

# Efficiency Improvement of PEM Fuel Cells via Humidity Control

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**Abstract:** PEM (Proton Exchange Membrane) fuel cell is a promising renewable energy source to a wide range of applications for its clean products and high power density. However, controlling its humidity is a challenging problem due to the interdependence of several phenomena contributing in membrane's water content. This work deals with efficiency improvement of PEM fuel cells via humidity control. An innovative strategy of control based on the model of Ref. [1] is proposed. It consists on regulating gas humidification rates according to the power demand so that to minimize power losses. The proposed control takes into consideration constraints related to humidification in order to avoid dry out or flooding of the membrane. Simulations results show that time-phasing between hydrogen and oxygen humidification rates plays an important role in minimizing power losses. The proposed control shows significant improvement in the fuel cell's efficiency up to 20%.

**Key words:** PEM fuel cell, mathematical modeling, humidity control, efficiency improvement.

## Notations

$A$	Area ( $m^2$ )	$\eta$	Voltage loss (Volt)
$a_w$	Water activity	$\lambda$	Water content
$C$	Concentration ( $mol/m^3$ )	<b>Subscripts</b>	
$C_{dl}$	Double layer capacitance ( $F$ )	$a, an$	Anode
$C_{geom}$	Geometrical capacitance ( $F$ )	$act$	Activation
$D$	Diffusion coefficient ( $m^2/s$ )	$bd$	Back diffusion
$E$	Nernst voltage ( $V$ )	$c, ca$	Cathode
$F$	Faraday constant ( $C/mol$ )	$conc$	Concentration
$i$	Current ( $A$ )	$e$	Electrode
$j$	Current density ( $A/cm^2$ )	$H^+$	Hydrogen protons
$j_o$	Exchange current density ( $A/cm^2$ )	$H_2$	Hydrogen
$j_l$	Limit current density ( $A/cm^2$ )	$H_2O$	Water
$J$	Gas diffusion flux ( $mol/s$ )	$in$	Input
$L$	Thickness ( $m$ )	$M, m$	Membrane
$P$	Pressure ( $Pa$ )	$o$	Ohmic
$R$	Resistance ( $\Omega$ )	$out$	Output
$R_g$	Gas constant ( $J/mole.K$ )	$O_2$	Oxygen
$T$	Temperature ( $K$ )	$t$	Total
$V$	Voltage (Volt)		
$X$	Mole fraction		

## Greek-Letters

$\alpha$  Charge transfer coefficient

## Superscripts

$BP$  Bipolar plate zone  
 $sat$  Saturation  
 $TC$  Triple contact zone

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## 1. Introduction

Fuel cell technology faces many obstacles retarding its marketing such as production cost and lifetime limitation. Among real obstacles is the control of physical phenomena occurring inside the cell in high dynamics of power demand. Particularly, controlling internal humidity of fuel cells is one of the most challenging problems. In fact, several interdependent phenomena contribute in membrane's water content. Placing a humidity sensor is unfeasible since the membrane is extremely thin and the sensor will block gas diffusion. Hence, a control-oriented mathematical modeling of water transport is needed to improve fuel cell's efficiency and lifetime.

Strong dependence and coupling between internal phenomena are the main cause of mathematical modeling limitation in literature. To avoid complexity, most of works focus on studying each phenomenon separately from the others. The existing models could be classified into three main categories:

(1) Static models describing the fuel cell's polarization curve [2-9]. These models are suitable for stationary applications where the operating point do not change rapidly. However, when it comes to dynamic applications these models are not convenient since they cannot follow the rapid change in power demand.

(2) Dynamic linear models such as state space linear models [10-13]. They result from the linearization of nonlinear models around an operating point. Being simple and easy to control are their main advantages. However, these works do not describe precisely highly nonlinear phenomena occurring inside the fuel cell. Results in Ref. [14] show a considerable error between linear models and experimental tests. Therefore, the fuel cell's control in dynamic applications cannot rely on this type of models.

(3) Dynamic nonlinear models that focus on studying each phenomenon separately from the others. For example, in the works of Refs. [15] and [16]

electrochemical modeling was not coupled with mass balance modeling. Furthermore, they do not consider the effect of temperature and humidity variations on the output voltage and voltage losses. Other works are based on fuzzy logic and artificial neural networks approaches to model the fuel cell [17-24]. Based on learning techniques, these computational models are not appropriate for fuel cells because they are not deterministic: they are black box; that is internal phenomena are not described. Moreover, model's learning requires time which is not convenient for controlling a critical parameter such as membrane's humidity.

Important works dealing with modeling membrane's water content in transient behaviors are the ones given by Refs. [25] and [26]. In their works, Wang studied the water distribution with respect to the space scale in a three-dimensional model as well as the two-phase flow. Fuller and Newman [27] examined experimentally the water transport number in Nafion 117 membrane and presented the relationship between transport number and electro-osmotic coefficient. Okada et al. [28] modeled the water transport and concentration profile at the anode side.

