An Online Self Cold Startup Methodology for PEM Fuel Cells in Vehicular Applications

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Abstract—This paper puts forward an adaptive cold start strategy for a proton exchange membrane fuel cell (PEMFC) based on maximum power mode. The proposed strategy consists of a water evacuation process after PEMFC shutdown and a self-heating process at PEMFC cold startup. To maximize the performance of the suggested strategy, an optimal operating condition for the cold start procedure is sought first. In this respect, an experimental parametric study is performed to explore the impact of fan velocity, micro-short circuit, anode pressure, and purge procedure on the PEMFC cold start performance. After laying down the proper conditions, the proposed cold start procedure is implemented on a test bench for experimental validations. The self-heating process is based on an online adaptive algorithm that maximizes the PEMFC’s internal heat depending on its operating parameters’ variation. In fact, this algorithm attempts to keep the current density at high levels, leading to PEMFC’s performance improvement achieved by membrane hydration and temperature increase. The experimental results confirm the effectiveness of the proposed strategy, which presents a fast and cost-effective PEMFC’s cold start.

Index Terms—Adaptive control, Cold start strategy, Kalman filter, Proton Exchange Membrane Fuel Cell, Vehicular applications

I. INTRODUCTION

TRANSPORTATION is a major pollution contributor, generating substantial quantities of nitrogen oxides, carbon monoxide, and other pollutants [1]. A wide range of technological solutions, such as hybrid electric vehicles (HEVs), battery vehicles, and fuel cell hybrid electric vehicles (FCHEVs) offer the potential to mitigate these impacts [2]. Among them, FCHEVs, which use a proton exchange membrane fuel cell (PEMFC) as the primary source and a battery pack as the secondary one, have gained great public and governmental recognition [3]. FCHEVs can be refueled in a few minutes, and used for a wide driving range [4, 5]. In addition, they can reduce dependency on fossil fuels and offer potential reductions of greenhouse gases, especially when fueled with hydrogen from electrolysis or biomass [6]. Although FCHEVs are becoming increasingly important in the automotive sector, their market penetration and economic viability depend on the technological progress in the area of fuel cell lifetime, cost decline, and resolving the cold start issue in countries with cold weather conditions [7, 8].

PEMFC is an electro-chemical device that generates electricity by discharging heat and water. At cold temperature, the produced water freezes on the cathode and anode catalyst layer, which can block the passage of oxygen and hydrogen [9]. Furthermore, the water absorbed by the membrane can freeze, which slows down the electro-chemical reactions [9]. These phenomena cause a voltage and power drop, leading to the failure of the cold start and acceleration of the PEMFC degradation [10, 11].

In vehicular applications, a number of countries and institutions have imposed specific requirements for the PEMFC’s cold startup. For instance, the US department of energy (DOE) has targeted a swift startup from -20 °C in less than 30 s while consuming a low of 5 MJ of energy in 2020 [12]. Moreover, they have aimed at a successful startup at -20 °C without utilizing an external heating source [12]. The European Union attempts to reach an effective cold startup at -25°C while sustaining a satisfying proton conductivity (over 10 mS.cm-1 at -20°C) [13]. Several studies have also proposed different strategies to improve the cold start quality of the PEMFC. These strategies fall into two groups of "Keep warm" and "Thaw at start" [14]. The former employs a thermal energy source to keep the PEMFC temperature near to its desirable range during the FCHEV parking time [15-18]. This methodology enhances the cold startup ability by protecting the PEMFC stack from freezing phenomenon. It can be effective for applications where the parking period is short and known a priori. However, when the parking period is long and not known, this method might become expensive concerning the required energy for heating the stack. The methods based on "Thaw at start" do not consume any energy during the parking time. Nevertheless, they demand a specific heat flow during the cold startup to rise the stack temperature from subfreezing condition to its desired operation range [19-32].

The strategies based on "Thaw at start" principle can be grouped into two categories of "Assisted cold start strategy" and "self-cold start strategies" according to the heating source [33, 34]. The first one requires an external heating source, such as heating element, hot air fan, battery, and catalytic burner, to escalate the PEMFC stack temperature during the startup [19-23]. This category of strategies improve the heating time at the cost of decreasing the energy efficiency and increasing the cost of the system [33]. Moreover, the DOE has demanded an effective approach without the use of an external heating source [35]. Therefore, several researchers have turned attention to the development of "self-cold start strategies". These strategies are based on two steps: 1) purging procedure during the shutdown
to reduce water accumulation and decrease ice formation in the cathode catalyst layer; 2) a heating process during the start-up based on a self-heating solution. The methods based on self-heating employ the heat produced by the exothermic reaction and are divided into three types: galvanostatic startup (fixing the PEMFC current to the best current regarding the production of the heat inside PEMFC), potentiostatic startup (setting the PEMFC voltage to the best voltage which maximizes the exothermic heat production), and the maximum power mode startup.

