly, the results of the MAC calculation for the structures. The corresponding switch-control schemes are shown as blue-highlighted electric currents in Figs. 4a and 4b are listed in Tabs. 2 and

36a, 6b, and 6c, respectively.

To verify and compare the results from the proposed greedy algorithm, we also used a the brute-force algorithmthat, which iterates through all possible switch states, and the heuristic algorithms (SA and GA) to calculate the MAC MACs of the same three RBSs. The final results of the brute-force algorithm are the same as the results shown in Tabs. 4-3. The method uses the greedy algorithm to calculate 11, 11, and 1 reconfigured structures for the RBS structure in Figs. 4c, 4a, and 4b, respectively. For the same RBS, the method those of the greedy algorithm and are shown in Tabs. 2, 3, and 4. However, the brute-force algorithm counts all possible switch states, which equates to 2^{19} , 2^{15} , and 2^{13} , and 2^{19} structures, respectively. The temporal evolutions of the objective values of the two heuristic algorithms during the iteration process are shown in Figs. 7a, 7b, and 7c, respectively, and compared with the proposed greedy algorithm. Compared with the SA and GA, the proposed greedy algorithm identifies the correct results within fewer iteration steps.

Table 2: The calculated MAC Calculating result of the four-battery RBS structure in Fig. 4a.

Structure	Figure 4a with 4-four batteries and 15 switches
Switch ON	$S_1, S_3, S_5, S_7, S_{10}, S_{13}, S_{14}, S_{15}$
I_o	$u_b/(R_o+r_b)$
\boldsymbol{I}_b	$[u_b/(R_o+r_b),0,0,0]$
$\max \eta$	1

Table 3: The calculated MAC Calculating result of the four-battery RBS structure in Fig. 4b.

Structure	Figure 4b with 4-four batteries and 13 switches
Switch ON	$S_1, S_2, S_3, S_4, S_5, S_9, S_{10}, S_{11}, S_{12}, S_{13}$
I_o	$4u_b/(4R_o+r_b)$
$oldsymbol{I}_b$	$[u_b/(4R_o+r_b), u_b/(4R_o+r_b), u_b/(4R_o+r_b), u_b/(4R_o+r_b)]$
$\max \eta$	4

Furthermore, the RBS with isolated batteries is taken into consideration and calculated. The MAC calculation

Table 4: The calculated MAC of the four-battery RBS structure in Fig. 4c.

Structure	Figure 4c with four batteries and 19 switches
Switch ON	$S_1, S_3, S_5, S_6, S_8, S_9, S_{10}, S_{12}, S_{18}, S_{19}$
$ec{I}_{\!\mathcal{Q}_{\!\sim}}$	$2u_b/(2R_o+r_b)$
$ ot\!\!{m I}_{b_{\!\sim}}$	$[u_b/(2R_o+r_b), u_b/(2R_o+r_b), 0, 0]$
$\max_{t} \underline{\eta}_{t}$	$\frac{2}{\sim}$

3.2.3 Structures with different numbers of batteries

We next examine the RBS configurations depicted in Fig. 4c, which consist of two, four, and six batteries. The results for the three structures under study, with varying numbers of isolated batteries,

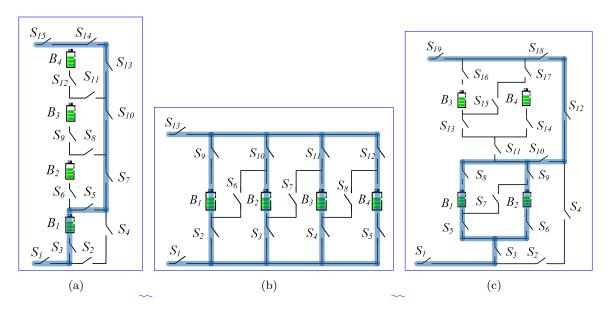


Figure 6: The RBS switch-control schemes with the output reaching the MAC.

