Mission Profile based System-level Reliability Analysis of DC/DC Converters for a Backup Power Application

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Abstract—Reliability analysis is an important tool for assisting the design phase of a power electronic converter to fulfill its life-cycle specifications. Existing converter-level reliability analysis methods have two major limitations: 1) based on constant failure rate models and 2) lack of considerations of long-term operation conditions (i.e., mission profile). Although various studies have been presented on power electronic component-level lifetime prediction based on wear-out failure mechanisms and mission profile, it is still a challenge to apply the same method to the reliability analysis of converters with multiple components. Component lifetime prediction based on associated models provides only a B, lifetime information (i.e., the time when X% items fail), but the time-dependent reliability curve is still not available. In this paper, a converter-level reliability analysis approach is proposed based on time-dependent failure rate models and long-term mission profiles. Two different methods to obtain the component-level time-to-failure are illustrated by a case study of dc/dc converters for a 5 kW fuel cell based backup power system. The reliability analysis of the converters with and without redundancy is also performed to assist the decision making in the design phase of the fuel cell power conditioning stage.

Index Terms—Fuel cell system, system-level reliability, Weibull distribution, power semiconductor, capacitor, reliability block diagram.

I. INTRODUCTION

Fuel cells are widely recognized as one of the most promising alternative power conversion technologies due to their high efficiency, and they provide very low pollution [1], [2]. In addition, the scalability of fuel cells allows extensive applications from low power to high power [3]. For instance, fuel cells have progressed from being a potential technology to a commercially viable power solution for mobile base stations [4]. Because telecom backup power systems are typically located in remote areas with harsh environments, frequent interruptions of the electrical grid place a high priority on the availability and reliability of the total system. As stated in [5], the reliability is the probability that a product can perform a required function under given conditions for a given time interval. The failure rate of a product, such a bathtub curve, consists of three major stages in its entire life cycle [6]: a decreasing failure rate caused by early infant mortality, a constant failure rate due to random failure, and an increasing failure rate because of the wear-out phenomenon. To design a backup power system to fulfill a certain service life, the failure due to component wear-out should be avoided or limited to an acceptable level.

![Fig. 1. Outage frequency of the subsystems in proton exchange membrane fuel cell power plants.](image)

According to a field study of residential-scale Proton-Exchange-Membrane (PEM) fuel cell power plants (1 – 5 kW) [7], the most prevalent means of failure in fuel cell systems are distributed as shown in Fig. 1. They are categorized from the fuel cell stack, power conditioning stage, and balance of the plant. Insight into the component-level durability and reliability of these subsystems of fuel cell systems are investigated in [7]-[9], especially for the fuel cell stack and the balance of the plant given their high outage frequency. In [7], the component failure is statistically analyzed based on a fleet of deployed PEM fuel cells, and the critical components can be identified with the recorded average lifetime. Unfortunately, the root cause of the failure mechanism cannot be identified by using this approach. In addition, it cannot easily evaluate the significant cause of the failures owing to the varying climate and operating conditions at the installation sites. Various efforts have been made to investigate the degradation mechanisms of fuel cell components to enhance durability [8], [9]. However, a quantitative lifetime model with respect to a certain stressor
of these components is lacking, which clearly affects the accurate reliability assessment of fuel cells facing different mission profiles. The reliability evaluation of the power conditioning stage is the main topic of this paper, which starts from the analysis of the critical components in the system. According to an industry survey [10], [11], the power semiconductors and capacitors are two types of fragile elements in the power electronics converter. Magnetic devices have potential reliability issues when applied in high-density power converters (i.e., thermal related degradation) or medium- to high-voltage applications (e.g., insulation degradation). For most of other applications, the thermal-related or insulation-related degradation is negligible for magnetic devices. Moreover, among the various stressors (vibration, humidity, temperature, etc.) for electronic equipment, temperature contributes up to 55% of the failure distribution [12]. Several research efforts have been devoted to the reliability prediction of electronic equipment [13]-[19]. On the one hand, the failure rate determined by using exponential distribution is adopted from various handbooks [15], [14]. This method is simple and inappropriate, considering only the operation period with a constant failure rate but neglecting the wear-out. On the other hand, the thermal stress of the critical components can be obtained according to the mission profile of the power converter used in renewable energy systems [15]-[19]. However, owing to the particular characteristics of the backup power application, this physics-of-failure based reliability evaluation cannot be performed with frequent switching between the standby mode and operation mode.

