

Using similar mathematical modelling with both single module IV curve measurements and array Inverter data

Juergen Sutterlueti¹ and Steve Ransome²

¹Gantner Instruments, Austria

J.Sutterlueti@gantner-instruments.com

²Steve Ransome Consulting Ltd, U.K.

steve@steveransome.com

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Levels	DC, AC	Performance Measurements	Variability, losses	Models	
Modules - Strings	DC	Can be full IV curves, indoor or outdoor	Minimal with single modules, mismatch in strings	1-diode, SAPM, LFM etc.	
Inverter	DC → AC	V _{MPP} tracking, Efficiency vs Pin, Thermal cut out, clipping, downtime etc.			ontner Instruments (GI) OTF in Tempe, AZ
Large Arrays	DC side or AC side	Usually just P _{MAX} , sometimes I _{MAX} and V _{MAX}	Module mismatch, Soiling, T _{MOD} , Wiring losses Cloud variability across array	Empirical e.g PVUSA, scaled module + inverter model, MPM etc.	37.8 MWp Kent, Great Britain

Comparison of two different models used by GI/SRCL



Model Introduced	DC, AC	Parameters	Pros and Cons	
LFM Loss Factors Model (2011 26th Hamburg PVSEC)	DC	Derives 6 normalised IV curve parameters and 2 curvature checks	 Optimised model for good IV traces. Cannot derive R_{sc}, R_{oc} from poor IV curves 	Curvature MMF Isc Infisc I
MPM Mechanistic Performance Model (2017 7 th PVPMC Canobbio)	DC side or AC side	Derives 5 normalised PR _{DC} parameters	 Optimised model for PR_{DC} for indoor matrix, outdoor data Only derives PR_{DC} 	Julia and the second se

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Comparison of two different models used by GI/SRCL



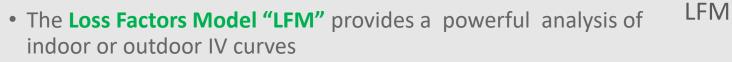
Model introduced	DC, AC	Parameters	Pros and Cons	Enhancements
LFM Loss Factors Model (2011 26th Hamburg PVSEC)	DC	Derives 6 normalised IV curve parameters and 2 curvature checks	 Optimised model for good IV traces. Cannot derive R_{sc}, R_{oc} from poor IV curves 	Simplify LFM parameters to work with lower quality IV curves and also just Imp and Vmp values
				✓ Overlap !
MPM Mechanistic Performance Model (2017 7 th PVPMC Canobbio)	DC side or AC side	Derives 5 normalised PR _{DC} parameters	 Optimised model for PR_{DC} for indoor matrix, outdoor data Only derives PR_{DC} 	Generalise MPM equations to work with LFM like parameters e.g. nI _{SC} , nR _{SC} , nV _{OC}



Are Coefficients	IFM and MPM	🗵 Empirical models		
<u>Meaningful</u> (depend on physical effects)	Yes e.g. • $P_{ACTUAL}/P_{NAMEPLATE} = P$ Tolerance • $1/P * dP/dT_{MODULE} = Temp Coeff \gamma$ • $nV_{oc} = measV_{OC}/refV_{OC}$ etc.	No <u>not</u> dependent on real behaviour e.g. X1, X2 $T_{MOD} * T_{AMBIENT}$ $T_{MOD} * LOG(G_1)^2$		
Orthogonal (independent of each other)	Yes Allow for <u>a unique fit.</u>	No Unique fits aren't possible.		
<u>R</u> obust (can fit data with noise)	Yes Fits sensibly <u>without being perturbed</u>	No <u>non realistic predictions</u> e.g. spikes and/or non linearity		
Normalised (divide by reference values)	Yes parameters can be compared and validated more easily e.g. nV _{OC} = V _{OC.MEAS} /V _{OC.NOMINAL} may be 95% module → string	No <u>hard to see what is good behaviour</u> at different levels e.g. V _{oc} =30V module, V _{oc} =600V string What about MPPT limits?		

Overview of The Loss Factors Model (LFM)

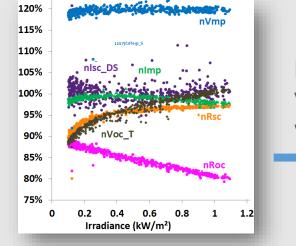
E.g [Stein et al 28th PVSEC Paris 2013] for comparison with 1-diode and SAPM



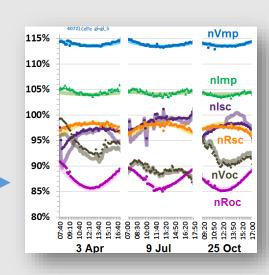
 $PR_{DC} = nI_{SC} * nR_{SC} * nI_{MP} * nV_{MP} * nR_{OC} * nV_{OC}$

 Two other parameters nl_c and nV_c show the deviation from expected I@V_{MP}/2 and V@I_{MP}/2 and give measured values indicating amounts of cell current mismatch and roll over respectively

Characterise a module vs. G_I, Tmod etc.



