

# New Approach on Mathematical Modeling of Photovoltaic Solar Panel

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## Abstract

Describing the Current-Voltage ( $I$ - $V$ ) characteristics that involves only one variable is the basic aim of this paper. A formula that describes the  $I$ - $V$  characteristics is found based on the information gathered, and the values of  $I$  and  $P$  (Power) are determined according to different values of  $V$ . Afterwards, the natural cubic spline interpolation method is used to build a mathematical model that can approximate those values. Finally, the bisection method is used as an optimization method in determining the value of  $V$  that can produce the maximum power. As a result, mathematical models that approximate the  $I$ - $V$  and  $P$ - $V$  characteristics are built. Through those models, the optimum values of  $V$  are determined. The major finding is the estimated values of  $V$ ,  $I$  and  $P$  that generate the most energy under a fix condition. For  $G = 221.19 \text{ W/m}^2$  and  $T_c = 28.59 \text{ }^\circ\text{C}$ , it is estimated that when  $V = 19.5724 \text{ V}$  and  $I = 1.0576 \text{ A}$ , the generated energy is the most which is  $20.3335 \text{ W}$ .

**Mathematics Subject Classification:** 62P30, 65Z05

**Keywords:** modeling; optimization; photovoltaic

## 1. Introduction

Resource for conventional energy decreases bit by bit with each passing day, stirring up worries among many. Economic and concerns over fossil fuels encourage the development of photovoltaic (PV) energy systems. As a kind of clean and renewable resource, PV energy has gained significant attention rapidly since the last decade, due to the high energy cost and adverse environmental impacts of conventional fossil fuels [3]. Malaysia's position in the equator zone enables a high amount of sunlight reception throughout the year, making PV energy a renewable energy resource that has a great potential.

Photovoltaic technology refers to the technology that converts solar energy directly into electricity, through the use of solar cells or similar devices. It is a technology that has been developed since the 20<sup>th</sup> century. This technology is growing rapidly, and is expected to reach full maturity in the 21<sup>st</sup> century [7].

The usage of this technology is very common and can be found just about everywhere. The simplest example would be some calculators in a house or an office. The small dark-colored panels that can be found on the calculator's surface are the ones we call *solar cells*. Other examples include traffic signals on the road, modern parking meters, roadside emergency telephones and many more.

A solar cell constitutes the basic unit of a PV generator which, in turn, is the main component of a solar generator. A photovoltaic generator, also known as a photovoltaic array, is the total system consisting of all PV modules connected in series or parallel with each other [6].

Solar energy, along with other renewable energy resources, does not deplete in source, is reliable, and environment-friendly. Grid-connected solar PV continued to be the fastest growing power generation technology, with a 70-percent increase in existing world capacity to 13 GW in 2008. This represents a sixfold increase in global capacity since 2004. Including off-grid applications, total PV existing worldwide in 2008 increased to more than 16 GW [9].

In an electrical circuit, the energy, or power generated is calculated through the equation

$$P = V \cdot I$$

where  $P$  representing power in watts (W),  $V$  the voltage in volts (V), and  $I$  the current in ampere (A). Figure 1 shows a model of a solar cell equivalent circuit.

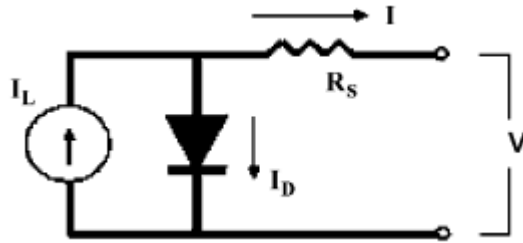


Figure 1. A four-parameter model of solar cell equivalent circuit.

The I-V characteristic of a PV cell is described by the following equations [8],

$$I = I_L - I_D \quad (1)$$

where  $I_L$  refers to the light current and  $I_D$  is the diode current. Using Shockley equation, the diode current can be expressed as

$$I_D = I_0 \left[ \exp\left(\frac{q(V + IR_s)}{\gamma k T_C}\right) - 1 \right]. \quad (2)$$

Substitute equation (2) into equation (1), and we get

$$I = I_L - I_0 \left[ \exp\left(\frac{q(V + IR_s)}{\gamma k T_C}\right) - 1 \right], \quad (3)$$

where

$$\gamma = A \cdot NCS \cdot NS, \quad (4)$$

$$I_L = \left(\frac{G}{G_R}\right) (I_{L,R} + \mu_{ISC} (T_C - T_{C,R})), \quad (5)$$

and

$$I_0 = I_{0,R} \left(\frac{T_C}{T_{C,R}}\right)^3 \exp\left[\left(\frac{q \varepsilon_G}{kA}\right) \left(\frac{1}{T_{C,R}} - \frac{1}{T_C}\right)\right]. \quad (6)$$

In the equations above,  $I_{L,R}$  refers to the light current at reference condition;  $I_0, I_{0,R}$  the reverse saturation current, actual and at reference condition respectively;  $T_C, T_{C,R}$  the cell temperature, actual and at reference condition respectively;  $G, G_R$  the irradiance, actual and at reference condition respectively;  $q$  the electron charge;  $R_S$  the series resistance;  $\gamma$  the shape factor;  $k$  the Boltzmann constant;  $A$  the completion factor;  $NCS$  the number of cells connected in series per module;  $NS$  the number of modules connected in series of the entire array;  $\mu_{ISC}$  the manufacturer supplied temperature coefficient of short-circuit current; and  $\varepsilon_G$  the material bandgap energy.

