

Overview of Photovoltaic Module Performance Modeling Approaches

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PV Module Modeling Approaches

■ Equivalent Circuit Models

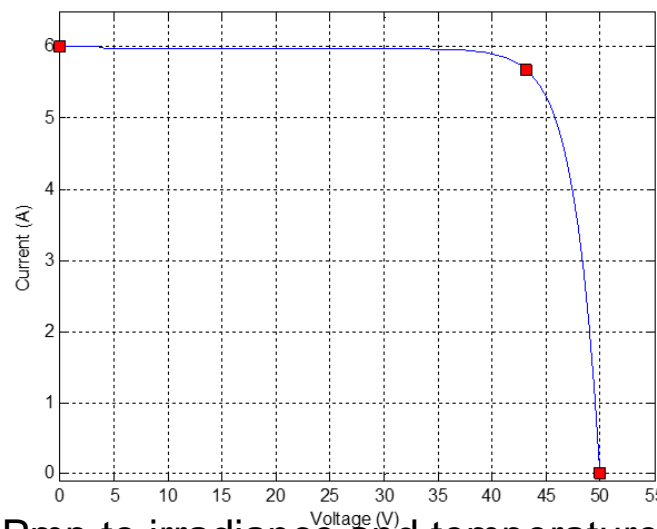
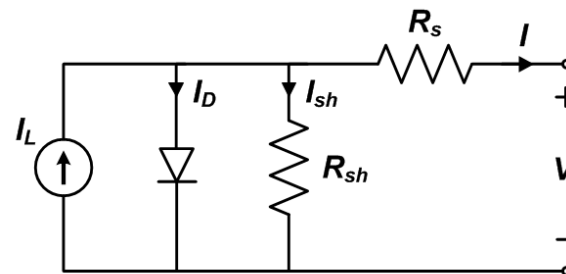
- Describe full I-V curve at desired irradiance and temperature conditions
- Single diode equation ('5 parameter') is most common
- Auxiliary equations relate the 5 parameters to irradiance and temperature
 - Different among PVsyst, De Soto, CEC, PV*SOL

■ Point Models

- Describe cardinal points on the IV curve: P_{mp} , I_{mp} , V_{mp} , I_{sc} , V_{oc}
- PVWatts
- Huld model
- Sandia PV Array Performance Model (SAPM)
- Loss Factors Model

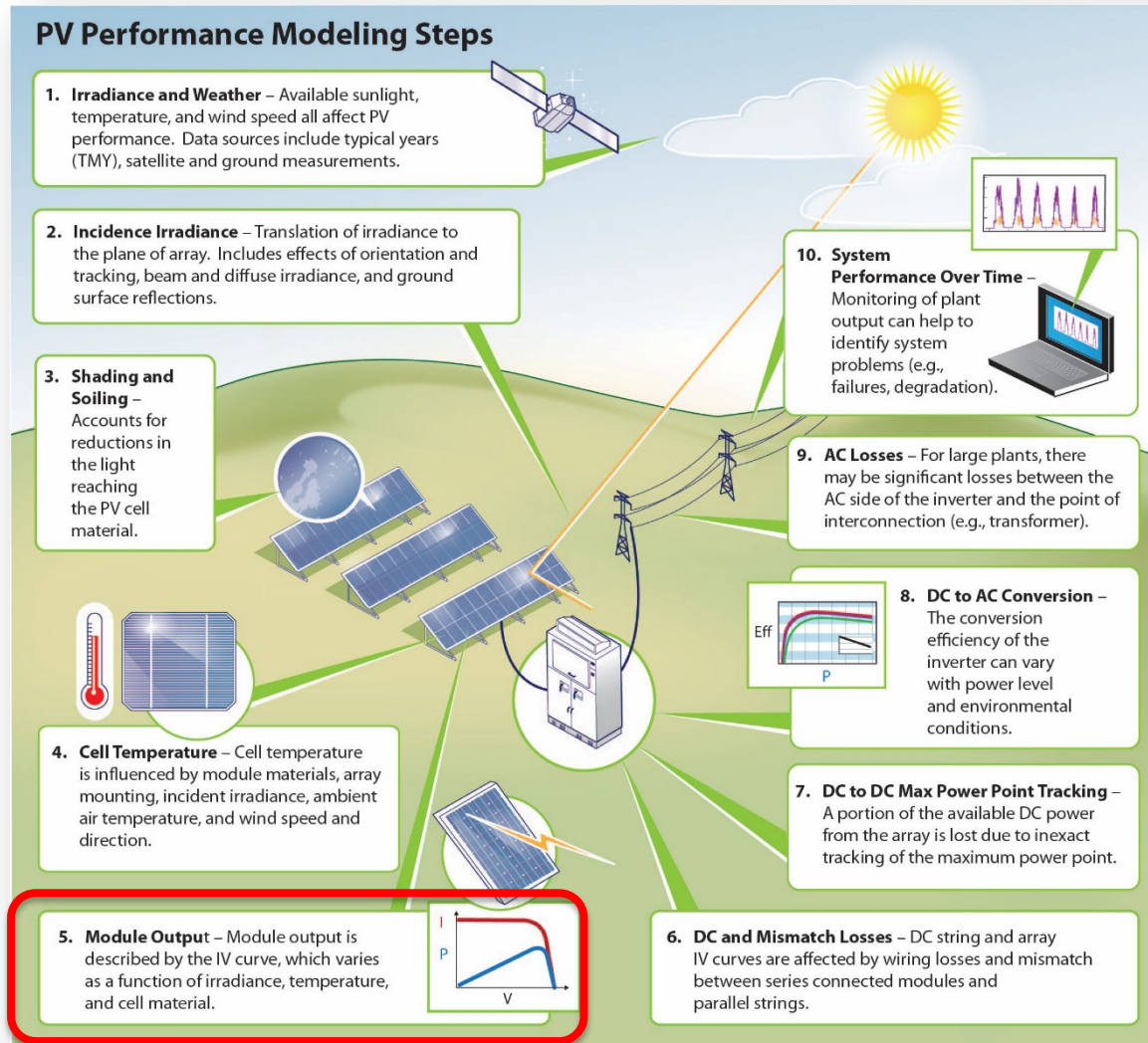
} Algebraic equations relating P_{mp} to irradiance and temperature