Predict performance vs. time and weather



MMF

nlsc

nRsc nImp **Icurvature**

MEASURED IV mVmp*mImp

=Pmax.meas

Gantner

nVmp

nRoc

LFM can easily find any discrepancies, degradation, poor measurements etc

SRCL Gantner Instruments

REFERENCE STC IV

rVmp * rImp

= Pmax.nom

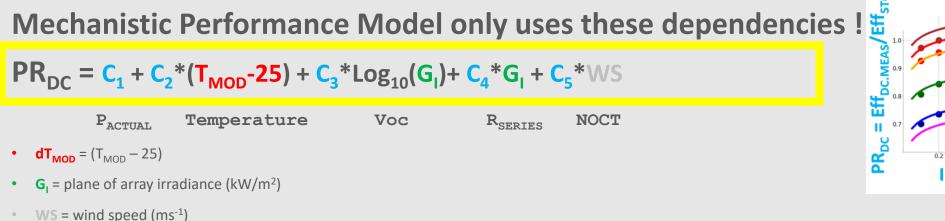
Vcurvature

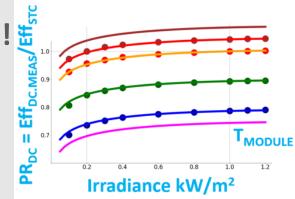
nVoc

6 LOSS FACTORS MODEL + 2 CHECKS



I_{MAX} ∝ G₁
 Module STC rating actual/nominal W
 P_{MAX} ∝ (1+γ*(T_{MOD}-25))
 V_{MAX} ∝ log(G₁)
 From diode equation
 ΔP_{MAX} ∝ I_{MAX}²* R_{SERIES}
 I².R_S loss
 T_{MOD} ~ T_{AMB} - fn(Windspeed)
 NMOT Thermal rise
 R_{SHUNT} vs. GI
 (depends on technology) - behaves like 3.V_{MAX}







[7th PVPMC Mar 2017 Canobbio CH][44th PVSC Jun 2017 Washington DC US],[PVSEC-27 Nov 2017 Shiga JAP]

 $PR_{DC}(y)$ vs $G_{I}(x)$ and T_{MOD} (coloured lines) Smooth Data Empirical model A **Empirical model D** All models can fit smooth data well echanistic model

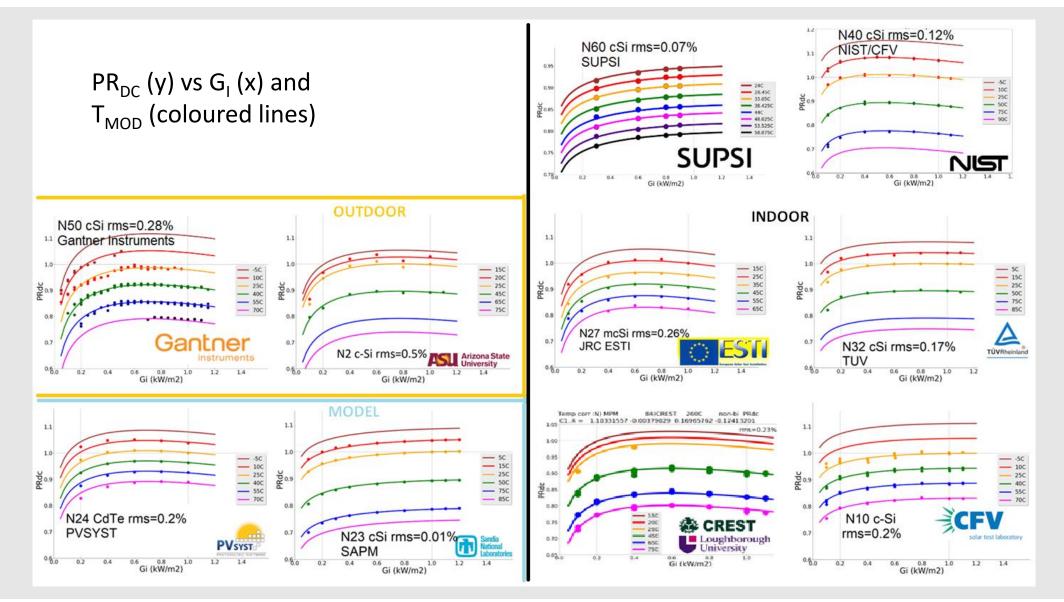
Added noise to mimic real measurements and test robustness

Empirical models can be affected by random scattered points and give non physical fits

Mechanistic models can be unaffected by random scattered points and still gives physical fit



(Please send us any more for confidential comparisons or to be included here !)



