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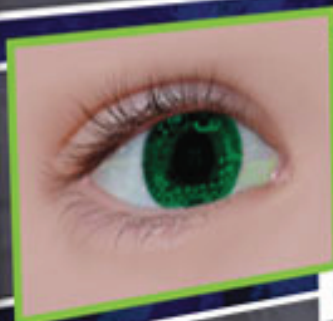
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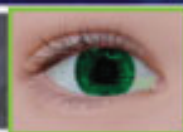
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ELECTRIC VEHICLE DRIVE SYSTEM WITH ADAPTIVE PID CONTROL

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ABSTRACT

The aim of this work is to implement an adaptive *PID* *SISO* feedback control to obtain a fine adjustment of an electric vehicle (*EV*) driving system. Our research work is done under different operating conditions, namely, variable battery voltage and variable load. A comparison between conventional and adaptive *PID* algorithms is established when they are applied to the above mentioned conditions. Experimental results indicate that the adaptive *PID* controller leads to a faster response and a better stability. Furthermore, the adaptive *PID* controller follows a given reference velocity faster and more smoothly than the conventional *PID* controller.

KEY WORDS

PID control, Electric vehicle, Drive system

1. Introduction

The development of *EVs* is motivated by global concerns over the need for environmental protection. McGraw-Hill's AccessScience Encyclopedia defines an *EV* as "a ground vehicle propelled by a motor that is powered by electrical energy from rechargeable batteries" [1].

In fact, *EVs* are playing an important role in solving problems caused by internal combustion engine vehicles (*ICEVs*) and are becoming more and more necessary all over the world. In recent years several researchers have investigated the use of *PID*, fuzzy logic, and adaptive control techniques for the driving control of *EVs* [2,3]. The drive system of *EVs* will suffer disturbances inevitably during its operating process. Therefore, in order to achieve an ideal dynamic behaviour, we must have good control techniques.

Though the use of *PID* control has been a long history in the field of control engineering, the three controller parameters are usually fixed. The disadvantage of the *PID* controller is its poor capability when dealing with system uncertainty, namely parameter variations and external disturbances.

Based on experimental data we show that classical linear control such as the *PID* algorithm cannot handle the

nonlinearities present in the system, resulting in a poor response.

This paper proposes an adaptive *PID* in order to stabilize the system and to minimize the error between the real velocity and the reference, in spite of the large perturbations in load rotational inertia in all kinds of driving cycles.

The paper is organized as follows. Section two presents the state of the art. Section three, provides an overview of the *EV* drive system. Section four analyses the traditional control, while section five studies the adaptive control. Finally, section six outlines the main conclusions.

2. State of the Art

With the development of navigation technologies for autonomous vehicles and the increase of the processing computers capacity, industrial mobile robot appeared as an important research area [4]. The *AGV* (Automatic Guided Vehicle) and *AS/RS* (Automatic Storage/Retrieval System) are devices with wider application in industry (Figure 1).



Figure 1. *AGV* and *AS/RS* in the industry

Rocha (Rocha et al. 2001) makes a description of mobile robots designed and installed in Portugal. The analysis is focused in *AGV* and *AS/RS* projects, making a classification of these systems and pointing-out the most relevant design, implementation and control aspects.

Butdee [5] presents an Automate Guided Vehicle (*AGV*)

involving an algorithm based on memorized path and the kinematics determination of the movement is developed. It is a three wheels vehicle. The front wheel is used for driving and steering with DC motors and the two rear wheels are free with encoders in order to measure the vehicle displacement, making possible to calculate its real time position and orientation. The choice of the positioning of the encoders on the free wheels provides to the vehicle an accurate measurement of its progression. A programmable logic control (PLC) is used and the control of the *AGV* motion is implemented through a *PID* algorithm. Displacement and the steering axis are separated to implement the motion control, while the position and orientation are estimated by a Kalman filter using a state-space model.

