

A Generic Battery Model and Its Parameter Identification

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Abstract

A new dynamic model is developed in this paper based on the generic MATLAB battery model. The battery capacity is expressed as a function of the self-discharge rate, the discharge current, the cycling life and the temperature of the battery. The dependence of the model parameters on cycle life and temperature are estimated from the first order approximation. The detailed procedures and formula to extract the model parameters are presented and the extraction relies only on the discharge curves at two different discharge currents, at two different life cycles, and at two different temperatures. These discharge curves are typically provided in the battery manufacturer's datasheet. The proposed model is verified for both nickel-metal hydride and lithium-ion batteries by comparing the calculated discharge curves with the results from the generic MATLAB model. The model is further validated for the Sinopoly lithium-ion battery (SP-LFP1000AHA) by comparing the model results with the discharge curves from the manufacturer's datasheet at different discharge currents, different cycling numbers, and different temperatures. Simulation results show that the new model can correctly predict voltage separation beyond the nominal zone while maintaining the same level of accuracy as the generic MATLAB model in the exponential and nominal zones.

Keywords

Battery Model, Parameter Identification, Self-Discharge,
Capacity Fade/Degradation, Cycling Life

1. Introduction

Battery-based energy storage systems (ESS) are playing an increasingly important role in modern electrical grids as they can serve a variety of functions in the

generation, transmission, and distribution of electric energy, especially in the case of the high penetration of intermittent renewable energy from wind and solar power generation [1]. When an ESS relies on an electrochemical storage technology like batteries, an accurate model of its voltage and capacity is essential to evaluating the system's ability to perform various function required for grid applications. The ESS controller must rely on this model in order to safely and reliably operate the ESS. The existing mathematical battery models vary widely in terms of chemistries, reactions, and geometries modelled. In general, as more detailed physiochemical phenomena are included in a battery model, its simulation results can be expected to be more reliable, but at the cost of increased computational effort. Therefore, an accurate yet relatively simple dynamic mathematical model is required for the battery management systems (BMS) to facilitate efficient ESS control in grid applications so that the charging/discharging processes can be properly managed to optimize battery performance and durability.

According to their modeling methodology, battery models can be classified as Electrochemical Multiphysics, Equivalent Circuit (EC), and Empirical or semi-empirical dynamic models [2]. Electrochemical Multiphysics battery models [3] consider the fundamental battery chemistry and multiphase transport theory to study the effects of the reaction kinetics and geometry on battery performance. The Electrochemical Multiphysics models are well developed but are not suitable for grid applications due to their complicated coupling of species reaction and transport equations. The equivalent circuit battery models [4] employ the passive circuit elements like capacitors, inductors and resistors to model electrolyte, electrode, polarization resistances along with active current and voltage sources. To improve EC model accuracy, additional resistor-capacitor (RC) networks can be integrated but at the cost of computational efficiency. The EC model with more RC networks also requires more detailed experimental data to determine its model parameters. Empirical models employ elementary mathematical functions such as polynomials, exponentials, power law, etc., to predict the battery performance. Some empirical models, also called OCV models [5], are just fitted mathematic functions for a specific battery without any physical meaning and therefore have limited popularity. Other empirical models, called generic models, are derived either from the Nernst equation [6] [7] or from the Butler-Volmer equation [8] and are now widely employed in various applications due to its simplicity and relatively high accuracy. The model parameter identification of the dynamic model developed in [8] [9] [10] is based on a batch least square approach that requires tedious testing and these data sets are not typically available or easy to obtain, especially for grid-scale battery systems. The identified model parameters vary with cycle life and no explicit expressions are presented for the parameters as functions of the battery's cycle life which make the model difficult to use for control purposes in grid applications. The newly released MATLAB software version R2017b [11] [12] includes a generic lithium-ion battery model that incor-

porates the temperature and aging functions and these effects on battery performance can be evaluated individually. This generic MATLAB battery model has been modified by many researchers to include some missing features, such as the self-discharge and internal resistance variation [13], nonlinear temperature dependence [14], and is commonly applied to the dynamic analysis for electric vehicles [12] [15]. The most advantageous feature of the generic MATLAB battery model is that the model parameters are estimated from a manufacturer's datasheet and only three points on one discharge curve are needed, namely the initial fully charged point (100% state of charge), the point at the end of exponential zone and the point at the end of nominal zone [7]. MATLAB's generic model did not consider the variation of the battery capacity as a function of its self-discharge, discharge current, cycle life and temperature so it is unable to predict the voltage behavior near the end of the discharge process.

As a lithium-ion battery ages, its voltage and capacity decreases or fades and its internal resistance increases due to the loss of active lithium and the thickness of the solid electrolyte interface (SEI) [16] [17] and these variations should be carefully considered when developing a the dynamic battery model. This paper describes the advancements made to the generic MATLAB battery model by introducing an additional parameter together with a variable capacity which is a function of the self-discharge rate, the discharge current, the cycle life and operating temperature. The effects of temperature and cycle life on battery performance are estimated using first-order approximations. An explicit and systematic approach is presented to determine the model parameters. More importantly, the identification of model parameters is based only on information typically provided in manufacturer's datasheets and no tedious battery testing is required.

2. The Modified MATLAB Model

In order to improve the accuracy of the generic MATLAB battery model beyond the nominal zone and to consider the capacity degradation, the equation of the generic MATLAB battery model [7] is modified as,

$$V_{\text{batt}}(t) = V_0 - Ri - K \frac{mQ(i)}{mQ(i) - it} + A \exp(-Bit) \quad (1)$$

and

$$Q(i) = \left(\frac{i}{i_0} \right)^\alpha Q_0 \quad (2)$$

where $V_{\text{batt}}(t)$ is the battery voltage (V), t is charging or discharging time (h), V_0 is a voltage constant (V), R is internal resistance of the battery (Ω), i is battery current (A) where positive current represents discharging the battery and negative current indicates charging the battery, K is a polarization constant (V/(Ah)), m is a capacity adjustment factor that accounts for capacity degradation, $Q(i)$ is battery capacity (Ah) at the specific discharge current i and

point in the battery's life being modeled, A is exponential zone voltage constant (V), and B is exponential zone inverse time constant (1/(Ah)), and Q_0 is the nominal or rated battery capacity at the beginning of life (*i.e.* $n = 1$) at a reference discharge current of i_0 .