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Why mechanistic model coefficient values are more useful than from empirical models



Compare fits vs. technology cSi aSi CdTe

[SUPSI data]

Technology	ID	C ₁	C ₂	C ₃	C ₄	C ₅	rms
c-Si	60)	-42.3	53.9	-10.7	-32.9	-8.0	0.22%
c-Si	62)	127.2	159.8	-31.7	-97.8	-23.9	0.22%
c-Si	64)	-71.5	90.5	-18.0	55.3	-13.5	0.09%
c-Si	66) :	-93.4	117.6	-23.4	-72.0	-17.6	1.84%
c-Si	67)	100.2	123.8	24.6	5 75.6	18.4	0.24%
c-Si	68)	-69.5	87.9	-17.9	5 -53.8	-13.1	0.16%
c-Si	70)	-37.3	131.4	-98.7	21.9	23.5	0.10%
c-Si	71)	6.4	-6.8	1.4	4.1	0.9	0.07%
c-Si	72)	60.7	132.5	76.7	10.1	-10.2	0.59%
c-Si	73)	53.8	-68.9	16.3	3 38.4	8.7	0.09%
TF a-Si	65)	0.2	1.1	-0.8	-0.5	-0.1	0.94%
TF a-Si	74)	90.8	121.1	31.9	62.8	13.2	0.32%
TF CdTe	63)	-0.6	2.2	- <mark>0.</mark> 6	5 -1.2	-0.3	0.27%
c-Si	60)	96.2%	-0.45%	8.3%	6 -2.1%	0.0%	0.07%
c-Si	62)	109.6%	-0.42%	20.5%	6 -10.0%	0.0%	0.09%
c-Si	64)	106.4%	-0.45%	8.5%	6 -6.4%	0.0%	0.09%
c-Si	66) :	107.7%	-0.48%	11.9%	6 -7.7%	0.0%	0.08%
c-Si	67)	115.2%	-0.48%	18.29	6 -15.4%	0.0%	0.11%
c-Si	68)	107.6%	-0.47%	10.49	6 -7.5%	0.0%	0.09%
c-Si	70)	103.7%	-0.46%	3.9%	6 -4.3%	0.0%	0.08%
c-Si	71)		-0.46%	24.4%	6 -12.4%	0.0%	0.08%
c-Si	72)	99.6 <mark>%</mark>	-0.44%	0.7%	6 1.2%	0.0%	0.20%
c-Si	73)			_	6 -9.2%	0.0%	0.09%
TF a-Si	65)	112.2%	-0.11%	31.69	6 -11.9%	0.0%	0.21%
TF a-Si	74)	122.7%		39.5%	-23.1%	0.0%	0.33%
TF CdTe	63)	127.3%	-0.23%	19.6%	-20.2%	0.0%	0.16%

Empirical model with 5 parameters

- no meaningful pattern to coefficients
- large variability
- self compensation (e.g. C₁ vs. C₂)

MPM Mechanistic model with 4 parameters

- Meaningful values of all coefficients
 more reduct fitting
 - = more robust fitting
- Better average rms fit

 $C_1 = P_{MAX}$ tolerance $C_2 = Realistic P_{MAX}$ Temperature coefficient (TF < c-Si)

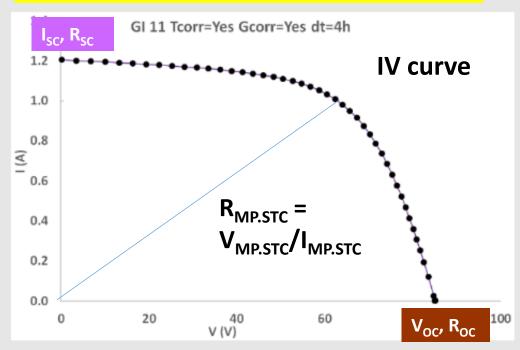
Checking smoothness and quality of IV data quality

With "normalised Resistance vs. Voltage" (nRV) curves shown on log scale

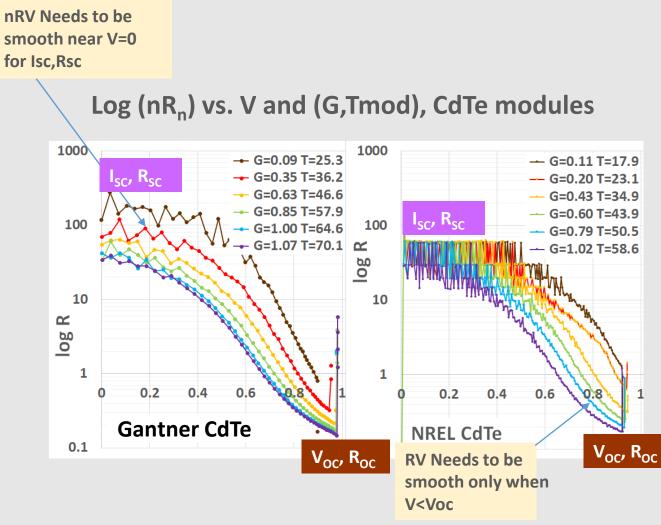


Smooth IV data is needed to extract precise values particularly for R_{sc}.

