

Original Research Paper

Robust and Efficient Controller to Design a Standalone Source Supplied DC and AC Load Powered by Photovoltaic Generator

Hassan Abouobaida

Laboratory of Engineering Sciences for Energy,
National School of Applied Sciences of El Jadida, Chouaib-Doukkali University, Morocco

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Email: hassanabouobaida@gmail.com

Abstract: The present work describes the analysis, modeling and control of a cascade DC-DC and DC-AC power conditioning stage to control a output voltage to supply a DC and AC load systems. To maximize energy extracted from PV generator and control output voltage a RCC MPPT and backstepping controller are designed. The stability of the control algorithm is demonstrated by means of Lyapunov analysis. The achievement of the DC-DC and DC-AC conversions and the efficient PV's energy extraction are validated with simulation results.

Keywords: MPPT, Backstepping Controller, DC-DC Converter, DC-AC Inverter

Introduction

Photovoltaic (PV) is a technology in which the radiant energy from the sun is converted to direct current. The photovoltaic process produces power silently and is completely self-contained, as there are no moving parts. These systems can withstand severe weather conditions including snow and ice. Photovoltaic systems for different applications can be either stand alone or grid connected. In a standalone system, the load has no connection to the utility grid and often relies on a set of batteries to secure an energy supply at night and other times when the solar panels do not produce electricity. A utility interactive or grid connected system is employed in applications where utility service is already available. In this case, there is no need for battery storage because the power station can be used to supplement photovoltaic generation when the load exceeds the available PV output. The use of photovoltaic as the power source with regulated voltage (DC 12V-24 or AC 220V) is considered as one of the most important application of PV system. Photovoltaic powered DC system allows to adapt the output voltage to the load even for climatic conditions that may change.

A power conditioning system linking the solar array and the power DC or AC load is needed to facilitate an efficient energy transfer between them, this implies that the power stage has to be able to extract the maximum amount of energy from the PV and to control a voltage of DC load.

In order to extract the maximum amount of energy the PV system must be capable of tracking the solar arrays Maximum Power Point (MPP) that varies with the solar radiation value and temperature. Several MPPT algorithms have been proposed, namely, Perturb and Observe (P&O), incremental conductance, RCC algorithms (Casadei *et al.*, 2006; Liu *et al.*, 2013), etc. They differ from its complexity and tracking accuracy but they all required sensing the PV current and/or the PV voltage.

Several controller strategies have been used in the literature, citing the PID that is generally suitable for linear systems, the sliding mode for which the chattering problem and fuzzy logic adapted to systems without a mathematical model (Lalouni and Rekioua, 2009).

In this study, the problem of controlling switched power converters is approached using the backstepping technique. While feedback linearization methods require precise models and often cancel some useful nonlinearities, backstepping designs offer a choice of design tools for accommodation of uncertain nonlinearities and can avoid wasteful cancellations. The backstepping approach is applied to a specific class of switched power converters, namely dc-to-dc converters. In the case where the converter model is fully known the backstepping nonlinear controller is shown to achieve the control objectives i.e., input voltage tracking and voltage control of DC or AC load with respect to climate

change. The desired array voltage is designed online using a Ripple Correlation Current (RCC) MPP tracking algorithm (Barth and Pilawa-Podgurski, 2013). The proposed strategy ensures that the MPP is determined, the voltage of DC or AC load is controlled to its reference value and the close loop system will be asymptotically stable. The stability of the control algorithm is analysed by Lyapunov approach. The rest of the paper is organized as follows. The dynamic model of the global system is described in Section II. A backstepping controller is designed along with the corresponding closed-loop error system and the stability analysis is discussed in Section III. In Section IV, a simulation results proves the effectiveness of this approach with constant solar radiation and temperature.

MPPT System Modelling

The solar generation model consists of a PV array module, dc-to-dc boost converter, dc-to-dc buck converter as shown in Fig. 1.

PV Model

PV array is a p-n junction semiconductor, which converts light into electricity. When the incoming solar energy exceeds the band-gap energy of the module, photons are absorbed by materials to generate electricity. The equivalent-circuit model of PV is shown in Fig. 2. It consists of a light-generated source, diode, series and parallel resistances (Rekioua and Matagne, 2012).

