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To cite this version:

HAL Id: hal-01896721
https://hal.archives-ouvertes.fr/hal-01896721
Submitted on 18 Oct 2018

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Stability study and dynamical analysis of the multicellular chopper

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Abstract: The dynamical properties of a two-cells chopper connected to a particular nonlinear load are described in this paper. Some interesting and complex attractors are obtained. We analyse the system by means of Lyapunov exponents, fractal dimension, Poincaré mapping, first return, bifurcation diagram and phase portraits, respectively. Our model is described by a continuous time three-dimensional non-autonomous system and displays two-scroll chaotic attractors for certain values of its parameters. The numerical simulation is used to figure out their chaotic attractors. The analysis results show clearly that this is a new chaotic system which deserves further detailed investigation. Finally, phase portraits are obtained by using MATLAB/Simulink, which validates the theoretical analysis results.

Keywords: chaotic behaviour; multicellular converter; nonlinear dynamics; dynamical properties; dissipative dynamics; equilibria and stability.


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Malek Ghanes received his MS degree and PhD in Applied Automatic and Informatics both from IRCCyN, Ecole Centrale Nantes, in 2002 and 2005, respectively. From September 2005 to September 2006, he was a Postdoctoral position at GREYC. Since September 2006 and November 2008, he is Associate Professor and Head of Automatic Department, respectively, at ENSEA, France. In November 2012, he obtained his Accreditation to Supervise Research (HDR) and qualified as a Full Professor. His research interests include observation and control of nonlinear system, with applications to electric and chaotic systems.
1 Introduction

Multicell converters have grown from an attractive theoretical concept to industrial applications, especially for use in higher power applications (Erickson and Maksimovic, 2001; Dargahi et al., 2014; Meynard and Foch, 1992; Jiangborani et al., 2015; Lobisger and Kolar, 2015; Hayashi and Ninomiya, 2016) and they are well suited to packaging of renewable energy sources. Indeed, due to their modular structure, they can be combined easily (Meynard, 2014). In recent decades, it was discovered that most of static converters were the seat of unknown nonlinear phenomena in power electronics (Bernardo and Chi, 2002; Vazquez et al., 2016; Choupan et al., 2017; Hayashi et al., 2016; Hayashi and Ninomiya, 2016). It is for example the case of multicellular choppers that can exhibit unusual and sometimes chaotic behaviours. Obviously this may generate dramatic consequences. However, the usual averaged models do not allow to predicting nonlinear phenomena encountered. By nature, these models obscure the essential nonlinearities. To analyses these strange behaviours, it is necessary to use a nonlinear hybrid dynamical model (Barbot et al., 2007; Benmansour et al., 2010; La Rosa et al., 2000; Ahmad et al., 2015).

There have been many methods for detecting chaos from order. They include Poincaré sections, Lyapunov exponents (Skokos, 2017; Maffione et al., 2011), fast Lyapunov indicators (FLIs) (Lega et al., 2016; Barrio, 2016), smaller alignment index (SALI) and its generalised alignment index (Skokos and Manos, 2016), bifurcations, power spectra, frequency analysis, 0-1 test, geometrical criteria, and fractal basin boundaries, etc. Each of them has its advantages and drawbacks in classifying the attractors. For a system of any dimension, the Lyapunov exponents, as a method of the average exponential deviation of two nearby orbits in the phase space, are efficient to distinguish regular from chaotic orbits, but quite a long integration times are often needed before obtaining reliable limit values. In addition, a more sensitive indicator, the FLI of Froeschlé (Lega et al., 2016; Barrio, 2016) is an ideal method. Distinction methods between chaotic and ordered orbits such as the small alignment index and fast Lyapunov indicator firstly used to discuss dissipative system in (Huang and Wu, 2012).

The main purpose of the present paper is to use numerical approaches to analyse the dynamical properties of a two-cells chopper connected to nonlinear load. The rest of this paper is organised as follows. Section 2, the two-cells chopper connected to a particular nonlinear load modelling and analysis of a switching cell are presented. Basic dynamical properties of the chaotic system are also investigated in Section 3. Finally, chaotic behaviour and simulation results are presented in Section 4.

2 Multicell converter model

The multicellular converters (Figure 1) are built starting from an association of a certain number of cells. At the output, one obtains \( (p+1) \) levels \( \left( \frac{E}{p}, \ldots, \frac{(p-1)E}{p}, \frac{E}{p} \right) \).