## 2. Algorithms

To produce the simulation data which would be used later, a formula is derived based on equation (3). By moving the variable  $I$  to one side of the equation, the following general formula is obtained:

$$I = \frac{-qV + \left( -LambertW \left( \frac{qR_S I_0 \exp \left( \frac{q(V + R_S I_L + R_S I_0)}{\gamma k T_C} \right)}{\gamma k T_C} \right) + \left( \frac{q(V + R_S I_L + R_S I_0)}{\gamma k T_C} \right) \right) \gamma k T_C}{qR_S} \quad (7)$$

By making all the variables except  $I$  and  $V$  constants, values of  $I$  can be calculated based on the values of  $V$ . The Lambert  $W$ -function, also called the *omega function*, is the inverse function of  $f(W) = W \cdot e^W$ . Banwell and Jayakumar [12] showed that a  $W$ -function describes the relation between voltage, current and resistance in a diode [5].

To simulate  $I$ , the natural cubic spline method is used to build the mathematical models using the simulation data produced earlier, and the bisection method is used to locate the optimal values of  $V$ ,  $I$  and  $P$  in the mathematical models of PV module.

### 2.1: Natural Cubic Spline Method

Equation (7) contains exponential function, thus making it difficult to locate the optimal point through its first derivative. In contrast, a polynomial interpolation has a much simpler form of first derivative. Interpolation plays

major roles in classical Numerical Analysis. Smooth interpolation functions like splines are used in data fitting, computer graphics, numerical differentiation, numerical integration of ordinary differential equations, numerical quadratures, etc. [1]. One of the most common methods of representing curves and surfaces in geometric modeling is the cubic spline interpolation with its parametric functions [11].

The natural cubic spline interpolation method is used to build the mathematical models that describe the I-V and P-V characteristics [10]. The algorithm for this method is given below:

Given a set of points  $x_0, x_1, \dots, x_n$ , where  $x_0 < x_1 < \dots < x_n$ , and  $a_0 = f(x_0), a_1 = f(x_1), \dots, a_n = f(x_n)$  for a function  $f$ .

(1) For  $i = 0, 1, \dots, n-1$  set  $h_i = x_{i+1} - x_i$ .

(2) For  $i = 1, 2, \dots, n-1$  set  $\alpha_i = \frac{3}{h_i}(a_{i+1} - a_i) - \frac{3}{h_{i-1}}(a_i - a_{i-1})$ .

(3) Set  $l_0 = 1, \mu_0 = 0, z_0 = 0$ .

(4) For  $i = 1, 2, \dots, n-1$  set

$$l_i = 2(x_{i+1} - x_{i-1}) - h_{i-1}\mu_{i-1},$$

$$\mu_i = h_i/l_i,$$

$$z_i = (\alpha_i - h_{i-1}z_{i-1})/l_i.$$

(5) Set  $l_n = 1, z_n = 0, c_n = 0$ .

(6) For  $j = n-1, n-2, \dots, 0$  set

$$c_j = z_j - \mu_j c_{j+1},$$

$$b_j = (a_{j+1} - a_j)/h_j - h_j(c_{j+1} + 2c_j)/3,$$

$$d_j = (c_{j+1} - c_j)/3h_j.$$

(7) Display the values  $a_j, b_j, c_j, d_j$  for  $j = 0, 1, \dots, n-1$ .

Algorithm is stopped.

From the values  $a_j, b_j, c_j, d_j$  above, a cubic polynomial  $S(x)$  is built where

$S(x) = S_j(x) = a_j + b_j(x - x_j) + c_j(x - x_j)^2 + d_j(x - x_j)^3$  for  $x_j \leq x \leq x_{j+1}$ .  $S(x)$  thus contains a number of piecewise functions with each function describing the characteristic to certain accuracy within the determined interval.

## 2.2: Bisection Method

The bisection method is used to determine the optimal point of the mathematical model built with the natural cubic spline interpolation method mentioned above. It has a simple concept, and can be applied easily in many situations. Some of its advantages are: the bisection method is always convergent. Since the method brackets the root, the method is guaranteed to converge. As an iteration is conducted, the interval gets halved. Therefore one can guarantee the error in the solution of the equation [2]. A journal written by Wu [13] also agrees that bisection method is globally convergent and has the asymptotic convergence of the sequence of interval diameters  $\{(b_n - a_n)\}_{n=1}^{\infty}$ , that is to say, the method has the important property that it will always converge to a real solution, although it is very slow in converging and only has linear convergence.

The algorithm for this method is as follows:

Given a function  $f$ , a tolerant value ( $TOL$ ), the left-end and right-end points of the interval containing the optimal point,  $a$  and  $b$ .

- (1) Determine the function's first derivative,  $f'$ .
- (2) Set  $L = a$  and  $R = b$ .
- (3) Find the value of  $M = (L+R)/2$ .
- (4) Determine the value of  $f'(M)$ . If it is 0 or  $|R - L| < TOL$ , stop the algorithm.
- (5) Determine the value of  $f'(L) \cdot f'(M)$ . If it is less than 0, set  $R = M$ .  
Else, set  $L = M$ .
- (6) Repeat the algorithm from step (3).

The final value of  $M$  before the algorithm ends is the voltage value which will give the P-V characteristic's maximum value or the optimal voltage value in the observed circuit.

