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***Abstract*—An accurate model to represent the photovoltaic modules is essential to facilitate the efficient deployment of these systems in terms of design, analysis, and monitoring considerations. In this respect, this study proposes a new approach to improve the accuracy of the widely used five-parameter single-diode model. Two new physical equations are introduced to represent the series and shunt resistances, while the other parameters are represented by well established physical expressions. In the proposed model, most of the parameters are in terms of the cell temperature, irradiance, and datasheet values, while a few parameters need to be tuned. The model is compared with four well-known methodologies to extract the parameters of the single-diode and double-diode models. The simulation studies make use of the different *I–V* characteristics provided in the photovoltaics (PVs) datasheets, characteristics extracted from an outdoor module, as well as the ones simulated with the software PC1D. The results show an improved precision of the proposed model to estimate the power characteristics for a wide range of temperatures and irradiances, not only in the maximum power point but also in the whole range of voltages. Furthermore, the proposed physical model can be easily applied to other kind of studies where a physical meaning of the PV parameters is of great importance.**

***Index Terms*—Bandgap energy, photovoltaic (PV), physical modeling, series resistance, shunt resistance, single-diode model, translating equations.**

I. Introduction

THE photovoltaic (PV) systems are one of the most promising technologies to produce green electricity and contribute to slow down the climate change due to their reliability and the unlimited availability of sunlight [1]. Thus, appropriate models to represent the PV modules are required to design, monitor, control, and operation management during their life times [2]. In this regard, the single-diode (SD) and double-diode (DD) models, which are two of the most widely used models [3]–[7], aim to estimate the actual PV-cell/module behavior not only under standard test condition (STCs)[1] but also under any environmental condition.

Therefore, many strategies based on SD and DD models have been presented. Using analytical methods is a widely used approach that usually makes assumptions that decrease the model accuracy, e.g., the ideal SD model [4], [5]. Another technique is to deploy curve-fitting methodologies to match the measured maximum power point (MPP) with the value computed under STCs [8], [9]. Although accurate results to model the current–voltage (*I–V*) characteristics are reported, inaccuracies still exist in points different from the MPP. An improved double parameter curve fitting method is presented in [10] to extract the initial parameters, while physical/empirical approximations are used to adjust the parameters to any condition. High accuracy in the whole *I–V* curve for different conditions of irradiance and temperatures above room temperature (RT) is observed, while the parameters are bounded to values with a physical meaning. However, the shunt resistance is assumed to be constant, while other studies suggest to consider its dependency on the irradiance and temperature [11]–[13]. Other studies solve a system of nonlinear equations using the PV-module's datasheet values, while using physical/empirical approximations to adjust the parameters to any condition [14]–[16]. These techniques are accurate in the MPP, but the solution is highly dependent on the initial guesses and it is quite likely to converge to a local minimum or, even not converge. Likewise, a new technique is presented in [17] that uses datasheet information and an adaptive algorithm. Such a technique formulates the problem in the form of a constrained convex optimization problem with two decision variables. The results show a good enough accuracy above the RT, while convergence to a unique solution is ensured. Nevertheless, the shunt resistance is expressed by an empirical approximation and the series resistance is considered

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[1]Irradiation 1000 [W/m^2], solar spectrum $AM1.5$ G, temperature 25 [°C].

constant when its dependency on the irradiance and temperature is emphasized in [11], [13], and [18]. Other strategies are based on heuristic algorithms [19]–[21], which are very accurate at any condition of irradiance and temperature, but they are slow and require a large measurement dataset. Furthermore, artificial neural network (ANNs) are proposed in [3], [22], and [23] to directly approximate the *I–V* characteristics or for computing the PV parameters. This technique could get similar accuracy to analytical methods without the need for mathematical formulations, but it requires a large set of measurement data for a specific module and the optimal ANN design could be different for each module.

Even though some of the existing techniques can provide good accuracy, expressions of several model parameters lack physical meaning, which limits their reliable application under different operating conditions. Therefore, such methods fail to accurately estimate the PV power in a wide range of temperature and irradiance or predict the PV-cell/module behavior. Moreover, the empirical models should be reshaped considerably, if not totally, for their application in other kind of studies. On the other hand, a physical model can be easily applied to other studies, e.g., degradation of PV-cells due to bombarding of energetic particles and photon recycling. Additionally, physics-based models could be used for modeling degradation process, diagnosing faults, and preventive and corrective maintenance. Further, the physical parameters might help PV-cell designers to optimize the PV modules in terms of price, efficiency, and lifetime.

In this regard, several attempts have been made during the past years to introduce a physical meaning for the PV parameters. In [24], a good accuracy of power estimation is presented in four different PV-modules for the whole range of voltage variation $[0 - V_{oc}$ (open circuit voltage)] at different irradiance conditions and temperatures above RT. However, the expressions for the series and shunt resistances are still semiempirical, which might limit the physical representation of the PV cell/module. A similar study is introduced in [25], but considering a DD configuration. The results show a good accuracy in two different PV modules for the whole range of voltage at different irradiance conditions and temperatures above RT. However, this method has also the limitation of using semiempirical approximations for the resistances. In [26], it is proposed to estimate the PV-cell parameters taking advantage of the dark *I–V* curve's derivative using a triple diode model. The results show good fitting accuracy with the experimental dark *I–V* curve. However, the study is limited to test just one kind of cell under darkness and under a specific temperature.

Therefore, the need for a PV model with four main features, namely, high accuracy in a wide range of temperature and irradiance, low processing time, use of limited experimental data, and physical expressions of the PV-cell/module parameters still exists. In this respect, this article proposes a new modeling technique that introduces new physical approximations for the series and shunt resistances, which depend on the irradiance, cell temperature, and a few tuning parameters; considers the narrowing effect in the bandgap energy due to the heavy doping; and uses physical expressions for the ideality factor, reverse saturation current, and photogenerated current. The goal is to increase the accuracy of the SD model in a wide range of irradiance and temperature conditions; give a physical description to the series-shunt resistances, while using well-established physical expressions for the other parameters to reduce computational burden; and use the PV panel's datasheet information, which could ease the model deployment in practical applications, where measurement data may not be available.

