Contents lists available at ScienceDirect

Data in Brief

journal homepage: www.elsevier.com/locate/dib



Ageing characterization data of lithium-ion battery with highly deteriorated state and wide range of state-of-health



Zhiyong Xia, Jaber A. Abu Qahouq*

Department of Electrical and Computer Engineering, College of Engineering, University of Alabama, Tuscaloosa, AL 35487 United States

ARTICLE INFO

Article history: Received 4 August 2021 Revised 7 December 2021 Accepted 13 December 2021 Available online 16 December 2021

Keywords: Battery Battery management system Impedance Lithium-ion State-of-health

ABSTRACT

This paper presents ageing characterization data of two lithium-ion battery cells that have been put through a deep ageing process. At the end of ageing process, the values of state-of-health (SOH) of the battery cells drop down to around 15%. The battery cells are aged using a developed autonomous ageing platform which performs functions such as constant current (CC) discharging, CC charging, and constant voltage (CV) charging. Each time the battery cell completes 30 ageing cycles, battery performance tests including dc impedance measurement, minimum impedance measurement, and capacity calibration are conducted to characterize the ageing or health status of the battery cell. The collected battery dc impedance data, minimum impedance data, capacity data, and CC-CV charging time for the deeply aged battery cells are presented in this paper. The presented data has the potential to help in identifying battery ageing behavior patterns. It can also be utilized to investigate the correlation or relationship between different battery ageing characterization data and to develop SOH estimation methods for lithium-ion batteries with high degradation conditions, for examples, for second-use battery and when battery health exhibits unexpected faster deterioration.

* Corresponding author.

E-mail address: jaberq@eng.ua.edu (J.A. Abu Qahouq).

https://doi.org/10.1016/j.dib.2021.107727

2352-3409/© 2021 The Author(s). Published by Elsevier Inc. This is an open access article under the CC BY license (http://creativecommons.org/licenses/by/4.0/)



© 2021 The Author(s). Published by Elsevier Inc. This is an open access article under the CC BY license (http://creativecommons.org/licenses/by/4.0/)

Nomenclature							
A. Acrony CC CV SOC SOH	vms constant current constant voltage state-of-charge state-of-health						
B. Symbols							
I _{Batt} Z _{dc}	battery electric current in Ampere (A) unit battery impedance magnitude when the phase of complex impedance is equal to zero in Ohm (Ω) unit						
Z_{\min}	battery minimum complex impedance magnitude in Ohm (Ω) unit						
Q_{avail}	a load (discharge) in Coulomb (C) unit						
Q _{nominal}	nominal capacity of battery (i.e., the amount of charges the battery can supply to a load or discharge when it is new or not aged) in Coulomb (C) unit						
Qcount	the amount of charges the battery discharges during a given time period from t_0						
to	to t_1 in Coulomb (C) unit (Battery Coulomb counting result) time instant t_0 in seconds unit						
t_0 t_1	time instant t_0 in seconds unit						
Δt	sampling time of battery electric current in seconds unit						

Subject	Electrical and Electronic Engineering
Specific subject area	Battery ageing characterization, battery impedance, and battery capacity
Type of data	Table
	Figure
How data were acquired	The battery dc impedance and battery minimum impedance are measured using Gamry Interface 5000E [1]. The battery capacity data and CC–CV charging time are measured and recorded using autonomous battery ageing platform which is described in [2].
Data format	Raw and analyzed
Parameters for data collection	Each time the battery cell completes 30 ageing cycles, the battery performance tests are conducted and CC–CV charging time is recorded. The dc impedance and minimum impedance values are measured at 100% state-of-charge (SOC), 60% SOC, and 20% SOC. The battery capacity is calibrated or calculated during the last ageing cycle of every 30 ageing cycles using Coulomb counting under CC discharging operation.
Description of data collection	CC-CV charging time and capacity data are collected using the developed
	battery autonomous ageing system. Battery dc impedance and minimum impedance values are recorded using Gamry Interface 5000E [1].
Data source location	Institution: The University of Alabama
	City/Town/Region: Tuscaloosa/Alabama
	Country: USA
	Latitude and longitude (and GPS coordinates, if possible) for collected samples/data: 33.21565034040115, -87.54360419954291
Data accessibility	Data is supplied with the article

