

Unknown Parameters Extraction and Characteristics Simulation of Photovoltaic Module using 2D-scanning and Datasheet at the Standard Test Condition

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Abstract

Original Research Article

Single-diode model is found to be more effective for photovoltaic module characterization in terms of computational time and accuracy compared to two-diode or three-diode models. To analyze the output characteristics of photovoltaic cell and module using single diode model, 5-unknown parameters including photo-current, reverse saturation current of diode, diode quality factor, series resistance and parallel resistance (or shunt resistance) have to be extracted. In this paper, we construct the two-dimensional search region with orthogonal basis of series resistance and diode quality factor, and present a mathematical modeling procedure and extraction algorithm for computing the photo current and the reverse saturation current corresponding to each point in the search region. The algorithm was implemented in MATLAB environment to extract the unknown parameters of R.T.C. France silicon photovoltaic cell and STP6-120/36 silicon photovoltaic module, and compare the results with previous extraction. The output characteristics of the mono-crystal silicon photovoltaic module JT100-12M are also presented.

Keywords: Photovoltaic module, Single-diode model, Unknown parameters extraction, 2D-scanning, I-V curve.

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1. Introduction

Today, when the global demand for energy is increasing and fossil fuel resource depletion is in the forefront, the active use of renewable energy sources is an important strategic task. Statistics show that oil and natural gas are available for 50 years and coal for 150 years and that, in total, fossil fuels will remain almost unchanged in the 21st century. In addition, with the attempt to abolish coal use by 2040 to prevent global warming, climate change and environmental pollution due to the widespread use of coal, power generation using river-hydro energy, wind energy, solar energy, sea tidal and wave energy, biomass energy, nuclear fission and fusion energy is gaining momentum worldwide. Among them, solar energy utilization plays an important role in reducing greenhouse gas emissions, mitigating climate change, and protecting the ecological environment of humans, animals and plants. Solar energy is the richest source on Earth and the increasing use of solar energy is due to the ever-decreasing rate of conversion efficiency and development costs.

Solar cell characterization is of great significance in evaluating the performance of solar cell and modules and ensuring ef-

ficient utilization. The unknown parameters obtained through solar cell characterization and current-voltage curves under given operating conditions are significantly used in the design of solar PV systems, the enhancement of power generation capability, the diagnosis of power system faults, and the optimization of solar array design on satellite. To date, there have been many reports of extracting the unknown parameters of solar cell and modules using analytical, numerical and artificial intelligence techniques and predicting the current-voltage curves related to the operating temperature and solar radiation intensity. The methods for extracting unknown parameters based on analysis and numerical calculations include continuous discretization[1], nonlinear least-squares approximation[2], open-circuit voltage point, short-circuit current point, and algebraic search algorithm using three points of optimum power point[3], Gauss-Seidel algorithm and analytical method[4], and artificial intelligence techniques introduced in recent years. The goal of the updating of the method of unknown parameter extraction of solar cell and module is to reduce the computational burden while reducing the mismatch between the current-voltage simulation curve and the real curve under different conditions. The solar cell power can be modeled by a single diode model, a double diode model, a



nd a triple diode model depending on the cell geometry. As the number of diodes in the model increases, the calculation accuracy increases, but long computation time is required. Previous studies have shown that the single diode model is superior to other models in terms of computational accuracy and computational complexity, and is well suited for unknown parameter extraction and current-voltage characteristic plot. We have reduced the fundamental dimension of the optimal solution to two by using the relationship between five unknown parameters, photocurrent and diode reverse saturation current, diode quality factor, series resistance and shunt resistance, for solar cell characterization based on the single-diode model. Generally, the diode quality factor of a solar cell is searched between 1 and 2, and the series resistance is found from 0 to the end of the user-defined value.

In this study, we set the range of search for series resistance from the characteristic parameters of the solar cell and module to the standard conditions, and set the range of search for the idealization coefficient of the diode to be the region of the region centered near the expected values according to the solar cell type. For each point of the square search region set above, the other three parameters were calculated and the current-voltage curve data of the solar cell were obtained from the five unknown parameters calculated as above. The best solution was chosen to find the parameter pair with the smallest error between the obtained current-voltage curve simulation data and the experimental data under the conditions. The algorithm of unknown parameter extraction and characterization described below, first extracts the unknown parameters under the given test conditions by using the initial data of electrical characteristics under certain test conditions (e.g., standard test conditions : 25 °C and AM1.5) of the solar cell, and then calculates the current-voltage simulation curve data under any operating conditions. The reliability of the algorithm was evaluated using the well-known R.T.C

France cell and the STP6-120/36 module measurement data. The key point of our proposed method of unknown parameter extraction and characterization is, first, the reduction of the search space dimension from 5D to 2D. For this purpose, the photocurrent, reverse saturation current and shunt resistance are treated as dependent parameters that can be calculated by the series resistance and the diode quality factor. Second, it reduces the computational burden and reduces the computational time by narrowing the search space reasonably. For the sake of computational time reduction, the central value of the diode quality factor search interval was chosen according to the cell geometry, using the maximum value that the series resistance can have from the current-voltage curve and the minimum value of the parallel resistance.