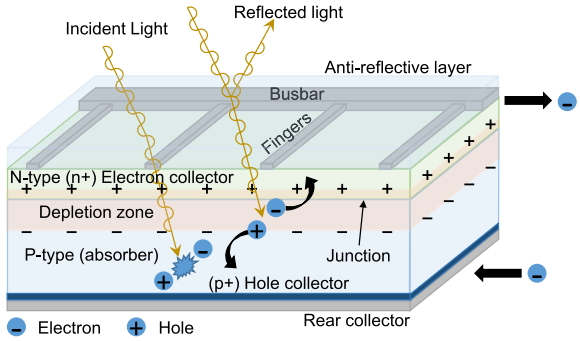

Fig. 1. Structure of a standard Si PV cell. Adapted from [27].

To verify the effectiveness of the proposed model, it is compared with four well-established techniques to extract the parameters of PV modules. For this purpose, datasheet/experimental data of five different PV modules is used. The results prove superiority of the proposed modeling approach in terms of accuracy over a wide range of temperature and irradiance throughout the whole range of voltage (0–V_{oc}).

The rest of the article is organized as follows. Section II presents the SD and DD models and some widely used techniques for the parameters extraction. Section III introduces the new approximations for the series-shunt resistances as well as the bandgap energy. Section IV gives a summary of the modeling procedure. Experimental results are discussed in Section V. Finally, Section VI concludes this article.

II. Photovoltaic-Cell Models

A PV cell based on mainstream silicon technology is fundamentally comprised of *n*-type and *p*-type semiconductor wafers, collectors, and antireflective coating, see Fig. 1. The SD and DD models, which are widely used to represent the PV system behavior are introduced in this section. Besides, four widely used techniques to compute the parameters of such models are presented, which will be used to compare the performance of the proposed models in Section V.

A. Single-Diode Model

This electrical circuit-based model is comprised of a current source to represent the photogenerated carriers; a series resistance to express losses caused by the load current; a shunt resistance to model the effect of the leakage current; and a diode to represent the diffusion and recombination [6], [9], [20], [28], see Fig. 2. Thus, assuming that all the cells of the PV module are similar, the module behavior is given as

$$I = I_{ph} - I_0 \Pi(a) - \frac{V + IR_s}{R_{sh}} \tag{1}$$

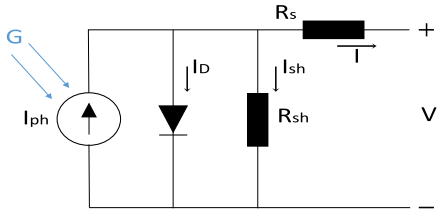

Fig. 2. SD equivalent circuit model for a PV module. V: Module voltage, I: Module current.

where $\Pi(a) = \exp(\frac{V+IR_s}{a}) - 1$. Besides, I_{ph} and I_0 give the equivalent photogenerated and reverse saturation current of the whole module, respectively. The equivalent series and shunt resistances for the whole module are given as R_s and R_{sh}, respectively. And the equivalent ideality factor as $a = mV_T = mn_skT/q$, where $k = 8.62 \times 10^{-5}\ [eV/K]$ is the Boltzmann constant, T is the cell temperature $[K]$, $q = 1.6021 \times 10^{-19}\ [C]$ is the elementary charge, m is the ideality factor, and n_s is the number of cells connected in series [5], [8], [14], [29].

1) W. D. Soto Solution: The widely used translating equations that are proposed by [14] are

$$a = a_{\text{stc}}\frac{T}{T_{stc}} \tag{2}$$

$$I_0 = I_0^{stc}\left(\frac{T}{T_{\text{stc}}}\right)^3 \exp\left[\frac{1}{k}\left(\frac{E_g}{T}\bigg|_{\text{stc}} - \frac{E_g}{T}\right)\right] \tag{3}$$

$$E_g = E_g^{\text{stc}}\left(1 - 0.0\,002\,677\Delta T\right) \tag{4}$$

$$I_{ph} = \frac{G}{G_{\text{stc}}}\frac{M}{M_{\text{stc}}}\left(I_{ph}^{\text{stc}} + \alpha_{\text{isc}}\Delta T\right) \tag{5}$$

where $\Delta T = T - T_{stc}$, E_g is the bandgap energy (1.12 $[eV]$ for Si-based cells at STC), G is the irradiance $[\frac{W}{m^2}]$, M is an air mass modifier, and α_{isc} is the short-circuit temperature coefficient. Further, the series and shunt resistances are expressed as $R_s = R_s^{\text{stc}}$ and $R_{sh} = R_{sh}^{\text{stc}}\frac{G_{\text{stc}}}{G}$, respectively. The reference parameters (at STCs) are found by solving a system of nonlinear equations.

2) M. G. Villalva Solution: This strategy is introduced in [8]. The technique consists in increasing the series resistance, while the shunt resistance is updated accordingly. The goal is to match the computed maximum power with the experimental value provided in the datasheet at STCs. The translating equations for this technique are

$$I_{ph} = \left(I_{ph}^{\text{stc}} + \alpha_{\text{isc}}\Delta T\right)\frac{G}{G_{\text{stc}}} \tag{6}$$

$$I_{ph}^{\text{stc}} = \frac{R_{sh} + R_s}{R_{sh}}I_{sc}^{\text{stc}} \tag{7}$$

$$I_0 = \frac{I_{sc}^{\text{stc}} + \alpha_{\text{isc}}\Delta T}{\exp\left[\left(V_{oc}^{\text{stc}} + \beta_{\text{voc}}\Delta T\right)/a\right] - 1} \tag{8}$$

where I_{sc} is the short-circuit current and β_{voc} is the temperature coefficient of the V_{oc}. In addition, R_{sh} is expressed in terms of R_s using (1) at STCs in the MPP, as follows:

$$R_{sh} = \frac{V_{mp}^{\text{stc}} + I_{mp}^{\text{stc}}R_s}{I_{ph}^{\text{stc}} - I_0^{\text{stc}}\exp\left[(V_{mp}^{\text{stc}} + I_{mp}^{\text{stc}}R_s)/a^{\text{stc}}\right] + I_0^{\text{stc}} - I_{mp}^{\text{stc}}} \tag{9}$$

where V_{mp} and I_{mp} correspond to the voltage and current at the MPP. After determining the resistances, they are kept constant. Recommended initial values for R_{sh} and R_s can be found in [8]. Furthermore, the ideality factor is arbitrarily chosen in the range $1.0 \leq m \leq 1.50$.

