

This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 792059

June. 22nd. 2022 6. Performance analysis Yield simulation at design phase

Daniel Valencia-Caballero GOPV Summer School. Catania

GLOBAL OPTIMIZATION OF INTEGRATED PHOTOVOLTAIC SYSTEM FOR LOW ELECTRICITY COST

co-organized with





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• The purpose (normally) is getting an economic profitability of the PV installation



• Here, only PV energy production simulation is addressed

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FROM SUN RADIATION TO AC ELECTRICITY







FROM SUN RADIATION TO AC ELECTRICITY



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FROM SUN RADIATION TO AC ELECTRICITY

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- Example of gains/losses along each modelling step and the related uncertainty.
- Typical output tables or diagrams in yield assessment show the contribution. Rarely the uncertainty is provided.
- The starting point of PR = 100 is considered after applying the horizon shading as this become the annual insolation seen by the PV modules
- Note that radiation in [kWh/m²] and electricity in [kWh/kWp] give comparable numbers.
- Uncertainty equal to 6.5 % is the result of the sum of the root mean square error of the relative uncertainty of each step.

Annual values	uncertainty	value	gains/loss	PR
	%	kWh/m ²	%	%
global irradiation on horizontal plane	4.0	1248	·	
irradiation on module plane	2.5	1448	16.0	
shading				
horizon shading	0.5	1445	-0.2	100.0
row shading	2.0	1422	-1.7	98.3
object shading	3.0	1422	0.0	98.3
soiling	0.5	1414	-0.5	97.9
deviations from STC				
reflection losses	0.5	1376	-2.7	95.2
	%	kWh/kWp	%	%
spectral losses	0.5	1363	-1.0	94.3
irradiation-dependent losses	0.8	1342	-1.5	92.9
temperature-dependent losses	1.0	1309	-2.5	90.5
mismatch losses	0.5	1298	-0.8	89.8
DC cable losses	0.5	1287	-0.8	89.1
inverter losses	1.5	1272	-1.2	88.0
inverter power limitation	0.5	1272	-0.1	88.0
additional consumption	0.5	1270	-0.1	87.9
AC cable losses low voltage	0.5	1265	-0.4	87.5
Transformer medium voltage	0.5	1253	-0.9	86.7
AC cable losses medium voltage	0.5	1252	-0.1	86.6
Transformer high voltage	0.0	1252	0.0	86.6
total	6.5	1252		86.6

Example done by Fraunhofer ISE

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J. Ascencio-Vásquez et al. "Methodology of Köppen-Geiger-Photovoltaic climate classification and implications to worldwide mapping of PV system performance," Solar Energy, vol. 191, pp. 672–685, Oct. 2019, doi: 10.1016/j.solener.2019.08.072.



THE PERFORMANCE RATIO

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J. Ascencio-Vásquez et al. "Methodology of Köppen-Geiger-Photovoltaic climate classification and implications to worldwide mapping of PV system performance," Solar Energy, vol. 191, pp. 672–685, Oct. 2019, doi: 10.1016/j.solener.2019.08.072.

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- Yield assessments (YA) and Long-Term Yield Predictions (LTYP) are a prerequisite for business decisions on long term investments into photovoltaic (PV) power plants.
- They should be provided with a related exceedance probability
- A reduction in the uncertainty of the energy yield can lead to a stronger business case.
- The main challenge in YA and LTYP relates to the trustworthiness of site-specific information
- The YA is **not only about the software** used; it is **mainly about the user**.
- There are always personal experience and assumptions: irradiance database selection and site adaptation, degradation/PLR assumption, total modelling uncertainty, soiling and far/near shading, ...



PV APPLICATIONS

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PV SOFTWARE FOR EACH APPLICATION

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- The reference in the market for big PV projects and bankability analysis
- Very detailed descriptions
- Confidence has been earned over time via repetition, experience, and accrued knowledge



- Useful to get meteo data of Europe and Africa
- Useful for first design and simple simulations

<mark>じ</mark>HelioScope

- Uses the same core as PVsyst for energy modeling.
- Unlike Pvsyst, it features an intuitive graphical user interface with Google Earth and SketchUp integration.



