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Energy Management Using a Rule-Based Control Strategy of Marine Current Power System with Energy Storage System

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Abstract: With the rapid development of renewable energy technology, marine current energy is treated as the most desirable form of ocean energies. Due to the nature of marine current energy, simple structure, high reliability, and good control performance are the primary consideration for the energy management strategy. This paper proposes an energy management control strategy based on rules to compensate for the fluctuating power caused by tidal motion. The hybrid energy storage system composed of vanadium redox flow battery (VRB) is applied to reallocate power. Supercapacitor banks (SCBs) are applied as the auxiliary power source to absorb or release the required power according to energy management strategy based on control rules in the marine current power system. SCB makes the grid-connected power track the grid command power and also improves the operational efficiency of the vanadium redox flow battery (VRB). VRB compensates for the low-frequency fluctuating power caused by tidal motion and plays an important role in compensating for the difference in power between the grid-connected power and the grid command power to ensure the reliability of the marine current power system. A simulation model of a 3 MW marine current power system is built to verify the effectiveness of the energy management strategy based on the real marine current velocity data.

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1. Introduction

With the rapid development of renewable energy technology, marine current energy is treated as the most desirable form of ocean energy due to its strong regularity, predictability, and sustainability. Under the same power level, the volume of system equipment is smaller than wind energy, and the corresponding manufacturing cost is lower. Additionally, the utilization of marine current energy has become a hot research field all over the world [1]. The change of marine current velocity will cause the fluctuation of marine current power, which will have a negative impact on the grid and users. In order to smooth the fluctuating power and improve the reliability of grid-connected power, using an energy storage system to smooth the fluctuating power is considered a more effective way [2–4]. Furthermore, the marine current fluctuating power could be smoothed by controlling the pitch angle of the marine current turbine [5,6]. Sliding mode control of nonlinear controller was used to manage marine current power and improve the efficiency of the marine current turbine (MCT) in [7,8]. In addition, energy management strategies involving wind power, solar power, and electric vehicle can also be analyzed and studied due to similar characteristics. The model predictive control (MPC) strategy was used to improve the efficiency of the electric vehicles [9–11], but the computational complexity may decrease the reliability of the system. Using other wind turbines to replace the energy storage system was studied in [12]. A fuzzy logic

control algorithm was applied to achieve a reasonable allocation of power and improve the utilization of a photovoltaic power system [13]. The energy management strategy based on a multiagent system could make energy distribution become much more intelligent [14–16]. Particle swarm optimization (PSO) had the advantages of easy hardware implementation, strong global search capability, fast convergence speed, and less computational complexity, and therefore, it has become a research hotspot in the optimization field [17–19]. In order to improve the accuracy of model state estimation in power systems, the Kalman filter algorithm was applied [20–22].

Due to the specific characteristics of the marine current power system, such as the fact that the system is underwater and its maintenance is difficult, simple structure, high reliability, and better performance are the main considerations of the energy management strategy. In this paper, an energy management strategy based on rules is proposed to compensate for the predictable fluctuating power caused by tidal motion. This energy management strategy can make the grid-connected power track the grid command well while improving the operational efficiency of the vanadium redox flow battery (VRB).

2. Description and Modeling of System

2.1. System Structure

As shown in Figure 1, the MCT converts the kinetic energy of marine current into electric energy by a permanent magnet synchronous generator (PMSG). SCB is used as the auxiliary power source to absorb or release power according to the energy management algorithm based on rules proposed in this paper. VRB enables the grid-connected power to follow the grid demand according to the control rules. SCB and VRB are connected to the DC bus through a bidirectional DC/DC converter (energy can flow in two directions). The maximum power tracking of marine current power is realized by controlling a machine-side converter (MSC). Grid-connected active power and reactive power are regulated by controlling a grid-side converter (GSC).

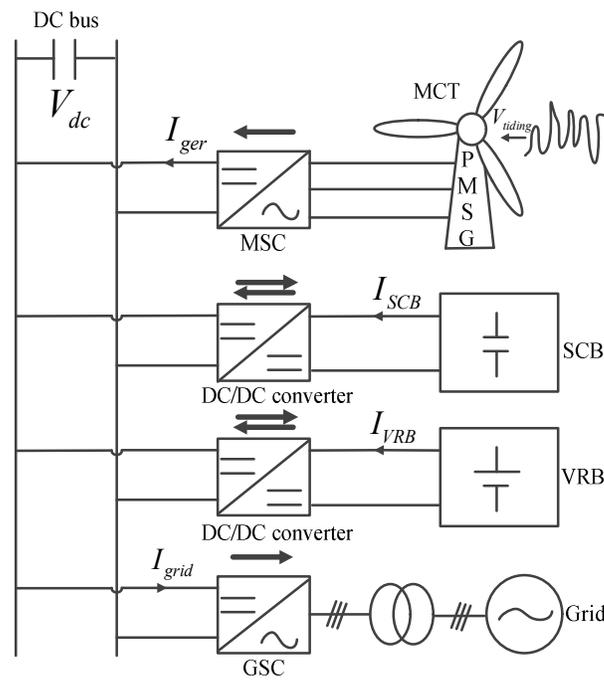


Figure 1. System structure.