Kuo [6] proposes a novel adaptive sliding mode control with *PID* tuning method for a class of uncertain systems. The goal is to achieve the system robustness against parameter variations and external disturbances. Suitable *PID* control gain parameters can be computed on-line according to the developed adaptive law. To reduce the high frequency chattering in the switching part of the controller, the boundary layer technique adopted. The proposed method is applied to a brushless DC motor control system. Simulation results demonstrate that satisfactory trajectory tracking is achieved effectively and the input chattering is eliminated.

One hybrid method system that is the most popular presently is neuro-fuzzy systems, which applies a combination of artificial neural networks and fuzzy systems. In fact, authors have proposed the dynamic control using fuzzy logic and neural networks. This type of control is today well established in the area of motion control and particularly in drive systems.

Cao [7] describe a neuro-fuzzy control method for

navigation of *AGV* robot. The significance of Jin Cao work lies in the development of a new method for mobile robot navigation. The system to be controlled is an electrically propelled mobile vehicle, named *Bearcat II* (Figure 2), consisting in an intelligent control system.



Figure 2. The *Bearcat II* mobile robot

3. EV Drive System

This work is based on experimental data acquired from an *EV* controlled with a Programmable Logic Controller (PLC) because cost, reliability and safety are a major issue. The objective is to control the *EV* over the ethernet communicating with the Siemens PLC adopting the PROFINET standard.

The reference velocity is sent to a specific memory word of the PLC. The real velocity is calculated based on the encoder pulses of the traction motor (Figure 3).

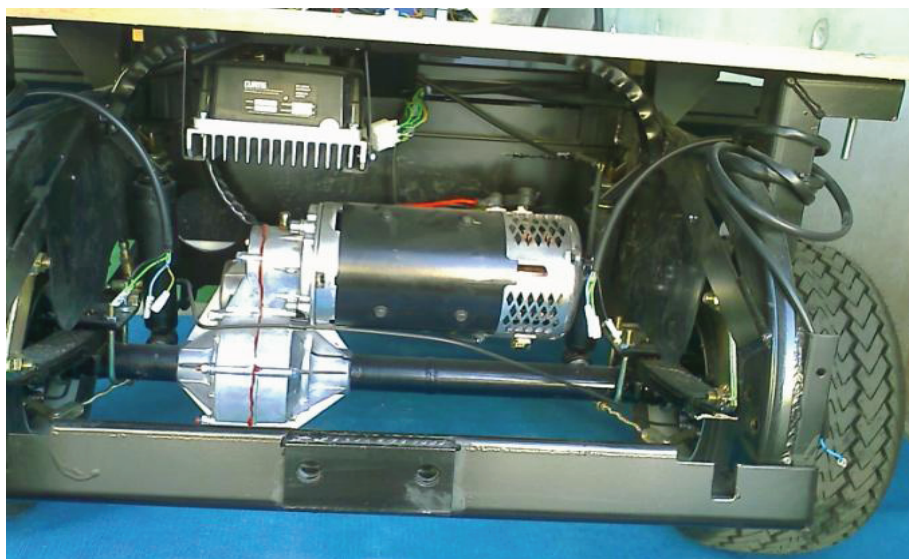


Figure 3. Traction motor



Figure 4. EV based on Melex golf cart

The electric vehicle is based on a *Melex* golf cart (Figure 4) with the characteristics presented in Table 1.

The *Melex* 48V Golf Cart is designed as a cost effective and reliable Golf Club Fleet cart with regenerative braking.

Table 1
EV Specifications

Dimensions [mm] Length x Width x Height	2350 x 1180 x 1750
Weigth [Kg]	550
Battery Voltage [V]	48
Motor [kW]	3
Maximum Velocity [km/h]	25

The difference between the reference velocity and the real velocity will be the error that needs to be minimized. In order to achieve a better response the controller is implemented in the *PLC* with Ladder. To optimize the time of the tuning process the designer can easily change the controller parameters online over the ethernet.

The traction motor velocity can be controlled with a DC signal logarithmically related with the desired velocity. In order to simplify the control algorithm it is considered a DC signal proportional to the desired velocity, since it is a good approximation for a specific load without any kind of disturbances.

To follow the reference of velocity faster it was decided to sum the controller response with the DC signal that would be necessary to follow the desired velocity leading to a feedforward scheme.

