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ABSTRACT
This paper investigates the control problem for static boost type converters using an high gain state feedback robust controller incorporating an integral action. The robust feature allows to achieve the required performance in presence of parametric uncertainties, while the integral action provides an offset free performance with respect to the desired levels of voltage. The adopted high gain approach is motivated by both fundamental as well as practical considerations, namely the underlying fundamental potential and the design parameter specification simplicity. The stability and convergence analysis has been carried out using an adequate Lyapunov approach, and the control system calibration is achieved throughout a few design parameters which are closely related to the desired dynamical performances. The effectiveness of the proposed control approach has been corroborated by numerical simulations and probing experimental results.

1 Introduction
The control of static converters has been the subject of an important research activity over the past decades. This interest is mainly due to the emergence of embedded electronics in everyday life, increasing thereby the need for more efficient converters. Indeed, this type of converter is used for many applications such as laptop computers [1], photo-voltaic [2], vehicular systems [3], fuel cell [4], etc. Furthermore in recent use of electrical devices, the control problem is not the only feature required from the user. There are more and more need of informations on the evolution/aging of the system. In the case of converters, these informations can be obtained by reconstructing voltages and currents thanks to suitable observers. Of particular interest is the Boost Converter which maintains a low power loss in the conversion process. The counterpart is that the average model presents strong nonlinearities due to the presence of bilinear terms and unstable zero dynamics which reduce the control system performances. These difficulties have given rise to several control solutions involving the converter output voltage. A widely spread method is the current-mode control which consists of two nested closed loop systems where the inner one is a current control system, while the outer one is a voltage control system ([5–7]). The feasibility of the nested control system require the inner control to be relatively faster. Furthermore, it stabilizes the unstable zero dynamics, allowing thereby to apply usual control designs. However the control system performances may be poor when working out of the nominal operating conditions. Even so, this approach, commonly used in the industry, has been developed in many regards since then ([8–11]).
Several approaches have been recently proposed to circumvent the limitations of the usual methods from both performance improvement and simplicity points of views. They are based on direct output voltage control without resorting to the current control system as well as the stabilization of unstable zeros dynamics. A common technique consists in using the linearized model of the system, which is valid in a neighborhood of the nominal point, with its uncertainties if needed. Some models for these uncertainties have been proposed in [12]. LQR controllers have been used in [13,14] and [15] under small signals constraints for implementation purposes. In order to cope with large signal disturbances, the linear techniques have been adapted in different ways that can be classified into two main streams. The first approach consists in designing a robust linear controller. Robust linear state feedback have been designed in [16] using passivity and using LMI in [17–20]. These approaches necessarily involve trade-offs between performances and robustness since the structure of the controller is linear. Different works have been done in order to keep the simplicity of the control law, while having better performances. Auto-tuning PID have been considered in [21, 22]. The extended linearization with a PI has been pursued in [23] and the model predictive control has been considered in [24]. Although, these approaches increase the robustness and performances to large disturbances, they would exhibit poor performances on the whole working space and can lead to unexpected behavior.

Given the nonlinear structure of the average model of the boost converter, numerous approaches using nonlinear control law have been proposed in the literature. Early approaches such as passivity based control and Lyapunov based control have been considered in [25–27]. These methods have shown to be more robust to large signals disturbances but their performances cannot be compared to the standard techniques. More complicated techniques have then been used in order to achieve better performances using Hammerstein models ( [28, 29]) or time optimal control ( [30, 31]). It is however worth noticing that these techniques need a very precise knowledge of the model which is difficult since the model is usually an average model (this model does not take into account all the uncertainties on the components). Backstepping control has been used in [32] for current control, the voltage value is indirectly determined from the current using the model knowledge and henceforth the precision on the desired voltage directly relies on the model’s accuracy. Fuzzy controllers have been used in [33–35] and [36] without any guarantee of stability and performance. Various contributions have been devoted to the sliding mode control approach for its robustness to bounded uncertainties [37–41]. They suffer from implementation issues because of the chattering behavior occurring near the sliding surface.

The objective of this paper is to propose a state feedback which is easy to calibrate, robust to uncertainties and competitive with respect to classical methods such as current mode control. Of fundamental interest, it performs well on the whole working space and is theoretically guaranteed. In order to cope with the unstable zeros dynamics, only regulation of the output voltage at a constant value is considered, as in most of the applications of this boost converter. The controller consists in combining an integral high-gain state feedback control law with an high-gain observer. Only three design parameters have to be specified and these parameters have physical meaning and are closely related to the underlying control dynamical performance. The admissible assumption of having access to on-line measurements of the current in the inductor and the output voltage is considered here. The current delivered to the load and the input voltage are not measured and are then estimated by a high-gain observer. In most approaches, the robustness toward changes in the current in the inductor and the output voltage is obtained through a degradation of the performances, since they are supposed to be unknown. This is not the case here, since these informations are reconstructed and used in the control law. It should be noted that these informations could be further use for fault detection purposes.

This paper is organized as follows. In section 2, the dynamical average model of the boost converter is presented. In the next section, an high gain type control law is synthesized, its construction is split into three parts. First the nominal case is considered and a control law is designed, then an observer reconstruct the current delivered to the load and the output voltage, finally these two parts are mixed in order to obtain a robust state feedback. Section 4 is dedicated to experimental results. Finally, section 5 concludes the paper.

2 Modeling of boost converter

The electrical model of the Boost converter considered in this paper is shown on Fig. 1. It consists of an inductor $L_c$, a capacitor $C$ and a controlled switch. The output current is denoted $i(t)$ and the control variable $u(t)$ represents the duty cycle of the controlled IGBT. The first objective is to keep the converter’s output voltage $v_c(t)$ at a constant reference value $v^*$, independently of the disturbances that could affect the operation of the converter, namely, the variations of the supply voltage of the battery and the load. This control objective is feasible, while tracking a reference sequence voltage is not possible in this form as the zero dynamics are not stable. For cost reasons and simplicity of implementation, the output voltage and the coil current are the unique available measurements. The second objective is to deliver an on line estimation of the unmeasured variables, namely the battery voltage and the output current for process monitoring considerations.