2.2. Modeling and Control of MCT and PMSG

Marine current power captured by horizontal axis MCT [6] can be calculated by (1).

$$P_{MCT} = \frac{1}{2} \rho C_p(\beta, \lambda) \pi R^2 V_{tiding}^3 \tag{1}$$

where ρ represents the water mass density, which is considered as 1025 kg/m³ in this paper, R means the length of blade radius of the MCT, and the specific value is 9 m, which can be found in Appendix A: Table A1. Additionally, V_{tiding} represents the marine current velocity near the blades of MCT. C_p is the energy captured coefficient of MCT, and its value range is 0.35–0.5 for typical MCTs [2]. C_p is a binary function of tip speed ratio (λ) and pitch angle (β) based on (2).

$$\begin{cases} C_p(\lambda, \beta) = 0.22 \left(\frac{116}{\gamma} - 0.4\beta - 5 \right) e^{-12.5/\gamma} \\ \frac{1}{\gamma} = \frac{1}{\lambda + 0.08\beta} - \frac{0.035}{\beta^3 + 1} \end{cases} \tag{2}$$

In this paper, $\beta = 0$ degrees is defined and according to (2), and the relational curve of C_p - λ can be expressed in Figure 2.

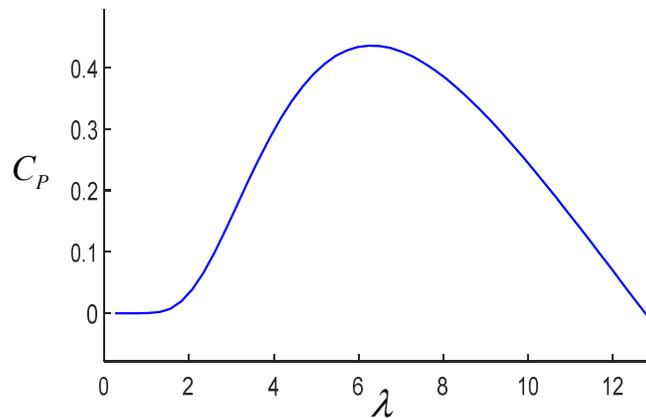


Figure 2. The curve of coefficient (C_p) and tip speed ratio (λ).

The maximum coefficient ($C_p = 0.44$) is reached when the optimum tip speed ratio λ_{opt} is 6.3. For the direct gearless system, according to (3), the λ_{opt} can be obtained when the change of generator rotor speed (ω_{ref}) keeps pace with the change of marine current velocity (V_{tiding}). This method is called the maximum power point tracking strategy based on the optimal tip speed ratio (TSR-MPPT) [23].

$$\lambda_{opt} = \frac{\omega_{ref} R}{V_{tiding}} \tag{3}$$

Higher reliability, high efficiency, and lower maintenance cost are the main characteristics of PMSG, compared with squirrel cage induction generator (SCIG); therefore, PMSG is gaining popularity [24]. In the three-phase coordinate system, it is difficult to design the controller due to the strong coupling and nonlinearity of PMSG. Therefore, coordinate conversion is introduced [25]. The voltage equation of PMSG in the d–q coordinate system is expressed by (4).

$$\begin{bmatrix} u_d \\ u_q \end{bmatrix} = \begin{bmatrix} R_s + pL_d & -\omega_e L_q \\ \omega_e L_d & R_s + pL_q \end{bmatrix} \begin{bmatrix} i_d \\ i_q \end{bmatrix} + \begin{bmatrix} 0 \\ \omega_e \Psi_f \end{bmatrix} \tag{4}$$

2.3. Mathematical Model of VRB and SCB

Supercapacitor banks (SCBs) and vanadium redox flow batteries (VRBs) [27] are used to absorb or release required power according to control rules in this paper. SCB has the characteristics of high power density and low energy density, while VRB is famous for high energy density, independent design of power and capacity, long service life, and high efficiency [28].

As shown in Figure 5a, V_{sta} , I_{sta} , V_{bat} , I_{bat} , R_{fixed} , C_{ele} , respectively, represent battery electromotive force, internal stack current, terminal voltage, terminal current, parasitic resistance, and internal electrode capacitance. R_{rea} and R_{res} are the equivalent resistance illustrated in [29,30]. The losses of VRB are mainly distributed in four aspects, namely, δ_1 , δ_2 , δ_3 , and δ_4 . They represent the internal resistive parasitic loss, external pump loss, internal reaction equivalent impedance loss, and equivalent resistance loss, respectively. The calculation formula of each loss factor [30] is expressed in Equation (6).

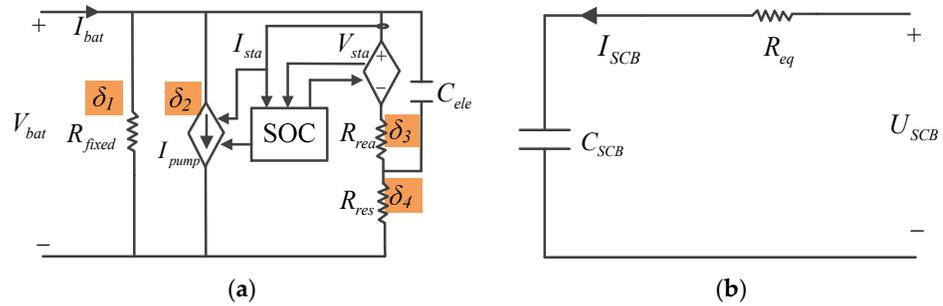


Figure 5. (a) Equivalent circuit model of VRB; (b) equivalent circuit model of SCB.

$$\begin{cases} \delta_1 = \frac{V_{bat_min}^2}{R_{fixed} \cdot P_{staN}} \times 100\% \\ \delta_2 = \frac{I_{pump} (P_N + 3\delta_1 P_{staN})}{0.2 P_{staN} |I_{sta}|} SOC_{VRB} \times 100\% \\ \delta_3 = k \frac{R_{rea}}{P_{staN}} \times 100\% \\ \delta_4 = k \frac{R_{res}}{P_{staN}} \times 100\% \end{cases} \quad (6)$$

where k is a constant; V_{bat_min} represents the minimum operating voltage of VRB; P_{staN} is the rated stack power of VRB. As the change of δ_1 , δ_3 and δ_4 is very small, they are set as constant values as $\delta_1 = 2\%$, $\delta_3 = 9\%$, $\delta_4 = 6\%$ [30] in this paper. The variation of pump loss (δ_2) is relevant to the operation power of VRB. Charge efficiency (η_1) and discharge efficiency (η_2) of VRB are defined as Equations (7) and (8).