Based on the component-level reliability metrics, the system-level reliability can instead be derived by using the Reliability Block Diagram (RBD), the Fault Tree Analysis (FTA), and the Markov Chain (MC) [20]-[27]. In [20]-[22], the reliability of an interleaved dc/dc boost converter, an induction motor drive, and a PEM fuel cell power plant are evaluated using the MC method. In addition, the RBD approach is used to analyze the reliability of a paralleled inverter system [23] and a multi-level converter [24]. However, a constant failure rate is applied in both situations, which neglects the effects introduced by the mission profile. An FTA for the PEM fuel cell is performed in [25], where again a constant failure rate is assumed. Moreover, the most fundamental difference is that the success combination is focused in the RBD, whereas the failure combination is considered in the FTA. In other words, an FTA can easily be converted into an RBD, because the FTA analyzes the fixed probabilities (i.e., each event that comprises a tree has a fixed probability of occurring) and the RBD may include time-varying distribution for success and other properties [28]. However, this research very seldom considers the mission profile.

The background of this paper is related to a 5 kW commercial fuel cell system for backup power. It is an application requiring high reliability (i.e., 0.99) within a service life of 5 years. The motivation is to predict the reliability of the power conditioners (i.e., dc/dc converter stage) at the end of service life to better size the key electronic components for the next generation product design. The outcome for the study is used to assist the design phase of product development. The novel aspects of the proposed method of reliability evaluation are as follows: 1) consider long-term (i.e., one year) mission profile (i.e., environmental conditions, occurrence of power grid outages, standby mode and active mode), and 2) obtain the lifetime distribution of power MOSFETs and capacitors using Monte Carlo analysis and the degradation testing data to consider the parameter variations in both components and lifetime models.

The rest of the paper is organized as follows: Section II describes the mission profile and the topology of the fuel cell power converter used in the backup power application. Sections III and IV address the way to obtain the time-to-failure of the power semiconductors and the capacitors. Section V considers the reliability of the whole power stage by using the RBD and concluding remarks are drawn in the last section.

II. DESCRIPTION OF POWER STAGE FOR BACKUP POWER APPLICATION

The most commonly used technologies for backup power systems in telecom applications are lead-acid battery systems and generator sets. PEM fuel cells are emerging as an important alternative, because of their low operation temperature, and relatively fast response time [4]. Compared to batteries, fuel cells provide longer continuous runtime and greater durability in outdoor environments under a wide range of temperature conditions. They require less maintenance than either generators or batteries because they have fewer moving parts. Because of the ac output of the generator and the variable dc output of the fuel cell, a power conversion unit is evitable in these two cases. In addition, power electronics are needed to charge the batteries and control the loading current. However, the loading stresses of the power electronic components are different owing to various output characteristics of the batteries, generators and fuel cells.

The fuel cell system consists of three subsystems: the balance of plant regulates the pressure of fuel and air, and maintains the fuel cell at a reasonable temperature by controlling the coolant loop. The fuel cell stack converts the chemical energy into electricity through an electrochemical reaction. Afterwards, a power stage is applied to provide a stable dc-bus used in telecom applications from the varying stack output voltage.