3) D. Sera Solution: Another solution is introduced in [15]. In this method, a system of three nonlinear equations is solved by a numerical solver. The result provides the value of the series resistance, shunt resistance, and ideality factor, which are considered constant at any ambient condition. In addition, a system of equations describe I_0, I_{ph}, I_{sc}, and V_{oc} at any condition of irradiance and temperature as

$$V_{oc}(T) = V_{oc}^{\text{stc}} + \beta\text{voc}\Delta T \tag{10}$$

$$I_{sc}(T) = I_{sc}^{\text{stc}} + \alpha_{\text{isc}}\Delta T \tag{11}$$

$$I_0(T) = \left[I_{sc}(T) - \frac{V_{oc}(T) - I_{sc}(T)R_s}{R_{sh}}\right]\exp\left[-\frac{V_{oc}(T)}{a^{\text{stc}}}\right] \tag{12}$$

$$I_{ph}(T) = I_0(T)\exp\left[\frac{V_{oc}(T)}{a^{\text{stc}}}\right] + \frac{V_{oc}(T)}{R_{sh}} \tag{13}$$

$$I_{sc}(G) = I_{sc}^{\text{stc}}\left(G/G_{\text{stc}}\right) \tag{14}$$

$$I_{ph}(G) = I_{ph}^{\text{stc}}\left(G/G_{\text{stc}}\right) \tag{15}$$

$$V_{oc}(G) = \ln\left|\frac{I_{ph}(G)R_{sh} - V_{oc}(G)}{I_0^{stc}R_{sh}}\right|a^{\text{stc}} \tag{16}$$

where (16) should be determined by a numerical solver. The method suggests to follow the superposition principle to consider simultaneously the effect of irradiance and temperature in any parameter.

B. Double-Diode Model

This model is used to improve the accuracy of the SD model [30], [31]. The second diode is located in parallel with the one in the SD model to represent the recombination in the depletion zone. The mathematical expression of this model for a module is given as

$$I = I_{ph} - I_{01}\Pi(a_1) - I_{02}\Pi(a_2) - \frac{V + IR_s}{R_{sh}}. \tag{17}$$

However, the improvement increases the complexity and the processing time [20].

1) Z. Salam Solution: This technique is introduced in [31] for a DD model. To reduce the computational burden to process the DD model parameters, an equal inverse saturation current in both diodes is assumed $I_{01} = I_{02}$. Besides, the ideality factor of D_1 and D_2 are set to $m_1 = 1$ and $m_2 \geq 1.2$, respectively. Thus, only four parameters are left and the translating equations are

$$I_{ph} = \frac{G}{G_{\text{stc}}}\left(I_{ph}^{\text{stc}} + \alpha_{isc}\Delta T\right) \tag{18}$$

$$I_{01} = I_{02} = \frac{I_{sc}^{\text{stc}} + \alpha_{\text{isc}}\Delta T}{\exp\left[\left(V_{oc}^{\text{stc}} + \beta_{\text{voc}}\Delta T\right)/V_T\right] - 1}. \quad (19)$$

R_{sh} is expressed in terms of R_s similarly to (9) but using (17) at STCs in the MPP. Recommended initial values for R_{sh} and R_s can be found in [8]. The goal is to match the computed MPP value with the value provided in the datasheet P_{mp}^{stc}.

III. New Approximation of the Bandgap Energy, Series, and Shunt Resistances

A. Bandgap Energy

It has been shown that the bandgap energy decreases if the temperature increases. Besides, a narrowing effect is observed, while the doping concentration of the impurities increases [32]–[35]. Thereby, the bandgap energy equation, along with a term ΔE_g to represent the narrowing effect, is given as

$$E_g = E_{g,0} - \frac{\alpha T^2}{T + \beta} - \Delta E_g \quad (20)$$

where $E_{g,0} \approx 1.169\ [eV]$ is the bandgap energy at zero K, $\alpha \approx 4.9 \times 10^{-4}\ [eV/K]$, and $\beta \approx 655\ [K]$ for Si [32]. Besides, for this study, ΔE_g is considered a tuning parameter taking into account that normally its value is in the order of [meV].

B. Series Resistance

The series resistance is comprised of two parts, one belongs to the conductors and the other one to the semiconductors.

1) Resistance in Conductors: Considering the range of temperatures on the earth surface, the series resistance of the conductive part is expressed as

$$R_{s1} \approx R_s^{\text{stc}}\left(1 + \alpha_0 \Delta T\right) \quad (21)$$

where α_0 is the collectors temperature coefficient at T_{stc} [36].

2) Resistance in the Semiconductive Part: The conductivity of a semiconductor is expressed as

$$\sigma = qn\mu_n + qp\mu_p \quad (22)$$

where n and p are the electron and hole densities, respectively. The corresponding mobilities are given as μ_n and μ_p [37].

