

## Application of adaptive fuzzy logic controller to improve photovoltaic pumping system performances

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ARTICLE INFO	ABSTRACT
<ul> <li>Article Type: Research Article</li> <li>Article History: Received: 8 October 2022 Revised: 11 December 2022 Accepted: 13 December 2022 Published: 31 December 2022</li> <li>Editor of the Article: M. E. Şahin</li> <li>Keywords: Photovoltaic, Water pumping system, Adaptive fuzzy logic control</li> </ul>	This research presents an optimization for a photovoltaic pumping system with an adaptive fuzzy logic controller (AFLC). This maximization power point tracking (MPPT) approach is compared with the traditional control Perturb &Observ (P&O) to demonstrate its efficiency. After a sizing study, an application is made under two different days, to satisfy water consumption in Bejaia city. Different models can be used in photovoltaic-pumping system. In our case, we apply a model where the electrical power input of the motor-pump is directly function of the water flow output for various total heads. A simulation study under MATLAB/Simulink has been made using P&O and AFLC approaches. The outcomes demonstrate that applying the AFLC method, particularly for low changes of solar irradiation, improves the system pumping performance. Thus, the amount of power extracted is more and consequently the volume of the pumped flow has been rised. It is also noticed a gain in pumping time. The different results show the superiority of AFLC approach in terms of power, water flow and speed.

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### **1. INTRODUCTION**

Solar energy, which is widely accessible and cost-free, is used in many applications such as electrification, and water pumping, but due to the solar irradiance variations, optimization power techniques are used to increase photovoltaic power [1-2]. Maximum power point tracking (MPPT) is applied to continuously boost the solar panel's production of power. Many different MPPTs are used for PV generators [3-14]. It can be classical methods which are divided into direct or indirect methods. The most popular are perturb & observe (P&O) [4, 5, 15-21]. The most used fuzzy logic control (FLC) [3, 6, 7, 18, 19], the adaptive fuzzy logic control (AFLC) [1, 3, 7], the artificial neural networks (ANN) [1, 7, 14, 16, 20]. The perturb and observe (P&O) strategy is the most employed to determine the MPP point for PV systems. The FLC optimizes the increment magnitude to obtain fast and fine tracking. This method is widely used because of its advantages and fast response and gives better performances than the P&O, but the controller depends on speed and power variations [3, 11-12]. The AFLC is primarily used to change the FLC duty cycle for dealing with various external factors [3, 13-14].

This study compares an advanced strategy (AFLC) with a classical approach (P&O) before applying the two approaches to a solar pumping system. In our case, the model applied uses the

electrical power input of the motor pump directly function of the water flow output for various total heads. Finally, an application is made to provide a family with water for two different days in the Bejaia area made under two different days, to satisfy water consumption in Bejaia city. The two MPPT methods have been compared in a study. The AFLC method performance is demonstrated by simulation results that were achieved under solar irradiation variations. The most significant improvements are a gain in pumping time and an increase in power.

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## 2. PROPOSED SYSTEM MODEL

A photovoltaic pumping system includes a pumping subsystem and a photovoltaic generator. The system investigated in this article is depicted in Figure 1. PV panels, a DC/DC and a DC/AC converter, an induction motor connected to a centrifugal pump, and a water tank make up this device. For the optimization, two MPPT techniques are used (P&O and AFLC).

## 2.1. Photovoltaic Array Modelling

The model based on one diode and the equivalent circuit is represented in Figure 2. The equivalent diagram of a photovoltaic cell includes a current generator that models the solar irradiance and a diode in parallel that models the PN junction [3].



Fig.1. Studied photovoltaic pumping system.



Fig. 2. One diode equivalent circuit.

Current characteristics are given by the following equation [5]:

$$I_{pv} = I_{ph} - I_0 \left[ e^{q \frac{(V_{pv} + I_{pv}, R_s)}{AkT_j}} - I \right] - \frac{V_{pv}}{R_{sh}}$$
(1)

Where: *E* solar irradiation (W/m<sup>2</sup>),  $T_j$  junction temperature (°C),  $I_{ph}$  the photocurrent (A),  $I_d$  the junction polarization current (A), and  $R_s$ ,  $R_{sh}$  are respectively serial and shunt resistances ( $\Omega$ ).