 $nR_n = -\Delta V / \Delta I / (R_{MP.STC})$



Analysing nRV data can show if limited precision, errors, drift, scatter exist



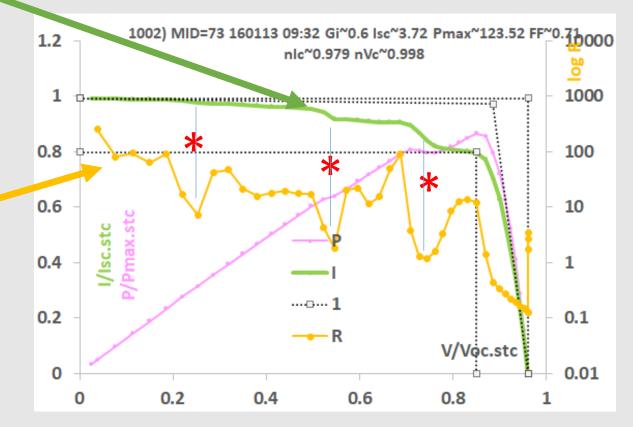
Gantner(L) much smoother I_{sc} than NREL Daystar (R)



Normalised IV curve

 $(I_{MEAS}/I_{SC.STC}/G_{I} vs. V_{MEAS}/V_{OC.STC})$ shows steps (*) between I_{SC} and I_{MP} indicative of cell current mismatch due to broken cells, variability or shading

- Derivation of nRV curves quantify and qualify the three steps (minima in nRV curve)
- Steps are also apparent in PV curve
- As nRV is normalised value should be ~1 at I_{MP} on a perfect (stepless) curve





- Accurate IV data such as GI's can be used to determine steps from shading and mismatch
- Each step has its own dip
 - Rule of thumb: if a dip is below 10 it is noticeable

MORNING 09:32 MID IRRADIANCE 0.6kW/m²

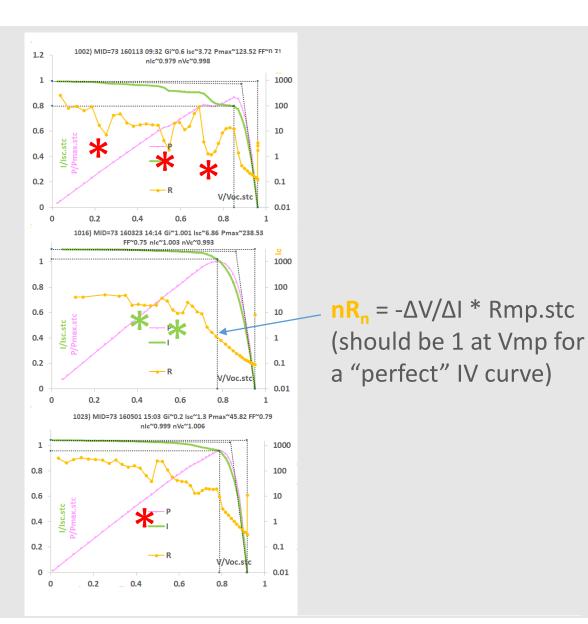
FF : Lower c-Si nRv : 3 Steps low<10

MID DAY 14:14 HIGH IRRADIANCE 1.0kW/m²

FF : High nRv : Smooth low>10

MID DAY 15:03 LOW IRRADIANCE 0.2kW/m²

FF : High nRv : Smooth low>10



cSi

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cSi and **Thin Film** normalised IV curves Imeas/Isc.stc vs. Vmeas/Voc.stc $nR_n = -\Delta V/\Delta I * rVmp/rImp$ (should be 1 at Vmp for a "perfect" device as top)



- Accurate IV data such as GI's can be used to determine steps from shading and mismatch
- Each step has its own dip
- Rule of thumb: if a dip is below 10 it is noticeable

Thin Film generally has lower FF than c-Si so doesn't show any large curvature effects

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MORNING 09:32 MID IRRADIANCE 0.6kW/m²

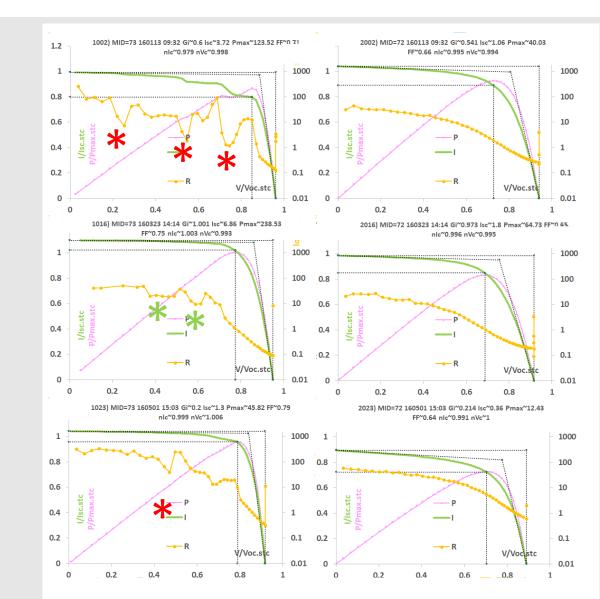
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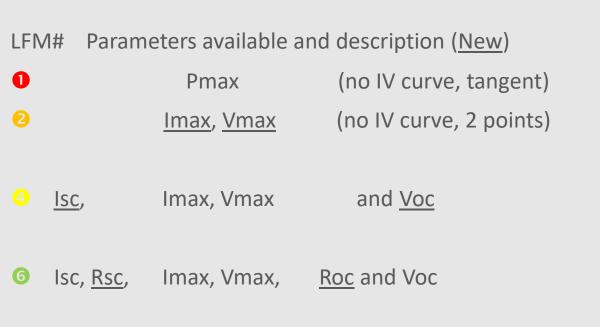