Specifications Table

Value of the Data

- The presented data helps to characterize the ageing behavior patterns of lithium-ion battery over wide range of degradation levels down to around 15% state-of-health (deeply aged with high degradation condition). The ageing characterization data of highly deteriorated lithium-ion battery is not publicly available and therefore the presented data fills a currently existing gap in the publicly available data.
- Researchers who have interests in battery management system and battery health estimation can utilize the provided data in this paper to further investigate battery ageing characterization and the development of SOH estimation methods especially for wide range of health conditions.
- The presented data can be further utilized to investigate the correlation or relationship between different battery ageing characterization data, to develop SOH estimation methods for lithium-ion batteries with high degradation conditions, for examples, for second-use battery and when battery health exhibits unexpected faster deterioration, and to help understand the ageing evolving trend of battery.

1. Data Description

The ageing characterization data for two lithium-ion battery cells is collected as the battery health (or capacity) decreases down to around 15%. The collected battery ageing characterization data at different ageing cycles is listed in Tables 1 and 2. The battery ageing characterization data includes battery dc impedance (Z_{dc}), battery minimum impedance (Z_{min}), CC charging time, CV charging time, total charging time (CC charging time + CV charging time), and SOH values. Fig. 1 illustrates the definition of Z_{dc} and Z_{min} . Z_{dc} refers to the magnitude of battery impedance when the phase of battery complex impedance is equal to zero. Z_{min} refers to the minimum magnitude value of the battery complex impedance within the frequency range from 0.01 Hz to 10 kHz. Over the ageing process, Z_{dc} and Z_{min} values are measured and recorded at different SOC conditions, i.e., 100% SOC, 60% SOC, and 20% SOC. The SOH values are calculated using Coulomb counting method, which is discussed in the next section of this paper (i.e., Experimental Design, Materials and Methods section). At the end of ageing process, i.e., when the ageing cycle is 630 in this paper, the battery cells are highly deteriorated with SOH value of 15.52% for battery #1 and SOH



Fig. 1. The illustration for the definition of Z_{dc} and Z_{min} .

	7 @10.0%	7 @60%	7 @20%	7 @10.0%	7 @60%	7 @20%	CC charge	CV charge	Total charge	
Cycle	$\Sigma_{dc} @ 100\%$ SOC (Ω)	$\Sigma_{dc} @ 00\%$ SOC (Ω)	$\Sigma_{dc} @ 20\%$ SOC (Ω)	$\Sigma_{\min} \otimes 100\%$ SOC (Ω)	$\Sigma_{\min} @ 00\%$ SOC (Ω)	$\Sigma_{\rm min} @ 20\%$ SOC (Ω)	(minutes)	(minutes)	(minutes)	SOH (%)
1	0.069733	0.06959	0.069624	0.068831	0.068709	0.068771	31.95	126.51	158.46	94.94
30	0.069729	0.069635	0.069631	0.068835	0.068751	0.069042	30.79	114.85	145.64	93.34
60	0.06988	0.069842	0.069713	0.068996	0.06894	0.069611	23.86	108.49	132.35	89.73
90	0.070795	0.070927	0.070634	0.069904	0.069983	0.070921	23.39	102.19	125.59	80.92
120	0.071607	0.071849	0.071552	0.070693	0.070882	0.071687	22.14	102.02	124.17	76.19
150	0.072111	0.072414	0.072137	0.071379	0.071436	0.072125	20.64	94.35	115.00	73.76
180	0.072617	0.072979	0.072734	0.071667	0.071991	0.072557	17.44	92.39	109.83	71.56
210	0.073279	0.073714	0.073523	0.072302	0.072713	0.073121	16.42	89.14	105.57	68.96
240	0.073828	0.074322	0.074183	0.073288	0.073311	0.073595	16.31	91.31	107.62	66.98
270	0.074544	0.075111	0.07505	0.073514	0.074089	0.074227	15.89	91.78	107.68	64.59
300	0.075668	0.076348	0.076423	0.07459	0.07531	0.075259	15.02	92.72	107.74	61.17
330	0.076019	0.077186	0.077714	0.075016	0.076133	0.076362	12.93	93.68	106.62	59.06
360	0.078239	0.078852	0.079071	0.076955	0.077454	0.077692	11.33	94.69	106.02	55.17
390	0.079037	0.081234	0.080756	0.077679	0.079887	0.079546	10.13	95.46	105.59	51.58
420	0.080462	0.080538	0.081688	0.079031	0.079491	0.080491	8.84	94.30	103.15	49.49
450	0.081474	0.082559	0.083034	0.080301	0.081296	0.081685	7.42	95.43	102.86	47.30
480	0.081558	0.083762	0.08467	0.080373	0.08254	0.083264	7.19	94.21	101.41	46.32
510	0.083566	0.083551	0.084077	0.082189	0.081972	0.082793	7.01	95.61	102.62	46.14
540	0.083372	0.085543	0.087031	0.081787	0.084198	0.085454	6.64	88.32	94.96	41.99
570	0.088103	0.091988	0.090913	0.086577	0.090521	0.089566	3.03	83.22	86.25	34.98
600	0.089273	0.095729	0.094826	0.087488	0.093726	0.092885	2.12	81.68	83.80	28.46
630	0.103558	0.102825	0.104732	0.101762	0.100861	0.102449	0.61	67.31	67.92	15.52