The used photovoltaic panels have the following characteristics as shown in Table 1.

Table.1 Paramete	er of the used PV panel.	
Paramters	Values	
$P_{PV}$	110Wp	
$I_{mpp}$	3.15A	
$V_{mpp}$	35V	
I <sub>sc</sub>	3.45A	
$V_{oc}$	43.5V	
$\alpha_{\rm sc}$	1.4mA/°C	
$\beta_{oc}$	-152mV/°C	
$\mathbf{P}_{\mathrm{mpp}}$	110W	

Electrical characteristics are obtained by taking into account the effects of solar irradiance and temperature variations as shown in Figures 3, and 4.



Fig. 3. Effects of solar irradiation variations on  $P_{pv}$ - $I_{pv}$  characteristics.



Fig. 4. Effects of temperature variations on  $P_{pv}$ - $I_{pv}$ characteristics.

# 2.2. MPPT Algorithms 2.2.1. P&O algorithm

The most common approach is the P&O approach [2–12]. A deliberate panel voltage disturbance is made and the power obtained after the disturbance is compared with that power. In particular, when the power panel is increased as a result of the disturbance, the following disturbance will happen in the same direction. Additionally, a new disturbance is created in the opposite direction if the power decreases as shown in Figure 5. This approach is quite simple, knowing the characteristics of a solar generator is not necessary [8–11].



Fig. 5. P&O MPPT algorithm.

### 2.2.2. Adaptive fuzzy logic controller

The AFLC, which is an upgraded version of the FLC is primarily used to change the FLC duty cycle. The PV module's voltage and current are added to the preceding values to produce the average value. The structure of AFLC is given in Figure 6. As seen in the figure, the AFLC method is made up of two parts: a fuzzy basic learning controller and a learning mechanism. This one aims to study the environmental parameters and modify the FLC accordingly so that the global system response is close to the optimum point. As it is illustrated in Figure 6 the learning mechanism consists of an inverse fuzzy model and a modifier of the learning base.



Fig. 6. AFLC structure.

The fuzzy parameters can be adapted using the following condition: If error  $< \varepsilon$  (limit value) then the learning basis modifier will be chosen. When a fuzzy variable's scaling factor is changed, each membership function's definition will be changed; therefore, changing any scaling factor can change the meaning of any rule. Any scaling factor can alter the meaning of any rule since it affects how each membership function is defined when a fuzzy variable's scaling factor is altered as shown in Table 2. Changing peak values can improve both speed and stability. A large error (NM and PM) can improve speed. While a small error (NS and PS) can improve stability. Rule base modification can affect the control system such as overshoot, catch time, and stability. When membership functions of a fuzzy set are changed, it may affect some rule bases. However, when a rule is changed, only that rule is involved.

Converted fuzzy model		The modified learning base		
Error (e)	Variation of error	Peak membership	Scaling factor	
$-\varepsilon \langle e(k) \rangle \langle \varepsilon$	$-\varepsilon \langle Ce(k) \rangle \langle \varepsilon$	c(k)	$e(k)=e(k)*\delta_3$	
e( k )⟩ε	$-\varepsilon \langle Ce(k) \rangle \langle \varepsilon$	$c(k) + \delta_2$	Inchanged	
e( k )}ε	$Ce(k)\rangle\varepsilon$	$c(k) + \delta_1$	Inchanged	
e( k )}ε	$Ce(k)(-\varepsilon$	c(k)	$e(k)=e(k)*\delta_3$	
$e(k)\langle -\varepsilon$	$-\varepsilon \langle Ce(k) \rangle \langle \varepsilon$	$c(k)$ - $\delta_2$	Inchanged	
$e(k)\langle -\varepsilon$	$Ce(k)\rangle\varepsilon$	c(k)	$e(k)=e(k)*\delta_3$	
$e(k)\langle -\varepsilon$	$Ce(k)\langle -\varepsilon$	$c(k) - \delta_1$	Inchanged	

Where:  $\varepsilon$  is the minimum of the error, and c(k) is the triangle peak of membership *k*.

The PV module's voltage and current are added to the preceding values to produce the average value. To alter the fuzzy parameters and improve system performance, the error and variation errors of the system are used [13-14].

$$e(k) = \frac{P_{pv}(k+1) - P_{pv}(k)}{V_{m}(k+1) - V_{m}(k)}$$
(2)

$$Ce(k) = e(k+1) - e(k)$$
 (3)

Table 3 illustrates the controller Mamdani type with functions for membership in seven classes.