Hishinuma et al. [30] and Lin et al. [27] propose a self-cold start strategy, based on purge procedure at shutdown and galvanostatic mode at start-up. This strategy is effective for a cold start from -5°C but ineffective for lower temperatures. In order to increase the heat flux supplied by the galvanostatic solution and to provide a self-cold start from -20°C, Guo et al. [26] introduce an O2/H2 mixture on the anode side during the cold start. The majority of studies regarding the self-cold start strategy propose purging the PEMFC after shutdown and considering the potentiostatic mode at startup [24, 25, 28, 36]. This strategy is shown to be effective at cold start from -20°C. Qing et al. [37] propose a self-cold start strategy based on the PEMFC maximum power mode, and compare it with both potentiostatic and galvanostatic modes for a cold start from -20°C. They demonstrate that the maximum power mode is capable of balancing the heat production and the ice formation, resulting in an enhanced cold start survivability compared to the galvanostatic and potentiostatic solutions [37]. The potentiostatic, galvanostatic, and maximum power modes are efficient in terms of energy consumption, but heating time remains long for automotive applications [33].

One common issue of the above-discussed methods is that they are based on a set of precomputed parameters which are prone to alteration as the PEMFC undergoes performance drifts owing to operating conditions variation (temperature, membrane hydration, etc) and stack degradation. The degradation of a PEMFC is typically defined by the loss of voltage, which directly affects the rated power, through time [38, 39]. Moreover, the extreme ambient conditions, such as subzero temperature, influence the output power of the PEMFC which in turn negatively impact the performance of the mentioned internal heating methods. So far, no PEMFC model has been proved to be able to accurately consider the degradation and the effect of ambient as well as the operating conditions, and it is still a study limitation in the literature. Therefore, some researches have been conducted to track the performance drifts of a PEMFC system in real-time based on extremum seeking strategies. These strategies offer two solutions. The first one mainly depends on the use of hill climbing methods [40, 41]. For instance, in [40], a fractional-order strategy is proposed to find an extremum value of a static nonlinear curve using a gradient optimization process. This strategy applies a periodic perturbation signal to the system input and then revises it toward the targeted objective point. Such methods are interesting due to their straightforward implementation. However, the perturbation signal is not simple to select during the cold startup of the PEMFC as it can increase the heating time and decrease the overall performance. The second solution is based on the use of an online parameter estimator coupled with an optimization algorithm [42, 43]. In this respect, the estimator updates the parameters of the mode online and then the required characteristics, such as maximum power, is extracted from the updated model. In [44], an adaptive cold start strategy is proposed to cope with the PEMFC parameters variation and to improve the cold start performance. However, the utilized model in this study does not take into account the temperature which is an important factor in cold start procedure.

The main objective of this paper is to develop an adaptive self-cold start strategy based on the maximum power mode of the PEMFC. In this regard, firstly, a parametric study is conducted to investigate the influence of PEMFC main parameters (anode pressure, fan speed, purge and micro short-circuit) over cold start performance. Afterwards, the adaptive cold start strategy is developed using a multi-input semi-empirical model of the PEMFC and Kalman filter, and its performance is experimentally validated by using three PEMFCs with different levels of degradation. Compared to other similar works, the main contributions of this paper are as follows:

- Performing a detailed parametric study to find the best operating parameters (anode pressure, fan speed, purge and micro short-circuit) for doing the PEMFC cold startup process (compared to [37, 44]).
- Employing a multi-input semi-empirical PEMFC model along with Kalman filter for accurate online estimation of maximum power in subfreezing condition (compared to [37, 44, 45]).
- Validating the performance of the proposed adaptive cold startup strategy by using three 500-W PEMFCs with different levels of degradation (compared to [37, 44]).

The rest of the paper is organized as follows. Section II details the parametric study. Section III explains the adaptive cold start strategy. Section IV discusses the experimental results, and finally, the conclusion is given in section V.

II. PARAMETRIC STUDY

During a self-cold startup, it is necessary to select the operating parameters in a way to maximize the generation of the heat inside the stack, which has a direct effect on the heating time. This section deals with the analysis of four important system parameters, namely anode pressure, fan velocity, purge, and short-circuit during the cold startup process to clarify the impact of each factor on the heating time. In this respect, the developed experimental set-up for testing the effect of each system’s parameter is explained first. Subsequently, the reason for selecting each parameter is discussed. Finally, the experimental protocol, and the results regarding the influence of parameters on the cold startup are presented.