Thin Film

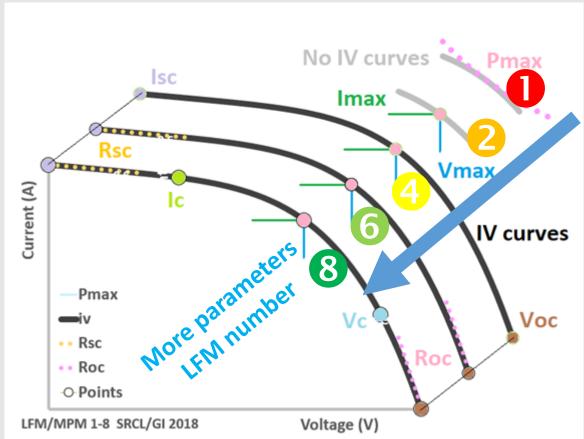
cSi

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Improving understanding from deriving more parameters from measurements (Generalise naming : previous LFM had 6 parameters and is called LFM6)



- Isc, Rsc, <u>Ic</u>, Imax, Vmax, <u>Vc</u>, Roc and Voc
 Where
 - Ic = Icurvature (measured vs. expected) at Vmp/2 due to cell mismatch, shading, non linear Rshunt
 - Vc = Vcurvature (measured vs. expected) at Imp/2 due to roll over or other non ideal behaviour

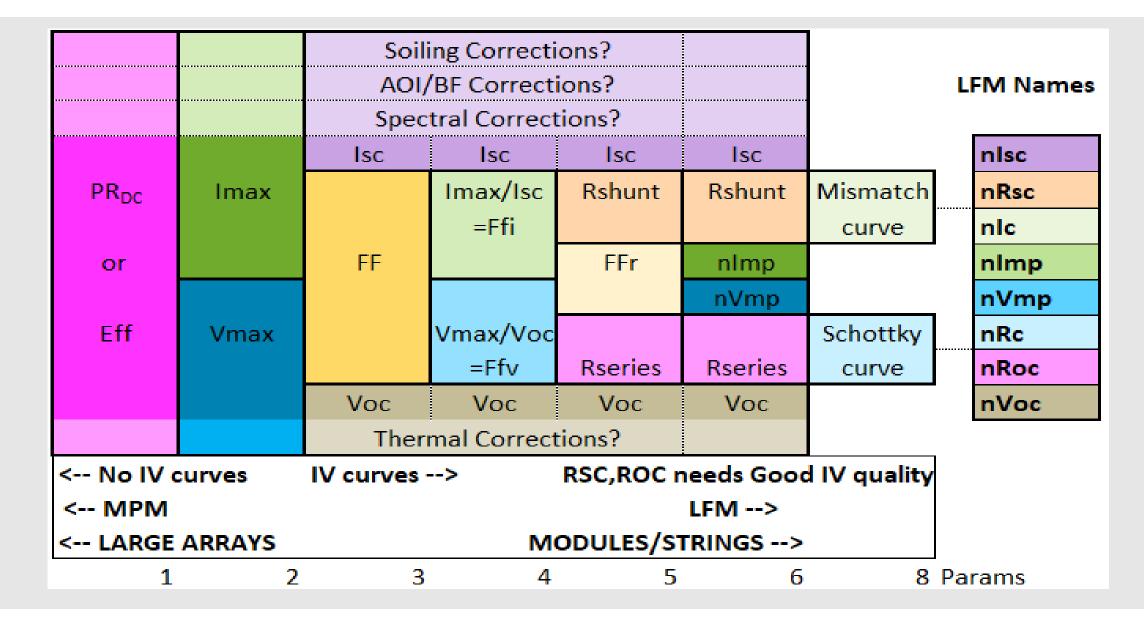


SRCL

Gantner



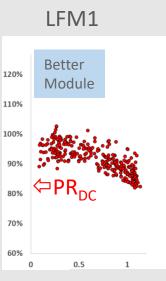


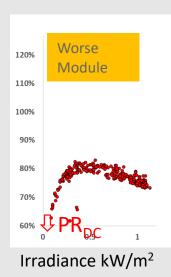




Model and #parameters = Which parameters make these modules differ?

PR_{DC} at all irradiances and especially at low light





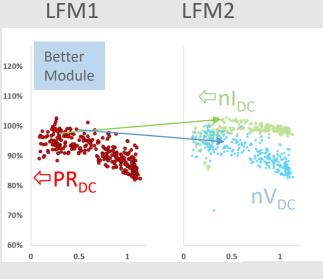
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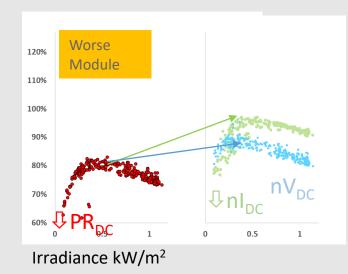
Model and #parameters = Which parameters make these modules differ?