■ 'Data' Models



} Set of equations describing I_{mp} , V_{mp} , I_{sc} and V_{oc}

Module model is only one part of a modeling process



Stein, J. S. and B. H. King (2013). Modeling for PV plant optimization. *Photovoltaics International, Solar Media Ltd.* 19th: 101-109.

“Single Diode” Models

- CEC, PVsyst, CEC, PV*SOL, others
 - IV curve described by single diode equation
 - “5 parameters” – for each IV curve
- Auxiliary equations describe how ‘5 parameters’ change with irradiance, temperature
 - Different equations for each PV model
 - Auxiliary equations contain the model parameters
 - E.g., De Soto model has 7 parameters, PVsyst v6 has 9 parameters

$$I = I_L - I_0 \left[\exp\left(\frac{V + IR_s}{nV_T}\right) - 1 \right] - \frac{V + IR_s}{R_{sh}}$$

$$I_L(E, T_C) = \frac{E}{E_0} [I_{L0} + \alpha_{Isc} (T_C - T_0)]$$

$$I_0 = I_{00} \left(\frac{T_C}{T_0}\right)^3 \exp\left(\frac{1}{k} \left(\frac{E_{g0}}{T_0} - \frac{E_g(T_C)}{T_C}\right)\right)$$

$$R_{sh} = R_{sh0} \frac{E_0}{E} \quad R_s, n \text{ constant}$$

De Soto et al, 2006

Ancillary Equations

Diode equation term (unit) symbol	PVsyst v6	CEC '6 parameter' model
light current (A) I_L	$I_L(E, T_C) = \frac{E}{E_0} [I_{L0} + \alpha_{isc} (T_C - T_0)]$	$I_L(E, T_C) = \frac{E}{E_0} [I_{L0} + \alpha'_{isc} (T_C - T_0)]$ $\alpha'_{isc} = \alpha_{isc} (1 - Adjust/100)$
dark current (A) I_o	$I_o(T_C) = I_{o0} \left[\frac{T_C}{T_0} \right]^3 \exp \left[\frac{q\epsilon_G}{k\gamma} \left(\frac{1}{T_0} - \frac{1}{T_C} \right) \right]$	$I_o(T_C) = I_{o0} \left[\frac{T_C}{T_0} \right]^3 \exp \left[\frac{1}{k} \left(\frac{E_g(T_0)}{T_0} - \frac{E_g(T_C)}{T_C} \right) \right]$ $E_g(T_C) = E_{g0} (1 - 0.0002677(T_C - T_0))$
series resistance (Ω) R_S	$R_S = R_{S0}$	$R_S = R_{S0}$
shunt resistance (Ω) R_{SH}	$R_{SH} = R_{SH,base} + (R_{SH,0} - R_{SH,base}) \exp \left(-R_{SH,exp} \frac{E}{E_0} \right)$ $R_{SH,base} = \max \left[\frac{R_{SH,ref} - R_{SH,0} \exp(-R_{shexp})}{1 - \exp(-R_{shexp})}, 0 \right]$	$R_{SH}(E) = R_{SH0} \frac{E_0}{E}$
ideality factor (unitless) γ or n	$\gamma = \gamma_0 + \mu_\gamma (T_C - T_0)$	$n = n_0$

