Abstract—This paper proposes a capacitor condition monitoring (CM) method for modular multilevel converter (MMC) in motor drive applications. The proposed method is based on waveform decomposition of transient voltage signals, which is independent of modulation schemes and control strategies. The equivalent series resistance (ESR) change can be detected with a moderate computation requirement. Both simulation and experimental results have verified the performance of the proposed method.

Index Terms—Capacitor, condition monitoring, modular multilevel converter, transient voltage analysis, wavelet.

I. INTRODUCTION

MODULAR multilevel converters (MMCs) are becoming popular in medium-voltage variable-speed motor drives due to its superior performance in modularity and scalability, etc. [1]. The difference of it compared to that of conventional high voltage direct current (HVDC) applications are mainly in two aspects: 1) the aluminum electrolytic capacitors (Al-caps) are usually adopted for its higher power density and low cost [2]; 2) the MMC in motor drives often integrate multiple control strategies or modulation schemes to adapt to varied operation conditions [3]–[5].

Condition monitoring (CM) for Al-caps in MMC is essential for system prognostics and health management [6], the proposed CM methods in previous studies are summarized in Table I. Firstly, many methods are mainly developed for HVDC-based applications [7]–[12]. The circulating current in the MMC generates capacitor voltage ripples, which are feasible to estimate the capacitor health status. For instance, ref. [7] utilizes the inherent circulating current to estimate the capacitance, while refs. [8], [9], [13] inject additional high-frequency circulating current to enhance the health-relevant signals. Meanwhile, some literature derive the capacitor health status based on specific modulation schemes, such as the estimation of the capacitance based on the phase-shifted carrier (PSC) [10], [14], and the nearest level modulation (NLM) [11], [12]. Therefore, the aforementioned CM methods based on specific modulation schemes and control strategies have limited applicability for the motor drives applications.

The CM methods in the state-of-the-art that have better applicability to different modulation schemes and control strategies are also summarized in Table I [15]–[21]. Some of these methods utilize additional hardware to measure the capacitor health-relevant information, such as the reference submodule (SM) in [17] and [18], the tunnel magnetoresistance sensor in [19]. However, the introduced hardware in both methods require additional sensors and signal conditioning circuits, the excessive cost might be questionable for drive applications. The other kind of these methods use multiple software-only based algorithms to estimate the capacitance, such as the recursive weight least square [15], [22], Kalman algorithm [16], etc. However, the complex algorithms are often prohibitively resource-consuming. Therefore, the motor drive-based MMC needs a cost-effective CM method that is flexible to different modulation schemes and control strategies with moderate implementation complexity.

Compared to the steady-state information used in the aforementioned literature, the transient information is less affected by the control strategies or modulation schemes. Refs. [23] and [24] point out that the capacitor charging transient voltage is able to reflect the equivalent series resistance (ESR), which is one of the capacitor health indicators. Specifically, ref. [23] estimates the capacitor parameters in a buck converter by utilizing the transient information. Furthermore, a wavelet-based CM method is proposed for a boost converter [24], which demonstrates the feasibility to estimate the capacitor ESR via charging transient. However, the above methods are challenging to be extended to the MMC directly. Firstly, the additional hardware cost in [23] and [24] is proportional to the number of capacitors in the converter, which is acceptable in buck or boost converter, however, excessive in the MMC. Secondly, the computational requirement of the wavelet transform is relevant to the chosen wavelet’s complexity. Accordingly, the dedicated first-order derivative of Gaussian function in [24] would bring heavy computations in MMC. Therefore, further study to extend the charging transient voltage-based method is still necessary.

This paper proposes a capacitor CM method for the MMC based on the charging transient voltage analysis. The contributions lie in three folds:

1) The relationship between the ESR and the capacitor charging transient voltage is established, and a CM method
TABLE I

