

Article

Electrical Characterization of Li-Ion Battery Modules for Second-Life Applications

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Abstract: The reuse and repurposing of lithium-ion batteries for transportation in stationary energy systems improve the economic value of batteries. A precise suitability test at the beginning of the second life is therefore necessary. Common methods such as electrochemical impedance spectroscopy (EIS) and current interrupt (CI) analysis, as well as capacity analysis, can be used for testing. In this paper, these methods are studied from the aspects of test duration, sensitivity and acquisition costs of the measuring instruments. For this purpose, tests are carried out on battery modules, which were used for transportation. It is shown that subtle differences are better detected with EIS and less accurately with the CI method. The test duration is fastest with the CI method, followed by EIS and the capacity test. Strongly aged modules are reliably detected with all methods.

Keywords: modules; characterization; aging; current interrupt; electrochemical impedance spectroscopy; second use; second life; lithium-ion battery



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1. Introduction

The new circular economy action plan from the European Commission aims to facilitate the reuse and repurposing of batteries [1]. Due to the high energy density, efficiency close to 95% and a low self-discharge when not in use, lithium-ion batteries are mainly used in many applications [2].

Kwade et al. investigated the current status and challenges for automotive battery production technologies. One aspect is the life cycle consideration from manufacturing to recycling [3]. To maximize the value of batteries for transportation, second use in stationary energy systems (SESSs) is essential. Second use in a SESS is called second life. Automotive manufacturers recommend replacing batteries when the remaining energy capacity reaches 70% to 80% of the original energy capacity [4]. Rallo et al. [5] found that the aging of batteries plays a significant role in economic results, besides the price of batteries. It is expected that the price of batteries will decrease due to the volume effect of growing electromobility. This development will make SESSs economically interesting [5]. Knowing the aging and speed of aging is relevant for the safe operation of battery modules of a second-life battery storage system, as the risk of damage increases with progressive aging [6,7].

To determine the aging of used battery modules, characterization of the state of health (SOH) is necessary. Access to battery management system (BMS) data from electric vehicles would be helpful for SOH prediction, but this access is usually protected by the manufacturer [8]. The most used indicator for the SOH is the ratio of the nominal capacity to the initial capacity. This method is time consuming and thus expensive, as capacity determination usually requires at least a complete charge and discharge cycle [9,10]. Two faster techniques are electrochemical impedance spectroscopy (EIS), where the impedance

of the battery is measured at different frequencies, and the current interrupt (CI) method, where a voltage drop (or rise) is analyzed by applying a load (or charge) pulse [10–12].

Waag et al. [12] investigated the dependency of battery impedance on various conditions (SOC, temperature, short-time previous history, current rate) and aging states. Their results show the influence of the internal and external conditions on the measurement methods. The sensitivity of the characterization methods was not compared, and the applicability of the tests for incoming goods inspection was beyond the scope of their study.

Wildfeuer et al. [13] used the CI and EIS methods to characterize battery cells in different interconnections. At the module level, they observed an increase in the internal resistance over the full SOC range after the cyclization of the module. Because different interconnections were analyzed, only one module was examined. EIS measurements at the module level were not performed.

Eddahech et al. [14] studied the influence of the SOH on the constant voltage (CV) phase during the charging process. They discovered a relationship between capacity loss and reduced charging time in the CV phase. However, a charging process is necessary to perform this characterization. Because the CV phase takes about 30% of the charging time due to overvoltages, fast characterization for an incoming goods inspection is not feasible. Furthermore, the CV voltage must be reached to perform this characterization. Weng et al. [15] studied the incremental capacity analysis (ICA) method to monitor the SOH of battery modules and packs. The voltage plateaus on the charging/discharging curve, which are displayed as peaks in the incremental capacity curve, are dependent on the SOH and are related to the intercalation process. For this characterization procedure, complete charge/discharge curves at low currents ($<1/25$ C) must be available. Therefore, this characterization method is not feasible for incoming goods inspection.

The influence of contact resistances at the cell pack (CP) level of modules is discussed by Campestrini et al. [16]. Campestrini et al. observed increased contact resistance on the CP level of the modules they built from cells. Cell resistance at the CP level seemed higher than it actually was. Therefore, this study focused on evaluating the module level and not each cell pack level of the modules. Disassembling modules into cell packs and reassembling equally aged cell packs also involves additional work; therefore, a general quality distinction of the modules should be sufficient for an inwards goods test.

To summarize, there are currently no studies showing the dependency of CI, EIS and capacity tests at the module level at the beginning of second life, and there is no comparison of their sensitivity to detect the SOH. The primary aim of this study is to investigate the dependencies of the named methods and to classify the applicability under the aspects of test duration, sensitivity and investment costs of the measuring devices. For this reason, five battery modules from an electric vehicle (EV) are analyzed with the CI and EIS methods. To apply the methods, the requirements of the measurement setup are considered in more detail. Subsequently, the results are compared to the measured capacity of the modules, and the sensitivity of each method is evaluated.

This makes it possible to select the most suitable measurement method for the desired application in practice. The CI method, which is an easy-to-integrate measurement method, can be applied without interruption during operation and is therefore considered more closely.

2. Material and Methods

The characterization procedure is divided into two parts. First, the CI method was applied using a test procedure, then the modules were tempered for 16 h and the EIS method was applied.

For both measurement techniques, it is crucial to use the four-wire measurement to reduce the voltage error caused by the current in the measurement wires.

2.1. Test Objects

The SOH investigation with the methods mentioned in the introduction was carried out on five automotive battery modules. The cell chemistry is NMC. The cells of the modules are interconnected in 4s3p (see Figure 1).