	Ta	ble. 3	AFLC	rules t	able.		
<b>F</b> ()	Variation error (Ce)						
Error (e)	NB	NM	NS	ZE	PS	РМ	РМ
NB	NB	NB	NM	ZE	ZE	ZE	ZE
NM	NB	NM	NM	ZE	NM	PS	PS
NS	NB	NB	NB	NB	PM	PS	PM
ZE	NB	NB	NS	ZE	PS	PM	PB
PS	NM	NS	ZE	PS	PM	PB	PB
РМ	NS	PB	PB	PB	PB	PB	PB
PB	ZE	PB	PB	PB	PB	PB	PB

Both the fuzzy logic control and the adaptive mechanism make up the AFLC technique. The fuzzification, fuzzy rules, and defuzzification as shown in Figure 7 (a-c) are three separate parts that make up the FLC [14–15].



Fig. 7. AFLC membership functions, (a) Input variable dP/dV,(b) Input variable DE, (c) Output variable Dd.

## 2.3. Pumping Subsystem Model 2.3.1.Machine modelling

The induction motor mathematical model is described as follows [15]:

$$\begin{cases} V_{S\alpha} = R_{st}I_{S\alpha} + \frac{d\Phi_{S\alpha}}{dt} \\ V_{S\beta} = R_{st}I_{S\beta} + \frac{d\Phi_{S\beta}}{dt} \end{cases}$$
(4)

where:  $I_{\sigma\alpha}$  and  $I_{s\beta}$  are for  $\alpha$ ,  $\beta$  stator currents respectively (A),  $\Phi_{s\alpha}$ ,  $\Phi_{\sigma\beta}$  are for  $\alpha$ ,  $\beta$  stator flux respectively (Wb), and  $R_{st}$  is the stator resistance ( $\Omega$ ).

$$\begin{cases} 0 = V_{R\alpha} = R_R I_{R\alpha} + \frac{d\Phi_{R\alpha}}{dt} + \frac{d\theta}{dt} \Phi_{R\beta} \\ 0 = V_{R\beta} = R_R I_{R\beta} + \frac{d\Phi_{R\beta}}{dt} - \frac{d\theta}{dt} \Phi_{R\alpha} \end{cases}$$
(5)

where:  $I_{R\alpha, I_R\beta}$  is for  $\alpha, \beta$  rotor current respectively (A),  $\Phi_{R\alpha}, \Phi_{R\beta}$  are for  $\alpha, \beta$  rotor flux respectively (Wb), and  $R_r$  is the rotor resistance ( $\Omega$ ).

$$\begin{cases} \boldsymbol{\Phi}_{S\alpha} = \boldsymbol{L}_{S}\boldsymbol{I}_{S\alpha} + \boldsymbol{L}_{m}\boldsymbol{I}_{R\alpha} \\ \boldsymbol{\Phi}_{S\beta} = \boldsymbol{L}_{S}\boldsymbol{I}_{S\beta} + \boldsymbol{L}_{m}\boldsymbol{I}_{R\beta} \\ \boldsymbol{\Phi}_{R\alpha} = \boldsymbol{L}_{R}\boldsymbol{I}_{R\alpha} + \boldsymbol{L}_{m}\boldsymbol{I}_{S\alpha} \\ \boldsymbol{\Phi}_{R\beta} = \boldsymbol{L}_{R}\boldsymbol{I}_{R\beta} + \boldsymbol{L}_{m}\boldsymbol{I}_{S\beta} \end{cases}$$
(6)

Mechanical equations:

$$T_{e} - T_{L} = J_{m} \cdot \frac{d\omega_{r}}{dt}$$

$$T_{e} = P \times (\phi_{s\alpha} \times I_{s\beta} - \phi_{s\beta} \times I_{s\alpha})$$
(7)

where  $T_e$  is the electromagnetic torque (N.m),  $T_L$  is the load torque (N.m),  $J_m$  is the inertia (N.m.s/rad), and P is the machine pole number.