The parameter α in Equation (2) can be determined by two battery capacities (Q_{i_1}, Q_{i_2}) at two different discharge currents i_1 and i_2 under normal temperature (typically 25°C) and near the beginning of life (*e.g.* $n < 5$), as shown in **Figure 1**.

$$\alpha = \frac{\log(Q_{i_2}/Q_{i_1})}{\log(i_2/i_1)} \quad (3)$$

There are a total of 7 unknowns in Equation (1): V_0 , R , K , A , B , Q_0 and m . The method described in [7] can be applied here to estimate the first 6 parameters once the needed three points on a discharge curve are accurately identified. The initial 100% state of charge point is at the beginning of the discharge process ($t = 0$). The other two points, one at the end of exponential zone and one at the end of nominal zone, are challenging to distinguish and need to be finely adjusted in order to obtain a good fit. The supporting documentation [11] for MATLAB's generic battery model describes that the capacity adjustment parameter (m) is set at a default value of 0.1 and no detailed procedure is given on how to determine its value. In addition, the determination of the internal resistance relies on the cell efficiency which is not typically provided on manufacturer's datasheets. In this paper, a new procedure is introduced to determine all model parameters based only on the specifications typically provided on manufacturer's datasheets. As compared to the MATLAB method, the new procedure proposed is more tolerant of the positions of the required points on discharge curves. In general, the nominal capacity (Q_0) of the battery can be found directly from manufacturer's datasheets. The following section will focus on how to determine the other 6 unknowns. **Figure 1** plots two typical discharge curves

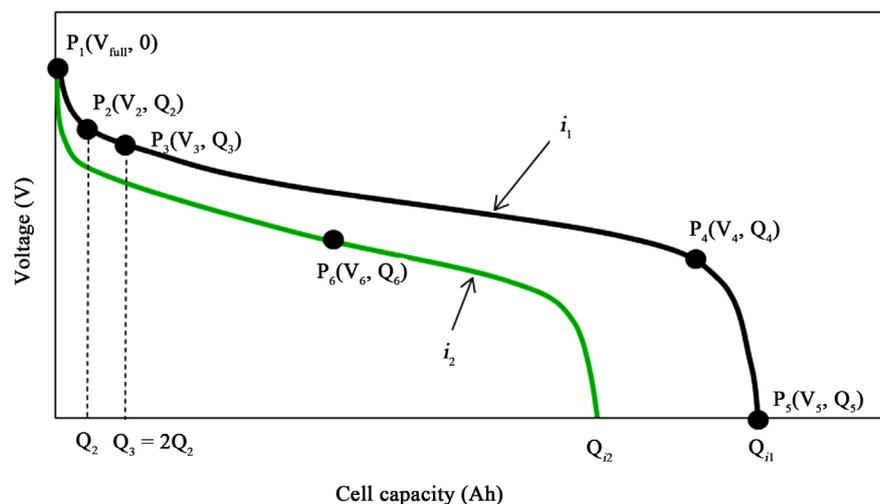


Figure 1. Typical discharge curves at two discharge currents, i_1 and i_2 .

at two different discharge currents that are commonly included in manufacturer's datasheets.

The first step is to determine the parameters A, B, K and m by using five points on the discharge curve at $i = i_1 (i_1 > i_2)$: $P_1(V_{\text{full}}, 0)$, the initial fully charged point; $P_2(V_2, Q_2)$ and $P_3(V_3, Q_3)$ are within the exponential zone and satisfy the condition of $Q_3 = 2Q_2$; $P_4(V_4, Q_4)$, one point near the end point of nominal zone; and $P_5(V_5, Q_5)$, the end point of discharge. Five equations are obtained by applying Equation (1) to these five points accordingly.

$$V_{\text{full}} = V_0 - Ri_1 - K + A \quad (4)$$

$$V_j = V_0 - Ri_1 - K \frac{mQ_i}{mQ_i - Q_j} + A \exp(-BQ_j), \quad j = 2, 3, 4, 5 \quad (5)$$

Equation (4) minus Equation (5) to eliminate the term $(V_0 - Ri_1)$,

$$V_{\text{full}} - V_2 = K \left[\frac{mQ_i}{mQ_i - Q_2} - 1 \right] + A [1 - \exp(-BQ_2)] \quad (6)$$

$$V_{\text{full}} - V_3 = K \left[\frac{mQ_i}{mQ_i - Q_3} - 1 \right] + A [1 - \exp(-BQ_3)] \quad (7)$$

$$V_{\text{full}} - V_4 = K \left[\frac{mQ_i}{mQ_i - Q_4} - 1 \right] + A [1 - \exp(-BQ_4)] \quad (8)$$

$$V_{\text{full}} - V_5 = K \left[\frac{mQ_i}{mQ_i - Q_5} - 1 \right] + A [1 - \exp(-BQ_5)] \quad (9)$$

Assuming $Q_2 \ll Q_i$, $Q_3 \ll Q_i$ and noticing the relation of $Q_3 = 2Q_2$, Equations (4) and (7) reduce to

$$V_{\text{full}} - V_2 \approx A [1 - \exp(-BQ_2)] \quad (10)$$

and

$$V_{\text{full}} - V_3 \approx A [1 - \exp(-BQ_3)] = A [1 - (\exp(-BQ_2))^2] \quad (11)$$

Parameters A and B can be solved from Equations (10) and (11) as

$$B = -\frac{1}{Q_2} \ln \left(\frac{V_{\text{full}} - V_3}{V_{\text{full}} - V_2} - 1 \right) \quad (12)$$

$$A = \frac{V_{\text{full}} - V_3}{1 - \exp(-BQ_3)} \quad (13)$$

Let

$$AB_4 = V_{\text{full}} - V_4 - A [1 - \exp(-BQ_4)] \quad (14)$$

$$AB_5 = V_{\text{full}} - V_5 - A [1 - \exp(-BQ_5)] \quad (15)$$

Equations (8) and (9) becomes

$$K \left[\frac{mQ_i}{mQ_i - Q_4} - 1 \right] = AB_4 \quad (16)$$

and

$$K \left[\frac{mQ_i}{mQ_i - Q_5} - 1 \right] = AB_5 \quad (17)$$

So the parameters m and K can be obtained from Equations (16) and (17) as

$$m = \frac{1 - \frac{AB_5}{AB_4}}{1 - \frac{AB_5 Q_4}{AB_4 Q_5}} \times \frac{Q_4}{Q_i} \quad (18)$$

$$K = AB_5 \left[m \frac{Q_i}{Q_5} - 1 \right] \quad (19)$$

The second step is to determine parameters R and V_0 . In some cases, the battery's internal resistance and efficiency are not provided in manufacturer's datasheets so the MATLAB method to determine R given in [7] is not applicable. Here a new procedure is proposed to calculate the internal resistance by using two points on two different discharge curves at two different discharge currents.