- Probably the best option among free software
- Complete options and accurate
- Most financiers do not accept SAM energy models in lieu of PVsyst reports



- Developed for BIPV, it uses a ray-tracing engine to calculate irradiance.
- Possible to import 3D files of buildings or other objects
- Compatible with BIM format



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MODELING STEPS

Example with SAM



Example done by Fraunhofer ISE

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Moser D. et al. "Uncertainties in Yield Assessments and PV LCOE," Nov. 2020. IEA-PVPS T13.



CONTR Control SAM 2020.2.29: C:\Users\108897\OneDrive - Fundacion Tecnalia Research & Innovation\Documentos\Proyectos\GOPV\Simulaciones Feb 2022 Cadarache 1axis vs fixed\SAM sim and results\1Row-28module



1.1.1			File 🗸 🔶 Add	untitled	
A CONTRACTOR OF			Photovoltaic, No fina	ncial	
			Location and Posou	Irco	Solar Kesource Library The Solar Resource Library is a list of weather files on your computer. Choose a file from the library and verify the weather data information below.
				lice	
Choose a performance model, and then choose fr	om the available financial models.		Module		The default library comes with only a few weather files to help you get started. Use the download tools below to build a library of locations you frequently model. Once you build your library, it is available for all of your work in SAM.
 Photovoltaic 	> Power Purchase Agreement		Inverter		Filter: Name V
Detailed PV Model	> Distributed		System Design		Name Latitude Longitude Time zone Elevation Station ID Source
PVWatts	Merchant Plant		Shading and Layout	t	imperial_ca_32.835205115.572398_psmv3_60_tmy 32.85 -115.58 -8 -20 72911 NSRDB phoenix_az_33.450495111.983688_psmv3_60_tmy 33.45 -111.98 -7 358 78208 NSRDB
High Concentration PV	LCOE Calculator (FCR Method)				tucson_az_32.116521110.933042_psmv3_60_tmy 32.13 -110.94 -7 773 67345 NSRDB
> Battery Storage	No Financial Model		Losses		Bilbao_tmy_43.2692.945_2007_2016 43.269 -2.945 1 1 unknown ECMWF/ERA
Concentrating Solar Power			Grid Limits		Meteo Cadarache albedo mensual 43.6974 5.7352 0 264 -
 Marine Energy 					SAM scans the following folders on your computer for valid weather files and adds them to your Solar Resource library. To use weather files stored on your computer, click Add/remove Weather File Folders and add folders containing the files.
Wind					C:\Users\108897/SAM Downloaded Weather Files Add/remove weather file folders
Fuel Cell-PV-Battery					V Refresh library
Geothermal					Downland Wasther File
Solar Water Heating					Download weatner rues The NSRDB is a database of thousands of weather files that you can download and add to your to your solar resource library: Download a default typical-year
Biomass Combustion					(TMY) file for most long-term cash flow analyses, or choose files to download for single-year or P50/P90 analyses. See Help for details.
biomass combustion					
Generic System		_			One location One location One location One location One location One location
					Type a location name, street address, or latitude and longitude Choose year V Download and add to library
					For locations not covered by the NSRDB, click here to go to the SAM website Weather Page for links to other data sources.
					Weather Data Information
					The following information describes the data in the highlighted weather file from the Solar Resource library above. This is the file SAM will use when you click Simulate.
					Weather file C:\Users\108897\SAM Downloaded Weather Files\Meteo Cadarache albedo mensual.csy
					Hander Data from Waather Ella
					- neader Data Hollin Weather File
					Station in - Latitude 43.0074 pp - or makers
					Data Source Longitude 5./332 DD coordinate DOU Can Larke the interest of data from
					Elevation 264 m Time zone GMT0 are the coo PVGIS and load it in SAM
					-Annual Averages Calculated from Weather File Data
		1			Global horizontal 4.38 kWh/m²/day Average temperature 10.6 °C Maximum snow dept
					Direct normal (beam) 5.21 kWh/m²/day Average wind speed 1.9 m/s Annual albedo 0.313085
			Simulate >	4	Diffuse horizontal 1.43 kWh/m²/day *NaN indicates missing data.
Help	OK Cancel		Parametrics Stor	chastic (Albedo - Shy Diffure Model - Irradiance Data (Advanced)
			P50 / P90 Ma	acros	∩ Albedo