Table 1Battery #1 ageing characterization data at different ageing cycles.

	Z _{dc} @100%	Z _{dc} @60%	Z _{dc} @20%	Z _{min} @100%	Z _{min} @60%	Z _{min} @20%	CC charge time	CV charge time	Total charge time	
Cycle	SOC (Ω)	SOC (Ω)	SOC (Ω)	SOC (Ω)	SOC (Ω)	SOC (Ω)	(minutes)	(minutes)	(minutes)	SOH (%)
1	0.066665	0.067224	0.067258	0.065802	0.066373	0.066003	32.90	132.77	165.68	93.79
30	0.067497	0.067267	0.067523	0.066632	0.066513	0.066011	31.09	115.66	146.75	89.73
60	0.067644	0.067781	0.06808	0.066788	0.066907	0.066088	25.13	111.00	136.13	84.82
90	0.06853	0.068834	0.068509	0.067667	0.067918	0.068233	22.60	109.06	131.67	75.39
120	0.069817	0.069729	0.069608	0.068925	0.068791	0.069477	22.35	99.24	121.59	70.33
150	0.070308	0.071111	0.070045	0.069594	0.07015	0.070033	20.86	98.36	119.22	69.35
180	0.070802	0.071666	0.070625	0.069875	0.070695	0.070453	17.76	98.27	116.03	66.27
210	0.071447	0.072387	0.071391	0.070495	0.071404	0.071001	17.36	97.80	115.16	65.13
240	0.072794	0.072984	0.072328	0.072262	0.071991	0.071755	15.77	96.46	112.23	64.72
270	0.0735	0.07406	0.073173	0.072485	0.073052	0.072371	14.64	96.33	110.97	57.95
300	0.074609	0.075279	0.074894	0.073546	0.074255	0.073754	14.17	94.94	109.12	57.64
330	0.074954	0.076106	0.07616	0.073965	0.075067	0.074835	13.22	94.20	107.42	52.60
360	0.077144	0.077748	0.077489	0.075878	0.076369	0.076138	11.40	89.65	101.06	52.31
390	0.078247	0.080146	0.079141	0.076902	0.078816	0.077955	9.60	89.48	99.08	48.85
420	0.079657	0.079459	0.080218	0.07824	0.078426	0.079042	8.76	89.13	97.89	48.76
450	0.08066	0.081452	0.081539	0.079498	0.080206	0.080215	8.37	88.55	96.93	45.17
480	0.080334	0.08264	0.083146	0.079168	0.081434	0.081765	7.15	88.41	95.56	40.97
510	0.082312	0.082632	0.082984	0.080956	0.08107	0.080717	5.54	86.81	92.36	40.27
540	0.082122	0.084602	0.085899	0.08056	0.083272	0.084343	5.00	85.41	90.41	39.58
570	0.086781	0.090976	0.089731	0.085278	0.089525	0.088402	4.62	83.25	87.88	28.69
600	0.087934	0.094675	0.093878	0.086175	0.092695	0.091956	2.14	79.05	81.20	27.75
630	0.102005	0.10118	0.103495	0.100235	0.099248	0.102327	0.68	64.33	65.02	12.40

Table 2 Battery #2 ageing characterization data at different ageing cycles.



