

An Interpretable AI Framework for Clear-Sky Detection in Photovoltaics Monitoring

Hugo Quest^{1*}, Bohan Li^{2,3}, Christophe Ballif^{1,3}, Antonin Faes^{1,3}, Alessandro Virtuani¹

1 – CSEM, Neuchâtel, Switzerland; 2 – North China Electric Power University, Beijing, China; 3 – EPFL, Neuchâtel, Switzerland

Summary

- Introduction of an interpretable AI framework for clear-sky detection (CSD) addressing the critical "black-box" problem in modern ML-based PV monitoring by integrating a SHAP-based diagnostic workflow.
- Final model achieves 97.3% accuracy with 1.99% FP rate, with reduced false alarms improving PLR calculation uncertainty.
- Identified error patterns used to develop real-time confidence metrics, allowing the system to flag ambiguous sky conditions where the CSD prediction may be less certain.

Context and Methods

- **Problem:** Clear-sky detection (CSD) often fails under haze or cloud enhancement, limiting PV data accuracy.
- **Solution:** use interpretable AI not as a "black box," but as a diagnostic tool to understand the physical reasons for model failure.

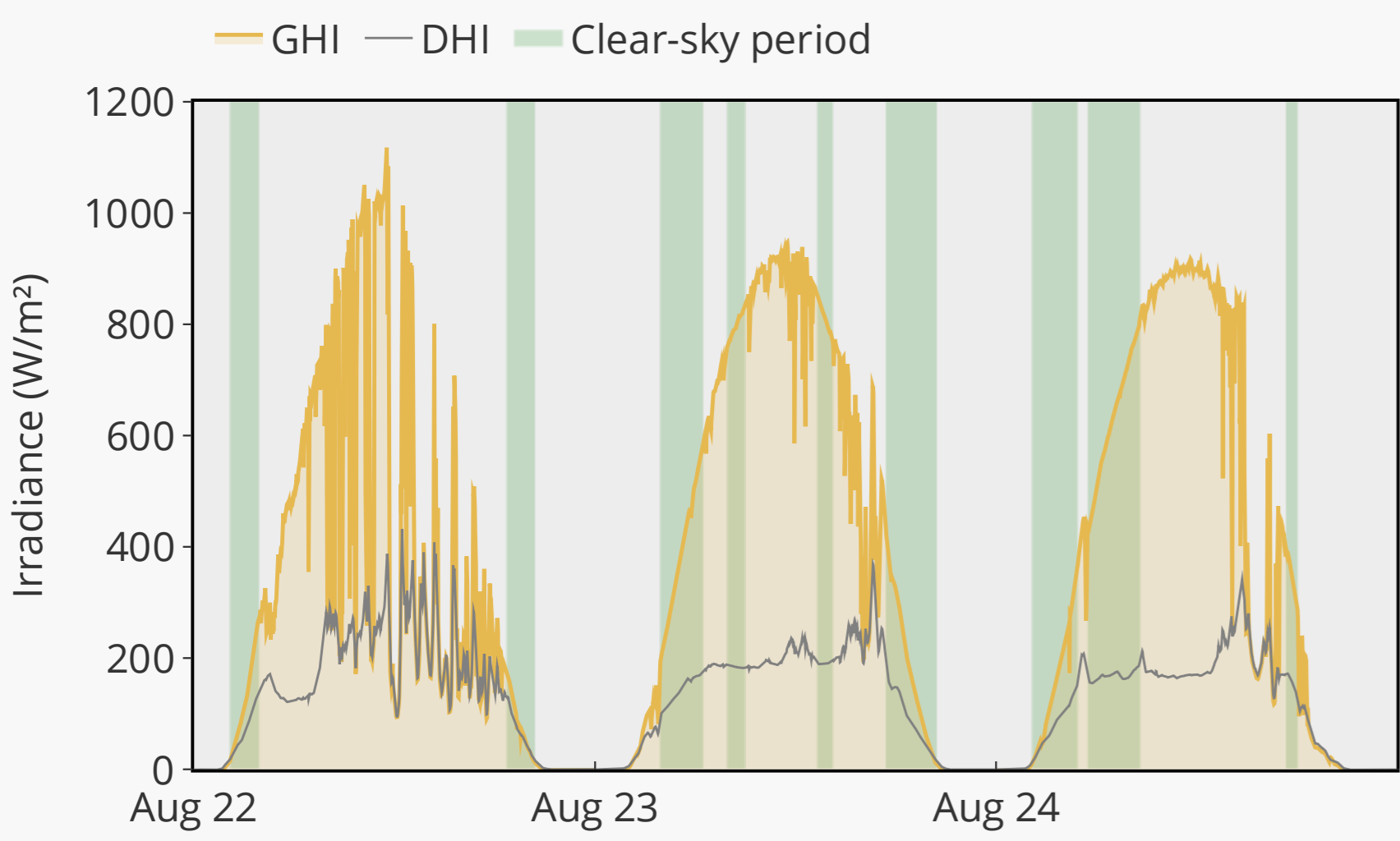


Fig. 1 | Example of Clear-Sky Detection (CSD).

Impact: Enables a *diagnose – refine* cycle: trace errors to physical causes and apply targeted, physics-based fixes.

Labelled dataset from Jordan & Hansen (2023)

“Diagnose-and-Refine” Framework

Workflow: Diagnose physical root-cause of errors → Refine model with physics-based insights (iterative)

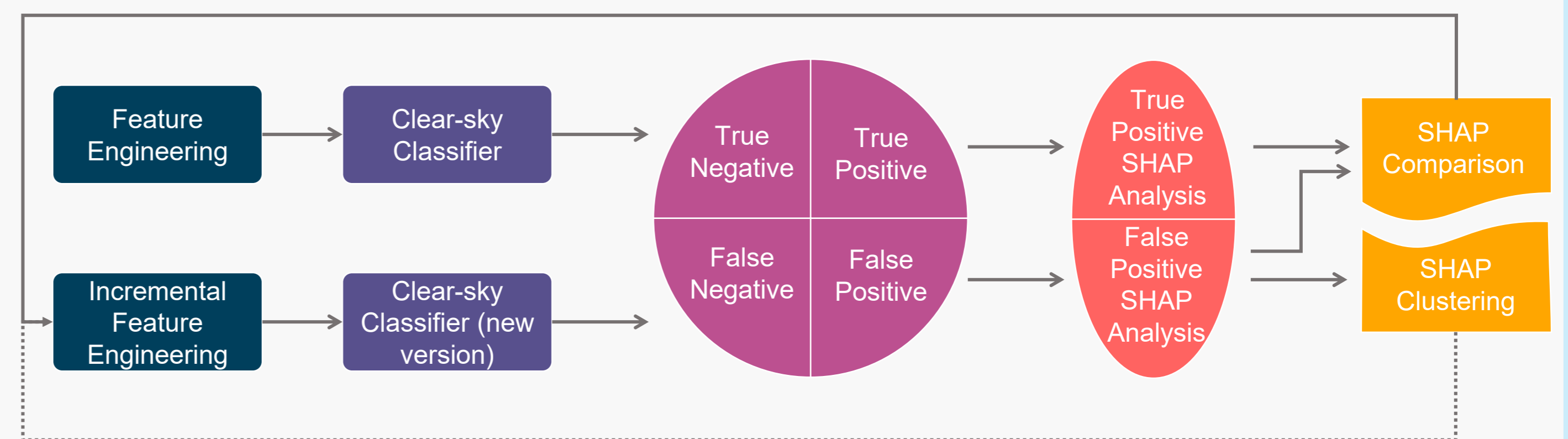


Fig. 2 | Interpretable CSD development framework

SHAP Comparison: Shows why the model failed by ranking feature impacts.

SHAP Clustering: Groups systematic error patterns for targeted fixes.

Interpretable AI for Deeper Insight

Using **Explainable AI (XAI)** to overcome the "black box" problem provides transparent insight into the model's decision-making.