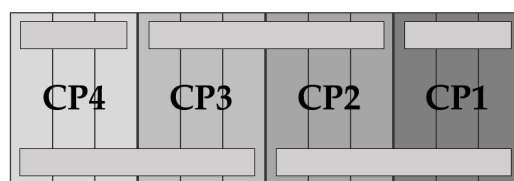


Figure 1. Schematic representation of the cell interconnection in the battery module showing the parallel connection results of 3 cells in a cell pack (CP).

The nominal capacity is given with 111 Ah. Voltage limits were given from 12.8 to 16.8 V [17]. No other aging and previous usage information were given.

2.2. CI Method and Capacity Determination

Cell voltages were measured with a National Instruments PXI System. The Chassis 1085 with a PXIe-4300 ADC measurement module was used to capture the cell pack voltages. As measurement wires, twisted-pair cables with shielded aluminum foil around the pairs and another shielding around the whole cable were used. The shielding was attached to the module housing to ensure potential equalization. To evaluate the module voltage, the cell pack voltages were summed up. The current was measured with a LEM HTA 100-S Hall effect sensor. The modules were placed in a temperature cabinet, which keeps the temperature at 20 °C to ensure that the battery modules are tested under equal conditions. A Greenlight module tester was used for the power supply.

Because the modules are supplied with 1C, very large currents (111 A) flow for the given capacity of the battery modules. The large currents next to the test leads for the voltages lead to a demanding measurement setup to reliably detect the very small internal resistance. In Figure 2, the experimental setup is shown.

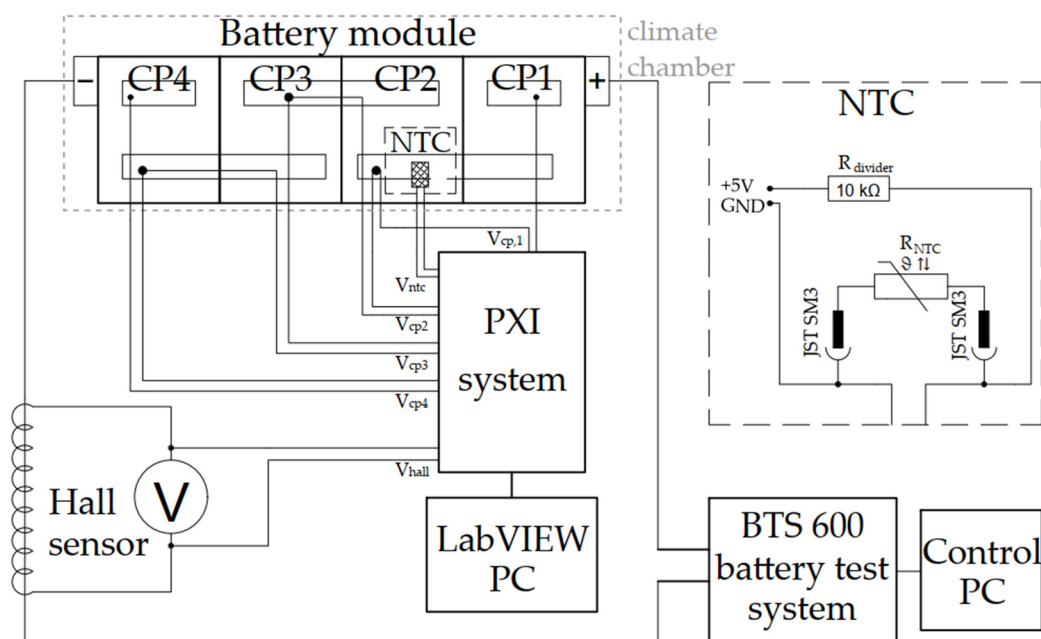


Figure 2. Schematic representation of the experimental setup for the testing procedure.

To evaluate the state of health of the modules, a testing procedure, illustrated in Figure 3, was applied to each module. The testing procedure principle is shown in Table 1.

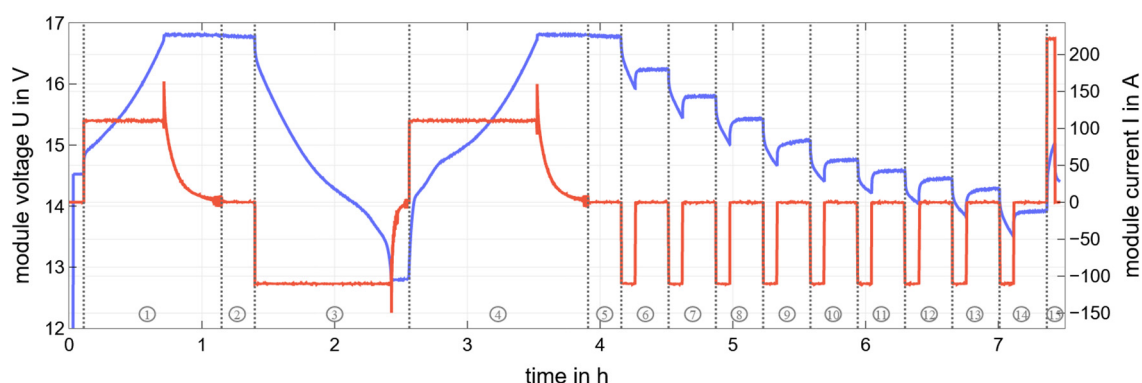


Figure 3. The testing procedure consists of different steps that can be seen over time: voltage curve of the battery module in the test (blue); current curve (red) of the battery module in the test.

Table 1. Testing procedure steps to evaluate the capacity and inner resistance of the battery modules at different SOC.