This association in series allows to the output source \( V_s \) to evolve on \( (p+1) \) possible levels. As the orders of the switches of the cells of commutation are independent, one obtains \( 2^p \) possible combinations. Thus, it is necessary to ensure an equilibrated distribution of the voltage of the floating condensers. Under these conditions, one obtains the following property: The converter has \( p - 1 \) floating voltages sources and the voltage of the capacity of index \( k \) is \( \frac{E}{k} \). The control signal associated with each commutation cell is noted as \( u_i \), where \( i \) represents the number of cells of the topology. This signal will be equal to 1 when the upper switch of the cell is conducting and 0 when the lower switch of the cell is conducting.

To simplify the study and the notations, we will study the overlapping operation of a converter with two cells (Figure 2). Its function is to supply a passive load (RL) in series with another nonlinear load (Fortuna et al., 2001) connected in parallel with a capacitor. Four operating modes are then possible as shown in Figure 3. Note that the floating source takes part in the evolution of the dynamics of the system only to the third and fourth mode. In the third mode, the capacity discharges and charge during the fourth mode. Thus, if these two modes last same time with a constant charging current, then the average power transmitted by this floating source over one period of commutation is null. We also notice that these two modes make it possible to obtain by commutation the additional levels \( \frac{E}{2} \) on the output voltage \( V_s \).

As the switches of each cell are regarded as ideals, their behaviour can be to model by a discrete state taking of the values 0 (on) or 1 (off). In practice, some of these states never will be visited for reasons of safety measures or following the strategy of order adopted or because of the structure of the converter him finally to even or comply with the rule of adjacency. The transitions are not necessarily controlled.

The system model can be represented by three differential equations giving its state space.

\[
\begin{align*}
L \frac{du_1}{dt} &= (u_1 - u_2) v_c - v_{c_1} - R i_L + u_2 D \\
C \frac{dv_c}{dt} &= (u_2 - u_1) i_L \\
C_i \frac{dv_{c_1}}{dt} &= i_L - g(v_{c_1})
\end{align*}
\]
Figure 1  p-cells chopper connected to RL load

Figure 2  Two-cell chopper connected to a nonlinear load

Figure 3  Switching cell and its configurations

$u_2 = 1  \quad u_1 = 1$

$u_2 = 0  \quad u_1 = 0$

$u_2 = 0  \quad u_1 = 1$

$u_2 = 1  \quad u_1 = 0$
where \( g(\cdot) \) is a piecewise-linear function
\[
g(v_c) = G_a v_c + \frac{1}{2} (G_a - G_b) (|v_c + 1| - |v_c - 1|)
\]
which is the mathematical representation of the characteristic curve of nonlinear load. The slopes of the inner and outer regions are \( G_a \) and \( G_b \), while \( B_p \) indicates break points. The parameters of the circuit elements are fixed as \( C = 0.1 \mu F, C_l = 40 \mu F, L = 50 \text{ mH}, R = 10 \Omega, E = 100 V. \)

3 Dynamical properties of system (1)

3.1 Equilibria and stability

The equilibria of (1) can be calculated by solving the following algebraic equations simultaneously
\[
\begin{align*}
L \frac{d i_l}{dt} &= (u_1 - u_2) v_c - v_c R i_l + u_2 E = 0 \\
C \frac{dv_c}{dt} &= (u_2 - u_1) i_l = 0 \\
C_l \frac{dv_c}{dt} &= i_l - g(v_c) = 0
\end{align*}
\]
Rescaling equation (1) as \( v_c = x_2 B_p, v_c = x_3 B_p, i_l = x_3 G B_p, \) \( G = \frac{1}{R}, t = \frac{C}{G} \tau \) and then redefining \( \tau \) as \( t \) the following set of normalised equations are obtained:
\[
\begin{align*}
\dot{x}_1 &= \beta (-p x_1 + e x_2 - x_3) + \alpha E \\
\dot{x}_2 &= e x_3 \\
\dot{x}_3 &= p (x_1 - g(x_3))
\end{align*}
\]
where \( e = u_2 - u_1, p = \frac{C}{C_l}, \beta = \frac{C}{L G^2}, \gamma = RG, \alpha = \frac{BE}{B_p}. \)