Applying the cell voltage Equation (1) to point P_6 on the discharge curve of current i_2 ,

$$V_6 = V_0 - Ri_2 - K \frac{mQ_{i_2}}{mQ_{i_2} - Q_6} + A \left[1 - \exp(-BQ_6) \right] \quad (20)$$

Subtracting Equation (5) (for $j = 3$) from Equation (20), and solving for the internal resistance R results in the following equation,

$$R = \frac{1}{i_1 - i_2} \left\{ V_6 + K \frac{mQ_{i_2}}{mQ_{i_2} - Q_6} - A \exp(-BQ_6) - \left[V_3 + K \frac{mQ_{i_1}}{mQ_{i_1} - Q_3} - A \exp(-BQ_3) \right] \right\} \quad (21)$$

Once the values of parameters, B, A, m, K and R are available, the constant voltage, V_0 can be obtained from Equation (4),

$$V_0 = V_{\text{full}} + Ri_1 + K - A \quad (22)$$

3. Model Equations with Aging and Temperature Effects

To include both aging and temperature effects, the model Equation (1) is rewritten as,

$$V_{\text{batt}}(t, i, n, t_r(n), T) = V(n, T) - R(n, T)i - K(n, T) \frac{mQ(i, n, t_r(n), T)}{mQ(i, n, t_r(n), T) - it} + A(n, T) \exp(-Bit) \quad (23)$$

where n is n -th discharge-charge cycle starting from discharging a cell to a fully discharged state and then charging the cell to a fully charged state. $t_r(n)$

is the n -th resting time of the cell after the $(n - 1)$ -th cycle. T is the cell temperature. Here the capacity adjustment (m) and exponential zone time constant (B) are assumed unchanged.

For simplicity, both the cycling- and temperature-dependences of the parameters in Equation (23) are estimated by the first-order approximation,

$$V(n, T) = V_0 (1 + k_{vn1} (n - 1)) \times (1 + k_{vT1} (T - T_{ref})) \quad (24)$$

$$R(n, T) = R_0 (1 + k_{rn1} (n - 1)) \times (1 + k_{rT1} (T - T_{ref})) \quad (25)$$

$$K(n, T) = K_0 (1 + k_{kn1} (n - 1)) \times (1 + k_{kT1} (T - T_{ref})) \quad (26)$$

$$A(n, T) = A_0 (1 + k_{an1} (n - 1)) \times (1 + k_{aT1} (T - T_{ref})) \quad (27)$$

where k_{xn1} refers to the first-order degradation rate constant of parameter x related to cycling life; k_{xT1} is the degradation rate constant of parameter x related to temperature; T_{ref} is a reference temperature.

3.1. Battery Capacity Degradation Considering Self-Discharge, Cycling Life and Temperature

A battery's capacity Q degrades not only due to the cycling life and temperature, but also as a function of the resting time between two consecutive discharge-charge cycles [18]. Furthermore, it also varies with the corresponding discharge current. These degradations must first be identified before calculating the degradation rate constants appearing in Equations (24) to (27). Assuming the self-discharge rate constant doesn't depend on cycling life, then $Q(n, t_r(n))$, the battery capacity at beginning of the n -th cycle after resting $t_r(n)$ hours, can be estimated as,

$$Q(n, t_r(n)) = (1 - k_{sd} t_r(n)) \tilde{Q}(n), \quad n = 1, 2, \dots \quad (28)$$

where k_{sd} (1/h) is the rate constant of the self-discharge of the battery, $\tilde{Q}(n)$ is the battery capacity at the beginning of the n -th cycle (before resting).

Once the battery is fully charged, its stored energy slowly diminishes due to internal self-discharge [13] (*i.e.* with no load on the battery). Self-discharge varies with battery types. For a lithium-ion battery, the self-discharge rate is around 5% in the first 24 hours and then reduces to 1% - 2% per month, plus 3% for safety circuit consumption [19]. The typical values of k_{sd} for a lithium-ion battery cell is given in [9] as,

$$k_{sd} = 8 \times 10^{-5} \text{ 1/h}$$

The degradation of $\tilde{Q}(n)$ due to cycling is an irreversible and cumulative process and can be approximated by second-order approximation as,

$$\begin{aligned} \tilde{Q}(n) &= \tilde{Q}(n-1) + (n-1)k_{Qn1}Q + (n-1)^2 k_{Qn2}Q, \quad n = 1, 2, \dots \\ \tilde{Q}(0) &= Q \end{aligned} \quad (29)$$

where Q is the battery's capacity given by Equation (2), k_{Qn1} and k_{Qn2} are the first-order and second-order degradation rates of battery capacity due to cycling,

respectively.

The summation of above Equations (29) from 1 to n leads to,

$$\tilde{Q}(n) = Q \left[1 + \frac{(n-1)n}{2} k_{Qn1} + \frac{(n-1)n(2n-1)}{2} k_{Qn2} \right] \quad (30)$$

Since the capacity degradation of the battery is also dependent on temperature, Equation (30) is modified as

$$\begin{aligned} \tilde{Q}(n, T) = Q & \left(1 + \frac{(n-1)n}{2} k_{Qn1} + \frac{(n-1)n(2n-1)}{2} k_{Qn2} \right) \\ & \times \left(1 + k_{QT1} (T - T_{ref}) + k_{QT2} (T - T_{ref})^2 \right) \end{aligned} \quad (31)$$

where k_{QT1} and k_{QT2} are the first and second-order degradation rates of the battery capacity related to temperature change.