2. Unknown parameters extraction method

Solar cells can be equivalently modeled using a single diode model, a two-diode model, and a three-diode model. As the number of diodes increases, the accuracy of the calculation of the unknown parameters increases, while the computational effort increases. Recent research results have shown that the single diode model is well suited for the uncertain parameter extraction and characterization of photovoltaic module [5-15].

The solar cell and module can be equivalent to the model shown in Fig. 1. The solar module is mainly composed of series connected solar cells for voltage expansion and there are few cases of parallel connection. Taking the diode quality factor of a solar cell module as the product of the number of series-connected cells, we can generally express the relationship between the output current and the output voltage for a solar cell and a module by Eq.(1).

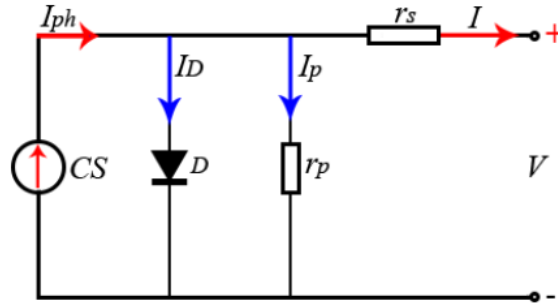


Fig. 1. Single-diode model of photovoltaic cell and module

$$I = I_{PH} - I_{SAT} \left(\exp \left(\frac{q(V + IR_s)}{AN_s kT} \right) - 1 \right) - \frac{V + IR_s}{R_p} \quad (1)$$

Where:

I : Output current of a solar cell or module (A)

V : Output voltage (V) of a solar cell or module

I_{PH} : Photocurrent of a solar cell or module (A)

I_{SAT} : The reverse saturation current of the diode (A)

A : The diode quality factor (no unit) of the solar cell that constitutes the module

R_s : The series resistance of the module (Ω)

R_p : Parallel resistance of the module (Ω)

N_s : Number of serially connected cells

q : The charge of the electron (C)

k : Boltzmann constant (1.38 e-23J/K)

2.1. Dimensionality Reduction of Unknown Parameters Matrix

Substituting the current and voltage corresponding to the two points A(0, I_{sc}) and B(V_{oc} , 0) of the current-voltage curve into Eq. (1), respectively, we obtain the following equations:

$$I_{sc} = I_{PH} - I_{SAT} \left(\exp \left(\frac{qI_{sc}R_s}{AN_s kT} \right) - 1 \right) - \frac{I_{sc}R_s}{R_p} \quad (2)$$

$$0 = I_{PH} - I_{SAT} \left(\exp \left(\frac{qV_{oc}}{AN_s kT} \right) - 1 \right) - \frac{V_{oc}}{R_p} \quad (3)$$

From Eqs. (2) and (3), the photocurrent can be expressed in two ways:

$$I_{PH} = I_{sc} \left(1 + \frac{R_s}{R_p} \right) + I_{SAT} \left(\exp \left(\frac{qI_{sc}R_s}{AN_s kT} \right) - 1 \right) \quad (4)$$

$$I_{PH} = I_{SAT} \left(\exp \left(\frac{qV_{oc}}{AN_s kT} \right) - 1 \right) + \frac{V_{oc}}{R_p} \quad (5)$$

From Eqs. (4) and (5), we have.

$$I_{sc} \left(1 + \frac{R_s}{R_p} \right) + I_{SAT} \left(\exp \left(\frac{qI_{sc}R_s}{AN_s kT} \right) - 1 \right) = I_{SAT} \left(\exp \left(\frac{qV_{oc}}{AN_s kT} \right) - 1 \right) + \frac{V_{oc}}{R_p} \quad (6)$$

By combining and aligning the reverse saturation current term of the diode in Eq. (6), the expression (7) in terms of open circuit voltage, short circuit current, series resistance and shunt resistance, and diode quality factor is obtained. Substituting Eq.

(7) into Eq. (5), Eq. (8) is obtained which shows the relationship between photocurrent, open circuit voltage, short circuit current, diode quality factor, series resistance and shunt resistance.

$$I_{SAT} = \frac{V_{oc} - I_{sc}(R_p + R_s)}{R_p \left(\exp \left(\frac{qI_{sc}R_s}{AN_s kT} \right) - \exp \left(\frac{qV_{oc}}{AN_s kT} \right) \right)} \quad (7)$$

$$I_{PH} = \left(\exp \left(\frac{qV_{OC}}{AN_S kT} \right) - 1 \right) \frac{V_{OC} - I_{SC}(R_P + R_S)}{R_P \left(\exp \left(\frac{qI_{SC}R_S}{AN_S kT} \right) \right)} + \frac{V_{OC}}{R_P} \quad (8)$$

From Eqs. (7) and (8), it can be concluded that the photocurrent and the reverse saturation current of the diode among the five unknown parameters of the solar cell and module can be calculated from the other three parameters, open circuit voltage, s

hort circuit current and operating temperature of the cell.

