

Volume 11, Issue 2, February 2022



Impact Factor: 7.569









|| Volume 11, Issue 2, February 2022 ||

| DOI:10.15680/IJIRSET.2022.1102076 |

Comparative Analysis between Fuzzy Logic Controller (FCL) and Artificial Neural Network (ANN) based MPPT Methods on PV Module in MATLAB/Simulink

Sudhanshu Shrivastava¹, Dr. Pankaj Kumar Shrivastava²

PG Scholar, Department of Mechanical Engineering, AKS University, Satna, M.P., India ¹ Professor, Department of Mechanical Engineering, AKS University, Satna, M.P., India ²

ABSTRACT: There exist numerous maximum power point tracking techniques in today's market to maintain the operation of PV module at maximum power such as off-line techniques, on-line or hill-climbing (HC) techniques and artificial intelligence (AI) techniques. Numerous approaches for improving, adapting, and optimizing these techniques have been published. However, they differ in many aspects such as tracking speed, tracking accuracy, steady-state efficiency and dynamic efficiency, number of sensors used, complexity, and cost. These MPPT techniques fail or deviate from tracking the correct maximum power point (MPP) under sudden or ramp variations of solar irradiation and ambient temperature as well as under partial shading with oscillations around MPP. The maximum power point tracking-based computational intelligence techniques that properly detect the global MPP are stated. This thesis present various soft computing technique like artificial neural Network, fuzzy logic controller to achieve maximum power point tracking solution in PV module. Also, a comparative study for MPPPT in PV module by using two different artificial intelligent controllers which are fuzzy logic controller (FLC) and artificial neural network (ANN). Intelligent controllers have gained wide popularity in the application of controlling system output voltage. We develop MATLAB/Simulink. Model to performed different global condition for maximum utilized solar radiation to achieved maximum power output. and we developed MATLAB program for modelling of the sun position, solar radiation, etc

KEYWORDS: Photovoltaic module, Artificial Neural Network (ANN), fuzzy logic controller (FCL), Maximum power point tracking (MPPT)

I. INTRODUCTION

Solar photovoltaic energy systems (PVES) have been used to feed loads in remote areas as well as in central power plants connected to the electric utility. Many research works have been done to reduce the cost of the generated energy from the PVES. One of the most important factors of research work is to reduce the cost of generated energy by increasing the generated energy from the PVES by modifying its performance operations.[01,02,] This can be achieved by tracking the maximum power available from the PV systems which can increase the generated energy from PVES considerably and reduce its cost. The device used to track the maximum power available from the PVES is called maximum power point tracker (MPPT). MPPT uses a controlled technique to control the power electronics converters to be sure it extracts the maximum power available from the PVES. Conventional MPPT techniques like perturb and observe, hill climbing, incremental conductance, etc., have been introduced a long time ago, and they were working for unshaded PVES very well. But, in case of partial shading conditions, multi-peaks in the P-V curve of the PV array are generated, and these peaks may trap the conventional techniques to fall within one of the local peaks. [05] With the advances of artificial intelligence, soft computing, and metaheuristic techniques, this limitation can be overcome, whereas metaheuristic techniques can track the global peaks in shaded or unshaded PVES which became a new trend in the tracking of the maximum power of the PVES. Many metaheuristic techniques are introduced and discussed in this thesis like Artificial Neural Network (ANN), fuzzy logic controller (FCL) particle swarm optimization (PSO), gray wolf optimization (GWO), genetic algorithm (GA), ant colony optimization (ACO), whale optimization (WO), simulated annealing (SA) [6]

1.1 Maximum Power Point Tracking (MPPT)

Maximum Power Point Tracking is electronic tracking - usually digital. The charge controller looks at the output of the panels and compares it to the battery voltage. It then figures out what is the best power that the panel can put out to



|| Volume 11, Issue 2, February 2022 ||

| DOI:10.15680/IJIRSET.2022.1102076 |

charge the battery. Assume your battery is low, at 12 volts. An MPPT takes that 17.6 volts at 7.4 amps and converts it down so that what the battery gets is now 10.8 amps at 12 volts. Now you still have almost 130 watts, and everyone is happy. Ideally, for 100% power conversion you would get around 11.3 amps at 11.5 volts, but you have to feed the battery a higher voltage to force the amps in. And this is a simplified explanation - in actual fact, the output of the MPPT charge controller might vary continually to adjust for getting the maximum amps into the battery.[7]

II. MODELING OF THE SOLAR SOURCE

The annual orbit of the earth around the sun . The distance between the earth and the sun changes throughout the year, the minimum being 1.471×1011 m at winter solstice (December 21) and the maximum being 1.521×1011 m at summer solstice (June 21). The year-round average earth—sun distance is 1.496×1011 m. The amount of solar radiation intercepted by the earth, therefore, varies throughout the year, the maximum being on December 21 and the minimum on June 21.