### 3. Case Study

The renewable energy station of Universiti Malaysia Terengganu(UMT) has been using PV panels to produce electricity for various purposes. The fixed parameter values used in this research were taken from the renewable energy station as:

$$A = 0.7, \quad \varepsilon_G = 1.5 \text{ eV}, \quad \gamma = 15.4, \quad I_{MP,R} = 3.88 \text{ A}, \quad I_{SC,R} = 4.80 \text{ A},$$

$$\mu_{ISC} = 0.10 \quad \%/^{\circ}\text{C},$$

$$NCS = 11, \quad NS = 2, \quad R_S = 1.713 \text{ } \Omega, \quad V_{MP,R} = 16.5 \text{ V}, \quad V_{OC,R} = 23.8 \text{ V}$$

where  $I_{MP,R}$  refers to the maximum power current at reference condition,  $I_{SC,R}$  the short-circuit current at reference condition,  $V_{MP,R}$  the maximum power voltage at reference condition, and  $V_{OC,R}$  the open circuit voltage at reference condition.

A solar cell's performance is normally evaluated under the standard test condition, where  $G = 1000 \text{ W}/m^2$  and  $T_C = 25 \text{ }^{\circ}\text{C}$  [8]. Considering the standard condition values, the I-V characteristic can be described by the following model:

$$I(V) = \begin{cases} 4.8 & 0 \leq V < 8.4 \\ 4.8 - 0.0001 (V - 8.4) - 0.0002 (V - 8.4)^2 + 0.0002 (V - 8.4)^3 & 8.4 \leq V < 9.8 \\ 4.8 + 0.0005 (V - 9.8) + 0.0006 (V - 9.8)^2 - 0.0007 (V - 9.8)^3 & 9.8 \leq V < 11.2 \\ 4.8 - 0.0018 (V - 11.2) - 0.0022 (V - 11.2)^2 + 0.0014 (V - 11.2)^3 & 11.2 \leq V < 12.6 \\ 4.797 + 0.0002 (V - 12.6) + 0.0036 (V - 12.6)^2 - 0.026 (V - 12.6)^3 & 12.6 \leq V < 14 \\ 4.733 - 0.1427 (V - 14) - 0.1057 (V - 14)^2 + 0.0105 (V - 14)^3 & 14 \leq V < 15.4 \\ 4.355 - 0.3768 (V - 15.4) - 0.0615 (V - 15.4)^2 + 0.0113 (V - 15.4)^3 & 15.4 \leq V < 16.8 \\ 3.738 - 0.4825 (V - 16.8) - 0.014 (V - 16.8)^2 + 0.001 (V - 16.8)^3 & 16.8 \leq V < 18.2 \\ 3.038 - 0.5155 (V - 18.2) - 0.0096 (V - 18.2)^2 + 0.0013 (V - 18.2)^3 & 18.2 \leq V < 19.6 \\ 2.301 - 0.5348 (V - 19.6) - 0.0041 (V - 19.6)^2 + 0.0003 (V - 19.6)^3 & 19.6 \leq V < 21 \\ 1.545 - 0.5446 (V - 21) - 0.0029 (V - 21)^2 & 21 \leq V < 22.4 \\ 0.777 - 0.5525 (V - 22.4) - 0.0027 (V - 22.4)^2 + 0.0006 (V - 22.4)^3 & 22.4 \leq V \leq 23.8 \end{cases}$$

(8)

while the following model describes the P-V characteristic:

$$P(V) = \begin{cases} 4.8V & 0 \leq V < 5.6 \\ 26.88 + 4.7999(V - 5.6) - 0.0002(V - 5.6)^2 + 0.0002(V - 5.6)^3 & 5.6 \leq V < 7 \\ 33.6 + 4.8005(V - 7) + 0.0006(V - 7)^2 - 0.0007(V - 7)^3 & 7 \leq V < 8.4 \\ 40.32 + 4.7982(V - 8.4) - 0.0022(V - 8.4)^2 + 0.0025(V - 8.4)^3 & 8.4 \leq V < 9.8 \\ 47.04 + 4.8067(V - 9.8) + 0.0083(V - 9.8)^2 - 0.0097(V - 9.8)^3 & 9.8 \leq V < 11.2 \\ 53.759 + 4.7728(V - 11.2) - 0.0325(V - 11.2)^2 + 0.0258(V - 11.2)^3 & 11.2 \leq V < 12.6 \\ 60.448 + 4.8334(V - 12.6) + 0.0757(V - 12.6)^2 - 0.3998(V - 12.6)^3 & 12.6 \leq V < 14 \\ 66.266 + 2.6945(V - 14) - 1.6035(V - 14)^2 + 0.0603(V - 14)^3 & 14 \leq V < 15.4 \\ 67.061 - 1.4405(V - 15.4) - 1.3501(V - 15.4)^2 + 0.1461(V - 15.4)^3 & 15.4 \leq V < 16.8 \\ 62.799 - 4.3618(V - 16.8) - 0.7366(V - 16.8)^2 + 0.0136(V - 16.8)^3 & 16.8 \leq V < 18.2 \\ 55.286 - 6.3444(V - 18.2) - 0.6795(V - 18.2)^2 + 0.0105(V - 18.2)^3 & 18.2 \leq V < 19.6 \\ 45.101 - 8.185(V - 19.6) - 0.6352(V - 19.6)^2 + 0.0164(V - 19.6)^3 & 19.6 \leq V < 21 \\ 32.442 - 9.8671(V - 21) - 0.5663(V - 21)^2 - 0.0452(V - 21)^3 & 21 \leq V < 22.4 \\ 17.394 - 11.7186(V - 22.4) - 0.7561(V - 22.4)^2 + 0.18(V - 22.4)^3 & 22.4 \leq V \leq 23.8 \end{cases}$$