Boost Converter Model

The dynamic model of the power boost converter presented in Fig. 3 can be expressed by an instantaneous switched model as follows Equation 1 and 2:

$$c_1 \dot{u}_{pv} = I_{pv} - I_{L1} \quad (1)$$

$$L_1 \dot{I}_{L1} = u_{pv} - (1 - u_1) u_{dc} \quad (2)$$

where, L_1 and I_{L1} represents the dc-to-dc converter storage inductance and the current across it, u_{dc} is the output voltage and u_1 is the switched control signal.

Using the state averaging method, the switched model can be redefined by the average PWM model as follows:

$$c_1 \dot{\bar{u}}_{pv} = \bar{I}_{pv} - \bar{I}_{L1} \quad (3)$$

$$L \dot{\bar{I}}_L = \bar{u}_{pv} - \alpha_1 \bar{u}_{dc} \quad (4)$$

where, α_1 is averaging value of $(1 - u_1)$, \bar{u}_{pv} and \bar{I}_{pv} are the average states of the output voltage and current of

the PV generator, \bar{I}_{L1} is the average state of the inductor current.

Buck Converter Model

The power converter here is like buck. It provides power to the DC load with a share of power supplied by the photovoltaic generator and the battery source. The main control objective of the buck converter is to accomplish constant voltage operation of the DC load. Figure 4 illustrate a buck converter connected to load.

Noticing that u_2 stand for the control signal of buck converter, the system can be represented by differential Equation 5 and 6:

$$c_2 \dot{u}_{s1} = I_{L2} - \frac{u_{s1}}{R_1} \quad (5)$$

$$L_2 \dot{I}_{L2} = u_2 u_{dc} - u_{s1} \quad (6)$$

where, u_{dc} and u_s designs a battery DC voltage and output buck converter voltage respectively.

L_2 , I_{L2} and R_L are the dc-to-dc converter storage inductance, the current across it and load resistor. u_2 is the switched control signal.

Using the state averaging method, the switched buck converter model can be redefined by the average PWM model as follows:

$$c_2 \dot{\bar{u}}_{s1} = \bar{I}_{L2} - \frac{\bar{u}_{s1}}{R_1} \quad (7)$$

$$L_2 \dot{\bar{I}}_{L2} = \alpha_2 \bar{u}_{dc} - \bar{u}_{s1} \quad (8)$$

where, α_2 is averaging value of u_2 and \bar{u}_s is the average states of the output.

Buck Inverter Model

The active power transfer from the PV panels is accomplished by regulating the voltage that supply the single phase AC load. The inverter operates as a Voltage-Control Inverter (VCI). Figure 5 illustrate a buck inverter connected to load.

Noticing that u_3 stand for the control signal of the inverter, the system can be represented by differential Equation 9 and 10:

$$c_3 \dot{u}_{s2} = i_{L3} - i_{R2} \quad (9)$$

$$L_3 \dot{i}_{L3} = (2u_3 - 1)u_{dc} - u_{s2} \quad (10)$$

where, u_{dc} , u_{s2} designs a battery voltage and load voltage respectively. i_{L3} , i_{L2} are inverter output current and load current respectively.

