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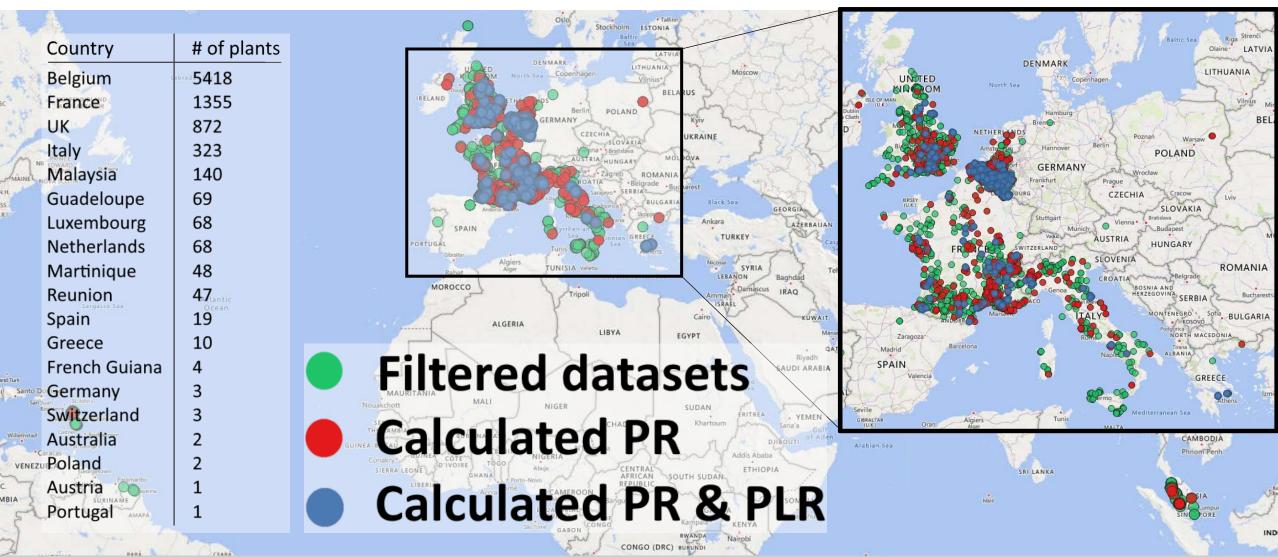
Evaluation of PV system data collected in COST Action Pearl PV Database –

Analysis of PR, Yield & PLR of a large fleet of PV systems

Sascha Lindig Eurac Research



Overview PV systems



02/12/21

Initial Data Quality Grading

Letter Grade	Outliers [%]	Missing percentage [%]	Longest gap [days]
Α	Below 10	Below 10	Below 15
В	10 to 20	10 to 25	15 to 30
С	20 to 30	25 to 40	30 to 90
D	Above 30	Above 40	Above 90

Pass/fail criteria	Time series > 24 months => PASS

[1] S. Lindig, et al., "International collaboration framework for the calculation of performance loss rates: Data quality, benchmarks, and trends (towards a uniform methodology)," *Progress in Photovoltaics: Research and Applications*, vol. 29, no. 6, pp. 573–602, 2021.

