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Smart Control of Buck Converters using a Switchingbased Clustering Algorithm

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Abstract— This paper proposes a new approach to the control of switching voltage regulators (buck converters). The method is performed using a switching-based clustering algorithm. The implementations of competing approaches such as a fuzzy-logic controller, proportional integral derivative controller and a neural network based controller are presented in order to compare and evaluate the performance of the switching-based clustering algorithm. The results of the approach show that the proposed method could improve the stability and the performance of the buck converter system by 2.7% in terms of settling time and by 0.6% in terms of the overshoot value as compared to other control methods for buck converters.

Keywords—control systems; power converters; unsupervised machine learning; switching; clustering

I. INTRODUCTION

As a type of switching voltage regulator, a buck converter, also known as a direct current (DC) step down converter, takes an input voltage, and outputs a lower voltage. Buck converters are extremely versatile in various applications where they are used to easily control an input voltage or input current regardless of a variety of factors such as frequency and temperature [1], [2]. Applications of the buck converter include the power supplies of laptops and mobile phones as well as the power type and level conversion from solar panels. As an example, in Figure 1a, a motherboard of a computer is shown that uses an on-board buck converter. A laptop which uses a buck converter in order to charge its battery is shown in Figure 1b. A foldable solar panel charger used by hikers which takes a non-linear input voltage and converts it into a useable voltage is shown in Figure 1c.



Figure 1a, 1b, 1c: Application areas of buck converters [3]

A more advanced buck controller system would allow for a more efficient and cost-effective use of the previous applications for the buck controller as well as an easier way to control the outputs of systems of all classifications. Recent literature regarding the control of buck converter systems has focused on the use of either proportional integral derivative (PID) controllers or fuzzy logic controllers that control the input to a pulse width modulator (PWM) generator in order to control the gain of a metal-oxide-semiconductor field effect transistor (MOSFET) [2], [4], [5]. This paper presents a comparison of the different controllers for a buck converter system and proposes a new, unsupervised machine learning controller which uses a switching-based clustering algorithm for the improvement of the stability of the buck converter output.

II. BUCK CONVERTER CONTROL

A conventional buck converter system consists of two switches, a MOSFET, a diode, and resistor-inductor-capacitor (RLC) elements that filter and lower voltage levels. The load resistor functions as an output load [2], [6]. In order to more effectively implement the conventional buck converter, the system equations and system differential equations were calculated.

A. State Space Equations

The state space equations of the buck converter system are given in (1), (2), and (3). The state variables were chosen to be the voltage through the capacitor $V_c(t)$ and the current across the inductor $i_L(t)$. The system differential equations that were found using Kirchhoff's voltage and current laws are given in (4) and (5), where u(t) is an input function. The output equation in terms of the state variables is given in (6). The transfer function, G(s) is generated from the system differential equations as given in (7), where the complex frequency $s = sigma + j^*omega$.

$$x(t) = [x_1(t), x_2(t)]$$
(1)

$$x_1(t) = V_c(t) \tag{2}$$

$$x_2(t) = i_L(t) \tag{3}$$

$$\dot{x}_1(t) = -\frac{1}{RC}x_1(t) + \frac{1}{C}x_2(t)$$
⁽⁴⁾

$$\dot{x}_2(t) = -\frac{1}{L}x_1 + \frac{1}{L}u(t)$$
⁽⁵⁾

$$y = V_R(t) = V_C(t) \tag{6}$$

$$G(s) = \frac{1}{s^2 CL + \frac{sL}{R} - 1} \tag{7}$$

B. Conventional Controllers

The PID controller is one example of conventional control approaches, which has been widely implemented and addressed in literature. The PID controller is implemented in series with a PWM generator leading to the gain of the MOSFET transistor [7], [8], [9]. The input to the PID controller was generated from the feedback loop of the converter.

As shown in Figure 2, the output of the controller was fed back and the error was calculated from a reference voltage of 3V. The output is given to the PID controller. The proportional gain of the controller was set to a value of 1. The integral gain of the controller was set to a value of 1. The derivative gain of the controller was set to a value of 0. The filter coefficient of the controller was set to a value of 100. Finally, the time domain of the controller was set to continuous time.

