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Article

A New Model for Constant Fuel Utilization and Constant Fuel Flow in Fuel Cells

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Abstract: This paper presents a new model of fuel cells for two different modes of operation: constant fuel utilization control (constant stoichiometry condition) and constant fuel flow control (constant flow rate condition). The model solves the long-standing problem of mixing reversible and irreversible potentials (equilibrium and non-equilibrium states) in the Nernst voltage expression. Specifically, a Nernstian gain term is introduced for the constant fuel utilization condition, and it is shown that the Nernstian gain is an irreversibility in the computation of the output voltage of the fuel cell. A Nernstian loss term accounts for an irreversibility for the constant fuel flow operation. Simulation results are presented. The model has been validated against experimental data from the literature.

Keywords: fuel cell model; constant fuel utilization; constant fuel flow; hydrogen; proton exchange membrane fuel cell (PEMFC); solid oxide fuel cell (SOFC)

1. Introduction

Fuel cells [1–4] convert chemical energy into electrical energy and are fast emerging as one of the major alternatives to fossil-fuel-based sources of energy. Modeling is often an integral part of fuel cell research, spanning areas from new material investigation [5] to system integration [6] to control of the plant [7]. Even though many models of fuel cells and fuel cell systems have already been proposed in the literature, new models continue to be developed, improving upon the state of the art. This paper presents an improved fuel cell model for two operating conditions: constant fuel utilization and constant fuel flow (with the constant fuel flow operation being analyzed in the context of two different modes for the oxidant—the constant oxygen flow mode and the constant oxygen stoichiometry mode). The need for a new model arose because of a recently identified problem [8] in a family of commonly used models in the literature. The problem involves the (erroneous) mixing of reversible and irreversible potentials in the computation of the cell output voltage. The present paper seeks to solve the problem by separating the irreversibilities from the reversible (equilibrium) potential of the fuel cell. The solution is achieved by introducing a novel "Nernstian gain" term and establishing it as an irreversibility for the constant fuel utilization condition. A Nernstian loss term, somewhat akin in spirit to traditional fuel cell losses, is used to analyze the constant fuel flow operation.

The remainder of this paper is organized as follows. A brief survey of previous work on fuel cell modeling is presented in Section 2. Section 3 describes the problem. Section 4 develops a model for the constant fuel utilization mode of operation of the cell. The constant fuel flow mode is modeled in Section 5. Conclusions are drawn in Section 6.

2. A Brief Review of the Literature

Typical classifications of fuel cells are based on the type of electrolyte used in the cell, the operating temperature range, the fuel used, and other factors [9,10]:

Appl. Sci. 2019, 9, 1066 2 of 24

- Direct methanol fuel cell (DFMC), about 60 °C
- Proton exchange membrane (or polymer electrolyte membrane) fuel cell (PEMFC), less than 120 °C, typically about 80 °C
- Alkaline fuel cell (AFC), about 100 °C or less
- Phosphoric acid fuel cell (PAFC), about 150–200 °C
- Molten carbonate fuel cell (MCFC), about 600–700 °C
- Solid oxide fuel cell (SOFC), about 600–1000 °C

Reversible fuel cells [11] represent another class of fuel cells that, in addition to producing electricity from hydrogen and oxygen, can use electricity to produce, by electrolysis, hydrogen and oxygen from water. Microbial fuel cells [12] are yet another category of fuel cells where biological organisms (bacteria, for example) act as the electrocatalyst.

The literature on the modeling of fuel cells is expansive (see, for example, [9,13–21]). The fuel cell is a complex, non-linear system, and not all aspects of the physics, chemistry and engineering of the system are completely understood yet. Given the enormity of the problem, a single model typically focuses on one particular aspect (e.g., liquid water removal from a PEMFC [22] or real-time simulation [23]) of the system.

Fuel cell models are usually characterized by the number of physical dimensions (number of different geometrical axes) they consider. Clearly, this number can vary from zero to three. Zero-dimensional (0-D) models (e.g., [24–26]) employ a concentrated (as opposed to distributed) or lumped representation of the parameters or metrics (e.g., voltage, current, temperature or species partial pressures) under consideration and therefore make more use of mathematical abstraction than the spatial geometry of the cell. In other words, the variation of parameters along/within the cell is ignored, that is, the "local" and "global" values are the same in such models. This type of modeling is particularly useful when the interaction of the cell with other components of the (integrated) system is to be investigated (e.g., the modeling of SOFC-based hybrid power generation systems) [27]. A simple, lumped, dynamic model of an integrated SOFC plant is developed for power system simulation in [28], building upon which a large number of fuel cell models have been proposed in the literature (see, e.g., [6,7,29–55]).

One-dimensional (1-D) models (e.g., [23,34,35,56,57]) take into account the variation of the relevant metrics (electrochemical, physical, or thermal) along only one dimension of the cell, ignoring the other two dimensions. For instance, a 1-D model can compute local changes in current density in axial directions of tubular fuel cells [58]. Zero- and one-dimensional models are often quite effective for control purposes.

Two-dimensional (2-D) models (e.g., [59–61]) consider a two-dimensional cross-section of the cell geometry, ignoring the variations along the third dimension.

Three-dimensional (3-D) models (e.g., [62–67]) use distributed (not lumped) metrics or parameters and are the closest to the physical geometry of the cell, describing spatial properties in each of the three coordinates. Because of their high computational costs, 3-D models are somewhat more esoteric than 2-D or 1-D models; in Reference [68], we find an example of the reduction of a 3-D model to 2-D without significant compromises with the quality of results.

Zero-dimensional or lumped models are also referred to as macro-models [27]. Because models in two or three dimensions look at finer levels of granularity, they are sometimes called micro-models [69]. Length-scales and time-scales are important issues in the modeling of fuel cells; for a review of multi-scale modeling of SOFCs from quantum (sub-atomic) to atomistic to continuum, see Reference [70].

On top of the dimensional categories, fuel cell models can also be broadly classified as:

- mechanistic (theoretical)
- empirical (experimental)
- semi-empirical

Appl. Sci. 2019, 9, 1066 3 of 24

data-driven (machine-learning-based)

Mechanistic models are obtained (derived) theoretically, using theories from, e.g., electrochemistry, thermodynamics, fluid dynamics and engineering; well-known examples include the use of fundamental, phenomenological equations such as the Butler-Volmer equation for voltage, the Stefan-Maxwell equation for diffusion, and the Nernst-Plank equation for species transport [71,72].