A. Test bench

The utilized test bench for the purpose of this work is presented in Fig. 1. This test bench includes an H500 Horizon open cathode PEMFC. It is self-humidified with a nominal power of 500 W (rated performance: 22 V@23.5 A), and 36 cells. This PEMFC has a dead-ended anode (DEA) configuration, which is fed by dry hydrogen at a controlled inlet
pressure. A DEA configuration benefits from lower cost, volume, and weight compared to a flow-through system since it does not need a hydrogen recirculation loop and its auxiliary modules, including a pump, humidifier, and water separator [46]. However, there are also some issues regarding the DEA operation. Water accumulates in the anode catalyst layer and gas diffusion layer. Moreover, nitrogen of the air in the cathode side is added to the hydrogen gas in the anode channels by passing through the membrane [47]. Consequently, the anode side needs to be purged recurrently in order to remove accumulated water and nitrogen. At subzero temperature, the accumulated water can also freeze and fail the cold startup. Therefore, purging the residual water after shutdown and during PEMFC operation is necessary. The utilized PEMFC in this study is equipped with a purge valve to evacuate the produced water from the anode side. The purge procedure is divided into two phases. The first phase is related to the activation of the purge valve for 10 ms every 10 s during the operation of the PEMFC. This phase is common for all PEMFCs as it allows the evacuation of water and impurities generated during the normal operation. The second phase is related to winter condition and consists in purging the PEMFC after shutdown for 90 s. Due to the open-cathode nature of the system, the cathode is always exposed to relatively dry ambient air. This configuration requires a system design which can keep the water in the MEA to achieve fast humidification [46]. Therefore, the bipolar plate on the anode side does not feature flow channels but an interdigitated flow field, which is integrated directly into the GDL. Moreover, fans are operated at high speed for a few seconds before each PEMFC shutdown to flush residual water out of cathode. It is also worth noting that both anode and cathode provide micro-porous layers (MPL) between GDL and CL, which avoid flooding phenomena and provide more space for ice formation. The anode and cathode flow field geometry of the studied stack as well as the structure of the diffusion layers have been discussed in details by Strahl et al. [48].

The minimum operating temperature of this PEMFC is around 0°C according to the manufacturer. It is equipped with two cooling fans to control the PEMFC temperature and to supply oxygen to the cathode side. Five thermistors are installed on different sides of the stack to measure the PEMFC temperature. One sensor is located in the center, two sensors are at the top-left and top-right corners, and two sensors are at the bottom-left and bottom-right corners. The stack temperature is equal to the average of all measurements. On the cathode side, the pressure is equal to the atmospheric pressure (1 bar). The pressure difference between the anode and the cathode must not exceed 0.6 bar to avoid membrane damage. PEMFC can also be short-circuited during normal operation, which promotes water production and heat. The generated water and heat generally favor the humidification of the membrane and improve the performance of the PEMFC.

The control of PEMFC auxiliaries, such as purge valve, anode pressure, fan speed, hydrogen valve, and short circuit, is performed through an NI CompactRIO programmable controller 9022. A programmable load with a maximum power of 1200 W is connected to the PEMFC to emulate the behavior of an electric motor. The PEMFC is placed inside a climatic room to simulate winter conditions. The volume of the climatic room is 0.39 m³ (85 cm × 55 cm × 85 cm), oxygen consumption rate is 3.5 L/min, and the amount of oxygen in the climatic room is 81.7 L, which is sufficient for the conducted tests.

\[ S_{O_2} = \frac{n_{in}O_2}{n_{cons}O_2} = 0.21 \left( \frac{P_{air}Q_{air}}{M_{air}} \right) \left( \frac{1}{N_{cell}} \right) \]  

(1)

where \( S_{O_2} \) is the oxygen stoichiometry, \( n_{in}O_2 \) is the amount of oxygen supplied to the PEMFC, \( n_{cons}O_2 \) is the consumed oxygen in the electrochemical reaction, \( P_{air} \) is the air density, \( Q_{air} \) is the air flow, \( M_{air} \) is the air molar mass, \( N_{cell} \) is the number of cells, and \( F \) is the Faraday constant. In (1), the only undefined parameter is the airflow \( (Q_{air}) \), which is obtained using the presented experimental measurements in Fig. 2.

![Fig. 1. The developed experimental setup for the cold startup test.](image-url)
According to (1) and Fig. 2, even at the minimum duty cycle of the fan (34%), a stoichiometry of around 33 is reached at the rated performance of the PEMFC (22V@23.5A).

- The micro-short circuit: The micro short-circuit provides a current peak for 10 ms every 10s in the utilized PEMFC stack. Activating the micro-short circuit leads to the increase of heat and water generation inside the PEMFC, which can improve the PEMFC cold startup. In normal operation, it is recommended to activate the micro-short circuit to enhance the performance of the stack in applications where the stack is not frequently used. However, the influence of this parameter on cold start performance has not been considered. Herein, the cold start performance is inspected with and without performing the micro short-circuit.

