



Solar irradiance and ramp events forecasting based on all-sky imagers

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This study was conducted under the framework of **IEA PVPS Task 16**

Aim



To evaluate the current and emerging solar forecasting techniques

Our task

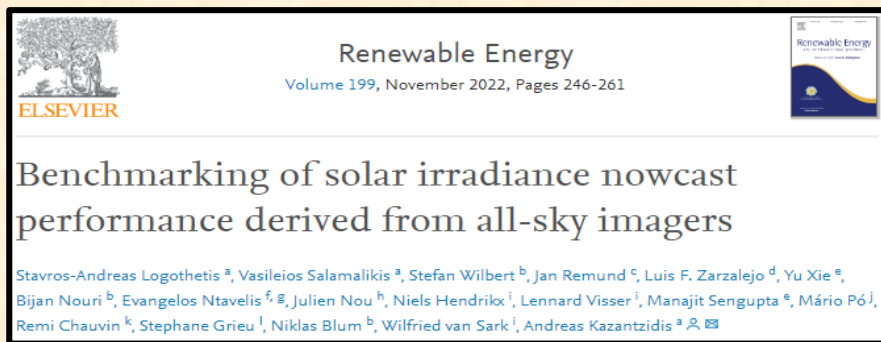


To qualify the current state-of-the-art of **all-sky imager** (ASI)-based solar nowcasting

Paper 1



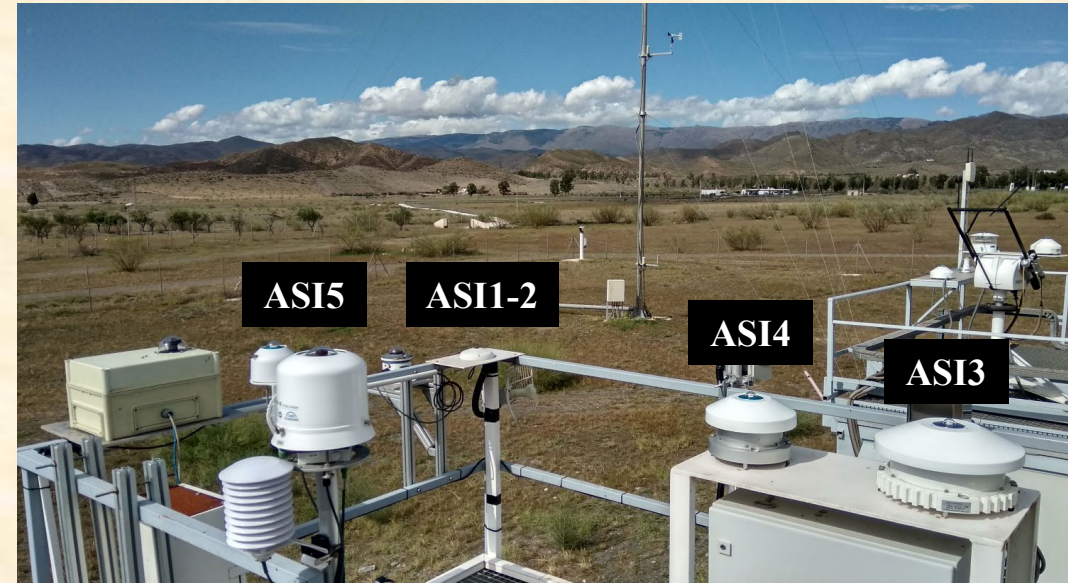
Paper 2



☒ Details of the campaign days and measurements

Information about the 5 ASI systems

Experimental setup at METAS at CIEMATs's PSA



Partners	Cameras Type	Number of Cameras	Image Acquisition time	Cloud detection method	Nowcast method
ASI1	Mobotix Q25 6MP	2	30s	Clear sky library (Wilbert et al., 2016)	Nouri et al. 2021
ASI2	Mobotix Q25 6MP	2	30s	Neural Network (Fabel et al., 2021)	Nouri et al. 2021
ASI3	Industrial	1	10s	N/A	Hybrid model (Knowledge/Neural Network)
ASI4	EKO 16	1	15s	Red/Blue Ratio (Ghonima et al., 2012)	LSTM (recurrent neural network) (Hendrikx et al., 2021)
ASI5	Prototype	1	10s	Deep learning (Pierer and Remund, 2019)	