(9)

Using equations (8) and (9), we can easily plot the relationship curves between  $V$ ,  $I$  and  $P$  as shown in Figure 2:

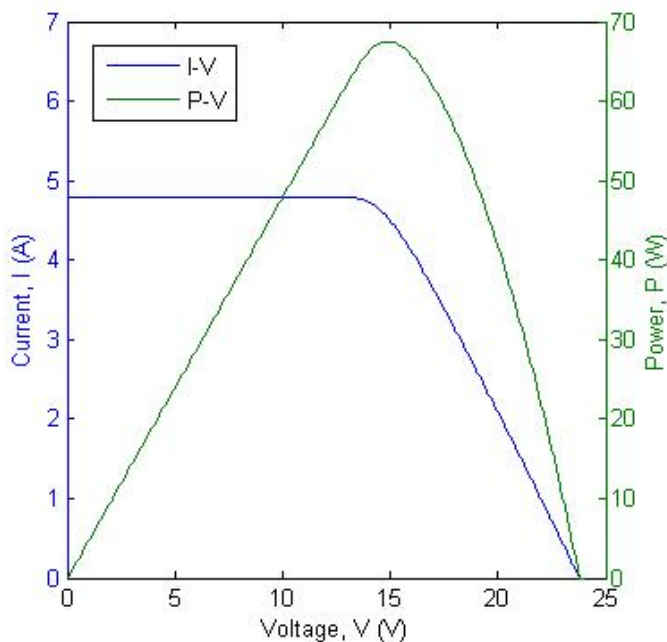


Figure 2. I-V and P-V characteristics curves of PV module under standard condition.



Further, Figures 3 and 4 show the characteristic curves of simulation data obtained through equations (1) and (7), and the mathematical models using the natural cubic spline method:

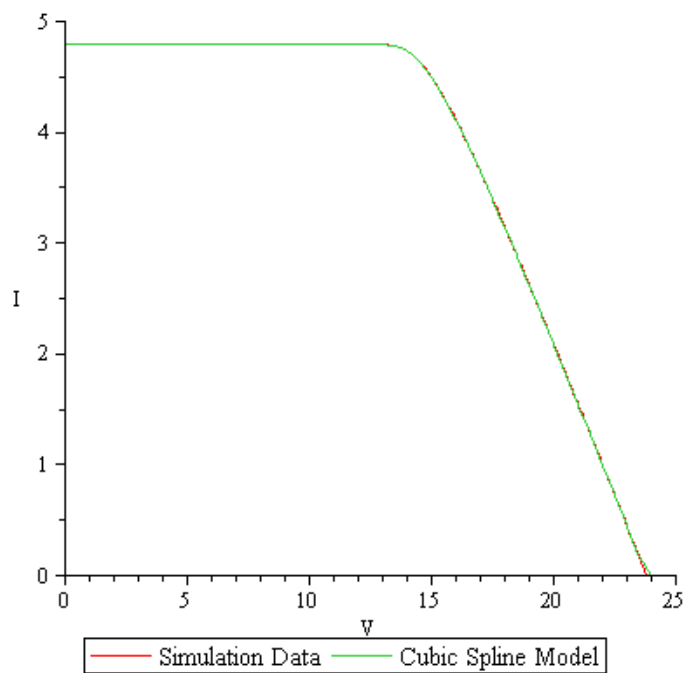


Figure 3. Comparison between simulation data and mathematical model for I-V characteristic under standard condition.

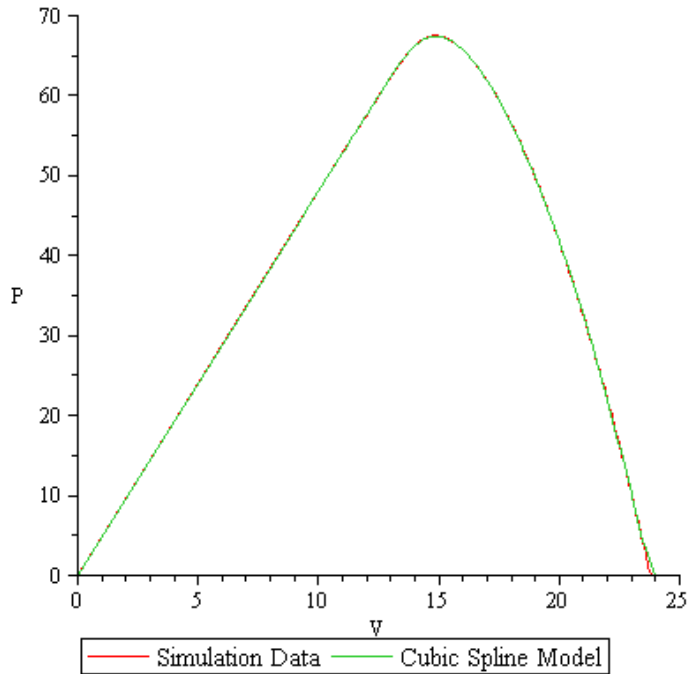


Figure 4. Comparison between simulation data and mathematical model for P-V characteristic under standard condition.