THE IMPORTANCE OF RADIATION DATA



- Prediction of weather in a particular place is the most difficult process.
- The selected **radiation model may provide variation up to ±4%** in the simulated output.
- Overall uncertainty of the energy yield to fall in a range between 5 and 11%. The main source of uncertainty is related to the insolation estimation.
- For comprehensive analysis of big PV plants:
 - High-quality long-term ground-based measurements have been are rare.
 - Site adaptation techniques combine **short-term measured data (about 1 year) and long-term satellite estimates**.
 - Assuming a strong correlation, the strengths of both data sets are captured and the uncertainty in the long-term estimate can be reduced.
 - Root mean square errors for satellite-based irradiation reported in literature are situated between 4 % to 8 % for monthly and 2 % to 6 % for annual irradiation values.
- Solar irradiation at the Earth's surface is not stable over time for all locations on earth.
- An increase in the average irradiance has been suggested. In Germany, about 1,1%/year



Solar irradiation at the Earth's surface is not stable over time for all locations on earth.



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EXAMPLE IN SAM – PV MODULE

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- Normally there is a module database.
- PAN files is a standard file format for PV modules information.
- It can be also defined from manufacturer datasheets.
- It is becoming common the characterization by independent parties.

	· ·													
Name	Technology	Bifacial	STC	PTC	A_c	Length	Width	N_s	l_sc_ref	V_oc_ref	l_mp_ref	V_mp_ref	alpha_sc	11
JA Solar JAM72S01-370/PR	Mono-c-Si	0	370.041000	343.600000	1.880000			72	9.910000	48.180000	9.380000	39.450000	0.004162	
JA Solar JAM72D09-375/BP	Mono-c-Si	1	374.768000	348.200000	1.970000			72	9.970000	48.500000	9.440000	39.700000	0.004088	
JA Solar JAM72S01-375/PR	Mono-c-Si	0	375.240000	348.400000	1.880000			72	9.980000	48.450000	9.440000	39.750000	0.004192	
JA Solar JAM72S09-375/PR	Mono-c-Si	0	375.315000	349.400000	1.940000			72	10.060	48.500000	9.550000	39.300000	0.003823	
JA Solar JAM72D09-380/BP	Mono-c-Si	1	380	353	1.970000			72	10.030	48.800000	9.500000	40	0.004112	
JA Solar JAM72S01-380/PR	Mono-c-Si	0	380.285000	353.200000	1.880000			72	10.050	48.710000	9.500000	40.030000	0.004221	
JA Solar JAM72S09-380/PR	Mono-c-Si	0	380.160000	354.200000	1.940000			72	10.120	48.800000	9.600000	39.600000	0.003846	
JA Solar JAM72D09-385/BP	Mono-c-Si	1	384.865000	357.700000	1.970000			72	10.090	49.100000	9.550000	40.300000	0.004137	· 🗸
<														>

Module Characteristics at Reference Conditions



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EXAMPLE IN SAM – TEMPERATURE MODEL

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• Here there are two model for calculating the PV module operating temperature. In general, this models works fine with the appropriate coefficients

• It is important to select the proper mounting configuration of the models, as the module temperature will impact the final PV production results

) Nominal operating cell temperature (NO	CT) method	∫ _[NOCT r	nethod parameters		
) Heat transfer method			Mounting standoff Ground or ra	ck mounted	\sim
ee Help for more information about CEC c	ell temperature models.		Array height One story bu	ilding height or lo	wer 🗸
Heat transfer method parameters					
Mounting configuration R	ack	\sim	Rows of modules in array	1	
Heat transfer dimensions	lodule Dimensions	\sim	Columns of modules in array	10]
Mounting structure orientation St	tructures do not impede flow und	derneath module \sim	Temperature behind the module	20	°C
Module width	1 m	Space	between module back and roof surface	0.05	m
Module length	1.97 m				-
Module length	1.97 m				
Module length ysical Characteristics Material Mono-c-Si	1.97 m Module ar	rea 1.970 m²	Number of cells	; 72	2
Module length ysical Characteristics Material Mono-c-Si ditional Parameters	1.97 m Module a	rea 1.970 m²	Number of cells	; 72	2
Module length ysical Characteristics Material Mono-c-Si ditional Parameters T_noct 45.1 °C	1.97 m Module an	rea <u>1.970</u> m ² _ref <u>10.096</u> A	Number of cells	s 72	2 2 9 Ohm