To achieve these objectives, the average model considered in this paper directly includes the losses and can thus be
written into the following form:

\[
\begin{align*}
\dot{v}_c(t) &= \frac{1}{C} ((1 - u(t))i_I(t) - i_E(t)), \\
\dot{i}_I(t) &= \frac{1}{L} (v_c(t) - (1 - u(t))v_c(t)),
\end{align*}
\]

where \( v_c(t) = v_c(t) + v_{loss}(t) \) and \( i_E(t) = i(t) + i_{loss}(t) \). The signals \( v_{loss}(t) \) and \( i_{loss}(t) \) have been introduced to take into account the modeling errors (stray resistor and voltage drop across the IGBT type switch). Given the sources of the losses, \( v_c(t), v_c(t) \) and \( i_E(t), i(t) \) have the same nature, that is, step-like. The different signals are assumed to belong to the following bounded sets

\[
\begin{align*}
v_c(t) &\in [v_{c_{\text{min}}}, v_{c_{\text{max}}}], \\
i_I(t) &\in [i_{i_{\text{min}}}, i_{i_{\text{max}}}], \\
v_c(t) &\in [v_{e_{\text{min}}}, v_{e_{\text{max}}}], \\
i_E(t) &\in [i_{e_{\text{min}}}, i_{e_{\text{max}}}], \\
i(t) &\in [u_{\text{min}}, u_{\text{max}}].
\end{align*}
\]

3 Control algorithm

In this section, the main theoretical development of the paper, namely a robust nonlinear state feedback control law for \( v_c \) is presented. The proposed algorithm is an integral high gain type controller, which has already been theoretically developed in the literature (see [42]). It should be noted that the explicit incorporation of an integral action, into the synthesis model, is common, whether for linear or nonlinear approaches. Usually, the integral action is introduced after shifting the state space coordinates that leads to the Brunovsky canonical form, because it greatly simplifies the analysis. Indeed, most of the existing nonlinear control are synthesized for systems under this form since every uniformly controllable and observable systems can be put into this form [43]. But, in presence of model uncertainties, the inverse transformation that is necessarily done on-line to calculate the value of the control variable, may cause a loss of the integral action effects. To cope with this difficulty and to ensure robust static performances, even in presence of uncertainties, the integral action is directly introduced in the original coordinates. This, then, leads to the following extended synthesis model

\[
\begin{align*}
\dot{v}_c(t) &= \frac{1}{C} ((1 - u(t))i_I(t) - i_E(t)), \\
\dot{i}_I(t) &= \frac{1}{L} (v_c(t) - (1 - u(t))v_c(t)), \\
\dot{u}(t) &= \mu(t),
\end{align*}
\]

where \( \mu(t) \) is the first derivative of the physical control variable \( u(t) \), and becomes the virtual input of the system. Another advantage of this structure is that it allows to deal with the negative effects of the actuator saturation on the control variable. This can now be easily solved using a standard anti-windup structure [44].

The presentation of the proposed algorithm is split in three parts. Firstly, a nonlinear control law is designed in the nominal case, that is to say, when the supply voltage and the output current are constant and a priori known. Secondly, an observer is designed in order to deliver an estimation of the variables \( v_c(t) \) and \( i_E(t) \) using the available on line measurements, namely \( v_c(t) \) and \( i(t) \). Finally, the two previous parts are combined to obtain both a robust control law that takes into account the changes in the voltage supply and the load and a diagnostic tool for the monitoring of these variables, the proposed control law is illustrated on Fig. 2.
3.1 Nominal case

In the nominal case, the supply voltage and the output current are constant and perfectly known. In this configuration, the power losses are equal to zero and system (3) can be written as:

\[
\begin{align*}
\dot{v}_c(t) &= f ((1 - u(t)) i(t) - i^*_e), \\
\dot{i}(t) &= \frac{1}{L} (v^*_e - (1 - u(t)) v_c(t)), \\
u(t) &= \mu(t).
\end{align*}
\]  

(4)

where \(v^*_e\) and \(i^*_e\) denote the nominal values of \(v_c(t)\) and \(i(t)\) respectively. The Brunovsky canonical form characterizes the class of nonlinear systems for which the nonlinearity has a triangular structure. For such a class of systems, the high-gain control law has been proposed in the literature using the duality with the high-gain type observer. The control law synthesis requires the existence of a diffeomorphism that puts the physical model into the Brunovsky form. For that purpose, let us consider the change of coordinates \(z = \psi(v_c, i, u; v_e, i_e)\) where \(\psi\) is defined by

\[
\psi(v_c, i, u; v_e, i_e) = \left( \begin{array}{c} \frac{C v^2_e + L i^2_e}{2 v^2_e + 2 i^2 e} + 2(2 v^2_e - v_e i_e) \\
\frac{2 v^2_e + 2 i^2 e}{2 v^2_e + 2 i^2 e} - 2(1 - u) (v^2_e + i^2 e) \end{array} \right).
\]  

(5)

Using (5), the physical model (4) can be put under the following canonical form

\[
\begin{align*}
\dot{z}_1(t) &= z_2(t), \\
\dot{z}_2(t) &= z_3(t), \\
\dot{z}_3(t) &= f(z(t); v^*_e, i^*_e) u(t) + \Phi(z(t); v^*_e, i^*_e),
\end{align*}
\]  

(6)

where \(f(z(t); v^*_e, i^*_e)\) and \(\Phi(z(t); v^*_e, i^*_e)\) have the following expressions

\[
\begin{align*}
f(z(t); v^*_e, i^*_e) &= f(z(t); v^*_e, i^*_e), \\
\Phi(z(t); v^*_e, i^*_e) &= \frac{2}{LC} \left( u - 1 \right)^2 \left( v^2_e - i^2 e \right).
\end{align*}
\]  

(7)

(8)
The nominal values \( v^c_e \) and \( i^c_e \) are considered as parameters, hence, the function \( \psi \) is seen as a function from \( \mathbb{R}^3 \) to \( \mathbb{R}^3 \), and \( \psi^{-1}(z; v^c_e, i^c_e) \) denotes the function verifying:

\[
\psi^{-1}(\psi(v_c, i_l, u; v^c_e, i^c_e); v^c_e, i^c_e) = (v_c, i_l, u),
\]

for all \((v_c, i_l, u)\) in \( \mathbb{R}^3 \).