Figure 5 depicts the *SISO* block diagram of the *EV* drive system.

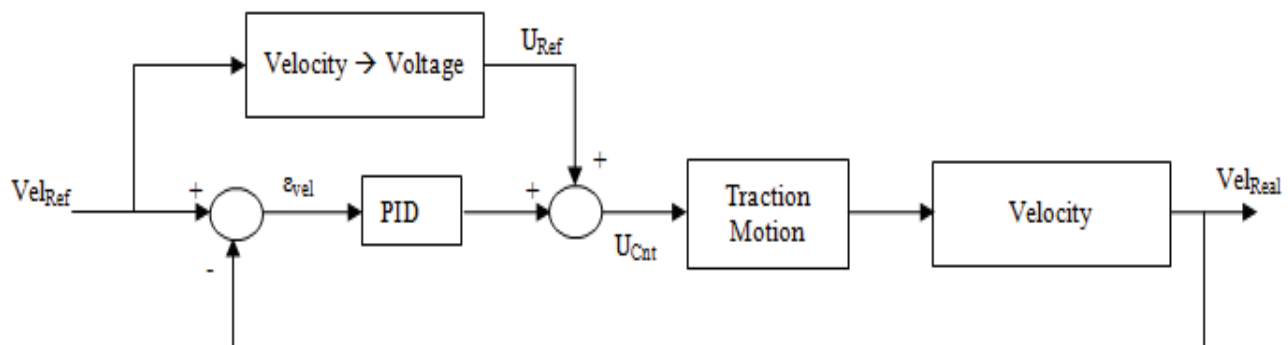


Figure 5. Control diagram

4. Classical Control

The classical *PID* are the most commonly used control algorithms in industry. Its parameters need to be adjusted according with the control process and remain unchanged during its operation. Among the various existing tuning schemes for *PID* controllers [8,9], the *Ziegler-Nichols* (*Z-N*) method is the most popular and is still extensively used for the determination of the *PID* parameters.

It is well known that the compensated systems, with controllers tuned by this method, have generally a step response with a high percent overshoot.

The transfer function of the *PID* controller is:

$$G_c(s) = \frac{U(s)}{E(s)} = K \left(1 + \frac{1}{T_i s} + T_d s \right) \quad (1)$$

where $E(s)$ is the error signal and $U(s)$ is the controller's output. The parameters K , T_i , and T_d are the proportional gain, the integral time constant and the derivative time constant of the controller, respectively.

The design of the *PID* controller using the second method of *Z-N* tuning consists on the determination of the optimum *PID* gains (K , T_i , T_d) based on the critical gain K_u and the corresponding period of sustained oscillation T_u (Table 2).

Table 2
Ziegler-Nichols recipe – Second method

Controller Type	K	T_i	T_d
<i>P</i>	$0.5 K_u$	∞	0
<i>PI</i>	$0.45 K_u$	$T_u/1.2$	0
<i>PID</i>	$0.6 K_u$	$T_u/2$	$T_u/8$

The critical gain K_u and the period of sustained oscillation T_u where experimentally acquired using only proportional feedback control ($T_i = \infty$ and $T_d = 0$). Increasing the proportional gain K leads to a state where sustained oscillation occurred at a velocity of 3 Km/h. Under those conditions, the critical gain obtained was $K_u = K = 400$ and the period of sustained oscillation $T_u = 3$ seconds.

Following the *Z-N* second method recipe for the *PID* controller we obtain the values represented in Table 3.

Table 3
***PID* controller parameters**

	K	T_i	T_d
<i>PID</i>	240	1.5	0.375

The values achieved with this method constitute a good first approach for a car operating velocity of 3 Km/h. However the controller does not yields a high performance.

In order to study the system dynamics under the action of the classical *PID*, during the operation we apply, separately, ramp pulses (ramp-up and ramp-down), at the reference.

The *PLC* returns the real velocity of the *EV* every 0.005 seconds.

The charts of Figure 6 depict the experimental results obtained.

Table 4 shows the time response characteristics of the system under the action of the classical *PID* controller, namely the percent overshoot *PO*, the rise time t_r , the peak time t_p and the settling time t_s .