In summary, the above models do not consider all parameters influencing the membrane's humidity, such as: water mole fractions in hydrogen and oxygen, water production at the cathode as well as the phenomena of diffusion and electro-osmosis between anode and cathode. Furthermore, they do not study the dynamic evolution of the humidity and its influence on the fuel cell's efficiency.

This paper is based on the mathematical model developed in our previous work [1] where the above limitations were covered. In that model all parameters affecting membrane's humidity including gas humidification rates were studied. Effect of humidity on electrical outputs and interactions between charge and mass transport along the fuel cell's components were considered. In this work, an innovative strategy of humidity control is proposed. It consists on

regulating gas humidification rates according to the power demand so that to improve fuel cell's efficiency.

## 2. Mathematical Model

In a previous paper we have developed a mathematical model that meet the objectives of the control proposed in this work [1]. The model built in Ref. [1] takes into account interactions between fuel cell's internal phenomena as well as their effects on the membrane's humidity. It also permits to regulate the humidity through humidification of input gas. We recall in Table 1 the input/output variables of the model. The inputs are hydrogen flow rate ( $J_{H_2}^A$ ), water mole fractions in hydrogen ( $X_{H_2O}^A$ ) and oxygen ( $X_{H_2O}^C$ ), as well as gas pressures ( $P_{H_2}^{BP}, P_{O_2}^{BP}, P_{H_2O}^{BP}$ ) at the bipolar plates zone (BP). At outputs we consider the output voltage per cell ( $V_{out}$ ) and the current density ( $j$ ).

We recall also the state variables of the model in the following state-space vector:

$$x = [J_{H_2}^{TC}, J_{O_2}^{TC}, J_{H_2O}^{TC}, V_e, v_1, v_2, v_3, v_4, v_5, v_6]^T$$

where  $v_1$  to  $v_6$  are the intermediate variables related to input derivatives  $\dot{u}_1$  to  $\dot{u}_6$  [1].

The global state-space model is:

$$\dot{x}_1 = \alpha_2 x_1 - \alpha_2 u_1 + \alpha_1 (Ku_4 - x_8);$$

$$\dot{x}_2 = \beta_3 x_2 - \beta_2 u_1 + \beta_1 (Ku_5 - x_9);$$

$$\dot{x}_3 = \gamma_2 x_3 - \gamma_2 u_1 + \gamma_1 (Ku_6 - x_{10});$$

$$\dot{x}_4 = -\frac{a_7 u_1 x_4}{\ln(a_8 u_1 - a_9 u_1^2)} - a_{10} u_1;$$

$$\dot{x}_5 = K^2 u_1 - Kx_5; \quad \dot{x}_6 = K^2 u_2 - Kx_6;$$

$$\dot{x}_7 = K^2 u_3 - Kx_7; \quad \dot{x}_8 = K^2 u_4 - Kx_8;$$

$$\dot{x}_9 = K^2 u_5 - Kx_9; \quad \dot{x}_{10} = K^2 u_6 - Kx_{10};$$

**Table 1 Input/output variables.**

Inputs	Description
$u_1 : J_{H_2}^{BP}$	Hydrogen diffusion flow
$u_2 : X_{H_2O}^A$	Water mole fraction in hydrogen
$u_3 : X_{H_2O}^C$	Water mole fraction in oxygen
$u_4 : P_{H_2}^{BP}$	Hydrogen pressure at the BP zone
$u_5 : P_{O_2}^{BP}$	Oxygen pressure at the BP zone
$u_6 : P_{H_2O}^{BP}$	Water pressure at the BP zone
Outputs	Description
$y_1 : V_{out}$	Output voltage
$y_2 : j$	Current density

$$\begin{aligned}
y_1 = & E^0 + \frac{R_g T}{2F} \ln \left[ \frac{P_{H_2}^{TC} \sqrt{P_{O_2}^{TC}}}{P_{H_2O}^{TC}} \right] + 2x_4 \\
& - 2Fu_1 \left( R_a + R_c + \frac{L_m}{A(a_5 \lambda_m - a_6)} \right); \\
y_2 = & 2Fu_1 \left[ 1 - \frac{L_m C_{geom} a_5}{A(a_5 \lambda_m - a_6) 2} \frac{d\lambda_m}{dt} \right] - 2C_{geom} \dot{x}_4 \\
& - 2FC_{geom} (Ku_1 - x_5) \left[ R_a + R_c + \frac{L_m}{A(a_5 \lambda_m - a_6)} \right] \\
& - \frac{R_g TC_{geom}}{2F} \left[ \frac{\dot{P}_{H_2}^{TC}}{P_{H_2}^{TC}} + \frac{\dot{P}_{O_2}^{TC}}{2P_{O_2}^{TC}} - \frac{\dot{P}_{H_2O}^{TC}}{P_{H_2O}^{TC}} \right].
\end{aligned} \tag{1}$$