Fig. 2. The distribution of Z_{dc} and charging time over the ageing process for battery #1: (a) Z_{dc} and CC charging time, (b) Z_{dc} and CV charging time, and (c) Z_{dc} and total charging time.

value of 12.4% for battery #2. Compared with common publicly available battery ageing datasets where battery cells are aged only up to approximately 70%–80% SOH [3–7], the battery ageing characterization dataset in this paper covers wide range of SOH down to around 15%. Therefore, the dataset presented in this paper can be utilized to investigate the correlation or relationship between different battery ageing characterization data and to develop SOH estimation methods for lithium-ion batteries that are deeply aged with high degradation conditions, for examples, for second-use and when battery health exhibits unexpected faster deterioration [8].

Figs. 2–5 show the distribution of battery impedance (Z_{dc} and Z_{min}) and battery charging time (CC charging time, CV charging time, and total charging time) over the ageing process. These figures can be utilized to identify or investigate the correlation between battery impedance and battery charging time over the ageing process.

In order to develop SOH estimators for the battery, the first step is to identify the features or indicators which can reflect the health status of the battery or have relevant correlation with SOH values. For this purpose, Figs. 6 and 7 illustrate the evolution trend of the battery impedance (Z_{dc} and Z_{min}) and SOH values, and Figs. 8 and 9 illustrate the evolution trend of the battery charging time (CC charging time, CV charging time, and total charging time) and SOH values.

Battery cells from different manufacturers may exhibit different battery ageing characteristics [9]. One of the reasons for this is because of the different chemistry compositions of different battery cells. However, the ageing effects on the presented parameters are expected to be same



Fig. 3. The distribution of Z_{dc} and charging time over the ageing process for battery #2: (a) Z_{dc} and CC charging time, (b) Z_{dc} and CV charging time, and (c) Z_{dc} and total charging time.

for the same battery type. The data presented in this paper is mainly used to reflect the ageing characteristics of the selected battery type from one manufacturer.

2. Experimental Design, Materials and Methods

The model of the utilized lithium-ion battery cell in the conducted ageing experiment is Tenergy ICR 16,650–2600 [10]. The main manufacturer datasheet parameters of interest of this type of battery include: nominal capacity is 2.6 Ah, initial impedance at 1 kHz is less than 65 m Ω , and average working voltage is 3.7 V [10].

The definition of SOH which is used in this paper is based on the concept of capacity fading [7], and it can be calculated using Eq. (1).

$$SOH = \frac{Q_{avail}}{Q_{nominal}} = 1 - \frac{Q_{nominal} - Q_{avail}}{Q_{nominal}} = 1 - capacity \ fading \tag{1}$$

Where Q_{avail} is the actual available capacity or total amount of charges the battery can supply to a load (discharge), which can be obtained using Coulomb counting method, and $Q_{nominal}$ is the nominal capacity of battery, which can be obtained from the documents of the manufacturer. Coulomb counting method is used to count the amount of charges the battery cell discharges



Fig. 4. The distribution of Z_{min} and charging time over ageing process for battery #1: (a) Z_{min} and CC charging time, (b) Z_{min} and CV charging time, and (c) Z_{min} and total charging time.

based on Eq. (2) [11].

$$Q_{count} = \int_{t_0}^{t_1} I_{Batt} dt = \sum_{t_0}^{t_1} (I_{Batt} \times \Delta t)$$

$$\tag{2}$$

Where Q_{count} is the amount of charges the battery cell discharges during time period from t_0 to t_1 , i.e., the Coulomb counting result, I_{Batt} is the battery electric current measured value during the time period Δt , and Δt is the sampling time (10 µs in this paper) of battery electric current. Capacity fading is another variable used to indicate the degradation level of battery cell, and it can be calculated based on Eq. (3). The range of the value of capacity fading is from 0 to 1 (or 0% to 100%). The larger the value of capacity fading is, the more deteriorated the health of the battery is (i.e., the lower SOH is).

$$capacity \ fading = \frac{Q_{nominal} - Q_{avail}}{Q_{nominal}} \tag{3}$$

The range of SOH value is from 0 to 1 (or 0% to 100%). The smaller the value of SOH is, the more deteriorated the battery is.