The control objective can be expressed in this new basis as: the state vector \( z(t) \) must be maintained in the vicinity of the following constant reference vector \( z^* \in \mathbb{R}^3 \)

\[
z^* = \begin{pmatrix} C(v^c_e)^2 + L(i^c_e)^2 \\ 0 \\ 0 \end{pmatrix}.
\]

For the class of systems (6), the following high-gain type control law is proposed:

\[
\mu(t) = -\frac{1}{f(v_c, i_l, u; v^c_e, i^c_e)} \times k_c \times \left[ \lambda^3, 3\lambda^2, 3\lambda \right] \times (z - z^*),
\]

where \( k_c > 0, \lambda > 0 \) are synthesis parameters. The control law can now be expressed using the physical variables

\[
\mu(t) = -\frac{1}{f(v_c, i_l, u; v^c_e, i^c_e)} v,
\]

where \( v \) is given by

\[
v = k_c \left[ \lambda^3, 3\lambda^2, 3\lambda \right] \times \left( \psi(v_c, i_l, u; v^c_e, i^c_e) - \psi(v^c_e, i^c_e, v^c_e, i^c_e) \right).
\]

Semicolons are used in the arguments of the different functions to enhance the fact that \((v_c, i_l, u)\) are the main variables while \( v^c_e \) and \( i^c_e \) are only parameters.

The nominal values of \( i_l \) and \( u \), are denoted \( i^*_l \) and \( u^* \), and can be obtained by replacing their expressions in (4):

\[
i^*_l = \frac{i^c_e}{v^c_e}, \quad u^* = 1 - \frac{v^c_e}{v^c_e}.
\]

One has the following result of convergence.

**Proposition 1.** If \( k_c \) and \( \lambda \) are chosen sufficiently large, then the control law (14)-(15) ensures that the tracking errors \( \tilde{v}_c(t) = v_c(t) - v^c_e, \tilde{i}_l(t) = i_l(t) - i^*_l \) and \( \tilde{u}(t) = u(t) - u^* \) converge to zero where \( v_c(t), i_l(t) \) and \( u(t) \) are the solutions of equation (4), \( v^c_e \) is the specified output voltage, \( i^*_l \) and \( u^* \) are given by (16).
place all the eigenvalues of the linear part of the closed loop system at \(-\lambda\). Then \(\lambda\) is closely related to the closed loop time constant. Moreover, \(\lambda\) has to be greater than a lower bound in order to dominate the nonlinearity \(\dot{\phi}\). This lower bound is proportional to the Lipschitz constant of \(\dot{\phi}\).

It is worth noting that the different variables are bounded (see (2)), thus, if \(\lambda\) is taken too large, the overshoot may lead to a saturation.

**Proof.** See section A in the appendix.

Now let us consider the more realistic case where the converter is not perfect \((v_{\text{loss}} \neq 0, i_{\text{loss}} \neq 0)\) and the supply voltage and the load are unknown but piecewise constant. In that case, the expression of the function \(f\) can be deduced from (3) and it depends on the exact values of \(v_e(t)\) and \(i_e(t)\) instead of their nominal values. As a consequence, the control law (14) will not ensure exact feedback linearization and the closed loop dynamical performances will be degraded. To overcome this problem, a nonlinear observer is synthesized in the next section to give an estimation of \(v_e(t)\) and \(i_e(t)\) and thus of \(f\). These estimations can also be useful for diagnostic purposes and then achieve the second objective of the proposed monitoring algorithm.

### 3.2 Observer for the perturbation

An observer is designed in this section to estimate \(v_e(t)\) and \(i_e(t)\) in the case where the supply voltage \(v_c(t)\) and the output current \(i(t)\) changes are piecewise constant. These estimates can be used to get useful informations about the state of charge of the energy storage component which supplies the DC/DC converter [45]. Since the signals \(i_e(t)\) and \(v_e(t)\) are piecewise constants and the measured signals are \(v_c(t)\) and \(i(t)\), an observer, based on [46] is given by

\[
\begin{align*}
\dot{\hat{v}}_c(t) &= \frac{1}{C} \left( (1-u(t))\hat{i}_i(t) - \hat{i}_e(t) \right) - 2\theta (\hat{v}_c(t) - v_c(t)), \\
\dot{\hat{i}}_i(t) &= \frac{1}{L} (\hat{v}_e(t) - (1-u(t))\hat{v}_c(t)) - 2\theta (\hat{i}_i(t) - i(t)), \\
\dot{\hat{i}}_e(t) &= C\theta^2 (\hat{v}_c(t) - v_c(t)), \\
\dot{\hat{v}}_e(t) &= -L\theta^2 (\hat{i}_i(t) - i(t)), \\
\end{align*}
\]  

(17)

where \(\theta > 1\) is the sole observer design parameter.

**Proposition 2.** Assume that the input \(u(t)\) belongs to a bounded set \(\mathcal{U}\) for all \(t > 0\), then there exists \(\theta_0 > 1\) such that for all \(\theta > \theta_0\) the observation error converges asymptotically to zero, independently of the control signal, as follows:

\[
\exists \theta_0 > 0, \quad \forall \theta > \theta_0, \exists \alpha > 0, \exists \mu_0 > 0, \forall u \in \mathcal{U}, \forall \dot{x}(0) \in \mathbb{R}^4,
\]

such that

\[
\|\dot{x}(t) - x(t)\| < \alpha e^{-\mu_0 t} \|\dot{x}(0) - x(0)\|,
\]  

(18)

where \(\lim_{\theta \to +\infty} \mu_0 = +\infty\), with

\[
x = [v_c, i_i, i_e, v_e]^T, \quad \text{and} \quad \dot{x} = [\dot{v}_c, \dot{i}_i, \dot{i}_e, \dot{v}_e]^T.
\]

**Remark 2.** As for the control law, the eigenvalues of the linear part of the observer error system are located at \(-\theta\). This design parameter must be greater than a lower bound \(\theta_0\), which is proportional to the Lipschitz constant of the nonlinear part.

**Proof.** See section B in the appendix.

Now that estimations of \(v_e\) and \(i_e\) are available, an estimation \(f_e(v_c, i_i, u; \hat{v}_e, \hat{i}_e)\) of \(f(v_c, i_i, u; v_e, i_e)\) is given as follows:

\[
f_e(v_c, i_i, u; \hat{v}_e, \hat{i}_e) = 2 \left( \frac{v_e \text{sat}_1(\hat{v}_e)}{L} + \frac{i_i \text{sat}_2(\hat{i}_e)}{C} \right),
\]

(19)
with

\[
sat_1(\sigma) = \begin{cases} 
    v_{\text{min}} & \text{if } \sigma < v_{\text{min}} \\
    \sigma & \text{if } \sigma \in [v_{\text{min}}, v_{\text{max}}] \\
    v_{\text{max}} & \text{if } \sigma > v_{\text{max}}
\end{cases}
\]

\[
sat_2(\sigma) = \begin{cases} 
    i_{\text{min}} & \text{if } \sigma < i_{\text{min}} \\
    \sigma & \text{if } \sigma \in [i_{\text{min}}, i_{\text{max}}] \\
    i_{\text{max}} & \text{if } \sigma > i_{\text{max}}
\end{cases}
\]

(20)

The saturating functions \(sat_1(\sigma)\) and \(sat_2(\sigma)\) have been introduced to ensure both that \(f_e(v_c, i_l, u; \hat{v}_E, \hat{i}_E)\) is a positive function, during the observer transient state and that \(f(v_c, i_l, u, v_c, \hat{i}_E) - f_e(v_c, i_l, u; \hat{v}_E, \hat{i}_E)\) converges to zero when \(v_c(t)\) and \(i_E(t)\) are constant [47].