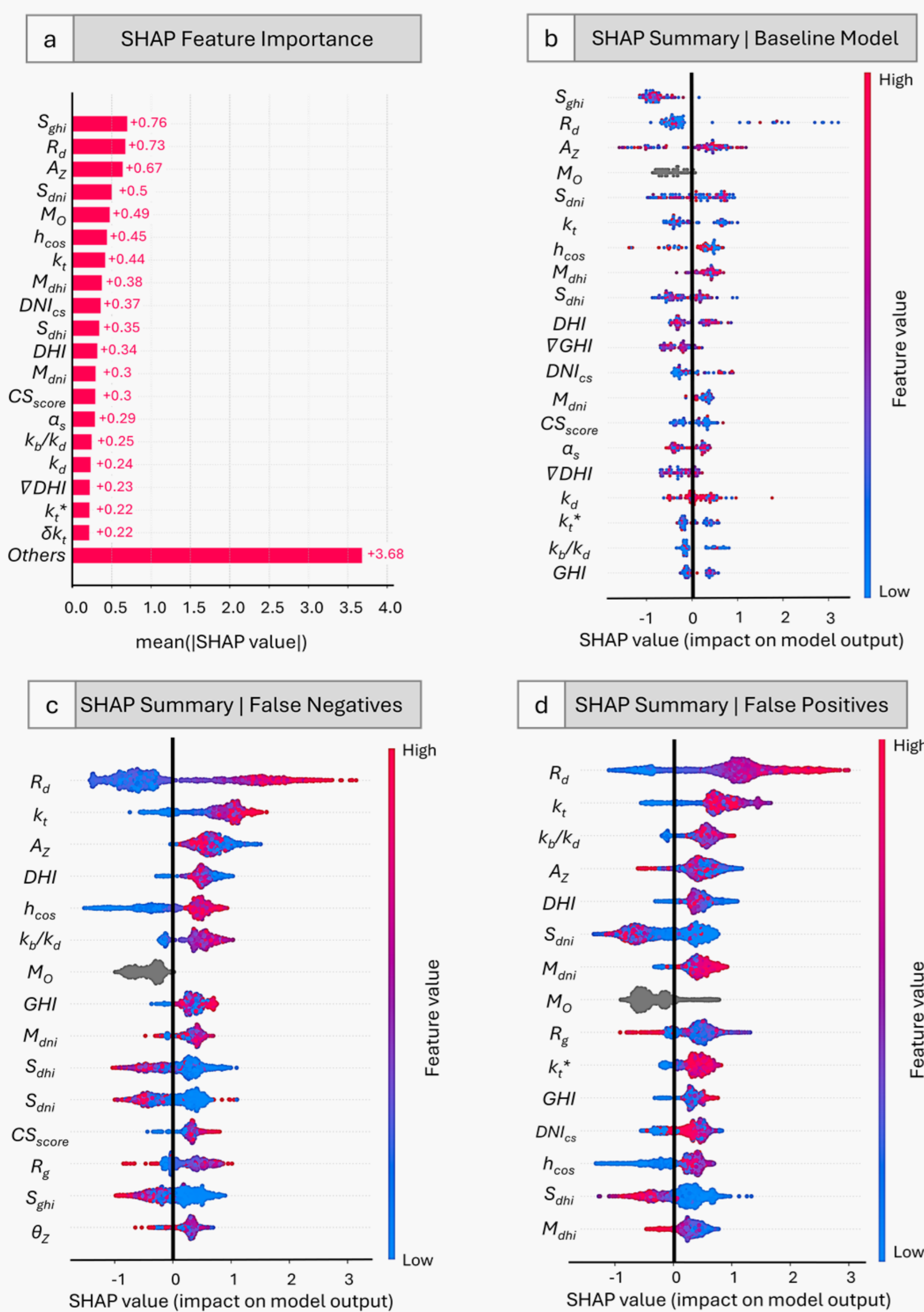


Fig. 3 | SHAP analysis: shows FPs stem from *cloud enhancement*, FNs from *haze*.

Pinpointing Failure Modes

Automatically categorise model errors, identifying "**cloud enhancement**" as the cause for >90% of false positives.



Fig. 4 | FP deep dive: clustering confirms *cloud enhancement* as main failure mode.

High-Impact Results & Application

Refined CSD model achieves **97.3%** accuracy, enabling more reliable fault diagnosis by significantly reducing false alarms.