- Purging after shutdown: Literature study shows that purging the PEMFC after shut down reduces ice formation on the cathode which in turn improves the PEMFC cold start [52, 53]. In this regard, the influence of purging on the heating time is of interest in this study. To justify the effect of purging, the cold startup performance is investigated with and without purging the PEMFC after shutdown.

- Anode pressure: The performed analysis of the anode pressure effect on the cell potential in [54] shows that the increase of hydrogen partial pressure to a certain range leads to the reduction of concentration loss which increases the cell potential. In fact, rising the pressure level in the anode side leads to the increase of hydrogen permeation through the membrane which promotes the reaction at the cathode and resulting in the increase of voltage and temperature. However, further rise of the hydrogen partial pressure results in lower membrane hydration owing to the reduction of water vapor diffusivity. In this study, the cold startup is tested for two anode pressure levels of 0.45 bar and 0.6 bar. 0.45 bar corresponds to the minimum pressure recommended by the manufacturer, and 0.6 bar is the maximum authorized pressure.

With respect to the discussed matters, two levels are considered for each parameter in this work. Therefore, the number of possibilities comes to 16 for four parameters with two levels. More levels for each factor could be selected. However, the number of possibilities would become very high. Moreover, heuristically, the selection of more levels is not required considering the aforementioned explanation for each parameter. The $2^k$ model (k factors with 2 levels each) is the most widely used model as it allows several factors to be inspected at no extra cost [55]. Moreover, the experimental tests show that two levels are adequate for adjusting the parameters.

C. Experimental protocol

An experimental protocol has been developed to determine the parameters that minimize the heating time. In this work, heating time refers to the required time for the evolution of the PEMFC temperature from -20 °C to the stack minimum operating temperature (0 °C). The experimental plan is composed of 16 tests representing all possible combinations of the discussed parameters. Each test has three or four steps:

1. Normal operation of PEMFC at the current level of 10 A and an ambient temperature of 20 °C for 15 minutes.
2. After 15 minutes of normal operation, the PEMFC is placed in the chamber of -20 °C for 8 hours. The 8-hour conditioning time corresponds to the required time to stabilize the temperature of the PEMFC at -20 °C.
3. At the beginning of the conditioning phase, the PEMFC is still hot and most of the produced water is in a gaseous state. When the PEMFC cools down and most of the water turns to a liquid state (around 5°C), the purge valve is activated for 90 seconds to evacuate the residual water. Purging temperature and duration are selected based on [33, 56-58]. This step presents "Purging at shutdown" factor.
4. After 8 hours of conditioning at -20 °C, the PEMFC cold startup is performed by asking a constant voltage of 15V. The constant voltage mode is chosen to keep the same electrical parameters for all the tests. The constant voltage value (15V) is in fact the rated voltage of the tested PEMFC at -20°C. The heating time is reported for each test. Each test has been repeated twice to ensure the accuracy of the results.

D. Results analysis

The factorial experiment is used to analyze the impact of the selected system parameters on the heating time. This experiment clarifies the effect of each parameter and the effect of interactions between parameters on the heating time. Minitab software is used for the design and analysis of the factorial experiment. The result of the factorial experiment is presented by a Pareto chart of effects in Fig. 3. The Pareto chart shows the absolute values of the normalized effects, ordered from the highest to the lowest. This diagram also includes a reference line which determines the statistically significant parameters. In Fig. 3, the bars that intersect the reference line are statistically important. That means the bars representing the "fan speed" (A) "purging at shutdown" (B), "micro short-circuit" (C) and "anode pressure" (D) intersect the reference line at 1.4 and have substantial impact on the heating time. Since the Pareto plot shows the absolute value of the effects, the level of parameters to increase or decrease the heating time cannot be determined.
In this respect, the normal probability plot of the parameters’ effect is presented in Fig. 4 to illustrate the importance and direction of effects. The normal probability plot indicates the position of the normalized effects with respect to the red line, which represents the case where all the effects would be 0. The positive effects increase the response when the factor state changes from the low level to the high level. Negative effects decrease response when factor state changes from low level to the high level. Effects farthest from 0 are statistically more significant. Fig. 4 confirms that the effects of A, B, C and D are statistically substantial. Concerning the direction of the effects, the parameters A and C have a positive normalized effect. As these parameters go from the low level to the high level, the heating time increases. The parameters B and D have negative normalized effects. When these two factors go from the low level to the high level, the heating time decreases.

To summarize, by decreasing the fan speed from high level to low level, the heating time considerably decreases (179 s). Therefore, during PEMFC cold start, the fan speed should be fixed at the minimum level which provides enough air to operate the PEMFC in the whole power range. Regarding the micro short-circuit, its activation increases the heating time (8 s). This increase is due to the fact that the water produced by micro short-circuit freezes at low temperature and affects the cold start performance. On the other hand, purging the PEMFC during shutdown reduces the heating time (11 s). Many studies have shown the effectiveness of the purge procedure in cold startup as it prevents the freezing of the residual water [59-62]. Concerning the anode pressure, the Pareto diagram shows that the pressure rise from 0.45 bar to 0.6 bar decreases the heating time (7 s). Therefore, the anode pressure should be set at its maximum authorized level. Table I shows the most suitable parameters for developing a cold startup strategy. These parameters are used in the adaptive cold startup strategy discussed in the next section.