3.3 Robust control law

In the non-nominal case, all the nominal values used in the control law equations (5)-(14)-(15)-(16) are replaced by the estimates of the corresponding variables. The scheme for the implementation of the control law is depicted on Fig. 2. The estimations of \(v_c(t)\) and \(i_E(t)\) can be directly obtained from observer (17) whereas (16) is replaced by:

\[
\dot{\hat{v}}_e(t) = \frac{\hat{i}_e(t)}{\hat{v}_e(t)} v_c^* \quad \hat{u}^*(t) = 1 - \frac{\hat{v}_e(t)}{v^*_c}.
\]

(21)

Using, in addition, the estimation \(f_e(v_c, i_l, u; \hat{v}_E, \hat{i}_E)\) delivered by (19), the state feedback (14)-(5) can be modified as follows:

\[
\mu = -\frac{1}{f_e(v_c, i_l, u; \hat{v}_E, \hat{i}_E)} \hat{v}_e.
\]

(22)

and

\[
ψ = k_c [3, 3\lambda^2, 3\lambda] \times \left(ψ(v_c, i_l, u; sat_1(\hat{v}_e), sat_2(\hat{i}_e)) - ψ(v_c^*, \hat{u}^*, sat_1(\hat{v}_e), sat_2(\hat{i}_e))\right)
\]

(23)

The main convergence result of the paper is stated below.

**Proposition 3.** The control law (17), (19), (22), (23) ensures that \(v_c(t)\), the solution of system (3), converges exponentially toward \(v_c^*\) provided that \(θ, k_c\) and \(λ\) are chosen sufficiently large.

**Remark 3.** The sole difference for the tuning of the design parameters compared to the nominal part concerns \(k_c\). Indeed, in the non nominal case, the multiplicative nonlinearity \(f\) is no longer exactly compensated. The parameter \(k_c\) is used in order to handle this issue and has to satisfy \(k_c > \frac{f_{\text{min}}}{2f_{\text{max}}}\) where \(f_{\text{min}}\) and \(f_{\text{max}}\) are respectively the lower and upper bounds of \(f\).

**Proof.** See section C in the appendix.

**Remark 4.** The conditions (51) and (56), established in appendix C, which ensure the convergence of the control algorithm may be very restrictive, since they provide a lower bound for the synthesis parameters \(k_c\) and \(λ\). Nevertheless, many simulations tests have shown that the proposed control law performs well for synthesis parameters values that are lower than the theoretical bounds.

4 Applications

The performances of the proposed state feedback control law have been firstly validated with simulations on MatLab/Simulink environment and compared to those obtained with a standard current mode controller [7]. For that purpose, a realistic boost converter simulator has been created to reproduce as closely as possible the behavior of the continuous system. Several stray inductors, resistors and capacitors associated to the connecting tracks, cables and semiconductor devices have
been introduced. The coil is highly nonlinear and displays hysteric behavior due to the magnetic cycle of its core. At usual operating points, effects of these phenomena and of the stray components are however mostly negligible. The IGBT has been considered as an ideal switch. Finally, the imperfections that have been taken into account in our simulation of the average model are the stray resistor of the coil and the saturation of the coil current. Please note that these operating points have usually been chosen in the literature in the more binding cases [17] and then gave us a satisfying benchmark test for the actual proposed control law. Table 1 reports the simulation parameters values which are the same than the experimental ones, presented next.

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Quantity</th>
<th>Value</th>
<th>Unity</th>
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<tbody>
<tr>
<td>$L$</td>
<td>Coil</td>
<td>2</td>
<td>mH</td>
</tr>
<tr>
<td>$C$</td>
<td>Capacitor</td>
<td>6.8</td>
<td>mF</td>
</tr>
<tr>
<td>$R_L$</td>
<td>Resistor</td>
<td>8</td>
<td>mΩ</td>
</tr>
<tr>
<td>$R_{esr}$</td>
<td>Resistor</td>
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<td>mΩ</td>
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<tr>
<td>$u_{max}$</td>
<td>Maximum control value</td>
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<td>-</td>
</tr>
<tr>
<td>$u_{min}$</td>
<td>Minimum control value</td>
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<tr>
<td>$v_c^e$</td>
<td>Nominal battery voltage</td>
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<td>Volts</td>
</tr>
<tr>
<td>$v_c^r$</td>
<td>Reference output voltage</td>
<td>24</td>
<td>Volts</td>
</tr>
<tr>
<td></td>
<td>PWM switch frequency</td>
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<td>kHz</td>
</tr>
</tbody>
</table>

Table 1: Experimental parameters values

4.1 Simulation results

First, let us examine the nonlinear control law design. Many design softwares are now available for the synthesis of linear control laws whereas this is not yet the case for nonlinear control schemes where the choice of the design parameters values is often a difficult task. The proposed control law gives a great answer to this challenge because the parameter values are readily determinable. Indeed, as long as $k_c$ remains constant, $\theta$ and $\lambda$ directly determine the values of the poles of the linear parts of the observer and of the state feedback respectively. The Fig. 3 shows the transient responses when the load switches from $R = 5\Omega$ to $R = 30\Omega$ at $75\,ms$ and switches back to $R = 5\Omega$ at $150\,ms$ for different values of $\theta$ and $\lambda$. The relationship between the values of these parameters and the response time is clearly shown there. Subject to compliance with conditions (18) and (56) an increase in the values of $\theta$ and $\lambda$ brings a better disturbance rejection but increases the closed loop sensitivity with respect to measurement noise.

![Figure 3](image)

**Fig. 3:** Simulated transient of the boost converter under a load step variation ($R = 5\Omega \rightarrow R = 30\Omega \rightarrow R = 5\Omega$) with different parameter values of the nonlinear controller.