On the other hand, substituting the current and voltage of the optimal operating point into Eq. (1), we obtain.

$$I_M = I_{PH} - I_{SAT} \left(\exp \left(\frac{q(V_M + I_M R_S)}{AN_S kT} \right) - 1 \right) - \frac{V_M + I_M R_S}{R_P} \quad (9)$$

Modification of Eq. (9) yields.

$$R_P = \frac{V_M + I_M R_S}{I_{PH} - I_M - I_{SAT} \left(\exp \left(\frac{q(V_M + I_M R_S)}{AN_S kT} \right) - 1 \right)} \quad (10)$$

Equation (10) shows the relationship between the series resistance and shunt resistance.

Using Eqs. (7), (8), and (10), the other three unknown para

meters can be calculated from the series resistance of the solar cell and module and the idealization coefficient of the diode. Rewriting Eqs. (7), (8), and (10) under standard test conditions, the following system of equations (11) is obtained:

$$\begin{cases} I_{SATN} = \frac{V_{OCN} - I_{SCN}(R_{PN} + R_{SN})}{R_{PN} \left(\exp \left(\frac{qI_{SCN}R_{SN}}{A_N N_S kT_N} \right) - \exp \left(\frac{qV_{OCN}}{A_N N_S kT_N} \right) \right)} : (11-A) \\ I_{PHN} = \left(\exp \left(\frac{qV_{OCN}}{A_N N_S kT_N} \right) - 1 \right) \frac{V_{OCN} - I_{SCN}(R_{PN} + R_{SN})}{R_{PN} \left(\exp \left(\frac{qI_{SCN}R_{SN}}{A_N N_S kT_N} \right) \right)} + \frac{V_{OCN}}{R_{PN}} : (11-B) \\ R_{PN} = \frac{V_{MN} + I_{MN}R_{SN}}{I_{PHN} - I_{MN} - I_{SATN} \left(\exp \left(\frac{q(V_{MN} + I_{MN}R_{SN})}{A_N N_S kT_N} \right) - 1 \right)} : (11-C) \end{cases} \quad (11)$$

The subscript N in the set of equations (11) means that the values are tested under standard test conditions. The manufacturer's side delivers the electrical and mechanical parameters of t

he solar cell and module measured under standard test conditions to the user side. The data measured under standard test conditions include open circuit voltage, short circuit current, voltag

e and current of the optimum operating point, maximum power and duty cycle, cell and module efficiency, temperature change coefficient of open circuit voltage, temperature change coefficient

ent of short circuit current, and cell temperature calculation coefficient ($NOCT$).

$$\begin{cases} R_{PN_MIN} = \frac{V_{MN}}{I_{SCN} - I_{MN}}, (12 - A) \\ R_{SN_MAX} = \frac{(V_{OCN} - V_{MN})}{I_{MN}}, (12 - B) \end{cases} \quad (12)$$