Reckoning of time	The time when the sun crosses the meridian of the observer is the local solar noon. It usually does not coincide with the 12:00 o'clock time of a locality			
Equation of time	Due to factors associated with the earth's orbit around the sun, the earth's orbital velocity varies throughout the year, so the AST varies slightly from the mean time kept by a clock running at a uniform rate. The variation is called the equation of time (ET). $E0T = 9.87\sin(2B) - 7.53\cos(B) - 1.5\sin(B)$			
Solar angles	$R = a(1 \pm e)$			
Declination angle δ	$\delta_s = 23.45^{\circ} \sin[360(284 + n)/365^{\circ}]$			
Solar altitude angle,	Ζ= 90- α			
α				
Solar azimuth angle,	$\sin \theta = \frac{\cos \delta \sin \omega}{}$			
$\boldsymbol{\theta}$	$\cos lpha$			
Hour angle	$\omega = 15^{\circ} (AST - 12h)$			
Incidence angle θ	$\cos(\theta) = \sin(L)\sin(\delta)\cos(B) - \cos(L)\sin(\delta)\sin(B)\cos(\delta_S)$			
Solar Time and	solar Time = LST + ET + $(l_{st} - l_{local})$. 4 $\frac{min}{degree}$			
Angles	, degree			

III. MODELLING OF PHOTOVOLTAIC SOURCE AND MPPPT

A solar cell is modelled as a p-n junction with nonlinear characteristics to describe its electrical response. To analyze these characteristics, a mathematical model of the solar cell is derived from mathematical equations in terms of solar cell inputs and outputs. The simplest equivalent circuit of a solar cell is a current source connected in parallel with a diode as shown in Figure 3.1The output of the current source is directly proportional to the light falling on the cell. During darkness, the solar cell is not an active device and it works as a diode. It produces neither current nor voltage. However, if light falls on the solar cell, it generates a diode current. The diode, D, determines the I-V characteristics of the cell. A series resistance, *RS*, represents the resistance inside each cell, while the shunt resistance, *RSH*, is neglected because it has a large resistance value. [01,02]

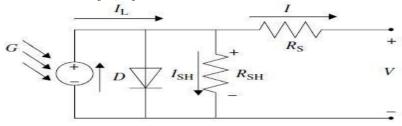


Fig 3.1Equivalent circuit of solar cell

In an ideal solar cell, it is assumed that R_S , = 0 and R_{SH} , = infinity. The net current of the cell is the difference between the photocurrent, IL, and the normal diode current, which is given by

$$I = I_L - I_0 \left\{ e^{\frac{q(V + IR_2)}{nKT}} \right\} - \frac{V + IR_S}{R_P}$$
 (1.2)

IJIRSET © 2022 | An ISO 9001:2008 Certified Journal | 1393

|| Volume 11, Issue 2, February 2022 ||

| DOI:10.15680/IJIRSET.2022.1102076 |

The photocurrent, I_L , depends on reference first and second temperatures, T_1 , and T_2 , respectively, and it is given by

refree first and second temperatures,
$$I_1$$
, and $I_L = I_L(T_1) + K_0(T_2 - T_1)$ (1.3)
$$I_L(T_1) = I_{SCT1,nom} \left(\frac{G}{G_{nom}}\right) (1.4)$$

$$K_0 = \frac{I_{SCT2} - I_{SCT1}}{T_1 - T_2} (1.5)$$
and C_1 is the solar rediction at the referen

where G is the present solar radiation and G_{nom} is the solar radiation at the reference test The saturation current of the diode, I_0 , is given by

$$I_0 = I_{0T1} \left[\frac{T}{T_1} \right]^{\frac{3}{n}} e^{\frac{qV_{qTI}}{nk\left(\frac{1}{T} - \frac{1}{T_1}\right)}}$$
 (1.6)

Where

$$I_{0T1} = \frac{I_{SCT_1}}{\frac{qV_{qTI}}{nk} - 1} (1.7)$$

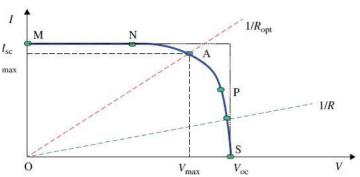


Fig 3.2I–Vcharacteristic curve of a solar cell.