To determine the best values of the synthesis parameters for the control law (22) with regard to the input/output closed loop performances, many simulations have been carried out using the previous simulator. The obtained values are:

$$\theta = 2000, \quad \lambda = 500, \quad k_c = 1.$$  \hspace{1cm} (24)
The performances of the control law (22) with the previous design parameters values have been evaluated in many configurations. For comparison purposes, a current mode controller has been designed. It was formed by a peak current modulator associated with a voltage control loop using a PI controller. This control structure is commonly used for static converters in the literature. A comparison of both algorithms allows us to quantify the performances of the nonlinear control law. The PI controller has been designed using the nominal average model and with specifications that guarantee a 60° phase margin and a 10 dB gain margin which are common specifications in power electronic [48].

A first experiment was conducted where the load changes from \( R = 5 \Omega \) to \( R = 30 \Omega \) and changes back to \( R = 5 \Omega \) with two operating point duty cycle values, respectively \( d = 0.5 \) (nominal value, Fig. 4a) and \( d = 0.7 \) (non-nominal value, Fig. 4b). Then, the performances of the control scheme have been tested in the presence of supply voltage variations (Fig. 4c). The step like corresponding profile consists of a change in level from \( v_c = 12V \) to \( v_c = 7V \) at \( t = 100 ms \) and again to \( v_c = 12V \) at \( t = 200 ms \).

The nonlinear control law obviously satisfies the transient and steady state performances requirements for both operating point duty cycles, while the current mode controller does not ensure the stability of the closed loop when working too far from the nominal operating point. Moreover, the nonlinear approach obviously gives the best settling time in the nominal case. It is worth pointing out that the converter, with the nonlinear algorithm, enters in discontinuous mode on figure 4b between \( t = 0.11s \) and \( t = 0.15s \). The robustness of our algorithm with respect to the change between continuous and discontinuous mode is due to the saturation functions which have been applied directly to the estimates provided by the observer (19).

Let us now examine the observer performances. This observer has first been designed to deliver an estimation of the supply voltage and the output current for monitoring purposes using the available measurements, namely, the output voltage and the coil current. An experiment has been conducted where the supply voltage changes from \( v_c = 12V \) to \( v_c = 7V \) at \( t = 250 ms \) and again to \( v_c = 12V \) at \( t = 550 ms \) (Fig. 5c). Moreover, for each supply voltage level, a load profile has been specified corresponding to a change from \( R = 5 \Omega \) to \( R = 30 \Omega \) and again to \( R = 5 \Omega \) (Fig 5d). Finally, a step-like output disturbance has been applied at \( t = 650 ms \) in order to simulate an offset on the sensor measurement. The observer performances are represented on Fig. 5a, 5b, 5c and 5d.

The convergence of the observer is clearly ensured for all the configurations that have been tested thanks to the specified disturbance profiles. The output voltage and the coil current are correctly estimated without any bias during all the steady state phases. A bias appears in the estimation of the supply voltage and the output current which can be easily explained by the model uncertainties that have been taken into account. In fact, the observer delivers estimations of both \( v_c(t) \) and \( i(t) \) instead of \( v_c(t) \) and \( i(t) \) respectively. In the actual simulation case, the difference between the previous signals is directly connected to the coil resistor that has been introduced in the simulator. A particular point needs to be emphasized on Fig. 5c and 5d. The sensor offset is responsible for an additional bias on the supply voltage and output current estimates. This test shows that the observer can be used to replace the sensors or to provide an oversight of the converter. Indeed, since \( v_c(t) \) and \( i(t) \) are constants, the differences between the measured and estimated signals are constant. A tabulation of the latter according to the variables can be used to measure the aging deviation of the components constituting the converter.
4.2 Experimental results

The proposed nonlinear control law has been tested on an existing DC/DC Converter. The experimental setup is composed of an Insulate Gate Bipolar Transistor (IGBT) power switches, a capacitor and an inductor. The physical parameter values are given in Table 1. This experiment was conducted with a sampling period of $0.75\,ms$, which is compatible with the imposed PWM switch frequency ($20kHz$), and the reference output voltage $v^*_c$ has been set to $24\,V$. Starting from the values (24) obtained in simulation with MATLAB/Simulink, the best input/output performances have been obtained with the following values:

$$\theta = 2000, \quad \lambda = 1500, \quad k_c = 1.$$  \hspace{1cm} (25)

The empiric rule for the choice of the parameters $\lambda$ and $\theta$ with respect to the sampling period is to have $T_s \lambda_{\text{max}}$ $\approx$ 1 and $T_s \theta_{\text{max}}$ $\approx$ 1 [49, 50]. Some new approaches allowing to overrule these limitation have recently been proposed in [51].

The control algorithm has then been implemented using the 1102 Dspace Environment. One can notice that the voltage and, especially, the current measurements are very noisy, indeed any anti-aliasing filters has been used. It is particularly visible on Fig. 6d, where the stabilized supply voltage measurements are shown.

The goal of this experiment is to show that our control law can keep the output voltage to the reference value even in presence of changes in load and supply voltage, all the results are presented on Fig. 6. The changes in the load and the supply voltage have been applied manually to the converter and can be seen on Fig. 6c and 6d.

The estimates of $v_c(t)$ and $i_c(t)$ are depicted on Fig. 6c and 6d. The difference between $v_c(t)$ and $v_c^*(t)$ clearly illustrates the power stage losses.

Similarly, the imposed variations of the output current are well estimated by the observer (see Fig. 6c). As shown in Fig. 6a, the regulation is offset position error free despite modeling errors, variations in supply voltage and load. The control objective is achieved with a slightly noisy input signal (Fig. 6c). The nonlinear process behavior is visible with the increase of noise on the output and input during the variation of the supply voltage ($3.2\,s - 4.5\,s$). The load variations cause a change in the power loss which results in an increase of the average value of the input signal (around $6.2\,s$). There is no important peak in the value of the control variables to compensate the absence of knowledge of the uncertain parameters. This allows to avoid saturation of the coil which would lead to a magnetic saturation and thus to a nonlinear behavior of the physical system, as shown on Fig. 6b.