A. Description of mission profile

Regarding the backup power, two major working modes can be identified. The power converter mainly works in the standby mode when the power grid is operational, whereas the power converter sometimes works in the operation mode in the case of a power outage.
The annual ambient temperature with a sample rate of one day is shown in Fig. 2(a), and it can be distributed into different temperature ranges. As shown in Fig. 2(b), the stability of the power grid is an essential mission profile, and daily outage with 4-hour duration, which determines the operation period of the fuel cell system, can be expected for severe users [29]. Meanwhile, the loading profile of the telecom application is repeated with a 10-hour quarter load and 2-hour full load, as stated in [4]. Moreover, the input of the power stage, determined by the V-I characteristics of the fuel cell stack, is shown in Fig. 2(c).

**Fig. 2.** Annual mission profile of telecom power. (a) Temperature distribution with a sample rate of one day; (b) Loading profile of the telecom application and operation period of the fuel cell system; (c) Output voltage and power of a fuel cell in relationship with current density.

B. Structure of power converter

Due to the variable output voltage of the fuel cell stack, a dc/dc power converter is required to match the voltage in telecom applications. A topology using galvanic isolation is shown in Fig. 3, where the rated power of the converter is 1 kW, and six 1 kW converters are connected in parallel for a 5 kW power stage to obtain the redundancy. Moreover, a synchronous rectification is adopted to achieve low conduction losses in the situation of low-voltage and high-current at the secondary-side of the transformer [29].

**Fig. 3.** Topology of isolated 1 kW dc/dc power converter used in a fuel cell backup telecom power.

The specification and main parameters of the 1 kW power converter are listed in Table I. Because the power semiconductors and dc capacitors are the major reliability-critical components in the power electronic converter, their performance will be evaluated and discussed in detail as follows.

**Table I**

<table>
<thead>
<tr>
<th>Power Converter Specification and Parameters</th>
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<tbody>
<tr>
<td>Input voltage $V_{in}$</td>
<td>30 – 65 V</td>
</tr>
<tr>
<td>Output voltage $V_o$</td>
<td>48 V</td>
</tr>
<tr>
<td>Maximum output power $P_o$</td>
<td>1000 W</td>
</tr>
<tr>
<td>Primary-side MOSFETs</td>
<td>100 V/74 A, ×8</td>
</tr>
<tr>
<td>Secondary-side MOSFETs</td>
<td>100 V/74 A, ×8</td>
</tr>
</tbody>
</table>

III. TIM-to-FAILURE OF POWER ELECTRONICS

This section addresses the method to estimate the time-to-failure of the power semiconductors. It starts with the B10 lifetime due to the limited lifetime data, which means that 10% of a sample will fail at this operation time. It can then be further extended to a lifetime distribution, if the parameter deviations are taken into account.
A. \( B_{10} \) lifetime estimation

As shown in Fig. 3, since the reflected voltage of the transformer primary-side can be higher or lower than the voltage of the fuel cell, the power converter can operate in both the step-up mode and the step-down mode [29]. In the case of step-up mode, the primary-side inductor is charged by the activation of all transistors, and it is discharged by the parallel connection of the two transformers. As described in [26], because the transistors in the middle bridge carry double of the current compared to the side bridges, two MOSFET \( Q_{31} \) and \( Q_{34} \) are connected in parallel to realize the balanced loading among the bridges of the primary-side. In the case of the step-down mode, the inductor is charged by the parallel connection of the transformers, and it is discharged by the series connection of the transformers. This results in a variable loading of the transistors among the bridges in the primary-side.

At the primary-side, although the loading of the transistors is unbalanced among the different bridges, owing to the symmetrical loading of the upper and lower transistors, only \( Q_{31} \), \( Q_{33} \) and \( Q_{35} \) are chosen. In addition to the symmetrical loading of the upper and lower transistors, the two rectifiers share the same current and voltage loading. As a result, \( Q_{ij} \) can represent the loading at the secondary-side.

![Flowchart to predict lifetime of power semiconductor in the operation mode.](image)

The power semiconductor reaches the end-of-life when an overlap occurs between its stress level and the strength model. From the power cycling perspective, the stress analysis is related to the mission profile (e.g., the ambient temperature, the loading profile, and the grid availability), whereas the strength model is determined by the selection of the power device and associated cooling. As discussed in [31], although the approach to evaluate the thermal stress of the power semiconductor at the standby mode and the operation mode is different, only the operation mode is considered, because much higher junction temperature leads to the dominating lifetime consumption compared to the standby mode.