Obviously \( g(x_3) = bx_3 + 0.5(a - b) [|x_3 + 1| - |x_3 - 1|], \) or
\[
g(x_3) = \begin{cases} 
  bx_3 + a - b & x_3 > 1 \\
  ax_3 & |x_3| \leq 1 \\
  bx_3 - a + b & x_3 < -1
\end{cases}
\]
where \( a = \frac{G_a}{G}, b = \frac{G_b}{G}. \)

Now the dynamics of equation (3) depends on the parameters \( e, p, \beta, \gamma, a, b, \) and \( \alpha. \) The circuit parameters used are then rescaled as: \( p = 25.10^{-4}, \alpha = 2.10^{-2}, \beta = 2.10^{-2}, a = -15, b = 5, \gamma = 1. \) We thus find the equilibria:

Case 1: \( e = 1, i.e., u_1 = 0; u_2 = 1 \)
\[
\begin{align*}
&\quad \quad x_3 > 1, \quad E_1 = \begin{pmatrix} 0 \\ \frac{b-a}{b} \alpha \beta \end{pmatrix} = \begin{pmatrix} 0 \\ 96 \\ 4 \end{pmatrix} \\
&\quad \quad x_3 \leq 1, \quad E_2 = \begin{pmatrix} 0 \\ -\frac{a}{\beta} \end{pmatrix} = \begin{pmatrix} 0 \\ -100 \\ 0 \end{pmatrix}
\end{align*}
\]

Case 2: \( e = -1, i.e., u_1 = 1, u_2 = 0 \)
\[
\begin{align*}
&\quad \quad x_3 > 1, \quad E_{11} = \begin{pmatrix} 0 \\ \frac{a-b}{b} \end{pmatrix} = \begin{pmatrix} 0 \\ -104 \\ -4 \end{pmatrix} \\
&\quad \quad x_3 < -1, \quad E_{21} = \begin{pmatrix} 0 \\ \frac{a-b}{b} \end{pmatrix} = \begin{pmatrix} 0 \\ -4 \\ -4 \end{pmatrix}
\end{align*}
\]

Case 3: \( e = 0 \)
\[
\begin{align*}
&\quad \quad x_3 > 1, \quad E_{12} = \begin{pmatrix} \frac{a+b}{\beta} \gamma \left(\frac{\alpha}{\beta} - 1\right) b \\ \frac{\gamma(a-b) + \alpha}{\beta} \end{pmatrix} = \begin{pmatrix} 80 \\ 20 \end{pmatrix}
\end{align*}
\]
Let us study the stability of different equilibrium points.

1. For $x_3 > 1$, the Jacobian matrix is defined as

$$
\begin{bmatrix}
\beta \frac{x_3 - 1}{\beta x_3 + 1} \\
\frac{1}{\beta x_3 + 1}
\end{bmatrix}
= \begin{bmatrix}
\frac{1}{2} \\
\frac{1}{2}
\end{bmatrix}
$$

The eigenvalues are:

\begin{itemize}
  \item a If $\beta = 1$; $\lambda_1 = 0.0140$; $\lambda_2 = -0.0143$; $\lambda_3 = -0.0375$
  \end{itemize}

Here $\lambda_1$ is a positive real number, $\lambda_2$ and $\lambda_3$ are two negative real numbers.

2. For $|x_3| \leq 1$, the Jacobian matrix is defined as

$$
\begin{bmatrix}
\beta \frac{x_3 - 1}{\beta x_3 + 1} \\
\frac{1}{\beta x_3 + 1}
\end{bmatrix}
= \begin{bmatrix}
\frac{1}{2} \\
\frac{1}{2}
\end{bmatrix}
$$

The eigenvalues are:

\begin{itemize}
  \item a If $\beta = 1$; $\lambda_1 = 0.0140$; $\lambda_2 = -0.0143$; $\lambda_3 = -0.0375$
  \end{itemize}

Here $\lambda_1$ is a positive real number, $\lambda_2$ and $\lambda_3$ are two negative real numbers.