$$\eta_1 = \frac{\int P_{sta} dt}{\int P_{bat} dt} \times 100\% = \frac{P_{sta}}{P_{bat}} \times 100\% \quad (7)$$

$$\eta_2 = \frac{\int P_{bat} dt}{\int P_{sta} dt} \times 100\% = \frac{P_{bat}}{P_{sta}} \times 100\% \quad (8)$$

where P_{bat} , P_{sta} , respectively, represent the terminal power and stack power of VRB; R_{eq} and C_{SCB} , respectively, represent equivalent resistance and equivalent total capacity [31], as shown in Figure 5b.

2.4. Modeling and Control of the Grid-Side Converter

In the d–q coordinate system, the mathematical model of the grid part can be written as Equation (9).

$$\begin{cases} e_{dg} = -R_g i_{dg} - L_g \frac{di_{dg}}{dt} + \omega_g L_g i_{qg} + V_{dg} \\ e_{qg} = -R_g i_{qg} - L_g \frac{di_{qg}}{dt} + \omega_g L_g i_{dg} + V_{qg} \end{cases} \quad (9)$$

The control structure of the grid-side converter is shown in Figure 6, where R_g, L_g, ω_g represent the equivalent resistance of the grid side, the equivalent inductance of the grid side, and angular frequency of the grid, respectively, while $i_{dg}, i_{qg}, V_{dg}, V_{qg}$ are the components of grid current and grid voltage in the d–q axis. In this paper, the GSC plays a role in making DC bus voltage constant and reallocates grid-connected power [26].

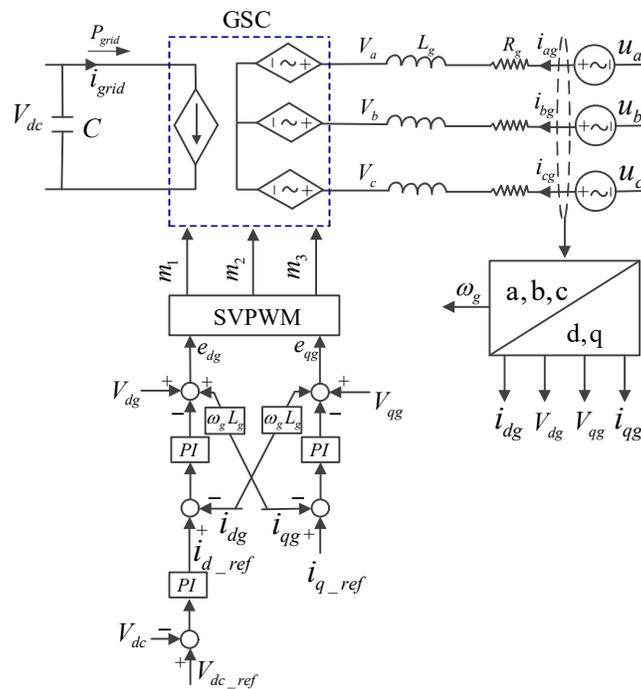


Figure 6. Control structure of the grid-side converter.

Active power in the grid-side converter is only related to the current component of the d-axis, while reactive power is only related to the current component of the q-axis. In this paper, the current reference value of the q-axis is set to zero to deliver the maximum active power into the grid. Additionally, in the d-axis, it is a double closed-loop structure of voltage and current. The DC bus voltage is stable by adjusting the grid-connected active power. The simulation result is shown in Figure 7.

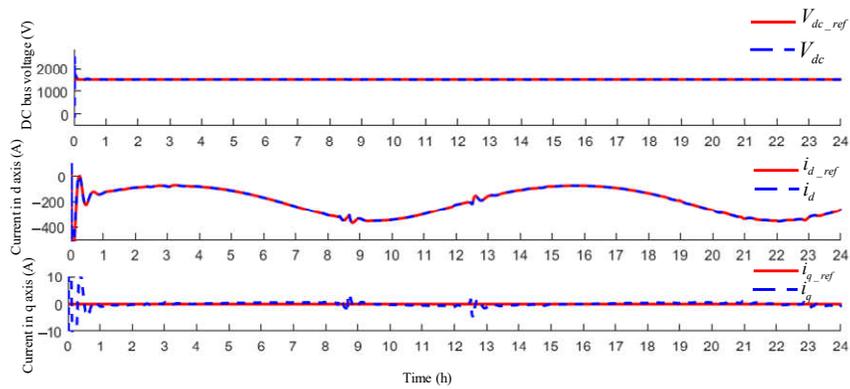


Figure 7. DC bus voltage and d–q axis current of GSC.

2.5. Mathematical Model of Marine Current Velocity

A detailed analysis of the marine current velocity model is made by Ref. [2] and Ref. [23]. The marine current velocity contains two parts, as shown in the formula presented in Equation (10). The first part C is decided by tidal speed (predictable speed), and C is considered as a constant within several minutes. The second part is induced by the swell effect [2], which is mainly caused by wind, waves, and other unknown factors.

$$V_{iding}(t) = C + \sum_{i=1}^n A_i \cos(\omega_i t + \varphi_i) \tag{10}$$

It is noticed that the frequency ω_i is around 0.05–0.2 Hz, A_i is velocity amplitude under swell effect, and φ_i is the initial phase angle.