5 Conclusion

In this paper, a nonlinear approach for controlling a static boost converter has been presented, with several improvements with respect to the previous approaches. The underlying control system ensures an offset free performance even in presence of uncertain parameters. Such a feature is achieved by a cancellation of the effect of a nonlinear term acting on the input which is appropriately estimated using a nonlinear high-gain observer. Three features are worth to be mentioned. Firstly, the involved estimates are of importance because they can be used for supervision purposes for batteries or super-capacitors [45]. Secondly, the implementation of the control law has been greatly simplified as the integral action is implemented directly in physical coordinates. And finally, The convergence has been proven and validated by probing experimental results.
A Proof of proposition 1

In the proof, for the sake of clarity, the dependence on the parameters $v^*_\epsilon$ and $i^*_\epsilon$ will be omitted, that is, the functions $\tilde{f}$ and $\tilde{\phi}$ will be denoted $\tilde{f}(z) = \tilde{f}(z; v^*_\epsilon, i^*_\epsilon)$ and $\tilde{\phi}(z) = \tilde{\phi}(z; v^*_\epsilon, i^*_\epsilon)$.

It is of particular importance to notice that the function $\tilde{f}(z)$ is positive on the physical working subspace because of the bound on the physical signals given by (2). Moreover, the function $\tilde{\phi}(z)$ is Lipschitz in its arguments over the domain of interest, since it is continuously differentiable and its arguments stay inside a bounded set, because of (2). The state vector $z(t)$ must be maintained in the vicinity of the constant reference vector $v^*$. This is equivalent to

$$\lim_{t \to +\infty} \tilde{z}(t) = 0,$$

where $\tilde{z}(t)$ is the tracking error defined by

$$\tilde{z}(t) = z(t) - z^*.$$  

Using (6), (12) and (27), the control model can be written in the following Brunovsky triangular canonical form

$$\dot{\tilde{z}}(t) = A_c \tilde{z}(t) + B_c \tilde{f}(z(t)) \mu(t) + \tilde{\phi}(z(t)),$$

with

$$A_c = \begin{pmatrix} 0_{2 \times 1} & I_{2 \times 2} \\ 0 & 0_{1 \times 2} \end{pmatrix}, \quad B_c = \begin{pmatrix} 0_{2 \times 1} \\ 1 \end{pmatrix},$$

$$\tilde{\phi}(z) = \begin{pmatrix} 0_{2 \times 1} \\ \tilde{\phi}(z) \end{pmatrix},$$

where $I_{2 \times 2}$, $0_{1 \times 2}$ and $0_{2 \times 1}$ are respectively the $2 \times 2$ identity matrix, the zero matrix of dimension $1 \times 2$ and the zero matrix of dimension $2 \times 1$.

The control law

$$\mu(t) = - \frac{1}{\tilde{f}(z(t))} \nu(\tilde{z}(t)),$$
provides an exponential convergence to zero of the tracking error for a sufficiently high value of the parameter synthesis \( \lambda \), according to Theorem 3.2 in [42], where

\[
\nu(\hat{z}(t)) = k_c B_c^T \Gamma_\lambda \hat{z}(t),
\]

(31)

\( k_c \) is a synthesis parameter, \( \Gamma_\lambda \) is the following diagonal matrix

\[
\Gamma_\lambda = \text{diag}(\lambda^3, \lambda^2, \lambda),
\]

(32)

and the matrix \( \bar{S} \) is the positive definite solution of the following algebraic Lyapunov equation:

\[
\bar{S} + A_c^T \bar{S} + S A_c = \bar{S} B_c B_c^T \bar{S}.
\]

(33)

This choice for the gain is a direct transposition of the high-gain observer from [52] exploiting the duality between control and observation. Let \( B_c^T = C_c \), since the algebraic Lyapunov equation \( S + A_c^T S + S A_c = C_c C_c^T \) has a unique symmetric positive definite solution \( S \), one can deduce that equation (33) has a unique symmetric positive definite solution \( S = TS^{-1}T \) with

\[
T = \begin{pmatrix}
0 & 0 & 1 \\
0 & 1 & 0 \\
1 & 0 & 0
\end{pmatrix}.
\]

With this choice for the Lyapunov equation, one has \( B_c^T S = C S^{-1}T = [1 \ 3 \ 3] \). This implies that all the eigenvalue of the matrix \( A_c - B_c B_c^T \Gamma_\lambda \) are equal to \( -\lambda \), this simplifies the implementation and the proof of convergence without introducing further constraints on the control performance.

In the original coordinates, the controller (30) is given by (14)-(5), which ends the proof.

**B Proof of proposition 2**

Our system can be written into the following form (studied in [46])

\[
\begin{aligned}
\dot{x}(t) &= A_o x(t) + \varphi(x(t), u(t)), \\
y(t) &= C_o x(t),
\end{aligned}
\]

(34)

where \( u(t) \) is the control variable, \( x(t) \) is the state of system (34), defined by

\[
x(t) = \begin{bmatrix} x^1(t) \\ x^2(t) \end{bmatrix} \in \mathbb{R}^4,
\]

(35)

with

\[
x^1(t) = \begin{bmatrix} x^1_1(t) \\ x^1_2(t) \end{bmatrix} = \begin{bmatrix} v_c(t) \\ i_i(t) \end{bmatrix},
\]

(36)

\[
x^2(t) = \begin{bmatrix} x^2_1(t) \\ x^2_2(t) \end{bmatrix} = \begin{bmatrix} -i_e(t)/C \\ v_e(t)/L \end{bmatrix},
\]

(36)

and the matrices \( A_o \) and \( C_o \) are defined by

\[
A_o = \begin{bmatrix} 0_{2 \times 2} & I_{2 \times 2} \\ 0_{2 \times 2} & 0_{2 \times 2} \end{bmatrix}, \quad C_o = [I_{2 \times 2} \ 0_{2 \times 2}].
\]

(37)
The function $\varphi(x(t), u(t))$ is globally Lipschitz with respect to $x(t)$ uniformly in $u(t)$ and is given by

$$
\varphi(x(t), u(t)) = \begin{bmatrix} \varphi^1(x(t), u(t)) \\ \varphi^2(x(t), u(t)) \end{bmatrix},
$$

with

$$
\varphi^1(x(t), u(t)) = \begin{bmatrix} (1 - u(t))x_2(t)/C \\ -(1 - u(t))x_1(t)/L \end{bmatrix},
$$

$$
\varphi^2(x(t), u(t)) = \begin{bmatrix} 0_{2 \times 1} \\ 0_{2 \times 1} \end{bmatrix}.
$$

Lemma 2.1 in [46] can be applied, since $v_\epsilon(t)$ and $i_\epsilon(t)$ are supposed piecewise constant, hence ensuring exponential convergence of the estimation error on every time interval where $v_\epsilon(t)$ and $i_\epsilon(t)$ are constants.