Through the P-V characteristic's mathematical model and the use of bisection method, it is found out that when  $V = 14.8844$  V, the optimal estimated value for  $P$  is obtained, that is 67.4365 W. The optimal value for  $I$  is 4.5314 A.

To determine how well a mathematical model's curve fits the simulation data, a correlation coefficient can be calculated to determine the goodness-of-fit between the two curves numerically. The correlation coefficient, denoted  $r$  (and sometimes  $R$ ), can be determined using the following equation taken from MathWorld website [4]:

$$r^2 = \frac{SS_{XY}^2}{SS_{XX} \cdot SS_{YY}} \quad (10)$$

In (10),  $X$  represents one set of data whereas  $Y$  contains corresponding data which matches with each data found in  $X$ .  $SS_{XY}$  is the sum of squared values obtained by multiplying the difference between a data in set  $X$  and the mean value of the set with the difference between the corresponding data in  $Y$  and the mean value of that set.  $SS_{XX}$  and  $SS_{YY}$  are obtained by summing each squared value of the difference between a data and the set's mean value in  $X$  and  $Y$ , respectively. The value of  $r^2$  lies between 0 and 1. The closer the value is to 1, the stronger is the relationship between  $X$  and  $Y$ .

Tables 1 and 2 represent the sets of data taken from I-V and P-V characteristics, respectively, each containing 20 pairs of corresponding data. The set  $X$  represents the data from simulation data, while  $Y$  is based on the mathematical model. The mean values for each set is included as well.

V	Simulation Data (X)	Cubic Spline Model (Y)
1.175	4.8	4.8
2.35	4.8	4.8
3.525	4.799999996	4.8
4.7	4.8	4.8
5.875	4.799999996	4.8
7.05	4.8	4.8
8.225	4.799999956	4.8
9.4	4.799999205	4.7999
10.575	4.799984535	4.800422036
11.75	4.799698837	4.798577425
12.925	4.794265687	4.796552719
14.1	4.719227221	4.7176835
15.275	4.401754334	4.400991993
16.45	3.903277117	3.904637412
17.625	3.331292762	3.330970266
18.8	2.725154503	2.7255248
19.975	2.100042879	2.099889258
21.15	1.463032309	1.46324475
22.325	0.817919089	0.818313688
23.5	0.166955469	0.1670786
Mean:	3.821130195	3.821189322

Table 1. Sets of data taken from I-V characteristics under standard condition

V	Simulation Data (X)	Cubic Spline Model (Y)
1.175	5.64	5.64
2.35	11.28	11.28
3.525	16.91999999	16.92
4.7	22.56	22.56
5.875	28.19999998	28.19996154
7.05	33.84	33.84002641
8.225	39.47999964	39.48022609
9.4	45.11999253	45.1185
10.575	50.75983646	50.76566249
11.75	56.39646133	56.37850122
12.925	61.965884	62.01312643
14.1	66.54110382	66.5194753
15.275	67.23679745	67.21977992
16.45	64.20890857	64.22911876
17.625	58.71403493	58.70680323
18.8	51.23290466	51.237008
19.975	41.94835651	41.94316484
21.15	30.94313334	30.9490407
22.325	18.26004367	18.26873768
23.5	3.923453531	3.91141
Mean:	38.75854552	38.75902713

Table 2. Sets of data taken from P-V characteristics under standard condition  
 Figures 5 and 6 show the data distribution based on the data sets in Tables 1 and 2.

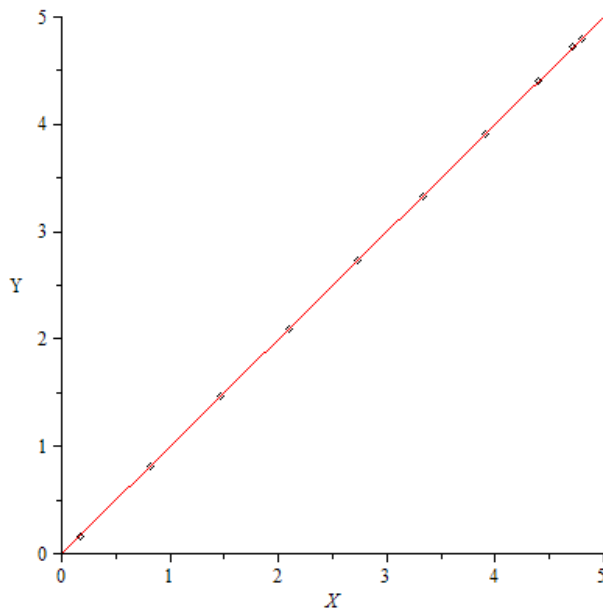


Figure 5. Distribution of data from Table 1, along with a straight line of  $X = Y$ .