Step	Description	Parameters and Conditions
1	Full charge	@1C, CC + CV
2	Pause	15 min
3	Capacity determination	Discharge @1C, CC + CV
4	Full charge	@1C, CC + CV
5	Pause	15 min
6–14	Recurring discharge to 10% SoC	@10% SOC steps + 15 min relaxation
15	Charge	Charge while battery voltage <14.9 V @2C

Due to the undefined state of charge at delivery, the modules were first fully charged to 16.8 V. For acclimatization, a pause of 15 min was integrated. Afterward, the capacity of the battery modules was tested with a CC/CV discharge with a current of 1C.

The next step consists of a charging process for further module evaluation. By discharging the modules in 10% SOC steps and a subsequent relaxation time of 15 min, the CI method was implemented.

To store the modules, they were charged with a current of 2C to 14.9 V. Due to overvoltage during charging, a voltage corresponding to a SOC of 30% was reached after 16 h when the EIS measurement for each module was performed. By choosing a SOC of 30%, peak heat can be reduced by 50% compared to a SOC of 50% in the event of a thermal runaway [17,18]. This is a recommended SOC from a safety point of view [19]. Because safe transport of second-life battery modules with a SOC of up to 30% is possible, this SOC was selected for the incoming goods test with the EIS measurements.

The capacity is calculated by the following equation:

$$Q = \int i(t) \cdot dt \quad (1)$$

The internal resistance is calculated as followed:

$$R_i(t) = \left| \frac{\Delta u(t)}{\Delta i(t)} \right| = \left| \frac{U(t_0 - 50 \text{ ms}) - u(t)}{I(t_0 - 50 \text{ ms}) - i(t)} \right| \quad (2)$$

In order to calculate and plot the internal resistances, a Python script was used. The Python script integrates the current to determine the capacity of the modules at a 1C discharge.

Wildfeuer et al. [13] studied internal resistance curves for the timepoints 0.1, 2 and 10 s at the module level. Because only a single module was considered, changes in the internal resistance curve could be studied at different SOC. The authors define the most meaningful timepoint by considering the difference of the internal resistance value of the worst module to the internal resistance value of the module with the next lower value for the timepoints 0, 0.1, 0.2, 0.5, 1 and 60 s during the relaxation phase. To ensure a comparison between the results of the CI and EIS methods, the evaluation of the internal resistances takes place at the same SOC that was selected for the EIS measurements.

2.3. EIS Method

A Reference 3000 AE (Gamry Inc., Warminster, PA, USA) is used in combination with a Gamry Reference 30k Booster to perform the EIS measurements. The modules were tested in a frequency range from 5 kHz to 20 MHz with 10 measurement points per decade. The measuring device of the company Gamry Inc. automatically readjusted the excitation amplitude. The used function is called “Hybrid-EIS.” Here, the maximum voltage change was set to 20 mV to ensure a pseudo-linear operating point of the four serial cell packs. The excitation currents vary in this range depending on the excitation frequency between 8 and 25 A. The EIS measurements were performed the next day after the CI measurements (approximately 16 h).

The experimental setup is shown in Figure 4.

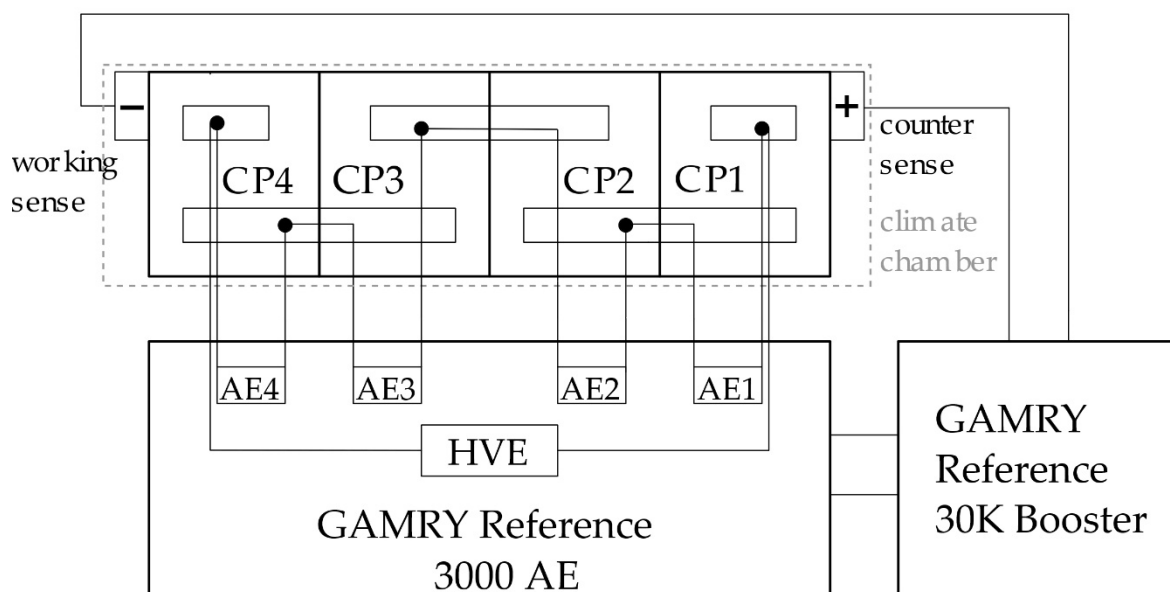


Figure 4. Experimental setup for the EIS measurements in the climate chamber.

To generate the impedance spectra, the battery modules or single-battery cells were stimulated by an alternating current with varying frequency, and the voltage response was measured. Here, it is important to pay attention to the causes of measurement errors such as parasitic inductive coupling, electromagnetic interferences, systematic errors caused by contact resistance change, violation of the system requirements (linearity, time invariance, causality) [6,20–22], different location of measurement lines and the use of uncalibrated additional adapters. A booster is needed to generate enough alternating current for EIS with battery modules or very large single cells. The effect on the impedance spectrum when the current excitation is too low is shown in Figure 5.