where,

$$\begin{aligned}
\lambda_m = & a_1 + N_1 u_2 \left[ \frac{a_2}{2} - \frac{N_1 a_3 u_2}{2} + \frac{N_1^2 a_4 u_2^2}{2} \right] \\
& + N_2 u_3 \left[ \frac{a_2}{2} - \frac{N_2 a_3 u_3}{2} + \frac{N_2^2 a_4 u_3^2}{2} \right]
\end{aligned}$$

$$\begin{aligned}
\frac{d\lambda_m}{dt} = & [N_1 \dot{u}_2 + N_3 u_2] \left[ \frac{a_2}{2} - \frac{N_1 a_3 u_2}{2} + \frac{N_1^2 a_4 u_2^2}{2} \right] \\
& + [N_2 \dot{u}_3 + N_4 u_3] \left[ \frac{a_2}{2} - \frac{N_2 a_3 u_3}{2} + \frac{N_2^2 a_4 u_3^2}{2} \right] \\
& + N_1 u_2 \left[ -\frac{N_1 a_3 \dot{u}_2}{2} - \frac{N_3 a_3 u_2}{2} + N_1^2 a_4 \dot{u}_2 + N_1 N_3 a_4 u_2^2 \right] \\
& + N_2 u_3 \left[ -\frac{N_2 a_3 \dot{u}_3}{2} - \frac{N_4 a_3 u_3}{2} + N_2^2 a_4 \dot{u}_3 + N_2 N_4 a_4 u_3^2 \right]
\end{aligned}$$

$$\alpha_1 = \frac{AD_{H_2}}{R_g TL}; \quad \alpha_2 = \frac{D_{H_2}}{L^2}; \quad \alpha_3 = \frac{R_g TL}{AD_{H_2}};$$

$$\beta_1 = \frac{AD_{O_2}}{R_g TL}; \quad \beta_2 = \frac{\beta_3}{2} = \frac{D_{O_2}}{2L^2}; \quad \beta_4 = \frac{1}{2\beta_1};$$

$$\gamma_1 = \frac{AD_{H_2O}}{R_g TL}; \quad \gamma_2 = \frac{D_{H_2O}}{L^2}; \quad \gamma_3 = \frac{R_g TL}{AD_{H_2O}};$$

$$\dot{u}_1 = Ku_1 - x_5; \quad \dot{u}_2 = Ku_2 - x_6; \quad \dot{u}_3 = Ku_3 - x_7;$$

$$\dot{u}_4 = Ku_4 - x_8; \quad \dot{u}_5 = Ku_5 - x_9; \quad \dot{u}_6 = Ku_6 - x_{10};$$

$$P_{H_2}^{TC} = u_4 - \alpha_3 u_1; \quad P_{O_2}^{TC} = u_5 - \beta_4 u_1;$$

$$P_{H_2O}^{TC} = u_6 - \gamma_3 u_1;$$

$$a_1 = 0.3; \quad a_2 = 10.8; \quad a_3 = 16; \quad a_4 = 14;$$

$$a_5 = 5139 \cdot 10^{-6} \exp[1268(\frac{1}{303} - \frac{1}{T})];$$

$$a_6 = \frac{3260a_5}{5139}; \quad a_7 = \frac{4\alpha F^2}{R_g TC_{dl}}; \quad a_8 = \frac{2F}{i_0};$$

$$a_9 = \frac{4F^2}{i_0 i_l}; \quad a_{10} = \frac{2F}{C_{dl}};$$

$$N_1 = \frac{R_g TL_a u_1}{D_{H_2} P_{T_{an,in}}^{sat}} + \frac{P_{H_2}^{TC}}{P_{T_{an,in}}^{sat}}; \quad N_2 = \frac{R_g TL_c u_1}{2D_{O_2} P_{T_{ca,in}}^{sat}} + \frac{P_{O_2}^{TC}}{P_{T_{ca,in}}^{sat}};$$

$$N_3 = \frac{R_g TL_a \dot{u}_1}{D_{H_2} P_{T_{an,in}}^{sat}} + \frac{\dot{x}_1}{P_{T_{an,in}}^{sat}}; \quad N_4 = \frac{R_g TL_c \dot{u}_1}{2D_{O_2} P_{T_{ca,in}}^{sat}} + \frac{\dot{x}_2}{P_{T_{ca,in}}^{sat}}.$$

Simulation results presented in our previous paper [1] showed that water mole fractions in hydrogen and oxygen have influence on the fuel cell's electrical efficiency. These results are used in the following section to build an appropriate humidity control aiming to minimize power losses inside the cell.

