



IN APPLIED SCIENCE & ENGINEERING TECHNOLOGY

Volume: 10 Issue: V Month of publication: May 2022

DOI: https://doi.org/10.22214/ijraset.2022.42951

www.ijraset.com

Call: 🕥 08813907089 🔰 E-mail ID: ijraset@gmail.com



Maximum Power Point Tracking using modified Particle Swarm Optimization Technique

Gangadhar Mahalingappa Akki¹, Srivani S. G²

^{1, 2}Department of Electrical and Electronics Engineering, RV College of Engineering, Bengaluru, India

Abstract: Maximum power point tracking is an essential and vital technique generally applied in SPV technology under uniform and partial shading conditions. The existence of partially shaded conditions leads to the presence of several peaks on PV curves, which decrease the efficiency of conventional techniques. Hence, the proposed algorithm, which is based on the modified particle-swarm optimization (MPSO) technique, increases the output power of PV systems under such abnormal conditions and has a better performance compared to other methods. The proposed method is examined under several scenarios for partial shading condition and non-uniform irradiation levels using Matlab and to investigate its effectiveness adequately. The experimental results show that the proposed method can decrease the interference of the local maximum power-point to cause the PV system to operate at a global maximum power-point. The efficiency of the MPSO is achieved with the least number of steady-state oscillations under partial shading conditions as compared with the other methods.

Keywords: Photovoltaic(PV), Particle Swarm Optimization(PSO), Modified Particle Swarm Optimization(MPSO), Partial shading condition(PSC), Velocity step function.

I. INTRODUCTION

The sustainability of solar energy and the resulting reduction in its material cost has led to the widespread application of photovoltaic (PV) systems in daily lives. However, during the practical implementation of PV systems, their short life cycles and low energy efficiency are the main associated problems. The main reasons for this are the power loss and hot-spots, which are caused by the presence of partial shadows. Under uniform irradiation levels, the tracking process of the maximum power-point of PV systems based on classical strategies can have a suitable performance. However, if a PV system operates under a partial shading condition (PSC), the power-voltage (P-V) characteristic curve of the PV system will have different local maximum points resulting from the connection of bypass diodes to reduce the impact of hot-spots.

The existence of several peaks on the PV characteristic curve increases the complexity associated with the extraction of the global maximum point under these conditions, and there is a need to propose a more suitable control system that can distinguish between local and global maxima to ensure the maximum possible power, thus enhancing the total system efficiency. For this reason, the key purpose of this study is to propose an intelligent maximum power-point tracking (MPPT) tracker that enables the efficient prediction of the global maximum power-point (GMPP) from a PV system, regardless of the condition of the surrounding atmosphere, whether under uniform or non-uniform solar irradiation levels.

In the literature, several global MPP search algorithms have been developed to determine the global MPP under conditions of partial shading. In PSO, particles are allowed to move in random directions and best value is arrived. This behavior is quite similar to change in duty cycle of maximum power point tracking [1]. In new ARMO based MPPT Algorithm, the coefficients in the ARMO will change adaptively [2]. In new ANN based MPPT tracking, the results of this method were accompanied by some drawbacks, such as the excessive complexity of the control scheme and a large number of computations [3]. Incremental conductance algorithm computes MPP by comparing the ratios of instantaneous conductance and incremental conductance. In Refs. [5-8], modifications of conventional MPPT methods are proposed to improve the efficiency and performance of these methods, which is an improved P&O (MPPT) method based on the adaptive duty cycle step of fuzzy logic controllers [9], and a new IC MPPT algorithm is proposed using direct control based on the fuzzy duty cycle change estimator [10].

According to the investigation, the PSO technique is easy, effective, and robust, and it is a population-based algorithm that can be used to treat optimization problems. Some modifications are necessary to enhance the performance of PSO. Therefore, MPSO modifies the velocity step function, controls the velocity limit, and controls the search space. Therefore, this work determines the performance of the MPSO based MPPT method under various solar irradiation conditions and PSCs.



International Journal for Research in Applied Science & Engineering Technology (IJRASET) ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538

Volume 10 Issue V May 2022- Available at www.ijraset.com

A. Principle Of Partial Shading

The PV module shading can be either partial or total considering the movement of objects that block the sunlight from the PV modules. Due to this, the output attributes of solar modules are more complex with various peak points. The shaded PV array has multiple local peaks on its P-V characteristics, rather than only one peak for non-shaded PV arrays. Hence only the global peak can result in highest power rather than the other different peaks P-V and V-I characteristics of solar PV are shown in Fig 1 and Fig 2.