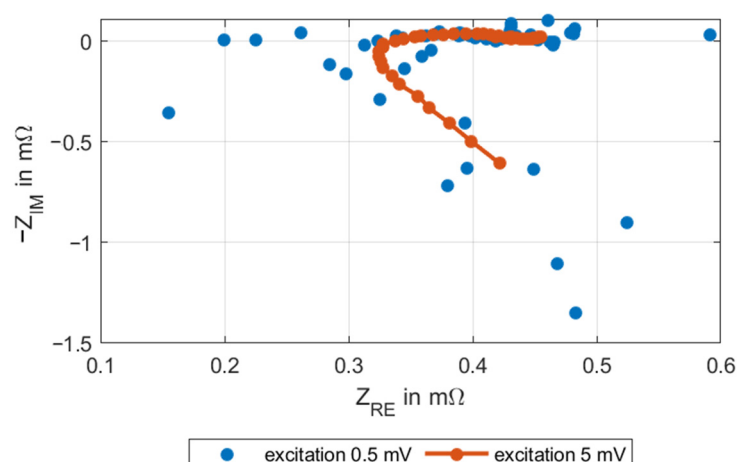


Figure 5. Excitation dependency for EIS measurements tested with a single test cell (64 Ah).

Before evaluating the module results, the influence of temperature, SOC and current excitation was investigated. For these tests, test cells that did not correspond to the cells in the modules were used.

If the current excitation is too small, the voltage change will be too small for meaningful measurements. A comparison with an excitation that is too low is shown in Figure 5.

As soon as an almost error-free EIS measurement is possible, it depends on the following three main factors: temperature, SOC, SOH. These dependencies were also studied by Waag et al. [12] and Zhu et al. [23] at the cell level.

The dependencies of temperature and SOC are shown in Figure 6.

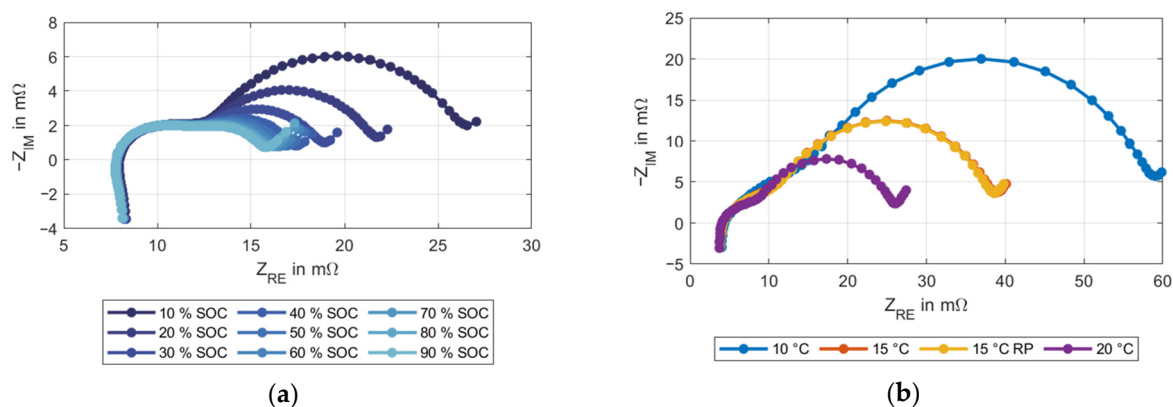


Figure 6. Schematic representation of spectra: (a) SOC dependence of the single test cell (5 Ah) measurement at 20 °C (b); temperature dependence of the single test cell (5 Ah) measurement at 20% SOC.

The temperature is assumed to be constant by using a climate chamber and 16 h of tempering time. The SOC must be determined precisely; for this purpose, OCV is consulted additionally.

Reversible aging effects, which are described by Petzl et al. [24], are brought to the same level for all modules by the previous CI test procedure. Thus, the temperature and SOC of the measurements are kept constant so that the SOH is the only parameter that affects the measurement results.

For the evaluation of the impedance spectra, the characteristic points from Figure 7 were defined.

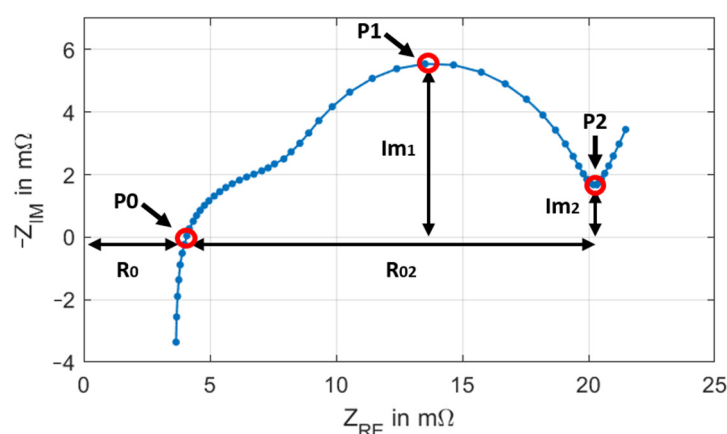


Figure 7. Schematic representation of an impedance spectrum with characteristic points that were used for evaluation.

R_{02} (see Formula (3)), Im_1 (see Formula (4)) and Im_2 (see Formula (5)) were defined as indicators, which were used for module evaluation.

$$R_{02} = Z_{RE}(P2) - Z_{RE}(P0) \quad (3)$$

$$Im_1 = Z_{IM}(P1) \quad (4)$$

$$Im_2 = Z_{IM}(P2) \quad (5)$$

Regarding evaluation, there are different challenges to precise the results, which can be seen in Table 2.

