

Service Lifetime Prediction of PV Modules and Systems: Progress of the SOLAR-TRAIN Project

 I. Kaaya¹, S. Lindig², N. Hrelja³, G. Oviedo Hernández⁴, F. Mariottini⁵, D. Moser², K.-A. Weiss¹, M. Van Iseghem³, P. V. Chiantore⁴ and T.R. Betts⁵
¹ Fraunhofer Institute for Solar Energy Systems ISE, Heidenhofstr.2, 79110, Freiburg, Germany

⁴ BayWa r.e. Operation Services S.r.l., 00139, Rome, Italy

² EURAC Research - Institute for Renewable Energy, 39100, Bolzano, Italy

⁵ CREST-Loughborough University, LE11 3TU, Loughborough, United Kingdom

³ Electricité de France (EDF), 77250, Moret-sur-Loing, France

INTRODUCTION

The project SOLAR-TRAIN aims to develop novel and validated models for the service life time and energy yield prediction of PV modules and systems. PV modules' & systems' performances are being investigated along the entire modelling chain: climatic degradation factors, analysis of degradation and failure modes and evaluation of polymeric materials. This work presents an overview of the current start-of-the-art and some preliminary results on the development of service lifetime prediction models for PV modules & systems.

DEGRADATION RATE EVALUATION OF A PV SYSTEM

Poly-crystalline Si system:

- Nominal power: 4.2kWp
- Parameter: P_{MPP} ; I_{MPP} ; V_{MPP}
- In operation since 2011

Weather station:

- Temperature
 - Ambient & module
- Irradiance
 - Plane of array, global horizontal & diffuse



Fig. 1 Studied PV system located in Bolzano (North of Italy).

BASED ON PREVIOUS STUDIES

Comparison of statistical methods to calculate Performance Loss Rates [1]

5 different statistical models for the calculation of Performance Loss Rates (PLR) were investigated on different PV systems & technologies. Seasonal & trend decomposition using LOESS (STL) as well as auto-regressive integrated moving average (ARIMA) were found to have the lowest uncertainties and consistent final values.