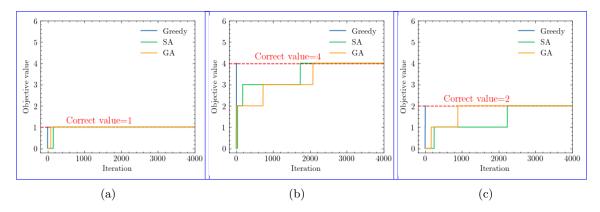


Figure 7: The temporal evolution of the objective values during the iteration process of calculating the RBS structures in (a) Fig. 4a, (b) Fig. 4b, and (c) Fig. 4c

four-battery configuration are presented in Tab. ??. Figs. 10a 10d illustrate the corresponding 4 and Figs. 6c and 7c. The structures and final switch-control schemes for the new structure proposed in this paper under different conditions of isolated batteries two-battery and six-battery systems are illustrated in Figs. 8a and 8b, respectively. Furthermore, the temporal evolutions of the objective values throughout the iteration process are shown in Figs. 9a and 9b, respectively. The proposed greedy algorithm still converges the fastest and achieves the correct MAC. The SA algorithm fails to obtain the correct MAC within the given number of iteration steps in the case of the six-battery RBS structure.

Variation of MAC with the number of isolated batteries for different RBS structures, including the structure proposed by Lawson et al., Visairo et al., and the structure proposed in this paper. This

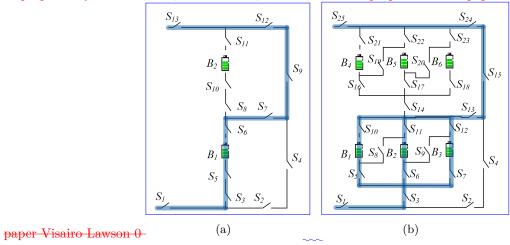


Figure 8: The (a) two-battery and (b) six-battery RBS switch-control schemes with the output reaching the MAC.

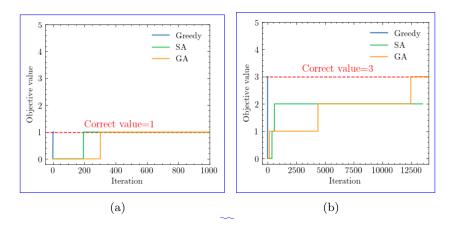


Figure 9: The temporal variation of the objective values during the iteration process of calculating the RBS structures in (a) Fig. 8a and (b) Fig. 8b.

3.2.4 Random isolated batteries

To assess the effectiveness of the proposed algorithm in the case of unhealthy batteries, the RBS with random isolated batteries is also taken into account and computed. In the case of the four-battery RBS structure depicted in Fig. 4c, there are four possible scenarios for isolated batteries: (a) a single unhealthy battery, (b) two unhealthy batteries located in different substructures, (c) two unhealthy batteries located in the same substructure, and (d) three unhealthy batteries. The resulting MAC (η) values for these four scenarios are 24 1 1, 23-, 12 2a or, and 1b 2 1 3 1 1 1, respectively. Furthermore, the corresponding switch-control schemes for the four scenarios are illustrated in Figs 10a-10d.

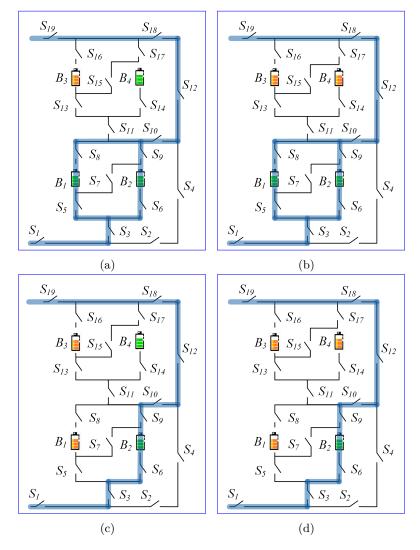


Figure 10: Circuit The circuit states of MACs when isolating (a) one, (b) two (best case in different substructures), (c) two (worst case in the same substructure), and (d) three batteries for the structure in Fig. 4c.