Finally, the degradation of a battery's capacity due to self-discharge, discharge current, life cycle and temperature is expressed by substituting Equations (31) and (2) into Equation (28) as

$$\begin{aligned} & Q(i, n, t_r(n), T) \\ & = (1 - k_{sd} t_r(n)) Q_0 \left(\frac{i}{i_0} \right)^\alpha \left(1 + \frac{(n-1)n}{2} k_{Qn1} + \frac{(n-1)n(2n-1)}{2} k_{Qn2} \right) \\ & \times \left(1 + k_{QT1} (T - T_{ref}) + k_{QT2} (T - T_{ref})^2 \right), \quad n = 1, 2, \dots \end{aligned} \quad (32)$$

Two battery capacities (Q_{n_1}, Q_{n_2}) at two different cycle numbers under normal or reference temperature are needed to calculate the cycling related degradation rates (k_{Qn1} and k_{Qn2}); another two battery capacities (Q_{T_1}, Q_{T_2}) at two different temperatures under in the first life cycle are required to estimate the temperature related degradation rates (k_{QT1} and k_{QT2}).

$$\begin{pmatrix} k_{Qn1} \\ k_{Qn2} \end{pmatrix} = \begin{pmatrix} \frac{(n_1-1)n_1}{2} & \frac{(n_1-1)n_1(2n_1-1)}{2} \\ \frac{(n_2-1)n_2}{2} & \frac{(n_2-1)n_2(2n_2-1)}{2} \end{pmatrix}^{-1} \begin{pmatrix} \frac{Q_{n_1}}{(1 - k_{sd} t_r(n_1)) \times Q_0 \times (i_n/i_0)^\alpha} - 1 \\ \frac{Q_{n_2}}{(1 - k_{sd} t_r(n_2)) \times Q_0 \times (i_n/i_0)^\alpha} - 1 \end{pmatrix} \quad (33)$$

where the two battery capacities, Q_{n_1} and Q_{n_2} , can be found on the two discharge curves provided by manufacturer's datasheet under two different cycling numbers n_1 and n_2 at normal temperature, as shown in **Figure 2**.

$$\begin{pmatrix} k_{QT1} \\ k_{QT2} \end{pmatrix} = \begin{pmatrix} T_1 - T_{ref} & (T_1 - T_{ref})^2 \\ T_2 - T_{ref} & (T_2 - T_{ref})^2 \end{pmatrix}^{-1} \begin{pmatrix} \frac{Q_{T_1}}{(1 - k_{sd} t_r(1)) \times Q_0 \times (i_T/i_0)^\alpha} - 1 \\ \frac{Q_{T_2}}{(1 - k_{sd} t_r(1)) \times Q_0 \times (i_T/i_0)^\alpha} - 1 \end{pmatrix} \quad (34)$$

where the two battery capacities, Q_{T_1} and Q_{T_2} , can be found on the two discharge curves from the manufacturer's datasheet under two different temperatures T_1 and T_2 at the first cycle, as shown in **Figure 3**.

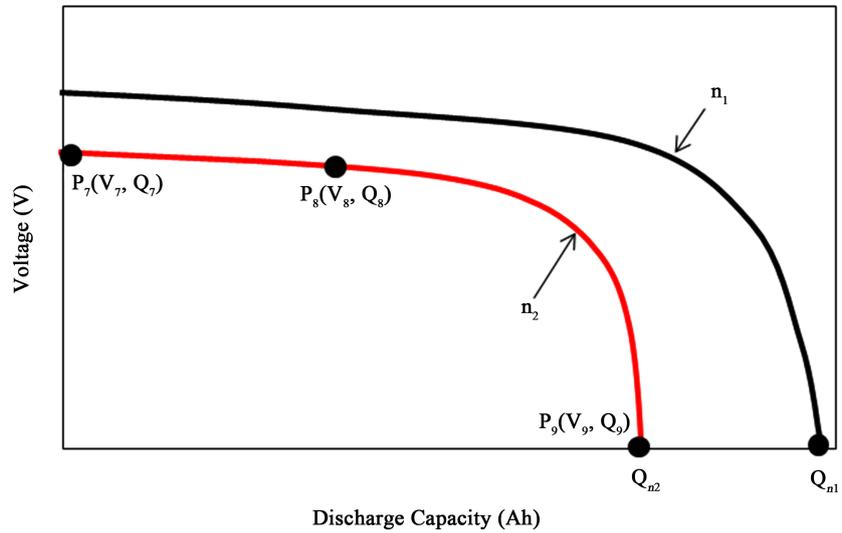


Figure 2. Two typical discharge curves of two cycle numbers n_1 and n_2 ($n_2 > n_1 > 1$) at normal temperature ($T = T_{ref}$) and same constant discharge current (i_n).

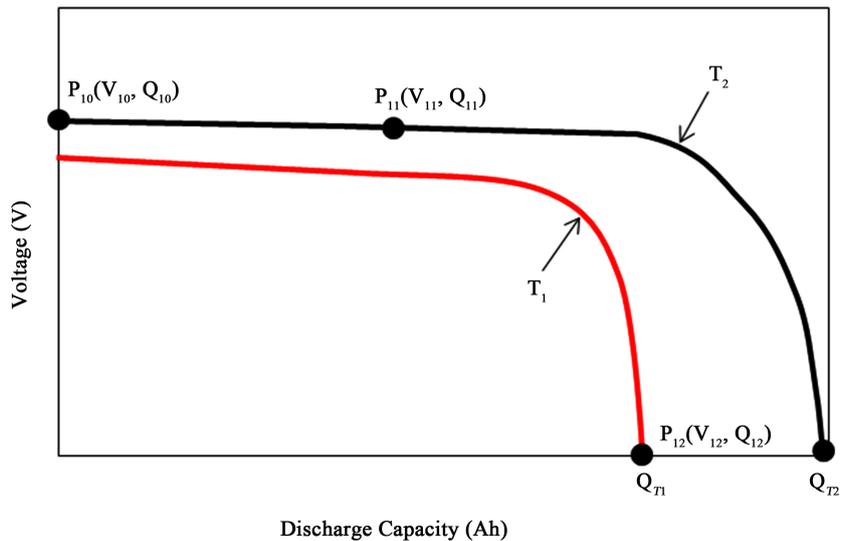


Figure 3. Two typical discharge curves of two temperatures T_1 and T_2 at the first cycle ($n = 1$) and same constant discharge current (i_n).