### Table I

<table>
<thead>
<tr>
<th>Parameters</th>
<th>Value or state</th>
</tr>
</thead>
<tbody>
<tr>
<td>Fan velocity</td>
<td>Minimum operational fan velocity</td>
</tr>
<tr>
<td>Micro short-circuit (C)</td>
<td>Disabled</td>
</tr>
<tr>
<td>Purge after Shutdown (P)</td>
<td>Enabled</td>
</tr>
<tr>
<td>Pressure (S)</td>
<td>The maximum authorized pressure</td>
</tr>
</tbody>
</table>

### III. ADAPTIVE COLD START STRATEGY

#### A. Maximum power mode

During the cold startup, it is important to maximize the generated exothermic heat inside the PEMFC to rapidly increase the stack temperature. Fig. 5 shows that as the drawn current from the PEMFC increases, more heat is generated by the exothermic reaction. However, reaching very high current level (low voltage level) can also result in the failure of the PEMFC cold start [24]. In fact, at high current levels, the concentration of oxygen and hydrogen decreases which leads to gases’ partial pressure drop inside the PEMFC [63, 64]. This phenomenon causes a drop in cell voltage commonly known as concentration loss [9]. The PEMFC stack has a current threshold (corresponding to the maximum power) and exceeding this limit results in an increase in concentration loss and PEMFC degradation [33]. Therefore, it is suggested to operate the PEMFC at its maximum power ($P_{max}$) during cold startup. This mode of operation does not require the external heating sources and maximizes the residual heat and electrical power while avoiding concentration loss and degradation of the stack. The maximum power mode is one of the most effective self-cold start strategies because it can better balance the heat generation and ice formation, leading to improved cold start survivability [37, 44]. Nevertheless, it cannot be assumed that the PEMFC maximum power is constant during a cold startup since it is strongly influenced by degradation and the variation of operating conditions, such as pressure, temperature, and so forth [45, 65-67]. Moreover, some phenomena, including the increase of concentration loss due to the ice formation in the cathode catalyst and the reduction of ohmic and concentration loss due to the generated heat [33], affect the magnitude and location of the maximum power in a PEMFC. Therefore, it is important to identify the PEMFC maximum power online during the cold startup.
In this work, an adaptive process is developed to identify the PEMFC maximum power depending on PEMFC operating conditions and health state. This adaptive process is based on a recursive identification algorithm that updates the parameters of a PEMFC semi-empirical model in order to plot the PEMFC power curve in real time. The maximum power point is updated in real time using a maximum power point tracker algorithm, and the current corresponding to the maximum power is imposed to the PEMFC during the cold start. The flow chart of the global cold start strategy is shown in Fig. 6.

**B. Purge procedure at shutdown**

In automotive applications, purge duration is very important because additional energy is consumed during the purge process and must be controlled to prevent the membrane dehydration [68]. Several studies have shown that a short purge time does not eliminate enough residual water which in turn increases the freezing potential inside the stack [69]. On the other hand, a long purge time increases the membrane resistance, which limits the PEMFC current and thus reduces the produced heat [70]. Based on literature, it is confirmed that the purge duration of 90-120 s is suitable for FCHEVs [71, 72]. In this respect, the duration of purge is set to 90 s in the adaptive cold start strategy of this paper.

The PEMFC temperature is also crucial for the effectiveness of purge procedure [56, 73]. After shutdown, the PEMFC is still warm and the produced water inside the PEMFC is in gaseous state. Therefore, it is preferable to wait until the PEMFC temperature decreases and water state changes from the gaseous to liquid. Once the water is in a liquid state, it must be purged before the PEMFC temperature reaches the freezing temperature. Based on [56], the PEMFC purge temperature is set to 5°C.

**C. Adaptive algorithm for maximum power identification**

The heating procedure is based on an adaptive PEMFC current control to maximize the produced exothermic heat during cold start. This process is performed in 3 steps, shown in Fig. 7, as follows:

1. Updating the parameters of a PEMFC model using an online identification algorithm to plot the present state PEMFC power curve.
2. Finding the operating current level corresponding to the PEMFC’s maximum power. This operating current is updated every 100 ms.
3. Applying the identified current level to the PEMFC stack.