The marine current velocity (as shown in Figure 8) in this paper is the real marine current data between Zhai Ruo Shan island and Xiao Zhai Ruo Shan island of Zhou Shan archipelago, Zhe Jiang Province [32]. In the simulation of this paper, the time scale is very large, and the core problem is to solve the operational efficiency of VRB over 24 h. Therefore, the power disturbance caused by the swell effect is not considered.

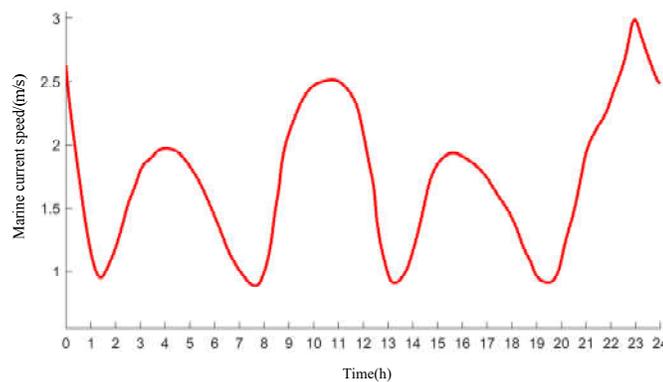


Figure 8. Tidal speed for 24 h.

3. Energy Management Strategy

3.1. Efficiency Analysis of VRB under Constant Current Charge–Discharge Test

As shown in Figure 9, with the increase of charge and discharge current, the efficiency of VRB initially increases, and after reaching the maximum efficiency, the

efficiency of VRB gradually decreases, which indicates that there is an extreme value of efficiency when charge and discharge currents are changed.

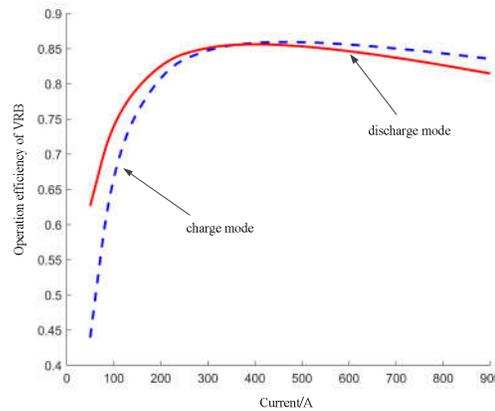


Figure 9. Current–efficiency curve of VRB.

The charging efficiency and discharge efficiency of VRB both have maximum values according to Figure 9. However, the maximum values of charging efficiency and discharge efficiency are not at the same point. Under the charging mode, when the current reaches 470 A, the efficiency of VRB reaches the maximum value of 85.96%. Under the discharge state, when the discharge current reaches 388 A, the efficiency of VRB reaches the maximum value of 85.61%. Therefore, controlling the charge and discharge current value of VRB can improve the operational efficiency of VRB.

In order to solve the problem that the efficiency of VRB is low when the charge current value and discharge current value of VRB is lower than the current threshold, an energy management strategy based on rules is proposed in this paper. The goal of the energy management strategy is to improve the efficiency of VRB in the marine current power system by controlling the auxiliary power source (SCB).

3.2. Energy Management Based on Rules Control Strategy

As shown in Figure 10, P_{grid} represents the power injected into the grid, and I_{SCB} and I_{VRB} represent the current of the SCB and the VRB, respectively. The dynamic average model is applied to replace the detailed power devices in the bidirectional DC/DC converter [26] in this paper. It is defined that the current flowing into the SCB or VRB is positive, and the current flowing out from the SCB or VRB is negative. The internal control structure of the rule-based control strategy is shown in Figure 11.

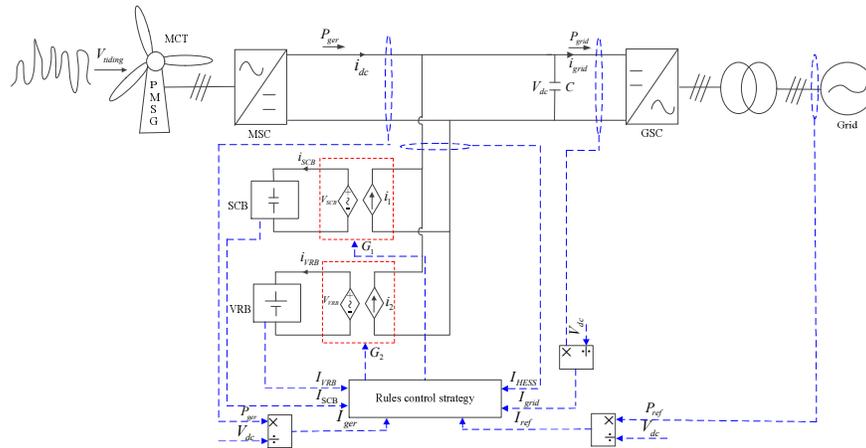


Figure 10. Control structure based on rule-based control strategy.