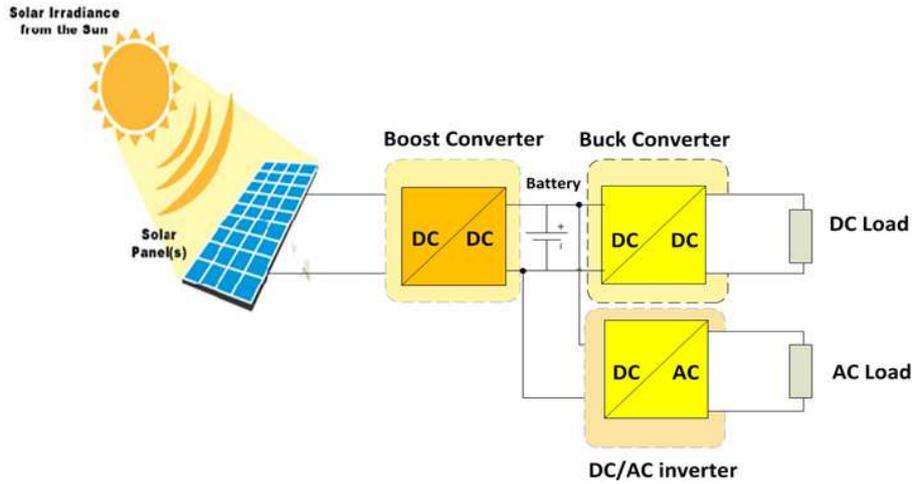


Fig. 1. Studied PV system

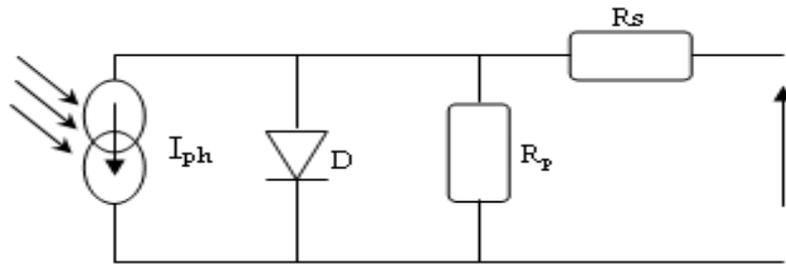


Fig. 2. Equivalent model of PV cell

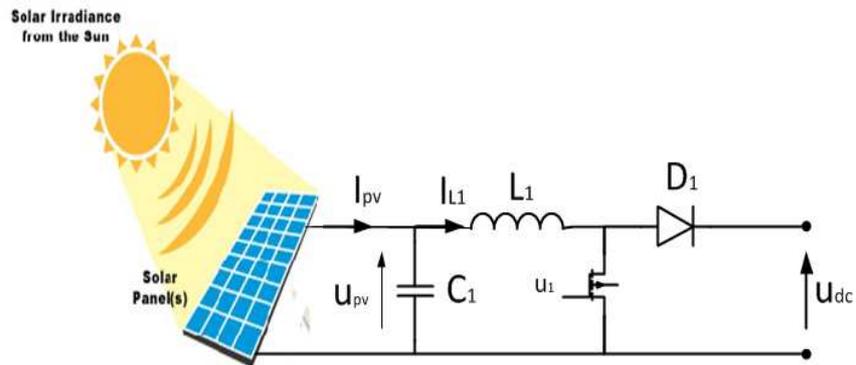


Fig. 3. PV generator and Boost converter

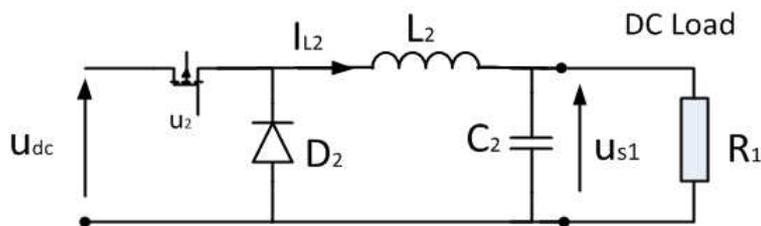


Fig. 4. Buck converter and DC load

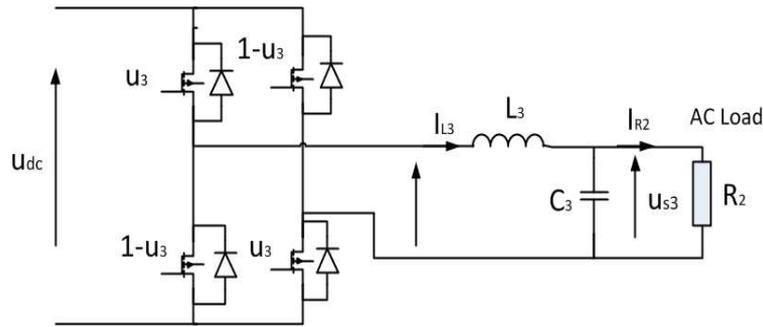


Fig. 5. Buck inverter

Using the state averaging method (on cutting period), the switched model can be redefined by the average PWM model as follows :

$$c_3 \cdot \dot{\bar{u}}_{s2} = \bar{i}_{L3} - \bar{i}_{R2} \quad (11)$$

$$L_3 \cdot \dot{\bar{i}}_{L3} = (2 \cdot \alpha_3 - 1) \cdot \bar{u}_{dc} - \bar{u}_{s2} \quad (12)$$

where, α_3 is averaging value of u_3

Nonlinear Controller Design

Two main objectives have to be fulfilled in order to transfer efficiently the photovoltaic generated energy into the DC and/or AC load are tracking the PV's Maximum Power Point (MPP) and control output voltage of DC and AC load. Figure 6 shows the control scheme used to accomplish the previous objectives.