Table 2. Evaluability of the characteristic points in comparison.

Point	$Re(Z)$	$Im(Z)$	Frequency
P0	Indirect informativeness	$\stackrel{\text{def}}{=} 0$, necessary to define P0	Depends on the experimental setup
P1	Vulnerable to measurement and evaluation errors	Robust to measurement- and evaluation errors	Robust but has to be interpolated
P2	Good measurability	Good measurability	Unequivocal allocation not possible

Point $P0$ is the intersection from the impedance spectra with the real axis, representing the ohmic resistance [25]. Following this, the imaginary value is 0 and has no further information. The measuring frequency at this point strongly depends on the experimental setup and the location of the measurement wires, which can produce additional inductive coupling. Thus, the resulting information content is given by the real part of $P0$. Figure 6a,b shows the stability of the real part under different influences of SOC and temperature, which results from a high gradient of the spectra at this point.

Point $P1$ is the local maximum between $P0$ and $P2$. Here, the gradient is 0; in addition, in a wide range near to this point, the gradient is small. As a result, small differences in the evaluation of the imaginary part will show a big impact on the real part of the measurement. On the contrary, big differences in the evaluation of the real part will only show small differences for the imaginary part, which ensures the robustness of the measurements. At high SOCs, it can be difficult to determine $P2$.

Point $P2$ is the local minimum at low measurement frequencies. This point represents the polarization resistance [13]. In this area, several measurement points of different frequencies are often very close to each other, especially at high SOCs. Due to the measurement uncertainty, it is not possible to define one of these points as the true minimum. Concurrently, the accumulation of measurement points with small spreading will lead to a sufficient result regarding a machine-based evaluation.

Using $P0$, $P1$ and $P2$, it is possible to derive three scales for comparing the impedance spectra of the modules with each other. First attempts showed that the real part of the

spectra can be distorted by offsets. These pure ohmic offsets (0.07 mOhm as usual for this experimental setup) can vary between the measurements of the modules but will be stable for a single module measurement. The authors expect these errors caused by contact resistances, despite all electrical contacts, were cleaned carefully. However, it is easy to erase this measurement error by regarding the differences from $P0$ and $P2$ in the spectra, as this measurement error will affect both real parts in the same way. Fortunately, the imaginary parts will not be distorted by these errors. As a result, the following three scales will be used to compare the impedance spectra of the modules with each other.

3. Results and Discussion

First, the results of the CI method are evaluated. Afterward, the results of the EIS methods are presented. Finally, the results are compared to the classic capacity test (classic SOH).

3.1. CI Method

The internal resistance curves of the battery modules at 30% SOC in the relaxation phase are shown in Figure 8. The course of the module resistances directly after changing the current show the greatest differences; therefore, this area is shown enlarged.

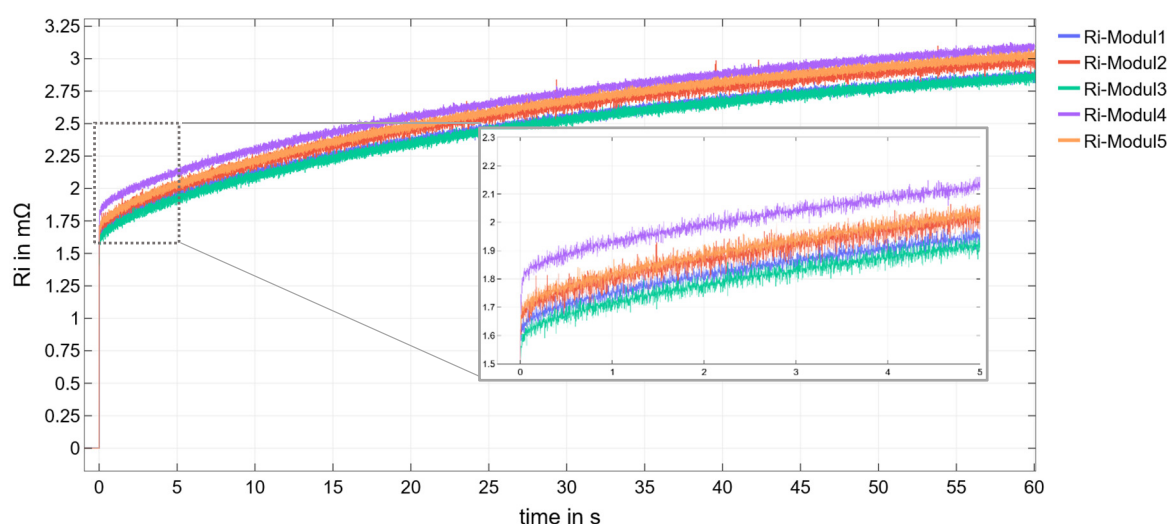


Figure 8. Internal resistance curves of the modules calculated from the voltage change in the relaxation phase at 30% SOC. In order to better recognize the differences in the first seconds, this is shown enlarged. The raw data display contains measurement noise, which underlines the importance of precise voltage measurement.

It can be seen that Module 4 always has the highest internal resistance values. The difference is greater in the front area than in the rear area. This raises the question of when the difference and thus the evaluation become best visible at what time.

To determine the most meaningful timepoint, the timepoints 0.1, 0.2, 0.5 and 1 s such as 60 s in the relaxation phase were evaluated (see Table 3). It can be seen that 0.5 s in the relaxation phase shows the largest difference and is therefore used for evaluation. A stationary current that lasts for 0.5 s is very likely to be found in the real operation of battery storage [26].