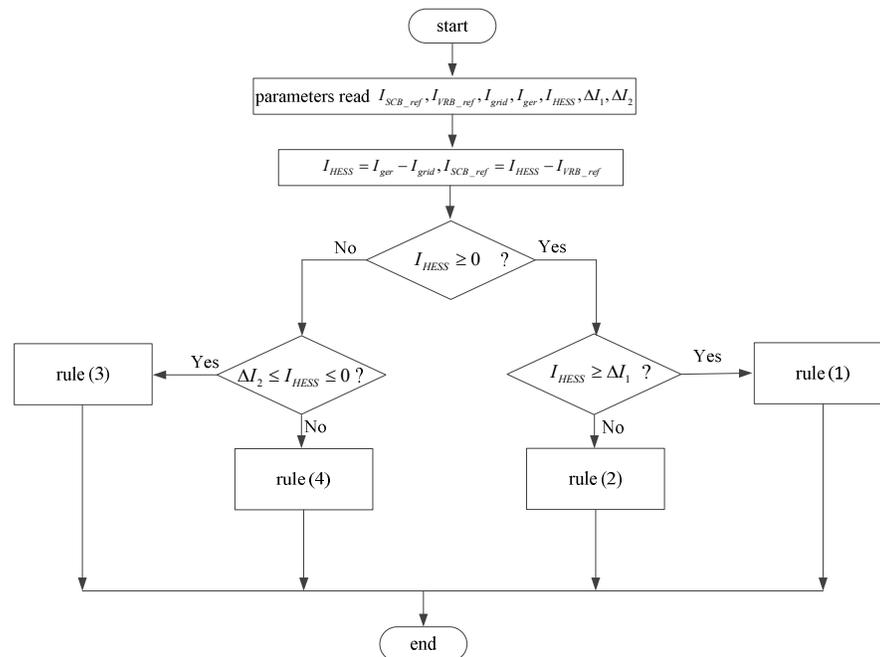


Figure 11. Internal control structure of rule-based control strategy.

As shown in Figure 11, I_{ger} is the current output by PWM rectifier on the PMSG side; I_{ref} represents the reference current of the grid. I_{HESS} represents the reference current of the hybrid energy storage system (HES). Their relationship is defined as in Equation (11).

$$I_{HES} = I_{ger} - I_{ref} \tag{11}$$

when " $I_{HES} > 0$ ", it means that the output current I_{ger} is greater than the grid reference value I_{ref} . At this moment, the HES should absorb the excess energy according to (11). Instead, if " $I_{HES} < 0$ ", which means the HES should release the required energy based on (11). If " $I_{HES} = 0$ ", the HES is in standby mode. According to the energy management strategy in Figure 11, "rule (1)" is applied if the charging current exceeds the threshold value under the charging state (" $I_{HES} > \Delta I_1$ "), and "rule (2)" applies in the situation that the charging current does not reach the threshold value under the charging state (" $0 < I_{HES} < \Delta I_1$ "). In addition, "rule (3)" is applied when the discharging

current does not reach the threshold value under the discharging state (“ $\Delta I_2 < I_{HESS} < 0$ ”). If the discharge current has exceeded the discharge threshold (“ $I_{HESS} < \Delta I_2$ ”), then “rule (4)” is running. Through the above four rules, the operational efficiency of VRB is improved.

Rule (1): when the current (I_{HESS}) is greater than the current threshold ΔI_1 under the charging mode, it means that the charging current is larger. VRB is operated in a high-efficiency state at this stage. Therefore, the current distribution of the hybrid energy storage system is shown in the formula presented in Equation (12).

$$\begin{cases} I_{VRB_ref} = I_{HESS} \\ I_{SCB_ref} = 0 \end{cases} \quad (12)$$

That is to say, the VRB takes on the task of absorbing all the surplus electric energy, and the SCB is in a standby state.

Rule (2): if the total current (I_{HESS}) is greater than zero but less than the current threshold ΔI_1 under the charging mode, it means that the system current is small at this stage, so the reference current distribution principle of VRB and SCB can be determined according to the formula presented in Equation (13).

$$\begin{cases} I_{VRB_ref} = \Delta I_1 \\ I_{SCB_ref} = I_{HESS} - \Delta I_1 \end{cases} \quad (13)$$

when the current of the hybrid energy storage system is small, the VRB is charged with the current threshold of ΔI_1 ; meanwhile, the SCB is in a discharged state, and the current of SCB flows into the VRB to maintain the current stability of VRB so as to keep the VRB running in a high-efficiency state.

Rule (3): when the current of the hybrid energy storage system (I_{HESS}) is less than zero but greater than the current threshold ΔI_2 , the reference current of VRB and SCB can be determined by the formula in Equation (14).

$$\begin{cases} I_{VRB_ref} = \Delta I_2 \\ I_{SCB_ref} = I_{HESS} - \Delta I_2 \end{cases} \quad (14)$$

In this state, in order to make VRB keep in a high-efficiency state, the SCB is in a charge mode, which means part of the current from the VRB flows into the DC bus side and another part flows into the SCB. The algebraic sum of the current flowing into the DC bus side and the current flowing into the SCB is the current threshold ΔI_2 .

Rule (4): if the current of the hybrid energy storage system (I_{HESS}) is less than the current threshold ΔI_2 , this rule applies. Under this condition, the reference current of VRB and SCB is shown in the formula in Equation (15).

$$\begin{cases} I_{VRB_ref} = I_{HESS} \\ I_{SCB_ref} = 0 \end{cases} \quad (15)$$

According to (15), the current required is all provided by VRB, and SCB is in the standby state. At this moment, the VRB can still be kept in the high-efficiency range.

The control structure is shown in Figure 12. The current reference values of SCB and VRB are determined by control rules. K_{p1} and K_{i1} are parameters of PI controller in the SCB control loop, and their specific values can be found in Appendix A: Table A1. The output value of the PI controller is proportional to the charge or discharge power of SCB. The value of normalization is “ $0.2 (V_{rated})^2$ ”. In the same way, K_{p2} and K_{i2} are parameters of PI controller in the VRB control loop, and their specific values can be found in Appendix A: Table A1. The value of normalization is “ $20 (V_{rated})^2$ ”. Additionally, the control signal gain G_1 and control signal gain G_2 are output values.