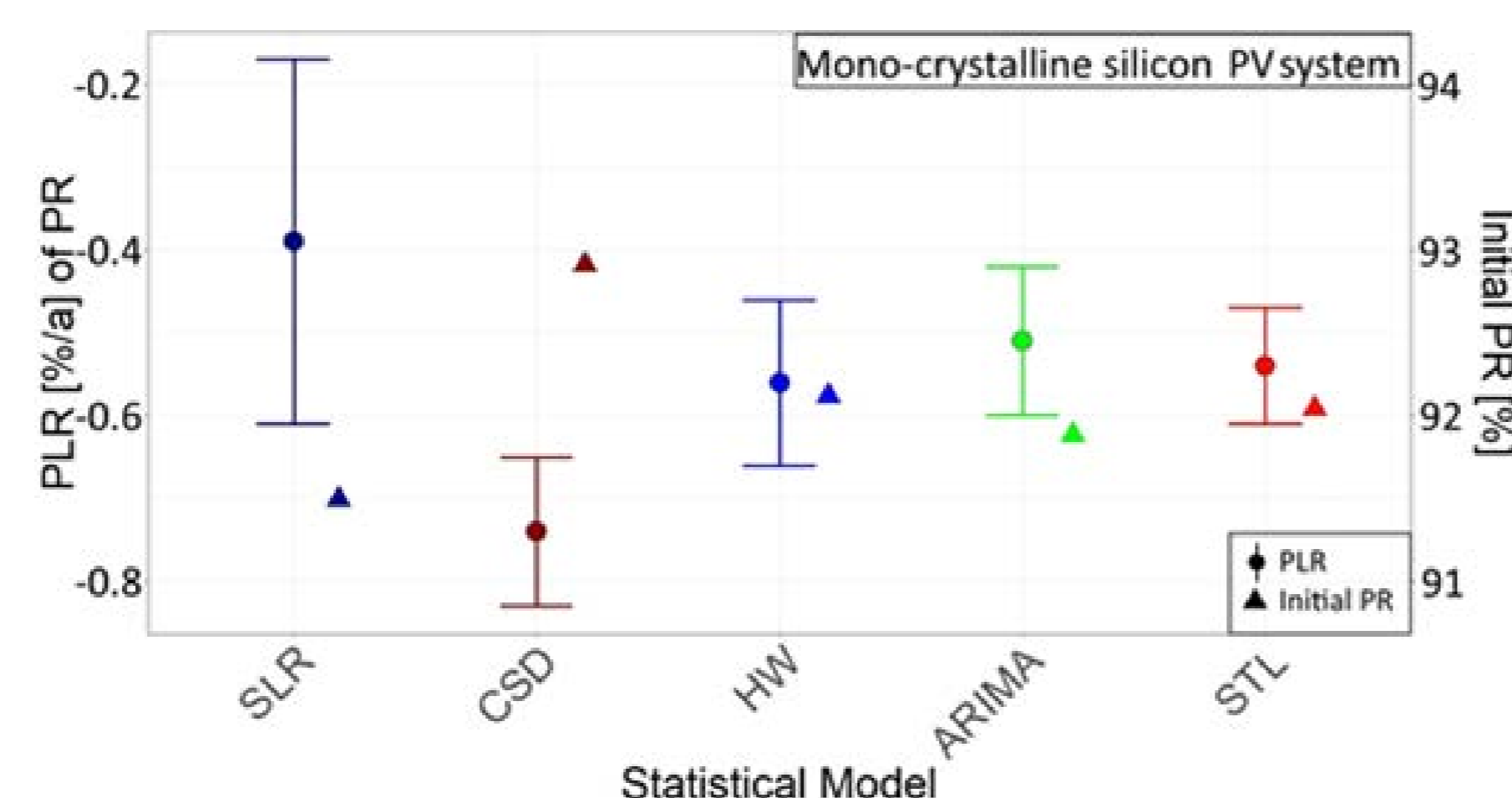
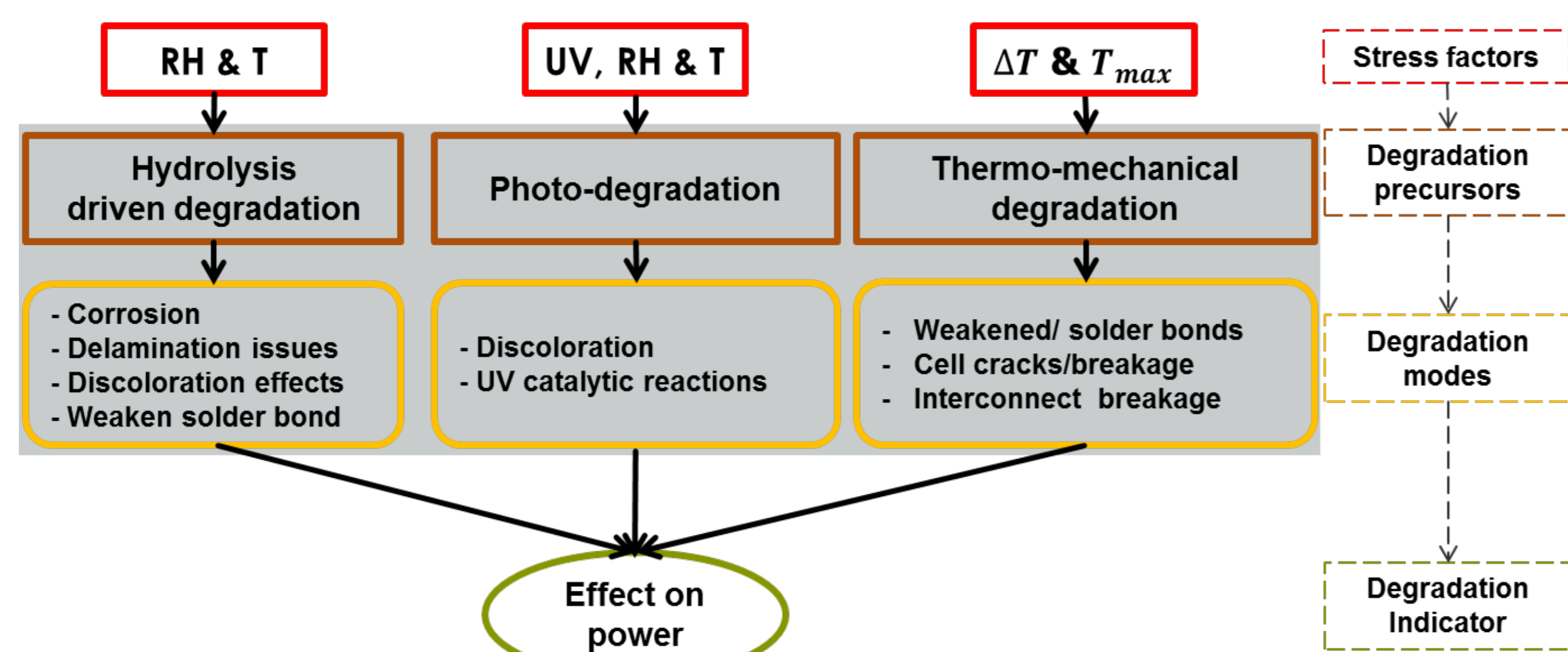