<table>
<thead>
<tr>
<th>Refs</th>
<th>Method</th>
<th>Software</th>
<th>Additional hardware</th>
<th>Applicable for control</th>
</tr>
</thead>
<tbody>
<tr>
<td>[7]</td>
<td>Second-harmonic impedance</td>
<td>+ +</td>
<td>N</td>
<td>N</td>
</tr>
<tr>
<td>[8], [9], [13]</td>
<td>High frequency source injection</td>
<td>+</td>
<td>unknown</td>
<td>Y</td>
</tr>
<tr>
<td>[10]</td>
<td>Fundamental frequency response in PSC</td>
<td>+ +</td>
<td>N</td>
<td>N</td>
</tr>
<tr>
<td>[11]</td>
<td>Capacitor voltage variation in NLM</td>
<td>+ +</td>
<td>N</td>
<td>N</td>
</tr>
<tr>
<td>[12]</td>
<td>Submodule switching times in NLM</td>
<td>+</td>
<td>N</td>
<td>N</td>
</tr>
<tr>
<td>[14]</td>
<td>Capacitor voltage phase in PSC</td>
<td>+ +</td>
<td>N</td>
<td>N</td>
</tr>
<tr>
<td>[15]</td>
<td>Recursive weight least square method</td>
<td>+ + +</td>
<td>N</td>
<td>N</td>
</tr>
<tr>
<td>[16]</td>
<td>Kalman filter</td>
<td>+ + +</td>
<td>N</td>
<td>N</td>
</tr>
<tr>
<td>[17], [18]</td>
<td>Reference submodule</td>
<td>+</td>
<td>Y</td>
<td>Y</td>
</tr>
<tr>
<td>[19]</td>
<td>Tunnel Magnetoresistance sensor</td>
<td>+ + +</td>
<td>Y</td>
<td>Y</td>
</tr>
<tr>
<td>[20]</td>
<td>Adaptive observer</td>
<td>+ + +</td>
<td>N</td>
<td>N</td>
</tr>
<tr>
<td>[21]</td>
<td>Fast-affine projection algorithm</td>
<td>+ + +</td>
<td>N</td>
<td>N</td>
</tr>
<tr>
<td>Proposed method</td>
<td>Charging transient voltage analysis</td>
<td>+</td>
<td>N</td>
<td>Y</td>
</tr>
</tbody>
</table>

"Y" means Yes, and "N" means No.

which is not limited to specific modulation schemes and control strategies is then proposed based on it.

2) Combined with a band pass filter, the haar wavelet-based discrete wavelet transform (DWT) is selected to extract the voltage step from the capacitor voltage [25]. As shown in Table I, the proposed method in this paper reduces the hardware complexity by only employing signal conditioning circuits. Meanwhile, the software computational cost is also reduced with the selected haar wavelet.

3) The further implementation methods have been investigated comprehensively, including the sampling frequency presetting, the potential error, and the effect of noise.

The outline of this paper is summarized as follows. Section II analyses the relationship between ESR and voltage step at the moment of capacitor charging. Section III proposes the wavelet-based CM method to estimate the ESR with the transient voltage step. Section IV presents the hardware design and analyses the estimation error considering practical implementation. Section V and VI verify the proposed method in the simulation and experiments, respectively. The conclusions are drawn at last.

II. CAPACITOR CHARGING TRANSIENT VOLTAGE IN MMC

This section analyzes the relationship between the capacitor charging transient voltage and the ESR. The ESR change serves as an indicator of the capacitor health status.

A. MMC configuration

The topology of a typical three-phase MMC system is shown in Fig. 1. The MMC consists of six arms. Each arm has \( N \) SMs and an arm inductor \( L_{\text{arm}} \). \( V_u \) and \( V_l \) are the upper and lower arm voltages, respectively, and \( V_{\text{dc}} \) is the dc voltage. The SM consists of two IGBTs and a capacitor bank (\( C_{\text{SM}} \)). \( i_{\text{arm}} \) and \( i_{\text{cap}} \) are the arm and the capacitor charging current. The capacitor terminal voltage is denoted as \( v \).

B. ESR effect on the capacitor charging transient

To illustrate the effect of the ESR on the capacitor’s charging transient voltage, a simulated case study is shown in Fig. 2. The ideal capacitor with ESR = 0 \( \Omega \) has a smoothly charging process. Whereas, the capacitor with ESR = 0.05 \( \Omega \) has a voltage step at the charging moment. It implies that the capacitor’s charging transient behavior is possible to be used to monitor the ESR.

To establish the analytical relation between the ESR and the voltage step, the capacitor’s charging transient is equivalent as shown in Fig. 3(a), and the capacitor terminal voltage is
When the SM is inserted \((S)\) which is better to extract the step information here and easy to apply in digital processors. Moreover, the reduced complexity alleviates the computations.

expressed as

\[ v(t) = v_C(t) + v_{ESR}(t). \]  

where \(v_C\) is the voltage across the pure capacitor, and \(v_{ESR}\) is the voltage drop of the ESR. Since the switching frequency in the MMC is mainly less than the order of kHz, the equivalent series inductance (ESL) is ignored in the following analysis.