C. Fuzzy Logic Controller

The fuzzy logic controller is another competing approach to buck converter control seen in contemporary literature. The fuzzy logic controller was implemented so that its output is the gain of the MOSFET transistor. The fuzzy logic controller is a method of control that utilizes various membership functions in order to generate a set of values between 0 and 1 that correspond to the inputs given [1], [2].

As shown in Figure 2, the output of the controller is given to the PWM generator. The rules of the fuzzy logic controller were designed comparably to the inputs and outputs of the PID controller. In the controller, the number of samples for output discretization was set to a value of 101.

D. Neural Network Predictive Controller

The neural network predictive controller is another competing approach to buck converter control. In implementing the neural network predictive controller, the feedback voltage was given as the reference for the neural network system, the load voltage was given as the plant output for the neural network, and the control signal from the neural network model was set as the input to the PWM generator of the buck converter.

In the implementation, the input and output arrays of the controller were taken from the input and output of the PWM from the original buck converter model and the number of training epochs was set to 100. After training the input and output, the testing data, validation data, and training data for the controller were recorded.

III. PROPOSED CONTROL APPROACH - SWITCHING BASED CLUSTERING ALGORITHM

Clustering is an unsupervised machine learning technique in which data is grouped into clusters based on their similarity. The clusters are mutually exclusive [10]. The most common types of clustering algorithms used include K-means clustering and hierarchical clustering. Being able to form clusters based on observations rather than numerical comparisons makes clustering a useful unsupervised machine learning approach for larger amounts of data [10], [11].

Although adaptive controllers have been used in conventional buck converter systems in the past [11], [12], the use of switching-based clustering algorithm is a novel method of control.

A model of the proposed switching-based clustering controller with the buck converter is shown in Figure 2. The switching-based clustering algorithm groups the output voltage of the converter into three clusters based on similarity and then based on the clustering coefficients and indices, the switching frequencies of the PWM were adjusted. The switching-based clustering algorithm is shown in Algorithm 1, where the output needed to be converted to a time-series and then, it could be grouped into switching clusters.



Figure 2: A model of the switching-based clustering controller

Algorithm 1:Switching-based Clustering Algorithm	
Input: Output Voltage (ç)	
Output: Clustering Index (idx); Coefficient	(y)
voutData = vout.data;	
<i>VoutMatrix =</i>	
<pre>zeros(length(voutData),length(voutData));</pre>	
<i>VoutTimeseries =</i>	
<pre>zeros(size(vout,1),size(vout,1));</pre>	
<pre>for m = 1:length(voutData)</pre>	
<pre>VoutMatrix(m,:) = voutData;</pre>	
end	
<i>VoutTimeseries = timeseries(VoutMatrix)</i>	
<pre>function y = fcn(VoutTimeseries, vout)</pre>	
u2 = zeros(12, 1)	
u2(1:11,1) = VoutTimeseries;	
$u^{2}(12,1) = vout;$	
idx3 = kmeans(u2,3);	
idx = cluster(u2, 3);	
y = idx3(12);	
End	

After obtaining pertinent data with the original signal, a new reference voltage signal shown in Figure 3 was implemented that was set to pulse three times between the values of 3V and 5V. This new reference was implemented in place of the constant 3V into the gain of the MOSFET. The purpose was to determine if the system is able to learn from the previous pulsed input and decrease the fall time to the desired output voltage.



Figure 3: A pulsed reference voltage signal

IV. RESULTS

After implementing the PID controller, the fuzzy logic controller and the k-means clustering algorithm into the buck converter system, the following results were observed.

A. Implementation of the PID controller in a buck converter

After implementating the PID controller into the buck converter described in Section II, the load current and voltage are shown in Figure 4. The load current and voltage are obtained by implementing the PID controller with a single input to the gain of the MOSFET. The data from the load current and voltage is used when comparing the PID controller to the other controllers implemented.



B. Implementation of Fuzzy Controller in the Buck Converter

The results of the implementation of the fuzzy logic controller are shown in Figure 5. The load voltage and current values are obtained from implementing the fuzzy logic controller with a single input to the gain of the MOSFET.