a) *Carrier concentration*: In the extrinsic range, the carrier concentration will be comprised of the thermally generated and the photogenerated carriers. For the special case of an *n*-type semiconductor with a moderate-heavy doping ($N_d < 10^{18}\ [1/\text{cm}^3]$), the carriers concentration could be expressed as [37]

$$n = N_d + \Delta n \quad (23)$$

$$p \approx \Delta n \quad (24)$$

where N_d is the donors concentration and Δn is the mean excess carrier concentration due to the photogeneration using the AM1.5 G spectral irradiance and the wavelength from 280 to 1200 [nm]. Therefore, replacing *n* and *p* in (22) with (23) and (24), the conductivity will be expressed as

$$\sigma_n = q\mu_n N_d\left[1 + \frac{\Delta n}{N_d}\left(1 + F_\mu\right)\right] \quad (25)$$

where $F_\mu = \mu_p/\mu_n$. Likewise, the conductivity for the *p*-type semiconductor can be expressed as

$$\sigma_p = q\mu_p N_a\left[1 + \frac{\Delta n}{N_a}\left(1 + F_\mu^{-1}\right)\right] \quad (26)$$

where N_a is the acceptors concentration and Δn is defined as

$$\Delta n = \frac{G\tau_r}{G_{stc}hcW}\int_{AM1.5\ G} \eta_\lambda F_\lambda \lambda d\lambda \quad (27)$$

where $\eta_\lambda = \frac{hc}{q\lambda}SR(\lambda)$ is the external quantum efficiency (EQE), $SR(\lambda)\ [A/W]$ is the spectral response, $h = 6.626 \times 10^{-34}\ [Js]$ is the Planck constant, and $c\ [m/s]$ is the speed of light; $F_\lambda\ [Js^{-1}m^{-3}]$ is the spectral irradiance; $W\ [m]$ is the wafer thickness; $\lambda\ [m]$ is the photon wavelength; $\tau_r\ [s]$ is the mean surface and bulk recombination time [37]–[39]. In this article, two scenarios are analyzed: With Δn, and without Δn. Table I shows the assumptions made throughout the article.

TABLE I
Assumptions of the Proposed Models

Parameter	Assumption
Operation	The semiconductors comprising the PV-cell operates in the extrinsic stage.
Doping concentration	Non-degenerate semiconductor, $n \ll N_c$ and $p \ll N_v$. Besides, $N_a = N_d$.
Doping profile	Uniform along the PV-cell transverse section.
SRV	Not considered (not passivated).
τ_r	Constant.
Wafer thickness (W)	W= Constant $\forall$ layers and $W \gg$ Abs. Depth.
Injection level	Low.
Mobility ratio	$F_\mu = \mu_p/\mu_n = 1$.

SRV: Surface recombination velocity. τ_r: Mean recombination time.

b) *Drift mobility*: The total mobility μ_n of electrons is obtained by using Matthiessen's rule as follows:

$$\frac{1}{\mu_n} = \frac{1}{\mu_I} + \frac{1}{\mu_L} \quad (28)$$

where μ_I and μ_L show the mobility due to the ionized donor impurities and the lattice vibrations, respectively. Besides, the mobility is defined as $\mu \propto \tau = 1/(SvN_s)$, where τ is the mean free time between scattering events, S is the cross section area of the scatterer, v is the mean speed of the electrons in the conductive band (CB) (thermal velocity), and N_s is the number of scatterers per unit volume [37].

Drift mobility due to lattice vibrations, μ_L: The scatterer cross-section area depends on the the atomic vibrations amplitude around the equilibrium point, which means $S \propto (3/2)kT$ [37]. Besides, it is considered that the electrons transferred to the CB will have a kinetic energy (KE) within $(3/2)kT$ and $(3/2)kT + \Delta E$. Thereby, the average total velocity of the electrons in the CB due to the action of temperature and energy gained from the photon

is proposed to be

$$v \propto \left(\frac{3}{2}kT + \alpha_{ph}\frac{G}{N}\right)^{1/2} \tag{29}$$

where the first term represents the thermal KE and the second term represents the average KE gained from the absorbed photon. Besides, α_{ph} is the average ratio of the KE gained by an electron to the energy of the photon and N is the global photon flux. N_s is assumed to be constant. Accordingly, the drift mobility due to the lattice vibrations can be expressed as

$$\frac{1}{\mu_L} \propto T g_\alpha^{1/2} \tag{30}$$

$$g_\alpha = \frac{3}{2}kT + \alpha_{ph}\frac{G}{N}. \tag{31}$$

Drift mobility due to ionized impurities, μ_I: The scattering cross-section area in this mobility is related to the Coulombic attraction between the electrons in the CB and the ionized impurities. The scattering cross-section area is modified by considering that the KE in the electrons is comprised of the thermal excitation and the energy gained from the photon as

$$S \propto g_\alpha^{-2} \tag{32}$$

while the carrier velocity is kept as $v \propto g_\alpha^{1/2}$. Besides, the concentration of the impurities is considered to be constant. Thus, the drift mobility due to the ionized impurities is given as

$$\frac{1}{\mu_I} \propto g_\alpha^{-3/2}. \tag{33}$$

Total mobility: Finally, using (28), the total mobility of the electrons can be calculated as follows:

$$\frac{1}{\mu_n} = \gamma_{n1} g_\alpha^{-3/2} + \gamma_{n2} T g_\alpha^{1/2} \tag{34}$$

where γ_{n1} and γ_{n2} represent the proportional constants of the mobility due to the ionized impurities (33), and the mobility due to the lattice vibrations (30), respectively.

c) *Resistance*. Considering the assumptions given in Table I, the resistance definition $R \propto 1/\sigma$, the conductivity expression in (25), and the mobility (34), the resistance of an *n*-type semiconductor can be expressed as

$$R_n = \left(\gamma_{n1} g_\alpha^{-3/2} + \gamma_{n2} T g_\alpha^{1/2}\right) \Big/ \left(1 + 2\frac{\Delta n}{N_d}\right). \tag{35}$$

Defining $R_n|_{stc} = R_n^{\text{stc}}$, it is straightforward to drive the following equations from (35):

$$\frac{R_n}{R_n^{stc}} = \frac{\Gamma_{n1} g_\alpha^{-3/2} + \Gamma_{n2} T g_\alpha^{1/2}}{\Omega} \tag{36}$$

$$\Gamma_{n_2} = \frac{\Omega_{\text{stc}} - \Gamma_{n1} g_{\alpha,\text{stc}}^{-3/2}}{T^{\text{stc}} g_{\alpha,\text{stc}}^{1/2}} \tag{37}$$

$$\Omega = 1 + 2\frac{\Delta n}{N_d} \tag{38}$$

where $\Gamma_{n1} = \gamma_{n1}/R_n^{\text{stc}}$. Following the same procedure but using (26), the *p*-type resistance is expressed as

$$\frac{R_p}{R_p^{\text{stc}}} = \frac{\Gamma_{p1} g_\alpha^{-3/2} + \Gamma_{p2} T g_\alpha^{1/2}}{\Omega} \tag{39}$$

$$\Gamma_{p_2} = \frac{\Omega_{\text{stc}} - \Gamma_{p1} g_{\alpha,\text{stc}}^{-3/2}}{T^{\text{stc}} g_{\alpha,\text{stc}}^{1/2}}. \tag{40}$$