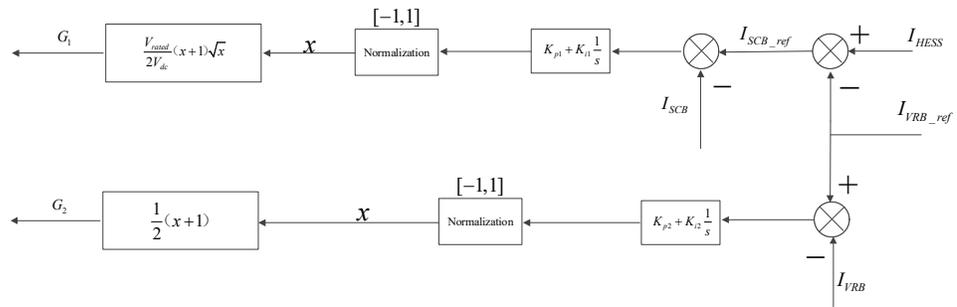


Figure 12. Control structure of SCB and VRB.

In addition, some constraints of SCB and VRB should be required as follows (16):

$$\begin{cases} |P_{VRB_min}| < |P_{VRB}| < |P_{VRB_rated}| \\ SOC_{VRB_min} < SOC_{VRB} < SOC_{VRB_max} \\ SOC_{SCB_min} < SOC_{SCB} < SOC_{SCB_max} \\ |P_{SCB_min}| < |P_{SCB}| < |P_{SCB_rated}| \end{cases} \quad (16)$$

SOC_{VRB_min} is set as 0.1 and SOC_{VRB_max} is 1, while the value of SOC_{SCB_min} is 0.2, and SOC_{SCB_max} is 1.

4. Simulation Results and Analysis

4.1. Simulation Results Based on Conventional Control Strategy

As shown in Figure 13, the conventional control strategy is applied as a comparison [29]. The power of VRB (P_{VRB}) flows to compensate for the power difference between marine current power (P_{ger}) and the grid command power (P_{ref}). However, when the charge current value and discharge current value of VRB are lower than the current threshold as the red dotted line mark in Figure 14, low operational efficiency appears (see Figure 14). Therefore, in order to improve the operational efficiency of VRB, the energy management strategy based on rules' control is proposed in this paper; detailed rules are illustrated in Section 3.

The charging current threshold " ΔI_1 " and discharge current threshold " ΔI_2 " are set as shown in Figure 14.

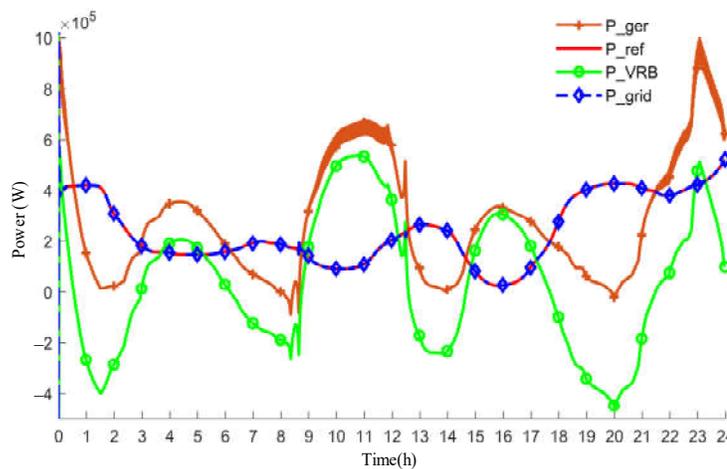


Figure 13. Power flow of marine current power system over 24 h based on conventional control strategy.

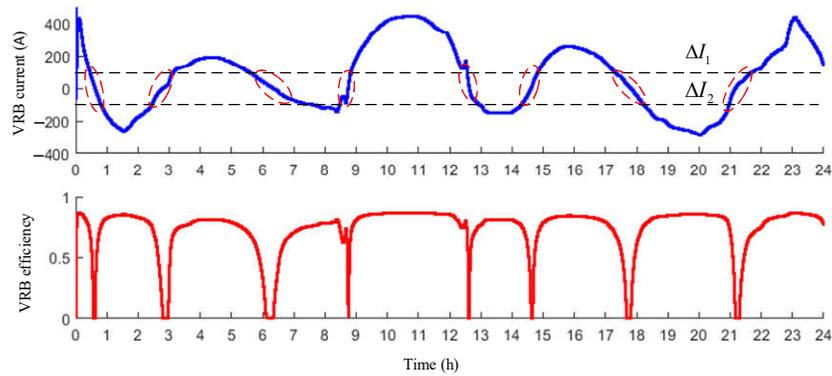


Figure 14. Current flow and operational efficiency of VRB over 24 h based on conventional control strategy.

As shown in Figure 14, the operational efficiency of VRB is in a high-efficiency operation state for most of the time over 24 h. There is a transient low-efficiency state when the charge current value and discharge current value of VRB are lower than the current threshold (“ ΔI_1 ” and “ ΔI_2 ”). The average efficiency of VRB over 24 h is defined according to the Equation (17).

$$\eta = \left(\frac{\int \eta_1 dt}{T_{charge}} + \frac{\int \eta_2 dt}{T_{discharge}} \right) \times 100\% \approx 70.83\% \tag{17}$$

where η represents the average efficiency of VRB; η_1 represents the charge efficiency of VRB; η_2 is the discharge efficiency of VRB; T_{charge} is the total charging time over 24 h, and $T_{discharge}$ is the total discharging time over 24 h. It can be seen that the average efficiency of VRB reaches 70.83% over 24 h.