The charging transient process is illustrated in Fig. 3(b). When the SM is inserted \((S = 1)\), the capacitor is charged or discharged depending on the direction of the arm current. When the SM is bypassed \((S = 0)\), the capacitor maintains its voltage. Assuming the SM is inserted at \(t_0\) and the arm current is positive, then the instantaneous charging current \(i_{cap}(t)\) at \(t_0\) is expressed as

\[ i_{cap}(t_0) = i_{arm}(t_0) \cdot H(t_0) \]  

where \(H(t_0)\) is the unit step function at \(t_0\).

According to (1) and (2), the capacitor voltage at \(t_0\) is expressed as

\[ v(t_0) = v_C(t_0) + \Delta V \cdot H(t_0). \]  

where \(\Delta V\) is the voltage step occurs at the capacitor charging transient, which has

\[ \Delta V = i_{arm}(t_0) \cdot ESR. \]  

This equation reveals the possibility to estimate the ESR with the capacitor charging transient voltage \(\Delta V\) and the corresponding arm current. However, capturing the transient voltage \(\Delta V\) has two challenges: 1) extraction of the transient voltage step from the capacitor voltage, with both information of the location and the amplitude. 2) the voltage step caused by the ESR is too small to be captured by the voltage sensors in MMC. To address them, the solutions are given in section III and IV.

III. PROPOSED WAVELET-BASED CONDITION MONITORING METHOD

This section proposes to use the DWT to extract the voltage step from the capacitor voltage. An essential challenge of the DWT is the wavelet selection. Based on the established analysis in Section II that the charging transient voltage is a step function, this paper selects to use haar wavelet instead of the conventional Gaussian wavelet [24]. The haar wavelet is similar to the step voltage at the capacitor charging moment, which is better to extract the step information here and easy to apply in digital processors. Moreover, the reduced complexity alleviates the computations.

The DWT uses a stretched wavelet to calculate the coefficients by shifting the wavelet along the time axis. The obtained wavelet coefficients contain the location and the amplitude information of the transient signals. The location information is given by the wavelet position, while the amplitude information is reflected by the detail coefficient, which is calculated by

\[ cD(\alpha, \tau) = \sum v(n)\psi_{\alpha,\tau}(n) \]  

where \(cD\) is the detail coefficient. \(\alpha, \tau\) represent the scale and shift factors, respectively. \(v(n)\) is the input capacitor voltage and \(n\) is a positive even integer that indicates the length of it. \(\psi\) is the wavelet function.

The wavelet selection is important for the detail coefficient calculation. The similarity between the original signal and the wavelet is one of the criteria in wavelet selection. According to the analysis in Section II, the capacitor charging transient voltage is a step signal, which is similar with the haar mother wavelet. Therefore, we choose the one-layer haar wavelet to decompose the capacitor voltage. The comparison of the DWT with haar wavelet and conventional Gaussian’s first derivation wavelet is shown in Fig. 4. Although the location and amplitude information of the transient voltage step can be extracted by both two wavelets, the haar wavelet used in this paper holds two advantages: 1) the haar wavelet is a kind of discrete wavelet, which fits better for decomposing the practical sampled discrete voltage; 2) the haar wavelet has a simple expression which alleviates the computation.

With the selected haar wavelet, the DWT transforms the original signal \(v(n)\) to two coefficients of half of its length. The one-layer haar wavelet used to calculate the detail coefficients is expressed by

\[ \psi_{1,1} = \left( \frac{1}{\sqrt{2}}, \frac{-1}{\sqrt{2}}, 0, 0, \ldots, 0 \right) \]
\[ \psi_{1,2} = \left( 0, 0, \frac{1}{\sqrt{2}}, \frac{-1}{\sqrt{2}}, 0, 0, \ldots, 0 \right) \]

\[ \vdots \]
\[ \psi_{1,n/2} = \left( 0, 0, \ldots, 0, \frac{1}{\sqrt{2}}, \frac{-1}{\sqrt{2}} \right) \]
In this way, the detail coefficient of the charging transient voltage is calculated with the haar wavelet by

\[ cD(1, m) = \frac{1}{\sqrt{2}} [v(2m - 1) - v(2m)] \]  