C. Implementation of Switching-based Clustering Controller

The results of the implementation of the switching-based clustering algorithm are shown in Figure 6, Figure 7, and Figure 8. The load current and voltage are shown in Figure 6. The data from the load was used when comparing the proposed controller to the other controllers as given in Table 1 and Table 2.



Figure 6: Load current and load voltage

The MOSFET gate current and voltage is shown in Figure 7 and the input and the output power are shown in Figure 8. The input and the output power values are obtained from implementing the unsupervised machine learning controller with a single input to the MOSFET gate terminal.





A comparison of the load voltages and load currents of the original converter (in yellow), the PID controlled converter (in pink), the fuzzy logic controlled converter (in purple), and the switching-based clustering controlled converter (in blue) are shown in Figure 9.



Figure 9: A Comparison of the load currents and load voltages

The rise time, fall time, overshoot, initial value, peak value, settling time and final value of the outputs of the controllers that were implemented are compared in

Table 1. The rise time was calculated as the time required for the voltage to rise from its initial value to 90% of its peak value. The fall time was calculated as the time it takes for the voltage to go from peak value V_{peak} to 98% of its stabilized value V_{stable} . Settling time was calculated as the time needed for the voltage to go from the initial value to within 98% of its final or stabilized value. The overshoot (OS) was calculated as given in (8).

$$OS = V_{peak} - V_{stable} \tag{8}$$

Table 1: Stability Comparison of Output Voltage at Fixed Reference

	Feedback Control	PID Control	Fuzzy Logic Control	Switching- based Clustering
Rise Time	1.752e-4s	1.754e-4s	1.776e-4s	1.751e-4s
Fall Time	3.264e-4s	3.39e-4s	3.26e-4s	3.13e-4s
Overshoot	2.604V	2.634V	2.641V	2.588V
Initial value	0V	0V	0V	0V
Peak value	5.639V	5.64V	5.635V	5.639V
Final value	2.863V	2.868V	2.863V	3.010V
Settling time	5.016e-4s	5.144e-4s	5.036e-4s	4.881e-4s

The rise time, fall time, overshoot, initial value, peak value, and final value of the input to the PWM of the controllers that were implemented is compared in Table 2.

Table 2: Stability Comparison of PWM Input Voltage

	Feedback Control	PID Control	Fuzzy Logic	Switching- based Clustering
Rise Time	Os	Os	Os	3.84e-4s
Fall Time	2.24e-3s	2.18e-3s	1.64e-3s	5.999e-4s
Overshoot	3.010V	2.998V	2.874V	6.33e3V
Initial Value	2.99V	2.999V	2.876V	6.024e2V
Peak Value	2.99V	2.999V	2.876V	1.291e4V
Final Value	-1.09e-2V	5.09e-4V	2.15e-3V	6.582e3V

D. Implementation of PID Controller with a Pulsed Reference

The results of the PID controller implementation with the pulsed reference are shown in Figure 10. The acquired load current and voltage both were as expected due to the introduction of the three pulsed input to the gain of the MOSFET.



Figure 10: Load current and load voltage with pulsed reference

The MOSFET current and voltage with the pulsed reference are shown in Figure 11 and the input and output power with the pulsed reference is shown in Figure 12. The input and output power were also both as expected due to the introduction of the new three pulsed input to the gain of the MOSFET.







Figure 12: Input power and output power with pulsed reference

E. Implementation of Switching-based Clustering Controller with a Pulsed Reference Voltage

The results of the implementation of the unsupervised machine learning controller which uses the switching-based clustering algorithm with the new pulsed reference are shown in Figure 13, Figure 14 and Figure 15. The data obtained from the load current and load voltage were used when comparing the unsupervised machine learning controller to the other controllers implemented using the pulsed reference. The load current and voltage with the new input are shown in Figure 13.



Figure 13: Load current and load voltage with pulsed reference

The MOSFET current and voltage with the new reference are shown in Figure 14 and the input and output power with the new input are shown in Figure 15. The input and output power were also both as expected due to the introduction of the new three pulsed input to the switching of the MOSFET.