Open issues with diode models

- Simulated IV curves often don't match measurements at low irradiance
- Methods to estimate model parameters are often not carefully scrutinized
 - Many methods return non-physical values, e.g., $n < 1$
 - As a consequence, the same data lead to different parameter sets
 - Could have reference cases to test estimation methods (a common practice in other disciplines)
- Can we validate the auxiliary equations?
 - Light, dark current forms derive from device physics
 - Others are eyeballed from data (e.g., R_{SH} in CEC model)
 - May involve corrections to match IV curves (e.g., γ in PVsyst v6)
 - Equations for R_S , n or γ are most promising targets

Single Point Models

- PVWatts :
$$P_{dc} = \frac{I_{tr}}{1000} P_{dc0} (1 + \gamma(T_{cell} - T_{ref}))$$
- Huld Model :
$$P(G', T') = G' (P_{STC, m} + k_1 \ln(G') + k_2 \ln(G')^2 + k_3 T' + k_4 T' \ln(G') + k_5 T' \ln(G')^2 + k_6 T'^2)$$
- PVUSA :
$$P = I \cdot (A + B \cdot I + C \cdot T_a + D \cdot WS)$$
 - Because P is P_{AC} , PVUSA combines cell temperature and inverter efficiency model with module performance model
- PVWatts is (almost) module-agnostic: the only term which derives from measurement is the temperature coefficient γ
- PVUSA is not a module model per se; it's a whole system model
- Huld model is intended to be fit to common measurements: back-of-module temperature and broadband plane-of-array

The Sandia Array Performance Model

- Describes module output at SC, OC and MP points
- As a function of beam and diffuse irradiance (E_b and E_{diff}), cell temperature (T_C), air mass (AM_a) and angle of incidence (AOI)
- 14 empirical coefficients, 2 empirical functions (f_1 and f_2)
- With exception of f_2 , coefficients must be determined for individual modules

$$V_{OC} = V_{OC0} + N_s n \delta(T_C) \ln(E_e) + \beta_{OC} (T_C - T_0)$$

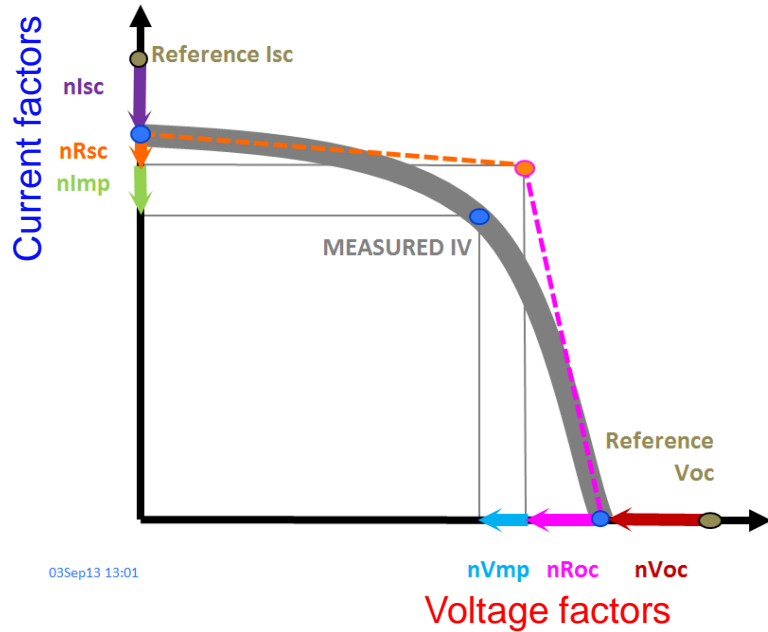
$$V_{MP} = V_{MP0} + C_2 N_s n \delta(T_C) \ln(E_e) + C_3 N_s (n \delta(T_C) \ln(E_e))^2 + \beta_{MP} (T_C - T_0)$$