Considering $R_{s2} = R_p + R_n$, $R_p = aR_{s2}$, $R_n = bR_{s2}$, and $a + b = 1$, it is possible to derive an expression for R_{s2} as

$$\frac{R_{s2}}{R_s^{\text{stc}}} = \left(\frac{\Omega_{\text{stc}}}{\Omega}\right)\left[A\phi^{-3/2} + (1-A)\left(\frac{T}{T^{\text{stc}}}\right)\phi^{1/2}\right] \tag{41}$$

where $\phi = g_\alpha/g_{\alpha,\text{stc}}$ and *A* is a tuning parameter.

3) Total Series Resistance: The total series resistance can be expressed as follows using the conductive part represented in (21) and the semiconductive part given by (41)

$$R_s = \Gamma_R R_{s1} + (1 - \Gamma_R) R_{s2} \tag{42}$$

where Γ_R is a tuning parameter, which corresponds to the fraction of the total series resistance in the conductive part.

C. Shunt Resistance

For the PV modeling purpose, R_{sh} is used to model the leakage current of the PV-modules, which flows across the crystal surface or through the grain boundaries for polycrystalline (PC) technology instead of along the load [37]. R_{sh} belongs to the semiconductive part of the PV-module, thus the same approach proposed to drive (41) is adopted as

$$\frac{R_{sh}}{R_{sh}^{\text{stc}}} = \left(\frac{\Omega_{\text{stc}}}{\Omega}\right)\left[B\phi^{-3/2} + (1-B)\left(\frac{T}{T^{\text{stc}}}\right)\phi^{1/2}\right] \tag{43}$$

where *B* is the tuning parameter of the shunt resistance.

IV. Proposed Single-Diode Model

In addition to the expressions for R_s, R_{sh}, and E_g, the translating equations for a, I_0, and I_{ph} are modeled using the proposed approach in [14] when air mass $M = M_{\text{stc}}$. These expressions are given by (2), (3), and (5). If the temperature is measured at the back surface of the module, a transformation to the actual cell temperature is also needed [40], [41]. In this article, the following transformation is used:

$$T = T_m + \frac{G}{G_{\text{stc}}}\Delta T \tag{44}$$

where T_m is the rear surface module temperature and ΔT is typically around $2 \sim 3$ [°C] for flat plate modules [42]. However, in this study a better performance is observed for $\Delta T = 4$ [°C]. Therefore, the translating equations of the proposed model are comprised of (2), (3), (5), (42), and (43) along with (20) and (44) for the bandgap energy and cell temperature, respectively. Table II shows the tuning parameters.

Similar to the other existing modeling approaches for PV modules [14], [29], [43], the proposed translating equations in

TABLE II
TUNING PARAMETERS OF THE PROPOSED MODEL

Parameter	Description	Range
A	Constant for R_s	[0,1.5]
B	Constant for R_{sh}	[0,1.5]
Γ_R	Portion of R_s (conductor part)	[0,1.0]
ΔE_g	Narrowing effect of E_g	$\leq 90\ meV$ (poly) $\geq 90\ meV$ (mono)
α_{ph}	Ratio of the electron KE in the CB to the photon energy	(0,1.0)

The range for A, B, and ΔE_g are suggested based on the experience.

TABLE III
PV-MODULES SPECIFICATIONS

	KK280P	**JAP60S01**	**REC245PE**	**PC1D**	**M-60**
Datasheet parameters					
I_{sc}^{stc}	9.53	9.18	8.80	7.9755	9.08
V_{oc}^{stc}	38.9	38.17	37.10	39.324	37.90
n_s	60	60	60	60	60
V_{mp}^{stc}	31.50	31.13	30.10	34.4085	30.80
I_{mp}^{stc}	8.89	8.67	8.23	7.6232	8.60
α_{isc}	0.00559	0.00532	0.002112	0.000251	0.003632
β_{voc}	-0.138	-0.12596	-0.10017	-0.12416	-0.12128
PV parameters under STCs (computed by the TLBO algorithm)					
I_{ph}^{stc}	9.52731	9.18789	8.69404	7.97837	9.07735
I_0^{stc}	1.45198	0.17515	61.7734	0.06194	19.32055
R_s^{stc}	0.27686	0.21724	0.26557	0.01382	0.15082
R_{sh}^{stc}	190.645	501.121	2094.48	5000	594.029
a^{stc}	1.72293	1.54662	1.91299	1.53731	1.89903

Note: α_{isc} is in $[A/°\text{C}]$, β_{voc} in $[V/°\text{C}]$, and I_0^{stc} in $[nA]$.

this article are a function of the corresponding parameter under the STC. To determine these parameters, this article uses the teaching-learning-based optimization (TLBO) algorithm.

V. RESULTS AND DISCUSSION

The proposed modeling methodology is validated for the PC and monocrystalline (MC) PV technologies since they are widely used nowadays. Two PC and one MC modules are studied using the *I–V* characteristics provided in their datasheets. A PC module located on the roof of the PV Lab at Aalborg University is also studied using experimental *I–V* curves. Another module is modeled with the software *PC1D*. Table III shows the information for each module. The parameters' values at STCs were computed by the TLBO algorithm and are used by the proposed model in this study and by the W. D. Soto solution [14].

The proposed model is considered without the excess of carriers due to the photogeneration effect (PGE) (*Prop. 1*) and with considering the excess of carriers (*Prop. 2*) as represented in (27). For comparison purposes, four widely cused techniques for PV modeling, which were introduced in Section II, have been implemented in addition to the proposed method. The comparison is performed by using the mean absolute error in power (MAEP) introduced by [10] and [24]

$$\text{MAEP}_i = \frac{\sum |P_{\text{mes},k} - P_{\text{mod},k}|}{N_p}, \qquad k = 1, \ldots, N_p \tag{45}$$

where N_p is the number of points on the power curve, P_{mes} is the PV power at a specific voltage from datasheet, and P_{mod} is the power computed by the respective model. The evaluation of the MAEP is performed from 0 to V_{oc} and around the MPP.