The flowchart to predict the lifetime of the power semiconductor in the case of the operation mode is shown in Fig. 4. Under this circumstance, the junction temperature of the power semiconductor is jointly determined by the ambient temperature and the loading profile. The loss distribution of the MOSFET mainly consists of the conduction loss \( P_{con} \) and the switching loss, which can be further divided into the turn-on losses \( P_{on} \) and the turn-off losses \( P_{off} \). The loss calculation of the MOSFET is consistent with Infineon application note [32], and the key waveforms of each MOSFET at different loading conditions during a switching period are discussed in detail by [31]. Owing to the discrete MOSFET, its own heat-sink results in an independent thermal system, and a thermal coupling can be neglected from the adjacent devices. Considering an ambient temperature of 40 °C as the worst case scenario, the mean junction temperature for a device is calculated, based on the thermal impedance from the junction to the ambient.

Regarding the lifetime model of the MOSFET, a discrete component is selected due to the low power rating of the converter and the cost. The chip die is soldered onto the copper base, and the bonding pads and die-attach are common failure mechanisms [33], [34], which are almost the same as the cases when the IGBT power module is applied [35], [36].

Due to the limited published power cycling data of the discrete MOSFETs [33], the lifetime model is consistent with the one mentioned in [34]. Compared with Bayerer’s lifetime model [35]-[37], the insignificant impacts from the mean junction temperature and on-time pulse duration are not considered, so the cycle to failure \( N_f \) can be expressed as,

\[
N_f = A \cdot dT_f^{b_1},
\]

where the power cycle is closely related to the junction temperature swing. In addition, \( A \) and \( b_1 \) can be obtained according to test data provided by [34]. Owing to small manufacturing variations of the devices, it is impossible to exactly predict the wear-out process when an individual device will fail. The only meaningful values are the failure probability for an individual device and the percentage of failed devices for entire populations. It is worth noting that these data are aimed for the \( B_{10} \) lifetime, which means that 10% of the sample fails when the power cycling reaches this number.

Together with the thermal profile of the power semiconductors, the cycle-to-failure can be calculated in the conditions of the full load and the quarter load, respectively. On the other hand, the annual power cycling of each loading...
condition can be estimated with the grid outage frequency. Based on Miner’s rule [38], [39], the annual damage $D_{MOS}$ can be calculated by the annual power cycling $n$ over the corresponding end-of-life power cycles:

$$D_{MOS} = \sum \frac{n(i)}{N_j(i)}$$

(2)

where $i$ indicates different loading conditions like the full load and the quarter-load.

The second type of uncertainty exists due to variances in the manufacturing process (like the typical, maximum and minimum on-state resistance of the MOSFET), which results in variation of the mean junction temperature and junction temperature fluctuation. In order to illustrate this, $Q_{S5}$ is selected as an example. As shown in Fig. 5, the junction temperature fluctuation experiences a variation of 5%. Each distribution is sampled by using Monte Carlo analysis, whose sample numbers results in the accuracy of the output distribution [5], [40]. Consequently, 10,000 samplings are chosen to establish the accumulated damage distribution.

In the condition that all parameter variations are taken into account, the annual damage distribution is depicted in Fig. 6(a). It is known that the time-to-failure data typically follow the Weibull distribution [41]:

$$f(t) = \frac{\beta}{\lambda} \left(\frac{t}{\lambda}\right)^{\beta-1} e^{-\left(\frac{t}{\lambda}\right)^\beta}$$

where $t$ is the time to failure, $\beta$ is the shape parameter, and $\lambda$ is the scale parameter.