It should be noted that as soon as the PEMFC temperature reaches 0°C, the control process will be stopped.
Fuel cell semi-empirical model

The choice of the semi-empirical model is crucial for the identification process because the model structure greatly influences the accuracy of the estimation. In [45], a comparative study of different semi-empirical models was performed and concluded that the model proposed by Amphlett et al. [74] is adequately precise for reproducing the PEMFC electrochemical behavior. Amphlett model estimates the PEMFC output voltage by taking several inputs (current, pressure, and temperature) into account. This model can be presented by the following equations.

\[ V_{fc} = N V_{cell} \]  
\[ V_{cell} = E_{Nernst} + \eta_{act} + \eta_{ohmic} + \eta_{con} \]  
\[ E_{Nernst} = 1.229 - 0.85 \times 10^{-3} (T_f - 298.15) + 4.3085 \times 10^{-5} T_f \ln(P_{H2}) + 0.5 \ln(P_{O2}) \]  
\[ \eta_{act} = \xi_1 + \xi_2 T_f + \xi_3 T_f \log(CO_2) + \xi_4 T_f \log(i_{fc}) \]  
\[ CO_2 = \frac{5.08 \times 10^6 \exp(-498/T_f)}{P_{O2}} \]  
\[ \eta_{ohmic} = -i_{fc} R_{internal} = -i_{fc} (\mu_1 + \mu_2 T_f + \mu_3 i_{fc}) \]  
\[ \eta_{con} = B \log \left(1 - \frac{J_{fc}}{J_{max}}\right) \]

Where, \( V_{fc} \) is the PEMFC output voltage (V), \( V_{cell} \) is the cell output voltage (V), \( N \) is the number of cells, \( E_{Nernst} \) is the reversible cell potential (V) [75], \( \eta_{act} \) is the activation loss (V), \( \eta_{ohmic} \) is the ohmic loss (V), \( \eta_{con} \) is the concentration loss (V), \( T \) is the stack temperature (K), \( P_{O2} \) is the cathode oxygen partial pressure (N m\(^{-2}\)), \( P_{H2} \) is the anode hydrogen partial pressure (N m\(^{-2}\)), \( E_s \) (n=1.4) are the semi-empirical coefficients based on electrochemistry, fluid mechanics and thermodynamics [76], \( CO_2 \) is the oxygen concentration (mol cm\(^{-3}\)), \( i_{fc} \) is the PEMFC operating current (A), \( R_{internal} \) is the internal resistor (\( \Omega \)), \( \mu_n \) (n=1…3) are the parametric coefficients, \( B \) is a parametric coefficient (V), \( J \) is the actual current density (A cm\(^{-2}\)), and \( J_{max} \) is the maximum current density (A cm\(^{-2}\)).

As explained in [77, 78], assuming that the PEMFC is a perfect isolated system, the entropy change becomes zero and the Gibbs free energy will be equal to the variations of enthalpy (\( \Delta H_f \)). In this case, the maximum voltage (\( V_{max} \)) given by the overall conversion of enthalpy into electrical energy is obtained by:

\[ V_{max} = -\frac{\Delta H_f}{n_e F} \]

where \( F \) is the Faraday constant (96485 C mol\(^{-1}\)), \( n_e \) is the number of electrons which is two in case of hydrogen in the overall PEMFC reaction (\( H_2 + \frac{1}{2}O_2 \rightarrow H_2O \)), \( \Delta H_f = -285.84 \) kj mol\(^{-1}\) if the water product is in liquid form, and \( \Delta H_f = -241.83 \) kj mol\(^{-1}\) if the water product is in vapor form. The higher value for \( \Delta H_f \) has been utilized in this study. It is called the higher heating value (HHV).

Under the condition \( i_{fc}(t) > 0 \), the delivered voltage by the cell (\( V_{cell} \)) is lower than the standard potential. Considering the fact that it is impossible to reach \( V_{max} \) in a PEMFC as there is always heat loss in form of entropy in the system which is calculated as follows.

\[ P_{loss} = i_{fc} (V_{max} - V_{cell}) N \]

Recursive identification algorithm

As previously described, the parameters of the PEMFC model are time-varying since its performance is influenced by degradation phenomenon and variation of operating conditions. In this respect, Kalman filter is utilized in this study to estimate the 8 parameters of the Amphlett model \( (\xi_1, \xi_2, \xi_3, \xi_4, \mu_1, \mu_2, \mu_3, B) \) online. The choice of the recursive identification algorithm mainly depends on the structure of the estimated parameters (linear or nonlinear). Amphlett model emulate the non-linear behavior of a PEMFC stack, but the targeted parameters for estimation present a linear structure. In [45], a comparative study of different recursive identification algorithms showed the effectiveness of Kalman filter to estimate the linear parameters of Amphlett model with the lowest estimation error. In this regard, Kalman filter is employed to update in real-time the unknown parameters of the PEMFC model. The updated model is then used to plot the real-time power curve to find the PEMFC maximum power point during the cold startup. The details about the structure and initialization of Kalman filter for online estimation of the PEMFC model are comprehensively discussed in our previous work [45].