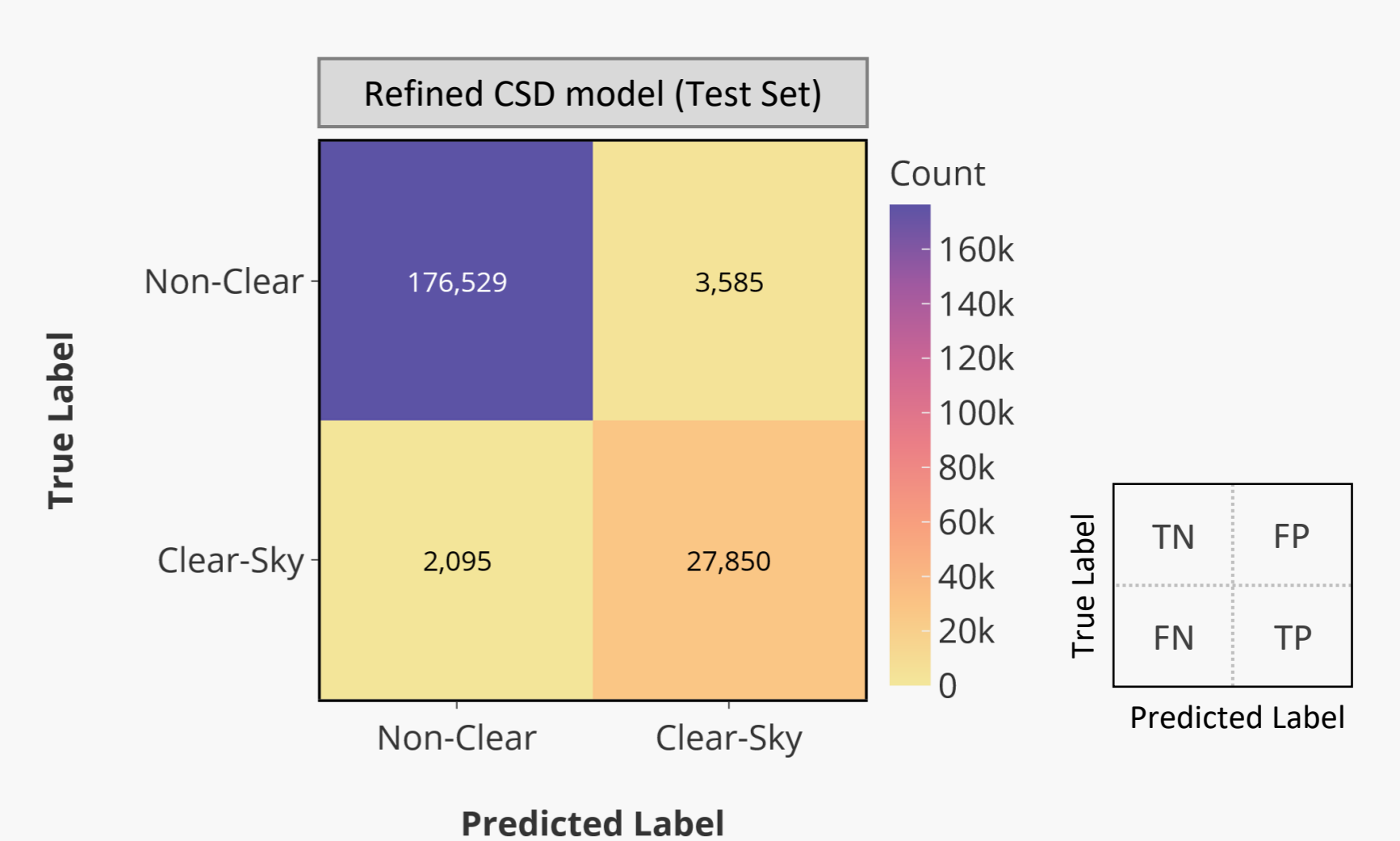


Fig. 5 | Confusion matrix: test performance of the refined CSD model.