3.2. First-Order Degradation Rate Constants Related to Cycling Life k_{xn1}

To determine the first-order degradation rate constants related to cycling life k_{xn1} , point P_1 in **Figure 1**, and three other points on the discharge curve of $n = n_2$ ($n_2 > n_1 > 1$), namely points P_7 , P_8 and P_9 , are required, as shown in **Figure 2**.

Applying the modified cell voltage Equation (23) with Equations (24)-(27) to point P_1 with $n = 2$, points P_7 , P_8 and P_9 respectively, and rearranging the resulting 4 equations into the following matrix form,

$$Gk_{xn} = b \tag{35}$$

where

$$\mathbf{k}_{Xn} = \begin{pmatrix} k_{Vn1} \\ k_{Rn1} \\ k_{Kn1} \\ k_{An1} \end{pmatrix}; \mathbf{b} = \begin{pmatrix} V_{full} - (V_0 - R_0 i_1 - K_0 + A_0) \\ V_7 - (V_0 - R_0 i_n - K_0 + A_0) \\ V_8 - \left(V_0 - R_0 i_n - K_0 \frac{mQ}{mQ - Q_8} + A_0 \exp(-B_0 Q_8) \right) \\ V_9 - \left(V_0 - R_0 i_n - K_0 \frac{mQ}{mQ - Q_9} + A_0 \exp(-B_0 Q_9) \right) \end{pmatrix} \quad (36)$$

and

$$\mathbf{G} = \begin{pmatrix} V_0 & -R_0 i_1 & -K_0 & A_0 \\ V_0(n_2 - 1) & -R_0 i_n(n_2 - 1) & -K_0(n_2 - 1) & A_0(n_2 - 1) \\ V_0(n_2 - 1) & -R_0 i_n(n_2 - 1) & -K_0(n_2 - 1) \frac{mQ}{mQ - Q_8} & A_0(n_2 - 1) \exp(-B_0 Q_8) \\ V_0(n_2 - 1) & -R_0 i_n(n_2 - 1) & -K_0(n_2 - 1) \frac{mQ}{mQ - Q_9} & A_0(n_2 - 1) \exp(-B_0 Q_9) \end{pmatrix} \quad (37)$$

Note that the value of Q in expressions (36) and (37) should be the value of $Q(i, n, t_r(n), T)$ at $i = i_n, n = n_2, T = T_{ref}$.

The first-order degradation rate constants related to cycling life in Equations (24) to (27), k_{Vn1} , k_{Rn1} , k_{Kn1} and k_{An1} can be obtained by solving Equation (35),

$$\mathbf{k}_{Xn} = \mathbf{G}^{-1} \mathbf{b} \quad (38)$$

The internal resistance variation is usually very small and can be neglected within a certain life span (for example, $n \leq n_0 = 1000$) and therefore, the degradation rate constant related to cycling life for internal resistance is modified as

$$k_{Rn1} = \begin{cases} 0, & n \leq n_0 \\ k_{Rn1}, & n > n_0 \end{cases} \quad (39)$$

3.3. First-Order Degradation Rate Constants Related to Temperature k_{XT1}

Similar to k_{xn1} , the degradation rate constants related to temperature k_{XT1} can be calculated by applying the procedures given in Section 3.2 to point P_1 with $T = T_{ref} + 1$, points P_{10} , P_{11} and P_{12} , as shown in **Figure 3**. The formulas for the calculation are given as follows.

$$\tilde{\mathbf{G}} \tilde{\mathbf{k}}_{XT} = \tilde{\mathbf{b}} \quad (40)$$

where

$$\tilde{\mathbf{k}}_{XT} = \begin{pmatrix} k_{VT1} \\ k_{RT1} \\ k_{KT1} \\ k_{AT1} \end{pmatrix}; \tilde{\mathbf{b}} = \begin{pmatrix} V_{full} - (V_0 - R_0 i_1 - K_0 + A_0) \\ V_{10} - (V_0 - R_0 i_T - K_0 + A_0) \\ V_{11} - \left(V_0 - R_0 i_T - K_0 \frac{mQ}{mQ - Q_{11}} + A_0 \exp(-B_0 Q_{11}) \right) \\ V_{12} - \left(V_0 - R_0 i_T - K_0 \frac{mQ}{mQ - Q_{12}} + A_0 \exp(-B_0 Q_{12}) \right) \end{pmatrix} \quad (41)$$

and

$$\tilde{\mathbf{G}} = \begin{pmatrix} V_0 & -R_0 i_1 & -K_0 & A_0 \\ V_0(T_2 - T_{\text{ref}}) & -R_0 i_T(T_2 - T_{\text{ref}}) & -K_0(T_2 - T_{\text{ref}}) & A_0(T_2 - T_{\text{ref}}) \\ V_0(T_2 - T_{\text{ref}}) & -R_0 i_T(T_2 - T_{\text{ref}}) & -K_0(T_2 - T_{\text{ref}}) \frac{mQ}{mQ - Q_{11}} & A_0(T_2 - T_{\text{ref}}) \exp(-B_0 Q_{11}) \\ V_0(T_2 - T_{\text{ref}}) & -R_0 i_T(T_2 - T_{\text{ref}}) & -K_0(T_2 - T_{\text{ref}}) \frac{mQ}{mQ - Q_{12}} & A_0(T_2 - T_{\text{ref}}) \exp(-B_0 Q_{12}) \end{pmatrix} \quad (42)$$

Note that the value of Q in expressions (41) and (42) should be the value of $Q(i, n, t_r(n), T)$ at $i = i_T, n = 1, T = T_2$.

The first-order degradation rate constants related to temperature in Equations (24) to (27), k_{VT1} , k_{RT1} , k_{KT1} and k_{AT1} can be obtained by solving Equation (40),

$$\tilde{\mathbf{k}}_{XT} = \tilde{\mathbf{G}}^{-1} \tilde{\mathbf{b}} \quad (43)$$

The final battery model takes into account the effects of self-discharge, cycling life and temperature on capacity degradation and battery voltage, and these dependencies are described by Equations (23) to (27), and (32). The identification of the model parameters depends only on the information typically provided in a manufacturer's datasheet and the procedures used to derive them are presented in detail.