3.3 Discussion

Consider first the results

3.3.1 Result validation

The correctness of the outcomes provided by the proposed greedy algorithm will now be discussed from two perspectives: circuit analysis and validation against the brute-force algorithm. The result of the four-battery RBS structure shown in Fig. ?? and listed in Tab. 4. 4c is determined as an example. When B_1 and B_2 or B_3 and B_4 are connected in parallel, the RBS outputs produces the maximum current, which is $\eta = 2$ (i.e., twice the current output of a single battery in the RBS). Adding more batteries to the main circuit only forms-creates a series structure and does not improve the MAC. Therefore, the state of the switches given switch-control scheme provided in Tab. 4 maximizes the RBS output current. The brute-force method, which go through examines all possible switch states, also gives the same result. yields the same η . This indicates that the proposed greedy algorithm successfully identifies the MAC among all the potential reconfigured structures. The literature contains no report on an algorithm for calculating the MAC of an RBS. The

3.3.2 Pros and cons analysis

The proposed greedy algorithm possesses a significant advantage in terms of its effectiveness and efficiency. In this paper, it is compared with the brute-force algorithm, which goes through SA, and GA. While the brute-force algorithm ensures the correctness of the results by exploring all possible switch states, is the most straightforward way to determine the MAC and is used as a benchmark for the it comes at a high computational cost. The SA and GA are commonly used heuristic algorithms for addressing NP-hard problems. They selectively generate solutions for the switching states to maximize the objective value η . However, neither of these two algorithms can determine whether the current η represents the final MAC or if there are better solutions. Moreover, as depicted in Figs. 7a–7c and Figs. 9a–9b, the SA and GA algorithms require more iterations to converge to the final solution than the proposed greedy algorithm.

To further elaborate on the efficiency of our algorithm, we analyze the time complexity of both the brute-force algorithm and the greedy algorithm. If an RBS has N_b batteries and N_s switches and the corresponding directed graph has N nodes, 2^{N_s} iterations are required to traverse all reconfigured possible structures. Calculating each reconfigured structure using Eqs. (7)–(10) requires matrix inversion and matrix multiplication, which has results in a time complexity of $O(N^3 + 2N^2N_b + N^2N_s + NN_b^2)$. Therefore, the time complexity of the brute-force algorithm is $O((N^3 + 2N^2N_b + N^2N_s + NN_b^2)2^{N_s})$. The greedy algorithm proposed in this paper requires that the SP be found for each battery, which requires N_b iterations. Each SP can be obtained by through several applications of Dijkstra's algorithms. Therefore, the total time complexity for calculating all SPs is $O(N_b(N_b + 2N_s)\log_{10}N)$. According to Appendix 1, the RBS can reconfigure $C_{N_b}^{N_{\text{set}}}$ structures by selecting N_{set} batteries from N_b batteries, which gives $\sum_{N_{\text{set}}=1}^{N_b} C_{N_b}^{N_{\text{set}}}/N_b \approx 2^{N_b}N_b^{-1}$ on average. Thus, with the bisection method, the time complexity of the greedy algorithm is $O((N^3 + 2N^2N_b + N^2N_s + NN_b^2)2^{N_b}N_b^{-1}\log_{10}N_b + N_b(N_b + 2N_s)\log_{10}N)$. Based on currently proposed RBS structures For

the existing RBS structures in the literature [39, 40, 41, 42, 43, 44], the number N_b of batteries N_s of switches N_s , and N_s of nodes of batteries N_b , the number of switches N_s , and the number of nodes N_s are quantitatively related as follows: $N_s \approx (3-5)N_b$, $N_s \approx N_s$. After simplifying, the time complexity of the method with the greedy algorithm is $O(2^{N_b}N_s^2\log_{10}N_b)$, while it is $O(2^{N_s}N_s^3)$ for that of the method with brute force algorithm the brute-force algorithm is $O(2^{N_s}N_s^3)$. Therefore, as the RBS grows, especially in terms of the number of switches, the greedy algorithm gains an advantage over the brute-force algorithm. This is confirmed by the number of structures required to determine the MAC in the previous section. Compared with the brute-force algorithm, the method based on the greedy algorithm is 3000 to 48000 times more efficient, which is theoretically $N_s 2^{N_s - N_b} \log_{10} N_b$ times according to the above time-complexity analysis. This benefits from two key points the result of two key factors:

- (1) The SPs guide the RBS to reconfigure reasonable structures rather than blindly going through all possible structures. This reduces the complexity from 2^{N_s} to 2^{N_b} , which is the main reason for the improvement in efficiency.
- (2) The bisection method further accelerates this process, reducing the complexity from 2^{N_b} to $2^{N_b}N_B^{-1}\log_{10}N_b$.