EXAMPLE IN SAM – INVERTER SELECTION

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- OND files is a standard file format for inverters information.
- It can be also defined from manufacturer datasheets.
- It is becoming common the characterization by independent parties.
- Inverter power (AC power) is normally lower than DC power for optimizing the inverter cost and efficiency
- Normally DC/AC power is 1.2 for monofacial, and 1 for bifacial.

Name	~													
Name	Paco	Pdco	Pso	Pnt	Vac	Vdcmax	Vdco	Mppt_high	Mppt_low	C0	C1	C2	C3	^
Fronius USA: Fronius Primo 10.0-1 [208V]	9995	10292.20	36.211	2.998500	208	800	660	800	100	-4.2756	-0.000029	-0.000759	-0.00	
Fronius USA: Fronius Primo 10.0-1 [240V]	9995	10295.99	44.270	2.998500	240	800	655	800	100	-7.9973	-0.000028	-0.000619	0.000	
Fronius USA: Fronius Primo 11.4-1 [208V]	11	11743.89	36.851	3.420000	208	800	660	800	240	-4.7172	-0.000032	-0.000730	-0.00	
Fronius USA: Fronius Primo 11.4-1 [240V]	11	11738.66	49.129	3.420000	240	800	660	800	240	-6.6822	-0.000033	-0.000724	-0.00	
Fronius USA: Fronius Primo 12.5-1 [208V]	12	12891.54	43.310	3.750000	208	800	665	800	260	-6.7086	-0.000033	-0.001410	-0.00	¥
<													>	

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Efficiency Curve and Characteristics

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EXAMPLE IN SAM – ELECTRICAL CONFIGURATION



• DC – AC ratio is normally 1.2 for monofacial installations. About 1 for bifacial due to extra production of rear side

AC Sizing	Sizing Summary				
Number of inverters 1	Nameplat	te DC capacity	10.776 kWdc	Total number of modules	28
DC to AC ratid 0.95	Tot	al AC capacity	11.400 kWac	Total number of strings	2
Size the system using modules per string and strings in parallel inputs below.	Total invert	er DC capacity	11.739 kWdc	Total module area	55.2 m²
Estimate Subarray 1 configuration					
DC Sizing and Configuration					
De sizing une comgaration					
To model a system with one array, specify properties f in parallel to a single bank of inverters, for each suba	or Subarray 1 and disa rray, check Enable and	ble Subarrays 2, 3, an specify a number of s	d 4. To model a syter trings and other pro	m with up to four subarrays connecte perties.	d
To model a system with one array, specify properties to in parallel to a single bank of inverters, for each suba	or Subarray 1 and disa rray, check Enable and Subarray 1	ble Subarrays 2, 3, an specify a number of s Subarray 2	d 4. To model a syter trings and other pro Subarray 3	m with up to four subarrays connecte perties. Subarray 4	:d
To model a system with one array, specify properties f in parallel to a single bank of inverters, for each suba -Electrical Configuration	or Subarray 1 and disa rray, check Enable and Subarray 1 (always enabled)	ble Subarrays 2, 3, an specify a number of s Subarray 2	d 4. To model a syter trings and other pro Subarray 3	m with up to four subarrays connecte perties. Subarray 4 □ Enable	:d
To model a system with one array, specify properties f in parallel to a single bank of inverters, for each suba -Electrical Configuration Modules per string in subarray	or Subarray 1 and disa rray, check Enable and Subarray 1 (always enabled) 14	ble Subarrays 2, 3, an specify a number of s Subarray 2	d 4. To model a syter trings and other pro Subarray 3 Enable	m with up to four subarrays connecte perties. Subarray 4 Enable	:d
To model a system with one array, specify properties f in parallel to a single bank of inverters, for each suba -Electrical Configuration Modules per string in subarray Strings in parallel in subarray	or Subarray 1 and disa rray, check Enable and Subarray 1 (always enabled) 14 2	ble Subarrays 2, 3, an specify a number of s Subarray 2	d 4. To model a syter trings and other pro Subarray 3	m with up to four subarrays connecte perties. Subarray 4 Enable	:d
To model a system with one array, specify properties f in parallel to a single bank of inverters, for each suba -Electrical Configuration Modules per string in subarray Strings in parallel in subarray Number of modules in subarray	or Subarray 1 and disa rray, check Enable and Subarray 1 (always enabled) 14 2 28	ble Subarrays 2, 3, an specify a number of s Subarray 2	d 4. To model a syter trings and other pro Subarray 3	m with up to four subarrays connecte perties. Subarray 4	:d
To model a system with one array, specify properties f in parallel to a single bank of inverters, for each suba -Electrical Configuration Modules per string in subarray Strings in parallel in subarray Number of modules in subarray String Voc at reference conditions (V)	or Subarray 1 and disa rray, check Enable and Subarray 1 (always enabled) 14 2 28 687.4	ble Subarrays 2, 3, an specify a number of s Subarray 2	d 4. To model a syter trings and other pro Subarray 3	m with up to four subarrays connecte perties. Subarray 4	:d