Empirical models (e.g., [73,74]), on the other hand, use data obtained from actual experiments on physical fuel cells to propose relationships among the variables of interest, especially relationships that would otherwise be difficult to arrive at from theory alone.

Semi-empirical models (e.g., [61,75]) employ a mix of theory and experimentation. The line between the theoretical and the semi-empirical is sometimes blurred.

Data-driven (or data-oriented) models are a relatively recent development in the field. This approach is capable of creating predictive models for fuel cells and fuel cell systems by using statistical and algorithmic analysis of experimental (or simulated) data. Data-driven models, sometimes referred to as "black-box" models, employ techniques from the machine learning [76] or artificial intelligence [77] paradigm. Examples of data-driven models include Reference [38], where genetic programming [78] is used in a supervised learning setting for static and dynamic (load-following) modeling of SOFCs, and Reference [79] where differential evolution [80] is used to optimize seven parameters for modeling the polarization curve of a PEMFC stack. Given the rather loose connection between theoretical (mechanistic) principles and the data-driven approach, it can be argued that data-driven models verge on the empirical or the semi-empirical.

In Reference [8], a critique of a major class of models was presented. zero-dimensional (lumped) model, the present paper solves the problem discussed in Reference [8].

3. The Problem

The Nernst equation, which is the cornerstone of fuel cell thermodynamics, provides an expression for the reversible thermodynamic potential, also known as the equilibrium voltage or the open-circuit electromotive force (EMF), of the fuel cell [1]:

$$E_{Nernst} = E_0 + \frac{RT}{nF} \ln \left(\frac{\prod_i a_{\text{reactant}_i}^{c_i}}{\prod_j a_{\text{product}_j}^{c_j}} \right)$$
(1)

where E_0 is the reference (standard) EMF at unit activity and atmospheric pressure, i and j are the numbers of reactant and product species, a represents the activity, c_i is the stoichiometric coefficient of species i, R is the universal gas constant, F is Faraday's constant, n is the number of electrons transferred for each molecule of the fuel participating in the reaction, and *T* is the temperature. For a hydrogen-oxygen fuel cell (e.g., solid oxide fuel cell (SOFC) or proton exchange membrane fuel cell (PEMFC)), hydrogen and oxygen are the reactants, and the product is water (or steam). The reference EMF, E_0 , depends on temperature T:

$$E_0 = E_0^0 + (T - T_0) \frac{\Delta s}{nF} \tag{2}$$

where E_0^0 is the standard EMF at temperature T_0 , and Δs is the change in entropy. The activity a of an ideal gas is expressed in terms of its pressure (or partial pressure) p:

$$a_{H_2} = \frac{p_{H_2}}{p^0}, (3)$$

$$a_{H_2} = \frac{p_{H_2}}{p^0},$$
 (3)
 $a_{O_2} = \frac{p_{O_2}}{p^0},$ (4)

Appl. Sci. 2019, 9, 1066 4 of 24

where p^0 is the standard-state pressure (1 atm). At high temperatures, such as 1000 °C (as in solid oxide fuel cells), steam can be assumed to behave as an ideal gas, and therefore

$$a_{H_2O} = \frac{p_{H_2O}}{p^0}. (5)$$

Using $p^0 = 1$ atm, and noting that n = 2 for a hydrogen fuel cell, we have the following version of the Nernst equation for solid oxide fuel cells:

$$E_{Nernst} = E_0 + \frac{RT}{2F} \ln \left(\frac{p_{H_2} \sqrt{p_{O_2}}}{p_{H_2O}} \right). \tag{6}$$

If the fuel cell is operated below 100 °C, so that liquid water is produced (as in proton exchange membrane fuel cells), the activity of water can be taken to be unity ($a_{H_2O} = 1$). In that case, the Nernst equation takes the form

$$E_{Nernst} = E_0 + \frac{RT}{2F} \ln \left(p_{H_2} \sqrt{p_{O_2}} \right). \tag{7}$$

The terminal (load) voltage is generally obtained by subtracting from E_{Nernst} the following types of losses (or "irreversibilities"):

- activation loss
- concentration loss
- ohmic loss
- losses due to fuel crossover and internal current.

The practice of expressing—either directly or indirectly via the partial pressures—the Nernst EMF as a function of cell current is rather common in the literature (e.g., [33,34]). Building upon References [28,29], a major piece of work [7] on control strategies for SOFC operation developed a model for the constant fuel utilization condition using the following form of the Nernst equation:

$$E_{Nernst} = E_0 + \frac{RT}{2F} \left\{ \ln \left(\frac{K_{H_2O}}{K_{H_2}} \sqrt{\frac{K_r}{r_{HO}K_{O_2}}} \right) + \frac{1}{2} \ln \left(I_{FC} \left(\frac{1}{u} - 1 \right)^2 \left(\frac{2}{u} - r_{HO} \right) \right) \right\}$$
(8)

where K_{H_2} , K_{O_2} , and K_{H_2O} are valve molar constants [28], $K_r = 1/(4F)$ [28,38,39], r_{HO} is the ratio of hydrogen to oxygen input molar flow rates [7], u is the fuel utilization ratio [7,28], and I_{FC} is the load current. This model, or one of its variants, has been used in many papers in the fuel cell literature (e.g., [6,38]). (A detailed re-derivation of Equation (8) can be found in, for example, Reference [8].) Of the four types of losses mentioned above, the models in References [7,28] consider only the ohmic loss and obtain the output (load) voltage, V, as

$$V = E_{Nernst} - rI_{FC} \tag{9}$$

where E_{Nernst} is given by Equation (8), and r is the ohmic resistance of the cell.

As explained in Reference [8], the problem with Equation (8) is that it mixes equilibrium and non-equilibrium expressions. The Nernst voltage (EMF) is the reversible thermodynamic potential that applies only to the equilibrium condition of the cell; the equilibrium is lost when current is drawn from the cell. In other words, the Nernst voltage is by definition the open-circuit EMF and cannot therefore be expressed in terms of the load current [8].

To alleviate the aforementioned problem presented by Equation (8), we present an improved model in the following two sections.

All fuel cell models in the literature make simplifying assumptions and approximations; the present model also makes the standard set of assumptions that includes the following:

The gases used in the chemical reactions are ideal and uniformly distributed.