However, the greedy algorithm proposed in this paper still contains

Furthermore, this approach can handle RBSs with arbitrary structures, which is another significant advantage. It can even do this when they have different battery variations or even random isolated batteries. Theoretically, each RBS structure can be transformed into a unique directed graph model using the methodology described in Section II, and the MAC can subsequently be calculated using the proposed greedy algorithm. This finding is supported by the findings in the previous subsection.

However, the suggested greedy algorithm still includes exponential terms in the its time complexity, which means it may not be able to handle extremely large RBS structures having large N_b . indicating that it struggles to perform at scale. Additionally, all batteries are assumed to be identical for the sake of simplification in the derivation. However, in reality, there may exist a small balancing current that could introduce a minor bias in the MAC due to variations in the open-circuit voltage u_b and the internal resistance r_b . Nevertheless, the proposed greedy algorithm remains a viable choice for RBS design and optimization in the early stage, and the issue of balancing current bias can be addressed by considering the inconsistency between batteries and replacing the internal resistance with impedance when constructing the directed graph model.

3.3.3 Application scenarios

Note that η is used as the objective function instead of I_o in solving for the MAC. This choice makes the resulting MAC more reasonable and applicable to practical scenarios. As shown in Tab. 4, I_o and I_b are functions of R_o , u_b , and r_b . However, when I_o is used as the objective function, even for the same RBS structure, the MAC solution and corresponding switch states could can change due to different external electrical appliances. This would increase increases the difficulty and uncertainty of involved in designing the RBS structure. To eliminate this problem, the ratio $\eta = I_o/\max I_b$ $\eta = I_o/\max (I_b)$ is adopted as the objective function in our research. Recall that η reflects only the structure's ability to output current, rather than the actual current outputing output by the battery system. Assuming that the MAC of the batteries in the RBS is I_m , the maximum output current of the RBS structure can be calculated as ηI_m by determining the value of η for the structure.

The method proposed in this paper facilitates the design of RBSs in the following ways: Most currently proposed. Most of the existing RBS structures [39, 40, 41, 42, 43, 44] have simple topological characteristics, so calculating the their MACs is relatively straightforward, or even intuitive. However, these simple structures do not always fully satisfy the requirements of complex applications, such as dynamically adapting the circuit to variable and random operating conditions or actively equalizing differences between batteries in the RBS. Moreover, isolating the batteries disrupts the original regularity and symmetry of the topology, which complicates the otherwise simple structure, and the maximum output current of the system becomes more challenging to obtain. In contrast, the proposed method calculates the MAC of arbitrary RBS structures, notably the most notably complex and flexible RBS structures.

To illustrate this point, the MACs of three RBS structures mentioned above the RBS structure in Fig. 4c are calculated after isolating one or more of the batteries, as shown in Tab. ??. Specifically, for the structure presented in Fig. 4c, the corresponding circuit states for the MACs when isolating one to three batteries are depicted in Figs. 10a-10d. This structure has two cases in which two When a single battery is isolated, the RBS is still capable of outputting the maximum current, denoted as $\eta = 2$. When two batteries are isolated, there are two scenarios: one is to isolate isolating two batteries within the same substructure (Fig. 10b), in which case resulting in $\eta = 2$; the other is to isolate isolating one battery in each of the two substructures (Fig. 10c), in which case resulting in $\eta = 1$. The results in Figs. 10a–10d show that the proposed method provides reasonable outcomes for isolating any number of batteries in any position. Furthermore, the output current for the three RBSs with isolated batteries is also shown in Tab. ??. For the structure proposed by Lawson et al., the MAC is independent of the number of isolated batteries. However, for Visairo's structure, the MAC decreases upon increasing the number of isolated batteries. Nevertheless, the MAC of the structure proposed in this work falls between the MACs of these two structures. This result indicates that the structure proposed in this paper has a larger MAC than Lawson's for the same number of batteries and has a wider range of regulation of the output current If three batteries are isolated, the RBS can only output the current of a single battery, which is $\eta = 1$. Therefore, the battery management system can adjust the output current and control the RBS to reconfigure the corresponding structure based on the isolated batteries.

