Maximum Power Point Tracking in SPV System under Linear and Non-Linear Condition

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Key words: Maximum Power Point Tracking (MPPT), Perturbation and Observation (P and O), Particle Swarm Optimization (PSO), linear and non-linear, direct irradiance and diffused irradiance, PV system

Abstract: In this study, maximum power point tracking is done in a solar photovoltaic system for maximum harvesting of power from the sun, so that, the solar power generation can be improved. There are various MPPT techniques used in solar PV system for extracting power from the sun but the most used is perturbation and observation and particle swarm optimization methods. In this study a comparative analysis of perturbation and observation and particle swarm optimization technique is done for maximum power point tracking of solar photovoltaic system under linear and non-linear condition. The efficiency of both the technique is compared and the best technique is applied for maximum power point tracking in a solar PV system under linear and non-linear conditions that prevail in the atmosphere. Linear and non-linear are conditions in the atmosphere that occurs due to direct sunlight irradiance and diffused irradiance on the Earth’s atmosphere that affects the solar PV panel insolation level. Under varying atmospheric conditions the solar PV generation is made more efficient by using the best maximum power point tracking technique.

INTRODUCTION

Now a days demand for renewable energy is rapidly increasing due depletion of non-renewable energy resources and increased pollution from the non-renewable system. In renewable energy systems solar energy is obtained very easily and it involves low capitals when compared to other forms of renewable energy systems. Therefore, tapping of solar energy plays an important role towards achieving long lasting, sustainable, environment friendly in fulfilling the energy needs for mankind which are given by Nepal (2012), Pavlovic et al. (2012), Eltawil and Zhao (2010). Generally, solar energy are tapped using photovoltaic cells. The whole combined setup of PV system is used for generating solar power. A photovoltaic source has several advantages such as low maintenance cost, no need for fuel, pollution free and contributes to standalone systems as by Gow and Manning (1996). Though, the solar photovoltaic system is very much beneficial it has some of its drawbacks such as low conversion efficiency, nonlinear characteristics of PV arrays and non-reliability on irradiation levels and atmospheric temperature has made some difficulties to extract the maximum power from them as by Alajmi et al. (2013). Over the years many papers are concentrated in finding a solution for maximum power harvesting in a solar PV system using many MPPT algorithms. The conventional methods used are incremental conductance, current sweep method, hill climbing methods and the most commonly used are perturbation and particle swarm optimization. Here, in this study a comparative analysis of Perturbation and Observation (P and O) and Particle Swarm Optimization (PSO) is performed to determine which is the best MPPT technique for achieving maximum power harvesting from a solar PV system under varying atmospheric condition similarly by Ishaque et al. (2011). The analysis is performed using the simulated results of MATLAB Simulink.
Fig. 1: Solar PV system

Fig. 2: Solar panel equivalent circuit

**MATERIALS AND METHODS**

**Photovoltaic system:** A PV system consists of a solar panel, DC-DC converter, controller, load and battery as shown in Fig. 1.

**Solar panel:** Solar panel is a P-N junction device in which the sunlight is absorbed by the semiconductor atoms, freeing electrons from the cell’s negative layer to flow through an external circuit and back into the positive layer like by Villalva et al. (2009). This flow of electrons produces electric current. Figure 2 shows the equivalent circuit of a solar panel. The equations of this circuit are as follows:

\[
I = I_p - I_s \left( \exp \left( \frac{V + R_s I}{aV_t} \right) - 1 \right) \frac{V + R_s I}{R_s} \tag{1}
\]

\[
V_t = \frac{N_s K T}{q} \tag{2}
\]

\[
I_p = \left( I_{pv,n} + K_s \Delta T \right) \frac{G}{G_n} \tag{3}
\]

\[
I_o = I_{o,n} \left( \frac{T}{T_n} \right)^\gamma \exp \left[ \frac{qE_g}{aK \left( \frac{T}{T_n} \right) - 1} \right] \tag{4}
\]

**Converter:** Converter is an essential part of any PV system regardless of load type. The main role of converter is conditioning the produced power of PV cells in order to meet the load requirements. However, it can be used for matching the panel curve with output load in order to extract maximum power from PV panel.

**Controller:** A MPPT solar charge controller is the charge controller embedded with MPPT algorithm to maximize the amount of current going into the battery from PV module. The major principle of MPPT is to extract the maximum available power from PV module by making them operate at the most efficient voltage (Alajmi et al., 2013). MPPT checks output of PV module, compares it to battery voltage then fixes what is the best power that PV module can produce to charge the battery and converts it to the best voltage to get maximum current into battery. MPPT is most effective under non linear conditions. Cold weather, cloudy or hazy days. MPPT

Where:

- \( V_t \): Thermal Voltage
- \( N_s \): The Number of series cells
- \( I_{pv} \): Photovoltaic current
- \( I_{pv,n} \): Photovoltaic current in nominal condition
- \( G \): Radiation intensity
- \( G_n \): Nominal radiation intensity = 1000 w/m²
- \( \Delta T \): \((\Delta T = T_n)\) Temperature change
- \( T_n \): Nominal Temperature = 25°C
- \( E_g \): The distance of energy bars
- \( S_i, I_{o,n} \): Inverse Saturation current in nominal condition
- \( K_i \): The ratio of short circuit

Current variation to temperature variation in nominal condition, \( a \) is a constant value between 1 and 1.5 that is determined by other cell’s parameters. By using the Eq. 6 instead of Eq. 4, the accuracy of model increases as a consequence of \( V_{oc} \) sensitivity to temperature. In these equations, \( K_v \) is the sensitivity of open circuit voltage with respect to temperature:

\[
I_o = \frac{I_{sc,n} + K_s \Delta T}{\exp \left( \frac{V_{oc,n} + K_s \Delta T}{aV_t} \right) - 1} \tag{6}
\]

\[
I_{sc} = (I_{sc,n} + K_s \Delta T) \frac{G}{G_n} \tag{7}
\]

\[
V_{oc} = V_{oc,n} + K_s \Delta T \tag{8}
\]

By using the Eq. 6 instead of Eq. 4, the accuracy of model increases as a consequence of \( V_{oc} \) sensitivity to temperature. In these equations, \( K_v \) is the sensitivity of open circuit voltage with respect to temperature:
Randomly select the voltage and compute the corresponding current

Compute the $P(i)$ using $V(i)$ and $I(i)$

If $P(i) > P(i-1)$

$V(i) = V(i-1)$

$V(i) = V(i+1)$

If error $< 0.001$

End

Fig. 3: Flow chart of P&O method

Fig. 4: Characteristic 1-5 curve of P and O (X Y plot)

solar charge controller is used to correct for detecting the variations in the current voltage characteristics of solar cell and shown by Nepal (2012), Pavlovic et al. (2012), Eltawil and Zhao (2010), Gow and Manning (1996), Villalva et al. (2009) curve (Alajmi et al., 2013; Kouchaki et al., 2013; Ishaque et al., 2011). MPPT solar charge controller is necessary for any solar power systems need to extract maximum power from PV module. It forces PV module to operate at voltage close to maximum power point to draw maximum power.

MPPT using perturbation and observation:

Perturbation and Observation (P and O) is based on a comparison of perturbation voltage and changing power. The Perturbation and Observation (P and O) has been used mostly for its simplicity and easy implementation. Applicable to both analog and digital designs. It is also compatible with any kind of PV modules. Oscillations occur during nonlinear conditions. In this method the controller adjusts the voltage by a small amount from the array and measures power. If the power increases, further adjustments in that direction are tried until power no longer increases. This is called the perturbation and observation method and is most common, although, this method can result in oscillations of power output. It is referred to as a hill climbing method because it depends on the rise of the curve of power against voltage below the maximum power point and the fall above that point. The P&O finds the global peak among the local peaks without scanning the entire of P-V curve (Sahnoun et al., 2013). The following Fig. 3 shows the flowchart of P&O method.

MPPT using particle swarm optimization:

If a PV array is partially shaded by the shadow of building, tree etc., realization of MPPT is a difficult task. The optimal current of each PV module is nearly proportional to the insolation falling on it. Under the partially shaded conditions the conventional MPPT controller may track to a local MPP instead of global MPP. Hence, the generated power may be reduced and PV system efficiency will decrease. Compared to other techniques in this technique a single pair of voltage and current sensors is sufficient to realize the MPPT scheme (Ishaque et al., 2012). The PSO uses several cooperative agents and each agent shares the information attained by each individual during the search process. In this method each agent moves in the search space with a velocity, $v_i^k$ and momentum $w$, according to, its own previous best solution and its groups previous best solution (Fig. 4):

$$V^{k+1} = w \cdot V_i^k + c_1 \cdot r_1 \cdot (p_{best}^k + c_2 \cdot r_2 \cdot g_{best}) + c_1 \cdot r_1 \cdot (s_i^k + V_i^k)$$

Here, $p_{best}^k = s_i^k$ and $f(s_i^k) = f(p_{best})$, where, $f$ (also known as objective function) is the generated power $P$ which represents the summation of power generated by each array. Only one current sensor is sufficient for
tracking the MPP and can be called as multidimensional MPPT control. Let $V$ represent the terminal voltage and can be represented in the form of N-dimensional row vector:

$$s^k = [V^k_1, V^k_2, ..., V^k_N]$$

$$v^k = [V^k_{1-1}, V^k_{2-1}, ..., V^k_{N-1}]$$

When the agents reach to the MPP in steady state, the velocity becomes zero. The agents are reinitialized whenever the following two conditions are satisfied. The researchers add the function of shifting the agents in order to adapt the agents to such situation. The direction of shifting is decided randomly. Hence, this function enables local search around the MPP:

$$|v_i+1| < \Delta V$$

$$\frac{P(s_i+1)-P(s_i)}{P(s_i)} < -\Delta P$$

The following Fig. 5 and 6 shows the flow chart for PSO method (Fig. 6).