Table 3. Comparison of internal resistances at different time points, measured in the relaxation phase at 30% SOC with the CI method.

t in s	Module 1 in mΩ	Module 2 in mΩ	Module 3 in mΩ	Module 4 in mΩ	Module 5 in mΩ	$\Delta R_{i-M4-MX}$ in mΩ
0	1.41	1.38	1.41	1.46	1.42	0.04
0.1	1.65	1.70	1.64	1.83	1.75	0.08
0.2	1.68	1.70	1.65	1.85	1.75	0.10
0.5	1.70	1.75	1.65	1.89	1.78	0.11
1	1.76	1.80	1.71	1.93	1.83	0.10
60	2.84	3.00	2.88	3.08	3.03	0.05

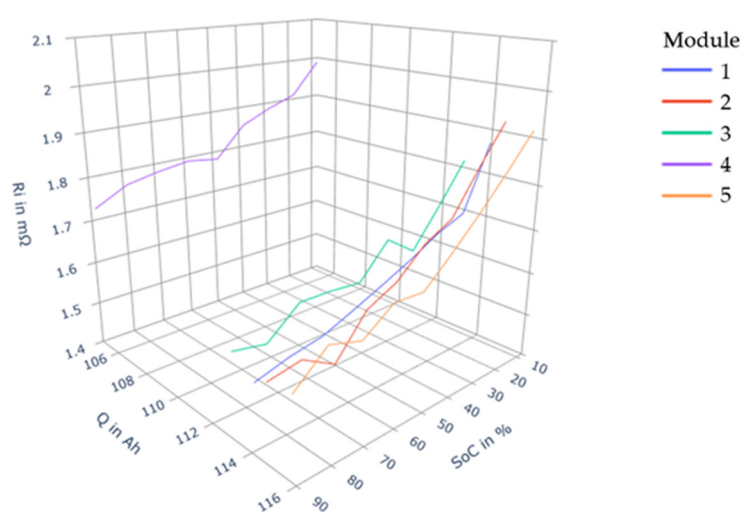
To examine the best SOC to evaluate the SOH of the modules, different working points are considered in Table 4.

Table 4. Comparison of different SOC working points at 0.5 s in the relaxation phase, internal resistances are normalized to the best module.

SOC in %	Module 1	Module 2	Module 3	Module 4	Module 5	$\Delta R_{i-M4-MX}$ in %
90	100% (1.55 mΩ)	101.03%	102.19%	111.73%	101.55%	9.54%
80	100% (1.57 mΩ)	100.51%	100%	112.79%	104.20%	8.59%
70	102.92%	100% (1.54 mΩ)	105.91%	116.11%	104.94%	10.20%
60	100% (1.61 mΩ)	100.68%	100.99%	111.48%	103.16%	8.31%
50	101.17%	102.10%	100% (1.62 mΩ)	110.17%	102.28%	7.89%
40	100% (1.67 mΩ)	102.57%	101.38%	111.18%	102.69%	8.49%
30	103.58%	106.13%	100% (1.65 mΩ)	114.68%	108.01%	6.67%
20	100% (1.73 mΩ)	106.35%	100.06%	110.57%	106.70%	3.87%
10	102.85%	105.70%	100% (1.82 mΩ)	109.27%	105.32%	3.57%

For this experimental setup, it is shown that the best working point to evaluate the module SOH is at 70% SOC after 0.5 s in the relaxation phase.

Figure 9 shows a 3D view of the measured values. Here, the amount of charge is integrated into the evaluation by the capacity measurement. The charge quantity describes the aging state according to the often-used classical SOH definition.

**Figure 9.** Three-dimensional (3D) characteristics of CI measurements. The increase in the internal resistance to low SOC is noticeable, and Module 4 (printed in purple color) has a significantly larger internal resistance curve. The charge quantity is plotted negatively because it is counted negatively in the discharge direction.

It can be seen that Module 4 always shows higher values than the other modules. This correlates with the consideration of the capacity test. The results of the internal resistances are, however, more differentiable in the further evaluation than the capacity test and can be determined during the operation of the system.

Because the comparability to the EIS measurement must be given, in the following, the evaluation of the EIS measurements is performed at 30% SOC (see Figure 10).

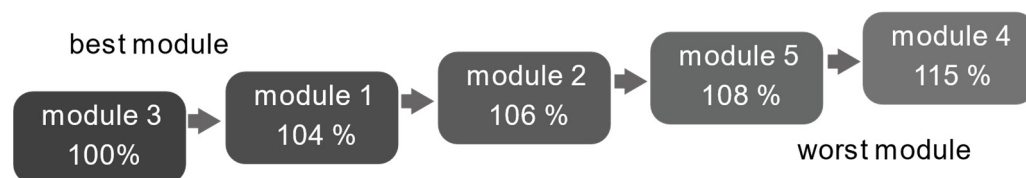


Figure 10. Quantitative evaluation of module aging by the CI method at 30% SOC.

3.2. EIS Method

Figure 11 shows that the impedance spectra vary between the measured modules. The larger the half arcs spread out, the older the battery. The measurand is big enough that module differences are visible regardless of measurement uncertainties. For easier visibility, the spectra from Modules 1 and 3 were shifted toward P0 from the other modules. This shifting has no effect on the spectra's evaluation because of the difference-based evaluation.