#### 2.3.2. Pumping system modelling

The PV-pumping system uses a wide range of different types of pumps. So, different models can be used. In our case, we apply a model where the electrical power input (P) to the motor pump is directly a function of the water flow output (Q) for various total heads. A polynomial fit of the third order is used [6, 7]:

$$P(Q,h) = a(h)Q^{3} + b(h)Q^{2} + c(h)Q + d(h)$$
(8)

Where a(h), b(h), c(h), and d(h) are the coefficients of the working total head.

$$a(h) = a_0 + a_1 h^1 + a_2 h^2 + a_3 h^3$$
(9)

$$b(h) = b_0 + b_1 h^1 + b_2 h^2 + b_3 h^3$$
(10)

$$c(h) = c_0 + c_1 h^1 + c_2 h^2 + c_3 h^3$$
(11)

$$d(h) = d_0 + d_1 h^1 + d_2 h^2 + d_3 h^3$$
(12)

With  $a_i$ ,  $b_i$ , and  $d_i$  are solar pumping system constants (i=0, 1, 2, 3)

The Newton-Raphson method is used to compute the instantaneous flow in terms of power. The flow Q is therefore provided by the following equation at the  $k^{th}$  iteration: For d – Pa (Q)> 0:

$$Q_{k} = Q_{k-1} - \frac{F(Q_{k-1})}{F'(Q_{k-1})}$$
(13)

With:

$$F(Q_{k-1}) = a Q_{k-1}^{3} + b Q_{k-1}^{2} + c Q_{k-1} + d - P_{a}(Q_{k-1})$$
(14)

### **3. SIMULATION RESULTS**

The P&O algorithm and the AFLC are compared in terms of maximum power point (MPP) tracking at various test conditions to demonstrate the proposed MPPT's robustness as shown in Table 4.

Table. 4. Different tests of solar irradiance and ambient temperature.

Tests	$E(W/m^2)$	$T(^{\circ}C)$
Test 1	1000	25
Test 2	600	30
Test 3	300	20

MATLAB/Simulink is used for simulation. The different obtained results for each test are shown in Figures 7-18. The two MPPT strategies (P&O and AFLC) were used to compare water flow, rotational speed, and electromagnetic torque in three different tests. We notice that the pump starts faster when using AFLC, which improves the speed and the electromagnetic torque. We can also have a good pump operation for very low irradiation values which is very interesting.







When compared to the P&O approach, it is seen that the AFLC method responds more quickly whatever the different test conditions (low, medium, and high solar irradiations).

## 4. APPLICATION OF AFLC TO THE PUMPING SYSTEM

4.1. Sizing Photovoltaic Pumping System

The pumping system's sizing is carried out in two steps. The pumping sub-system must first be sized based on the flow rate, tank size, and dynamic total head. Then, an appropriate size of the PV generator must be determined based on whether conditions in the Bejaia area as in Figure.19.



The different equations used to calculate the PV energy are:

$$\begin{cases}
P_{Hydro} = \rho.g.H.Q_{v} \\
P_{mec} = P_{Hydro} / \eta_{pump} \\
P_{elect} = P_{elec} / \eta_{IM} \\
P_{inv} = P_{elec} / \eta_{inv} \\
\tau_{pump} = V_{tan k} / Q_{v} \\
E_{pv,m} = \tau_{pump}.P_{inv}
\end{cases}$$
(15)

Where  $P_{hydro}$  is the hydropower (W),  $P_{mec}$  is the mechanical power (W),  $P_{elect}$  is the electrical power (W),  $P_{inv}$  is the inverter input power (W),  $\tau_{pump}$  is the pumping time (hours/day),  $Q_v$  is the volumetric water flow (m<sup>3</sup>/s), *H* is the dynamic level head (m),  $\eta_{pump}$ ,  $\eta_{inv}$ ,  $\eta_m$  are respectively the pump, inverter and motor efficiency (%), and  $E_{pv,m}$  is the monthly photovoltaic energy (kWh/m<sup>2</sup>).

#### 4.2. Application

The photovoltaic pumping system's size has been determined. It comprises a 100 m<sup>3</sup> water tank to satisfy a family's domestic needs in the Bejaia area on two different days. The nominal flow rate is chosen to be  $34 \text{ m}^3/\text{h}=0.0094 \text{ m}^3/\text{s}$ , while the dynamic level head measures approximately 10 meters. The obtained results are summarized in Table 5.