Fig. 1: Solar PV P-V characteristic under partial shading condition



Fig. 2: Solar PV V-I characteristic under partial shading condition

The PV array may be defective owing to different aging effects. These effects arise from different outer reasons like include dust, dirt, surrounding plants, and bird droppings, which cause partial or total shading conditions and increased internal temperature. These are the reasons for more optical and physical cell degradation phenomena.

II. PSO ALGORITHM

Particle Swarm Optimization (PSO) was invented by Russell Eberhart and James Kennedy in the year 1995 and the technique inspired by the social behavior and dynamic movement of flock of birds, holds high potential due to its simple structure and fast computation capability. An individual bird must move away from its neighbor so as to avoid any chance of collision. The position and velocity is adjusted to determine the best possible position for individual particles (pbest) and the best one from the swarm(gbest).



International Journal for Research in Applied Science & Engineering Technology (IJRASET) ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538

Volume 10 Issue V May 2022- Available at www.ijraset.com

The velocity and position of the particle is updated by the following equation:

w is the inertia weight factor, c1 and c2 are the learning coefficients, rand() is the random variable generated, and pbest and gbest are the personal best position and global best position of the particle respectively.

The updation of velocity and position of the PSO particle in search space is defined by using equations (1) and (2) and it is represented in fig 3 and the flowchart of the PSO algorithm is shown in fig 4.



Fig. 3: Particle movement in search space



Fig. 4: Flowchart for Particle swarm optimization



International Journal for Research in Applied Science & Engineering Technology (IJRASET)

ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538 Volume 10 Issue V May 2022- Available at www.ijraset.com

III.PROPOSED MODIFIED PSO ALGORITHM

The proposed method is named as Modified PSO (MPSO) algorithm computes the values of initial particles i.e.duty cycle in case of MPPT, using equation (3). In equation(3), the value of *Rin* is computed based on voltage to current ratio at maximum power. Hence, it starts with an initial value closer to MPP.

The expression for initial duty cycle of the MPSO algorithm is:

$$d = 1/\left(1 + \sqrt{\frac{R_{in}}{R_o}}\right).$$
(3)

where

is the internal resistance of the PV module and R_o is output load resistance

The equation for voltage at maximum power point is given by:

The equation for current at maximum power point is given by:

where

 k_1 and k_2 are the constant of proportionalities that changes with the change in irradiation and temperature.

The equation corresponding to k1 and k2 can be derived for a particular panel as:

$$k_{1} = \left(\frac{v_{mppn}}{v_{ocn}}\right) * \left[1 + \left(\alpha^{*}\ln\left(\frac{G}{G_{n}}\right)\right) + \beta^{*}(T - T_{n})\right].$$

$$k_{2} = \left(\frac{l_{mppn}}{l_{scn}}\right) * \left[1 + \left(\gamma^{*}\ln\left(\frac{G}{G_{n}}\right)\right)\right].$$
(8)

The values are selected in such a way that they will match the solar module characteristics. MPSO method starts with initial set of particles (duty cycles) and are defined as:

 $x_{i}^{t} = d_{i}^{t} = \left[d_{1}, d_{2}, d_{3}, \dots, d_{N_{p}} \right].$ (9)

where,

N p is the number of particles, and *t* is the number of iteration.

The objective function can be formulated as:

 $P(d_i^t) > P(d_i^{t-1}).$ (10)

The search process begins by sending three duty cycles to the power converter. In the first iteration, all the three duty cycles will be considered as the *pbest* value. The duty cycle which is close to the module maximum power (fitness value) will be taken as the *gbest* value. Based on the *pbest* and *gbest* values, the velocity and position of the duty cycles will be updated. In the next iteration, the present fitness value will be compared with that of previous one and a small perturbation is provided updating the duty cycle. This process undergoes changes until all the duty cycle attains a better fitness value. At this point, the velocity component becomes nearly zero. Hence, all the duty cycle will converge to a single point at MPP.

International Journal for Research in Applied Science & Engineering Technology (IJRASET)



ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538 Volume 10 Issue V May 2022- Available at www.ijraset.com

IV.SIMULATION AND RESULTS

Matlab/Simulink is one of the significant tool to determine the performance of the recommended MPPT algorithm. The modeling of PV system with a boost converter and MPPT algorithms is shown in Fig 5.