IV. EXPERIMENTAL VALIDATION OF THE GLOBAL ADAPTIVE STRATEGY

A. Experimental procedure

In order to investigate the performance of the proposed adaptive cold startup strategy, three H500 Horizon PEMFCs with different degradation levels are utilized in this paper. These PEMFCs have already been utilized in different projects and conditions. Therefore, some of them have been under operation more than the others. Since the exact age of each fuel cell in terms of hour is not available, their present-day polarization and power curves in normal ambient temperature are presented in Fig. 8 to classify them in terms of degradation. The original polarization curve (i.e. initial polarization curve of a brand-new Horizon 500 W PEMFC) is also shown as the reference. It is well-known that as the PEMFC gets aged, its rated power decreases. Therefore, the relative aging level of each utilized PEMFC is assumed based on the experimentally obtained power curves.
Hereinafter in this manuscript, these PEMFCs are referenced by their rated powers at ambient temperature: 300 W, 400 W and 500 W. The 500-W PEMFC is a brand-new one. However, the other two PEMFCs have been degraded through time.

The three PEMFCs are purged according to the previously-explained procedure. After conditioning at -20 °C for 8 h, the adaptive algorithm is lunched to start the PEMFCs from -20 °C. In the first step, Kalman filter receives the measured current, voltage, and temperature data from the PEMFC via Ethernet communication every 100 ms. In the second step, it estimates the parameters of the Amphlett model. Afterwards, the updated model is used to plot the power curve. Finally, the operating current corresponding to the maximum power is identified and sent to the programmable load via USB communication. The process is done in real-time until PEMFC temperature reach 0°C.

B. Experimental results

In this section, the performance of the suggested adaptive cold start strategy is scrutinized from different perspectives. Initially, the evolution of the main operating parameters (current, voltage, and temperature) as well as the generated electrical and thermal power are discussed for each PEMFC during a cold startup test from -20 °C. Subsequently, the reached heating time and consumed energy of each case study are compared. Moreover, to illuminate the impact of initial temperature on the consumed energy and heating time during cold startup, four cold startup tests with different initial temperatures are conducted over the 400-W PEMFC. This section comes to an end by verifying the performance of the online identification method which has a significant role in the performance of the proposed cold startup method.

Fig. 9 demonstrates the evolution of the current, voltage, and temperature during the performed cold startup test for the three PEMFCs. Fig. 9a shows that in all the PEMFCs, the current increases during the first seconds, followed by a rapid drop. After that it increases smoothly until becoming almost stable. In fact, the algorithm increases the current during the first few seconds due to the membrane hydration, which lowers the ohmic resistance and enhances the reaction kinetics, and the following decrease is mainly caused by the ice formation [37]. Owing to the sharp voltage drop, the algorithm decreases the PEMFC current to adapt to the state of the frozen PEMFC and avoid the cold start failure. As the current has a rising tendency, the stack temperature is almost on the rise during the whole test (Fig. 9c).

Fig. 10 shows the electrical and thermal power trends. The thermal power has been calculated based on (9) and (10). According to Fig. 10a, from 0 to almost 6 s, the electrical power has a sharp rise followed by a drop. It then gradually increases through time for each PEMFC. The intensity of this increase is different in each PEMFC as they have different degradation levels. Fig. 10b illustrates that the thermal power has also experienced a similar trend to the electrical power as they both depend on the drawn current from the stack. In fact, the produced heat, during the first seconds, raises the PEMFC temperature which improves the PEMFC performance and therefore allows the algorithm to gradually increases the PEMFC current.
Table II presents the required current to reach the maximum power and the corresponding thermal power of each utilized PEMFC stack at two different temperatures. An important point that needs to be underlined here is the influence of temperature variation and PEMFC degradation on the maximum power point values. According to this table, the 500-W PEMFC reaches higher current levels than the other ones during cold startup. The PEMFC current level, which significantly changes during the cold start process, directly reflects the rate of water and heat production inside the stack. Table II also indicates that the 500-W PEMFC produces more heat compared to other two PEMFCs during the cold start which makes this PEMFC faster in terms of reaching the temperature level of 0°C. Therefore, this analysis justifies the rapid temperature rise of the 500-W PEMFC from -20°C to 0°C in 43 s, compared to 51 s for the 400-W PEMFC and 68 s for the 300-W PEMFC as shown in Fig. 11. From Fig. 11, the effect of degradation on the performance of the proposed strategy can be realized. The degradation of the PEMFC causes a rise on heating time because the maximum power as well as the produced thermal power decrease when the PEMFC is degraded. The heating time and the PEMFC power are different for the three PEMFCs during cold start, which affects the energy consumption. It is important to note that the consumed energy for the purge procedure is included in the presented energy consumption of the adaptive cold start strategy in Fig. 11. The experimental validation of the proposed strategy was repeated 20 times for each PEMFC with a standard deviation of 5 s regarding the heating time and 54 J cm⁻² concerning the energy consumption.