Table 4
Time response parameters of the system
under the action of the classical *PID* controller

V_{refmax} (Km/h)	<i>PO</i> (%)	t_r	t_p	t_s
1	25.83	1.01	1.80	-
2	8.89	2.00	2.15	6.50
3	5.78	3.05	3.50	7.00
4	5.36	4.00	4.50	-
5	5.98	5.07	5.91	-

5. Adaptive Control

The classical *PID* would not be enough for the system we are handling because it isn't a linear system. The start up of the *PID* controller requires a not always simple work in the parameters adjustment, besides the existence of some methodologies [10]. Despite the help of these schemes it is necessary an observation period to survey the controller performance, which requires, in some cases, a substantial amount of time. This may be interpreted as a disadvantage, or a difficulty, in the controller start-up service.

In more complex cases, dynamical phenomena compromise the *PID* controller performance, making necessary to readjust the controller parameters.

What we will do next is to divide our nonlinear system in multiple linear subsystems based on the reference and real velocity. Like we previously made we will now obtain, for each different condition, a critical gain and a period of sustained oscillation.

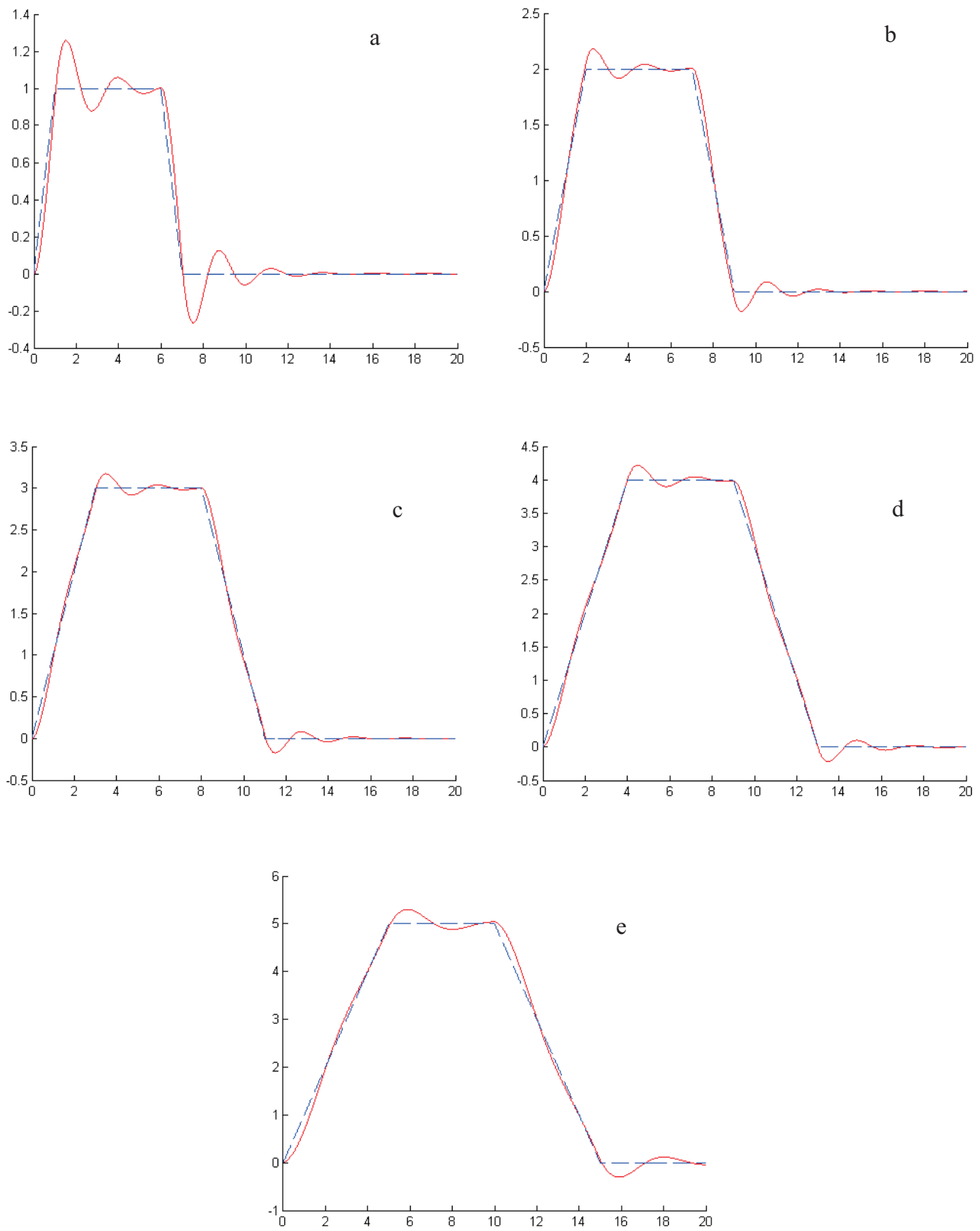


Figure 6. Time response of the EV under the action of the classical PID controller for ramp with an amplitude of a) $Vel_{Ref} = 1$ km/h b) $Vel_{Ref} = 2$ km/h c) $Vel_{Ref} = 3$ km/h d) $Vel_{Ref} = 4$ km/h e) $Vel_{Ref} = 5$ km/h