3. For $x_3 < -1$, the Jacobian matrix is defined as

$$
\begin{bmatrix}
\beta \frac{x_3 - 1}{\beta x_3 + 1} \\
\frac{1}{\beta x_3 + 1}
\end{bmatrix}
= \begin{bmatrix}
\frac{1}{2} \\
\frac{1}{2}
\end{bmatrix}
$$

The eigenvalues are:

\begin{itemize}
  \item a If $\beta = 1$; $\lambda_1 = 0.0140$; $\lambda_2 = -0.0143$; $\lambda_3 = -0.0375$
  \end{itemize}

Here $\lambda_1$ is a positive real number, $\lambda_2$ and $\lambda_3$ are two negative real numbers.
The eigenvalues are:

\( \lambda_1 = 0.0140 \), \( \lambda_2 = -0.0141 \), \( \lambda_3 = -0.0126 \).

Here \( \lambda_1 \) is a positive real number, \( \lambda_2 \) and \( \lambda_3 \) are two negative real numbers.

Therefore, the equilibrium \( E_3 = \left( \begin{array}{c} 0 \\ -104 \\ 4 \end{array} \right) \) and \( E_3 = \left( \begin{array}{c} 0 \\ 4 \\ -4 \end{array} \right) \) are a saddle point in the plane \((x_1, x_2)\); these equilibrium are unstable.

b) If \( \varepsilon = 0; \lambda_1 = -0.0002; \lambda_2 = -0.0125; \lambda_3 = 0 \)

Then, \( E_{12} = \left( \begin{array}{c} 80 \\ x_2 \\ 20 \end{array} \right) \) and \( E_{13} = \left( \begin{array}{c} 10 \\ x_2 \\ 10 \end{array} \right) \) are stable equilibrium points within the meaning of Lyapunov for all \( x_2 \).

### 3.2 Symmetry and invariance

Symmetry on the \( x_2 \)-axis is due to invariance of the equations under the transformation \((x_1, x_2, x_3) \rightarrow (x_1, -x_2, x_3)\). The system has rotational symmetry in the plane \((x_1, x_3)\) (Wei and Yang, 2011; Dadras and Momeni, 2009).

### 3.3 Dissipativity

For system (3), we can obtain

\[
\nabla V = \begin{bmatrix} \frac{\partial x_1}{\partial x_1} + \frac{\partial x_2}{\partial x_2} + \frac{\partial x_3}{\partial x_3} \end{bmatrix} \begin{bmatrix} -\beta y - pb \quad \text{if} \quad |x_3| > 0 \\
-\beta y - pa \quad \text{if} \quad |x_3| \leq 0 
\end{bmatrix}
\]

As long as \(( -\beta y - pb ) < 0 \) our system (3) is dissipative. Thus the volume elements are contracting. After a time unit this contraction reduces a volume \( V_0 \) by a factor \( e^{-(\beta y + pb)t} \). Therefore, all system orbits are ultimately confined to a specific subset having zero volume and the asymptotic motion settles onto an attractor (Wei and Yang, 2011; Dadras and Momeni, 2009; Zhou et al., 2008; Li et al., 2009; Djondiné et al., 2014).

### 3.4 Lyapunov exponent and fractional dimension

The Lyapunov exponents generally refer to the average exponential rates of divergence or convergence of nearby trajectories in the phase space. If there is at least one positive Lyapunov exponent, the system can be defined to be chaotic. Thus,

- for the case \( \varepsilon = \pm 1 \), we find that the Lyapunov exponents have the values: \( \lambda_1 = 0.0318 \), \( \lambda_2 = -0.0319 \), \( \lambda_3 = -0.0128 \). Therefore, the Lyapunov dimension of this system is:

\[
D_L = j + \sum_{j=1}^{l} \frac{l_j}{|l_j|} = 2 + \frac{0.0318 - 0.0319}{-0.0128} = 2.0078125
\]

- for the case \( \varepsilon = 0 \), we find that the Lyapunov exponents have the values: \( \lambda_1 = -1.4029^{-4} \), \( \lambda_2 = 0 \), \( \lambda_3 = -0.0126 \). The Lyapunov dimension of this system is:

\[
D_L = j + \sum_{j=1}^{l} \frac{l_j}{|l_j|} = 2 + \frac{1.4029^{-4}}{-0.0126} = 1.9886587
\]

This means that our system is chaotic since one of the Lyapunov exponents is positive and the Lyapunov dimensions of the system are fractional. The null Lyapunov exponent is obviously related to the critical nature between expansion and contracting nature of different directions in phase space. The chaotic behaviour for this system can also be shown by plotting the separation of two nearby trajectories.