Backstepping Controller to Extract Maximum Power

The boost converter is governed by control signal α_1 generated by a backstepping controller that allow to extract maximum of photovoltaic generator control by regulating the voltage of the photovoltaic generator to its reference provided by MPPT algorithm.

Step 1. Let us introduce the input error Equation 13:

$$e_1 = \bar{u}_{pv} - u_{pv}^* \quad (13)$$

Deriving e_1 with respect to time and accounting for (3), implies Equation 14:

$$\dot{e}_1 = \dot{\bar{u}}_{pv} - \dot{u}_{pv}^* = \left(\frac{\bar{I}_{pv}}{c_1} - \frac{\bar{I}_{L1}}{c_1} \right) - \dot{u}_{pv}^* \quad (14)$$

In Equation 14, i_{L1} behaves as a virtual control input. Such an equation shows that one gets $\dot{e}_1 = -k_1 \cdot e_1$ ($k_1 > 0$ being a design parameter) provided that:

$$I_{L1} = k_1 \cdot c_1 \cdot e_1 + \bar{I}_{pv} - c_1 \cdot \dot{u}_{pv}^* \quad (15)$$

As i_{L1} is just a variable and not (an effective) control input, (14) cannot be enforced for all $t \geq 0$. Nevertheless, Equation 12 shows that the desired value for the variable I_{L1} is Equation 16:

$$I_{L1}^* = k_1 \cdot c_1 \cdot e_1 + I_{pv} - c_1 \cdot \dot{u}_{pv}^* \quad (16)$$

Indeed, if the error Equation 17:

$$e_2 = I_{L1} - I_{L1}^* \quad (17)$$

Vanishes (asymptotically) then control objective is achieved i.e., $e_1 = u_{pv} - u_{pv}^*$ vanishes in turn. The desired value I_{L1}^* is called a stabilization function.

Now, replacing i_{L1} by $(e_2 + I_{L1}^*)$ in (14) yields Equation 18:

$$\dot{e}_1 = \left(\frac{\bar{I}_{pv}}{c_1} - \frac{\bar{I}_{L1}^* + e_2}{c_1} \right) - \dot{u}_{pv}^* \quad (18)$$

Which, together with (16), gives Equation 19:

$$\dot{e}_1 = -k_1 \cdot e_1 - \frac{e_2}{c_1} \quad (19)$$

Step 2. Let us investigate the behavior of error variable e_2 .

In view of (4) and (17), time-derivation of e_2 turns out to be Equation 20:

$$\dot{e}_2 = \dot{\bar{I}}_{L1} - \dot{I}_{L1}^* = \left(\frac{\bar{u}_{pv}}{L_1} - \frac{\alpha_1 \cdot \bar{u}_{dc}}{L_1} \right) - \dot{I}_{L1}^* \quad (20)$$

From (16) one gets Equation 21:

$$\dot{I}_{L1}^* = k_1 \cdot c_1 \cdot \dot{e}_1 + \dot{\bar{I}}_{pv} - c_1 \cdot \ddot{u}_{pv}^* \quad (21)$$