TABLE IV
TUNING PARAMETERS USED FOR EACH MODEL AT EACH PV MODULE

PV-models proposed (SD)					
	Value used for each module				
Param.	KK280P	JAP60S01	REC245PE	PC1D	M-60
A	1.004240	0.18	0.0064	-	-
B	0.516975	0.55	0.32	1.50	0.121
Γ_R	0.954569	0.15	0.30	1.0	1.0
ΔE_g	80.645	32.25	70	6.5	223.75
α_{ph}	0.25	0.099	0.1485	0.03542	0.3264
Salam solution (DD) [31]					
m_2	1.53	1.20	1.20	1.20	1.60
Villalva solution (SD) [8]					
m	1.10	1.0	1.35	1.025	1.25

Note: m_2 is the ideality factor of diode 2. ΔE_g is in [meV].

Besides, the average MAEP is used as follows:

$$\text{MAEP}_{av} = \frac{\sum \text{MAEP}_i}{N_{\text{curves}}}, \qquad i = 1, \ldots, N_{\text{curves}} \tag{46}$$

where N_{curves} is the total number of curves used. Further, the root mean square error (RMSE) of the power from 0 V to V_{oc} and around the MPP is also analyzed. Table IV shows the values for the tuning parameters used in this study.

Furthermore, *Prop. 2* requires additional data: The global spectral irradiance at AM1.5 G given by the ASTM [44] to determine the total photon flux as $N \approx 2.90 \times 10^{21}\ [\#/sm^2]$, the irradiance at the specific wavelength $G_{0,\lambda}\ [Js^{-1}m^2]$ between 280 and 1200 [nm]; The spectral response of the MC and PC PV cells to compute the EQE [45]; cell thickness $W \approx 200\ [\mu\text{m}]$; And the mean recombination time, $\tau_r = 5 \times 10^{-5}\ [s]$. For all the modules, an average doping concentration of $N_d = 10^{16}\ [\text{cm}^{-3}]$ is considered. Also, it is assumed that the fingers and busbars are made of Silver, $\alpha_0 \approx 3.72 \times 10^{-3}\ [K^{-1}]$ [36].

A. MAEP and RMSE of Power

First, the PV power characteristics of all modules are derived. Afterward, the MAEP and RMSE of power are computed in the whole range of voltage from 0 V to V_{oc} and in a few sampling points around the MPP.

Fig. 3 represents the MAEP of all the models for the different modules at different conditions. The accuracy enhancement can be observed in Fig. 3(5), 3(6), and 3(10) when (27) is included. Table V presents the average MAEP of each model obtained for each of the modules at different levels of irradiance for the whole range of voltages and around the MPP. The results show the superiority of the proposed models *Prop. 1* and *Prop. 2* over the other models in the most of the conditions for all the modules in terms of the minimum average MAEP. Likewise, the average MAEP for the whole range of voltages and around the MPP at different levels of temperature is given in Table VI. It can be noticed that *Prop. 1* outperforms other modeling approaches in reducing the average MAEP. In addition, a considerable difference between *Prop. 1* and *Prop. 2* is not observed.

Similarly, Table VII shows the average RMSE of power obtained from each model for each of the modules in different irradiances. Likewise, Table VIII shows the average RMSE of power obtained from each model for each of the modules in different temperatures. The averages, in both tables, cover