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EXAMPLE IN SAM – ORIENTATION

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	-Shade Loss Tables				
utomatically generate shade data from a drawing the array and shading objects.	Edit and import shade of software and devices, o	lata. Data may be entered r generated by the 3D shac	by hand, imported from s le calculator.	hade analysis	
Open 2D shade salsulator	Subarray 1	Subarray 2	Subarray 3	Subarray 4	
	Edit shading	Edit shading	Edit shading	Edit shading	
If Shading for Fixed Subarrays and One-axis Tracl	cers				
elf shading is shading of modules in the array by m	odules in a neighboring ro	Ν.			
Self shading	None	None	V	∨ None	\sim
rray Dimensions for Self Shading, Snow Losses, an	d Bifacial Modules				
he product of number of modules along side and bo	ttom and number of rows	should be equal to the nur	nber of modules in subar	ray.	
Module orientation	Landscape \checkmark	Portrait 🗸 🗸	Portrait ~	Portrait 🗸 🗸	
Number of modules along side of row	2	0	0	2	
Number of modules along bottom of row	14	0	0	9	
-Calculated System Layout					
Number of rows	1	NaN	NaN	0	
			0	0	
Modules in subarray from System Design page	28	0	v		
Modules in subarray from System Design page Length of side (m)	28	0	0	3.66005	
Modules in subarray from System Design page Length of side (m) GCR from System Design page	28 2.15297 0.35	0 0 0.25	0	3.66005	

Snow losses are caused by snow covering the array. When your weather file includes snow depth data, SAM can estimate losses due to snow. Losses are calculated for each subarray.

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Estimate snow losses

ΕΥΛΝΛΟΙ Ε ΙΝΙ ΟΛΝΛ

or PV

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EXAMPLE IN SAM – SOILING AND DC LOSSES

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☐ Irradiance Losses

 Soiling is an uncertainty which strongly depends on the environment of the system, raining conditions, etc.

- In medium-rainy climates (like middle of Europe) and in residential zones, this is usually low and may be neglected (less than 1%)
- Common values are:
 - 5% for regions with long dry seasons
 - 2% for regions with year-round rain
- O&M costs are in part dictated by the frequency of cleaning required
- In bifacial, soiling on rear side is normally 10 times lower than in front side.





DC Losses

DC losses apply to the electrical output of each subarray and account for losses not calculated by the module performance model.

Soiling losses apply to the total solar irradiance incident on each subarray. SAM applies these losses in addition to any losses on the

Module mismatch (%)	2	2	2	2	
Diodes and connections (%)	0.5	0.5	0.5	0.5	
DC wiring (%)	2	2	2	2	
Tracking error (%)	0	0	0	0	
Nameplate (%)	0	0	0	0	
DC power optimizer loss (%)	0	All four subarrays are sub	ject to the same DC pow	er optimizer loss.	
Total DC power loss (%)	4.440	4.440	4.440	4.440	
	Total DC power loss = 100% * [1	- the product of (1 - loss/100%)	1		
- Default DC Losses Apply default losses to replace DC losses f	or all subarrays with defa	ult values.	een 0,01%-3%. Indu	istry consensus	is 2%
Apply default losse	s for: Central inve	ters Microinverters	s DC optimizers		