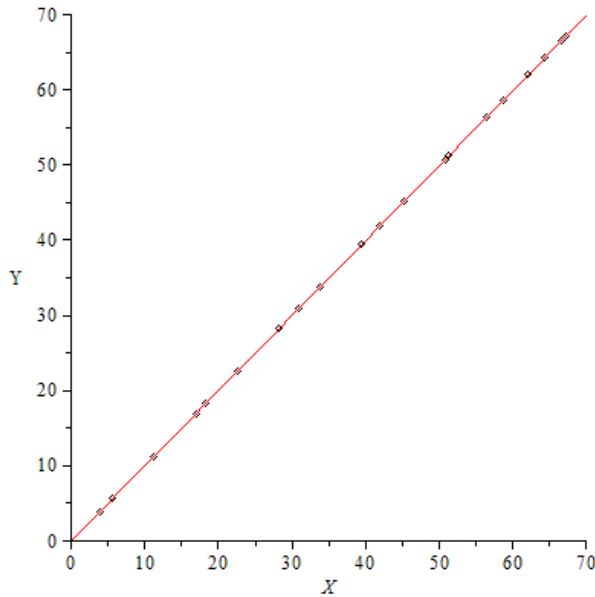


Figure 6. Distribution of data from Table 2, along with a straight line of  $X = Y$ .

From Table 1,  $SS_{XX} = 44.01482347$  ,  $SS_{YY} = 44.01008072$  , and  $SS_{XY} = 44.01244613$  . Therefore the correlation coefficient for I-V characteristic,

$$r^2 = \frac{(44.01244613)^2}{(44.01482347)(44.01008072)} = 0.999999732 .$$

As for the P-V characteristic, calculation based on Table 2 gives  $SS_{XX} = 8250.955890$  ,  $SS_{YY} = 8251.669508$  , and  $SS_{XY} = 8251.310647$  , thus

$$r^2 = \frac{(8251.310647)^2}{(8250.955890)(8251.669508)} = 0.999999504 .$$

Since both values of  $r^2$  are very close to 1, it can be said that the model curve fits the simulation data very well, almost the entire line.

Now we consider the conditions of the area surrounding the renewable energy station of UMT. Averaging the data values obtained from 2004 to 2006, we get the values of  $G = 221.19 \text{ W}/m^2$  , and  $T_C = 28.59 \text{ }^\circ\text{C}$ . Based on these values, the mathematical model for I-V characteristic is as follows:

$$I(V) = \begin{cases} 1.066 & 0 \leq V < 11.5 \\ 1.066 - 0.0002(V - 11.5) - 0.0001(V - 11.5)^2 + 0.0001(V - 11.5)^3 & 11.5 \leq V < 13.8 \\ 1.066 + 0.0007(V - 13.8) + 0.0005(V - 13.8)^2 - 0.0004(V - 13.8)^3 & 13.8 \leq V < 16.1 \\ 1.066 - 0.0026(V - 16.1) - 0.002(V - 16.1)^2 + 0.0013(V - 16.1)^3 & 16.1 \leq V < 18.4 \\ 1.065 + 0.0084(V - 18.4) + 0.0068(V - 18.4)^2 - 0.0165(V - 18.4)^3 & 18.4 \leq V < 20.7 \\ 0.92 - 0.2215(V - 20.7) - 0.1067(V - 20.7)^2 + 0.0155(V - 20.7)^3 & 20.7 \leq V \leq 23 \end{cases} \quad (11)$$

and for P-V:

$$P(V) = \begin{cases} 1.0658V & 0 \leq V < 2.3 \\ 2.451 + 1.0654(V - 2.3) - 0.0002(V - 2.3)^2 + 0.0001(V - 2.3)^3 & 2.3 \leq V < 4.6 \\ 4.901 + 1.0654(V - 4.6) + 0.0002(V - 4.6)^2 & 4.6 \leq V < 6.9 \\ 7.352 + 1.0658(V - 6.9) & 6.9 \leq V < 9.2 \\ 9.803 + 1.0654(V - 9.2) - 0.0002(V - 9.2)^2 + 0.0001(V - 9.2)^3 & 9.2 \leq V < 11.5 \\ 12.254 + 1.0666(V - 11.5) + 0.0007(V - 11.5)^2 - 0.0005(V - 11.5)^3 & 11.5 \leq V < 13.8 \\ 14.704 + 1.061(V - 13.8) - 0.0031(V - 13.8)^2 + 0.0022(V - 13.8)^3 & 13.8 \leq V < 16.1 \\ 17.155 + 1.0821(V - 16.1) + 0.0123(V - 16.1)^2 - 0.0098(V - 16.1)^3 & 16.1 \leq V < 18.4 \\ 19.59 + 0.9836(V - 18.4) - 0.0551(V - 18.4)^2 - 0.2072(V - 18.4)^3 & 18.4 \leq V < 20.7 \\ 19.04 - 2.5577(V - 20.7) - 1.4845(V - 20.7)^2 + 0.2151(V - 20.7)^3 & 20.7 \leq V \leq 23 \end{cases} \quad (12)$$

The relationship between  $V$ ,  $I$  and  $P$ , also the characteristic curves of simulation data and cubic spline model for both I-V and P-V characteristics are shown in Figures 7, 8 and 9:

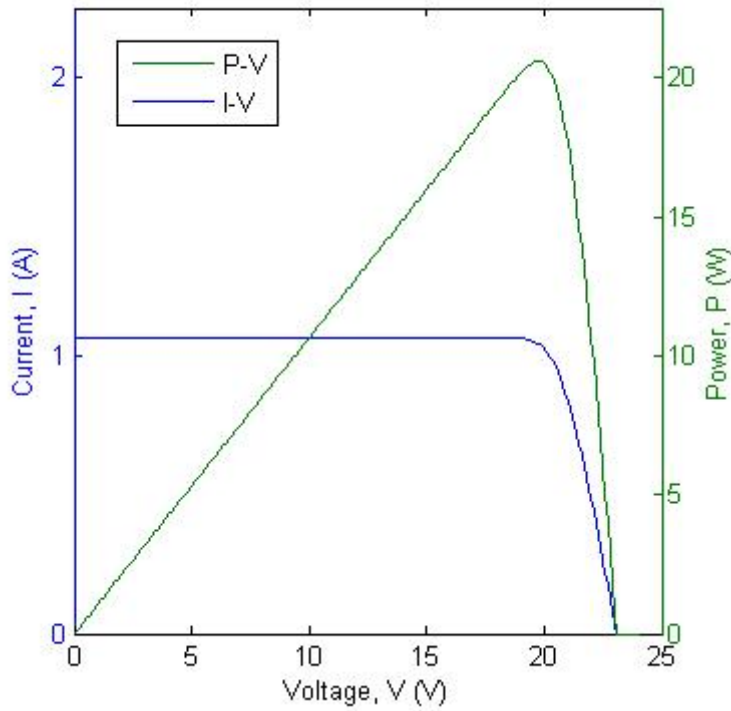


Figure 7. Relationship between I-V and P-V characteristics when  $G = 221.19 \text{ W/m}^2$  and  $T_c = 28.59 \text{ }^\circ\text{C}$ .

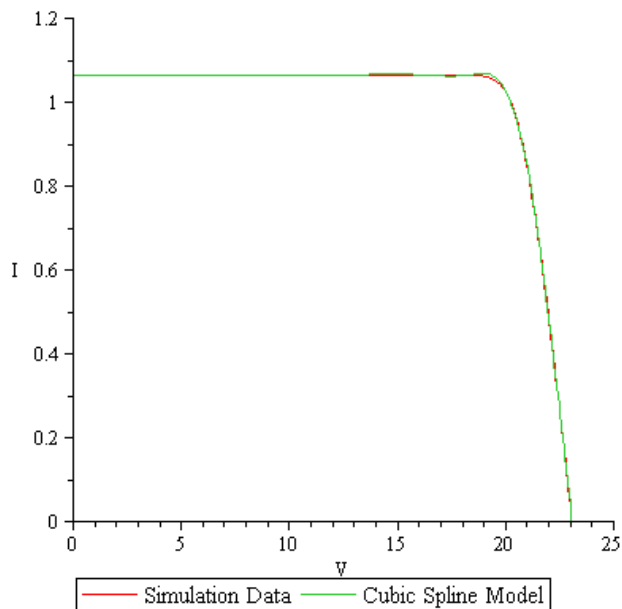


Figure 8. Comparison between simulation data and mathematical model for I-V characteristic when  $G = 221.19 \text{ W/m}^2$  and  $T_c = 28.59 \text{ }^\circ\text{C}$ .

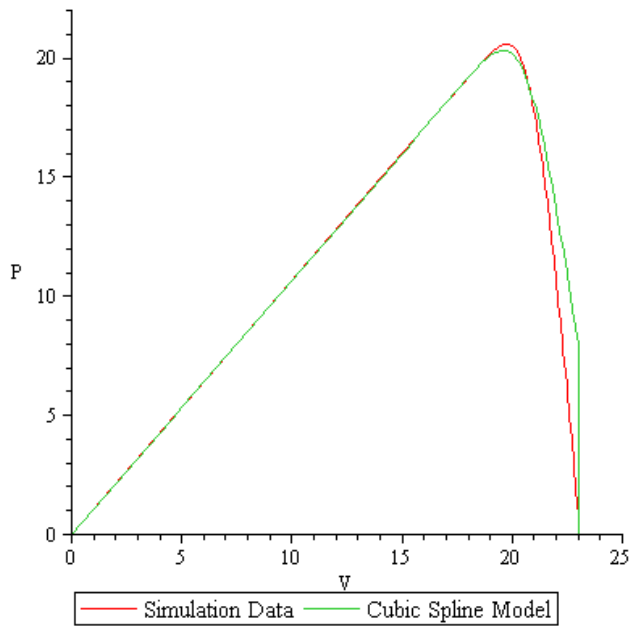


Figure 9. Comparison between simulation data and mathematical model for P-V characteristic when  $G = 221.19 \text{ W/m}^2$  and  $T_c = 28.59 \text{ }^\circ\text{C}$ .

The bisection method gives the optimal estimation of voltage value,  $V = 19.5724 \text{ V}$ . The matching values of  $P$  and  $I$  are  $20.3335 \text{ W}$  and  $1.0576 \text{ A}$ .

Once again, data are sampled from Figures 8 and 9 to test the goodness-of-fit. Tables 3 and 4 show the values and means for the sampled data, while Figures 10 and 11 give graphical presentation of their distributions.