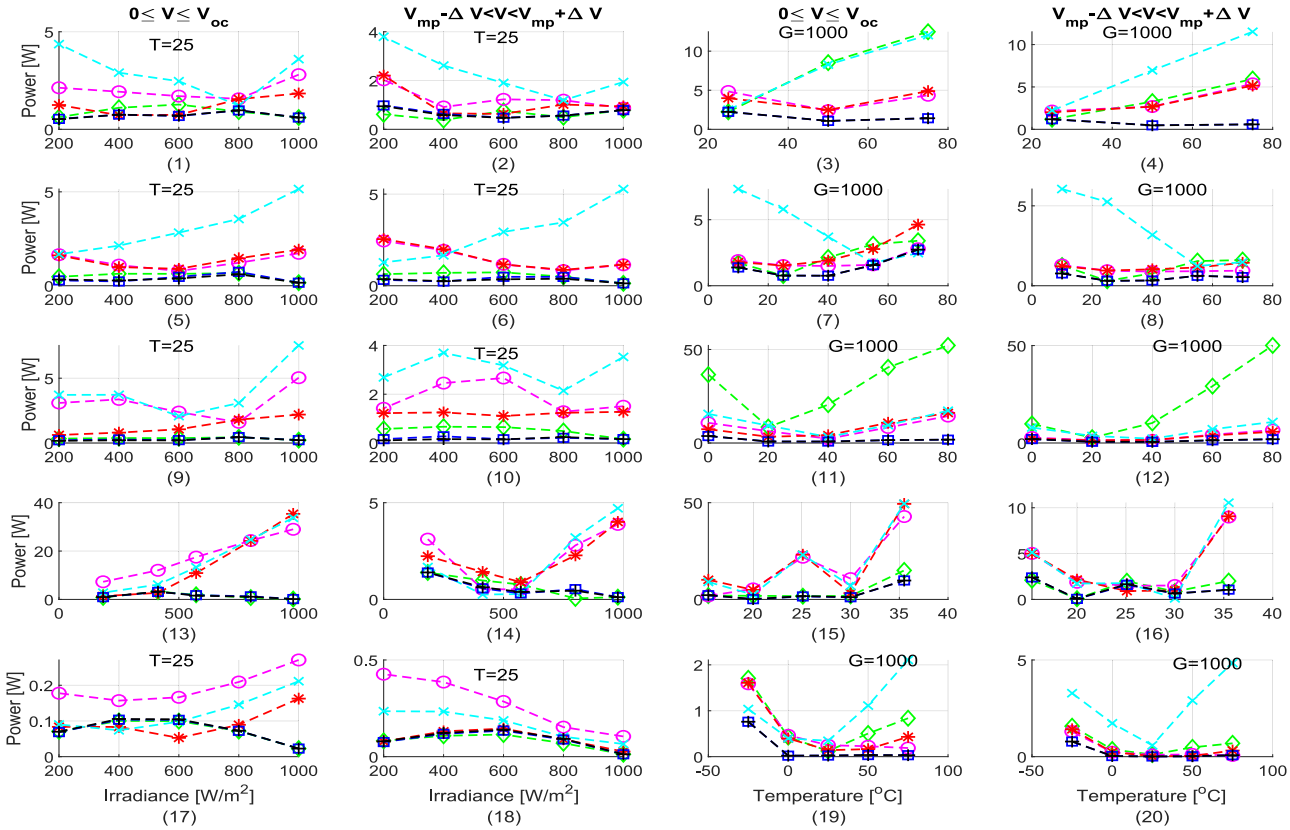

Fig. 3. Diamond: D. Soto solution (SD); Circle: Salam solution (DD); Asterisk: Villalva solution (SD); Cross: Sera solution (SD); Square: Prop. 1 (SD); Plus: Prop. 2 (SD). From left to right. Column 1: MAEP for $0 \leq V \leq V_{oc}$ for different levels of irradiance. Column 2: MAEP around MPP for different levels of irradiance. Column 3: MAEP for $0 \leq V \leq V_{oc}$ for different levels of temperature. Column 4: MAEP around MPP for different levels of temperature. From top to bottom. Row 1: Module KK280P3CD3CG. Row 2: Module JAP60S01270SC. Row 3: Module M60. Row 4: Module REC245PE. Row 5: Module PC1D. The points in (13)–(14), for each level of irradiance, correspond to rear surface temperatures of $\{25.9, 25.3, 24.1, 25.1, 25.3\}$ [°C], from left to right, respectively. The points in (15)–(16) for each level of temperature correspond to irradiances of $\{205\,255, 842\,219, 1007\}$ [W/m^2], from left to right, respectively.

TABLE V
$MAEP_{av}$ FOR $0 \leq V \leq V_{oc}$ AND AROUND THE MPP. AVERAGE IS COMPUTED FOR SEVERAL IRRADIANCES

Module	$0 \leq V \leq V_{oc}$						$V_{mp} - \Delta V < V < V_{mp} + \Delta V$					
	A	B	C	D	Prop.1	Prop.2	A	B	C	D	Prop.1	Prop.2
KK280P	0.9018	2.0244	1.2194	2.9094	0.7090	**0.7086**	**0.6062**	1.2475	1.0887	2.2887	0.6870	0.6697
JAP60S01	0.5161	1.2969	1.3790	3.0688	0.3767	**0.3537**	0.5377	1.5028	1.5417	2.9210	0.3301	**0.2937**
REC245PE	**1.3757**	18.0345	14.9079	16.1447	1.4438	1.4019	0.6619	2.1668	2.1661	2.0229	**0.5833**	0.5948
PC1D	**0.0730**	0.1962	0.0943	0.1238	0.0741	0.0751	**0.0773**	0.2710	0.0941	0.1649	0.0859	0.0892
M-60	0.3509	3.0948	1.2941	4.0221	0.2741	**0.2497**	0.5091	1.8660	1.2206	3.0480	0.1913	**0.1626**

A: D. Soto solution [14]; B: Salam solution [31]; C: Villalva solution [8]; D: Sera solution [15]. Note: The numbers in bold represent the minimum MAEPav obtained from different models for each PV module.

TABLE VI
$MAEP_{av}$ FOR $0 \leq V \leq V_{oc}$ AND AROUND THE MPP. AVERAGE IS COMPUTED FOR SEVERAL TEMPERATURES

Module	$0 \leq V \leq V_{oc}$						$V_{mp} - \Delta V < V < V_{mp} + \Delta V$					
	A	B	C	D	Prop.1	Prop.2	A	B	C	D	Prop.1	Prop.2
KK280P	7.7474	3.8610	3.7828	7.6119	**1.5745**	**1.5745**	3.4641	3.4009	3.2835	6.8969	**0.7464**	**0.7464**
JAP60S01	2.2357	1.8638	2.5242	4.2067	**1.4417**	**1.4417**	1.0997	0.9957	1.1559	3.4388	**0.5130**	**0.5130**
REC245PE	4.4608	16.5264	18.1640	18.2104	**2.9509**	2.9967	1.4043	3.7778	3.6210	3.8585	**1.1483**	1.1652
PC1D	0.7196	0.5435	0.5482	1.0006	**0.1748**	**0.1748**	0.6456	0.3602	0.4063	2.6676	**0.1846**	**0.1846**
M-60	31.6314	8.4136	8.3520	10.9760	**1.7181**	1.7187	20.4090	3.2953	2.9785	6.3183	**1.3026**	1.3038

A: D. Soto solution [14]; B: Salam solution [31]; C: Villalva solution [8]; D: Sera solution [15]. Note: The numbers in bold represent the minimum MAEPav obtained from different models for each PV module.

TABLE VII
AVERAGE RMSE OF POWER FOR $0 \leq V \leq V_{oc}$ AND AROUND THE MPP. AVERAGE IS COMPUTED FOR SEVERAL IRRADIANCES

Module	$\frac{1}{N}\sum_{\forall N} RMSE_N$ $0 \leq V \leq V_{oc}$						$\frac{1}{N}\sum_{\forall N} RMSE_N$ $V_{mp}-\Delta V < V < V_{mp}+\Delta V$					
	A	B	C	D	Prop.1	Prop.2	A	B	C	D	Prop.1	Prop.2
KK280P	1.3137	2.8525	1.5932	3.9898	**1.0095**	1.0111	**0.7261**	1.4178	1.2936	2.6797	0.7829	0.7621
JAP60S01	0.7141	1.6432	1.7735	4.6595	0.5725	**0.5338**	0.5597	1.5985	1.6444	3.7217	0.3637	**0.3293**
REC245PE	1.6662	20.9017	17.4257	18.8156	1.6931	**1.6378**	0.6686	2.2575	2.2036	2.0675	**0.5877**	0.5988
PC1D	**0.2240**	0.3041	0.2313	0.2674	0.2269	0.2280	**0.0808**	0.2790	0.0988	0.1692	0.0890	0.0922
M-60	0.4484	4.9867	1.6724	6.0045	0.4360	**0.3891**	0.5389	2.3731	1.3554	3.4790	0.2402	**0.1983**

A: D. Soto solution [14]; B: Salam solution [31]; C: Villalva solution [8]; D: Sera solution [15]. Note: The numbers in bold represent the minimum average RMSE obtained from different models for each PV module.