$$I_{SC} = I_{SC0} f_1(AM_a) E_e (1 + \alpha_{SC} (T_C - T_0))$$

$$I_{MP} = I_{MP0} (C_0 E_e + C_1 E_e^2) (1 + \alpha_{MP} (T_C - T_0))$$

$$E_e = E_b f_2(AOI) + E_{diff} f_d$$

Loss Factors Model



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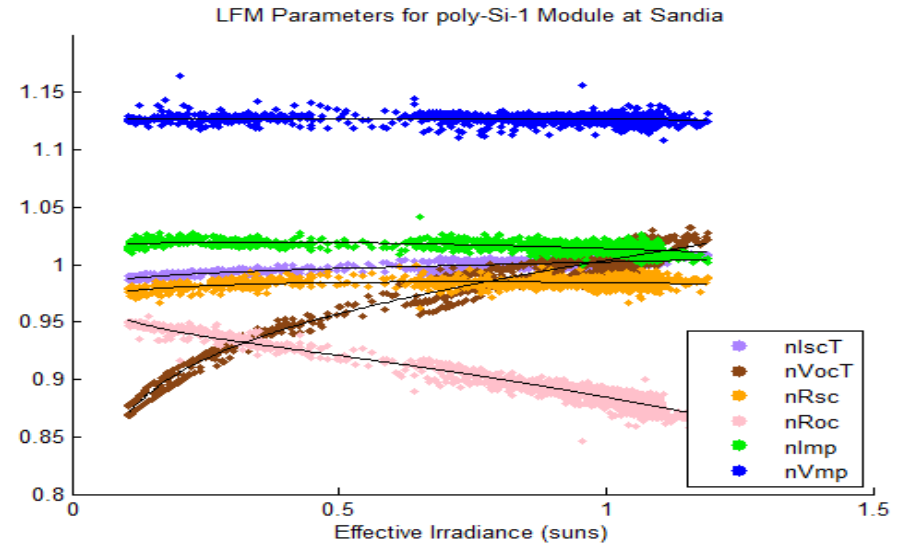
Six Normalized LFM Variables

1. $nIscT =$
 2. $nRsc =$
 3. $nImp =$
 4. $nVmp =$
 5. $nRoc =$
 6. $nVocT =$
- } Current factors
- } Voltage factors

$$pImp = nIscT * nRsc * nImp * rImp * MMF * G_i \div T_{CORR,IsC}$$

$$pVmp = nVmp * nRoc * nVocT * rVmp \div T_{CORR,Voc}$$

MMF = spectral mismatch, G_i = irradiance



LFM fits each of the LFM variables to a function of irradiance for a total of $6 \times 3 = 18$ coefficients

$$nf(G_i) = c_1 + c_2 \log(G_i) - c_3 \times G_i^2$$

$$T_{CORR,Voc} = 1 + \beta_{Vmp} \times (25 - T_c)$$

$$T_{CORR,IsC} = 1 + \alpha_{Imp} \times (25 - T_c)$$

'Data' models

- These models are based directly on measure data (e.g., 61853-1) or outdoor IV curves collected over time or trained on these data.
- Models describe how to interpolate and extrapolate from a measured reference data set.
 - *Janine Freeman* will describe one such model developed at NREL.
 - These models do not have parameters but rather are descriptions of a procedure that results in a result.
- Another type of “Data” model is Machine Learning
 - *Birk Jones* from Sandia has used a Gaussian Process Regression algorithm to simulate IV curves
 - Jones, C. B., M. Martinez-Ramonz, R. Smith, C. K. Carmignani, O. Lavrova and J. S. Stein (2016). Automatic Fault Classification of Photovoltaic Strings Based on an In-Situ IV Characterization System and a Gaussian Process Algorithm. 43rd IEEE Photovoltaic Specialist Conference. Portland, OR.

Issues common to all performance models

- Parameter estimation is generally not transparent nor reproducible
- Data requirements are not well understood:
 - Thought to be either:
 - Minimal, e.g., fit a diode model using single IV curve and a temperature coefficient, or
 - Extensive, e.g., requiring days of measurements outdoors on a two-axis tracker
 - Both are wrong
 - 61853-1 is enough for diode or point models, with the exception of the spectral mismatch function
- Multiple models in use but underlying data is not shared
 - Some resort to creating “models of models”, e.g., fitting the SAPM to a PVsyst simulation
 - A modeled model inherits the problems of the source model

Useful Module Model Characteristics

- Parameters are easy/straightforward to obtain
 - Standards exist for the collection of the characterization data
 - Calculations yield consistent parameter values (solutions are unique)
 - Estimation is transparent and reproducible from a common data set
 - Parameter values are easily shared and published
 - Values of parameters with physical meaning do not violate physics

- Model predicts performance well across the entire range of environmental conditions
 - High and low irradiance
 - High and low temperatures
 - Spectral effects

Questions?



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