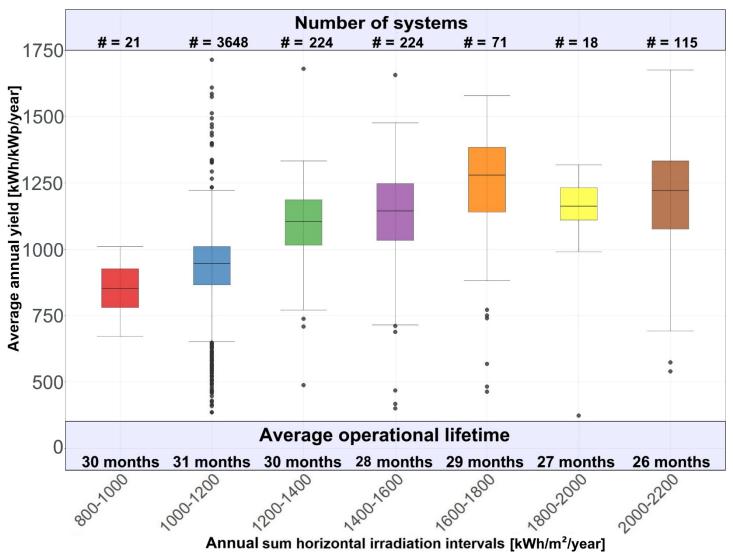
Initial Data Quality Grading

Letter Grade	Outliers	Missing percentage	Longest gap
Α	8,367	5,773	7,655
В	0	2,216	280
С	0	164	291
D	0	214	141
D	Above 30	Above 40	Above 90

Pass/fail criteria P: 4,323 F: 4,044

[1] S. Lindig, et al., "International collaboration framework for the calculation of performance loss rates: Data quality, benchmarks, and trends (towards a uniform methodology)," *Progress in Photovoltaics: Research and Applications*, vol. 29, no. 6, pp. 573–602, 2021.