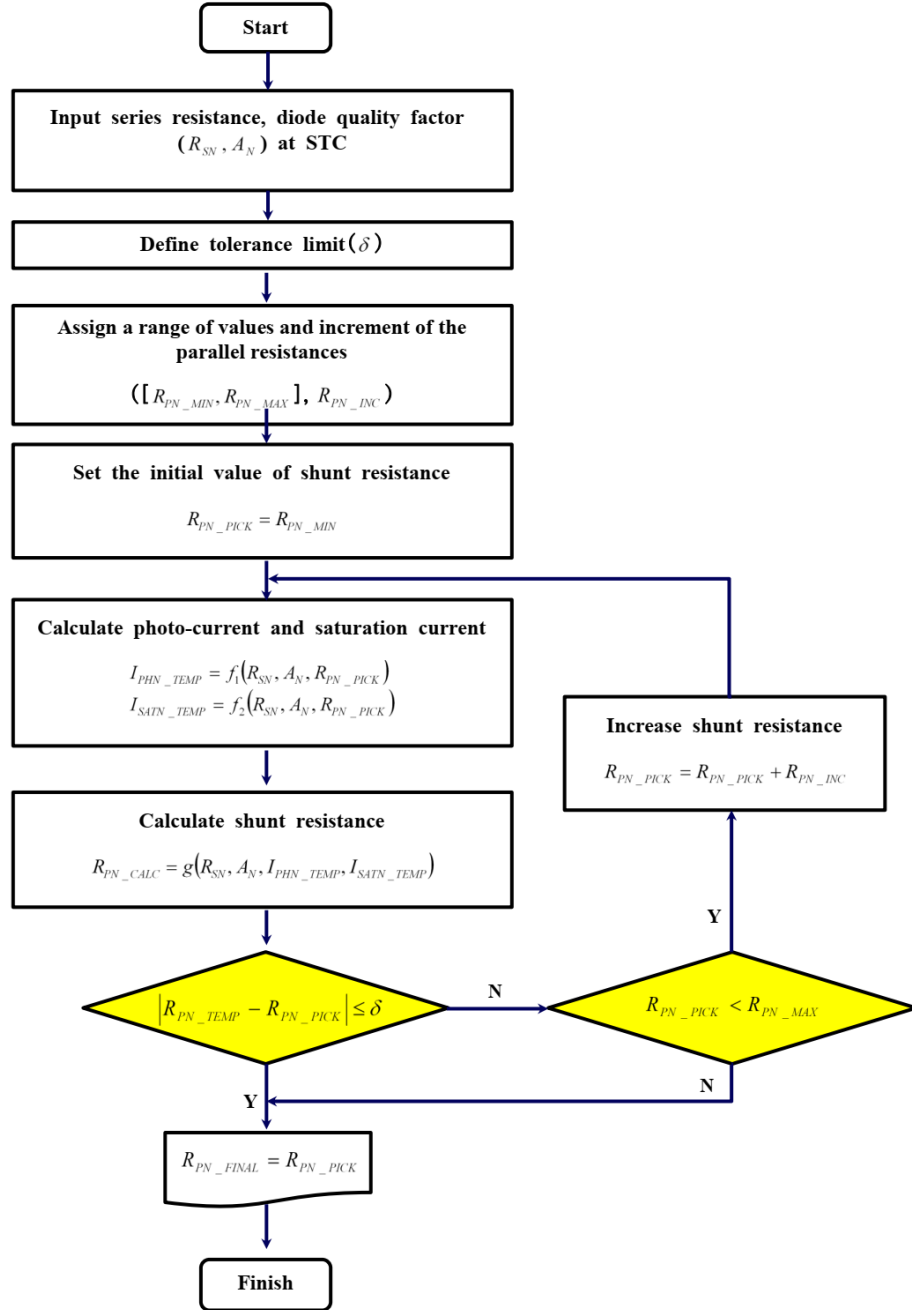


Fig. 2. Algorithm flow diagram for calculating shunt resistance from series resistance and diode quality factor.

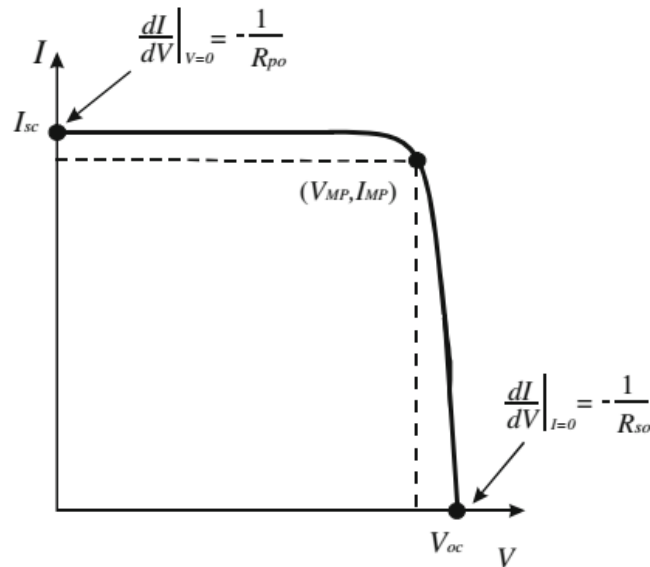


Fig. 3. Relationship between current-voltage curve and parasitic resistance value of solar cell.

Equations (11-A) and (11-B) are the expressions for calculating the reverse saturation current and photocurrent of the diode from the series resistance and shunt resistance of the solar cell or module and the idealization coefficient of the diode. On the other hand, as shown in Eq. (11-C), shunt resistance is calculated by the photocurrent represented by the shunt resistor itself as well as the series resistance and diode quality factor, and the reverse saturation current of the diode. This relationship suggests that, given the series resistance and diode quality factor, the shunt resistance, photocurrent and reverse saturation current calculations of the diode are not straight-line but ring-like. Given the series resistance of the solar cell or module and the idealization coefficient of the diode, the algorithm flow chart (F) for extracting the other three unknown parameters is shown in Fig. 2. In the flowchart, f1, f2, and g represent Eqs. (11-A), (11-B), and (11-C), respectively. The factors determining the accuracy and time of shunt resistance calculation in the algorithm are the minimum, maximum and increment of shunt resistance. The ideal value of shunt resistance is infinite, but in the simulation calculations it is usually set to 10000ohms, and the increment is chosen according to the accuracy required by the computer. Figure 3 shows the relationship between the current-voltage curve of the solar cell and the series resistance and shunt resistance values. As can be seen from the figure, the series resistance and shunt resistance are expressed as the inverse of the value of the derivative of the voltage versus the current at the open-circuit voltage point and the short-circuit current point, respectively, so that the maximum and minimum values of shunt resistance can be selected using Eq. (12). In Eq. (12), V_{MN} and I_{MN} are used as initial data in the calculation of unknown parameters as optimal operating point voltage and current values under normal conditions. Using the above algorithm, the diode quality factor

and shunt resistance corresponding to series resistance are determined.