Figure 14: MOSFET current and voltage with pulsed reference



Figure 15: Input power and output power with pulsed reference

F. Implementation of fuzzy logic controller with pulsed input

The results of the implementation of the fuzzy logic controller with the new pulsed reference are shown in Figure 16. The data obtained from the figure of the load current and load voltage was used when comparing the fuzzy logic controller to the other implemented controllers.



A comparison of the load voltages and the load currents of the converter with the PID implemented (in pink), with the fuzzy logic controller implemented (in purple) and with the switching-based clustering implemented (in blue) for the system with the pulsed reference is shown in Figure 17.



pulsed reference

The rise time, fall time, overshoot, initial value, peak value, and final value of the controller outputs for the first two peaks recorded with the pulsed reference are compared in Table 3.

	PID	Fuzzy Logic	Switching-based Clustering
Rise Time-Peak 1	1.73e-4s	1.738e-4s	1.735e-4s
Fall Time –Peak 1	7.96e-4s	7.936e-4s	8.030e-4s
Overshoot- Peak 1	2.649V	2.651V	2.491V
Initial Value	1.2e-3V	1.2e-3V	1.2e-3V
Peak Value – Peak 1	5.634V	5.634V	5.585V
Final Value-Peak 1	2.894V	2.983V	3.094V
Rise Time-Peak 2	9.39e-4s	9.36e-4s	9.39e-4s
Fall Time –Peak 2	1.38e-3s	1.3874e-3s	1.404e-3s
Overshoot- Peak 2	3.155V	3.156V	2.940V
Peak Value - Peak 2	6.154V	6.156V	6.115V
Final Value-Peak 2	2.999V	3.00V	3.175V

Table 3: Stability Comparison of Output Voltage at Pulsed Reference

V. DISCUSSION

A comparison of the initial experiment with a constant reference voltage showed that the unsupervised machine learning controller with the switching-based clustering algorithm performed better than the other control approaches as shown in Table 4. The switching-based clustering controller had the quickest rise time and fall time of 1.75×10^{-4} s and 3.13×10^{-4} s respectively and had the lowest overshoot and a final value that was closest to the desired output voltage of 3V. The settling time of the PID, the fuzzy logic, and the switching-based clustering controller were 5.14×10^{-4} s, 5.04×10^{-4} s, and 4.88×10^{-4} s respectively. The switching-based clustering controller allowed for the fastest settling time as compared to the other approaches.

Table 4: A Summary of the	Best Controllers	for the Initial Test
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	Controller and Values
Rise Time	Switching-based Clustering 1.75x10 ⁻⁴ s (fastest)
Fall Time	Switching-based Clustering 3.13x10 ⁻⁴ s (fastest)
Overshoot	Switching-based Clustering 2.58V (lowest)
Initial Value	Same value for all approaches
Peak Value	PID 5.64V (highest)
Final Value	Switching-based Clustering 3.01V (closest to 3V)

A comparison of the second test with a pulsed reference is shown in Table 5. The fuzzy logic controller had the fastest rise time but also had the highest peak value. The PID controller had the fastest fall time. The switching-based clustering controller had the smallest overshoot value. The settling time for the PID, fuzzy logic and the switching-based clustering controller with the new three pulsed input to the gain of the MOSFET were 2.33×10^{-3} s, 2.32×10^{-3} s, and 2.34×10^{-3} s.

Table 5: A Summary of the Best Controllers for the Second Test

	Controller and Values
Rise Time	Fuzzy Controller 9.36x10 ⁻⁴ s (fastest)
Fall Time	PID 1.39x10 ⁻³ s (fastest)
Overshoot	Switching-based Clustering 2.94V (lowest)
Initial Value	Same value for all approaches
Peak Value	Fuzzy Controller 6.16V (highest)
Final Value	Fuzzy Controller 3V (closest to 3V)

VI. CONCLUSION

The smart controller approach presented in this paper was implemented using a new switching-based clustering algorithm. This method of control provided a stable voltage output more efficiently than competing methods. The results showed that the proposed method could improve the performance of the buck converter system by 2.7% in terms of its settling time and by 0.6% in terms of the overshoot value as compared to the other control methods for buck converters. Future work could address implementing the controller for higher voltage and multi-level converter systems.

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