4. Model Validation

The modified model is verified by two steps. First, the model is applied to the two batteries used in [7] and the predicted battery voltages are compared with those generated by the generic MATLAB model. Secondly, the modified model with the inclusion of aging and temperature is applied to a commercially available lithium-ion battery and the model parameters are extracted. The new model's calculated battery voltage is then compared with the performance figures provided in the manufacturer's datasheet.

4.1. Comparison with the MATLAB Model

Both the nickel-metal hydride (NiMH) and lithium-ion batteries from [7] are used here to check the improvement of battery voltage prediction of the modified model Equations (1) and (2). For the NiMH battery, **Figure 4** shows the comparison of voltage prediction by the modified model with the prediction by the generic MATLAB model, and with the values on the manufacturer's datasheet as well. It can be seen that the overall accuracy of the modified model is better than the generic MATLAB model for all three discharge rates, especially beyond the nominal region.

Figure 5 plots the voltages from both MATLAB and the modified model calculation and from manufacturer's datasheet for the lithium-ion battery. It is clearly seen that the voltage prediction from the modified model at high C-rate is much better than the MATLAB model prediction. Another significant improvement

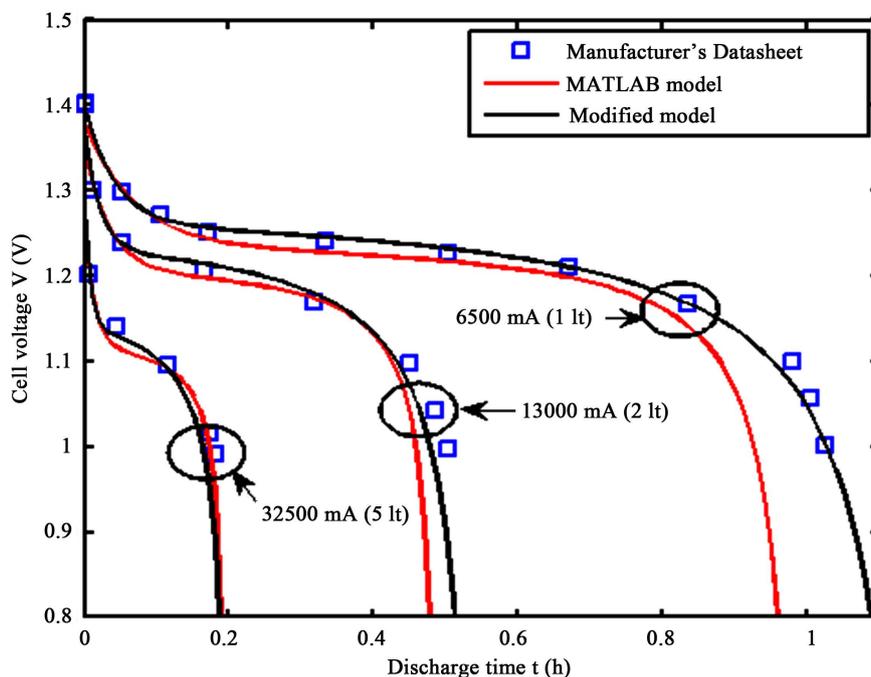


Figure 4. Comparison of voltage prediction under different discharge currents for the NiMH battery [7].

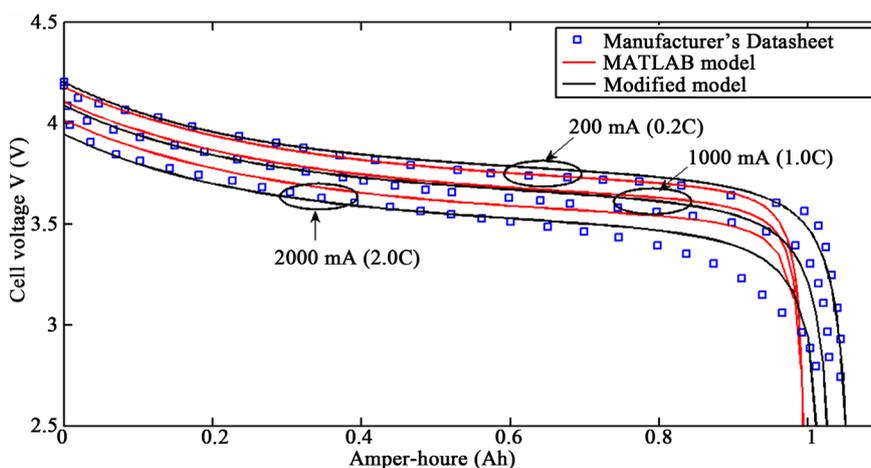


Figure 5. Comparison of voltage prediction under different discharge currents for the lithium-ion battery [7].

is the voltage prediction beyond the nominal zone. All of the three discharge curves (red lines) predicted by the generic MATLAB model converge to the rated capacity which is not the case comparing with the reference data. The modified model demonstrates much higher accuracy within this zone and has correctly predicted the separation of the three discharge curves (black lines) at the end of discharge process.

Figure 6 plots the battery capacity as the function of discharge current and in comparison with the data extracted from the manufacturer's data for both NiMH and lithium-ion batteries from [7].

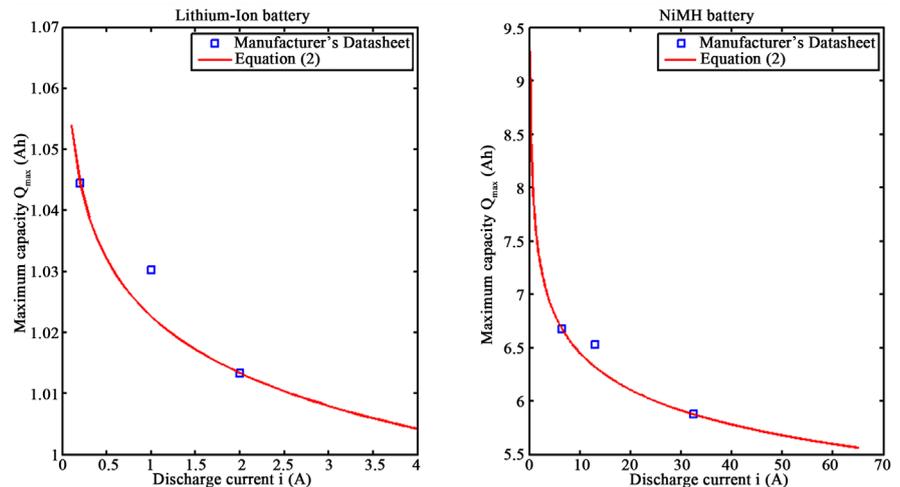


Figure 6. Battery capacity changes with discharge current for both lithium-ion and NiMH batteries from [7].