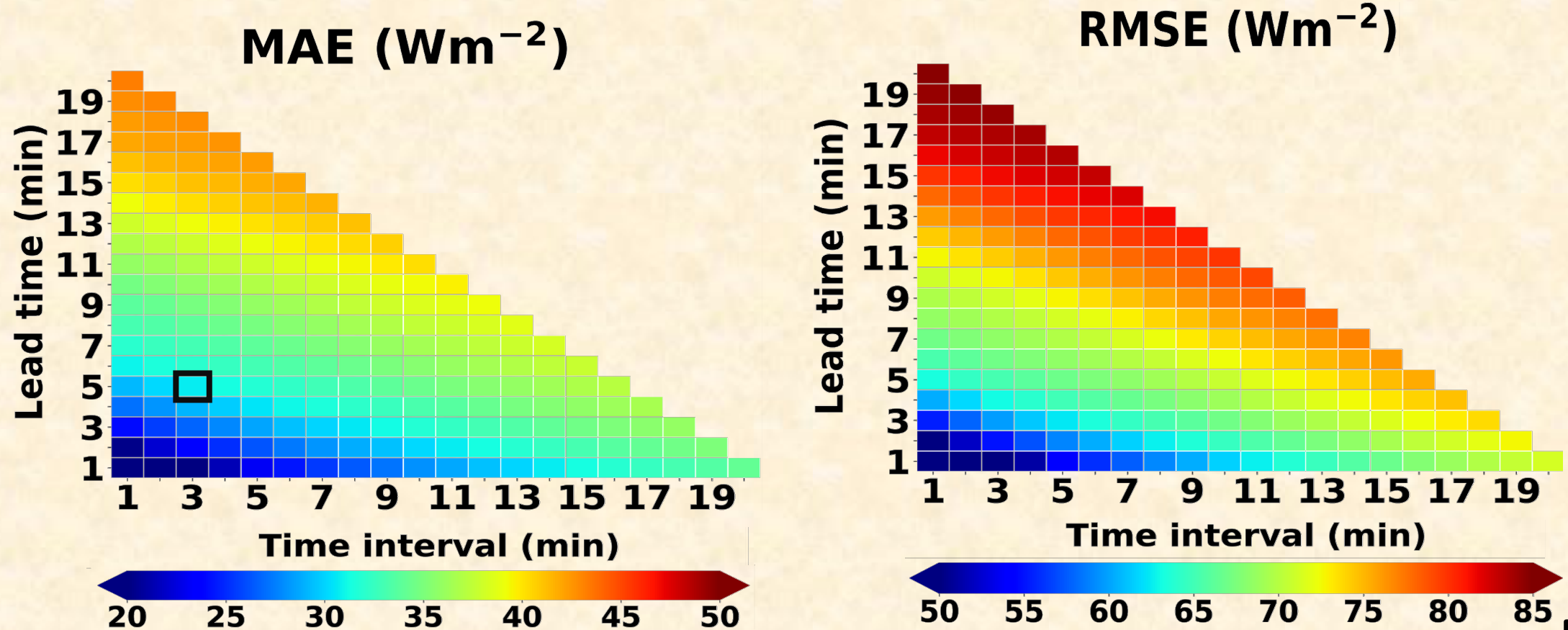
GHI nowcast systems

- 1) ASI: 4 camera setups, 5 nowcasting methods (ASI1–5).
- 2) Two persistence algorithms: PSPI and smart persistence.

❖ Time horizon = Nowcasts up to 20 min with 1 min step

❖ Time period = from August to November 2019.

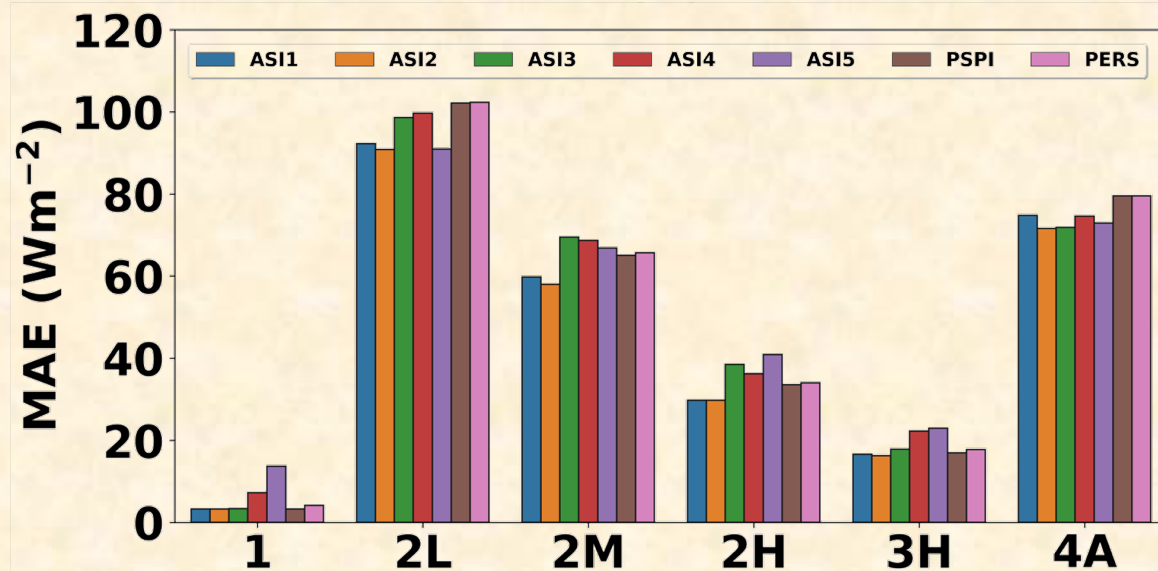
☒ ASI2 system nowcast performance at various time intervals



The vertical axis shows starting forecast minute (lead time).
 The horizontal axis shows the number of minutes included (time interval).