(7)

where \( m = 0, 1, \ldots, n/2 \). In this way, the relationship between the detail coefficient and the voltage step at the charging transient is

\[ cD(1, t_0) = \frac{\Delta V(t_0)}{\sqrt{2}}. \]  

(8)

Substituting (8) to (4), the ESR can be estimated by

\[ ESR = \left| \frac{\sqrt{2}cD(1, t_0)}{i_{arm}(t_0)} \right|. \]  

(9)

From (9), the ESR is only relevant with the capacitor transient information at the charging moment. Although the control strategies and the modulation schemes may change in MMC-based motor drives, the transient information can be extracted as long as the capacitor is charged.

IV. IMPLEMENTATION OF THE PROPOSED CM METHOD

The aforementioned analysis shows a theoretical feasibility to extract the capacitor ESR by the charging transient voltage. This section further demonstrates the implementation, including the minimum hardware design, the potential error analysis, the CM monitoring interval selection and a flowchart.

A. Transient voltage measurement circuit

The small magnitude of the charging transient voltage is challenging to be captured by the voltage sensors of the MMC directly. To extract the transient voltage with minimum hardware, this paper utilizes a band-pass filter and the spare channels of the amplifier, as shown in Fig. 5.

The high-amplitude and low-frequency voltage fluctuations are bypassed by a designed band-pass filter to provide a better resolution for the charging transient behavior. Specifically, \( R_h \) and \( C_h \) form a high-pass filter firstly. The crossover frequency is set as 2 kHz \((R_h = 2.5 \, \text{k}\Omega \text{ and } C_h = 33 \, \text{nF})\) to filter out the low-frequency voltage. Next, \( R_l \) and \( C_l \) form a low-pass filter against the high-frequency background noise. It should be noted that the crossover frequency of the low-pass filter should be higher than the switching frequency, because the charging transient voltage is highly relevant to the switching behavior. Therefore, the crossover frequency of 80 kHz \((R_l = 68 \, \text{Ω} \text{ and } C_l = 33 \, \text{nF})\) is selected in this case study.

To fully utilize the measuring range of the voltage sensors or ADs, an amplifier is designed in Fig. 5 with three resistors, i.e., \( R_1, R_f \) and \( R_s \). The amplifier ICs in MMC SMs usually have spare channels. For instance, a SOIC14 package amplifier (OP4177ARZ) shown in Fig. 5 is one of the typical selections. If the channel A, C and D are used for capacitor voltage, temperature and humidity measurement, the spared channel B can be used to capture the charging transient voltage. By this way, the proposed solution only needs several resistors and capacitors to achieve a minimum hardware requirement. In this case, a gain of the amplifier is set as 5 with \( R_f = 4 \, \text{k}\Omega \) and \( R_1 = 1 \, \text{k}\Omega \). The \( v_{apo} \) is the output of the amplifier.

B. Sampling frequency selection

The above analysis is based on the assumption that the sampling instant is synchronized with the voltage step, which rarely happens in a practical system. Therefore, the sampling error caused by unsynchronized sampling instant is analyzed in detail, as shown in Fig. 6. Assuming the first sampling instant is \( t_0 \) and the second sampling instant is \( t_1 \), the sampled error is analyzed under three different situations.

Fig. 6(a) shows the situation where the transient voltage step is included in the two sampling instants and the voltage step happens immediately after the first sampling instant. The blue curve represents the ideal sampled voltage with infinite high sampling frequency, while the red curve represents the sampled voltage with a low sampling frequency. Therefore, the \( \Delta V_0 \) and \( \Delta V \) are regarded as the ideally and the practically sampled voltage step. The \( \Delta V \) can be expressed as

\[ \Delta V = \Delta V_0 + \Delta v_{cap} \]  

(10)

where \( \Delta v_{cap} \) is the additional voltage step caused by the delayed sampling period \( \Delta t \). As the transient analysis in Section II, \( \Delta v_{cap} \) is dominated by capacitance, which thus has an expression of

\[ \Delta v_{cap} = \frac{1}{C} \int i_{cap}(t)dt = \frac{1}{C}i_{arm}(t_0)\Delta t \]  