PR_{DC} at all irradiances and especially at low light

low light nl_{DC}



PR_{DC}→nl_{DC}*nV_{DC}



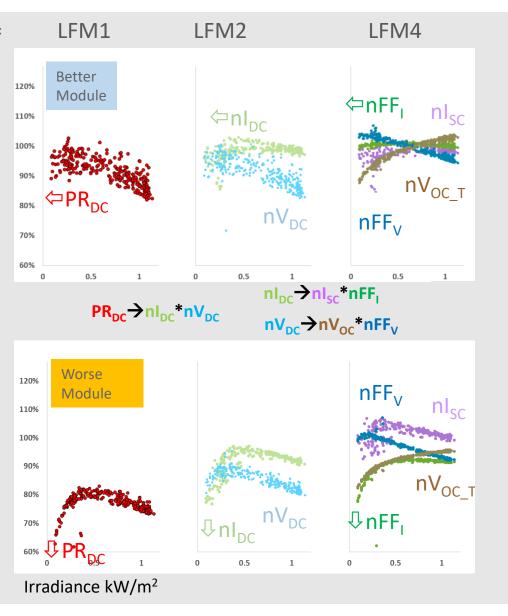
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Model and #parameters = Which parameters make these modules differ?

PR_{DC} at all irradiances and especially at low light

Iow light nl_{DC}
 Iow light nFF₁



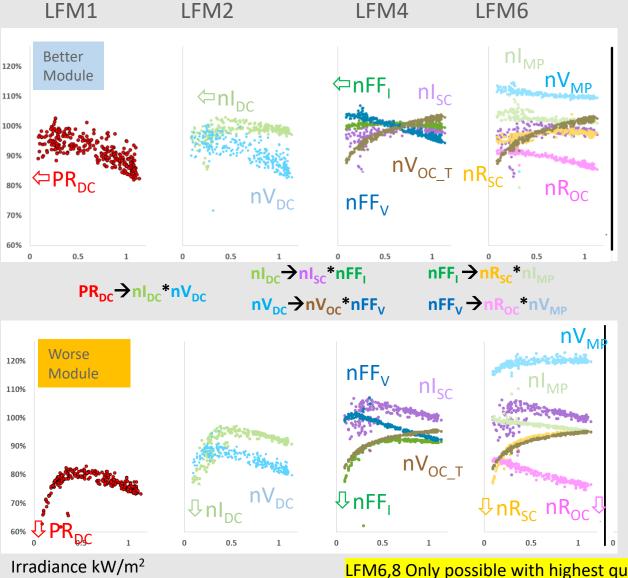


Model and #parameters = Which parameters make these modules differ?

- **PR**_{DC} at all irradiances and especially at low light
- low light nl_{DC}
- **4** low light **nFF**_I

Analysing Resistance parameters needs good quality IV curves

Iow light nR_{sc} (better R_{SHUNT}) high light nR_{oc} (lower R_{SERIES})



LFM6,8 Only possible with highest quality measurements



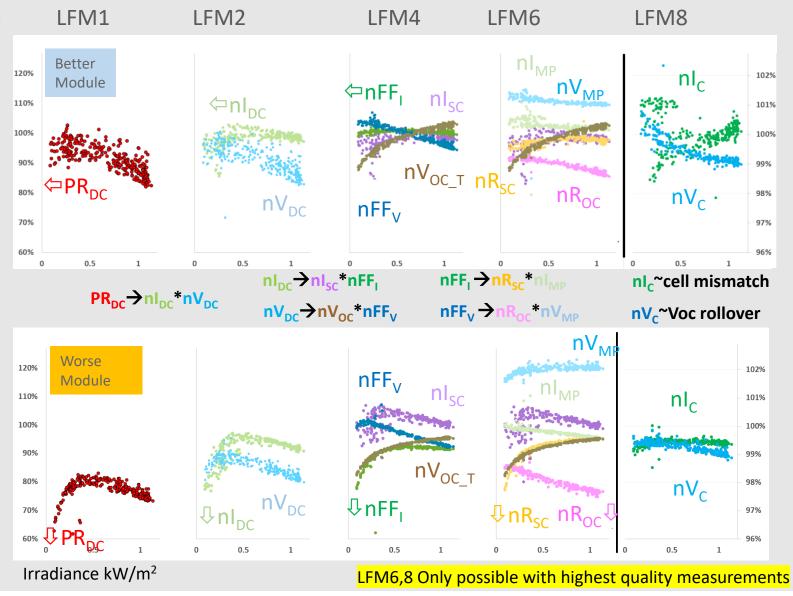
Model and #parameters = Which parameters make these modules differ?