Fig 5 : Solar PV modelling with Boost Converter

```
function output=PV Model1(input)
 %PV characteristics func
Va=input(1);
G=input(2);
Tcell=input(3);
 %This function uses the output voltage, irradiance and ambient temper
 %to calculate the current of the PV array
 k = 1.38e-23; % Boltzmann's const
 q = 1.60e-19; % Electron charge
 % Photovoltaic constants at STC
A = 2; % "diode quality" factor (from PV-DesignPro-S)
Egap = 1.12; % band gap voltage for silicon devices
Num Series = 72; % series connected cells
Voc = 43.2; % open circuit voltage at STC
 Isc = 8.71; % short circuit current at STC
TempCoefI = 0.05; %current temperature cvoefficient
TempCoefV = -0.34; %voltage temperature coefficient
deltaVdeltaI Voc = -0.45; % values at Voc from manufacturers curves
 % Initial temperature T1 variables at STC
T1 = 273 + 25; % convert ambient temperature to Kelvin
 Voc_T1 = Voc ./Num_Series; % open cct voltage per cell at T1
 Isc_T1 = Isc; % short cct current per cell at T1
T2 = 273 + 75; % convert temperature to Kelvin
Voc T2 = Voc + (50.*TempCoefV)./Num Series; % Voc per cell at T2
 Isc T2 = Isc + (50.*TempCoefI); % Isc per cell at T2
TaK = 273 + Tcell; % module temperature in Kelvin at any temperature
Vc = Va./Num Series; % determine the cell voltage
```

Fig 6 : Matlab code part_1 for modelling of Solar PV module



International Journal for Research in Applied Science & Engineering Technology (IJRASET) ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538 Volume 10 Issue V May 2022- Available at www.ijraset.com

```
TaK = 273 + Tcell; % module temperature in Kelvin at any temperature
Vc = Va./Num Series; % determine the cell voltage
Iph T1 = Isc T1 * (G./1000);%current produced by the cell at temp 1
a = (Isc T2 - Isc T1)/Isc T1 * 1/(T2 - T1);%a=constant
Iph = Iph T1 * (1 + a*(TaK - T1));%current produced by cell
Vt T1 = k * T1 / q; %Define thermal properties(Vt) at Temp1
Ir T1 = Isc T1 / (exp (Voc T1/ (A*Vt T1))-1);%diode reverse saturation curent
b = Egap * q / (A * k);
Ir T1 * (TaK/T1)^(3/A) .* exp(-b *(1/TaK - 1/T1));%reverse saturation current at T1
Xv = Ir T1/(A*Vt T1) * exp(Voc T1/(A*Vt T1));%cell series impedance
deltaVdeltaI_Voc_per_cell = (deltaVdeltaI Voc)/Num Series;
Rs = - deltaVdeltaI Voc per cell - 1/Xv;
%define thermal properties at temperature TaK
Vt TaK = A * k * TaK / q;
Ipv = zeros(size(Vc));
for j=1:5; %calculates Ia using Newton's method
Ipv = Ipv - (Iph - Ipv - Ir.*( exp((Vc+Ipv.*Rs)./Vt TaK) -1))...
./ (-1 - (Ir.*( exp((Vc+Ipv.*Rs)./Vt TaK) -1)).*Rs./Vt TaK);
end
output(1)=Ipv;
end
```

Fig 7 : Matlab code part_2 for modelling of Solar PV module

```
[] function Ipv=PVcharacteristics func(Va,G,Tcell)
□ %PV characteristics func
 %This function uses the output voltage, irradiance and ambient temperature
 -%to calculate the current of the PV array
 k = 1.38e-23; % Boltzmann's const
 q = 1.60e-19; % Electron charge
  % Photovoltaic constants at STC
 A = 2; % "diode quality" factor (from PV-DesignPro-S)
 Egap = 1.12; % band gap voltage for silicon devices
 Num Series = 72; % series connected cells
 Voc = 43.2; % open circuit voltage at STC
 Isc = 8.71; % short circuit current at STC
 TempCoefI = 0.05; %current temperature cvoefficient
 TempCoefV = -0.34; %voltage temperature coefficient
 deltaVdeltaI Voc = -0.45; % values at Voc from manufacturers curves
  % Initial temperature T1 variables at STC
 T1 = 273 + 25; % convert ambient temperature to Kelvin
 Voc_T1 = Voc ./Num_Series; % open cct voltage per cell at T1
 Isc_T1 = Isc; % short cct current per cell at T1
 T2 = 273 + 75; % convert temperature to Kelvin
 Voc T2 = Voc + (50.*TempCoefV)./Num Series; % Voc per cell at T2
 Isc T2 = Isc + (50.*TempCoefI); % Isc per cell at T2
 TaK = 273 + Tcell; % module temperature in Kelvin at any temperature
 Vc = Va./Num Series; % determine the cell voltage
 Iph_T1 = Isc_T1 * (G./1000);%current produced by the cell at temp 1
 a = (Isc_T2 - Isc_T1)/Isc_T1 * 1/(T2 - T1);%a=constant
            Fig 8 : Matlab code part_1 for modelling of Solar PV module characteristics
```