(11)

where the charging current \( i_{cap}(t) \) during \( \Delta t \) is considered as a constant and equals to the instantaneous value \( i_{arm}(t_0) \). Therefore, the practically ESR is overestimated as

\[ ESR = \frac{\Delta V + \Delta v_{cap}}{i_{arm}(t_0)} = ESR_0 + \frac{1}{C}\Delta t. \]  

(12)

where \( ESR_0 \) denotes the initial \( ESR \). The second item in (12) represents the estimation error caused by the delayed sampling. The estimation error \( \sigma \) is defined as

\[ \sigma = \frac{ESR - ESR_0}{ESR_0} = \frac{\Delta t}{ESR_0}. \]  

(13)

The minimum sampling frequency can be preset with a given maximum error according to (13). However, the practical sampling frequency must be higher than the theoretical value considering the additional noise, which is expressed by

\[ f_{sample} > \frac{1}{\sigma \cdot C \cdot ESR_0} \]  

(14)
The estimation error would change when a capacitor is degraded. In a practical degraded capacitor, the capacitance would decrease simultaneously with the increase of the ESR. The end-of-life (EOL) for an Al-cap is 80% capacitance decrease and two times increase of the ESR. Therefore, the variation of estimation error is discussed with two worst situations: only capacitance changes and only ESR changes. When only the capacitance decreases, the estimation error would be 1.25 times than the initial one, while the error would be 0.5 times than the initial one when only the ESR increases. Therefore, the estimation error would vary 0.5 ∼ 1.25 times of the initial one during the degradation.

Fig. 6(b) shows the situation where the transient voltage step is included in the two sampling instants and the voltage step happens after the first sampling instant. This situation is similar but better than the aforementioned one because capacitor voltage before the first sampling instant is a constant, therefore, the additional voltage step caused by the delayed sampling period is smaller.

Fig. 6(c) shows the situation where the transient voltage step rising process is sampled by the second sampling instant. In this situation, the estimated ESR would be lower than the real value. The estimation results in this situation are considered as outliers.

Accordingly, the estimated ESR would fluctuate around the real value due to the unsynchronized sampling instants. Therefore, further data analysis methods could be employed to improve the accuracy according to the data distribution characteristics.

C. Condition monitoring interval selection

The arm current also contributes to additional errors to the estimated ESR according to (9). Specifically, a small drift $\varepsilon$ would be introduced to the arm current due to the unsynchronized sampling. The drift is relevant with the current change rate and the unsynchronized time $t_e$, which is expressed by

$$\varepsilon = i'_{\text{arm}}(t) \cdot t_e = \omega A \cdot \cos(\omega t) \cdot t_e$$  \hspace{1cm} (15)$$

where $\omega$ is the MMC ac-side line frequency and $A$ is the amplitude of the arm current. $i'_{\text{arm}}(t)$ is the current change rate and equals to the derivation of the arm current. The current change rate is close to 0 when the arm current is measured around its maximum value $[t = \pi/2 \pm 2k\pi \ (k = 0, 1, 2, \cdots)]$. Therefore, the estimation error caused by the small drift can be eliminated by utilizing a preset threshold $i_{\text{armth}}$ to extract the maximum arm current. It should be noted that for variable-load condition motor drive-based MMC systems, the threshold should be set according to the current change rate due to the variable current amplitude.

D. Proposed CM method flowchart

A flowchart shown in Fig. 7 illustrates the implementation of the proposed method. The dashed blue box shows the existed hardware in a practical MMC system, while the dashed red box

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Symbols</th>
<th>Simulation (Full-scale)</th>
<th>Experiment (Down-scale)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Rated power</td>
<td>$P$</td>
<td>1.2 MW</td>
<td>1.4 kW</td>
</tr>
<tr>
<td>Power factor</td>
<td>$\cos\varphi$</td>
<td>0.8</td>
<td>0.8</td>
</tr>
<tr>
<td>Phase voltage</td>
<td>$v_j (j = a, b, c)$</td>
<td>4000 V</td>
<td>80 V</td>
</tr>
<tr>
<td>dc-bus voltage</td>
<td>$V_{dc}$</td>
<td>8000 V</td>
<td>200 V</td>
</tr>
<tr>
<td>Number of SM per arm</td>
<td>$N$</td>
<td>10</td>
<td>4</td>
</tr>
<tr>
<td>SM capacitance</td>
<td>$C_{\text{SM}}$</td>
<td>2 mF</td>
<td>0.8 mF</td>
</tr>
<tr>
<td>Arm inductance</td>
<td>$L_{\text{arm}}$</td>
<td>1 mH</td>
<td>1 mH</td>
</tr>
<tr>
<td>carrier frequency of PSC</td>
<td>$f_c$</td>
<td>500 Hz</td>
<td>1.5 kHz</td>
</tr>
<tr>
<td>sampling frequency</td>
<td>$f_s$</td>
<td>100 kHz</td>
<td>25 kHz</td>
</tr>
</tbody>
</table>
Fig. 8. Extracted signals and ESR estimation results with PSC modulation in the simulation. (a) without circulating current suppressing control and (b) the corresponding estimated ESR. (c) with circulating current suppressing control and (d) the corresponding estimated.