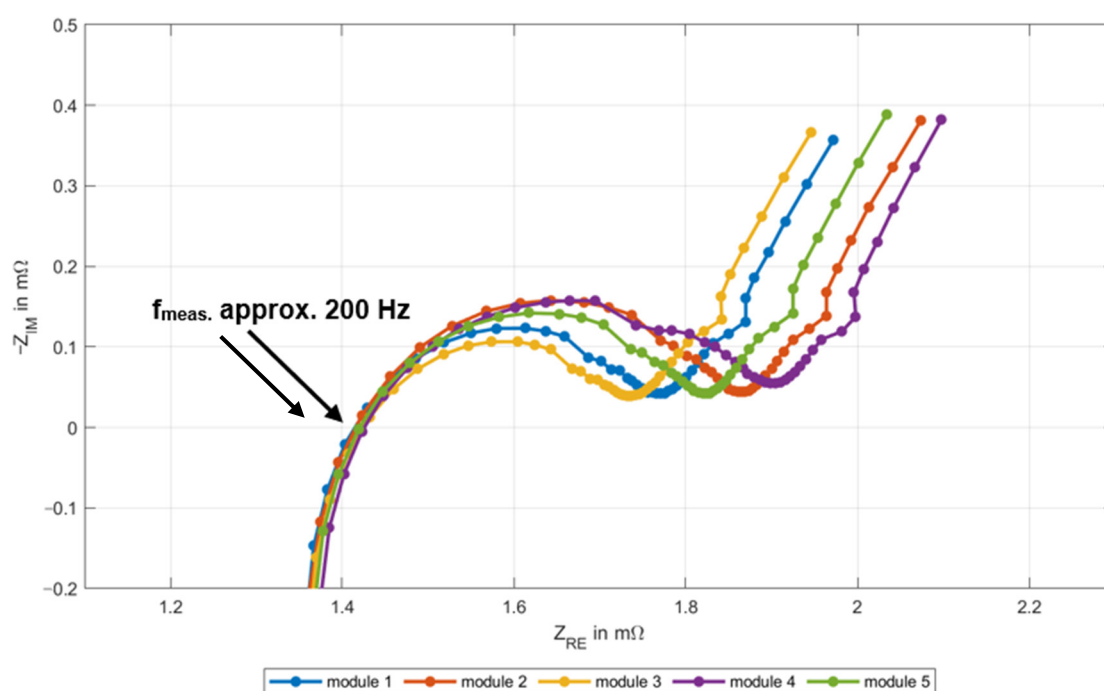


Figure 11. Impedance spectra at 30% SOC of the modules. Modules 1 and 3 were shifted.

From the available spectra, it is clear that the modules have a measurable scattering. Modules 3 and 4 are particularly noticeable. Even an optical evaluation of the spectra shows a widening of the spectrum (R_{02}) of Module 4 compared to the other modules. The two other evaluation characteristics, Im_1 and Im_2 , are also the largest in this comparison. This suggests that the aging of Module 4 is greater (and therefore its performance is lower) than that of the other modules. Module 3 is the opposite of Module 4; it has the best measurements of this comparison, which indicates less aging and potentially longer use in the second life compared to Module 4. An absolute statement about aging is not possible because all modules of the comparison could have aged strongly. For absolute statements

about aging, the comparison with a new battery module is necessary as a reference. Table 5 shows the relative deviations of the modules compared to the best value. Formula (6) describes the equal weighting of the parameters Im_1 , Im_2 and R_{02} .

$$EIS_{SOH,\emptyset} = \frac{Im_1 + Im_2 + R_{02}}{3} \quad (6)$$

Table 5. Evaluation of the module spectra with the characteristic points. Shading represents the order of the aged modules.

Module No.	Im_1	Im_2	R_{02}	$EIS_{SOH,\emptyset}$
1	115%	107%	121%	114%
2	148%	112%	148%	136%
3	100% (0.106 mΩ)	100% (0.039 mΩ)	100% (0.300 mΩ)	100%
4	148%	139%	158%	148%
5	133%	106%	135%	125%

The best module of this test method is Module 3, and the worst is Module 4. It is clear that modules perform consistently well or poorly on the selected evaluation criteria. More differentiated evaluations would be desirable but are not possible due to the instability and evaluability of other partial ranges of the impedance spectra. Im_1 and R_{02} proved to be particularly suitable aging indicators. Here the differences in measurement results were most pronounced.

3.3. Capacity Test

During the capacity tests, the modules used here show measurement results that exceed the nominal capacity of new modules. This may be due to the additional integration over the CV step, but the discharge amount in this step is small. Due to the increased capacity, the authors assume that the modules are underestimated in the data sheet.

For the capacity evaluation, the module with the largest capacity was used as a reference and all other modules were normalized to this module (see Figure 12).

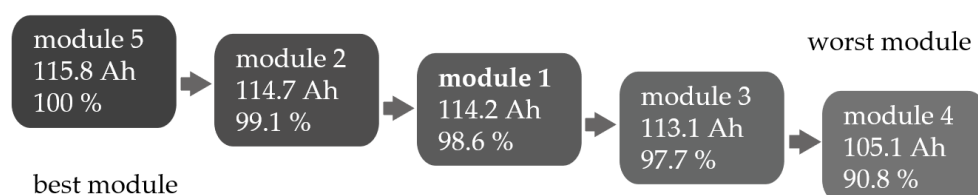


Figure 12. Evaluation of the capacity of the battery modules from Step 3 of the test procedure.

The percentage deviation from the best to the worst module is 9.2%. A heavily aged module is thus well recognizable by the capacity test. More subtle differences (e.g., between Modules 1 and 2) are difficult to detect.

3.4. Comparison of Methods

The modules were characterized using EIS, the internal resistance tests and the capacity tests. From the measurement results, a relative aging condition evaluation of the modules could be generated. The results of the relative aging condition assessments vary depending on the measurement procedure. The results are summarized in Table 6.