2.2. Unknown parameters extraction of photovoltaic module under the standard test condition

Under standard test conditions, the operating temperature of the solar cell is 25 °C and the atmospheric mass is 1.5 (solar radiation intensity 1000 W/m²). The measured electrical parameters of the solar cell and module under standard conditions and the open-circuit voltage and short-circuit current variation coefficients with temperature are provided by the manufacturer. However, the operating environment of solar cells varies with season and time, so the current-voltage curve changes. Accurate prediction of the output characteristics of solar modules and arrays is an indispensable priority step in the design and management of PV systems. The calculation flow chart for the extraction of unknown parameters of solar cell modules under normal conditions is shown in Fig. 4. The green highlighted part in the flowchart indicates the algorithm F mentioned above.

The maximum value of series resistance is calculated using Eq. (12-B), and the increment value is chosen according to the required calculation accuracy. The diode quality factor values are generally chosen in the interval (1, 2), but we refer to Table 1. Table 1 gives a rough estimate of the diode quality factor values for the solar cell fabrication technique, but it cannot be said that the exact value is. Hence, we set up a diode quality factor search interval centered on the values concluded by the authors according to the solar cell topology. The upper and lower bounds that define the search range were chosen as values normally deviated from the central value to ± 0.2 .

Table 1. Diode quality factor for solar cell types

Fabrication techniques	Diode quality factor
Mono crystal silicon	1.2
Poly crystal silicon	1.3
Single-junction amorphous silicon	1.8
Double-junction amorphous silicon	3.3
Triple-junction amorphous silicon	5
CdTe	1.5
CIS	1.5
GaAs	1.3

In the step, indicated in red in the unknown parameter extraction flowchart shown in Fig. 4, the mean square error of the calculated current-voltage curve points of a solar cell or module

from a given parameter pair using Eq. (13) and the experimentally measured values is calculated.

$$RMSE = \sqrt{\frac{\sum (I_{i_CAL} - I_{i_MEASURE})^2}{n}} \quad (13)$$

In Eq. (13), n means the number of measurement points, and I_{i_CALC} and $I_{i_MEASURE}$ mean the calculated current and measured current intensities for the same operating voltage, respectively. Given the unknown parameters of a solar cell or module, one can use either the Newton-Raphson method or the Lambert W function to plot the current-voltage curve. Using Lambert W function, the current-voltage relationship in the implicit form can be obtained by obtaining the explicit form of the function $I=f(V)$ to calculate the current intensity at a given voltage. Only Lambert W functions are available in MATLAB environment, and in a general C++ environment, the user can override

the computational module or implement it through interlocking with MATLAB. The Newton-Raphson method is an approximate solution that can obtain with high accuracy the solution of equations that are difficult to solve analytically, and can be applied to almost all equations. The point to consider in using the Newton-Raphson method is that the solution obtained is valid only if the initial value of the calculation and the solution are in the same monotone interval. As shown in Fig. 4, the main factors driving the search for unknown parameters are the series resistance and the diode quality factor, which implements the full search in a two-dimensional search space consisting of these two parameters.