- **PR**_{DC} at all irradiances and especially at low light
- low light nl_{DC}
- **4** low light **nFF**_I

Analysing Resistance parameters needs good quality IV curves

- **6** low light nR_{SC} (better R_{SHUNT})
 - high light nR_{OC} (lower R_{SERIES})

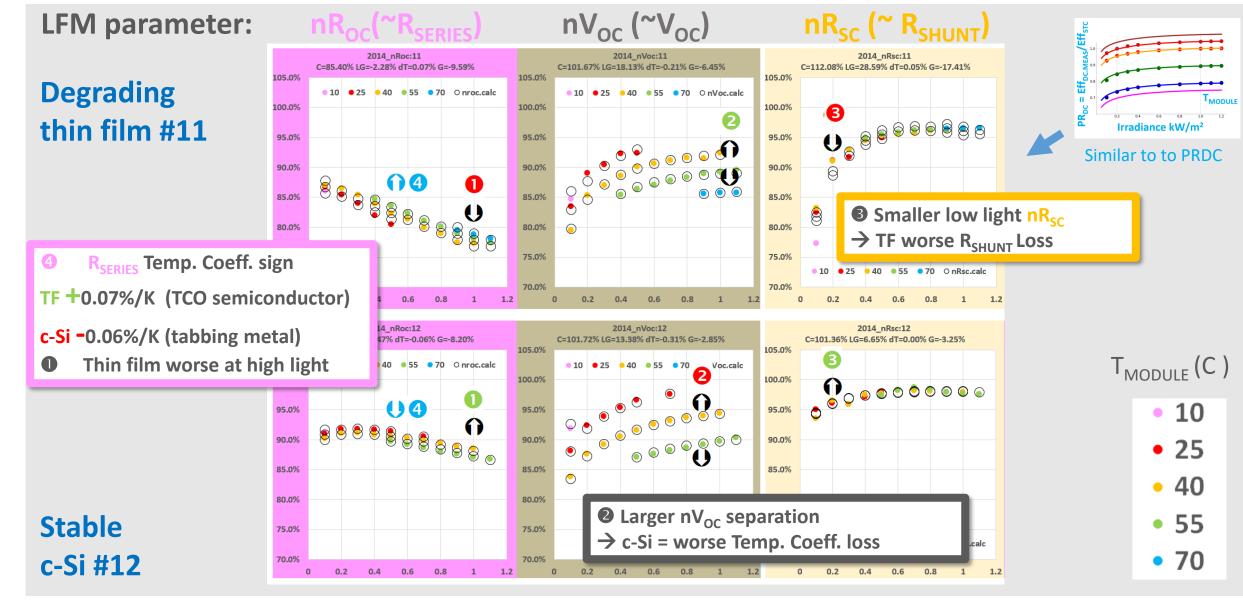
Ourvature parameters show more nlc scatter in good module particularly at low light (cell mismatch and/or shading), poor module has lower FF so is less affected



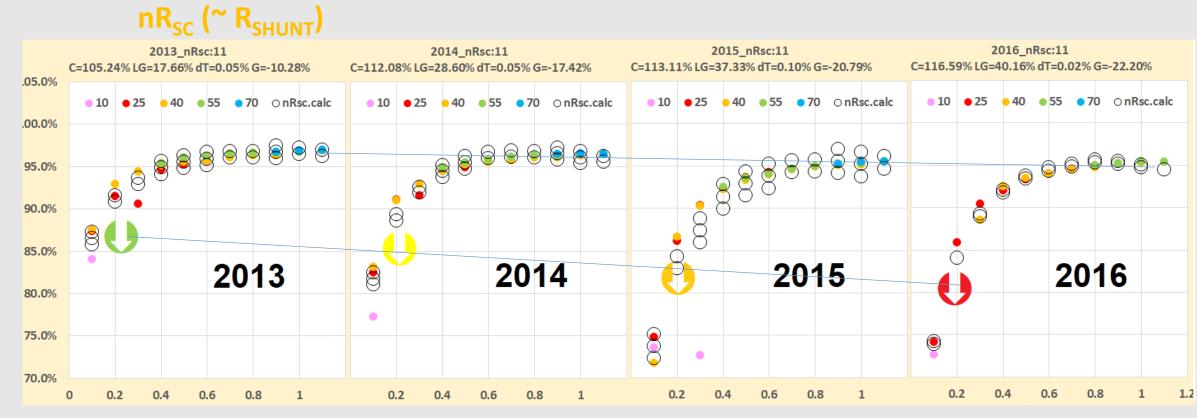
Using yearly matrix averages to determine outdoor performance



3 LFM parameters Θ vs. G_I \bigcirc and T_{MOD} (coloured dots) for two technologies (GI, 2014)



 nR_{sc} • vs. Gi • : Finding the cause and rate of drop of a degrading module 2013-16 (GI)



O 7% Performance fall at low light in 3 years is caused by degrading R_{SHUNT} causing nR_{sc} to reduce but at high light remains similar (GI data)

As PR_{DC} is the product of six coefficients –

<u>any</u> drop or change has a direct influence on PR_{DC} and therefore energy yield.