### 3. Humidity Control

We develop in this part a strategy for controlling

the membrane's humidity in view of improving its performance. The proposed strategy is based on regulating the membrane water content  $\lambda_m$  calculated from the state-space model, with respect to  $\lambda_{ref}$  calculated from reference water activities  $a_{w,ref}^A$  and  $a_{w,ref}^C$ . The regulation consists on adjusting water mole fractions in hydrogen  $X_{H_2O}^A$  and oxygen  $X_{H_2O}^C$  according to the power demand. The proposed control takes into consideration constraints related to humidification in order to avoid dry out or flooding of the membrane.

The suggested regulation process is illustrated by the block diagram in Fig. 1. Indeed, the hydrogen flow rate  $u_1$  is considered as free input representing the power demanded by the load. The reference water mole fractions  $u_{2,ref}$  and  $u_{3,ref}$  are computed with respect to  $u_1$ ,  $a_w^A$  and  $a_w^C$ . The commands related to humidification ( $u_2$  and  $u_3$ ), are determined with respect to  $u_{2,ref}$  and  $u_{3,ref}$  plus a correction based on the error calculated between  $\lambda_{ref}$  and  $\lambda_m$ . The latter is computed from the model's state variables, the input  $u_1$  and the output  $y_1$ .

As is widely used in the literature, the control of the humidity inside fuel cells is realized by fixing a constant reference for water activities  $a_w^A$  and  $a_w^C$  regardless of the operating conditions. In fact, maintaining a constant humidity in case of dynamic power demand, such as in automotive applications, is not optimum as we will show in this work. The remainder of this section is divided into two parts: In the first part, we elaborate the control laws that ensure static water activities regardless of the power demand. In the second part, the evolution of the power demand is taken into consideration when building the control laws.

#### 3.1 Control for Static Water Activities

In case of stationary applications, the dynamic of load variation is slow and therefore the power demand can be considered as piecewise constant. Hence, the control objective is to maintain the water activities at

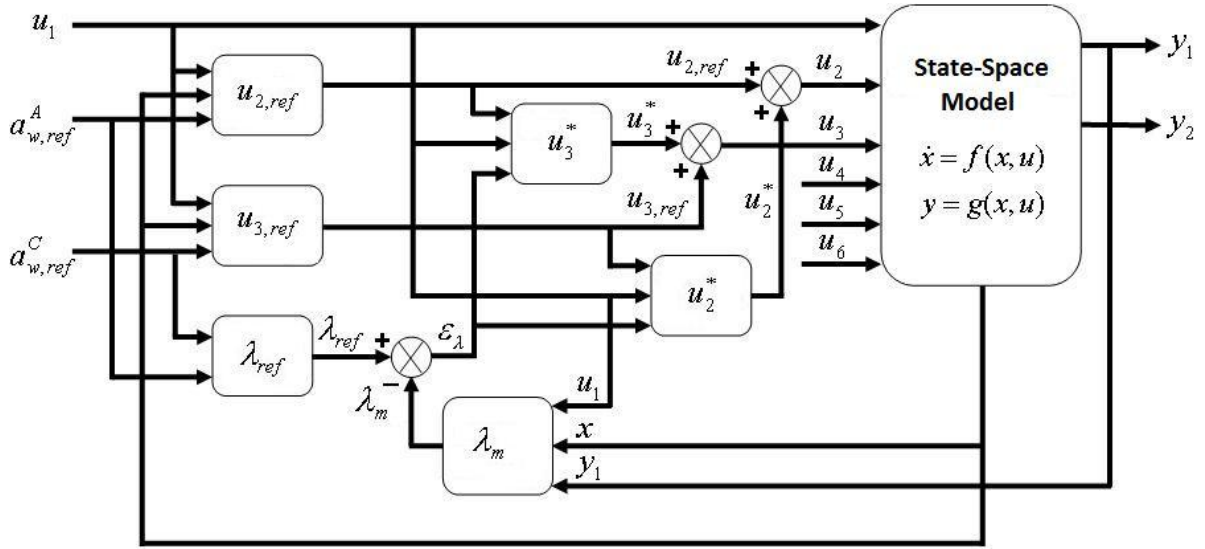


Fig. 1 Block diagram of the regulation process.

anode and cathode sides at convenient levels such that to avoid dry out or flooding of the membrane. For this reason, we aim to have  $a_w^A = a_w^C = 80\%$  and we determine the inputs  $X_{H_2O}^A$  and  $X_{H_2O}^C$  ensuring these activities. We consider a sinusoidal evolution of  $J_{H_2}^A$  between  $J_{H_2\min}^A = \frac{j_0}{nF}$  and  $J_{H_2\max}^A = \frac{j_l}{nF}$  in order to cover the entire cell operating range (between exchange  $j_0$  and limit  $j_l$  current densities).