As an important performance characterization parameter of battery, battery impedance can be utilized to study and monitor the battery operation status and to estimate other internal state variables of the battery (for examples SOC, SOH, and battery internal temperature) [12–15]. The

Fig. 5. The distribution of Z_{min} and charging time over ageing process for battery #2: (a) Z_{min} and CC charging time, (b) Z_{min} and CV charging time, and (c) Z_{min} and total charging time.

values of Z_{dc} and Z_{min} are two important characteristic parameters of the battery impedance [7,12–15], and they are utilized as two ageing characterization parameters in this paper. During the ageing process of the lithium-ion battery in this paper, Z_{dc} and Z_{min} are measured. Battery charging time is also utilized to reflect the operation or health status of the battery [4–6]. Therefore, battery charging time is also measured in the conducted ageing experiment in this paper to characterize the ageing or performance of battery cell. The collected data for the battery charging time in this paper includes CC charging time, CV charging time, and total charging time.

Fig. 10 shows the utilized ageing and performance testing procedure used to obtain the presented data in this paper. Before conducting the ageing experiment, the initial battery capacity is calibrated using Coulomb counting method.

After the initial capacity calibration, the battery cell is subjected to 30 ageing cycles using the developed autonomous ageing platform. The details about this ageing platform can be found in [2]. During each ageing cycle, the battery cell goes through one full CC discharging operation and one complete CC–CV charging operation. The CC discharging current is set to be 1 C (i.e., 2.6 A) during the ageing process. The discharging operation is terminated when the battery voltage drops down to 2.7 V. The CC charging current is also set to be 1 C (i.e., 2.6 A) during the ageing operation is terminated when the battery voltage ing process. The CC charging operation is initiated to charge reaches 4.2 V. After the CC charging operation, CV charging operation is initiated to charge the battery until the battery current decreases down to 0.05 A (which is the condition to end the CV charging

Fig. 6. The distribution of battery impedance value and SOH values over the ageing process for battery #1: (a) Z_{dc} and SOH values and (b) Z_{min} and SOH values.

Fig. 7. The distribution of battery impedance value and SOH values over the ageing process for battery #2: (a) Z_{dc} and SOH values and (b) Z_{min} and SOH values.

operation). For every 30 ageing cycles, CC charging time, CV charging time, and total charging time for the last ageing cycle are recorded by the developed ageing platform.

After completing 30 ageing cycles, the battery cell is put in rest state for 12 hours in order to achieve thermal equilibrium before calibrating the available capacity of the battery. This can help to obtain accurate capacity calibration results. After the capacity calibration, the battery is fully charged again to 100% SOC using CC–CV charging method and then the battery cell is put in rest state again for 12 hours. In the next step, Gamry interface 5000E [1] equipment is used to measure Z_{dc} and Z_{min} at 100% SOC, 60% SOC, and 20% SOC. CC discharging is utilized to discharge the battery cell from one SOC value to another SOC value, e.g., from 100% SOC to 60% SOC. After each discharging operation and before measuring battery impedance at one specific SOC value, the battery cell is always put in rest state for 12 hours to reach the thermal equilibrium state. The ageing process and battery performance tests are conducted until the SOH value of battery cell decreases to around 15%.

Fig. 8. The distribution of battery charging time and SOH values over the ageing process for battery #1: (a) CC charging time and SOH values, (b) CV charging time and SOH values, and (c) total charging time and SOH values.

Fig. 9. The distribution of battery charging time and SOH values over the ageing process for battery #2: (a) CC charging time and SOH values, (b) CV charging time and SOH values, and (c) total charging time and SOH values.

Fig. 10. The flowchart of battery ageing and performance testing procedure utilized to measure and collect the presented data in this paper.

Declaration of Competing Interest

The authors declare that they have no known competing financial interests or personal relationships which have or could be perceived to have influenced the work reported in this article.

CRediT Author Statement

Zhiyong Xia: Investigation, Methodology, Data curation, Visualization, Writing – original draft; **Jaber A. Abu Qahouq:** Funding acquisition, Project administration, Investigation, Methodology, Visualization, Supervision, Writing – review & editing.

Acknowledgment

This material is based upon work supported in part by the National Science Foundation under Grant No. 1509824. Any opinions, findings and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the National Science Foundation.