A low internal resistance value is an indicator for a good module using the CI method; the same applies to the measured value R_{02} using the EIS method. In the capacity test, this is reversed. Here, a large charge quantity is an indicator of a good module.

Table 6. Comparison of the measurement methods. Shading represents the order of the aged modules.

Measurement Method	Best Module	\leftrightarrow				Worst Module
CI/internal resistance	Module 3 100%	Module 1 104%	Module 2 106%	Module 5 108%	Module 4 115%	
EIS	Module 3 100%	Module 1 114%	Module 5 125%	Module 2 136%	Module 4 148%	
Capacity test	Module 5 100%	Module 2 99.1%	Module 1 98.6%	Module 3 97.7%	Module 4 90.8%	

In all tests, Module 4 was clearly identified as the most aged module. The differences compared to the other tested modules lead to the conclusion that Module 4 is least suitable for long-term second-life use. The assessment of the other modules with regard to their suitability must be more differentiated, as the different measuring methods produced inconsistent module evaluations. In this context, Module 3 is to be highlighted. While Module 3 often generated the best measurement results of a series of measurements during the EIS and internal resistance measurements, the remaining capacity of Modules 1, 2 and 5 is higher. It should be noted that the deviation from the best module is only approximately 2.3%, and the capacity rating is therefore given a low weighting in this comparison. Module 3 is therefore regarded as the module with the lowest aging of the five tested modules. In addition, it can be seen that EIS provides significantly more sensitive results than the other used methods.

Another inconsistency is the EIS rating of Modules 2 and 5 compared to the internal resistance rating. While EIS measurements of Module 2 showed similarly poor results as Module 4, the internal resistance measurements of Module 2 always appear better than those of Module 5, which was rated better by EIS. In addition, it is noticeable that Modules 2 and 5 are the best modules with regard to capacity rating. An explanation for these deviations is not known, but this observation underlines the thesis that capacity measurements do not allow a generally valid statement about the aging condition of a battery module [7].

The evaluation of the EIS measurements was so far performed with an equal weighting of the parameters Im_1 , Im_2 and R_{02} . In future research, the electrochemical significance of the parameters Im_1 , Im_2 and R_{02} must be weighted so that the significance of the measurement can be increased.

If we look at the initial cost of the measuring devices, we can see that the CI method and the capacity test can be easily integrated by measuring current and voltage. Measuring devices with which EIS can be performed are significantly more expensive in comparison. Considering the necessary test duration, the acquisition costs and the sensitivity, the CI method is the most suitable method.

The CI method has shown that at lower charge states, the internal resistances increase. This is due to a steeper OCV curve so that the voltage jump is larger and was also described by Wildfeuer et al. [13] An increasing module resistance at high SOC could not be observed, which indicates good preservation of the low-contact resistances for the measurements with the CI method. Furthermore, a low internal resistance results in a higher power capability, as more current can flow until the lower discharge voltage is reached due to the voltage drop at the internal resistance.

The results show that the methods are suitable for the quantitative evaluation of the aging states of the modules. However, because only five modules were tested, significantly more modules must be tested to increase the statistical certainty. Further measurements with more modules are planned and also tests with as-new modules so that a statement in comparison with the new condition of the modules can be made. This allows a more quantitative SOH specification and increases the based dataset and statistical outliers can be identified. Additional cycling is important to check how the reversible aging processes may further change the results. Cyclization makes it possible to determine an aging process

so that the condition of delivered modules can be identified more easily. Following this, a more accurate estimation of the remaining usage time should be possible.

It is shown that an individual measuring technology alone provides limited information, but the combination of methods increases the expressiveness. A capacity test cannot properly describe the true aging of the cells, battery modules with a similar aging history can be more difficult to distinguish. Subtle differences can be better detected by using EIS or CI methods. Both methods determine similar module ratings and hence a better gradation of the module aging is possible. Heavily aged modules can be reliably detected with any of the named methods.

It is conceivable that in the future, large single cells will be built and used. This will require improved test equipment for EIS, as the current state of the art (Gamry Inc.) will probably not be precise enough, as the internal resistance of the cells will be very low and the capacity very high. EIS measurement techniques of single cells are basically applicable at the system level, but appropriate measuring equipment for EIS at higher voltages is missing. The electrometer's voltage range is the limiting factor [27].

A further challenge is the low μ -ohm range, as contacting errors of any kind will then have an increasing impact on CI and EIS measurement methods. As a consequence, the results will inevitably become less accurate.

4. Conclusions

In summary, the measurement techniques CI, EIS and the capacity test were experimentally compared to each other to classify their applicability under the aspects of test duration, sensitivity and investment costs of the measuring devices. The applicability of the measurement methods at the module level is given. The measurement technique places high demands on the test setup in order to measure the module properties and not the measurement setup.

The detailed sensitivity analysis shows that heavily aged modules can be detected with all the methods mentioned. Normal aged modules cannot be sufficiently distinguished by a capacity test, and the CI and EIS methods are more sensitive and allow more precise aging determination. In terms of time, CI and EIS analyses are significantly faster to perform than a capacity test.

Considerable differences can also be observed in the cost of the measuring equipment used to perform the measurement methods. Due to their higher sensitivity, the measuring instruments used to perform EIS measurements place high demands on the measuring setup and are therefore usually more expensive.

For integration into SESSs, the CI method is best suited, as current peaks can occur during operation due to the loads. If the current peak is 1C and reference data are available for the SOC, the aging can be determined. This allows in-service aging monitoring so that important reference data can be recorded for the particular battery type, providing predictability for modules of the same type that have not aged as much. Depending on the application, the selection of the most suitable measurement method must be determined on the basis of the investment costs, the desired sensitivity (which, in turn, imposes measurement setup requirements), the number of pieces to be tested and the available test time.

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