4 Conclusion

472

473

474

475

476

477

479

480

481

482

483

485

486

487

488

489

490

491

492

493

494

495

496

497

498

499

500

502

503

506

507

This paper proposes a pervasive has proposed a reliable and automated method to efficiently compute the MAC of an RBS. The method is implemented by using a greedy algorithm combined with an improved directed graph model. Not only does the method provides provide the same global MAC calculation results as the brute force brute-force method, but it also improves the calculation

efficiency by 3 000 to 48 000 times for three RBS structures in the case studydemonstrates superior computational efficiency to both the brute-force algorithm and the heuristic algorithms (SA and GA). Theoretically, for an RBS with N_s switches and N_b batteries, the efficiency of the proposed method is $N_s 2^{N_s - N_b} \log_{10} N_b$ times that of the brute-force method, which is mainly because of using . This is primarily due to the utilization of the batteries' SPs to guide guiding the RBS to reconfigure reasonable structures rather than blindly going through all possible structures. The main Another advantage of this method is its ability capability to calculate the MAC MACs of RBSs with arbitrary structures and varying batteries. Even in scenarios with random isolated batteries, the proposed method remains effective. This method helps to fully tap the current output potential can facilitate the full utilization of the RBS's current output potential, guide the RBS structure design and optimization in the design stage of the RBS structure, and assist in evaluating the current-overload risk of the system risk of current overload in practical applications.

Appendix

510

511

512

513

515

516

518

519

521

525

526

527

528

533

Acknowledgments

Author Contributions

B. Xu conceived the main idea, formulated the overarching research goals and aims, designed the algorithm, and reviewed and revised the manuscript. G. Hua developed and analyzed the model, implemented the code and supporting algorithms, and wrote the initial draft. C. Qian provided critical review, commentary, and revisions. Q. Xia contributed to shaping the research, analysis, and manuscript. B. Sun conducted the research and investigation process. Y. Ren secured the funding and supervised the project. Z. Wang verified the results and provided necessary resources. 530

Funding 531

This work was supported by the National Natural Science Foundation of China (NSFC, No.52075028). 532

Conflicts of Interest

The authors declare that there is no conflict of interest regarding the publication of this article. 535

Data Availability

This work does not require any data to be declared or publicly disclosed.

Algorithm 1: Get Obtain the $\frac{max}{max}$ available $\frac{max}{max}$ available

```
Data: Directed graph model G(V, E) of the RBS
    Result: \max \eta
 1 for i \in E_b do
      P_i \leftarrow \{path | \text{starts at } v_1 \text{ and ends at } v_n\};
         SP_i \leftarrow p_i which has the minimum \omega(p_i) among all p_i \in P_i.
 4 end
 \mathbf{5} get Get \mathbf{A} by Eq. 1;
 6 while not yet determined \max \eta do
 7 end
 8 N_{\text{set}} \leftarrow \text{number of } \frac{\text{setected selected SPs calculated by dichotomy;}}{N_{\text{set}}}
 9 C_b \leftarrow set of all combinations of N_{\text{set}} batteries from N_b;
10 for c_b \in C_b do
         \boldsymbol{x}_s \leftarrow \text{list of all switches' states: } x_s[j] = 1 \text{ if } j \in \bigcup_{i \in c_b} SP_i \text{ else 0};
          \boldsymbol{X} \leftarrow diag[1, 1, \cdots, 1, \boldsymbol{x}_s];
12
          get Y_n by Eq. 9;
13
         if Y_n is invertible then
14
              pass
15
         {f else}
16
              construct an effective solution
17
18
          get I_o by Eq. 7;
19
          get I_b by Eq. 8;
20
         if \max(I_b) \leq I_m then
21
              \eta \leftarrow I_o/\max(\boldsymbol{I}_b);
22
          else
23
              break
24
          \mathbf{end}
25
26 end
```