Fig. 2 Comparison of statistical models on PR-data of mc-Si system, circles represent PLR (primary axis), triangles represent initial PR.

Climate based modelling [2]



A total degradation rate (k_T) was proposed assuming 3 degradation precursors:

$$k_T = (1 + k_h(T, RH)) * (1 + k_p(UV, T, RH)) * (1 + k_{Tm}(\Delta T, T_{max})) - 1$$

REFERENCES

- [1] S. Lindig, I. Kaaya, K. Weiß, D. Moser, and M. Topic, "Review of Statistical and Analytical Degradation Models for Photovoltaic Modules and Systems as Well as Related Improvements," IEEE J. Photovolt., vol. 8, no. 6, pp. 1773–1786, Nov. 2018
- [2] I. Kaaya, M. Koehl, A. P. Mehili, S. d C. Mariano, and K. A. Weiss, "Modeling Outdoor Service Lifetime Prediction of PV Modules: Effects of Combined Climatic Stressors on PV Module Power Degradation," IEEE J. Photovolt., pp. 1–8, 2019.
- [3] F. Mariottini, J. Zhu, T. R. Betts, and R. Gottschalg, "Evaluation of Uncertainty Sources and Propagation from Irradiance Sensors to PV Energy Production", PVSAT-14, 2018.

EFFECT OF FILTERING ON PLR EVALUATION

Filter	Irradiance	Other	$\overline{PR} \pm \sigma$	% Filtered
Positive Values	Remove NA & negative	Remove NA & neg Power	0.825 ± 0.141	55.28%
Clear sky instants		$0.75 < P/G < 1.25$	0.845 ± 0.071	65.3%
PR statistical	$500 < G < 1200$	$0.1 * P_{nom} < P_{DC} < 1.2 * P_{nom}$ $PR_{mode} \pm 2\sigma$	0.850 ± 0.049	83.4%