A Policy Science of Contracting of C

International Journal for Research in Applied Science & Engineering Technology (IJRASET)

ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538 Volume 10 Issue V May 2022- Available at www.ijraset.com

```
a = (Isc_T2 - Isc_T1)/Isc_T1 * 1/(T2 - T1);%a=constant
Iph = Iph T1 * (1 + a*(TaK - T1));%current produced by cell
Vt_T1 = k * T1 / q; %Define thermal properties(Vt) at Temp1
Ir_T1 = Isc_T1 / (exp (Voc_T1/ (A*Vt_T1))-1);%diode reverse saturation curent
b = Egap * q / (A * k);
Ir = Ir T1 * (TaK/T1)^(3/A) .* exp(-b *(1/TaK - 1/T1));%reverse saturation current at T1
Xv = Ir_T1/(A*Vt_T1) * exp(Voc_T1/(A*Vt_T1));%cell series impedance
deltaVdeltaI_Voc_per_cell = (deltaVdeltaI_Voc)/Num_Series;
Rs = - deltaVdeltaI_Voc_per_cell - 1/Xv;
%define thermal properties at temperature TaK
Vt TaK = A * k * TaK / q;
Ipv = zeros(size(Vc));
for j=1:5; %calculates Ia using Newton's method
Ipv = Ipv - (Iph - Ipv - Ir.*( exp((Vc+Ipv.*Rs)./Vt_TaK) -1))...
./ (-1 - (Ir.*( exp((Vc+Ipv.*Rs)./Vt_TaK) -1)).*Rs./Vt_TaK);
- end
- end
```

Fig 9 : Matlab code part_2 for modelling of Solar PV module characteristics

[function output=MPPT_PSO(input)

```
%%%----- PV 1 -----
Tcell1 =input(1);% 25; %Ambient cell temperature (degrees celcius)
G1 = input(2);%1000%Solar Irradiation (W/m2)
Voc1 =43.2; % open circuit voltage at STC
Va1 = 0:0.1:Voc1;
Ipv1 = PVcharacteristics func(Val,G1,Tcell1);
factor1=Ipv1>=0;
Ipv1=factor1.*Ipv1;
Ppv1 = Va1.*Ipv1;
888-----
Pmax1 = max (Ppv1); %Finding the maximum power of the array
[row1,cumn1]=find(Ppv1<=Pmax1 & Ppv1>=Pmax1);
V max1=Val(row1, cumn1);
I max1=Ipv1(row1,cumn1);
%%%----- PV_2 -----
Tcell2 = input(3);%33; %Ambient cell temperature (degrees celcius)
G2 = input(4);%850;%Solar Irradiation (W/m2)
Voc2 = 43.2; %Open circuit voltage
Va2 = 0:0.1:Voc2;
Ipv2 = PVcharacteristics func(Va2,G2,Tcell2);
              Fig 10 : Matlab code part_1 for MPSO Algorithm
factor2=Ipv2>=0;
Ipv2=factor2.*Ipv2;
Ppv2 = Va2.*Ipv2; %Calculating PV output power
Pmax2 = max (Ppv2); %Finding the maximum power of the array
[row2,cumn2]=find(Ppv2<=Pmax2 & Ppv2>=Pmax2);
```

```
V_max2=Va2(row2,cumn2);
I_max2=Ipv2(row2,cumn2);
```

Fig 11 : Matlab code part_2 for MPSO Algorithm



International Journal for Research in Applied Science & Engineering Technology (IJRASET) ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538 Volume 10 Issue V May 2022- Available at www.ijraset.com

```
%%%%%------ PV total calculation -----
PV_total=Va3.*(Ipv1+Ipv2+Ipv3);
Pmax_total = max (PV_total); %Finding the maximum power of the array
[row_total,cumn_total]=find(PV_total<=Pmax_total & PV_total>=Pmax_total);
V_max_total=Va3(row_total,cumn_total);
I_max_total=Ipv1(row_total,cumn_total)+Ipv2(row_total,cumn_total)+Ipv3(row_total,cumn_total);
%%------
output(1)=V_max_total(1);
-output(2)=I_max_total(1);
```

Fig 12 : Matlab code part_3 for MPSO Algorithm

The matlab codes for modelling of Solar PV module and Solar PV charactersistics and MPSO algorithm are shown in Fig 6, Fig 7, Fig 8, Fig 9, Fig 10, Fig 11, Fig 12.