References

- Yuqing Yang, Stephen Bremner, Chris Menictas, and Merlinde Kay. Battery energy storage
 system size determination in renewable energy systems: A review. Renewable and Sustainable
 Energy Reviews, 91:109–125, August 2018.
- [2] Luanna Maria Silva de Siqueira and Wei Peng. Control strategy to smooth wind power output
 using battery energy storage system: A review. Journal of Energy Storage, 35:102252, March
 2021.
- [3] Eugene Schwanbeck and Penni Dalton. International Space Station Lithium-ion Batteries for
 Primary Electric Power System. In 2019 European Space Power Conference (ESPC), pages 1–1.
 IEEE, September 2019.
- [4] Lihua Zhang. Development and Prospect of Chinese Lunar Relay Communication Satellite.
 Space: Science & Technology, 2021, January 2021.
- Jaephil Cho, Sookyung Jeong, and Youngsik Kim. Commercial and research battery technologies for electrical energy storage applications. Progress in Energy and Combustion Science,
 48:84–101, June 2015.
- ⁵⁵³ [6] Naixing Yang, Xiongwen Zhang, BinBin Shang, and Guojun Li. Unbalanced discharging and aging due to temperature differences among the cells in a lithium-ion battery pack with parallel combination. *Journal of Power Sources*, 306:733–741, February 2016.
- Fei Feng, Xiaosong Hu, Lin Hu, Fengling Hu, Yang Li, and Lei Zhang. Propagation mechanisms and diagnosis of parameter inconsistency within Li-Ion battery packs. Renewable and Sustainable Energy Reviews, 112:102–113, September 2019.
- [8] J. A. Jeevarajan and C. Winchester. Battery Safety Qualifications for Human Ratings. *Interface*magazine, 21(2):51–55, January 2012.
- [9] Daniel Vázquez Pombo. A Hybrid Power System for a Permanent Colony on Mars. Space:
 Science & Technology, 2021, January 2021.
- [10] Weiji Han, Torsten Wik, Anton Kersten, Guangzhong Dong, and Changfu Zou. Next Generation Battery Management Systems: Dynamic Reconfiguration. *IEEE Industrial Electronics Magazine*, 14(4):20–31, December 2020.
- [11] H. Visairo and P. Kumar. A reconfigurable battery pack for improving power conversion efficiency in portable devices. In 2008 7th International Caribbean Conference on Devices, Circuits and Systems, pages 1–6. IEEE, April 2008.
- [12] Song Ci, Ni Lin, and Dalei Wu. Reconfigurable battery techniques and systems: A survey.
 IEEE Access, 4:1175–1189, 2016.

- [13] Nejmeddine Bouchhima, Matthias Gossen, Sascha Schulte, and Kai Peter Birke. Lifetime of
 self-reconfigurable batteries compared with conventional batteries. *Journal of Energy Storage*,
 15:400–407, 2018.
- [14] Song Ci, Jiucai Zhang, Hamid Sharif, and Mahmoud Alahmad. A novel design of adaptive reconfigurable multicell battery for power-aware embedded networked sensing systems.
 In IEEE GLOBECOM 2007-IEEE Global Telecommunications Conference, pages 1043–1047.
 IEEE, 2007.
- Jan Engelhardt, Tatiana Gabderakhmanova, Gunnar Rohde, and Mattia Marinelli. Reconfigurable stationary battery with adaptive cell switching for electric vehicle fast-charging. In 2020 550 55th International Universities Power Engineering Conference (UPEC), pages 1–6, 2020.
- [16] Jan Engelhardt, Jan Martin Zepter, Tatiana Gabderakhmanova, Gunnar Rohde, and Mattia
 Marinelli. Double-string battery system with reconfigurable cell topology operated as a fast
 charging station for electric vehicles. *Energies*, 14(9):2414, 2021.
- [17] Barrie Lawson. A Software Configurable Battery. EVS26 International Battery, Hybrid and
 Fuel Cell Electric Vehicle Symposium, pages 252–263, 2012.
- [18] Liang He, Linghe Kong, Siyu Lin, Shaodong Ying, Yu Gu, Tian He, and Cong Liu.
 Reconfiguration-assisted charging in large-scale lithium-ion battery systems. In 2014
 ACM/IEEE International Conference on Cyber-Physical Systems (ICCPS), pages 60-71. IEEE,
 2014.
- [19] Hahnsang Kim and Kang G Shin. On dynamic reconfiguration of a large-scale battery system.
 In 2009 15th IEEE Real-Time and Embedded Technology and Applications Symposium, pages
 87–96. IEEE, 2009.
- [20] Weiji Han and Anton Kersten. Analysis and Estimation of the Maximum Circulating Current
 during the Parallel Operation of Reconfigurable Battery Systems. In 2020 IEEE Transportation
 Electrification Conference & Expo (ITEC), pages 229–234. IEEE, June 2020.
- [21] Jan Engelhardt, Jan Martin Zepter, Tatiana Gabderakhmanova, Gunnar Rohde, and Mattia
 Marinelli. Double-String Battery System with Reconfigurable Cell Topology Operated as a Fast
 Charging Station for Electric Vehicles. *Energies*, 14(9):2414, 2021.
- Weiji Han, Anton Kersten, Changfu Zou, Torsten Wik, Xiaoliang Huang, and Guangzhong
 Dong. Analysis and estimation of the maximum switch current during battery system reconfiguration. IEEE Transactions on Industrial Electronics, 69(6):5931–5941, 2021.
- Lidiya Komsiyska, Tobias Buchberger, Simon Diehl, Moritz Ehrensberger, Christian Hanzl,
 Christoph Hartmann, Markus Hölzle, Jan Kleiner, Meinert Lewerenz, Bernhard Liebhart,
 Michael Schmid, Dominik Schneider, Sascha Speer, Julia Stöttner, Christoph Terbrack, Michael
 Hinterberger, and Christian Endisch. Critical Review of Intelligent Battery Systems: Challenges, Implementation, and Potential for Electric Vehicles. Energies, 14(18):5989, 2021.