Appl. Sci. 2019, 9, 1066 5 of 24

• The fuel cell is considered a single "lumped" system; the species enter the cell through one end and exit through the other end.

- The only reaction that contributes to voltage takes place between hydrogen and oxygen. There is no parasitic reaction.
- All reactants generate their ideal number of electrons.
- No fuel or oxidant crosses the electrolyte.
- The gases have a uniform concentration in the supply channels; there are no pressure losses.
- The flows are incompressible.
- The fuel and oxidant are available as soon as they are needed (no delays).
- The partial pressures of the gases in the bulk of the anode and the cathode channels are assumed to be the same as those at the triple-phase-boundaries.
- The temperature is fixed (stable) at all times, and temperature variations across the cell are ignored.
- The operating voltage is uniform over the cell.
- Heat losses are negligible (the fuel cell is well insulated).
- There is no gas leakage.
- The Nernst equation can be applied.

4. Constant Fuel Utilization

Following Reference [7], the following two parameters are kept constant in this section: u (Equation (11)) and r_{HO} (Equation (17)). This corresponds to the operating condition known as constant fuel utilization ratio/factor condition or, equivalently, constant stoichiometry condition.

Hydrogen partial pressure is given by [28,38]

$$p_{H_2} = \frac{1}{K_{H_2}} \left(q_{H_2}^{in} - q_{H_2}^{react} \right) \tag{10}$$

where K_{H_2} is a valve molar constant [28], $q_{H_2}^{in}$ represents the input molar flow rate (mol/s) of hydrogen fuel, and $q_{H_2}^{react}$ is the hydrogen flow rate (mol/s) that takes part in the reaction.

Using the definition of fuel utilization ratio u,

$$u = \frac{q_{H_2}^{react}}{q_{H_2}^{in}} \tag{11}$$

we have

$$p_{H_2} = \frac{1}{K_{H_2}} \left(\frac{1}{u} - 1 \right) q_{H_2}^{react}. \tag{12}$$

The ratio of two different hydrogen partial pressures is then given by

$$\frac{p_{H_2}^{final}}{p_{H_2}^{initial}} = \frac{q_{H_2}^{react,final}}{q_{H_2}^{react,initial}}.$$
(13)

Since the reaction rate is given (from electrochemistry) by

$$q_{H_2}^{react} = \frac{I_{FC}}{2F},\tag{14}$$

the pressure ratio reduces to

$$\frac{p_{H_2}^{final}}{p_{H_2}^{initial}} = \frac{I_{FC}^{final}}{I_{FC}^{initial}}.$$
 (15)

Appl. Sci. 2019, 9, 1066 6 of 24

Next, the partial pressure for oxygen is expressed as

$$p_{O_2} = \frac{1}{K_{O_2}} \left(q_{O_2}^{in} - q_{O_2}^{react} \right) \tag{16}$$

where K_{O_2} is a valve molar constant [28], $q_{O_2}^{in}$ denotes the oxygen input molar flow rate, and $q_{O_2}^{react}$ is the oxygen flow rate that reacts. Noting that r_{HO} is the ratio of hydrogen-to-oxygen input flow rates,

$$r_{HO} = \frac{q_{H_2}^{in}}{q_{O_2}^{in}} \tag{17}$$

and that

$$q_{H_2}^{react} = 2q_{O_2}^{react} \tag{18}$$

we have

$$p_{O_2} = \frac{1}{K_{O_2}} \left(\frac{1}{r_{HO}} q_{H_2}^{in} - \frac{1}{2} q_{H_2}^{react} \right). \tag{19}$$

Now, by using the definition of u, we obtain

$$p_{O_2} = \frac{1}{K_{O_2}} \left(\frac{1}{ur_{HO}} - \frac{1}{2} \right) q_{H_2}^{react}.$$
 (20)

Two different oxygen partial pressures, then, have the following ratio:

$$\frac{p_{O_2}^{final}}{p_{O_2}^{initial}} = \frac{I_{FC}^{final}}{I_{FC}^{initial}}.$$
 (21)

Water (steam) partial pressure is expressed as

$$p_{H_2O} = \frac{1}{K_{H_2O}} \left(q_{H_2O}^{react} \right) \tag{22}$$

$$= \frac{1}{K_{H_2O}} \left(q_{H_2}^{react} \right) \tag{23}$$

$$= \frac{1}{K_{H_2O}} \frac{I_{FC}}{2F} \tag{24}$$

where K_{H_2O} is a valve molar constant [28], and $q_{H_2O}^{react}$ is the flow rate of the water vapor (steam) that is produced in the reaction. The final and initial pressures are then related by

$$\frac{p_{H_2O}^{final}}{p_{H_2O}^{initial}} = \frac{I_{FC}^{final}}{I_{FC}^{initial}}.$$
 (25)

The change in the voltage corresponding to the changes in the pressures is given by

$$\Delta V = \frac{RT}{2F} \ln \left(\frac{\frac{p_{H_2}^{final} \sqrt{p_{O_2}^{final}}}{p_{H_2O}^{finital}}}{\frac{p_{H_2}^{finital} \sqrt{p_{O_2}^{initial}}}{p_{H_2O}^{initial}}} \right)$$
(26)

Appl. Sci. 2019, 9, 1066 7 of 24

which, by Equations (15), (21) and (25), yields

$$\Delta V = \frac{RT}{2F} \ln \left(\sqrt{\frac{I_{FC}^{final}}{I_{FC}^{initial}}} \right)$$
 (27)

$$= \frac{RT}{4F} \ln \left(\frac{I_{FC}^{final}}{I_{FC}^{initial}} \right). \tag{28}$$