EXAMPLE IN SAM – AC & TRANSMISSION LOSSES



AC Losses					1	
AC losses apply to the electrical output of the inverter	and account for losses not ca	alculated by the inverter perfo	ormance model.			
	AC wiring 1]%→	Potential losses be (AC cabling resistar	tween inverters and nce,)	d transformers	
Transformer Losses					1	
The transformer loss model is intended for distributi and assume a power factor of 1. The transformer ca	on or substation transformers pacity is equal to the total inve	in large PV systems. Losses a erter AC power rating.	apply to the electrical ou	utput of the inverter		
				 In big PV plantas 	there are transformers to	
Transformer no load loss	0 % Tra	ansformer load loss	0 %	increase the Volt	age	
·			/	 This data should datasheet 	be taken from transforme	rs
Transmission Losses				Gatasheet		
Transmission losses apply to the system generated po	wer output.					
Transr	nission loss 0	%				
System Availability					-	
System availability losses reduce the system output	C Losses	-AC Los	ses			
Availability losses may be applied either on the DC or	Edit losses Constant loss: 0	0.0 % Edit los	sses Constant loss: 0.	.0 %		
AC side of the system.	Hourly losses: N Custom periods	None s: None	Hourly losses: N	one None		
	custom penous	s none	custom penous	, none		

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THE AVAILABILITY

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- PV power plant effective availability
 - Availability of our power plant because is working properly
 - Availability of the grid to accept power
- O&M contractor guaranteed availability values typically are 99% or higher
- Availability is between 92% and 99.5%, typically 99 % ±1 %
- Unavailability has decreased over time, possibly based on improved O&M protocols
- Plants in areas with constrained grids can have availabilities below 90%



Figure 14: Statistical indicators of mean monthly unavailability of 533 PV plants in 2019 (majority installed in Belgium, Germany and Switzerland)

"Market Notices - Update to system strength requirements in North Queensland, Market Notice 76455 CONSTRAINTS,"Jul. 27, 2020. https://aemo.com.au/Market Notices

C. Tjengdrawira and M. Richter, "Review and Gap Analyses of Technical Assumptions in PV Electricity Cost - Report on Current Practices in How Technical Assumptions are Accounted in PV Investment Cost Calculation," Solar Bankability WP3 Deliverable D3.1, Jul. 2016. K. Hunt, A. Blekicki, and R. Callery, "Availability of utility-scale photovoltaic power plants," in 2015 IEEE 42nd Photovoltaic Specialist Conference (PVSC), Jun. 2015, pp. 1–3, doi: 10.1109/PVSC.2015.7355976. M. García, J. A. Vera, L. Marroyo, E. Lorenzo, and M. Pérez, "Solar-tracking PV plants in Navarra: A 10 MW assessment," Progress in Photovoltaics: Research and Applications, vol. 17, no. 5, pp. 337–346, 2009, doi: 10.1002/pip.893. PERFORMANCE ANALYSIS

E. Muñoz-Cerón, J. C. Lomas, J. Aguilera, and J. de la Casa, "Influence of Operation and Maintenance expenditures in the feasibility of photovoltaic projects: The case of a tracking pv plant in Spain," Energy Policy, vol. 121, pp. 506–518, Oct. 2018, doi: 10.1016/j.enpol.2018.07.014



EXAMPLE IN SAM – RESULTS

Metric	Value
Annual energy (year 1)	20,961 kWh
Capacity factor (year 1)	22.2%
Energy yield (year 1)	1,945 kWh/kW
Performance ratio (year 1)	0.90





POA front-side shading loss POA front-side soiling loss POA front-side reflection (IAM) loss DC module deviation from STC DC inverter MPPT clipping loss DC mismatch loss DC diodes and connections loss DC wiring loss DC tracking loss DC nameplate loss DC power optimizer loss DC performance adjustment loss AC inverter power clipping loss AC inverter power consumption loss AC inverter night tare loss AC inverter efficiency loss AC wiring loss Transformer loss percent AC performance adjustment loss

Summary Data tables Losses Graphs Time series Profiles Statistics Heat map PDF / CDF Notices



Summary Data tables Losses Graphs Time series Profiles Statistics Heat map PDF / CDF Notices



EXAMPLE IN SAM - RESULTS

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This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 792059

Thank you for your attention!