- [24] Luis D. Couto and Michel Kinnaert. Partition-based Unscented Kalman Filter for Reconfigurable Battery Pack State Estimation using an Electrochemical Model. In 2018 Annual
 American Control Conference (ACC), pages 3122–3128. IEEE, June 2018.
- [25] Anton Kersten, Manuel Kuder, Weiji Han, Torbjorn Thiringer, Anton Lesnicar, Thomas Weyh,
 and Richard Eckerle. Online and On-Board Battery Impedance Estimation of Battery Cells,
 Modules or Packs in a Reconfigurable Battery System or Multilevel Inverter. In *IECON 2020* The 46th Annual Conference of the IEEE Industrial Electronics Society, pages 1884–1891. IEEE,
 October 2020.
- [26] Michael Schmid, Emanuel Gebauer, Christian Hanzl, and Christian Endisch. Active Model Based Fault Diagnosis in Reconfigurable Battery Systems. IEEE Transactions on Power Electronics, 36(3):2584–2597, March 2021.
- [27] Jan Kacetl, Jingyang Fang, Tomas Kacetl, Nima Tashakor, and Stefan Goetz. Design and
 Analysis of Modular Multilevel Reconfigurable Battery Converters for Variable Bus Voltage
 Powertrains. IEEE Transactions on Power Electronics, 38(1):130–142, January 2023.
- [28] Feng Yang, Fei Gao, Baochang Liu, and Song Ci. An Adaptive Control Framework for Dynamically Reconfigurable Battery Systems Based on Deep Reinforcement Learning. *IEEE Transactions on Industrial Electronics*, 69(12):12980–12987, December 2022.
- [29] Weiji Han, Changfu Zou, Liang Zhang, Quan Ouyang, and Torsten Wik. Near-Fastest Battery
 Balancing by Cell/Module Reconfiguration. *IEEE Transactions on Smart Grid*, 10(6):6954–6964, November 2019.
- [30] Xinghua Liu, Guoyi Chang, Jiaqiang Tian, Zhongbao Wei, Xu Zhang, and Peng Wang. Flexible
 path planning-based reconfiguration strategy for maximum capacity utilization of battery pack.
 Journal of Energy Chemistry, 86:362–372, November 2023.
- [31] Si-Zhe Chen, Yule Wang, Guidong Zhang, Le Chang, and Yun Zhang. Sneak Circuit Theory
 Based Approach to Avoiding Short-Circuit Paths in Reconfigurable Battery Systems. IEEE
 Transactions on Industrial Electronics, 68(12):12353-12363, 2021.
- [32] Kailong Liu, Zhongbao Wei, Chenghui Zhang, Yunlong Shang, Remus Teodorescu, and Qing Long Han. Towards Long Lifetime Battery: AI-Based Manufacturing and Management. *IEEE/-* CAA Journal of Automatica Sinica, 9(7):1139–1165, July 2022.
- [33] Morteza Mollajafari. An efficient lightweight algorithm for scheduling tasks onto dynamically
 reconfigurable hardware using graph-oriented simulated annealing. Neural Computing and Applications, 35(24):18035–18057, August 2023.
- [34] Liang He, Linghe Kong, Siyu Lin, Shaodong Ying, Yu Gu, Tian He, and Cong Liu.
 Reconfiguration-assisted charging in large-scale Lithium-ion battery systems. In 2014
 ACM/IEEE International Conference on Cyber-Physical Systems (ICCPS), pages 60-71. IEEE,
 April 2014.