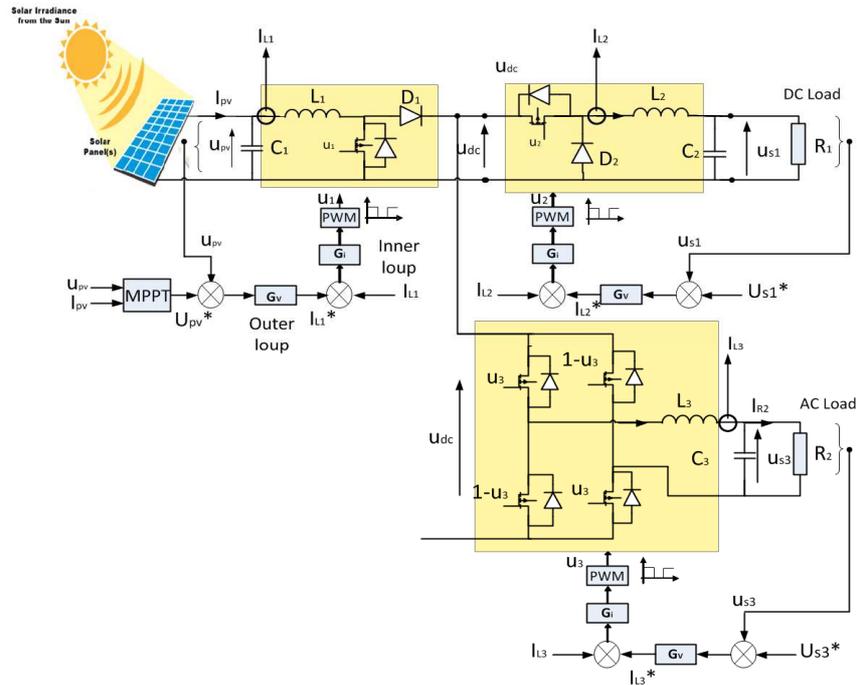


Fig. 6. Control strategy

Which together with (20) implies Equation 22:

$$\dot{e}_2 = \frac{\bar{u}_{pv}}{L_1} - \frac{\alpha_1 \bar{u}_{dc}}{L_1} - k_1 c_1 \dot{e}_1 - \dot{\bar{I}}_{pv} + c_1 \ddot{u}_{pv}^* \quad (22)$$

Buck Controller Design

This controller consists of an inner current loop and an outer voltage loop. The inner current loop is responsible to regulate the current of storage inductance in buck converter. The outer loop assures a steady-state control a output voltage.

Step 1. Let us introduce the input error Equation 23:

$$e_3 = \bar{u}_{s1} - u_{s1}^* \quad (23)$$

where, u_{s1}^* is a derivable reference signal of output voltage of DC load. Deriving e_3 with respect to time and accounting for (7) implies Equation 24:

$$\dot{e}_3 = \dot{\bar{u}}_{s1} - \dot{u}_{s1}^* = \frac{\bar{I}_{L2}}{c_2} - \frac{\bar{u}_{s1}}{R_2 c_2} \quad (24)$$

where, I_{L2} is a virtual control input. Such an equation shows that one gets $\dot{e}_3 = -k_3 e_3$ ($k_3 > 0$ being a design parameter) provided that:

$$I_{L2}^* = \frac{\bar{u}_{s1}}{R_1} - k_3 e_3 c_2 \quad (25)$$

As I_{L2}^* is just a variable (not effective) control input, Equation 25 cannot be enforced for all $t \geq 0$. Indeed, a new error is introduced Equation 26:

$$e_4 = \bar{I}_{L2} - I_{L2}^* = \bar{I}_{L2} - \left[\frac{\bar{u}_{s1}}{R_1} - k_3 e_3 c_2 \right] \quad (26)$$

Vanishes (asymptotically) then control objective is achieved i.e., $e_3 = \bar{u}_{s1} - u_{s1}^*$ vanishes in turn. The desired value I_{L2}^* is called a stabilization function.

Now, replacing I_{L2} by $(I_{L2}^* + e_4)$ in (24) yields Equation 27:

$$\dot{e}_3 = \frac{I_{L2}^* + e_4}{c_2} - \frac{u_{s1}}{R_1 c_2} \quad (27)$$

Which, together with (25), gives Equation 28:

$$\dot{e}_3 = -k_3 e_3 + \frac{e_4}{c_2} \quad (28)$$

Step 2. Let us investigate the behavior of error variable e_4 .