First, we derive from Eq. (27) in the previous work [1]:

$$\begin{aligned} u_{2,ref} &= \frac{a_{w,ref}^A}{M_1 u_1 + M_2 x_1}; \\ u_{3,ref} &= \frac{a_{w,ref}^A}{M_3 u_1 + M_4 x_2}. \end{aligned} \quad (2)$$

where,

$$\begin{aligned} M_1 &= \frac{R_g T L_a}{D_{H_2} P_{T_{an,in}}^{sat}}; \quad M_2 = \frac{1}{P_{T_{an,in}}^{sat}}; \\ M_3 &= \frac{R_g T L_c}{2 D_{O_2} P_{T_{ca,in}}^{sat}}; \quad M_4 = \frac{1}{P_{T_{ca,in}}^{sat}}; \end{aligned}$$

The reference membrane water content  $\lambda_{ref}$  is then deduced from Eq. (24) in the previous paper [1] as

follows:

$$\lambda_{ref} = \frac{\lambda_{ref}^A + \lambda_{ref}^C}{2}, \quad (3)$$

where,

$$\begin{aligned} \lambda_{ref}^A &= a_1 + a_2 a_{w,ref}^A - a_3 (a_{w,ref}^A)^2 + a_4 (a_{w,ref}^A)^3; \\ \lambda_{ref}^C &= a_1 + a_2 a_{w,ref}^C - a_3 (a_{w,ref}^C)^2 + a_4 (a_{w,ref}^C)^3. \end{aligned}$$

Now,  $\lambda_m$  can be calculated from  $g$  function given by  $y_1 = g(x, u_1, \lambda_m)$  in the state-space model (Eq. (1)):

$$\begin{aligned} \lambda_m &= h(x, u_1, y_1) = \\ &= \frac{1}{a_5} \left[ \frac{2 F L_m u_1}{E + 2 x_4 - y_1 - 2 A F (R_a + R_c) u_1} + a_6 \right]. \end{aligned} \quad (4)$$

We emphasize that  $u_2^*$  and  $u_3^*$  depend on the error between  $\lambda_{ref}$  and  $\lambda_m$ . However,  $\lambda_m$  is a third degree polynomial function (see  $\lambda_m$  expression in the state-space model). Consequently, its solution with respect to  $u_2$  and  $u_3$  is impossible. For this reason, we calculate first  $u_2^*$  by replacing  $u_3$  with  $u_{3,ref}$ , then we state  $u_3^*$  by replacing  $u_2^*$  with  $u_{2,ref}$ . Hence, we have to resolve the following relations separately:

$$\begin{aligned}\epsilon_\lambda &= \lambda_{ref} - \lambda_m = P(x_1, x_2, u_1, u_2^*, u_{3,ref}); \\ \epsilon_\lambda &= \lambda_{ref} - \lambda_m = Q(x_1, x_2, u_1, u_{2,ref}, u_3^*).\end{aligned}\quad (5)$$

The solution of these equations gives one real root and two complex roots. Then, the real root of the  $P$  function gives:

$$u_2^* = \frac{[12C_4H_{1\lambda} + 8C_3^2]^{1/3} + 4C_4(C_3^2 - 3C_2C_4)}{6C_4^2[12C_4H_{1\lambda} + 8C_3^2]^{1/3} + 2C_3}, \quad (6)$$

where,

$$H_{1\lambda} = 18(\lambda_{ref} - \lambda_m)C_4 - 3C_2C_3 - 9C_1C_4 + \sqrt{F_{1\lambda}};$$

$$F_{1\lambda} = (\lambda_{ref} - \lambda_m)(12C_3^2 - 54C_2C_3C_4 - 162C_1C_4^2) + C_2^2 + C_4(4C_2^3 + 18C_1C_2C_3 + 27C_1^2C_4) - C_3^2(4C_1).$$

$$C_1 = \frac{a_2}{2}[M_3u_1 + M_4x_1]u_{3,ref} - \frac{a_3}{2}[M_3u_1 + M_4x_2]^2u_{3,ref}^2 + \frac{a_4}{2}[M_3u_1 + M_4x_2]^3u_{3,ref}^2 + a_1;$$

$$C_2 = \frac{a_2}{2}[M_1u_1 + M_2x_1];$$

$$C_3 = \frac{a_3}{2}[M_1u_1 + M_2x_1]^2;$$

$$C_4 = \frac{a_4}{2}[M_1u_1 + M_2x_1]^3.$$

Similarly, the real root of the  $Q$  function gives:

$$u_3^* = \frac{[12C_8H_{2\lambda} + 8C_7^2]^{1/3} + 4C_8(C_7^2 - 3C_6C_8)}{6C_8^2[12C_8H_{2\lambda} + 8C_7^2]^{1/3} + 2C_7}, \quad (7)$$