Fig. 9. Extracted signals and ESR estimation results with NLM modulation in the simulation. (a) without circulating current suppressing control and (b) the corresponding estimated ESR. (c) with circulating current suppressing control and (d) the corresponding estimated.

shows the additional hardware introduced by the proposed CM method. The transient voltage is extracted by the band-pass filter and enhanced by the amplifier. Meanwhile, the measured arm current is compared with the preset threshold $i_{armth}$ to
ensure the capacitor is monitored around its maximum value. Then, the ESR is estimated combined with the arm current and the DWT results.

### V. SIMULATION VERIFICATION

A three-phase MMC simulation model is established in Plecs to validate the proposed method’s robustness to different modulation schemes and control strategies. The parameter of the full-scale simulation model is shown in Table II. The ESR are set as 50 mΩ, 75 mΩ and 100 mΩ to represent different degradation levels of the capacitor. The influence of system noise on the proposed method is also studied.

#### A. Effect of circulating current suppressing control

The effect of circulating current suppressing control strategy is analyzed both in NLM and PSC. The simulation results are shown in Fig. 8 and Fig. 9, respectively. The sub-figures (a) and (c) show the extracted voltage step and the corresponding detail coefficient without and with the circulating current suppressing control. From the results, the charging transient voltage can be extracted regardless of whether the circulating current suppressing control is activated or not. The DWT further extracts the voltage step from the signal, the location and amplitude information is then obtained by the detail coefficient. According to (13), if the maximum allowable error is set to be 10%, then the minimum sampling frequency should be 100 kHz. The ESR estimation results are shown in (b) and (d), the individual ESR estimation error is limited within 10%. What’s more, since the estimated results are evenly distributed, the averaging method is chosen to do the further data analysis according to the aforementioned analysis. The maximum error when taking the average value as the final result is near 0.00% under PSC and 3.5% under NLM.

#### B. Effect of different modulation schemes

The effect of modulation scheme on the proposed CM method can be analyzed by comparing the results of Fig. 8 with Fig. 9. Different modulation schemes do affect the capacitor’s steady-state voltage, where the prior arts are difficult to adapt to the largely varied capacitor behavior. However, the ESR estimation results in (b) and (d) show that the proposed CM method has a good robustness to different modulation schemes. The maximum error when taking the average value as the final result is 0.13% under PSC and 2.7% under NLM.

The above ESR estimation results show that the estimation error under NLM is larger than the PSC, which is caused by the different data size in the average value calculation. The switching frequency in NLM is much lower than the PSC, therefore, the charging transients are less in NLM during the same time. The estimation error can be reduced by expanding the estimation time. Fig. 10 shows the estimation results under
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Fig. 13. Experimental results with circulating current suppressing control. (a) Measured signals of group I and II. (b) zoomed view. (c) ESR estimation results.

Fig. 14. Experimental results without circulating current suppressing control. (a) Measured signals of group I and II. (b) zoomed view. (c) ESR estimation results.

the NLM scheme in 2 s, which takes the without circulating current control strategy as an example. The results show that the estimation error can be reduced to 0.2% with longer timescale.

C. Effect of system noises and switching frequency

The ideal charging transient voltage in the simulation proves the feasibility of the proposed method. However, the system noise in a practical system would unavoidably affect the charging transient. To investigate noise’s effect, noise with different levels [signal-to-noise ratio (SNR) = no noise, 40 dB, 30 dB and 20 dB] are added into the original simulated voltage and current signals. The CM results distribution is shown in Fig. 11. From the results, the standard deviation (std) of the CM results increases with the increased noises level. However, the average value of the results always keep close to the real ESR. Therefore, taking the average value as the final result could improves the ESR estimation accuracy.