To confirm the accuracy of the online parameter identification process, the physical relevance of the estimated PEMFC resistance is studied for the 400-W PEMFC. Table III compares the estimated and the calculated membrane resistance at two different temperatures. The estimated membrane resistance ($R_{\text{internal}}$) by the Amphlett model depends on the current, temperature, and the three empirical parameters ($R_{\text{internal}} = \mu_1 + \mu_2 T_{fc} + \mu_3 i_{fc}$). The calculated membrane resistance is deducted from the slope of the linear region of the measured polarization curve at the given temperature. This
slope is approximately equal to the PEMFC internal resistance. From Table III, the calculated values of $R_{\text{internal}}$ are in agreement with the estimated values. Fig. 13 also shows the evolution of the estimated resistance by the online PEMFC model for the 400-W PEMFC. As is seen in this figure, $R_{\text{internal}}$ gradually decreases during the cold startup process. This is due to the fact that $R_{\text{internal}}$ is dependent on the stack temperature, and therefore, the rise of temperature results in the hydration of the membrane.

<table>
<thead>
<tr>
<th>Temperature</th>
<th>Estimated resistance</th>
<th>Calculated resistance</th>
</tr>
</thead>
<tbody>
<tr>
<td>-15°C</td>
<td>0.315 Ω</td>
<td>0.265 Ω</td>
</tr>
<tr>
<td>1°C</td>
<td>0.273 Ω</td>
<td>0.265 Ω</td>
</tr>
</tbody>
</table>

Fig. 13. Evolution of the estimated resistance of the Amphlett model during an adaptive cold startup test with the 400-W PEMFC from -20°C.

V. CONCLUSION

This paper presents an adaptive cold startup strategy for a PEMFC based on the prediction of the PEMFC maximum power point during the cold startup. The core of this strategy is a purge procedure before shutdown and a PEMFC maximum power point tracking algorithm. The maximum power point is identified using an online PEMFC model composed of a multi-input semi-empirical model and a Kalman filter. From the predicted power curve, the current corresponding to the maximum power is extracted and imposed to the PEMFC during the cold start. The proposed strategy is experimentally validated for three 500-W PEMFCs with different degradation levels. In order to improve the performance of the cold startup strategy, a parametric study is also performed to tune the operating parameters of the PEMFC (fan speed, anode pressure, purge and micro short-circuit) for the cold startup test. The conducted parametric study reveals that the fan speed must be set to the minimum level to keep the produced heat inside the stack. Moreover, it shows that higher anode pressure is favorable as it decreases the heating time. Regarding the purge procedure after shutdown, it is confirmed that it is effective in improving cold start performance. However, the PEMFC short circuit during cold start shows a negative impact on the cold start performance and should be deactivated.

The experimental results show that the proposed adaptive strategy is perfectly functional for a cold startup from -20°C. It is capable of coping with the PEMFC parameters variation during cold startup and keeps the current density at high levels, resulting in the PEMFC performance improvement caused by membrane hydration and temperature increase. The adaptive strategy is economical in terms of energy consumption because it avoids external heating by maximizing the PEMFC internal heat and avoiding operating the PEMFC at low voltage levels.

The outcomes of this paper put forward the following directions for future studies:

- The proposed adaptive strategy has been experimentally validated for three 500-W PEMFCs. The latest DOE target for the cold startup of the FCHEV is an 80-kW PEMFC at -20 in less than 30s and less than 5 MJ of energy. Therefore, the adaptive strategy of this work should be scaled up in future.
- The impact of the suggested adaptive strategy on the PEMFC lifetime should be studied.
- The operating temperature of the proposed strategy is limited to -20°C. For lower operating temperatures, more focus should be given to the development of an online ice estimation algorithm combined with a hybrid cold start solution. This algorithm should be based on an adaptive maximum cold start mode, as the main cold start solution, and an external heating method used when cathode catalyst layer is blocked by ice.
- In this article, an open-cathode PEMFC is used because it benefits from a simpler and faster experimental implementation. Unlike the typical liquid-cooled automotive PEM fuel cells, an open-cathode PEMFC system requires no compressor, humidifier, hydrogen recirculation pump or water coolant loop, which dramatically simplifies system design. Therefore, more concrete studies are required in future to focus on experimental performance evaluation of the proposed strategy in sub-freezing conditions with a water-cooled PEMFC.

ACKNOWLEDGEMENTS

This work was supported in part by the Bureau de l’efficacité et de l’innovation énergétique, Ministère des Ressources Naturelles et de la Faune du Québec, the Natural Sciences and Engineering Research Council of Canada (NSERC), the Fonds de recherche du Québec–Nature et technologies (FRQNT), and Canada Research Chairs program.

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