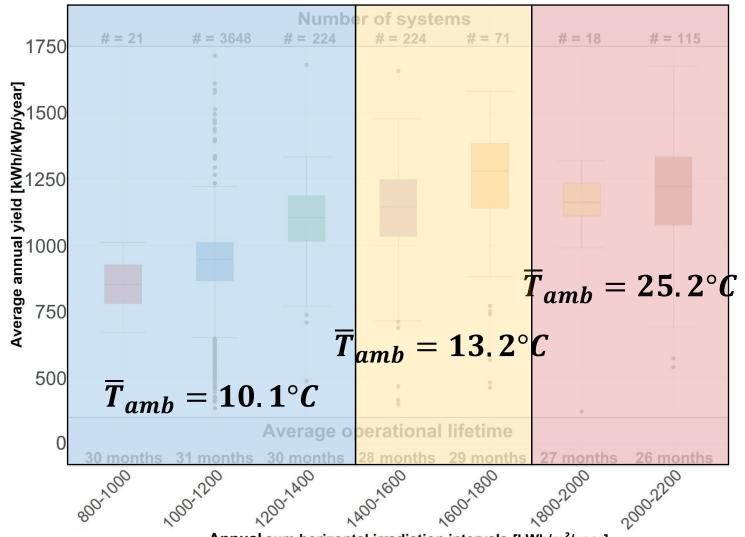
Energy Yield



$$\overline{yield} = 954.9 \frac{kWh}{kWp}$$
 per year

$$\widetilde{yield} = 961.5 \frac{kWh}{kWp}$$
 per year

Energy Yield

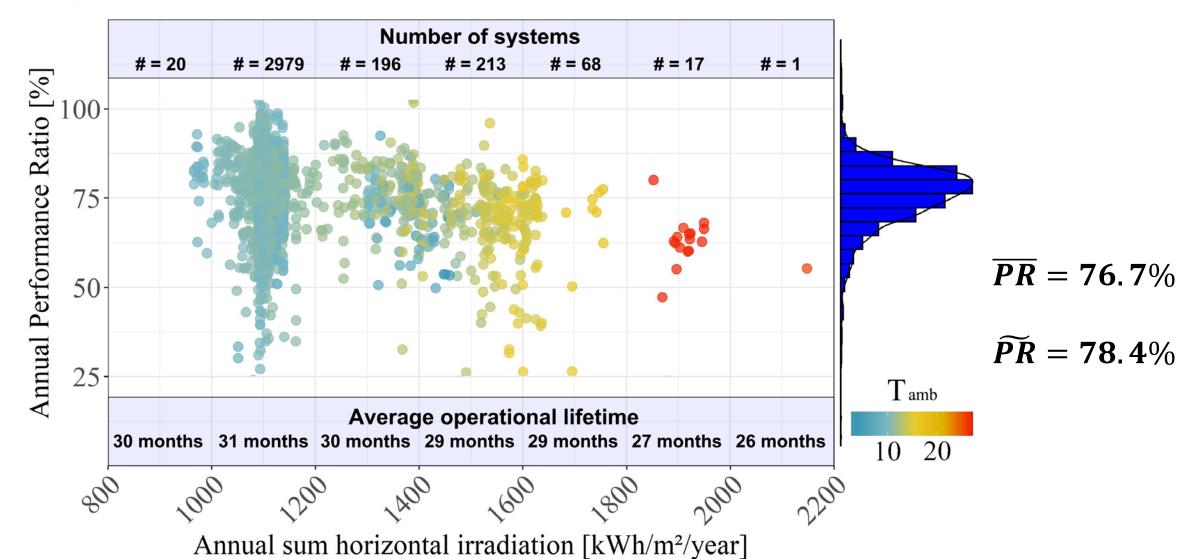


$$\overline{yield} = 954.9 \frac{kWh}{kWp}$$
 per year

$$\widetilde{yield} = 961.5 \frac{kWh}{kWp}$$
 per year

Annual sum horizontal irradiation intervals [kWh/m²/year]

Performance Ratio



0. Exploratory data analysis for data quality & grading

0.a Data availability

 P_{mpp} ; G_{POA} ; T_{mod} ; T_{amb} ; wind speed

0.b Data quality assessment & grading

Outliers, missing, gaps

1. Input data cleaning/filtering

1.a Data assembly

Data imputation, Timestamp validation

1.b Filter application

 P_{mpp} ; G_{POA} ; T_{mod} ; PR; $clear\ sky$

2. Performance metric selection, corrections & aggregation

2.a Perf. Metric

Predicted Power, Performance Ratio

2.b Temp. corrections

IEC61724-1, UTC

2.c Data aggregation

Daily; weekly; monthly; yearly

3. Timeseries feature corrections

3.a Seasonal decomp.

CSD; STL; HW

3.b Imputation of Power *P*; *PR*

3.c Outlier Removal Z-score; Interquartile

Z-score; Interquartile range 4. PLR Result

4.a Statistical models

Regression; YoY; CPLR

4.b \overline{PLR} Determination

4.c Assess Confid. Int.

 P_{mpp} – PV power in the maximum power point

 G_{POA} – irradiance in the plane of array

 T_{mod} – module temperature

 T_{amb} – ambient temperature

PR – Performance Ratio

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Data input & Data quality analysis

- 10 min power time series
- Hourly ERA5 in-plane irradiation²

Input data cleaning & filtering

• SCSF³ : Clear-sky filter

• YoY : Filter provided by RdTools^{4,5}

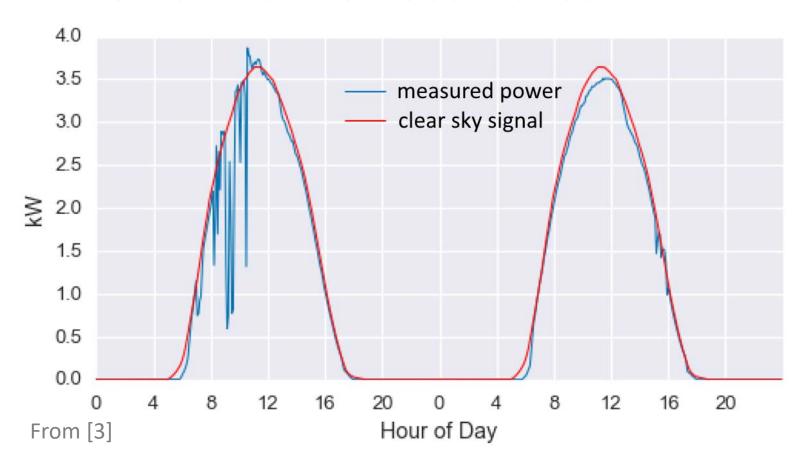
• STL : Tailored threshold & statistical filter