### C Proof of proposition 3

As in the proof of proposition 1, the system (3) is first transformed into the Brunovsky canonical form (6) through the change of coordinates $z = \psi(v_\epsilon, i_\epsilon, u, v_\epsilon, i_\epsilon)$:

$$
\begin{align*}
\dot{z}_1(t) &= z_2(t), \\
\dot{z}_2(t) &= z_3(t), \\
\dot{z}_3(t) &= \tilde{f}(z(t); v_\epsilon, i_\epsilon)\mu(t) + \Phi(z(t); v_\epsilon, i_\epsilon),
\end{align*}
$$

where $\tilde{f}$ and $\Phi$ are defined as in (7) and (9) respectively. For the sake of clarity, the notation $\tilde{f}(z)$ and $\Phi(z)$ is considered instead of $\tilde{f}(z; v_\epsilon, i_\epsilon)$ and $\Phi(z; v_\epsilon, i_\epsilon)$ in the rest of the proof.

The reference trajectory, is no longer constant in the non-nominal case, it is given by

$$
z^*(t) = \psi(v_\epsilon^*, i_\epsilon^*(t), u^*(t), v_\epsilon(t), i_\epsilon(t)),
$$

$$
= \psi \left( v_\epsilon^*, \frac{i_\epsilon(t)}{v_\epsilon(t)}, v_\epsilon^*, 1 - \frac{v_\epsilon(t)}{v_\epsilon^*}, v_\epsilon(t), i_\epsilon(t) \right),
$$

$$
\triangleq \kappa(v_\epsilon^*, v_\epsilon(t), i_\epsilon(t)).
$$

Similarly, the estimate of the reference trajectory $\hat{z}^*(t)$ is given by

$$
\hat{z}^*(t) = \psi(v_\epsilon^*, i_\epsilon^*(t), \hat{u}^*(t), sat_1(\hat{v}_\epsilon(t)), sat_2(\hat{i}_\epsilon(t))),
$$

$$
= \kappa(v_\epsilon^*, sat_1(\hat{v}_\epsilon(t)), sat_2(\hat{i}_\epsilon(t))),
$$

$$
= \begin{bmatrix} C(v_\epsilon^*)^2 + L(sat_2(\hat{i}_\epsilon(t))/sat_1(\hat{v}_\epsilon(t)))v_\epsilon^* \end{bmatrix},
$$

where the function $sat_1$ and $sat_2$ are define in (20).

The aim of this proof, is to show that the error $\tilde{z}(t) = z(t) - z^*(t)$ converges exponentially to zero on any time interval where $z^*(t)$ is constant. Note that saying that $z^*(t)$ is constant is equivalent to say that $v_\epsilon(t)$ and $i_\epsilon(t)$ are constant since $z^*(t)$ is a function of $v_\epsilon^*, v_\epsilon$ and $i_\epsilon$.

In order to do that, it is sufficient to prove that the next two signals

$$
\begin{align*}
\tilde{z}_e(t) &= z(t) - \hat{z}^*(t), \\
\tilde{z}_o(t) &= \hat{z}^*(t) - z^*(t),
\end{align*}
$$

(41)
converge to zero, since \( \xi(t) = \tilde{z}_e(t) + \tilde{z}_o(t) \).

**Convergence of \( \tilde{z}_o(t) \)**

The signal \( \tilde{z}_o(t) \) can be seen as the observation error on the reference trajectory. It is directly related to the observation error since:

\[
\|\tilde{z}_o(t)\| = \|\kappa(v^*_e, sat_1(\hat{v}_e(t)), sat_2(\hat{z}(t))) - \kappa(v^*_e, v^*_e(t), \hat{z}(t))\|.
\] (42)

The function \( \kappa \) is Lipschitz in its arguments over the domain of interest, then one can use the mean value theorem which states that there exists \( L_1 > 0 \) such that

\[
\|\tilde{z}_o(t)\| \leq L_1(\|sat_1(\hat{v}_e(t)) - v^*_e(t)\| + \|sat_2(\hat{z}(t)) - \hat{z}(t)\|),
\] (43)

Since \( v^*_e(t) \) and \( \hat{z}(t) \) belong to the unsaturated domain and the \( sat() \) function is 1-Lipschitz, equation (43) becomes

\[
\|\tilde{z}_o(t)\| \leq L_1(\|v^*_e(t) - v^*_e(t)\| + \|\hat{z}(t) - \hat{z}(t)\|).
\] (44)

Using the convergence of the observer, one obtain that if \( \theta \) is taken sufficiently large, one has

\[
\|\tilde{z}_o(t)\| \leq L_2\alpha\theta e^{-\mu_0}(\|v^*_e(t) - v^*_e(t)\| + \|\hat{z}(t) - \hat{z}(t)\|).
\] (45)

where \( \alpha \) and \( \mu_0 \) are given by proposition 2.

**Convergence of \( \tilde{z}_e(t) \)**

According to (40) and (41), the error \( \tilde{z}_e(t) \) is ruled by the following dynamical system:

\[
\dot{\tilde{z}}_e(t) = A_e \tilde{z}_e(t) + B_e \overline{J}(z(t)) \mu(t) + \overline{\Phi}(z(t)) - \hat{z}_e(t),
\] (46)

where \( A_e, B_e, \overline{J} \) and \( \overline{\Phi} \) are the same as in the proof of proposition 1. Similarly to the proposition 1, the control law is taken as:

\[
\mu(t) = -\frac{1}{\overline{J}(z(t); \hat{v}_e, \hat{\xi}_e)} k_e B_e^T \overline{\Gamma}_\lambda(\hat{z}_e(t) - \hat{z}_e(t)),
\] (47)

where \( \overline{J}(z(t); \hat{v}_e, \hat{\xi}_e) \) which will be simply denoted \( \overline{J}_e(z) \), in the sequel. The matrices \( \overline{\Gamma}_\lambda \) and \( \overline{S} \) are defined by (32) and (33) respectively and \( \hat{z} \) by \( \hat{z} = \overline{w}(v_e; \hat{v}_e, \hat{\xi}_e) = \overline{\Phi}(z(t)) - \hat{x} \)

Note that the corrective term in the proposed control law is not \( \tilde{z}_e \), because \( z \) is not available, indeed, it depends on \( v_e \) and \( i_e \). Let \( \overline{z}_e = \overline{\Gamma}_\lambda \tilde{z}_e \), then

\[
\overline{z}_e = \lambda A_e \overline{z}_e - k_e \frac{\overline{J}(z(t); \hat{v}_e, \hat{\xi}_e)}{\overline{J}(z(t); \hat{v}_e, \hat{\xi}_e)} \lambda B_e B_e^T (\overline{z}_e + \hat{z}_e(t) - \hat{x}(t) + \hat{z}(t)) + \overline{\Gamma}_\lambda(\overline{\Phi}(z(t)) - \hat{x}(t))
\] (48)

since \( \Gamma_e A_e \Gamma_e^{-1} = \lambda A_e \) and \( \Gamma_e B_e = \lambda B_e \).