TABLE VIII
AVERAGE RMSE OF POWER FOR $0 \leq V \leq V_{oc}$ AND AROUND THE MPP. AVERAGE IS COMPUTED FOR SEVERAL TEMPERATURES

Module	$\frac{1}{N}\sum_{\forall N} RMSE_N$ $0 \leq V \leq V_{oc}$						$\frac{1}{N}\sum_{\forall N} RMSE_N$ $V_{mp}-\Delta V < V < V_{mp}+\Delta V$					
	A	B	C	D	Prop.1	Prop.2	A	B	C	D	Prop.1	Prop.2
KK280P	12.3243	4.9558	4.7456	9.8631	**2.0624**	**2.0624**	4.0382	3.5711	3.5024	7.4920	**0.9169**	**0.9169**
JAP60S01	3.2788	**2.5414**	3.5234	6.0553	2.7530	2.7530	1.3150	1.1779	1.4171	4.2604	**0.5914**	**0.5914**
REC245PE	5.3401	19.2347	21.1252	20.9988	**3.4061**	3.4472	1.4154	3.8430	3.6723	3.9007	**1.1597**	1.1768
PC1D	2.5819	1.7673	1.8260	2.7153	**0.5664**	**0.5664**	0.6689	0.4025	0.4243	2.7067	**0.1916**	**0.1916**
M-60	47.7531	13.2887	12.8402	16.0188	**2.5192**	2.5198	23.7117	3.9325	3.3512	7.4103	**1.4958**	1.4969

A: D. Soto solution [14]; B: Salam solution [31]; C: Villalva solution [8]; D: Sera solution [15]. Note: The numbers in bold represent the minimum average RMSE obtained from different models for each PV module.

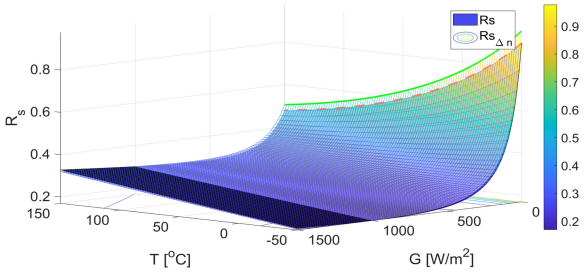

Fig. 4. PV-module JAP60S01. R_s : Series resistance for model *Prop. 1*. $R_{s,\Delta n}$: Series resistance for model *Prop. 2*. Solid green-line: $R_{s,\Delta n}$ at $G = 0$. Dashed red-line: R_s at $G = 0$.

the whole range of voltages and around the MPP. The results once again show the superiority of the proposed models under different irradiance and temperature conditions.

B. Series and Shunt Resistances

The mathematical representation of the resistances is derived by considering an *n-p* junction. Therefore, the shunt resistance characteristics of a PV module should be similar to the one of a semiconductor in the dark ($G = 0$), while the series resistance characteristics should be similar to the one of a linear combination between a semiconductor and a metal. Fig. 4 shows R_s of the PV-module JAP60S01-270-SC for both models: *Prop. 1* and *Prop. 2*. According to Fig. 4, R_s at $G = 0$ decreases when temperature increases. Besides, the effect of Γ_R is clearly observed for large irradiances with a linear increment, while the temperature increases. The shunt resistances have a similar behavior without the linear component.

When the photons hit the atoms inside the semiconductor lattice with enough energy ($G/N \geq E_g$), the generation of electron–hole pairs starts and the carriers concentration density increases. The electrons "jump" from the valence band (VB) to the CB with an average velocity described by (29). Thereby, the increment of the carriers concentration and velocity will result in reducing the resistances of the semiconductors for a given temperature. The reduction is sharper for low temperatures, see Fig. 4. However, the contribution of the metallic parts will increase the resistance, while the temperature increases. The resistances will reach a minimum value that will be in lower irradiances for higher temperature levels. The minimum values can be found by using (42) and (43). The shadowed area in Fig. 4 is the zone where $R_s \geq R_{s,\Delta n}$.

According to Fig. 3 and Tables V–VIII, the inclusion of the approximations proposed in this article, has resulted in improving the modeling accuracy throughout the whole range of the module voltage in a wide range of T and G.

VI. CONCLUSION

In this article, new approximations for the series and shunt resistances of PV modules are proposed. Besides, an expression for the bandgap energy considering the narrowing effect for heavy doping was used along with physical expressions for the remaining parameters. The proposed approximations were applied to the SD model. Furthermore, it was analysed with and without considering of the excess of carriers due to the PGE. The performance of the proposed models were compared with four well-known models using the *I–V* characteristics of different PC and MC PV modules. The results show the superiority of the proposed models in terms of accuracy in almost all the conditions of temperature and irradiance for the five modules analyzed. Although the proposed models contain a few tuning parameters, they vary within very narrow ranges, which allow to tune them, even manually. However, to reach a good approximation for the tuning parameters, additional *I–V* curves are required, which can be obtained either from the PV datasheets or field measurements. In case the lower number of tuning parameters is desired, other information like doping concentration, doping distribution profile, and so on are required which are not normally provided by PV manufacturers. Thus, a satisfactory tradeoff between accuracy and complexity is needed, considering the model applications. Finally, considering the proposed physics-based modeling approach for representing the series and shunt resistances, the

model has the potential to be used in extreme operating conditions through accurately considering the effect of temperature and irradiance on the PV resistances behavior. Besides, it can be used for degradation tracking, performance monitoring, PV design improvement, and other applications, where the physical meaning of the PV parameters is of great importance.

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