**Table II**

<table>
<thead>
<tr>
<th>Equivalent Static Value for Each Mosfet</th>
<th>$Q_{S1}$</th>
<th>$Q_{S3}$</th>
<th>$Q_{S5}$</th>
<th>$Q_{S1}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of cycles per year $n$</td>
<td>365</td>
<td>365</td>
<td>365</td>
<td>365</td>
</tr>
<tr>
<td>Annual damage $D$ (year$^{-1}$)</td>
<td>5.20E-3</td>
<td>5.20E-3</td>
<td>6.30E-3</td>
<td>2.87E-6</td>
</tr>
<tr>
<td>Number of cycles to failure $N_f$</td>
<td>7.02E4</td>
<td>7.02E4</td>
<td>5.79E4</td>
<td>1.27E8</td>
</tr>
<tr>
<td>Mean junction temperature $T_{jm}$ ($^\circ C$)</td>
<td>62.5</td>
<td>61.0</td>
<td>102.7</td>
<td>43.3</td>
</tr>
<tr>
<td>On-state time $t_{os}$ (second)</td>
<td>14,400</td>
<td>14,400</td>
<td>14,400</td>
<td>14,400</td>
</tr>
<tr>
<td>Junction temperature fluctuation $dT_j$ ($^\circ C$)</td>
<td>83.9</td>
<td>85.5</td>
<td>54.4</td>
<td>14.4</td>
</tr>
</tbody>
</table>

It is noted that $\beta$ denotes the mean value of the distribution, and $\sigma$ denotes the standard deviation. To simplify the thermal stress from both the full load and the quarter load, the equivalent static values of the lifetime semiconductors can be estimated by considering the parameter variations.

**B. Time-to-failure of individual power semiconductors due to parameter deviation**

The previous discussion gives a $B_{10}$ annual damage of MOSFETs used in a power converter, but the uncertainties due to the statistic properties of the applied lifetime model and the parameter variations of the power device should also be taken into account. Therefore, a statistical approach to analyze the lifetime performance subject to parameter variations is carried out in detail by means of Monte Carlo analysis. Finally, the time-to-failure distribution of the power semiconductors can be estimated by considering the parameter variations.

Fig. 5. Normal distribution of the factors from the lifetime model. (a) $A$ – scaling factor; (b) $\beta$ – exponential factor of temperature swing; (c) $dT_j$ – junction temperature fluctuation.

Since the lifetime model is obtained from the accelerated test results based on a specific number of testing samples, there are uncertainties in the derived constant parameters. All parameters in the lifetime model as stated in (1) are distributed by means of a normal probability density function (pdf), assuming that $A$ and $\beta$ experience a variation of 5%.
where $\eta$ denotes the scale parameter, and $\beta$ denotes the shape parameter.

\[ f(t) = \frac{\beta}{\eta} \left(\frac{t}{\eta}\right)^{\beta-1} \exp\left[-\left(\frac{t}{\eta}\right)^\beta\right] \quad (3) \]

As a result, the fitting curve of the annual damage can be obtained with a scale parameter of $6.7E-3$ and a shape parameter of 2.60. Assuming that the mission profile is repeated every year, the Weibull distribution of the lifetime, which is defined as one over the annual damage, can be obtained as well. Thus, the unreliability or failure of the power switch $Q_{s5}$ can be calculated as shown in Fig. 6(b), which is the integration of the lifetime distribution. It is noted that 1% of MOSFETs are predicted to fail over more than 30 years.

With the static equivalent values of each component as listed in Table I, the lifetime distributions of the key MOSFETs can be calculated, considering 5% parameter variations from the lifetime model and the stress analysis. Since the scale parameter of the Weibull function denotes the value when 63.2% failure occurs, it is predicted that $Q_{s5}$ has the lowest scale parameter of 220.5 according to the accumulated damage estimation as shown in Fig. 4. For the accumulated failure shown in Fig. 7, it can be seen that the 5-year operation (i.e. desired lifetime of the fuel cell backup power system) results in 5.8E-5, 4.3E-5, and 3.5E-5 failure in $Q_{p5}$, $Q_{p3}$ and $Q_{p1}$, respectively.