References

- Gamry Instruments Inc., Warminster, PA, Gamry interface 5000E user manual, https://www.gamry.com/assets/ Uploads/Interface5000-Users-Manual.pdf. Accessed July 18, 2021.
- [2] Z. Xia, J.A. Abu Qahouq, E. Phillips, R. Gentry, A simple and upgradable autonomous battery aging evaluation and test system with capacity fading and AC impedance spectroscopy measurement, IEEE App. Power Electron. Conf. Expos. (2017) 951–958, doi:10.1109/APEC.2017.7930811.
- B. Saha, K. Goebel, Battery Data Set, NASA Ames Prognostics Data Repository, 2021 https://ti.arc.nasa.gov/tech/dash/ groups/pcoe/prognostic-data-repository/. Accessed July 18, 2021.
- [4] Z. Xia, J.A. Abu Qahouq, Lithium-ion battery ageing behavior pattern characterization and state-of-health estimation using data-driven method, IEEE Access 9 (2021) 98287–98304, doi:10.1109/ACCESS.2021.3092743.
- [5] J. Yang, B. Xia, W. Huang, Y. Fu, C. Mi, Online state-of-health estimation for lithium-ion batteries using constantvoltage charging current analysis, Appl. Energy 212 (2018) 1589–1600, doi:10.1016/j.apenergy.2018.01.010.
- [6] Y. Wu, Q. Xue, J. Shen, Z. Lei, Z. Chen, Y. Liu, State of health estimation for lithium-ion batteries based on healthy features and long short-term memory, IEEE Access 8 (2020) 28533–28547, doi:10.1109/access.2020.2972344.
- [7] Z. Xia, J.A. Abu Qahouq, State-of-charge balancing of lithium-ion batteries with state-of-health awareness capability, IEEE Trans. Ind. Appl. 57 (1) (2020) 673–684, doi:10.1109/tia.2020.3029755.
- [8] A. Podias, A. Pfrang, F.D. Persio, A. Kriston, S. Bobba, F. Mathieux, M. Messagie, L. Boon-Brett, Sustainability assessment of second use applications of automotive batteries: ageing of Li-ion battery cells in automotive and grid-scale applications, World Electr. Veh. J. 9 (2) (2018) 1–15, doi:10.3390/wevj9020024.
- [9] F. Yang, D. Wang, Y. Zhao, K. Tsui, S. Bae, A study of the relationship between coulombic efficiency and capacity degradation of commercial lithium-ion batteries, Energy 145 (2018) 486–495, doi:10.1016/j.energy.2017.12.144.
- [10] [datasheet] Tenergy Corp., Fremont, CA, Tenergy cylindric lithium-ion cell 30005-0 datasheet, https://www. shorepowerinc.com/media/wysiwyg/files/Tenergy-30005-0-datasheet.pdf. Accessed July 18, 2021.
- [11] G. Fathoni, S. Widayat, P. Topan, A. Jalil, A. Cahyadi, O. Wahyunggoro, Comparison of state-of-charge (SOC) estimation performance based on three popular methods: coulomb counting, open circuit voltage, and Kalman filter, in: Proceedings of the International Conference on Automation, Cognitive Science, Optics, Micro Electro-Mechanical System, and Information Technology, 2017, pp. 70–74, doi:10.1109/ICACOMIT.2017.8253389.
- [12] D. Stroe, M. Swierczynski, A. Stroe, S.K. Kaer, R. Teodorescu, Lithium-ion battery power degradation modeling by electrochemical impedance spectroscopy, IET Renew. Power Gener. 11 (9) (2017) 1136–1141, doi:10.1049/iet-rpg. 2016.0958.
- [13] X. Wang, X. Wei, J. Zhu, H. Dai, Y. Zheng, X. Xu, Q. Chen, A review of modeling, acquisition, and application of lithium-ion battery impedance for onboard battery management, ETransportation (2020) 100093, doi:10.1016/ j.etran.2020.100093.
- [14] A.A. Hussein, A.A. Fardoun, S.S. Stephen, An ultrafast maximum power point tracking technique for optimal battery charging, IEEE Trans. Sustain. Energy 8 (3) (2017) 1321–1329, doi:10.1109/TSTE.2017.2679161.
- [15] J. Zhu, Z. Sun, X. Wei, H. Dai, A new lithium-ion battery internal temperature on-line estimate method based on electrochemical impedance spectroscopy measurement, J. Power Sources 274 (2015) 990–1004, doi:10.1016/j. jpowsour.2014.10.182.