B. Under Uniform Irradiation Conditions



Fig 13 : Performance of the SPV modules with Boost converter under constant radiation using MPSO



Fig 14 : Voltage curve of the PV modules under constant radiation



Fig 15 : Current curve of the PV modules under constant radiation



Fig 16 : Power curve of the PV modules under constant radiation

Under uniform irradiation conditions, solar PV modules were able to capture 810.9 W of maximum power using MPSO Algorithm. The performance curves of voltage, current and power are shown in Fig 14, Fig 15 and Fig 16

C. Under Partial Shading Conditions



Fig 17 : Solar PV modelling with Boost converter under partial shading conditions using MPSO



International Journal for Research in Applied Science & Engineering Technology (IJRASET)

ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538 Volume 10 Issue V May 2022- Available at www.ijraset.com



Fig 18 : Voltage curve of the PV modules under partial shading conditions





Fig 20 : Power curve of the PV modules under partial shading conditions

Under partial shading conditions, solar PV modules were able to capture 681W of maximum power using MPSO Algorithm. The performance curves of voltage, current and power are shown in Fig 18, Fig 19 and Fig 20.

V. CONCLUSION

This work presents the performance analysis of Solar PV Boost converter fed circuit under constant irradiation conditions and under partial shading conditions. Under uniform radiation conditions, the peak power point was obtained at 810.9 W. Under partial shading conditions, the maximum power point was obtained at 680.9 W. Two trials were tested for partial shading conditions, i.e under different temperature and irradiation conditions. It was observed that in both the trials, SPV panel was operated at a peak power of 680.9 W. It can be concluded that MPSO algorithm has superior and better tracking efficiency as compared to other MPPT tracking techniques.

REFERENCES

- M. M. Shehu, M. Dong and J. Hu, "Optimization of Particle Swarm based MPPT under Partial Shading Conditions in Photovoltaic Systems," 2021 IEEE 16th Conference on Industrial Electronics and Applications (ICIEA), 2021, pp. 267-272.
- [2] M. Seyedmahmoudian, T. K. Soon, B. Horan, A. Ghandhari, S. Mekhilef and A. Stojcevsk, "New ARMO-based MPPT Technique to Minimize Tracking Time and Fluctuation at Output of PV Systems under Rapidly Changing Shading Conditions," IEEE Transactions on Industrial Informatics, Volume 10 Issue 09, 24th January 2019.
- [3] M. Manas, A. Kumari and S. Das, An Artificial Neural Network based Maximum Power Point Tracking method for photovoltaic system, "2016 International Conference on Recent Advances and Innovations in Engineering (ICRAIE), 2016, pp. 1-6.
- [4] M. Forouzesh, Y. P. Siwakoti, S. A. Gorji, F. Blaabjerg, and B. Lehman, "A comprehensive review of voltage-boosting techniques, topologies, and applications", IEEE Trans. Power Electron., Vol. 32, No. 12, Pp. 9143–9178, Dec. 2017.
- [5] A W Ibrahim, M Ding, X Jin, et al, "Artificial neural network based maximum power point tracking for PV system," Chinese Control Conference(CCC), 27-30 July, 2019, Guangzhou, China. IEEE, 2019: 6559-6564.
- [6] N A Ahmed, M Miyatake, "A novel maximum power point tracking for photovoltaic applications under partially shaded insolation conditions," Electric Power System Research, 2016, Vol 78, No 5, pp.777-784.
- [7] K S Tey, S Mekhilef, M Seyedmahmoudian, "Improved differential evolution-based MPPT algorithm using SEPIC for PV systems under partial shading conditions and load variation," IEEE Transactions on Industrial Informatics, 2018, Vol 14, Issue 10, pp.4322-4333.
- [8] F Li, Y Huang, F Wu, "Research on clustering equivalent modeling of large-scale photovoltaic power plants," Chinese Journal of Electrical Engineering, 2018, 4(4): 80-85.
- [9] A Jubaer, S Zainal, "A maximum power point tracking, MPPT for PV system using Cuckoo search with partial shading capability," Applied Energy, 2014, 119: 119-130.
- [10] M B Shafik, H Chen, G I Rashed, "Adaptive multi objective parallel seeker optimization algorithm for incorporating TCSC devices into optimal power flow framework," IEEE Access, 2019, 7: 36934-36947.











45.98



IMPACT FACTOR: 7.129







INTERNATIONAL JOURNAL FOR RESEARCH

IN APPLIED SCIENCE & ENGINEERING TECHNOLOGY

Call : 08813907089 🕓 (24*7 Support on Whatsapp)