IV. SOFT COMPUTING MPPT TECHNIQUES

4.1 The MPPT based Fuzzy logic controller:

Renewable energy resources have a wide range of applications for the Fuzzy Logic controllers (FLC). From the last decade, the requirement of FLC has been increased due to simplicity. FLC also deals with the imprecise input, which does not need any accurate mathematical model for the controller. FLC can easily handle nonlinearity conditions for obtaining the maximum power from PV Modules. It is capable of performing in any weather condition or any change in the temperature or irradiance level [10]

4.2 Modeling for Experimental Investigation of Photovoltaic on Fuzzy logic Controller (FCL) Based MPPT Module Under Outdoor condition:

A10 Green technology A10J-S72-175PV module is experimented. The measuring instruments are digital mustimeter with DC accuracy: $\pm 0.5\% + 3$ digits and AC accuracy: 1% + 3 digits and auto digital lux meter with measuring range varying from 2000to 50,000 lx which are utilized for measuring I–V. The measurement under STCprovided by manufacturers

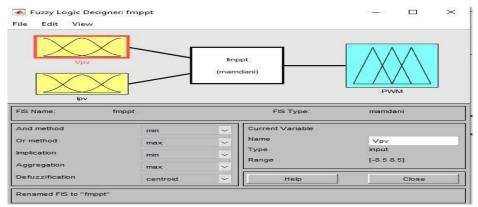


Fig 4.1 fuzzy interface system to import fis file



|| Volume 11, Issue 2, February 2022 ||

| DOI:10.15680/IJIRSET.2022.1102076 |

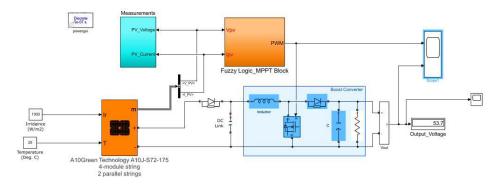
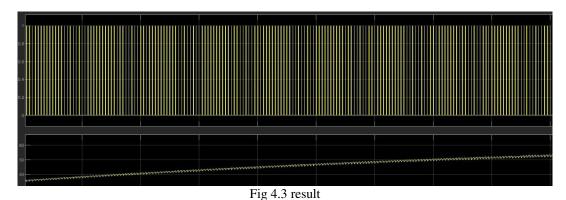


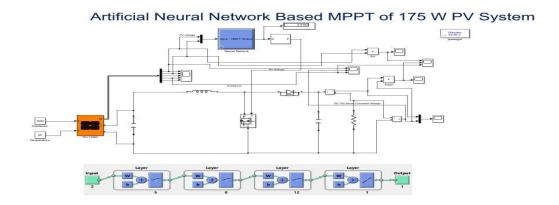
Fig 4.2Fuzzy Logic Controller Based MPPT of PV Systems Using Matlab Simulink



V. ARTIFICIAL NEURAL NETWORK

Recently, Artificial Neural Network (ANN) has been strongly developed not only in theory but also in application. A common AN, showed in Fig 5.2, has many layers: input, hidden and output layers. For MPPT, ANN input can be PV array parameters like PV voltages and currents, environmental data like irradiance and temperature, or any combination of these. The output is usually one or several reference signal(s) like a duty cycle signal used to drive the electronic converter to operate at or close to the MPP. The input and output data are from experimental measurement or model-based simulation results. Some studies used ANN to approximate nonlinear function were published

5.1ANN Based Modelling for Experimental Investigation of Photovoltaic Module Maximum Power Production Under Outdoor Condition.