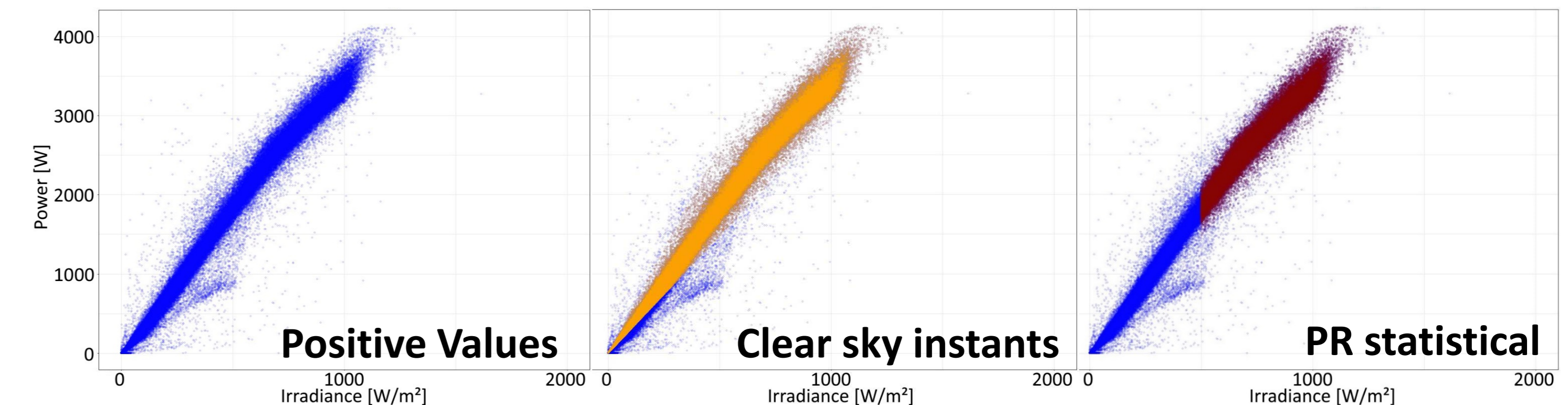


Fig. 3 Power against irradiance of investigated pc-Si system for 3 different filters tested.

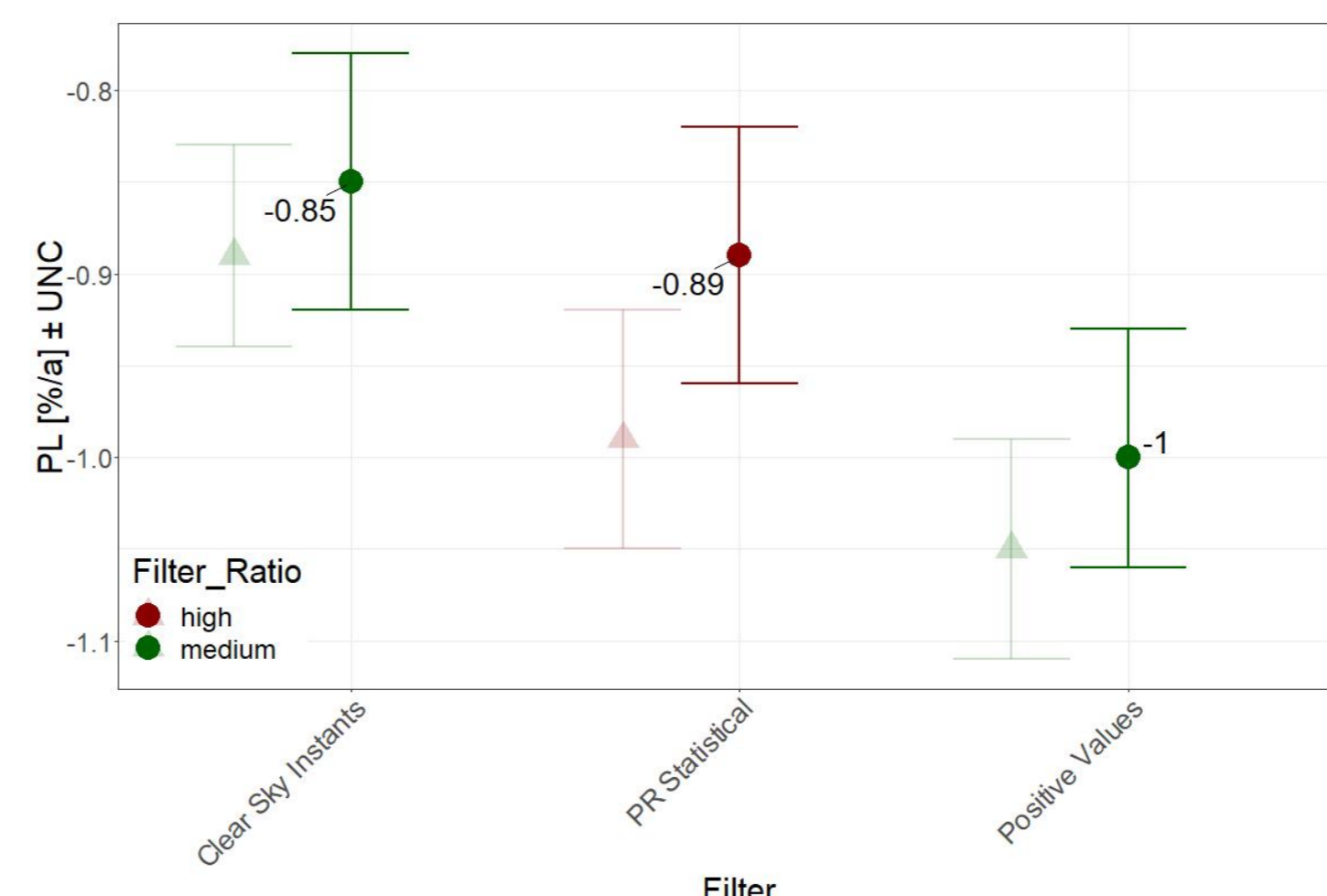


Fig. 4 PLR including related uncertainties (circle: T-corrected PR; triangle: PR).