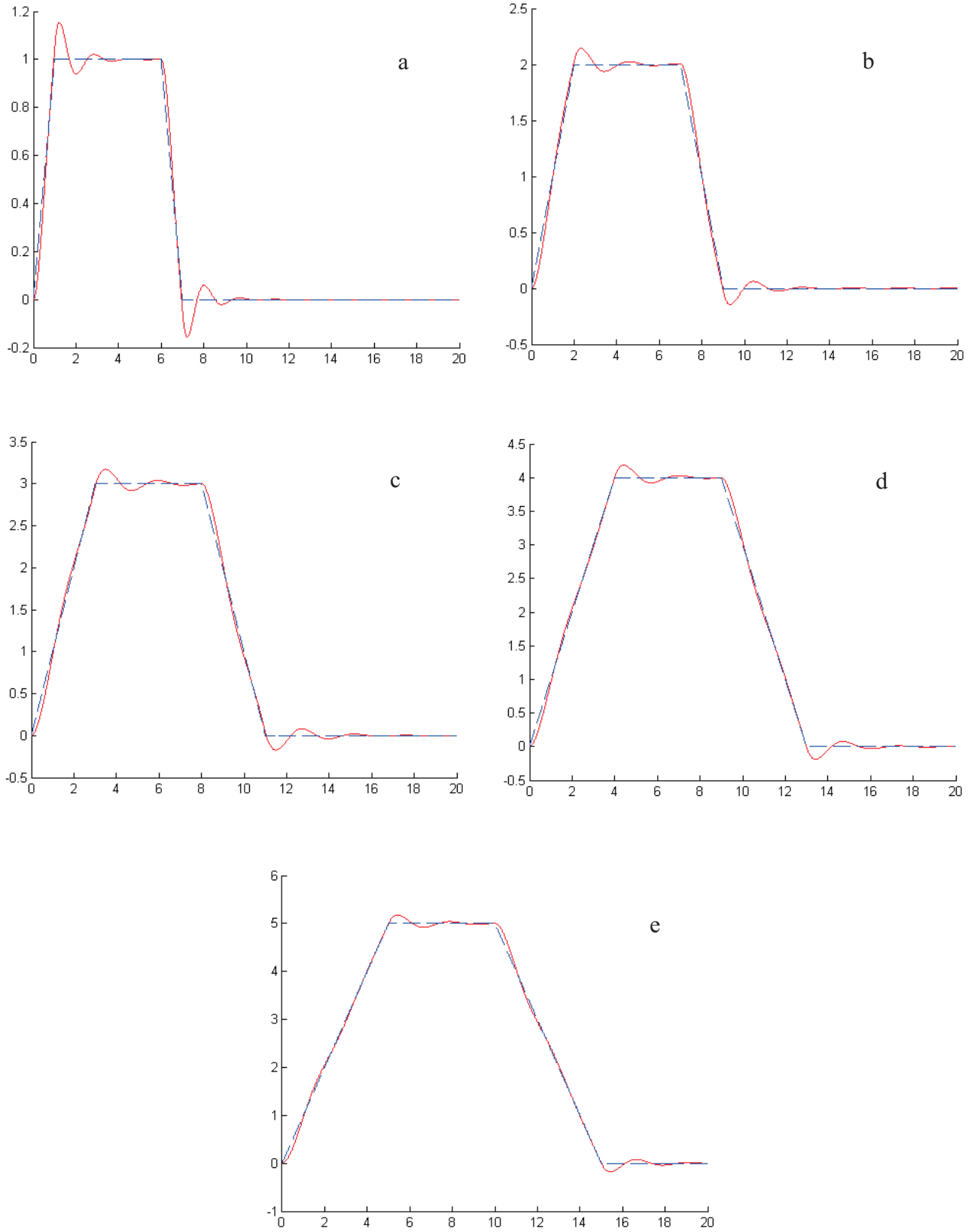


Figure 7. Time response of the EV under the action of the adaptive *PID* controller for ramp with an amplitude of
a) $Vel_{Ref} = 1$ km/h b) $Vel_{Ref} = 2$ km/h c) $Vel_{Ref} = 3$ km/h d) $Vel_{Ref} = 4$ km/h e) $Vel_{Ref} = 5$ km/h

Table 5 refers to the *PID* parameters obtained experimentally. For the proportional gain K_p we verify that it increases when the reference velocity Vel_{Ref} increases. The derivative time T_d will depend on the error between the reference velocity and the real velocity. The integral time will depend either the vehicle is accelerating or decelerating (in other words it will depend on the relation between the reference and the real velocity).

As previously analyzed with the classical *PID* we will now study the system dynamics under the action of the adaptive *PID* controller when applying the ramp signals at the reference (Figure 7).

Table 6 shows the time response characteristics of the adaptive *PID*, namely the percent overshoot PO , the rise time t_r , the peak time t_p and the settling time t_s .

Table 5
***PID* parameters experimentally obtained**

$Vel_{Ref} (Km/h)$	$Vel_{Real} (Km/h)$	K	T_i	T_d
$Vel_{Ref} \leq 1$	$\leq Vel_{Ref}$	50	1.5	$0.375 \cdot \varepsilon_{vel}$