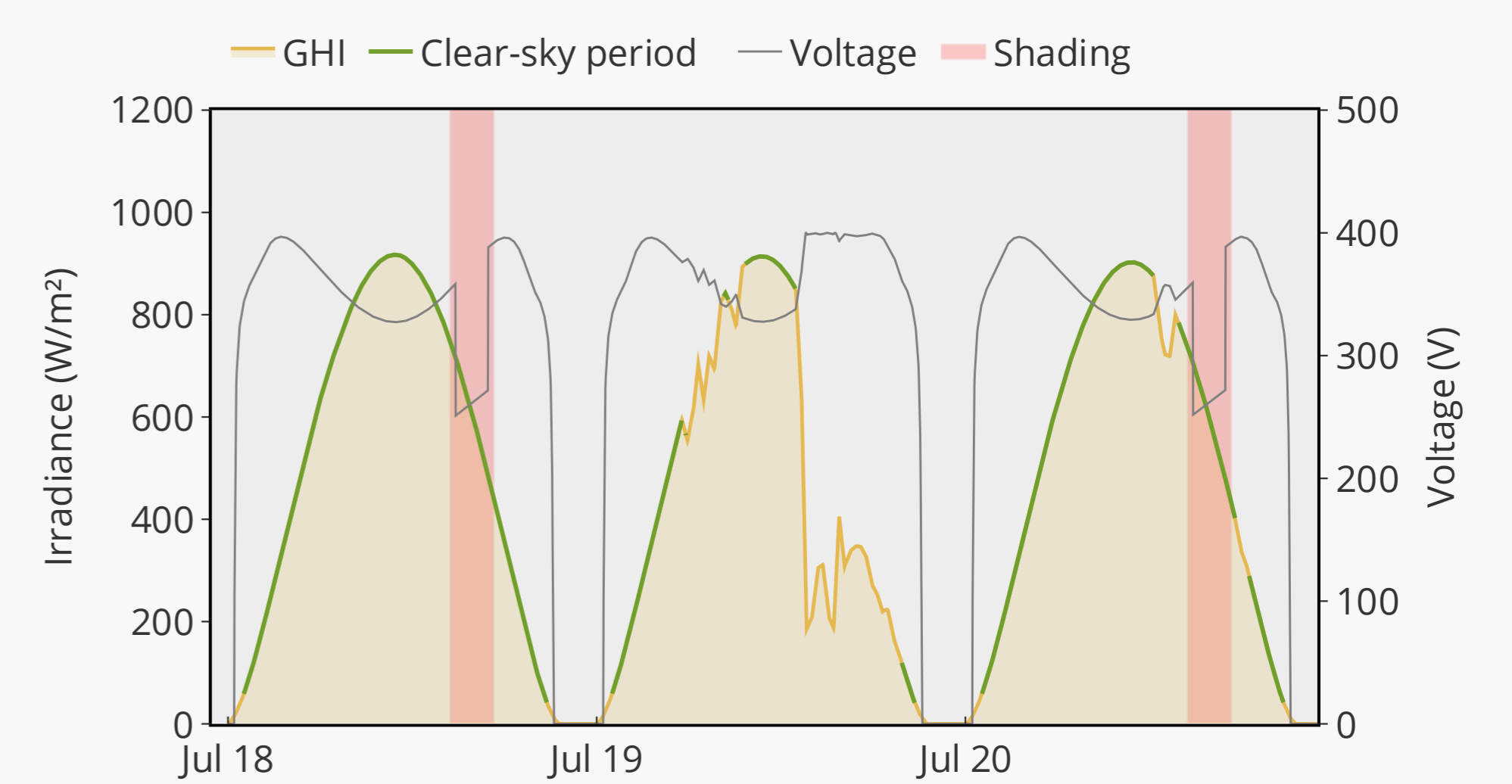


Fig. 6 | CSD for fault diagnosis: reveals voltage drops from bypass diode activation (shading).

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References

- [1] Quest et al. (2023), doi: 10.1002/pip.3762
- [2] Reno, M. J. et al. (2016), doi: 10.1016/j.renene.2015.12.031
- [3] Lundberg, S. M. et al. (2020), doi: 10.1038/s42256-019-0138-9
- [4] Holmgren, W. F. et al. (2018), doi: 10.21105/joss.00884
- [5] Jordan, D. C. et al. (2023), doi: 10.1016/j.renene.2023.04.035

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info@csem.ch • www.csem.ch