|| Volume 11, Issue 2, February 2022 ||

| DOI:10.15680/IJIRSET.2022.1102076 |

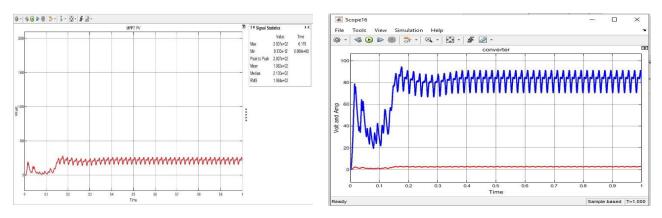
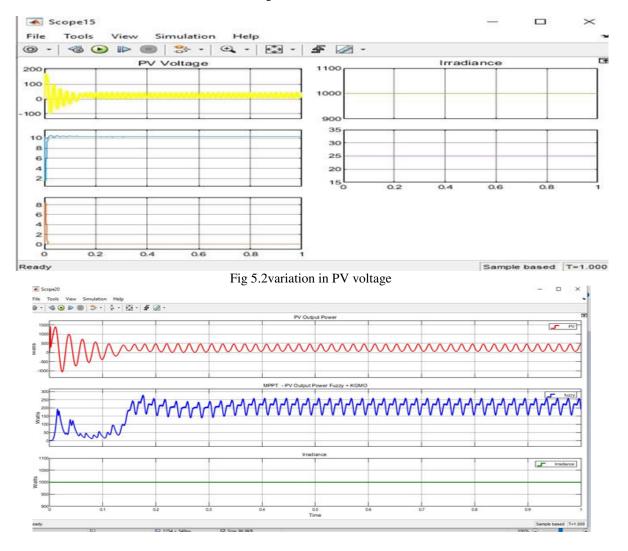


Fig 5.1 a change of power with respect to time in ANN based MPPT controller b.voltage and current variation





|| Volume 11, Issue 2, February 2022 ||

| DOI:10.15680/IJIRSET.2022.1102076 |

VI. RESULT AND CONCLUSION

The comparison of the convergence time and generated output power for various MPPT control methods under fixed environmental condition

MPPT Method	Converging Time (s)	Output voltage(v)	Output Power (Watts)
MPPT Controller	0.1	56.2	172.25
FCL	0.05	53.7	173.4
ANN	0.035	50.12	174.56

Solar energy is one of the clean resources of energy but it has challenging issues of low efficiency and low power extraction due to varying environmental conditions. There are many conventional MPPT techniques such as Hill climb, IC and P&O but their associated issues hinder the effective power extraction. AI based MPPT techniques alleviate the challenges of conventional MPPT techniques and are a suitable solution to the low efficiencies in PV systems. This work presented the comparison of various MPPT methods (FLC, ANN, for a PV system under fixed and varying environmental conditions in MATLAB/Simulink environment. The comparison results showed that ANN based MPPT control has better performance in terms of converging time and output voltage power extraction when compared to other MPPT methods. TheANN based control outperformed other methods under fixed as well as varying environmental conditions which makes it a suitable candidate for the MPPT control in PV systems.

REFERENCES

- [1] Fuzzy Logic MPPT for Grid Tie Solar Inverter by Margeel L Shah1 AsheeshDhaneria 2020 IEEE International Conference for Innovation in Technology (INOCON) Bengaluru, India. Nov 6-8, 2020
- [2] Wu Z, Yu D (2018) Application of improved bat algorithm for solar PV maximum power point tracking under partially shaded condition. Appl Soft Comput 62:101–109
- [3] Schoeman JJ, VanDyke JD (1982) A simplified maximal power controller for terrestrial photovoltaic panel arrays. In: Proceeding 13th annual IEEE power electronics specialists conference, pp 361–367
- [4] Punitha K, Devaraj D, Sakthivel S (2013) Development and analysis of adaptive fuzzy controllers for photovoltaic system under varying atmospheric and partial shading condition. Appl Soft Comput 13:4320–4332
- [5] PHOTOVOLTAICS SYSTEM DESIGN AND PRACTICE Heinrich Haberlea Berne University of Applied Sciences, Switzerland Translated by Herbert Eppel HE Translations, Leicester, UK
- [6] Huang C, Wang L, Yeung RS-C, Zhang Z, Chung HS-H, Bensoussan A (2018) A prediction model-guided Jaya algorithm for the PV system maximum power point tracking. IEEE Trans Sustain Energy 9(1):45–55
- [7] Cristea M, Chiritoiu V, Costache M, Zaharie I, Luminosu I (2010) On a model of the typical cell from a solar panel. In: AIP Conference Proceedings, vol 1203, no 1. AIP, pp 433–438
- [8] MODELING OF PHOTOVOLTAIC SYSTEMS USING MATLAB by Tamer Khatib Wilfried Emmerich
- [9] MATLAB / SIMULINK Simulation of PV System based on MPPT in Variable Irradiance with EV Battery as Load Arshdeep Singh978-1-5090-6621-6/17/\$31.00 ©2017 IEEE











Impact Factor: 7.569

INTERNATIONAL JOURNAL OF INNOVATIVE RESEARCH

IN SCIENCE | ENGINEERING | TECHNOLOGY







📵 9940 572 462 囟 6381 907 438 🔼 ijirset@gmail.com