where,

$$H_{2\lambda} = 18(\lambda_{ref} - \lambda_m)C_8 - 3C_6C_7 - 9C_5C_8 + \sqrt{F_{2\lambda}};$$

$$F_{2\lambda} = (\lambda_{ref} - \lambda_m)(12C_7^2 - 54C_6C_7C_8 - 162C_5C_8^2) + C_6^2 + C_8(4C_6^3 + 18C_5C_6C_7 + 27C_5^2C_8) - C_7^2(4C_5).$$

$$C_5 = \frac{a_2}{2}[M_1u_1 + M_2x_1]u_{2,ref} - \frac{a_3}{2}[M_1u_1 + M_2x_1]^2u_{2,ref}^2 + \frac{a_4}{2}[M_1u_1 + M_2x_1]^3u_{2,ref}^2 + a_1;$$

$$C_6 = \frac{a_2}{2}[M_3u_1 + M_4x_2];$$

$$C_7 = \frac{a_3}{2}[M_3u_1 + M_4x_2]^2;$$

$$C_8 = \frac{a_4}{2}[M_3u_1 + M_4x_2]^3.$$

Given that, the control laws ensuring constant water activities of 80% are:

$$\begin{aligned}\bar{u}_2 &= u_{2,ref} + u_2^* = \frac{a_{w,ref}^A}{M_1u_1 + M_2x_1} + u_2^*; \\ \bar{u}_3 &= u_{3,ref} + u_3^* = \frac{a_{w,ref}^C}{M_3u_1 + M_4x_2} + u_3^*.\end{aligned}\quad (8)$$

Fig. 2 presents the control inputs  $\bar{u}_2$  and  $\bar{u}_3$  that ensure static water activities in case of sinusoidal power demand represented by  $J_{H_2}^A$ . One can observe that  $\bar{u}_2$  and  $\bar{u}_3$  behave conversely to  $J_{H_2}^A$ . This is due to the fact that we need to humidify less when  $J_{H_2}^A$  increases and vice versa because the water production at the cathode side is proportional to gas flow rates. Fig. 2 shows also that  $a_w^A$ ,  $a_w^C$  and  $\lambda_m$  are constants and at their desired values. Hence, the proposed control laws  $\bar{u}_2$  and  $\bar{u}_3$  provide a good regulation.

Fig. 3 shows that the output current density  $j$  behaves proportionally to the hydrogen flow rate  $u_1$

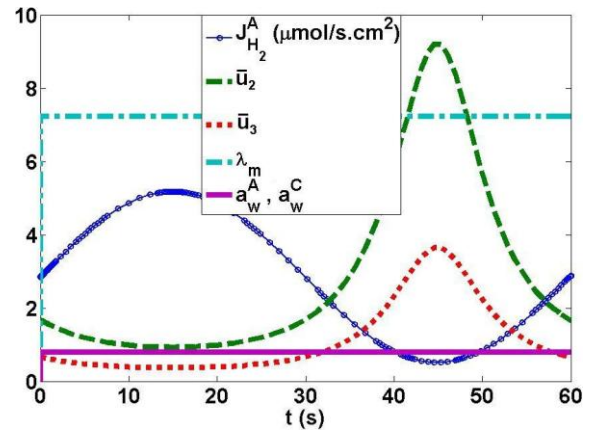


Fig. 2 Static water activities control.

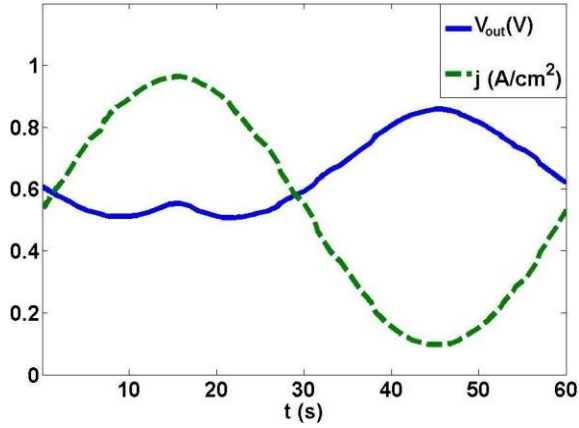


Fig. 3 Responses of electrical outputs for static water activities control.