The above expression for voltage change shows that, for an increase in the cell current, ΔV is positive, representing a gain. (Following the present approach, an expression for ΔV is derived in the next section for the constant fuel flow condition. For an increase in current, ΔV for the constant fuel flow case is negative, representing a loss.) Now, setting $I_{FC}^{final} = I_{FC}$, and calling

$$\frac{I_{FC}^{final}}{I_{FC}^{initial}} = \frac{I_{FC}}{I_{FC}^{initial}} = I_{FC}^{ratio},$$

where I_{FC}^{ratio} is a unit-free number, we can express the output voltage as follows:

$$V \approx E_{Nernst} + \frac{RT}{4F} \ln I_{FC}^{ratio} - rI_{FC}$$
 (29)

where E_{Nernst} , determined from the initial cell conditions at open circuit, is given by Equation (6), $I_{FC}^{ratio} > 1$ (corresponding to an increase in current), and $I_{FC}^{initial}$ is small (ideally zero, corresponding to the open-circuit condition). A zero value of $I_{FC}^{initial}$, however, will render I_{FC}^{ratio} undefined. Since "it must be borne in mind that the logarithmic model does not work at very low currents, especially at zero" [1], for practical purposes, we set $I_{FC}^{initial} = 1$ A. This allows us to be "lazy" ([4], p. 50) and "simplify" ([1], p. 37) Equation (29) to

$$V \approx E_{Nernst} + \frac{RT}{4F} \ln I_{FC} - rI_{FC}$$
 (30)

where I_{FC} in expressed in Amperes. (Mathematically, the argument of logarithm is dimensionless (unitless). The propriety of taking the logarithm of a dimensioned quantity has long been debated [81–84]. In the fuel cell literature, however, it seems standard practice to do so (see, for example, References [1,3,4,85]). A classic example of the logarithm of a dimensioned quantity is the famous Tafel equation [86], which expresses the activation loss as $a + b \log(J)$, where J stands for the current density.)

Equation (30) holds for $I_{FC} > 1$ A and $I_{FC}^{ratio} > 1$, and smoothes out, by approximation, the transition from $I_{FC} = 0$ to $I_{FC} = 1$ A.

The second term in Equation (30), which is positive for $I_{FC} > 1$ A, represents a "Nernstian" gain. This can be given a physical interpretation. The fact that both u and r_{HO} are constant implies a constant fuel (hydrogen) utilization ratio and a constant oxidant (oxygen) utilization ratio. This means that: (a) hydrogen and oxygen are replenished as soon as they are consumed in the fuel cell; and (b) as current increases, so do hydrogen and oxygen supplies, causing the reactant concentrations to go up, thereby producing a higher potential according to the Nernst equation. The product (water or steam) concentration also increases with current under the constant-u-and-constant-v-u-condition, but any decrease in potential brought about by the increase in the product concentration is more than offset by the increase in potential caused by the corresponding increase in the reactant concentrations. This Nernstian gain is irreversible, much the same way that the traditional concentration loss (or mass transport loss or Nernstian loss) is irreversible.

Figures 1 and 2 present polarization curves for the constant fuel utilization condition using fuel cell parameters shown in Table 1. The data in Table 1 were taken from References [7,28]. E_{Nernst} , to be

Appl. Sci. 2019, 9, 1066 8 of 24

found from Equation (6) by plugging in the initial partial pressures and the E_0 , is taken to be 1 V in Figures 1 and 2. Figure 1 shows the two irreversibilities—Nernstian gain and ohmic loss—separately. The magnitude of the ohmic loss is plotted, ignoring the sign. The Nernstian gain, being a logarithmic function, is monotone increasing. The shape of the output voltage curve is determined by how much of the Nernstian gain is offset by the ohmic loss for a given current. The effect of the Nernstian gain can be seen rather conspicuously in the hump in the output voltage at low currents. The output voltage equals the Nernst voltage at the point where

$$\frac{RT}{4F}\ln I_{FC} = rI_{FC},\tag{31}$$

an equation that is difficult to solve analytically. Figure 2 shows, for low currents, how the polarization plot's hump varies for different values of the cell resistance. (An r of 3.28125 \times 10⁻³ Ω is used in Figure 1.)

Parameter	Value
	1273 K
u	0.8
K_{H_2}	0.843 mol/(s·atm)
K_{H_2O}	0.281 mol/(s·atm)
K_{O_2}	2.52 mol/(s·atm)
r^{-}	$3.28125 imes 10^{-4} \ \Omega$
r_{HO}	1.145
n	2
Constants	
F	96,485 Coulombs/mol
R	8.31 I/(mol K)

Table 1. Numerical values of parameters and constants.

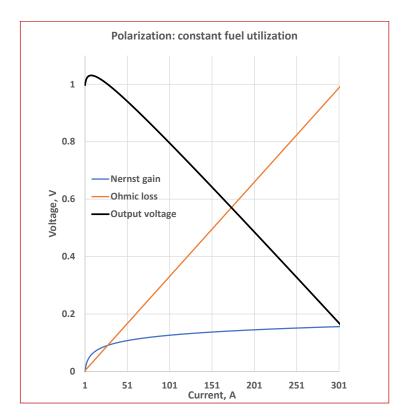


Figure 1. Polarization characteristics for the constant fuel utilization condition; the Nernstian gain and the ohmic loss are plotted separately.

Appl. Sci. 2019, 9, 1066 9 of 24

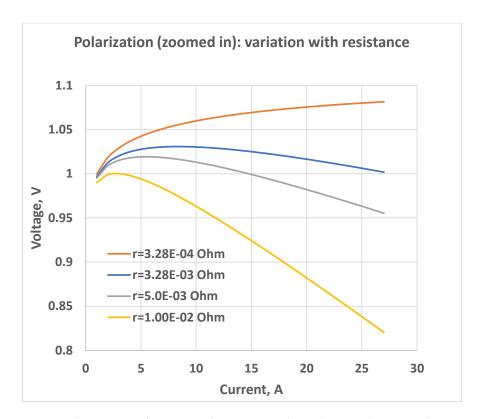


Figure 2. Zoomed in version of Figure 1 at low current values, showing changes in the curvatures.

It can be argued that, in theory, in the constant hydrogen and oxygen utilization condition, the concentration loss (mass transport loss) is mitigated or almost eliminated. (Of course much depends on the physical characteristics of the reactant supply systems, reformer and control mechanisms, time-delays in the diffusion process between the channel and the three-phase-boundary, etc.) Under the assumption of the ideal conditions in our lumped model, the concentration loss is negligible or zero in Equation (30).

For the constant fuel utilization operation, Equations (11), (12), (20) and (23) show that p_{H_2} , p_{O_2} , and p_{H_2O} all are directly proportional to (i.e., increase linearly with) the input fuel flow rate. This is shown in Figure 3, where the straight lines have been obtained using the parameter values in Table 1. All three lines have a zero intercept, and their slopes are given by

$$p_{H_2}: \frac{1}{K_{H_2}}(1-u)$$

$$p_{O_2}: \frac{1}{K_{O_2}} \left(\frac{1}{r_{HO}} - \frac{u}{2}\right)$$

$$p_{H_2O}: \frac{u}{K_{H_2O}}$$

The polarization characteristics produced by the constant fuel utilization model are similar in nature to those in the literature.