Selection performance metric & statistical model

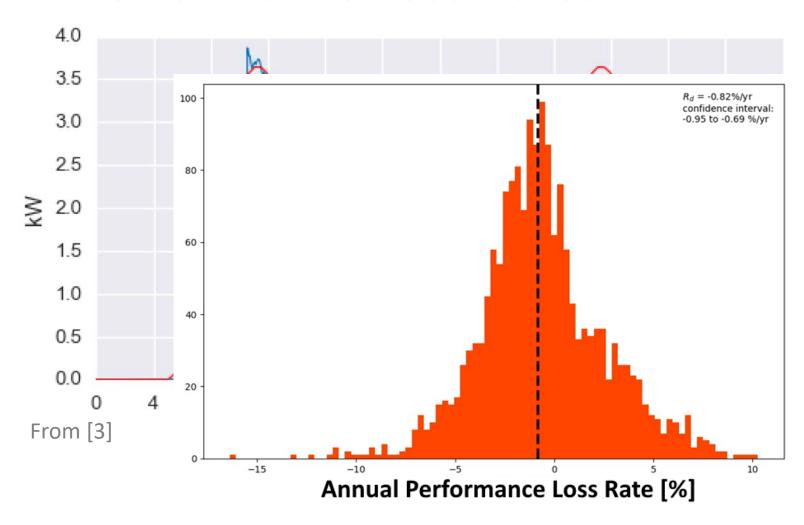
- Daily clear-sky energy : SCSF³
- Daily normalized energy: YoY⁵
- Monthly PR : STL⁶
- [2] Copernicus Climate Change Service (C3S) ERA5: Fifth generation of ECMWF atmospheric reanalyses of the global climate. Copernicus Climate Change Service CDS 2017.
- [3] B. Meyers, et al, "Signal Processing on PV Time-Series Data: Robust Degradation Analysis Without Physical Models," IEEE Journal of Photovoltaics, 2019.
- [4] RdTools, "Version 2.0.5." [Online]. Available: https://github.com/NREL/rdtools
- [5] D. Jordan, et al, "Robust PV Degradation Methodology and Application," IEEE Journal of Photovoltaics, 2017.
- [6] R. B. Cleveland et al., "STL: A Seasonal-Trend Decomposition Procedure Based on LOESS," Journal of Official Statistics, 1990.







[3] B. Meyers, et al, "Signal Processing on PV Time-Series Data: Robust Degradation Analysis Without Physical Models," IEEE Journal of Photovoltaics, 2019.



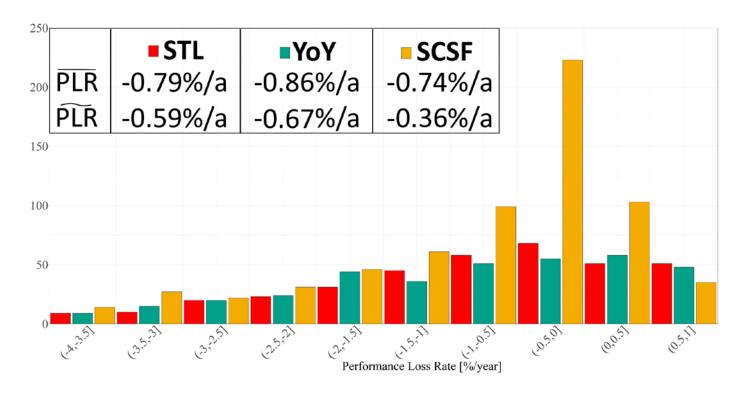
[3] B. Meyers, et al, "Signal Processing on PV Time-Series Data: Robust Degradation Analysis Without Physical Models," IEEE Journal of Photovoltaics, 2019.

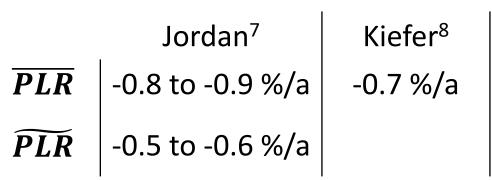


	Filter			
Statistical method	G_{POA} [W/m²]	T_{mod} [°C]	Power	Performance Ratio
SCSF	Strict clear-sky filter			
YoY	200-1200	-50 – 110	P > 0	
STL	100-1200		$(0.01 - 1.05)* P_{nom}$	$\pm 2\sigma$ around daily mean PR