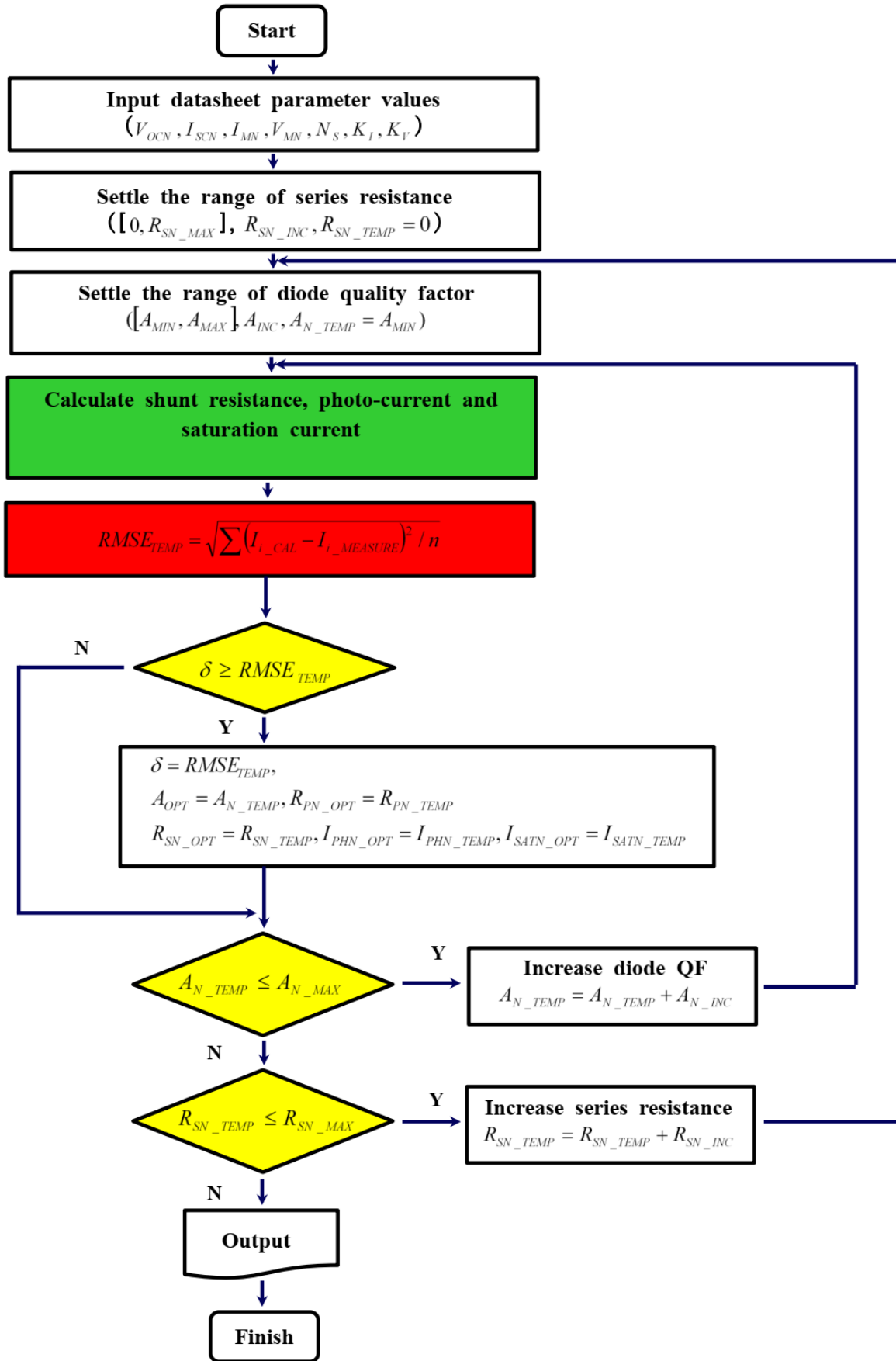


Fig. 4. Unknown parameter extraction flow diagram of solar cell and module at STC

3. Algorithm implementation and results analysis in Matlab environment

According to the calculation flow chart shown in Figs. 2 and 4, the unknown parameters of the solar cell and module are extracted, and the code is written in MATLAB environment to calculate the current-voltage curve points for any operating conditions and the results are analyzed.

3.1. Mean square error comparison

To perform a comparative analysis with the previous calculation methods, we performed simulation calculations on the RTC France cell and STP6-120/36 module and measured the computation time.

The electrical parameters of the RTC France cell and STP6-120/36 module are shown in Table 3.

Table 3. Electrical characteristics of the R.T.C. France cell and STP6-120/36 module (experimental measurements - 1000 W/m²)[2]

	R.T.C France Cell(33°C)	STP6-120/36 Module(55°C)
I_{sc} (A)	0.7603	7.48
I_M (A)	0.6894	6.83
V_{oc} (V)	0.5728	19.21
V_M (V)	0.4507	14.93
N_s	1	36

The results of the unknown parameter calculations of the R.T.C. France cell and STP6-120/36 module are shown in Tables 4 and 5, respectively.

The unknown parameters of R.T.C.France cell and STP6-1

20/36 module correspond to the case of 33 °C and 55 °C cell temperature and 1000 W/m² solar radiation, respectively. Fig. 5 shows the distribution of the mean square error calculated from the current values calculated in the 2D search region and the measured current values for the R.T.C.France cell.

Table 4. Results of the calculation of the unknown parameters of the R.T.C.France cell and comparison of mean square error

	I_{ph} (A)	I_{sat} (μA)	A	R_s (Ω)	R_p (Ω)	$RMSE(A)$
2DSCAN	0.76080	0.36118	1.4940	0.0390	52.34974	7.5278e-4
SMA(2021)[17]	0.76079	0.31068	1.51770	0.03655	52.89000	7.7301e-4
SSA(2020)[18]	0.76116	0.8987	1.5900	0.031595	96.9350	2.8726e-3
SDO(2019)[20]	0.7608	0.3230	1.4812	0.03640	53.71850	9.8602e-4
FFO(2020)[16]	0.7608	0.3223	1.5215	0.0364	53.8989	7.75184e-4
LAPO(2020)[19]	0.76071	0.96105	1.5980	0.031142	99.1440	2.9350e-3

In the figure, there is an optimal search point in the region marked with deep blue.