### 3.5 Poincaré section, first return and bifurcation diagram

The Poincaré section is computed for \( x_1 = 0 \) when the frequency is 20 Hz. From Figure 4, it can be seen that the symmetry (Djondiné et al., 2013) is around the nominal voltage capacitor \( V_{CN} \) that is 50 V.

The first return map is computed by the voltage of nonlinear load \( (x_3) \) for each minimum of \( x_3 \) which is detected with \( x_1 = 0 \) and \( |x_1| < 0.01 \). From Figure 5(a), it can be seen that there is a folding point at each end of the cross and obviously symmetry occurs (Dang-Vu and Delcarte, 2000).

**Figure 4** The Poincaré section of \( x_1 - x_2 \) plane
Figure 5  (a) First return map (b) Bifurcation diagram (see online version for colours)

Figure 6  Phase portraits of the system, (a) $C = 0.01 \ \mu F$ (b) $C = 0.1 \ \mu F$ (c) $C = 0.5 \ \mu F$ (d) $C = 1 \ \mu F$ (e) $C = 10 \ \mu F$ (f) $C = 100 \ \mu F$
Figure 6  Phase portraits of the system, (a) $C = 0.01$ μF (b) $C = 0.1$ μF (c) $C = 0.5$ μF (d) $C = 1$ μF (e) $C = 10$ μF (f) $C = 100$ μF
(continued)

From bifurcation diagram [Figure 5(b)] it can be seen that the system is in the chaos between 1 Hz and 30 Hz. As the frequency increases, the system behaviour stays chaotic and as maximum amplitude at the resonance frequency

$$f_s = \frac{\sqrt{C + C_1}}{2\pi \sqrt{LCC_1}} = 2,253.6 \text{ Hz},$$

the load current $x_1$ gets the maximum value of 5 A. When the switching frequency is higher than the resonance frequency, the amplitude decreases and finally the behaviour passes from strange attractor to an equilibrium point.

3.6 Observation of chaotic dynamics

This nonlinear system exhibits the double-scroll chaotic attractor. It should be pointed out that, with different parameters, system (3) can evolve to other complex dynamics such as single-scroll chaotic attractor and periodic orbit (Huang and Zhao, 2016). In this system, the capacitor $C$ is a parameter of control and the value of $C$ can be changed within a certain range. When the parameter $C$ is changed, the chaos behaviour of this system can effectively be controlled. In the numerical simulation, the initial values of the system (3) are $(0, 5, 4)$. Using MATLAB program, the numerical simulation have been completed. This nonlinear system exhibits the complex and abundant of the chaotic dynamics behaviours, the strange attractors (Tremori et al., 2016) are shown in Figure 6. These phase portrait are obtained by solving equations (3) by means of Runge-Kutta method for step size of 0.000001. At the beginning of simulation $u_1 = u_2 = 0$. Now is clear that the double scroll attractor has a structure quite different from well-known Lorenz, Rössler and Chua attractors (Buscarino et al., 2011) since the double scroll (Koubaa, 2016) structure has not observed with the latter attractors.

- for $C = 0.01$ to 0.1 μF, system (3) is chaotic, and there are several periodic windows in the chaotic band, as shown in Figures 6(a)–6(b)
- for $C = 0.5$ to 1 μF, systems (3) are some quasi-periodic loops and also call a reverse period-doubling bifurcation window, as shown in Figures 6(c)–6(d).
- for $C = 10$ to 100 μF, there is a reverse period-doubling bifurcation window too, as shown in Figures 6(e)–6(f).

4 Conclusions

In this paper, the dynamics of a two-cells chopper associated to a particular nonlinear load has been reported and confirmed analytically and numerically. Some basic properties of the system including invariance, dissipativity, equilibria and their stability and Lyapunov exponents have been investigated. The particular interest is that our system can generate different scroll chaotic attractors with variation of a parameter of the circuit. The topological structure of the new system should be completely and thoroughly investigated. According to the analysis above, the two-cells converter with nonlinear load structure belongs to strongly nonlinear system, and its performance can be easily affected when the circuit parameters are varied. Therefore, the parameters should be chosen appropriately according to the research results. The analysis methods and research findings will possess an important reference value to engineering design and performance analysis. It is expected that more detailed theory analyses and simulation investigations will be provided in a forthcoming paper.
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