**Fig. 7.** Monte Carlo analysis of four typical power switches in terms of the accumulated failure.

**IV. TIME-TO-FAILURE OF DC CAPACITORS**

Due to the lack of complete failure data, the previous section calculated the $B_{10}$ lifetime of the power semiconductors, and then the lifetime distribution can be obtained by considering the parameter deviations. In this section, the lifetime distribution of the dc capacitors is studied, based on the complete failure data of the capacitors.

**A. Weibull lifetime data**

The capacity change, dissipation factor and leakage current are generally considered as key indicators during the healthy condition of the electrolytic capacitors [42]. To obtain the failure statistics of the used capacitor, a degradation test is performed with a series of nine capacitors (56 µF/ 35 V) at the rated voltage, an upper category temperature (105 ºC) and rated ripple current, and the normalized capacitance and Equivalent Series Resistance (ESR) are regularly measured during 4,000 testing hours. As shown in Fig. 8, the results are analyzed by using the software tool Weibull++ [41]. It can be observed that during the process of the testing hours, the initial capacitance and ESR increase or decrease smoothly until they reach the turning point after 4,000 hour. This agrees with the end-of-life criteria of the individual capacitor, which typically is 20 % drop of initial capacitance.

Generally, the exact reliability and the probability of the failure can never be known unless the failure data of every unit in the population can be obtained. Since this usually is
not a realistic solution, the testing with a certain number of samples is used to estimate the reliability, which introduces the Confidence Level (CL) – a range within which these reliability values are likely to occur with a certain percentage of time. The widely-adopted 50\% CL is applied by using Median Rank (MR) [41], where the lifetime is neither over-estimated nor under-estimated. The time-to-failure of 9 samples are located as shown in Fig. 9, and they can be fitted in terms of a Weibull distribution with a shape factor of 5.12 and scale factor of 6,804. It can be seen that the claimed lifetime of 5,000 hour in the datasheet means almost the $B_{20}$ lifetime. Moreover, the $B_X$ lifetime at any condition can be calculated based on this.

\[
L = L_0 \cdot 2^{\frac{T-T_0}{\eta V_0}}
\]

where $L_0$ and $L$ are the lifetime under the reference condition and the use condition, $V_0$ and $V$ are voltage at reference condition and use condition, and $T_0$ and $T$ are the temperature at the reference condition and use condition, respectively, $n_1$ is the temperature dependent constant, and $n_2$ is the voltage stress exponent. Based on the leading capacitor manufacturers [43], $n_2$ equals 0 for small radial capacitors (below a rated voltage of 160 V), as the temperature dependent electrolyte loss dominates the lifetime model. For medium or large sizes (snap-in and screw terminal types), the more closely the operating voltage approaches the rated voltage, the more the electrolyte is consumed for self-healing of small flaws within the dielectric layer. In addition, $n_1$ equals 10 for the analyzed capacitors, which follows the “10-Kelvin-rule” deduced from the Arrhenius law. According to the above equation, the lifetime estimation at other operational temperature stresses can be calculated.

![Fig. 8. Capacitor degradation testing results at rated voltage, rated ripple current and upper category temperature (105 °C). (a) Normalized capacitance; (b) Normalized ESR.](image)

As discussed in [42], [43], the state-of-the-art lifetime model for the electrolytic capacitors is affected by the temperature stress and the voltage stress, and it is given by,

![Fig. 9. Time-to-failure of capacitors at 50\% Confidence Levels (CL) by using Weibull distribution using upper category temperature of 105 °C.](image)

### B. Time-to-failure distribution of individual capacitor

According to the loading condition and thermal model of the electrolytic capacitor, the operation temperature of the capacitor can be calculated in the cases of the standby mode and the operation mode. Afterwards, based on the lifetime model of the capacitor, the expected $B_X$ lifetime for the input-side and output-side capacitor can be calculated.