V	Simulation Data (X)	Cubic Spline Model (Y)
1.15	1.065523545	1.066
2.3	1.065523544	1.066
3.45	1.065523544	1.066
4.6	1.065523543	1.066
5.75	1.065523543	1.066
6.9	1.065523542	1.066
8.05	1.065523542	1.066
9.2	1.065523542	1.066
10.35	1.065523541	1.066
11.5	1.065523541	1.066
12.65	1.06552354	1.065789838
13.8	1.065523531	1.066



14.95	1.065523382	1.0668579
16.1	1.065520755	1.066
17.25	1.065474339	1.062342138
18.4	1.064656852	1.065
19.55	1.051068011	1.058558562
20.7	0.919796939	0.92
21.85	0.545704035	0.547737813
23	0.034443262	0.0346955
Mean:	0.979923504	0.980649088

Table 3. Sets of data taken from I-V characteristic when  $G = 221.19 \text{ W/m}^2$  and  $T_c = 28.59 \text{ }^\circ\text{C}$

V	Simulation Data (X)	Cubic Spline Model (Y)
1.15	1.225352077	1.22567
2.3	2.450704151	2.451
3.45	3.676056227	3.676097588
4.6	4.901408298	4.901
5.75	6.126760372	6.1264745
6.9	7.35211244	7.352
8.05	8.577464513	8.57767
9.2	9.802816586	9.803
10.35	11.02816865	11.02809759
11.5	12.25352072	12.254
12.65	13.47887278	13.48075531
13.8	14.70422473	14.704
14.95	15.92957456	15.92339618
16.1	17.15488416	17.155
17.25	18.37943235	18.40077718
18.4	19.58968608	19.59
19.55	20.54837962	20.33314495
20.7	19.03979663	19.04
21.85	11.92363317	14.46253396
23	0.792195026	7.9214067
Mean:	10.94675216	11.4203012

Table 4. Sets of data taken from P-V characteristic when  $G = 221.19 \text{ W/m}^2$  and  $T_c = 28.59 \text{ }^\circ\text{C}$

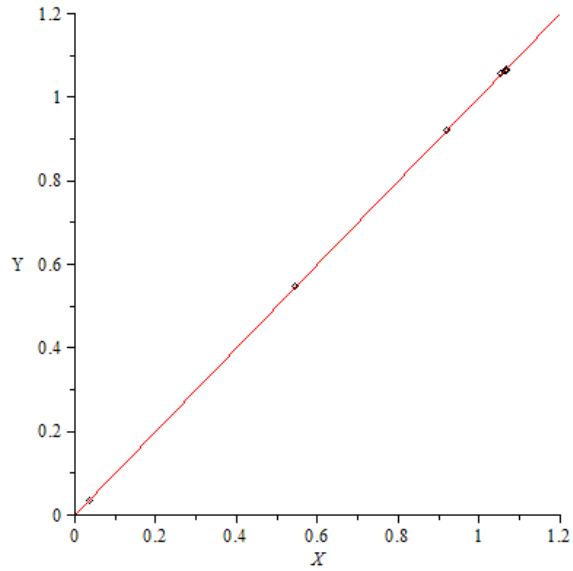


Figure 10. Distribution of data from Table 3, along with a straight line of  $X = Y$ .

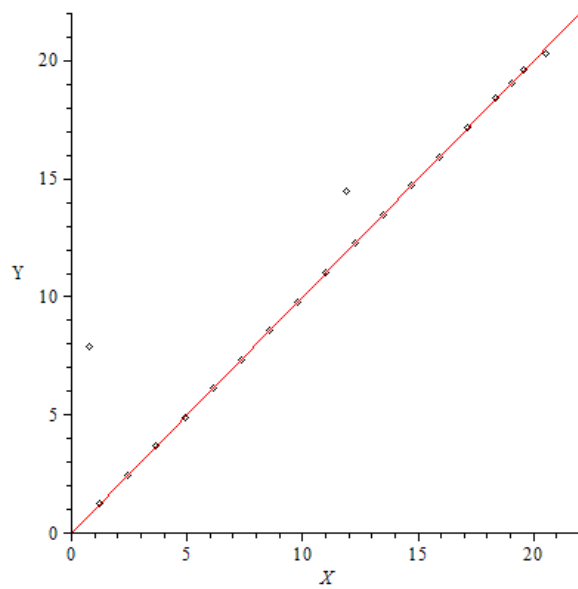


Figure 11. Distribution of data from Table 4, along with a straight line of  $X = Y$ .

For I-V characteristic based on Table 3,

$$SS_{xx} = 1.208237491, SS_{yy} = 1.207875636, \text{ and } SS_{xy} = 1.208024400,$$

so  $r^2 = 0.999946775$ . As for the P-V characteristic,

$$SS_{xx} = 776.4904379, SS_{yy} = 685.6334589, \text{ and } SS_{xy} = 704.6451852,$$

thus  $r^2 = 0.932637476$ . Although both values are not as good as the ones under standard condition, they are still very close to 1, meaning that the model curve fits the simulation data very well in this case study.

#### **4. Conclusion and Discussion**

In both case studies, built mathematical models fit closely to the simulation data, and their curves resemble that of a typical I-V and P-V characteristics. The values of  $r^2$  calculated from sampled data from each curve also support this observation. However, it is noticeable that the larger error exists near the end in Figure 9, with the largest error having a value close to 6.0476 W. This is due to the reason that the size of subintervals used when building the mathematical models was not small enough, and this has caused  $S(x)$  to be unable to give better approximations for intervals where the value of the simulation data changes drastically. The number of decimal places used in calculations might be another factor contributing to the problem occurred. Nevertheless, the value of  $r^2$  which is close to 1 strongly suggests that the cubic spline model can be accepted as a way to describe the characteristic curve.

By using more subintervals of smaller size and also more decimal places in calculations, it is believed that mathematical models which are able to fit the simulation data better can be built.

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