To investigate the influence of switching frequency, the proposed method is verified with different carrier frequency. The test condition is with PSC modulation scheme, without circulating current suppressing control, and the ESR value is 0.05 Ω. The ESR estimation results (average value) are shown in Table. III. The results show that the ESR is estimated precisely under different switching frequency.

VI. EXPERIMENTAL VERIFICATION

A three-phase down-scaled MMC platform is built in laboratory to verify the proposed method, as shown in Fig. 12. The experimental platform configuration is shown in Table. II. As the NLM modulation scheme is usually applied to MMC that the number of SMs is larger than ten. The following experiments are based on the PSC modulation scheme only. The experiments test the effectiveness with and without activating the circulating current suppressing control. To emulate the health status of the capacitor, two groups of capacitor banks are implemented. Group I uses the two parallel identical capacitors while Group II uses a capacitor only. Compared to Group I, the ESR of Group II increases by 100% which typically regards reaching the one of the end-of-life criteria of Al-Caps.

A. Without circulating current suppressing control

The experiment results with the circulation current suppressing control are shown in Fig. 13(a) and (b). It shows that the voltage step can be extracted and amplified with the designed hardware in each charging transient, which is $v_{apo}$. Fig. 13(c) shows the ESR estimation results with $v_{apo}$, the results fluctuate in a certain range due to the effects of temperature, frequency, noise, etc.. To avoid the effect of operation conditions and individual difference, we use two repeated tests under three
different load conditions (load resistor $R_L = 12.4 \, \Omega$, $8.7 \, \Omega$ and $4.6 \, \Omega$) to show the results distribution. The red central mark indicates the median value, the bottom and top edges of the box indicate the 25th and 75th percentiles, respectively. The outliers are plotted individually using the red ‘+’ marker symbol. It is clear that the noises would lead to many outliers in the estimated ESR, which results in the unevenly distributed results. Therefore, the median value is chosen as the further data analysis method. The median values of the estimated ESR are shown in the table, which are approximately equal in each test. It can be inferred that the ESR estimation is independent of the operation conditions of the converter. Moreover, the average median value of group II is approximately as that of group I, which means the capacitor is degraded.

B. With circulating current suppressing control

The experimental results without circulating current suppressing control are shown in Fig. 14, where the ESR estimation results are shown in Fig. 14(c). Compared with the results with circulating current suppressing control, the median values of the estimated ESR are not influenced by the control strategy. In addition, the median value of group II is approximately twice of the group I’s as expected. Therefore, the robustness to different control strategies of the proposed CM method is verified.

With the experimental results with and without circulating current suppressing control, the degraded capacitor is identified in both situations. Although it is not experimentally proved with the NLM scheme due to the power limit in laboratory, the lower switching frequency of the NLM would make the proposed method easier to be applied. Therefore, the feasibility of the proposed CM method is proved.

The proposed method aims to improve the robustness of previous CM methods to different system control strategies and modulation schemes. However, it still has limitations in practical applications: 1) metallized polypropylene film (MPPF) capacitors are usually employed in high power MMC applications for higher voltage capability. In such cases, the ESR is hard to be detected due to significantly lower values than that of Al-caps. Therefore, the proposed CM method is only applicable in low to medium MMC applications (microgrid, motor drive, etc.). 2) although simple and economic, the minimum hardware design is still invasive to existed MMC projects. However, it can still provide possibility for upcoming MMC projects to monitor capacitor health status.

VII. CONCLUSION

Previous CM methods for capacitors in MMC are suffering from the effect of varied control strategies and modulation schemes. This article provides a novel CM method by estimating the ESR with the capacitor’s charging transient voltage. Based on a 1.2 MW MMC simulation model with different control strategies and modulation schemes, the proposed CM method’s robustness is verified. The ESR estimation error with different control strategies and modulation schemes is limited within 10%, and the maximum error is further reduced to 3.5% by using the medium value as the final result. Moreover, the proposed method is further designed considering the practical implementation, including the minimum hardware design, the potential error and noise’s effect. Finally, the feasibility of the proposed method is proved in a down-scaled MMC platform.

REFERENCES


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