4.2. Simulation Results Based on Rule-based Control Strategy

Compared with the conventional control strategy, under the control of the rule-based strategy proposed in this paper, the current flows of VRB and SCB over 24 h are shown in Figure 15, respectively. It can be seen that in the rule-based control strategy, the current becomes constant value (see marker “a” of Figure 15) and in order to maintain the current of VRB at a constant state, the SCB is discharged, which is corresponding to the “rule (2)”. As VRB is discharged by constant value “ ΔI_2 ” (shown in marker “a” of Figure 15), the SCB is charged, which indicates that “rule (3)” is running at this stage. As shown in marker “b” of Figure 15, if the current of VRB is higher than the threshold value “ ΔI_1 ”, the current of VRB remains the same as the current of the hybrid energy storage system (I_{HES}), and the SCB is in a standby state, which is corresponding to the “rule (1)”. When the VRB is in discharge state and the discharge current is less than the threshold value “ ΔI_2 ”, the current of VRB is equal to the current of the hybrid energy storage system (I_{HES}), as shown in marker “c” of Figure 15, and the SCB is still in a standby mode according to the “rule (4)”. Simulation results verify the effectiveness of the energy management strategy based on rules. Due to the constant current control near the zero-crossing point, the charging and discharging efficiency of VRB is significantly improved, which can be maintained in the range of 70–88% (shown in Figure 15). According to (17), the average efficiency of VRB reaches 80.01% over 24 h.

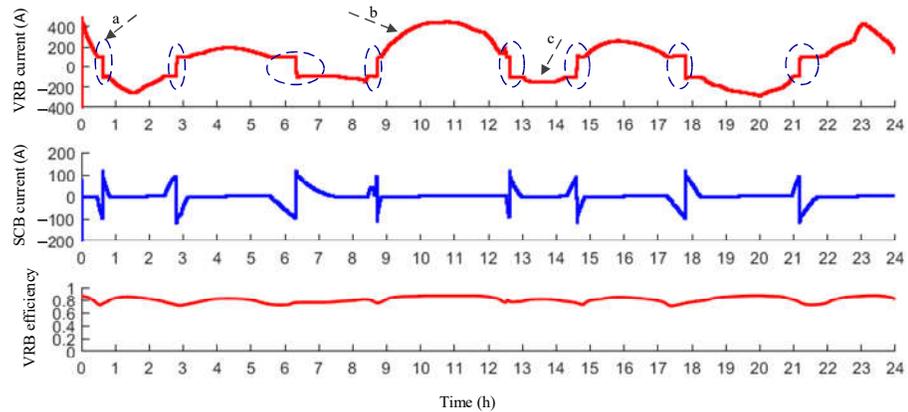


Figure 15. Current flow of HESS and operational efficiency of VRB over 24 h based on rule-based control strategy.

It is worth noting that in this paper, the time scale is over 24 h, and the core problem is to solve the operational efficiency of VRB over 24 h; therefore, the power disturbance caused by the swell effect is not considered.

BMOD0094 P075 B02 Maxwell module ($C_{SC} = 94 \text{ F}$, $V_{SC} = 75 \text{ V}$, $R_{SC} = 13 \text{ m}\Omega$) is modelled in this paper, the rated voltage is set to 750 V, which means 10 SC cells are needed in series ($N_s = 10$). The SOC_{SCB} range is set from 0.2 to 1. According to (18), V_{min} and V_{max} can be calculated.

$$\begin{cases} SOC_{SCB} = \left(\frac{U_{SCB}}{V_{rated}}\right)^2 \\ E_r = \frac{C_{series}}{2} (V_{max}^2 - V_{min}^2) \end{cases} \quad (18)$$

The available capacity E_r is 0.56 kWh. The rated power and capacity of the SCB are determined according to the mark “d” in Figure 16. As the charging time of the SCB is the longest over 24 h, it is reasonable to select this time period to determine the rated capacity. The maximum charging power of SCB in Figure 16 is 150 kW. The safety margin is considered in the primary estimation; therefore, the rated power of SCB is 200 kW. The time period under a rated-capacity power is half an hour, so the rated capacity of SCB is 100 kWh. The above series capacitor banks are regarded as a group, and then 500 groups ($N_p = 500$) are paralleled to meet the requirements of rated capacity. Therefore, the equivalent model of the SCB system is shown in Equation (19).

$$\begin{cases} C_{SCB} = \frac{N_p}{N_s} \times C_{SC} = \frac{500}{10} \times 94 = 4700(F) \\ R_{SCB} = \frac{N_s}{N_p} \times R_{SC} = \frac{10}{500} \times 13 = 0.26(m\Omega) \end{cases} \quad (19)$$

VRB plays an essential role in compensating the difference between marine current power and grid command power so that the final grid-connected power follows the grid reference power. The rated power of VRB is decided based on Equation (20).

$$P_{VRB_rated} = \max(|P_{ger} - P_{ref}|) \approx 800(kW) \quad (20)$$

Considering battery efficiency and safety margin, the rated power P_{VRB_rated} is set as 1 MW, and the rated capacity is 3 MWh. This size makes it possible for VRB to compensate for power differences under extreme marine conditions.

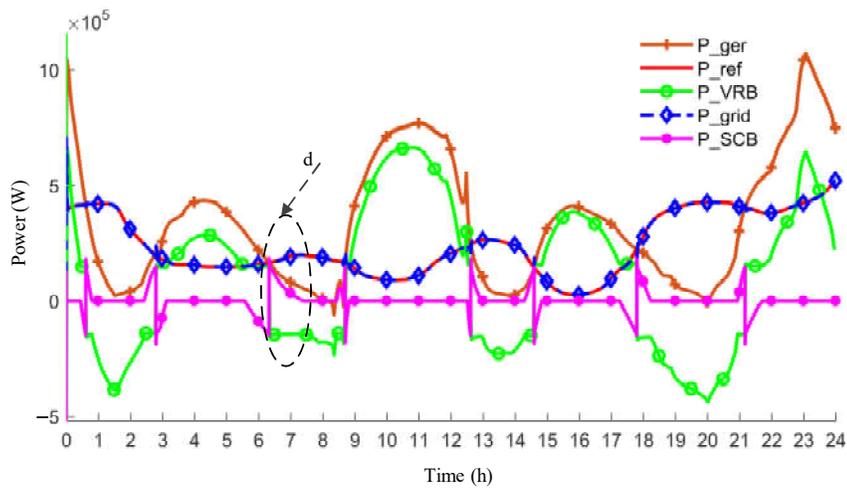


Figure 16. Power flow of marine current power system over 24 h in the rule-based control strategy.