Table 5. Unknown parameter estimation results and mean square error comparison of STP6-120/36 modules

	$I_{ph} (A)$	$I_{sat} (\mu A)$	A	$R_s (\Omega)$	$R_p (\Omega)$	$RMSE(A)$
2DSCAN	7.4905	2.2016	47.0957	0.1638	1392.2985	1.5706e-2
SMA(2021)[17]	7.4648	2.1907	49.7412	0.1671	1500.000	1.3990e-2
SSA(2020)[18]	7.5035	32.3970	54.9396	0.1032	691.1640	5.6021e-2
SDO(2019)[20]	7.4725	2.3350	45.3636	0.1656	799.9164	1.6601e-2
HISA(2019)[7]	7.4753	1.9309	44.8004	0.1689	570.1977	1.4251e-2
LAPO(2020)[19]	7.5116	2.2000	53.2476	0.1170	1407.9000	4.1306e-2

The minimum value of the mean square error in the search interval is unique, and the corresponding diode quality factor and series resistance are 1.490 and 0.390 Ω , respectively.

In the figure, the horizontal axis is the axis of the diode quality factor and the vertical axis is the axis of series resistance.

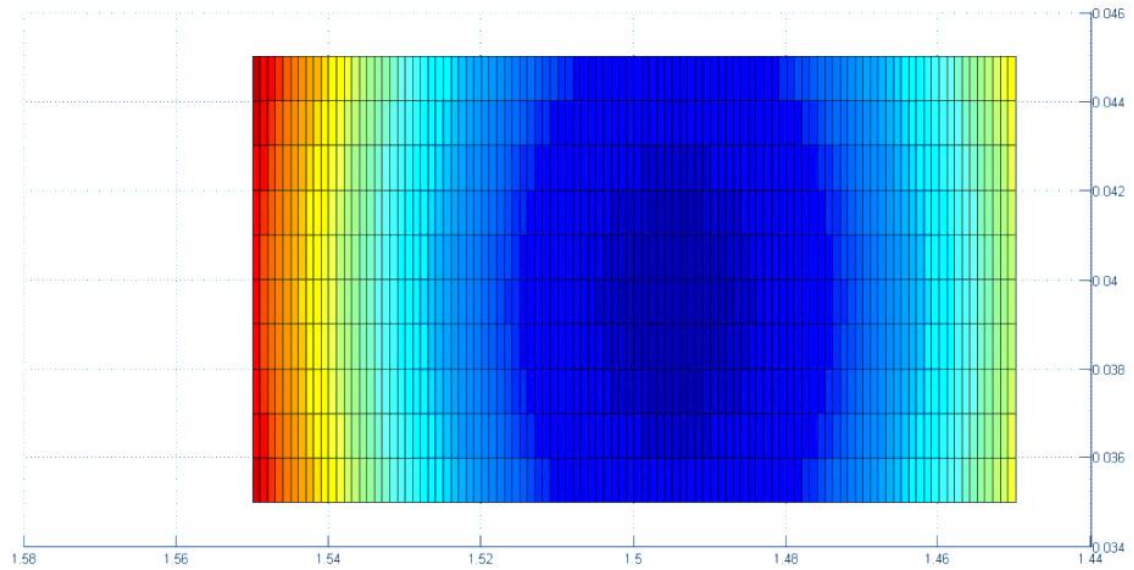


Fig. 5. Mean square error distribution in the 2D search region

3.2. Single-Crystal Silicon Solar Cell Module JT100-12M Unknown Parameter Extraction and Output Characterization

The output characteristics change at different temperatures and solar radiation intensities were simulated from the measured electrical and temperature change characteristics data under the standard test conditions of the commercial solar cell module JT100-12M. The electrical and temperature variation characteristics data given by the manufacturer are shown in Table 6.

To precisely simulate the changes of open circuit voltage and short circuit current, current-voltage curve and power-voltage curve with variation of temperature and solar radiation intensity, the above five unknown parameters must be extracted first. Since there are no current-voltage measurement points data of the proposed solar cell module, the optimization objective function of the unknown parameter extraction was taken as the mean square error at the short-circuit current point, open-circuit voltage point, and maximum power point.

Through the simulation, the series resistance of the module

is considered to be 0.06Ω , the parallel resistance is 8256.245Ω , and the idealization coefficient of the diode is considered to be 1.2010. Based on this, the current-voltage curve and power-vol

tage curve are obtained by varying the solar radiation intensity from 200 W/m^2 to 1200 W/m^2 at the reference temperature, as shown in Fig. 6.