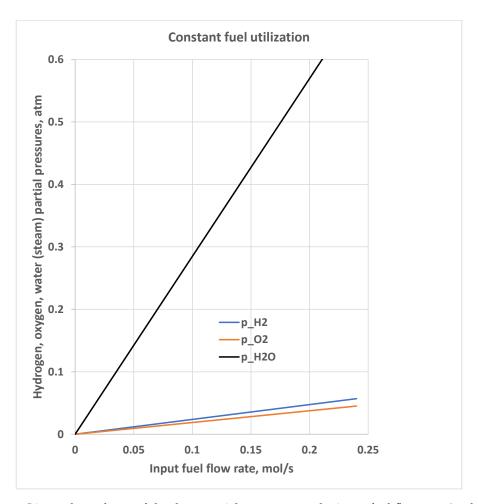


Figure 3. Linear dependence of the three partial pressures on the input fuel flow rate (under the constant fuel utilization condition).

The constant fuel utilization model is validated with experimental data, taken from Reference [87], on a Siemens-Westinghouse high-power-density solid oxide fuel cell with ten channels (APS HPD10). The data were obtained graphically (as in References [88–91]) from the results of Test 1018 in Figures 9 and 10 of Reference [87]. Figure 4 presents a model-versus-experimental comparison. The average absolute relative error was found to be 3.74%, a figure that is generally considered in the literature to indicate good agreement between theory and practice. Since the experimental current reported in Figures 9 and 10 of Reference [87] starts at 87 A for Test 1018, the Nernst voltage was estimated, by an approximate extrapolation (as in Figure 7-17 of Reference [85]), to be 0.74 V. The temperature was 1273 K, and the cell area was 870 cm² [87]. The resistance was estimated (following Reference [61]), from the slope of the (approximately) linear part of the experimental polarization (current-voltage) curve, to be 0.0006 Ω . The slope was obtained from the equation of the straight line derived by a least-squares linear regression fit of all but the first five experimental data points.

Voltage transients (dynamics) due to hydrogen concentration changes are an important issue in fuel cell operation and are investigated in References [92,93], where the constant fuel utilization mode is used to implement a current-based fuel control strategy (changing the fuel flow rate in proportion to the cell current) to keep voltage transients under check. Constant fuel utilization is also shown to be beneficial for control purposes of SOFC systems in Reference [94], where it is reported that a suitably chosen constant fuel utilization operation leads to nearly maximum-efficiency performance for variable loads.

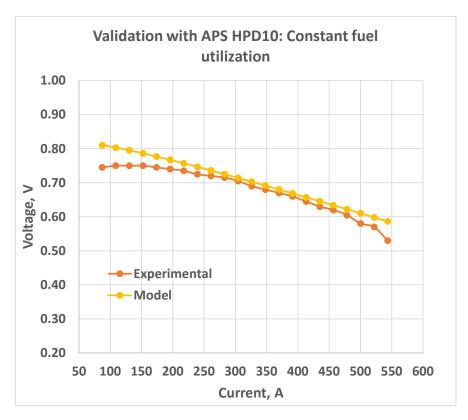


Figure 4. Validation results for the constant fuel utilization operation; mean absolute relative error = 3.74%.

5. Constant Fuel Flow

This section considers a constant input fuel flow rate; that is, $q_{H_2}^{in}$ does not vary with I_{FC} . The two operating modes—constant fuel utilization (the preceding section) and constant fuel flow (the present section)—are summarized in Figure 5 where the line parallel to the current axis represents a constant input fuel flow rate. The line starting at the zero current point represents a constant fuel utilization ratio and has a slope of $\frac{1}{2uF}$, a fact that is easily established from Equations (11) and (14). (Figure 5 uses data from Table 1.)

While the fuel flow rate is held constant, oxygen flow control can be done in one of two ways: either maintain a constant ratio of hydrogen-to-oxygen input flow rates or maintain a constant stoichiometric ratio for oxygen. We analyze these two modes in the following two subsections.

5.1. Constant Hydrogen-Oxygen Input Flow Ratio

This subsection treats the operating condition of r_{HO} (Equation (17)) being held constant. The constancy of both $q_{H_2}^{in}$ and r_{HO} means that the oxygen input molar flow rate, $q_{O_2}^{in}$, is also constant. From Equations (10) and (14), we have

$$p_{H_2} = \frac{1}{K_{H_2}} \left(q_{H_2}^{in} - \frac{I_{FC}}{2F} \right) \tag{32}$$

The ratio of two different partial pressures, corresponding to two different current values, is obtained as

$$\frac{p_{H_2}^{final}}{p_{H_2}^{initial}} = \frac{q_{H_2}^{in} - \frac{I_{FC}^{final}}{2F}}{q_{H_2}^{in} - \frac{I_{FC}^{initial}}{2F}}$$
(33)

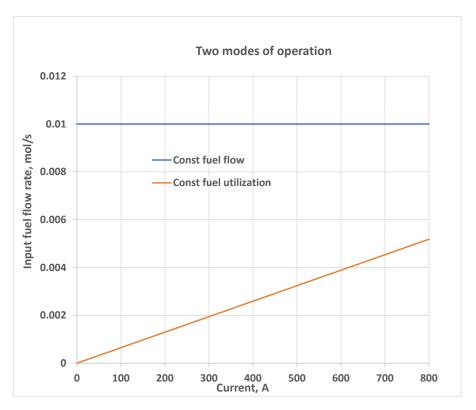


Figure 5. Variation in the input fuel flow rate with current in the two modes of operation: no variation (constant fuel flow mode) and linear variation (constant fuel utilization mode).