Let us denote \( V \) the following candidate Lyapunov function:

\[
V(\overline{z}_e) = \overline{z}_e^T \overline{S} \overline{z}_e,
\] (49)

its derivative is given by:

\[
\dot{V}(\overline{z}_e) = 2\overline{z}_e^T \overline{S} \dot{\overline{z}}_e + \overline{z}_e^T \overline{S} \dot{v}_e - 2k_e \frac{\overline{J}(z(t); \hat{v}_e, \hat{\xi}_e)}{\overline{J}(z(t); \hat{v}_e, \hat{\xi}_e)} \lambda B_e B_e^T (\overline{z}_e + \hat{z}_e(t) - \hat{x}(t) + \hat{z}(t)) + \overline{\Gamma}_\lambda(\overline{\Phi}(z(t)) - \hat{x}(t))
\] (50)
Denoting \( v_1=2\tilde{\sigma}_c^\top S_k z - 2k_1 T_c(z) \tilde{\sigma}_c^\top S_k B \tilde{\sigma}_c \), using the expression of \( \tilde{\Lambda}_c \) deduced from the algebraic Lyapunov equation (33) and taking \( k_c \) such that

\[
k_c > \frac{1}{2} \left( \frac{f_{\max}}{f_{\min}} \right),
\]

leads to

\[
V_1 \leq -\lambda V(z_c)
\]

where \( f_{\max}, f_{\min} > 0 \), are such that \( f_{\min} \leq f(v_c, i_t, u, v_E, \tilde{i}_E) \leq f_{\max} \), these bounds exist since \((v_c, i_t, u, v_E, \tilde{i}_E)\) belong to a bounded set given by (2).

Noticing that \( \phi(z^*) = 0 \) when \( z^* \) is constant, and denoting \( V_2 = 2\tilde{\sigma}_c^\top S_k \phi(z) - 2k_1 T_c(z) \tilde{\sigma}_c^\top S_k B \tilde{\sigma}_c \),

\[
V_2 = -2k_1 T_c(z) \tilde{\sigma}_c^\top S_k B \tilde{\sigma}_c \tilde{\sigma}_c^\top S_k \lambda (\hat{z} - z)
\]

using the Cauchy-Schwarz inequality and applying the mean value theorem, since \( \phi \) has a triangular structure and is Lipschitz over its definition range, gives the existence of \( c_1, c_2, c_3 > 0 \) such that

\[
v_2 \leq c_1 V(z_c) + c_2 \sqrt{V(z_c)} \| \Gamma_k z_c \| + c_3 \sqrt{V(z_c)} \| \hat{z} - z \|
\]

where \( c_1 \) and \( c_2 \) does not depend on \( \lambda \).

Denoting \( W(t) = \sqrt{V(z_c(t))} \), using inequalities (50), (52), (53) and Cauchy-Schwarz inequality for the term \(-2\tilde{\sigma}_c^\top S_k \phi \) yields

\[
W(t) \leq -\frac{\lambda - c_1}{2} W(t) + \frac{c_2}{2} \| \Gamma_k z_c \| + \frac{c_3}{2} \sqrt{V(z_c)} \| \hat{z} - z \|.
\]

It has already been proven that \( \tilde{z}_o(t) \) converges exponentially toward zero, the same can be done for \( \hat{z}^* \) (t) and \( \hat{z} - z \) by using the fact that

\[
\dot{\hat{z}}^* (t) = \dot{\tilde{z}}_o(t),
\]

\[
\| \hat{z} - z \| = \| \psi(v_c, i_t, u; sat_1(\hat{v}_E), sat_2(\hat{i}_E)) - \psi(v_c, i_t, u; v_E, i_E) \|
\]

\[
\leq L_3 (\| v_E - \hat{v}_E \| + \| i_E - \hat{i}_E \|),
\]

where \( L_3 > 0 \) exists since \( \psi \) is Lipschitz over its domain of definition. Thus, there exist \( \alpha, \beta > 0 \) such that

\[
W(t) \leq -\frac{(\lambda - c_1)}{2} W(t) + \alpha e^{-\beta t}.
\]

Note that \( \beta \) does not depend on \( \lambda \) but \( \alpha \) may depend on \( \lambda \). Furthermore, using the comparison lemma 3.4 in [53], one obtains :

\[
W(t) \leq W(0) e^{-\frac{(\lambda - c_1)}{2} t} + W(0) \int_0^t e^{-\frac{(\lambda - c_1)}{2} (t-s)} \alpha e^{-\beta s} ds,
\]

\[
\leq W(0) e^{-\frac{(\lambda - c_1)}{2} t} + \alpha W(0) e^{-\frac{(\lambda - c_1)}{2} t} \left[ e^{-(\beta - \frac{\lambda - c_1}{2}) t} - e^{-(\beta + \frac{\lambda - c_1}{2}) t} \right]_0,
\]

\[
\leq W(0) e^{-\frac{(\lambda - c_1)}{2} t} + \alpha W(0) \left( \frac{\lambda - c_1}{2} - \beta \right) \left( e^{\frac{\lambda - c_1}{2} t} - e^{\beta t} \right),
\]

\[
\leq W(0) e^{-\frac{(\lambda - c_1)}{2} t} + \alpha W(0) \left( \frac{\lambda - c_1}{2} - \beta \right) \left( e^{-\beta t} - e^{\lambda - c_1 t} \right).
\]
Finally, if $\lambda$ is chosen so that

$$\lambda > c_1, \quad (56)$$

then equation (56) implies the exponential convergence of $W$ toward zero and thus of $V$. If $\lambda$ is taken sufficiently large

$$\|z_c(t)\| \leq \alpha_0 e^{-\min(\lambda-c_1)\beta t} \quad (57)$$

where $\alpha_0 > 0$. This ends the proof.

References