In view of (8), time-derivation of e_4 turns out to be Equation 29:

$$\dot{e}_4 = \dot{I}_{L2} - \dot{I}_{L2}^* = \left(\frac{\alpha_2 \bar{u}_{dc}}{L_2} - \frac{\bar{u}_{s1}}{L_2} \right) - \dot{I}_{L2}^* \quad (29)$$

From (25) one gets Equation 30:

$$\dot{i}_{L2}^* = \frac{\dot{\bar{u}}_{s1}}{R_1} - k_3 \cdot \dot{e}_3 \cdot c_2 \quad (30)$$

Substituting (30) in (29) implies Equation 31:

$$\dot{e}_4 = \frac{\alpha_2 \cdot \bar{u}_{dc}}{L_2} - \frac{\bar{u}_{s1}}{L_2} - \left(\frac{\dot{\bar{u}}_{s1}}{R_1} - k_3 \cdot \dot{e}_3 \cdot c_2 \right) \quad (31)$$

Control Design of VCI Inverter

This controller consists of an inner current loop and an outer voltage loop. The inner current loop is responsible of producing the averaging value of switched control signal. The outer voltage loop assures a steady-state maximum input-output energy transfer ratio and regulate a desired steady-state output voltage that supply the AC load.

Step 1. Let us introduce the input error Equation 32:

$$e_5 = u_{s2} - u_{s2}^* \quad (32)$$

where, u_{s2}^* is a reference signal of the output voltage.

Deriving e_5 with respect to time and accounting for (11) implies:

$$\dot{e}_5 = \dot{\bar{u}}_{s2} - \dot{\bar{u}}_{s2}^* = \frac{\bar{i}_{L3} - \bar{i}_{R2}}{c_3} - \dot{\bar{u}}_{s2}^* \quad (33)$$

In Equation 33, \bar{i}_{L3} behaves as a virtual control input. Such an equation shows that one gets $\dot{e}_5 = -k_5 \cdot e_5$ ($k_5 > 0$ being a design parameter) provided that:

$$\bar{i}_{L3} = -c_3 \cdot k_5 \cdot e_5 + c_3 \cdot \dot{\bar{u}}_{s2}^* + \dot{i}_{r2} \quad (34)$$

As \bar{i}_{L3} is just a variable and not (an effective) control input, (34) cannot be enforced for all $t \geq 0$. Nevertheless, Equation 34 shows that the desired value for the variable \bar{i}_{L3} is Equation 25:

$$\bar{i}_{L3}^* = -c_3 \cdot k_5 \cdot e_5 + c_3 \cdot \dot{\bar{u}}_{s2}^* + \dot{i}_{r2} \quad (35)$$

Indeed, if the error Equation 36:

$$e_6 = \bar{i}_{L3} - \bar{i}_{L3}^* \quad (36)$$

Now, replacing i_{L3} by $(e_6 + \bar{i}_{L3}^*)$ in (33) yields Equation 37:

$$\dot{e}_5 = \frac{(e_6 + \bar{i}_{L3}^*) - \bar{i}_{R2}}{c_3} - \dot{\bar{u}}_{s2}^* \quad (37)$$

Which, together with (35), gives Equation 38:

$$\dot{e}_5 = -k_5 \cdot e_5 + \frac{e_6}{c_3} \quad (38)$$

Step 2. Let us investigate the behavior of error variable e_4 .

In view of (12), time-derivation of e_6 turns out to be Equation 39:

$$\dot{e}_6 = \dot{\bar{i}}_{L3} - \dot{\bar{i}}_{L3}^* = \frac{(2 \cdot \alpha_3 - 1) \cdot \bar{u}_{dc} - \bar{u}_{s2}}{L_3} - \dot{\bar{i}}_{L3}^* \quad (39)$$

Which together with (26) implies Equation 40:

$$\dot{e}_6 = \frac{(2 \cdot \alpha_3 - 1) \cdot \bar{u}_{dc} - \bar{u}_{s2}}{L_3} + c_3 \cdot k_5 \cdot \dot{e}_5 - c_3 \cdot \ddot{\bar{u}}_{s2}^* - \dot{\bar{i}}_{r2} \quad (40)$$

In the new coordinates $(e_1, e_2, e_3, e_4, e_5, e_6)$, the controlled system is expressed by the couple of Equation 3, 4, 7, 8, 11 and 12. We now need to select a Lyapunov function for such a system. As the objective is to drive its states $(e_1, e_2, e_3, e_4, e_5, e_6)$ to zero, it is natural to choose the following function Equation 41:

$$V_2 = \frac{1}{2} \cdot e_1^2 + \frac{1}{2} \cdot e_2^2 + \frac{1}{2} \cdot e_3^2 + \frac{1}{2} \cdot e_4^2 + \frac{1}{2} \cdot e_5^2 + \frac{1}{2} \cdot e_6^2 \quad (41)$$