Next, substituting Equation (14) into Equation (19), we have

$$p_{O_2} = \frac{1}{K_{O_2}} \left(\frac{1}{r_{HO}} q_{H_2}^{in} - \frac{1}{2} \frac{I_{FC}}{2F} \right) \tag{34}$$

$$= \frac{1}{K_{O_2} r_{HO}} \left(q_{H_2}^{in} - r_{HO} \frac{I_{FC}}{4F} \right) \tag{35}$$

Two different oxygen partial pressures, corresponding to $I_{FC}^{initial}$ and I_{FC}^{final} , are then related by:

$$\frac{p_{O_2}^{final}}{p_{O_2}^{initial}} = \frac{q_{H_2}^{in} - r_{HO} \frac{I_{FC}^{final}}{4F}}{q_{H_2}^{in} - r_{HO} \frac{I_{FC}^{initial}}{4F}}$$
(36)

From Section 4, we know that the final and initial steam pressures have a ratio given by Equation (25). The change in voltage corresponding to the changes in the three pressures is given by

$$\Delta V = \frac{RT}{2F} \ln \left(\frac{\frac{p_{H_2}^{final} \sqrt{p_{O_2}^{final}}}{p_{H_2O}^{finil}}}{\frac{p_{H_2}^{finil} \sqrt{p_{O_2}^{initial}}}{p_{H_2O}^{initial}}} \right)$$
(37)

which, by Equations (25), (33) and (36), becomes

$$\Delta V = \frac{RT}{2F} \ln \left(\frac{I_{FC}^{initial}}{I_{FC}^{final}} \times \frac{q_{H_2}^{in} - \frac{I_{FC}^{final}}{2F}}{q_{H_2}^{in} - \frac{I_{FC}^{initial}}{2F}} \sqrt{\frac{q_{H_2}^{in} - r_{HO} \frac{I_{FC}^{final}}{4F}}{q_{H_2}^{in} - r_{HO} \frac{I_{FC}^{initial}}{4F}}} \right)$$
(38)

Appl. Sci. 2019, 9, 1066 13 of 24

The above expression shows that, for an increase in the cell current, ΔV is negative, representing a loss. As in Section 4, expressing all currents in units of Amperes, and setting $I_{FC}^{initial}=1$ A and $I_{FC}^{final}=I_{FC}$ in the above equation, we have

$$\Delta V = \frac{RT}{2F} \ln \left(\frac{1}{I_{FC}} \times \frac{q_{H_2}^{in} - \frac{I_{FC}}{2F}}{q_{H_2}^{in} - \frac{1}{2F}} \sqrt{\frac{q_{H_2}^{in} - r_{HO} \frac{I_{FC}}{4F}}{q_{H_2}^{in} - r_{HO} \frac{1}{4F}}} \right)$$
(39)

The output (load) voltage, *V*, for the constant fuel flow mode is then given by

$$V \approx E_{Nernst} + \frac{RT}{2F} \ln \left(\frac{q_{H_2}^{in}/I_{FC} - \frac{1}{2F}}{q_{H_2}^{in} - \frac{1}{2F}} \sqrt{\frac{q_{H_2}^{in} - r_{HO} \frac{I_{FC}}{4F}}{q_{H_2}^{in} - r_{HO} \frac{1}{4F}}} \right) - rI_{FC}$$
 (40)

where E_{Nernst} , the reversible (equilibrium) potential, is given by Equation (6), and the second term on the right side represents an irreversible loss (Nernstian loss). (Equation (40) holds for $I_{FC} > 1$ A.) This Nernstian loss is explained by the fact that, with $q_{H_2}^{in}$ and r_{HO} held constant (note that the constancy of both $q_{H_2}^{in}$ and r_{HO} implies that of $q_{O_2}^{in}$, too), reactant concentrations decrease and product concentration increases with increasing current. Equation (32) shows that, for a fixed $q_{H_2}^{in}$, hydrogen pressure p_{H_2} decreases linearly with increasing current. Similar is the behavior of p_{O_2} with respect to current, when the two input flow rates are kept fixed (see Equation (35)). The behavior of p_{H_2O} is opposite; as shown by Equation (24), p_{H_2O} grows linearly with current. Interestingly, given current, p_{H_2O} is (conditionally) independent of $q_{H_2}^{in}$. Figure 6 shows an example of how the three pressures change with current when the flow rates are held constant (the plots are produced with the parameter values in Table 1 and a fixed $q_{H_2}^{in} = 0.1$ mol/s).

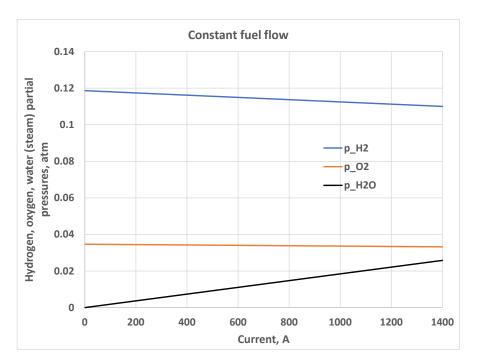


Figure 6. Linear dependence of the three partial pressures upon current (hydrogen and oxygen input flow rates are held constant).

The polarization curve for the constant flow rate condition (see Equation (40)) is shown in Figure 7 where the Nernst loss and ohmic loss are shown separately. The positive magnitudes of the two losses are shown; their signs are accounted for in the output voltage calculation. This figure uses

the parameter values in Table 1 along with $E_{Nernst} = 1$ V, $q_{H_2}^{in} = 0.004$ mol/s and $r = 3.28125 \times 10^{-3}$ Ω . The polarization plot shows the same trend that is generally seen in constant flow polarization in the literature.

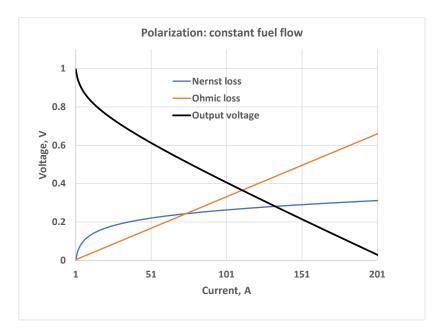


Figure 7. Polarization curve when hydrogen and oxygen input flow rates are held constant; the Nernst loss and the ohmic loss are plotted separately.

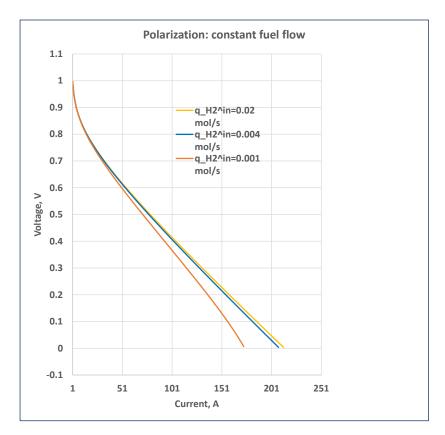


Figure 8. Effect of the input hydrogen flow rate on polarization (both input flow rates are held constant, with the oxygen flow rate proportional to the hydrogen flow rate).