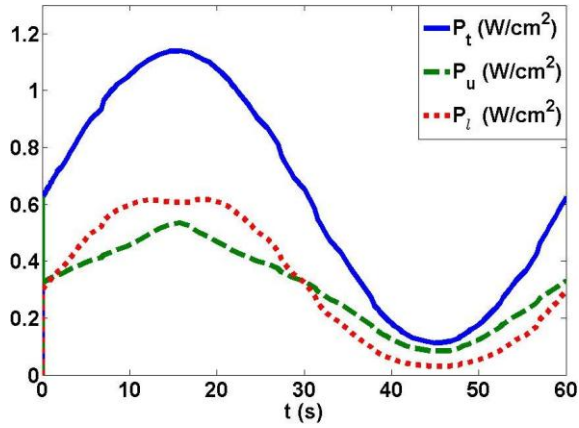


Fig. 4 Responses of powers for static water activities control.

while the output voltage  $V_{out}$  is conversely proportional to it. This is in accord with fuel cell operating principle since voltage losses are proportional to the current. Furthermore, Fig. 4 reveals an overflow of the lost power  $P_l$  with respect to the useful power  $P_u$  in the zone of high power demand. This is due to the high voltage drop caused by the current increase. Therefore, simulation results show that maintaining constant water activities in variable operating power is not optimum for the fuel cell's efficiency.

### 3.2 Control for Dynamic Water Activities

In order to improve fuel cell's efficiency in variable operating power (case of automotive applications), we are going to build in this part a control providing an appropriate dynamical evolution for water activities.

In particular, we propose a sinusoidal variation of  $a_w^A$  and  $a_w^C$  similar to  $J_{H_2}^A$ . The desired variations of  $a_w^A$  and  $a_w^C$  are supposed to be between 60% and 100% to avoid dry out or flooding of the membrane.

Since we consider a sinusoidal power demand,  $u_1$  can be written in the following form:

$$u_1 = k_1 \sin(\omega t) + k_2, \quad (9)$$

with

$$k_1 = \frac{u_{1\max} - u_{1\min}}{2}; \quad k_2 = \frac{u_{1\max} + u_{1\min}}{2}.$$

The objective of the control aims to have reference water activities similar to  $u_1$ . Therefore,

$$a_{w,ref} = M_5 \sin(\omega t) + M_6, \quad (10)$$

with

$$M_5 = \frac{(a_{w,ref})_{\max} - (a_{w,ref})_{\min}}{2};$$

$$M_6 = \frac{(a_{w,ref})_{\max} + (a_{w,ref})_{\min}}{2}.$$

Introducing time-phasing between  $a_{w,ref}^A$  and  $a_{w,ref}^C$ , we can write:

$$a_{w,ref}^A = M_5 \sin(\omega t) + M_6 = k_3 u_1 + k_4; \quad (11)$$

$$a_{w,ref}^C = M_5 \sin(\omega t + \phi) + M_6.$$

with

$$k_1 = \frac{M_5}{k_1};$$

$$k_4 = M_6 - \frac{M_5 k_2}{k_1}.$$

Hence, the appropriate control laws ( $\tilde{u}_2$  and  $\tilde{u}_3$ ) for sinusoidal water activities are calculated according to Eqs. (2) and (11) as follows:

$$\tilde{u}_2 = \frac{k_3 u_1 + k_4}{M_1 u_1 + M_2 x_1} + u_2^*; \quad (12)$$

$$\tilde{u}_3 = \frac{M_5 \sin(\omega t + \phi) + M_6}{M_3 u_1 + M_4 x_2} + u_3^*.$$



Figs. 5 and 6 present the control inputs applied to the model for static and dynamic water activities with different time-phasing. The results in Fig. 7 show an exact regulation of the water activities with respect to their desired variations and time-phasing. In particular, simulations show that static control inputs  $\bar{u}_2$  and  $\bar{u}_3$  results in constant water activities of 80% as desired. Dynamic control inputs  $\tilde{u}_2$  and  $\tilde{u}_3$  results in sinusoidal water activities varying between 60% and 100% as also desired. Furthermore, the proposed control inputs related to the desired time-phasing  $\phi$  ensure an exact regulation of  $a_w^A$  and  $a_w^C$  for  $\phi = 0, \phi = \frac{\pi}{3}, \phi = \frac{\pi}{2}$  and  $\phi = \pi$ .

In Fig. 8 we simulate the evolution of the electrical efficiency with respect to the power demand.

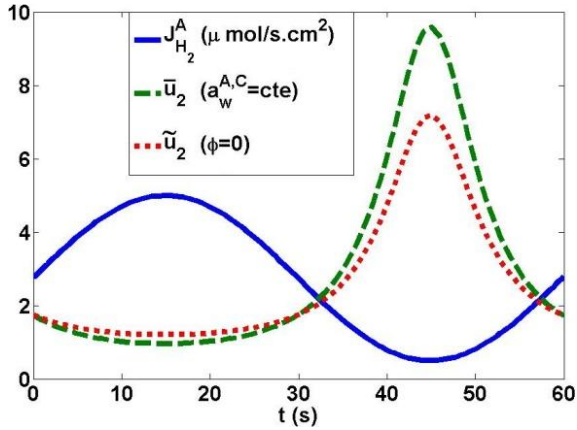


Fig. 5 Static versus dynamic control of water mole fraction in hydrogen.