	SCSF	YoY-ERA5	STL-ERA5
All systems	8,367		
FINAL	661	361	366

Minimum 3 years of data -4%/a < PLR < 1%/a

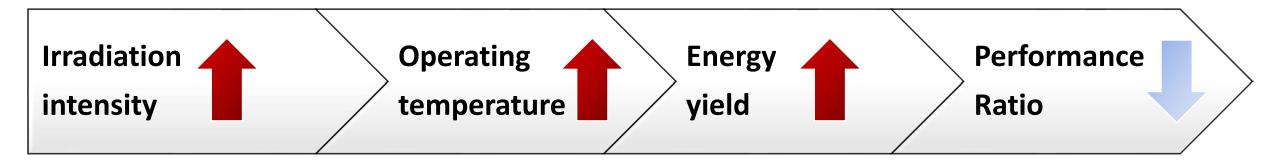




[7] D. C. Jordan, et al, "Compendium of photovoltaic degradation rates," *Progress in Photovoltaics Research and Application*, vol. 24, no. 7, pp. 978-980, 2016.
[8] K. Kiefer, et al, "Degradation in PV Power Plants: Theory and Practice," in *36th EU PVSEC, Marseille*, 2019.

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Conclusion



PLR is an important parameter to assess the health status of a PV system

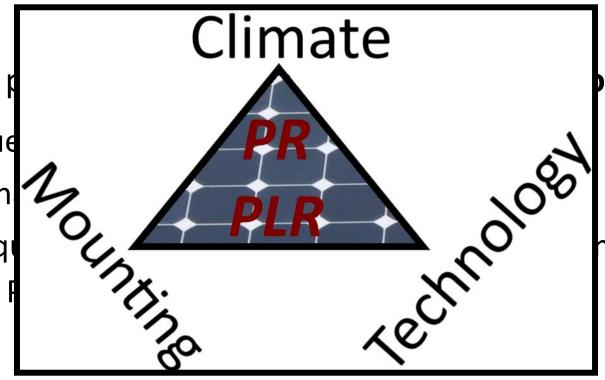
- Calculating PLR values is not straightforward
 - Many variables have to be considered
 - The length and quality of the PV system time series is the most important characteristic of PV performance analyses

Conclusion

Irradiation Operating Energy Performance Intensity Yield Ratio

PLR is an important p

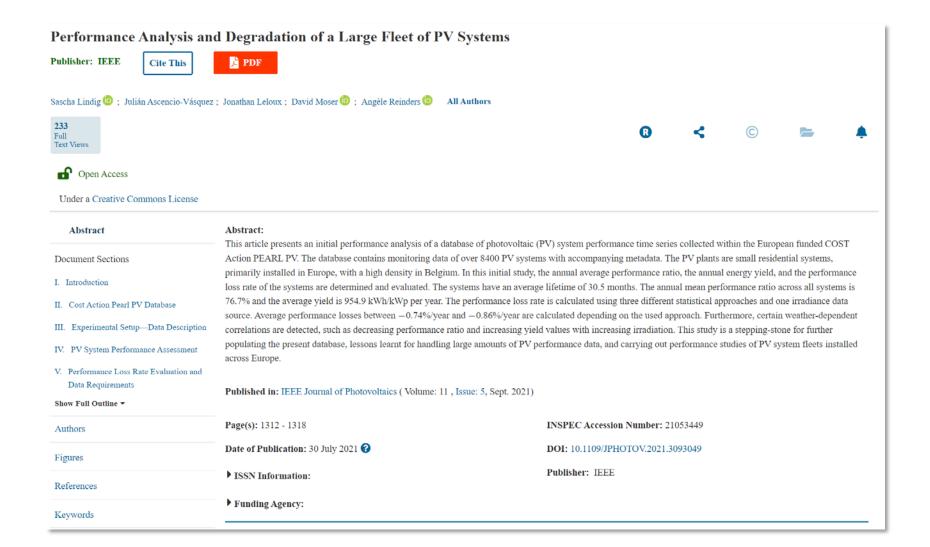
- Calculating PLR value
 - Many variables h
 - The length and questions
 characteristic of I



of a PV system

host important

Performance Analysis and Degradation of a Large Fleet of PV Systems



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Thank you for your attention

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Sascha.Lindig@eurac.edu