Appl. Sci. 2019, 9, 1066 15 of 24

The effect of the (constant) input fuel flow rate on the polarization behavior is presented in Figure 8, where three different $q_{H_2}^{in}$ values are used: 0.02 mol/s, 0.004 mol/s, and 0.001 mol/s. The figure shows that a higher input flow rate leads to a higher voltage for the same load current, an observation that fits in with the concept of a higher reactant concentration leading to a higher potential. Figure 9 shows the variation in power characteristics corresponding to different (constant) fuel flow rates (the fuel flow rates in this figure are the same as the ones in Figure 8); a higher fuel flow rate results in higher power, as expected. Figures 8 and 9 use $E_{Nernst} = 1$ V and $r = 3.28125 \times 10^{-3} \Omega$.

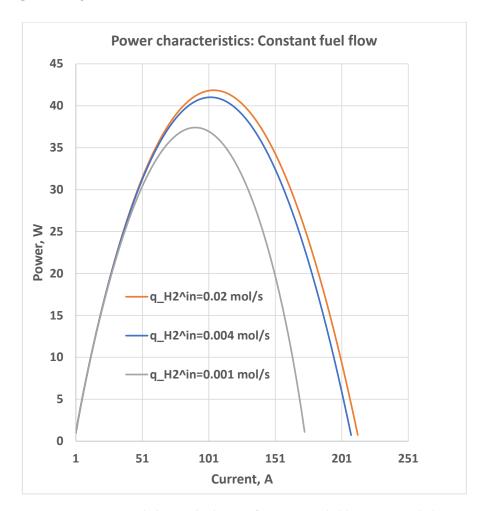


Figure 9. Power-versus-current behavior (both input flow rates are held constant, with the oxygen flow rate proportional to the hydrogen flow rate).

Results of model validation are shown in Figure 10, where the experimental polarization data (on a planar single SOFC at ENEA laboratories) were graphically obtained from Reference [61]. The results show reasonably good agreement, with an average absolute relative error of 3.96%. The validation was done under the following operational condition [61] for both the model and the experiment: temperature: 923 K; pressure: atmospheric; q_{H2}^{in} : 0.00112 mol/s; q_{O2}^{in} : 0.00038 mol/s; anode: 96% hydrogen, 4% water; cathode: 21% oxygen, 79% nitrogen; cell area: 121 cm². The calculated value of the model E_{Nernst} was 1.122 V, with the value of E_0 in Equation (2) estimated from the following relationship [61,95,96]:

$$E_0 \approx 1.253 - 2.4516 \times 10^{-4} \times T$$

Given the "systematic error of about 0.018 V" in Reference [61] between the theoretical Nernst voltage and its experimental counterpart, the experimental E_{Nernst} in the present validation was taken as 1.122 V - 0.018 V = 1.104 V. A resistance of 0.0043 Ω was used for validation; this value was obtained

Appl. Sci. 2019, 9, 1066 16 of 24

as the approximate slope of the (nearly) linear part of experimental polarization plot, as computed from a least-squares linear fit to the experimental data points (excluding the first four points).

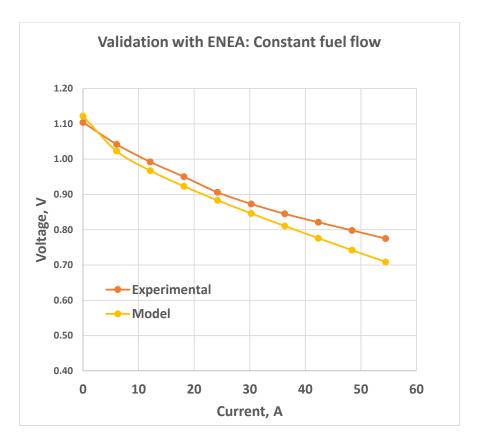


Figure 10. Validation results for the constant fuel flow operation; mean absolute relative error = 3.96%.

5.2. Constant Oxygen Stoichiometry Ratio

We consider a stoichiometric ratio, *S*, for oxygen:

$$S = \frac{q_{O_2}^{in}}{q_{O_2}^{react}} \tag{41}$$

where S is held constant (unlike Section 5.1 where r_{HO} is kept constant). The fuel flow rate is kept constant, as before.

Using the stoichiometric ratio, we can express oxygen partial pressure as:

$$p_{O_2} = \frac{1}{K_{O_2}} \left(q_{O_2}^{in} - q_{O_2}^{react} \right)$$

$$= \frac{1}{K_{O_2}} (S - 1) q_{O_2}^{react}$$
(42)

$$= \frac{1}{K_{O_2}} (S-1) q_{O_2}^{react} \tag{43}$$

$$= \frac{1}{K_{O_2}} (S-1) \frac{I_{FC}}{4F} \tag{44}$$

Two different oxygen partial pressures have the following ratio:

$$\frac{p_{O_2}^{final}}{p_{O_2}^{initial}} = \frac{I_{FC}^{final}}{I_{FC}^{initial}} \tag{45}$$

Appl. Sci. 2019, 9, 1066 17 of 24

From Equations (25), (33) and (45), we obtain the following expression for ΔV for the constant-S mode:

$$\Delta V = \frac{RT}{2F} \ln \left(\frac{q_{H_2}^{in} - \frac{I_{FC}^{final}}{2F}}{q_{H_2}^{in} - \frac{I_{FC}^{initial}}{2F}} \sqrt{\frac{I_{FC}^{initial}}{I_{FC}^{final}}} \right)$$
(46)

The above equation shows that for $I_{FC}^{final} > I_{FC}^{initial}$, the change in voltage is negative, implying a loss. As before, expressing currents in Amperes and setting $I_{FC}^{initial} = 1$ A and $I_{FC}^{final} = I_{FC}$, we have for $I_{FC} > 1$ A the following expression for the output voltage:

$$V \approx E_{Nernst} + \frac{RT}{2F} \ln \left(\frac{q_{H_2}^{in} - \frac{I_{FC}}{2F}}{q_{H_2}^{in} - \frac{1}{2F}} \sqrt{\frac{1}{I_{FC}}} \right) - rI_{FC}$$

$$(47)$$

where E_{Nernst} is given by Equation (6).