In the case of standby mode, the capacitor temperature is almost the same as the ambient temperature. Based on the ambient temperature distribution and the stability of the power grid, the annual damage can be calculated for each temperature range with the help of the lifetime model listed in (4), which can be accumulated as the total annual damage by using Miner’s rule. In the case of the operation mode, the capacitor temperature is jointly determined by the loading condition and the ambient temperature. As shown in Fig. 10,
the loss of the capacitor is caused by the ripple current across
the ESR, whereas the temperature of the capacitor can be
estimated based on the thermal impedance of the capacitor as
well as its cooling solutions.

\[ D_{cap} = \sum_{i=1}^{2} \frac{P_i \cdot 365 \cdot 24 \cdot P_2(i)}{L(i)} = \frac{P_i \cdot 365 \cdot 24}{L} \sum_{i=1}^{2} \frac{P_2(i)}{2} \cdot \frac{1}{10} \quad (5) \]

where \( i \) indicates the temperature sequence ranging from 5
°C to 35 °C as shown in Fig. 2(a). \( P_i \) denotes the percentage
of the power converter working in the standby-operation
mode, \( P_2 \) denotes the percentage of the temperature
distribution, and \( L \) denotes the predicted lifetime at the
temperature level, which is the ambient temperature in the
case of the standby mode and the sum of the ambient
temperature and core temperature rise in the case of the
operation mode.

Although the types of input-side and output-side
capacitors are different from the tested capacitor, all of them
belong to the family of the radial electrolyte capacitor, whose
lifetime expectancy is 5,000 hour as mentioned in their
datasheets. Consequently, the lifetime model as shown in
Fig. 9 is suitable for both the input capacitor and the output
capacitor. Based on the \( B_{10} \) and \( B_1 \) lifetime model, the
damage of the output-side capacitor is shown in Fig. 11, in
which the standby mode and the operation mode are
compared. Although the loading condition leads to a rise of
the capacitor temperature, when the \( B_{10} \) lifetime is taken into
account, the higher proportion of standby mode causes a
higher annual damage of 6.4E-3 compared to the operation
mode of 5.2E-3. Meanwhile, if the \( B_1 \) lifetime model is
considered, the annual damage in both the standby mode and
the operation mode is higher than that of \( B_{10} \) because of its
lower cycle-to-failure at the same stress level. Similarly, the
same approach can be extended to the input-side capacitor,
and the \( B_{10} \) and \( B_1 \) damage can be calculated as well.

Fig. 10. Relationship between the electrical and thermal models of
electrolytic capacitors.

![Fig. 11. Annual damage of the output-side capacitor in the case of
the standby mode and operation mode. (a) \( B_{10} \) damage; (b) \( B_1 \)
damage.](image)

The annual damage of the capacitor \( D_{cap} \) is defined as,
factors for the individual output capacitor are 1.93 and 199, respectively. Similarly, in the case where the 5-year operation (desired lifetime of the fuel cell power stage) is in focus, the unreliability of the capacitor along with the operation time is shown in Fig. 12. It is evident that 5-year operation of the fuel cell system induces 11.1E-4 and 8.1E-4 failure for the input capacitor and output capacitor, respectively.

V. RELIABILITY ASSESSMENT OF POWER STAGE

In order to assess the reliability metrics of the entire power stage in the fuel cell system, major steps can be divided into the reliability analysis of a 1 kW power converter and a 5 kW power stage. By using RBD, the procedure to calculate the reliability function is shown in Fig. 13. It is evident that the reliability evaluation of the total power semiconductors and capacitors is calculated by each component. Afterwards, the 5 kW power converter stage can be investigated based on the 1 kW power converter. Due to the same time-to-failure characteristic of the used power devices and dc capacitors, only these components are considered in this paper.