Table 1 lists the model parameters of the generic MATLAB model and the modified model for both the NiMH and lithium-ion batteries.

4.2. Model Validation Using a Commercially Available Large-Format Lithium-Ion Battery

To further validate the modified battery model, a large-format high capacity (1000 Ah) lithium-ion battery from SINOPOLY was used. The specifications and discharge curves of the battery (Model: SP-LFP1000AHA) can be found in [20]. Based on the procedures given in Sections 2 and 3, the parameter values for the modified model are extracted and listed in **Table 2**. For comparison, the generic MATLAB model parameter values are also estimated and presented in **Table 2**.

The capacity of the SP-LFP1000AHA battery varies with the discharge current, temperature and cycling number. Comparing with the experimental data, Equation (32) gives a very good approximation, as shown in **Figure 7**.

Figure 8 shows the voltage vs capacity curves from the manufacturer's datasheet, the MATLAB model, and the modified model for three different discharge currents. The modified model has comparable accuracy with the generic MATLAB model in terms of the voltage prediction in the exponential and nominal zones but has much higher accuracy beyond the nominal zone. Near the end of the discharge process, the new model predicts the battery's output voltage much closer to the results provided in the manufacturer's datasheet while the generic MATLAB model incorrectly converges the three (red) curves to a single capacity value due to the MATLAB model doesn't consider the battery capacity variation with discharge currents.

The voltage predictions by the new model are compared with the manufacturer's datasheet for three different cycling numbers and the results are plotted in **Figure 9**.

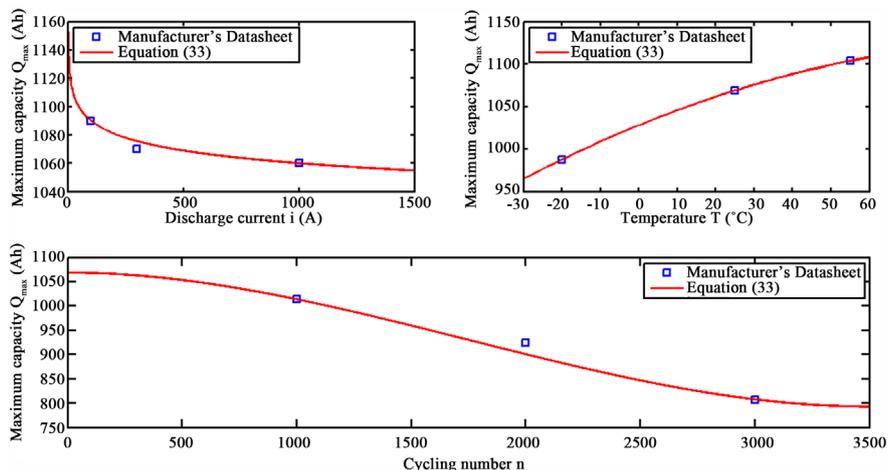


Figure 7. Comparison of capacity prediction with experimental data for the SP-LFP1000AHA battery.

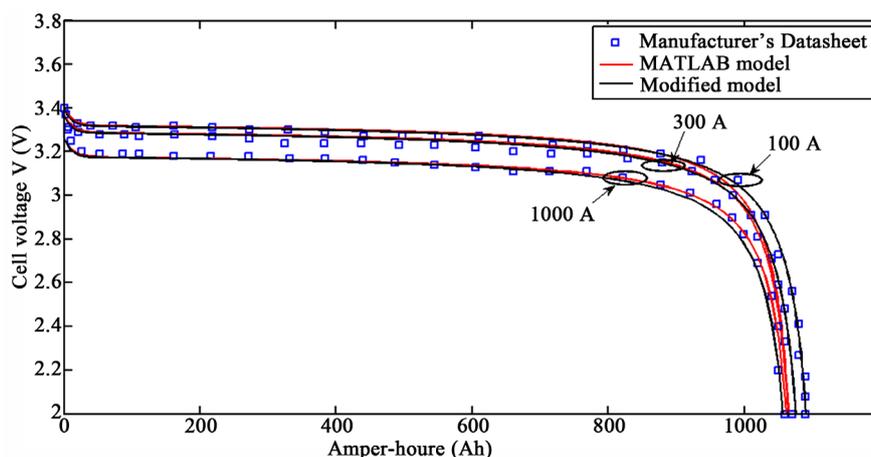


Figure 8. Comparison of voltage prediction under different discharge currents for the SP-LFP1000AHA lithium-ion battery.

Table 1. Model parameters.

Battery	Nickel Metal Hybrid (1.2 V 6.5 Ah)		Lithium-Ion (3.6 V 1 Ah)	
	MATLAB Model [7]	Modified Model	MATLAB Model [7]	Modified Model
V_0 (V)	1.2848	1.3298	3.7348	3.82626
R (Ω)	0.0046	0.0047	0.09	0.14222
K (V)	0.01875	0.03842	0.00876	0.02054
A (V)	0.144	0.14117	0.468	0.42273
B ($A \cdot h$) ⁻¹	2.3077	3.79613	3.5294	4.208
m	-	1.14849	-	1.02115
α	-	-0.07897	-	-0.01313
$Q_{max,0}$ ($A \cdot h$)	-	6.67206	-	1.04445
i_0 (A)	-	6.5	-	0.2

Table 2. Model parameters for the SP-LFP1000AHA battery.