Tabl	le 5. Sizing res	ults of the pumping sy	stem.
	Parameters	Results	
	P <sub>Hydro</sub>	926,50 W	
	P <sub>mec</sub>	1684.5 W	
	$\mathbf{P}_{elec}$	2105.62 W	
	$\mathbf{P}_{inv}$	2477.20 W	
	$ au_{ m pump}$	2.94 hours/day	
	E <sub>pv,m</sub>	9103.72 Wh/day	
	$\mathbf{P}_{\mathbf{pv}}$	3096.50 Wp	

MATLAB/Simulink was used to carry out the simulation investigation and the photovoltaic power. The chosen profile solar irradiance is given in Figure 20 (a) and the flow pump's results with and without MPPT are provided respectively in Figures 20 (b) and 20 (c). The AFLC technique improves the system pumping operation particularly in low changes of solar irradiation, as seen by the increased pumped flow from 0.005 m<sup>3</sup>/s to 0.006 m<sup>3</sup>/s. As the speed increases the extracted power becomes more significant, and the pumped flow follows. The operation starts before 2.5 s and before 4.5 s (on a time axis of 24 s) for high and low irradiation.

The system's operation is improved by using the AFLC method compared to direct coupling, especially under low solar irradiation, such that the extracted power is higher so the speed increases which increases the pumped flow rate.



Fig. 20. (a) Solar irradiance variations during two days, (b) Photovoltaic power variations with and without MPPT, (c) Water flow pump variations with and without MPPT.

### **5. CONCLUSION**

In this study, P&O and AFLC MPPT techniques used in solar pumping systems, have been compared. An application is made to satisfy a family's water needs in the Bejaia area. It has been observed that increased mechanical power caused by the AFLC results in increased water flow. This has improved the system pumping operation, particularly in low changes of solar irradiation, as seen by the increased pumped flow from 0.005 m<sup>3</sup>/s to 0.006 m<sup>3</sup>/s. As the speed increases the extracted power becomes more significant, and the pumped flow follows. The operation starts before 2.5 s and before 4.5 s (on a time axis of 24 s) for high and low irradiation. The control with the AFLC technique is superior in terms of power, water flow, and speed, according to the simulation results. It will be interesting in the future to compare the obtained simulation results to experimental ones.

## Nomenclature

Tomene	Autur C
E	Solar irradiation (W/m <sup>2</sup> )
$E_{pv,m}$	Monthly photovoltaic energy (kWh/m <sup>2</sup> )
$J_m$	Inertia (N.m.s/rad)
$I_{ph}$	Light-generated current (A)
$I_d$	Diode current (A)
$I_{Rlpha,} I_{Reta}$	$\alpha$ , $\beta$ rotor current (A)
$I_{Rsh}$	Shunt-leakage current (A)
$I_{s\alpha}, I_{s\beta}$	$\alpha$ , $\beta$ stator currents (A)
H	Dynamic level head (m)
Р	Pole number
$P_{ele}$	Electrical power (W)
$P_{hydro}$	Hydropower (W)
$P_{inv}$	Inverter input power (W)
$P_{mec}$	Mechanical power (W)
$R_s$	Series resistance ( $\Omega$ )
$R_{sh}$	Shunt resistance ( $\Omega$ )
$R_{st}$	Stator resistance ( $\Omega$ )
$R_r$	Rotor resistance ( $\Omega$ )
Т	Ambient temperature (°C)
$T_e$	Electromagnetic torque (N.m)
$T_{j}$	Junction temperature (°C)
$T_L$	Load torque (N.m)
$Q_{v}$	Volumetric water flow (m <sup>3</sup> /s)
$\Phi_{slpha}$	α stator flux (Wb)
$arPhi_{\sigma\!eta}$	$\beta$ stator flux (Wb)
$\Phi_{R\alpha}$	α rotor flux (Wb)
${\cal P}_{Reta}$	$\beta$ rotor flux (Wb)
ωr	Rotor angular speed (rad/s)
$ au_{pump}$	Pumping time (hours/day)
$\eta_{pump}$	Pump efficiency (%)
$\eta_{inv}$	Inverter efficiency (%)
n n	Motor efficiency (%)
<i></i>	

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