For the reliability analysis of a 1 kW power converter, the existence of any failed MOSFET or capacitor results in abnormal operation of the power converter, which indicates that all MOSFETs and capacitors are serially connected in RBD. As the reliability of the series block is the product of all components, the failure function of sub-system $F_{\text{sub}}$ can be expressed by the component failure function $F_{\text{com}}$ [5],

$$F_{\text{sub}}(t) = 1 - \prod_{i} (1 - F_{\text{com}(i)}(t))$$

(a)

Fig. 13. System-level reliability calculation by using reliability block diagram. (a) Composition of MOSFETs and capacitors in 1 kW power converter; (b) Composition of 5 kW power stage considering redundancy.

As mentioned before, the MOSFETs are not evenly stressed and the four representing MOSFETs can be found. The reliability of the 1 kW power converter can then be calculated by considering all MOSFETs used in the primary-side and secondary-side. As shown in Fig. 14(a), the reliability of the total MOSFETs can be deduced from each

(a)

(b)

(c)

Fig. 14. Accumulated percentage of unreliability from component-level to system-level. (a) From MOSFETs to 1 kW power converter; (b) From capacitors to 1 kW power converter; (c) From 1 kW power converter to 5 kW power stage with and without redundancy.
component. It can be seen that the damage of the 5-year operation (desired lifetime of the backup power) increases from less than 0.01% of the most stressed MOSFET to almost 0.04% of all MOSFETs existed in the 1 kW power converter. Meanwhile, the reliability of the total capacitors from the input capacitor and output capacitor is shown in Fig. 14(b).

The reliability of the whole power stage can be estimated from the reliability analysis of the 1 kW power converter, where six 1 kW power converters are connected in parallel for a 5 kW application. In the case of m-out-of-n redundancy, the failure function of the system $F_{\text{sys}}$ can be expressed as [5],

$$F_{\text{sys}}(t) = \sum_{i=0}^{n-1} \frac{n!}{i!(n-i)!} (1-F_{\text{sub}}(t))^i \cdot F_{\text{sub}}(t)^{n-i}$$  \label{eq:sys}

The unreliability of the whole power stage is shown in Fig. 14(c), where the cases with and without redundancy are compared as well. Due to the fact that five reliability blocks are serially connected in the condition without redundancy, the lifetime of the power stage is significantly reduced compared to the 1 kW power converter. However, in the case of using redundancy, the reliability of the power stage can be enhanced compared with no redundancy. It can be seen that the accumulated damage reaches 1% (i.e. $R=0.99$) for approximately 2 years of the power stage without redundancy, whereas it can be enhanced to 7.5 years with $N+1$ redundancy. It is noted that 5-year operation results in 1.3% failure of the 1 kW power converter, whereas it has 0.3% and 6.3% failure whether the redundancy is applied. It is concluded that, in order to achieve the high reliability (i.e., 0.99) within a service life of 5 years, the $N+1$ configuration must be employed. Alternatively, it can be achieved by sizing the dc capacitors to be more reliable.

VI. CONCLUSION

A system-level reliability analysis method of dc/dc converters for the power conditioning stage of a fuel cell stack is proposed in this paper. Mission profile and Weibull distribution based analysis is applied by using the long-term electro-thermal stress profiles and the time-to-failure distribution of key components. A case study of an industrial designed 5 kW fuel cell system with multiple dc/dc converters is presented. The reliability curves of the MOSFETs and capacitors used in the dc/dc converters is predicted by considering the associated electro-thermal loadings and statistical properties in lifetime models and component parameters. The unreliability of the group for 16 MOSFETs and the group of 14 capacitors reaches 1% when the single dc/dc converter operates 18 years and 4.5 years, respectively. Without redundancy, it will last only 2 years for the power conditioning stage with five paralleled 1 kW dc/dc power converters. To extend its lifetime, a redundant design with additional 1 kW dc/dc converter is analyzed and the operational time is increased to 7.5 years, which fulfills the high reliability demands (i.e., 0.99) for a service life of 5 years. Although the study case is carried out in a fuel cell application, a general approach is to evaluate the system-level reliability. By knowing the mission profile, and the thermal stress and lifetime distribution of the critical components, the reliability of the power converter used in popular applications (e.g., renewable energy system, drive system) can be investigated in a similar way.

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