Parameter	MATLAB model	Modified model	
V_0 (V)	3.36757	3.36964	
R_0 (Ω)	0.00016	0.00016	
K_0 (V)	0.03157	0.03546	
A_0 (V)	0.08	0.08165	
B ($A \cdot h$) ⁻¹	0.07712	0.1003	
m		1.0269	
T_{ref} ($^{\circ}C$)		25 (assumed)	
k_{Vn1}		9.71249×10^{-6}	For Equations (23)-(27)
k_{Rn1}		7.51635×10^{-4}	
k_{Kn1}		-8.59363×10^{-5}	
k_{An1}		-2.92489×10^{-4}	
k_{VT1}		1.05135×10^{-3}	
k_{RT1}		1.83721×10^{-2}	
k_{KT1}		-7.72438×10^{-3}	
k_{AT1}		-4.31833×10^{-2}	
n_0		1000 (assumed)	For Equation (39)
t_r (n)		0 (assumed)	
α		-0.01212	
Q_0 ($A \cdot h$)	1090	1090	
i_0 (A)		100	
k_{sd} (1/h)		8×10^{-5} [9]	For Equation (32)
k_{Qn1}		-1.27571×10^{-7}	
k_{Qn2}		1.22095×10^{-11}	
k_{QT1}		1.32729×10^{-3}	
k_{QT2}		-7.9763×10^{-6}	

Figure 10 depicts the temperature effects on the discharge curves calculated by the new model under three different temperatures and compares them with the manufacturer's datasheet.

It can be seen from **Figure 9** and **Figure 10** that the new model has the capability to predict the discharge curves for different cycling numbers and at different temperatures with a reasonable accuracy.

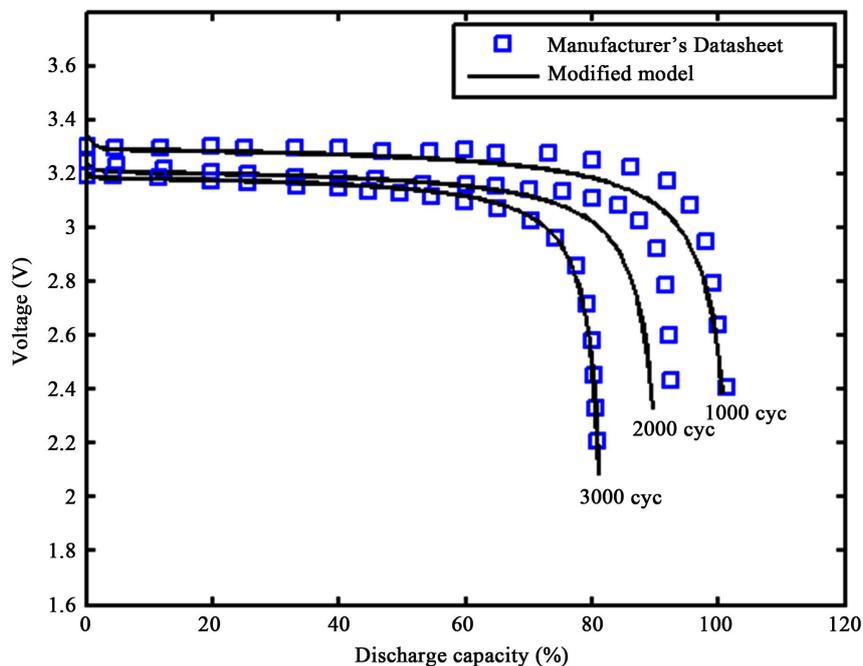


Figure 9. Comparison of voltage prediction over cycle life under different discharge conditions for the SP-LFP1000AHA lithium-ion battery.

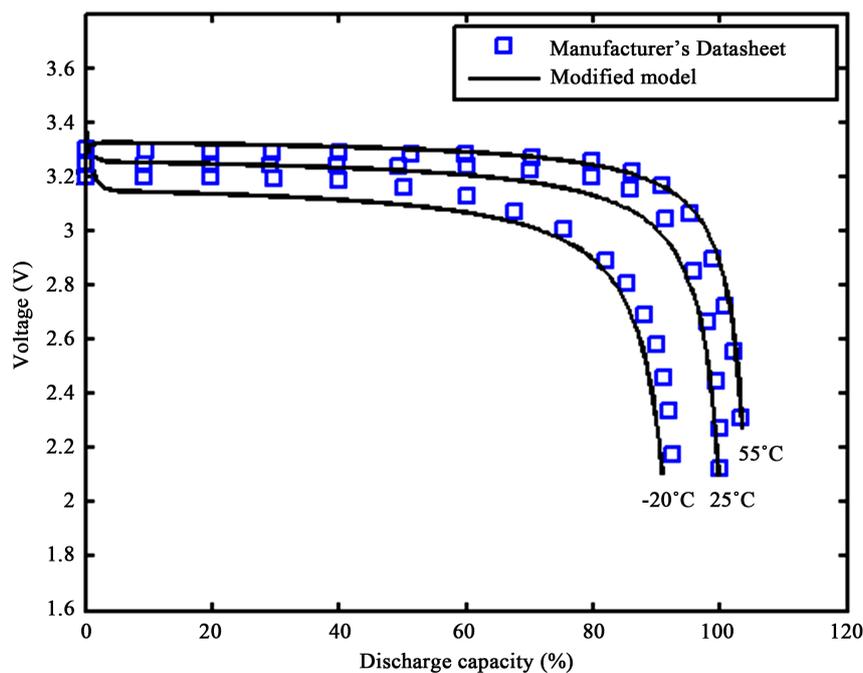


Figure 10. Comparison of voltage prediction under different temperatures for the SP-LFP 1000AHA lithium-ion battery.

5. Conclusion

The well accepted generic MATLAB model is improved by adopting a variable battery capacity which is expressed as a function of self-discharge, discharge current, cycle number and temperature. The dependence of the model parame-

ters on cycle life and temperature are linearly estimated. The detailed procedures are developed to determine the model parameters and the capacity function. The model parameter identification depends only on information typically available from manufacturer's datasheets. The new model was compared with the generic MATLAB model for a nickel-metal hydride battery and the lithium-ion battery that are used for the generic MATLAB battery model validation. The simulation results demonstrated that the newly developed model has similar accuracy within the exponential and nominal zones compared to the generic MATLAB model but can correctly capture the discharge curve separation behavior beyond the nominal zone where the generic MATLAB model failed. Further validation on a large capacity lithium-ion battery has shown that the modified model can accurately predict the discharge curves at different discharge currents, at different points in cycle life, and at different temperatures. Future work will be devoted to further validate the model with various battery types, integrate it into a grid-scale energy storage control system with the goal of improving its reliable and economic operation.

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