PLR is between -0.8%/a to -0.9%/a while a degradation rate of -0.45%/a has been calculated considering climate effects only. The deviation between both values can be attributed to technical failures, which are not classified as a physical degradation of the system (e.g. soiling, shading, ...).

Climate based modelling:

Rate	k_h	k_p	k_{Tm}	k_T
Value	-0.030%/a	-0.080%/a	-0.301%/a	-0.45%/a

YIELD ASSESSMENT ACCURACY [3]

- Filter: physically possible (extremely rare) limits & theoretical sun path
- Clear sky days identification based on correlation with clear sky model (Perez)
- Uncertainty evaluation (JCGM 100:2008) from interpolated values based on calibration

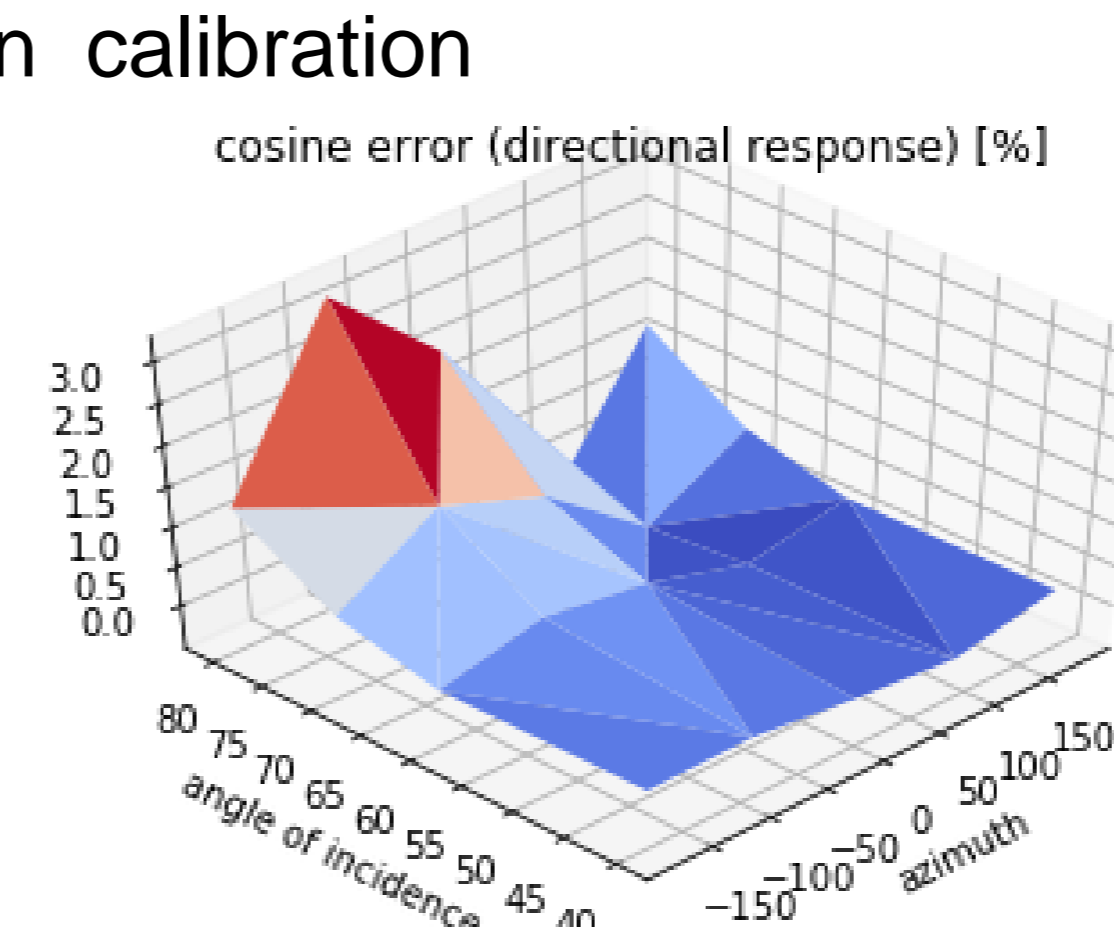


Fig. 5 Interpolation of cosine error of pyranometer based on response values extracted from the calibration certificate.