The time-derivative of the latter, along the $(e_1, e_2, e_3, e_4, e_5, e_6)$ trajectory is Equation 42 and 43:

$$\dot{V}_2 = e_1 \cdot \dot{e}_1 + e_2 \cdot \dot{e}_2 + e_3 \cdot \dot{e}_3 + e_4 \cdot \dot{e}_4 + e_5 \cdot \dot{e}_5 + e_6 \cdot \dot{e}_6 \quad (42)$$

$$\dot{V}_2 = -k_1 \cdot e_1^2 - k_2 \cdot e_2^2 - k_3 \cdot e_3^2 - k_4 \cdot e_4^2 - k_5 \cdot e_5^2 - k_6 \cdot e_6^2 \quad (43)$$

The equilibrium $(e_1, e_2, e_3, e_4, e_5, e_6) = (0, 0, 0, 0, 0, 0)$ is globally asymptotically stable, so doing, one gets the following control law Equation 44 to 46:

$$\alpha_1 = \frac{L_1}{\bar{u}_{dc}} \left[\frac{\bar{u}_{pv}}{L_1} - \frac{e_1}{c_1} + k_2 \cdot e_2 - k_1 \cdot c_1 \cdot \dot{e}_1 - \dot{I}_{pv} + c_1 \cdot \ddot{u}_{pv}^* \right] \quad (44)$$

$$\alpha_2 = \frac{L_2}{\bar{u}_{dc}} \left[\frac{\bar{u}_{s1}}{L_2} - \frac{\dot{\bar{u}}_{s1}}{R_1} - k_3 \cdot \dot{e}_3 \cdot c_2 - \frac{e_3}{c_2} - k_4 \cdot e_4 \right] \quad (45)$$

$$\alpha_3 = \frac{1}{2} + \frac{L_3}{\bar{u}_{dc}} \left[\frac{e_5}{c_3} + \frac{\bar{u}_{s2}}{L_3} + c_3 \cdot \ddot{\bar{u}}_{s2}^* + \dot{\bar{i}}_{r2} - c_3 \cdot k_5 \cdot \dot{e}_5 - k_6 \cdot e_6 \right] \quad (46)$$

Simulation Results

The PV model, DC-DC and DC-AC converters and the backstepping controllers are implemented in Matlab/Simulink as illustrated in Fig. 6. In the study, RSM-60 has been selected as PV source. The PV generator is chosen to deliver maximum 1kW by

connecting 16 module parallelly. The specification of the system and PV module are respectively summarized in the Table 1 and 2.

A Matlab/Simulink simulation of the PV system, the MPPT algorithm and the backstepping controller been carried out using the parameters in Table 2.

The studied PV system is evaluated on two aspects: The first one to extract the maximum power according to solar radiation and temperature change. The second one is the ability to regulate the outputs DC voltage and AC voltage to a references.

Figure 7a and b shows the irradiation and temperature change respectively.

Figure 7c shows the PV power properly following its maximum value according to irradiation and temperature change.

Figure 7d illustrates the output DC voltage properly following its reference value (12V or 24V).

Table 1. Main characteristics of the PV generation system

Maximum power	Output voltage at P_{max}	Open-circuit voltage	Short current circuit
60w	16v	21.5v	3.8A