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EXTRA INFORMATION



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Moser D. et al. "Uncertainties in Yield Assessments and PV LCOE," Nov. 2020. IEA-PVPS T13.



ANALYSIS P50-P90-PXX

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- Probability analysis of having a minimum yearly production.
- P90 data is the minimum PV production that we will get with 90% probability.
- Mainly required by banks and investment firms.
- Main contribution will be the uncertainty and variability of the meteo data. But other uncertainties should be considered.
- It is also used in wind farms. Not specific of PV.
- TMY weather files includes 1-year hourly data choosing "typical" months to represent the long-term properties. TMY files are not valid for P50-P90 analysis.
- It is required several years weather data to ensures that the performance prediction accounts for potential worst-case years.



https://www.pvsyst.com/help/p50_p90evaluations.htm

https://www.pvsyst.com/help/meteo_notes_annual_variability.htm A. Dobos, P. Gilman. P50/P90 Analysis for Solar Energy Systems Using the System Advisor Model. 2012 World Renewable Energy Forum. https://www.nrel.gov/docs/fy12osti/54488.pdf



HOW MANY YEARS OF METEO DATA?



- Uncertainty in yield assessment can be reduced by decreasing uncertainty of irradiation.
- The more years of irradiation data, more uncertainty reduction
- Reference scenario is defined for the case of 20 years of satellite data on the horizontal plane (20 y sat GHI DiffHI)
- P90 values which could be up to 20 % higher for the case with reduced uncertainty



Figure 16: Reduction of uncertainty in case of improvement of the quality of irradiation source. Taken from [47], courtesy of Eurac Research



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• Uneven terrain represent both a construction and yield assessment challenge



Figure 13 - Distribution of mounting structure tilt (left) and azimuth (right) angles for a PV system installed on a site characterized by uneven terrain





THE BIFACIAL CASE



- Challenge of calculating rear side irradiance
- Two main approaches: ray tracing and view factor
- Published errors compared to in-plane irradiance measurements range from 5 % to 40 % showing that both methods have high accuracy potential if appropriately implemented.
- Challenging task to obtain such measurements as albedo
- Albedo variations:
 - o intra-day due to anisotropic ground reflectivity
 - seasonal variations due to changing vegetation, ground moisture or snow cover
- Albedo difference of 0.1 results in an approx. 1 % bifacial gain difference
- Non-uniform rear irradiance is inevitable in bifacial PV systems







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THE FLOATING PV CASE

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- Two benefits: mitigation of land-use conflicts and improved energy performances
- Lack of simulation tools specific to the FPV
- Most installers/researchers are still using conventional simulation tools that do not account for the PV-water interaction and cooling effect caused
- The temperature model should be changed
- There are temperature models depending on whether the FPV system is on a lake/river or on seawater



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- Two targets: electricity production + crop production
- This application is still under study
- Impact of lower irradiance received by the crops?
- Potential protection of the crops by the PV modules (hailstorms)?
- Optimal inclination/tracking of the PV modules
- Performance depends on the type of plants/crops
- No specific software currently available
- Currently, common PV tools are being used.



BayWa r.e.



METEO INPUTS AND CLIMATE CHANGE



- Growing number of publications is using data from global climate models (GCM's) for predictions in years 2050 or 2100.
- They can predict either higher and lower irradiation depending on locations, but in general higher temperatures that impacts the PR.
- The regional analysis and more realistic PV performance modelling including uncertainties need further exploration.



J. Ascencio-Vásquez et al. "Methodology of Köppen-Geiger-Photovoltaic climate classification and implications to worldwide mapping of PV system performance," Solar Energy, vol. 191, pp. 672–685, Oct. 2019, doi: 10.1016/j.solener.2019.08.072.

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GLOBAL OPTIMIZATION OF INTEGRATED PHOTOVOLTAIC SYSTEM FOR LOW ELECTRICITY COST



This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 792059

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