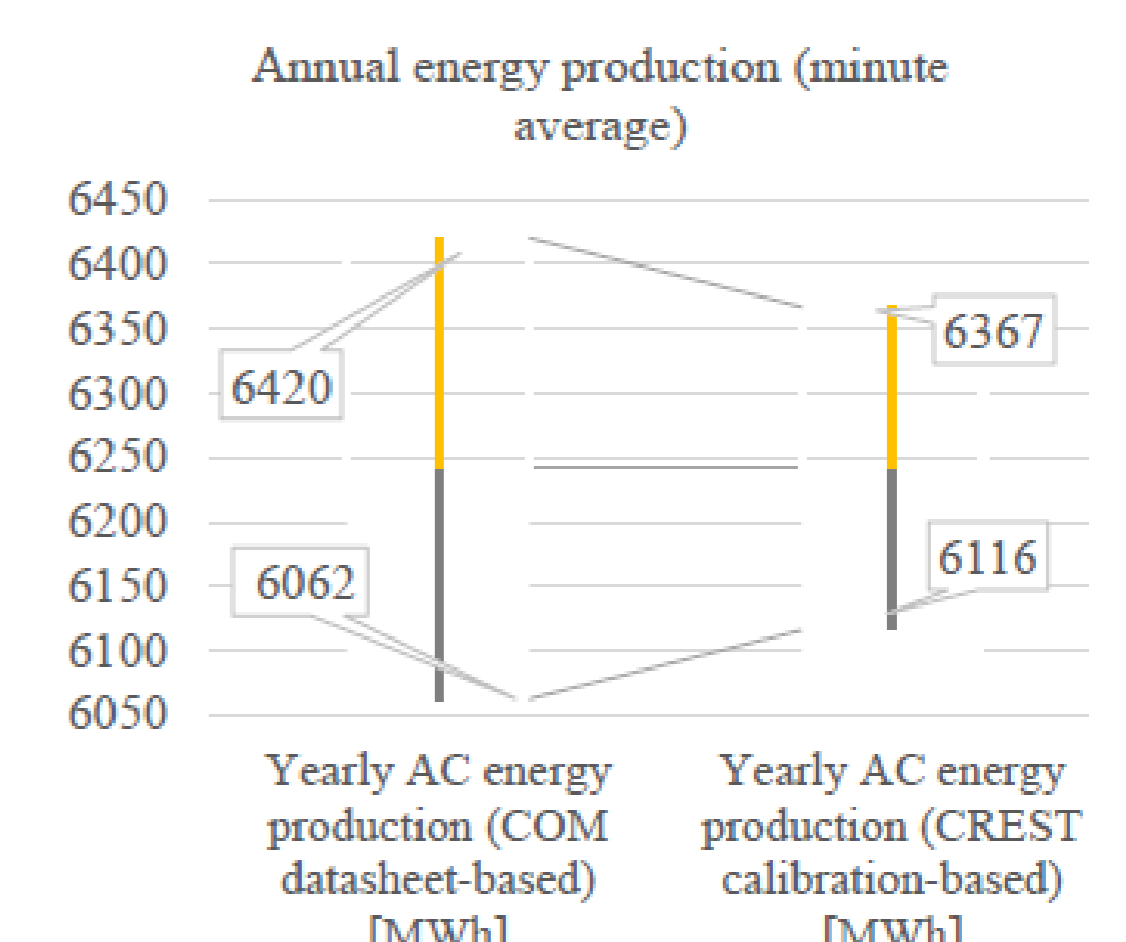


Fig. 6 Decrease of energy production uncertainty with help of calibration information in a 7389 kWp solar farm.