As shown in Figure 16, the mark “d” is used to determine the rated power and capacity of the SCB; P_{ger} is the marine current power captured by PMSG; P_{ref} is the grid-connected power command; P_{grid} represents the final real = 0 time grid-connected power; P_{VRB} is the real-time power of VRB. If P_{VRB} is positive, VRB is in charge mode; otherwise, it is in discharge mode. P_{SCB} is the real-time power of the SCB and its power is positive in the charging state and negative in discharging state. Under the energy management strategy based on a rule-based control strategy, the final actual grid-connected power (P_{grid}) can track the grid power command (P_{ref}) very well.

The operational state of the four control rules over 24 h is shown in Figure 17. It can be seen that “rule (1)” and “rule (4)” are running for a long time over 24 h, which validates SCB as an auxiliary power source because it is at a standby state for most of the time during the whole day.

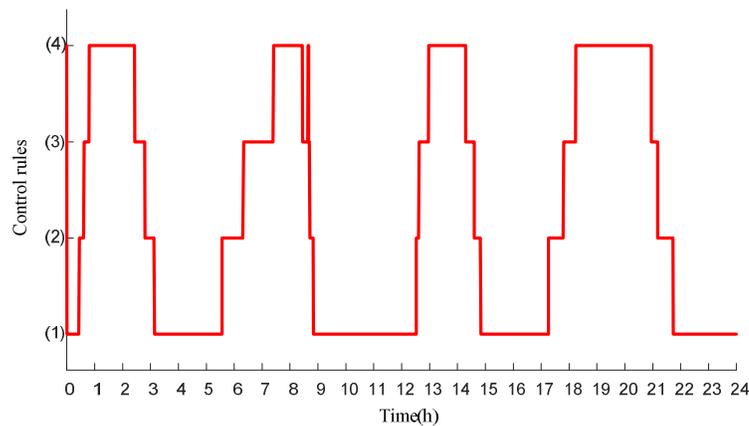


Figure 17. Operational state of rules over 24 h.

The SOC of VRB and SCB over 24 h is shown in Figure 18. It can be seen that under the control of the rule-based control energy management strategy, the SOC of VRB and SCB are both in a reasonable range. The initial SOC of VRB is set for 0.4, while the initial SOC of SCB is 0.5.

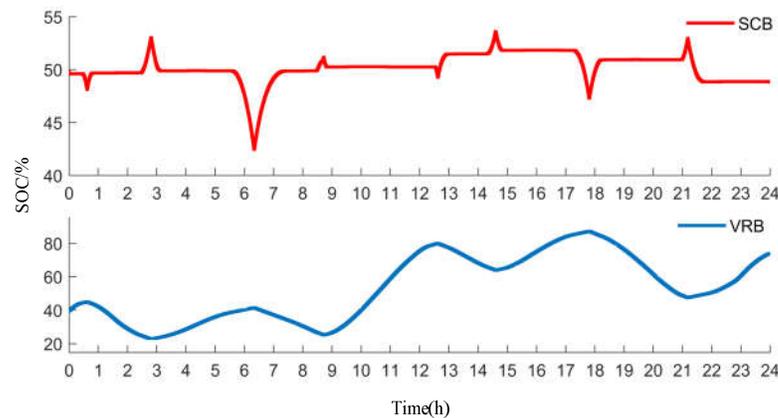


Figure 18. SOC of VRB and SCB over 24 h based on rule-based control strategy.

5. Conclusions

This paper proposes an energy management strategy based on rules to compensate for the fluctuating power caused by tidal motion while improving the operational efficiency of VRB. The marine current power system is underwater and its maintenance is not convenient. Therefore, a simple structure, higher reliability, and better control performance are the main considerations of the energy management algorithm. The simulation results indicate that VRB can make the grid-connected power track the grid command power based on both the conventional control strategy and the rule-based control strategy proposed in this paper. However, in this paper, SCB is applied as the auxiliary power source to absorb or release power according to control rules. Therefore, the operational efficiency of VRB is significantly improved, which can be maintained in the range of 70–88%. The average efficiency is 80.01%, while the average efficiency is 70.93% based on the conventional control strategy. Therefore, the energy management based on the rule-based control strategy has better control performance than the conventional control strategy. The feasibility and effectiveness of the energy management strategy are verified based on a 3 MW marine current power system established in MATLAB/SIMULINK.

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Appendix A

Table A1. Main simulation parameters of the system.

Order Number	Parameters	Values
1	R	9 m (space)
2	R_s	0.008 Ω
3	$L_d = L_q$	1.2 mH
4	Ψ_f	2.46 Wb
5	Pole pair number (P_n)	125
6	J	1.3×10^6 kg.m ²
7	Rated phase voltage of PMSG	690 V (RMS)
8	Rated phase current of PMSG	1600 A (RMS)
9	Rated voltage of MSC and GSC	690 V (RMS)
10	Rated current of MSC and GSC	3320 A (RMS)
11	R_{fixed}	54 Ω
12	R_{rea}	0.14 Ω
13	R_{res}	0.09 Ω
14	C_{ele}	5.6 mF
15	R_g	0.1 m Ω
16	L_g	1.5 mH
17	ω_g	100 π Hz
18	K_{p1}	7.6
19	K_{i1}	883
20	K_{p2}	0.6
21	K_{i2}	0.8
22	ΔI_1	100 A
23	ΔI_2	-100 A

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