- [35] Zoltan Mark Pinter, Dimitrios Papageorgiou, Gunnar Rohde, Mattia Marinelli, and Chresten
 Traholt. Review of Control Algorithms for Reconfigurable Battery Systems with an Industrial
 Example. In 2021 56th International Universities Power Engineering Conference (UPEC),
 pages 1–6, August 2021.
- [36] Liang He, Lipeng Gu, Linghe Kong, Yu Gu, Cong Liu, and Tian He. Exploring Adaptive
 Reconfiguration to Optimize Energy Efficiency in Large-Scale Battery Systems. In 2013 IEEE
 34th Real-Time Systems Symposium, pages 118–127, December 2013.
- [37] Hongwen He, Rui Xiong, Xiaowei Zhang, Fengchun Sun, and JinXin Fan. State-of-Charge
 Estimation of the Lithium-Ion Battery Using an Adaptive Extended Kalman Filter Based on
 an Improved Thevenin Model. *IEEE Transactions on Vehicular Technology*, 60(4):1461–1469,
 May 2011.
- [38] S.M. Mousavi G. and M. Nikdel. Various battery models for various simulation studies and
 applications. Renewable and Sustainable Energy Reviews, 32:477–485, April 2014.
- [39] Song Ci, Jiucai Zhang, Hamid Sharif, and Mahmoud Alahmad. A Novel Design of Adaptive
 Reconfigurable Multicell Battery for Power-Aware Embedded Networked Sensing Systems. In
 IEEE GLOBECOM 2007-2007 IEEE Global Telecommunications Conference, pages 1043–1047,
 November 2007.
- [40] Mahmoud Alahmad, Herb Hess, Mohammad Mojarradi, William West, and Jay Whitacre. Battery switch array system with application for JPL's rechargeable micro-scale batteries. *Journal of Power Sources*, 177(2):566–578, March 2008.
- [41] Hahnsang Kim and Kang G. Shin. Dependable, efficient, scalable architecture for management
 of large-scale batteries. In *Proceedings of the 1st ACM/IEEE International Conference on Cyber-* Physical Systems, ICCPS '10, pages 178–187, New York, NY, USA, April 2010. Association for
 Computing Machinery.
- Younghyun Kim, Sangyoung Park, Yanzhi Wang, Qing Xie, Naehyuck Chang, Massimo Poncino, and Massoud Pedram. Balanced reconfiguration of storage banks in a hybrid electrical energy storage system. In 2011 IEEE/ACM International Conference on Computer-Aided Design (ICCAD), pages 624–631, November 2011.
- Taesic Kim, Wei Qiao, and Liyan Qu. A series-connected self-reconfigurable multicell battery capable of safe and effective charging/discharging and balancing operations. In 2012 TwentySeventh Annual IEEE Applied Power Electronics Conference and Exposition (APEC), pages 2259–2264, February 2012.
- [44] Liang He, Linghe Kong, Siyu Lin, Shaodong Ying, Yu Gu, Tian He, and Cong Liu.
 Reconfiguration-assisted charging in large-scale Lithium-ion battery systems. In 2014
 ACM/IEEE International Conference on Cyber-Physical Systems (ICCPS), pages 60–71, April 2014.