The polarization plot for the constant oxygen stoichiometry (and constant fuel flow) condition is shown in Figure 11 where the Nernstian loss is plotted separately (the Nernstian loss is plotted as a negative value). The polarization curve shows characteristics similar to those in the literature: after the initial drop, the voltage shows an approximately linear decay for much of the operating range before dropping heavily at high currents. In Figure 11, $q_{H_2}^{in}=0.0009$ mol/s. Figure 12 shows the effect of the (constant) fuel flow rate on the polarization performance. Power-versus-current behavior for the constant-fuel-flow-constant-oxygen-stoichiometry condition is presented in Figure 13. Figures 11–13 use the relevant data from Table 1 and $E_{Nernst}=1$ V. Figures 12 and 13 use the following $q_{H_2}^{in}$ values: 0.004 mol/s, 0.002 mol/s, and 0.0009 mol/s.

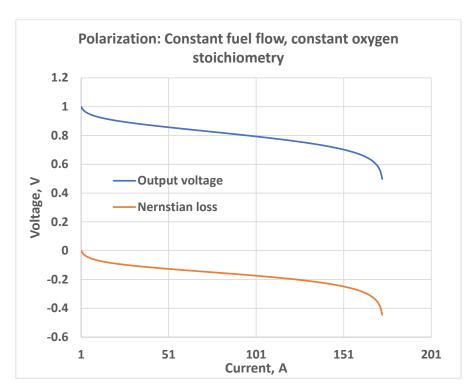


Figure 11. Polarization characteristics (constant hydrogen flow and constant oxygen stochiometry); the Nernst loss is plotted separately.

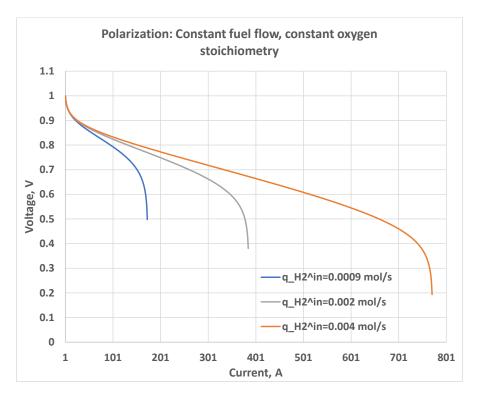


Figure 12. Effect of the hydrogen flow rate on polarization (constant hydrogen flow and constant oxygen stochiometry).

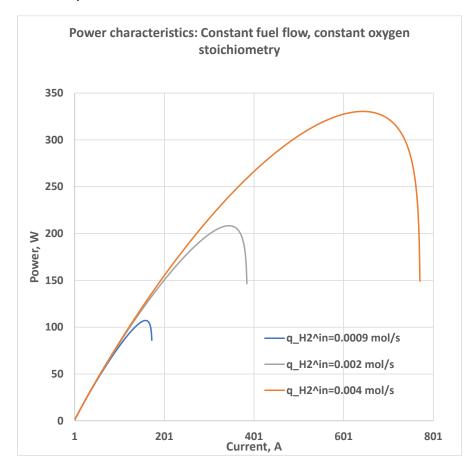


Figure 13. Effect of the fuel flow rate on the power-versus-current behavior (constant hydrogen flow and constant oxygen stochiometry).

6. Conclusions

An improved fuel cell model for constant fuel utilization control and constant fuel flow control was developed in this paper. By deriving expressions for two irreversibilities—a Nernstian gain (for the constant fuel utilization mode) and a Nernstian loss (for the constant fuel flow mode)—the model solves the problem of mixing up of reversible and irreversible potentials in the Nernst EMF expression. To our knowledge, the concept of the Nernstian gain and the characterization of the Nernstian gain as an irreversibility has not yet been presented in the fuel cell literature. The model thus eliminates the (flawed) need in previously published models for the open-circuit Nernst voltage to be expressed as a function of the load current. The model was validated against experimental data from the literature, with the average absolute relative error being less than 4%.

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Conflicts of Interest: The author declares no conflict of interest.

Abbreviations

The following abbreviations are used in this manuscript:

SOFC Solid oxide fuel cell

PEMFC Proton exchange membrane fuel cell

EMF Electromotive force

Nomenclature

*E*_{Nernst} Nernst potential (open-circuit EMF) of a single cell, V

*E*₀ Standard (reference) EMF of a single cell, V

 E_0^0 Standard (reference) EMF of a single cell at temperature T_0 , V

V Output terminal voltage of a single cell, V

Temperature, K

n Number of electrons transferred

a Activity

 a_{H_2} Activity of hydrogen a_{O_2} Activity of oxygen

 a_{H_2O} Activity of water vapor (steam) Δs Change in entropy, J/(mol K)pPressure or partial pressure, atm p^0 Standard-state pressure, atm p_{H_2} Partial pressure of hydrogen, atm p_{O_2} Partial pressure of oxygen, atm p_{H_2O} Partial pressure of water vapor, atm

 I_{FC} Fuel cell current, A u Fuel utilization ratio

 r_{HO} Ratio of hydrogen-to-oxygen input flow rates K_{H_2} Valve molar constant for hydrogen, mol/(s atm) K_{O_2} Valve molar constant for oxygen, mol/(s atm) Valve molar constant for water vapor, mol/(s atm)

 K_r Modeling constant, mol/(s A) $q_{H_2}^{in}$ Hydrogen input flow rate, mol/s $q_{H_2}^{out}$ Hydrogen output flow rate, mol/s

 $q_{H_2}^{react}$ Hydrogen flow rate that takes part in the reaction, mol/s

 $q_{O_2}^{in}$ Oxygen input flow rate, mol/s $q_{O_2}^{out}$ Oxygen output flow rate, mol/s $q_{O_2}^{react}$ Oxygen reacting flow rate, mol/s

Appl. Sci. 2019, 9, 1066 20 of 24

- $q_{H_2O}^{out}$ Water vapor output flow rate, mol/s
- $q_{H_2O}^{react}$ Water vapor flow rate produced in the reaction, mol/s
- *r* Ohmic resistance of a single cell, Ohm
- R Universal gas constant, J/(mol K)
- F Faraday's constant, Coulombs/mol

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Appl. Sci. 2019, 9, 1066 24 of 24

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