	$> Vel_{Ref}$		5	
$1 < Vel_{Ref} \leq 2$	$\leq Vel_{Ref}$	150	1.2	$0.375 \cdot \varepsilon_{vel}$
	$> Vel_{Ref}$		6	
$2 < Vel_{Ref} \leq 3$	$\leq Vel_{Ref}$	200	1	$0.375 \cdot \varepsilon_{vel}$
	$> Vel_{Ref}$		7	
$3 < Vel_{Ref} \leq 4$	$\leq Vel_{Ref}$	250	0.8	$0.375 \cdot \varepsilon_{vel}$
	$> Vel_{Ref}$		8	
$Vel_{Ref} > 4$	$\leq Vel_{Ref}$	300	0.6	$0.375 \cdot \varepsilon_{vel}$
	$> Vel_{Ref}$		9	

Table 6
Time response parameters of the system under the action of the adaptive *PID* controller

$V_{refmax} (Km/h)$	$PO(\%)$	t_r	t_p	t_s
1	15.43	1.00	1.23	4.80
2	7.19	2.00	2.33	6.20
3	5.78	3.05	3.50	7.00
4	4.63	4.00	4.43	8.50
5	3.47	5.00	5.45	9.70

The adaptive *PID* controller presents better results than the classical *PID* controller revealing a smaller overshoot and a better time response to the perturbation.

6. Conclusion

A real-time adaptive *PID* digital controller has been developed that controls the speed of a DC motor in a closed loop manner. A Siemens PLC is used to implement the control.

In this paper were obtained satisfactory results in the velocity control of the EV proving that the implementation of the adaptive *PID* is more efficient than the classical *PID* and leads to a superior response.

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