Table 6. Electrical properties and temperature variation data (STC) of single-crystal silicon solar module JT100-12M

Electrical parameters	value
Open-circuit voltage (V_{OC})	22.44V
Short-circuit current (I_{SC})	5.69A
Optimum voltage (V_m)	18.05V
Optimum current (I_m)	5.54A
Number of eries connected cells (N_s)	36
Temperature variation coefficient of open circuit voltage (K_v)	-0.336%/K
Temperature variation coefficient of short circuit current (K_I)	0.042%/K

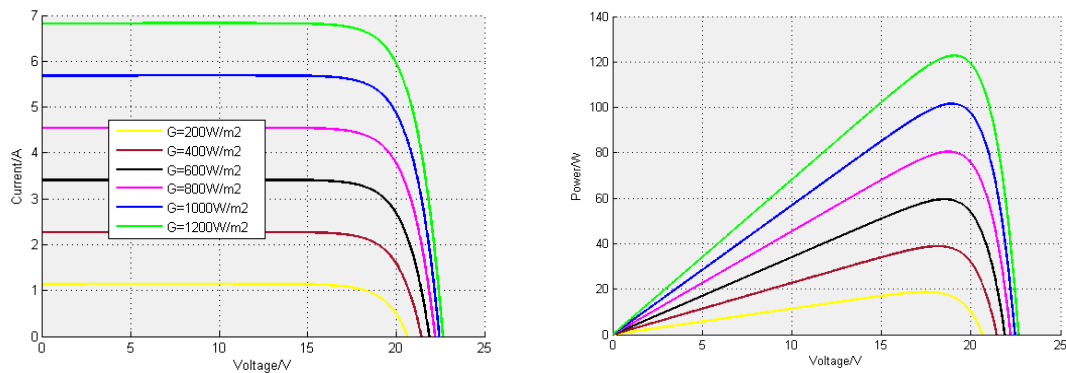


Fig. 6 Current-voltage curve and power-voltage curve of single crystal silicon solar cell model JT100-12M

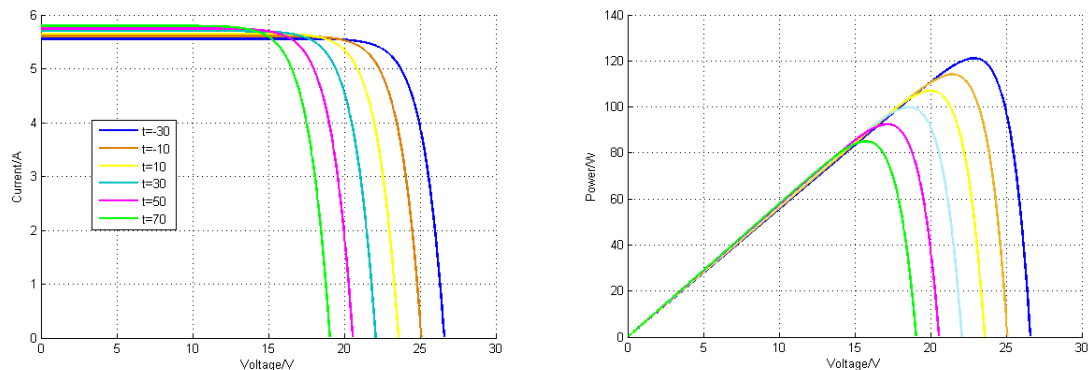


Fig 7. Variation of power curve with temperature variation

It can be seen that when the solar radiation intensity is 1000 W/m² and the temperature is changed from -30 to 50 °C, the current-voltage curve and the power-voltage curve change as shown in Fig. 7. As shown in the simulation results, the open circuit voltage of the solar cell increases with increasing solar radiation intensity, with lower temperature, and changes more sensitively with temperature. On the other hand, short-circuit current increases with increasing solar radiation intensity, with higher temperature, and changes sensitively to solar radiation intensity.

As above mentioned, we can extract the unknown parameters of the solar cell and module from the measured electrical characteristics of the solar cell or module under standard test conditions and obtain the current-voltage curve and power-voltage curve under any conditions.

4. Conclusions

In order to make the solar cell and module efficient in various fields such as power generation, temperature measurement and illumination measurement, it is essential to predetermine the unknown parameters such as parasitic resistances and diode quality factor. The number of unknown parameter extraction algorithms introduced to date has reached tens of times and the accuracy and speed of calculation are steadily increasing. The extraction algorithm presented in this paper can be said to have relatively high accuracy of the calculation results, but it has the disadvantage of long computation time. This is due to the fact that the dimensionality of the search space is reduced to two dimensions and the optimal point search is not efficient. However, in this study, we have confirmed the uniqueness of the optimal points in the search space. The goal of the next study is to design an algorithm that can lead to an optimal solution of a project starting at any point in the search space and to measure the computational accuracy and time.

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