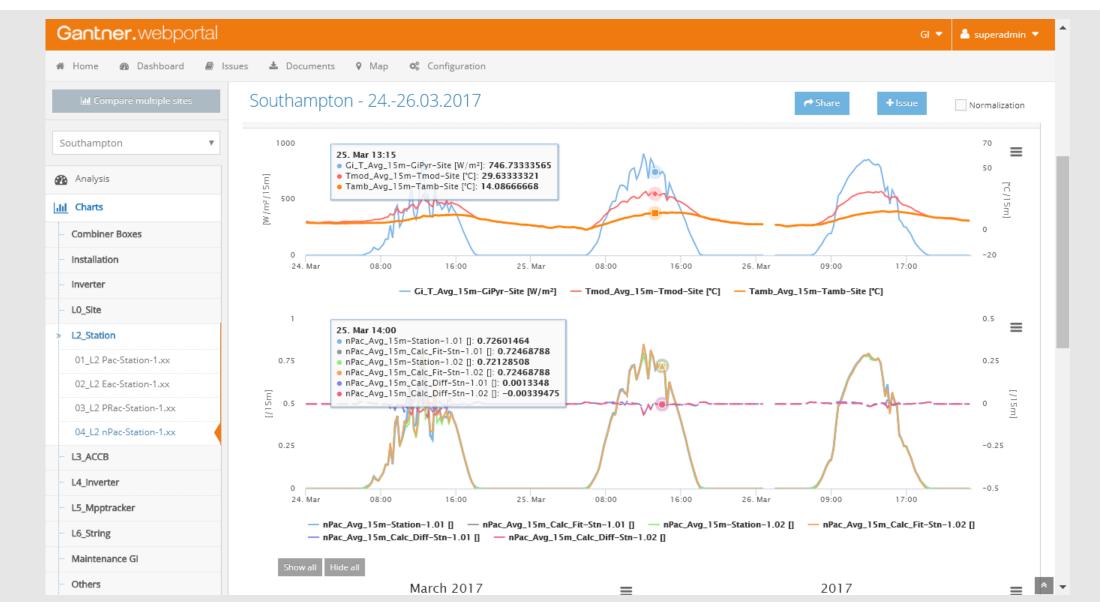
• 25

• 70



Large array UK variable weather day LFM 2 (I_{DC} and V_{DC}) predicted performance (real time)





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(as seen with Marios Theristis talk yesterday)

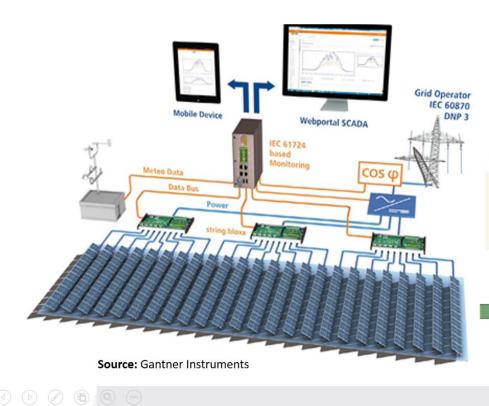
FOSS

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Gantner instruments University of Cyprus PV Technology

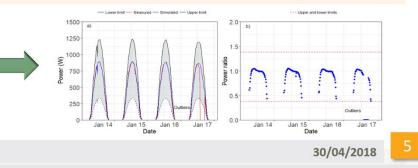
IPERMON



Partners: Gantner Isntruments and University of Cyprus
Project: IPERMON [Solar-ERA.net project]
Budget: €400,000
Duration: 36 Months
Weblink:

http://www.pvtechnology.ucy.ac.cy/projects/ipermon/

Development of innovative condition monitoring platform (algorithms and devices to quantify performance loss, diagnose faults and estimate degradation from acquired data).



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- Sophisticated modelling and analysis can be performed using the combined MPM/LFM method
- The normalised RV method shows data quality and also analyses steps from cell mismatch or shading
- If IV curves are not available (e.g. large arrays) then a good understanding of PV performance can be derived.
- With good IV curves (as from Gantner) the causes of any underperformance and rates of any PV degradation (e.g. "R_{SHUNT} at low light") can easily be found using these methods
- Scalable fault detection and analytics are available on the Gantner webportal available serving industry requirements

Thank you for your attention

[1] Ransome et al, 44th PVSC 2017 Washington DC "How to Choose the best Empirical Model for Optimum Energy Yield Predictions"

[2] Shiga "7TuPo "QUANTIFYING AND ANALYSING THE VARIABILITY OF PV MODULE RESISTANCES RSC AND ROC TO UNDERSTAND AND OPTIMISE KWH/KWP MODELLING"
 [3] Shiga "7MoO.5.4 "OPTIMISED FITTING OF INDOOR (E.G. IEC 61853 MATRIX) AND OUTDOOR PV MEASUREMENTS FOR DIAGNOSTICS AND ENERGY YIELD PREDICTIONS"
 [4] http://www.gantner-webportal.com

[5] Optimized PV Performance using State of the Art Monitoring for Increased Asset Value PVPMC8 2017



Spare slides



SaaS platform | Data processing – real time