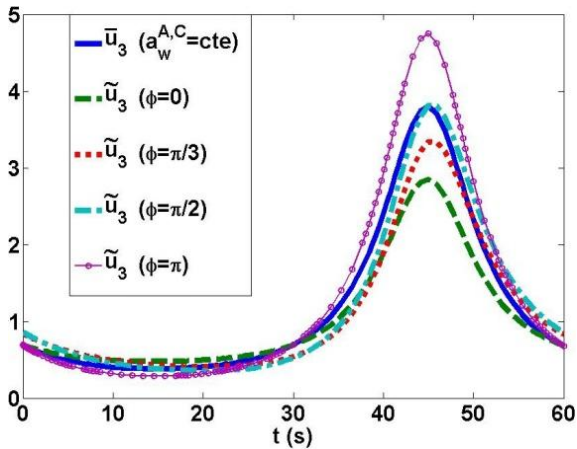


Fig. 6 Static versus dynamic control of water mole fraction in oxygen with different time-phasing.

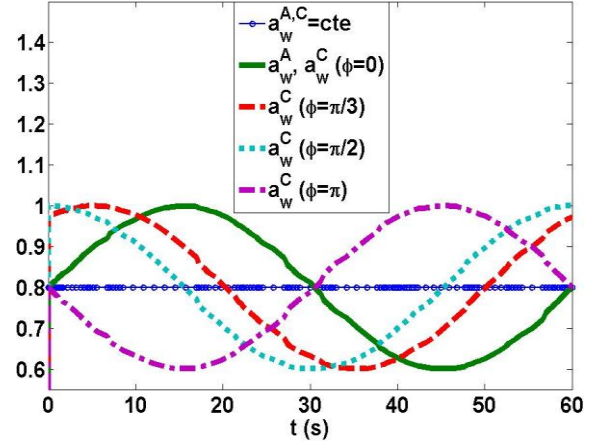


Fig. 7 Anode and cathode water activities for static and dynamic control with different time-phasing.

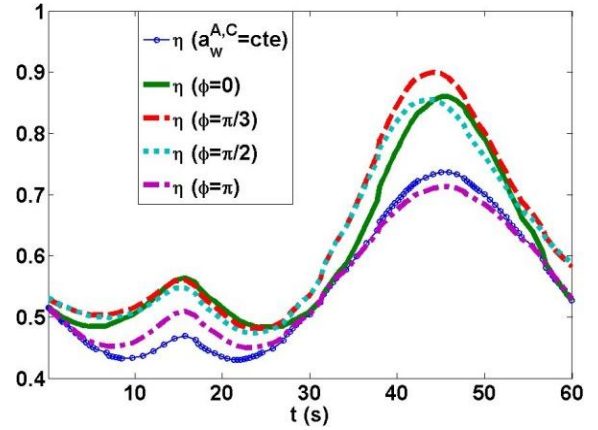


Fig. 8 Electrical efficiency variation with respect to the power demand-comparison between static and dynamic control with different time-phasing.

First of all, we notice that the efficiency is conversely proportional to the power demand. Indeed, for high power demand the current generated by the fuel cell is large and so are the losses. In opposition, when the power demand is low the losses are also low and therefore the efficiency is higher. Second, the simulation shows that dynamic control of water activities ensure better efficiency comparing to static control. In addition, the result proved that the best dynamic efficiency is obtained for a time-phasing of  $\frac{\pi}{3}$  between anode and cathode water activities.

In conclusion, simulation results show that static humidity control do not provide optimum efficiency especially in case of dynamic power demand such as in automotive applications. A better control consists



of dynamic humidity regulation similar to the power demand evolution. Furthermore, time-phasing between anode and cathode water activities plays an important role in the efficiency improvement and its best value is  $\phi = \frac{\pi}{3}$ . It should be noted that this

value is drawn from simulations and do not represent the optimal value. However, it shows that taking into account time-phasing between water activities make an important improvement in fuel cell's efficiency up to 20% as shown in Fig. 8.

#### 4. Conclusion

In this work, an innovative strategy of humidity control in PEM fuel cells is proposed. It consists on regulating gas humidification rates according to the power demand so that to improve fuel cell's dynamic efficiency. The proposed control uses the model of Ref. [1] to calculate the membrane's water content thereby solving the problem of placing a humidity sensor inside the cell. Humidification constraints are taken into account to avoid dry out or flooding of the membrane. Control for static and dynamic water activities are examined. Simulation results show that static humidity regulation do not provide optimum efficiency in case of dynamic power demand. A better regulation consists of dynamic humidity regulation similar to the power demand variation. Furthermore, effect of time-phasing between anode and cathode water activities is studied. Simulations results prove that time-phasing plays an important role in minimizing power losses. Results show a significant improvement in fuel cell's efficiency up to 20% for a time-phasing of  $\frac{\pi}{3}$ .

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