Table 2. Control parameters used in simulation

Parameters	Value	Unit
c_1	220	μF
c_2	470	μF
c_3	22	μF
L_1	2	mH
L_2	2	mH
L_3	1	mH

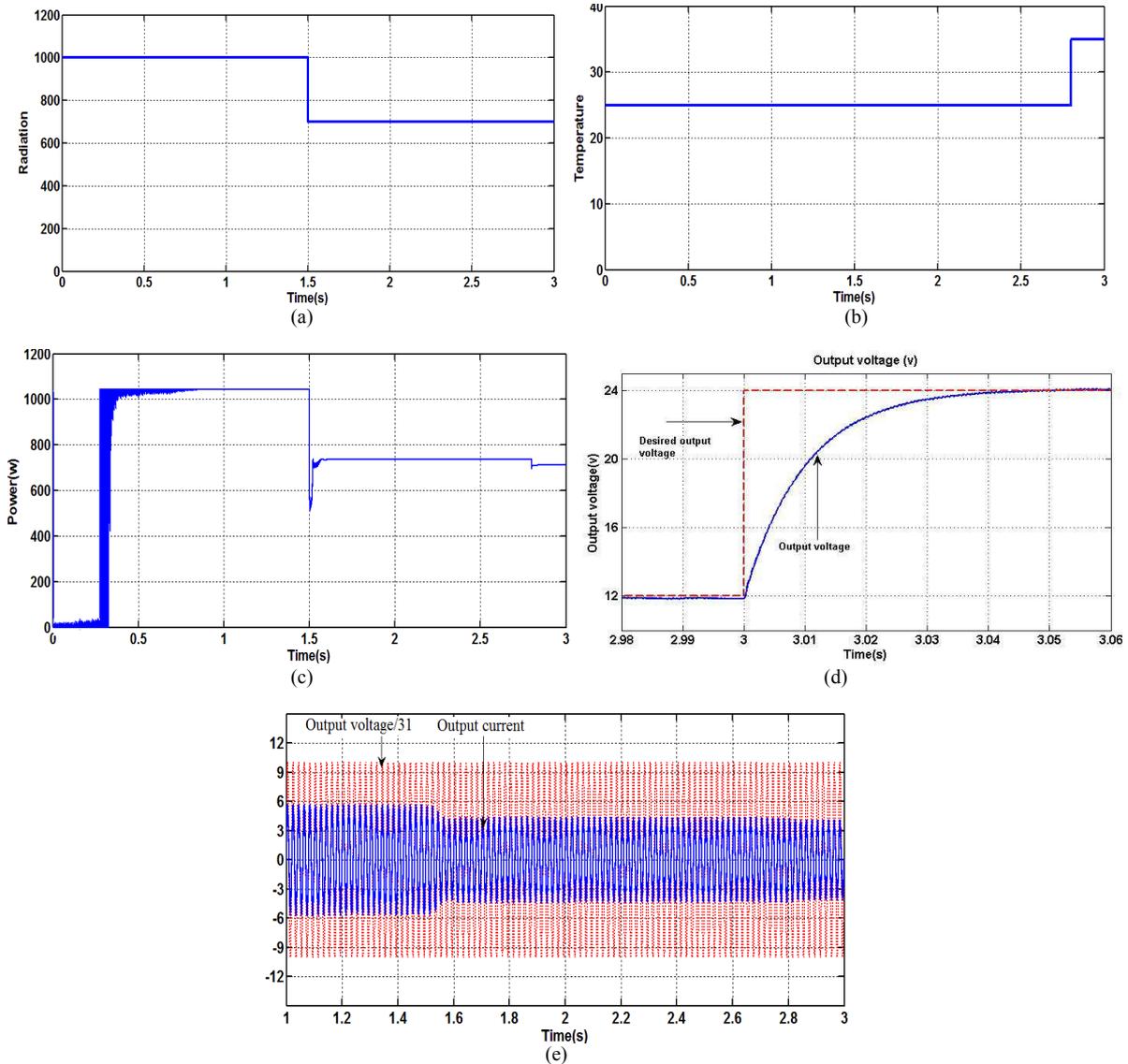


Fig. 7. (a) Radiation, (b) Temperature, (c) PV Power, (d) Output voltage u_{s1} , (e) Output voltage u_{s2}

Figure 7e shows that the output AC voltage and output current. This figure illustrates the output voltage well achieved to its reference (220V/50Hz).

These simulations show the effectiveness of the combination of backstepping controller approach and MPPT to search the maximum power and ability to adjustment of outputs voltages without influence of climate change.

Conclusion

A backstepping control strategy has been developed in the PV generating system. The first objective is to ensure operation at maximum power of solar panels. The second objective is to regulate the output voltage of the continuous and alternative load. The simulations results have shown that the objectives are achieved and that the PV system is stable and better efficiency against climate change.

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There is no funding for this work. this work is developed in a personal capacity

Ethics

This scientific research work is developed in the direction of improving the energy production for a global population increasingly growing and also to respond to the global economic demand. This work is considered part of a thesis on the ability of H. Abouobaida and will be published shortly.

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