**RESULTS AND DISCUSSION**

The simulated results for P and O and PSO under linear condition and non-linear is given in the following tables.

**Comparative analysis of P and O and PSO under linear and non-linear conditions:** A comparative maximum power output for P and O and PSO algorithm is executed under linear and non-linear condition (Heydari-doostatabad et al., 2013). The linear condition is nothing but SPV panel exposed to direct beam of sunlight irradiance where as non-linear condition is that when the SPV panel is exposed to diffused sunlight irradiance, here, the term diffused irradiance occur due to cloudy weather condition or due to reflection of sunlight by huge mountains lying low. The simulated results are obtained to check which algorithm among PSO and P&O is doing better in maximum power point tracking and the comparative output is shown in tables where Table 1 illustrates the maximum power comparative output of PSO and P and O under linear condition and Table 2 illustrates the maximum power comparative output of PSO and P and O under non-linear condition.
Table 1: Simulated results using PSO and P&O technique under linear condition

| GW/m² | °C | Voltage (V) | Current (A) | Power (W) | Voltage (V) | Current (A) | Max power (W) |
|-------|----|-------------|------------|-----------|-------------|------------|---------------|
| 1000  | 25 | 38.56       | 5.44       | 210.00    | 39.70       | 5.29       | 210.05        |
| 800   | 25 | 38.24       | 4.30       | 164.65    | 39.10       | 4.22       | 165.37        |
| 600   | 25 | 28.24       | 3.40       | 97.76     | 38.22       | 3.17       | 121.07        |
| 400   | 25 | 19.12       | 2.27       | 43.58     | 37.08       | 2.10       | 78.09         |
| 1000  | 35 | 35.50       | 5.36       | 190.30    | 36.69       | 5.40       | 198.26        |
| 1000  | 40 | 33.83       | 5.15       | 181.00    | 35.23       | 5.48       | 193.06        |
| 750   | 25 | 28.24       | 3.40       | 97.76     | 33.99       | 5.48       | 186.59        |
| 800   | 35 | 38.23       | 3.94       | 150.87    | 36.15       | 4.30       | 155.56        |
| 600   | 40 | 28.67       | 3.50       | 100.39    | 33.87       | 3.24       | 110.07        |
| 950   | 45 | 34.32       | 4.72       | 162.00    | 33.87       | 5.20       | 176.40        |
| 500   | 39 | 23.89       | 2.94       | 70.47     | 33.53       | 2.70       | 90.74         |

Table 2: Simulated results using PSO and P&O technique under non-linear condition

| GW/m² | °C | Voltage (V) | Current (A) | Power (W) | Voltage (V) | Current (A) | Max power (W) |
|-------|----|-------------|------------|-----------|-------------|------------|---------------|
| 550   | 25 | 26.40       | 3.00       | 79.20     | 34.60       | 2.99       | 103.45        |
| 800   | 25 | 38.24       | 4.30       | 164.65    | 39.10       | 4.22       | 165.37        |
| 600   | 25 | 28.68       | 3.40       | 97.76     | 38.22       | 3.17       | 121.07        |
| 400   | 25 | 19.12       | 2.27       | 43.58     | 37.08       | 2.10       | 78.09         |
| 1000  | 35 | 35.50       | 5.36       | 190.30    | 36.69       | 5.40       | 198.26        |
| 1000  | 40 | 33.83       | 5.15       | 181.00    | 35.23       | 5.48       | 193.06        |
| 1000  | 45 | 32.79       | 5.29       | 173.50    | 33.99       | 5.48       | 186.59        |
| 1000  | 25 | 38.23       | 5.44       | 210.05    | 39.70       | 5.29       | 210.05        |
| 750   | 29 | 35.84       | 4.09       | 146.88    | 37.69       | 3.99       | 150.43        |
| 800   | 35 | 38.23       | 3.94       | 150.87    | 36.15       | 4.30       | 155.56        |
| 600   | 40 | 28.67       | 3.50       | 100.39    | 33.87       | 3.24       | 110.07        |
| 950   | 45 | 34.32       | 4.72       | 162.00    | 33.87       | 5.20       | 176.40        |
| 500   | 39 | 23.89       | 2.94       | 70.47     | 33.53       | 2.70       | 90.74         |

CONCLUSION

Thus the comparative analysis of maximum power point techniques: perturbation and optimization and particle swarm optimization is done under linear and non-linear conditions using MATLAB and the maximum power point tracking is carried out and from the results it is seen that particle swarm optimization technique is best suitable under linear and non-linear conditions with reduced steady state oscillations because P&O fails to track maximum power point when the difference between the irradiation level of the shaded modules and non-shaded modules is large. Therefore, PSO is best suited.

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