☒ Nowcast performance at different cloud conditions



Summary of the algorithms with the **best performance** for each cloud cluster.

	Cloud Cluster					
	1	2L	2M	2H	3H	4A
Metric						
MAE	PSPI	ASI2	ASI2	ASI1	ASI2	ASI2
RMSE	ASI2	ASI5	ASI2	ASI2	ASI2	ASI3

Cloud clustering of campaign days

All campaign days were flagged in the following **6 cloud** classes:

1: Cloud-free (or almost cloud-free)

Scattered/broken cloudiness with:

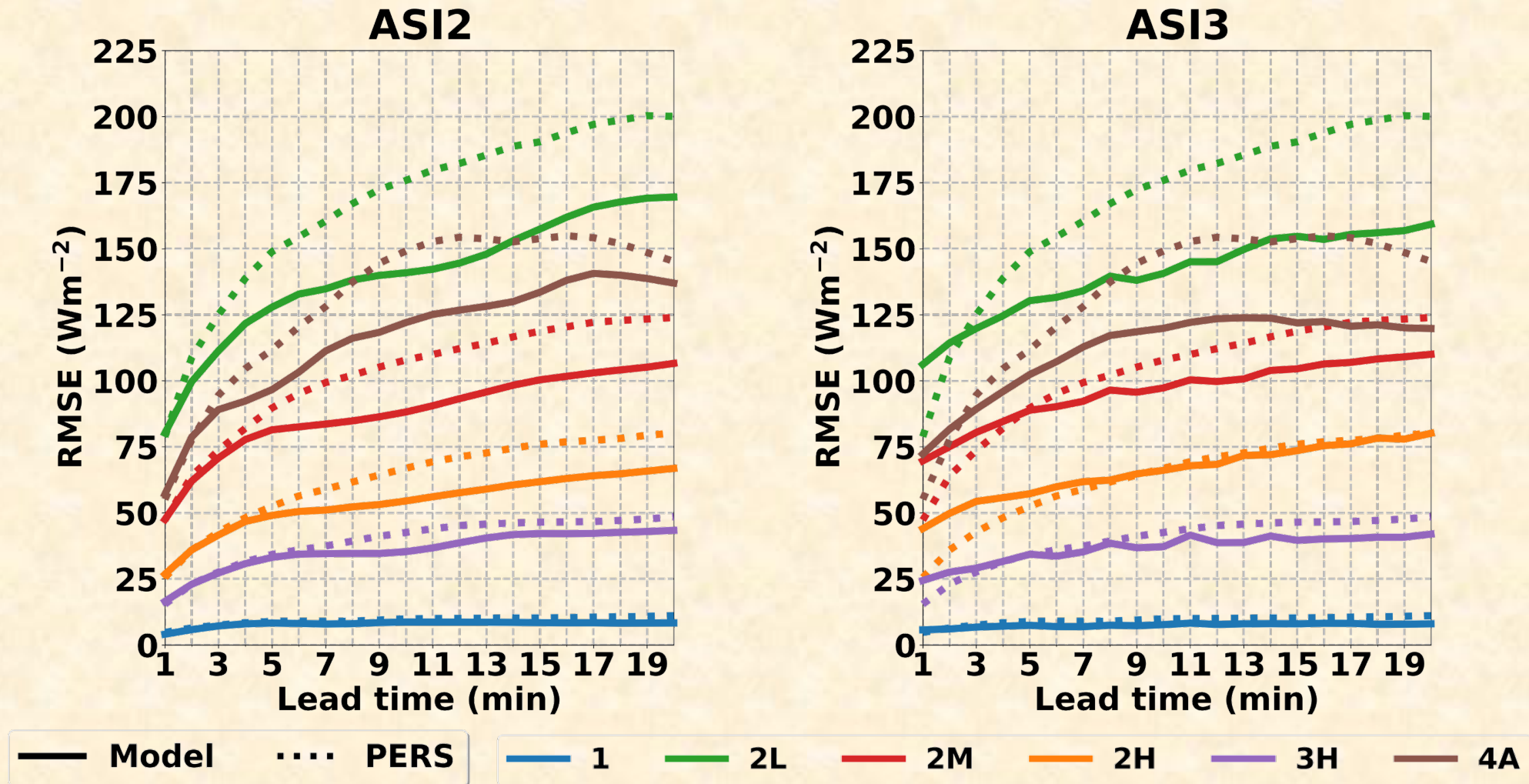
2L: Low clouds | **2M:** Multiple clouds | **2H:** High/Middle clouds

3H: Scattered/broken cloudiness with High/Middle clouds during half of the day, cloud-free during the other half.

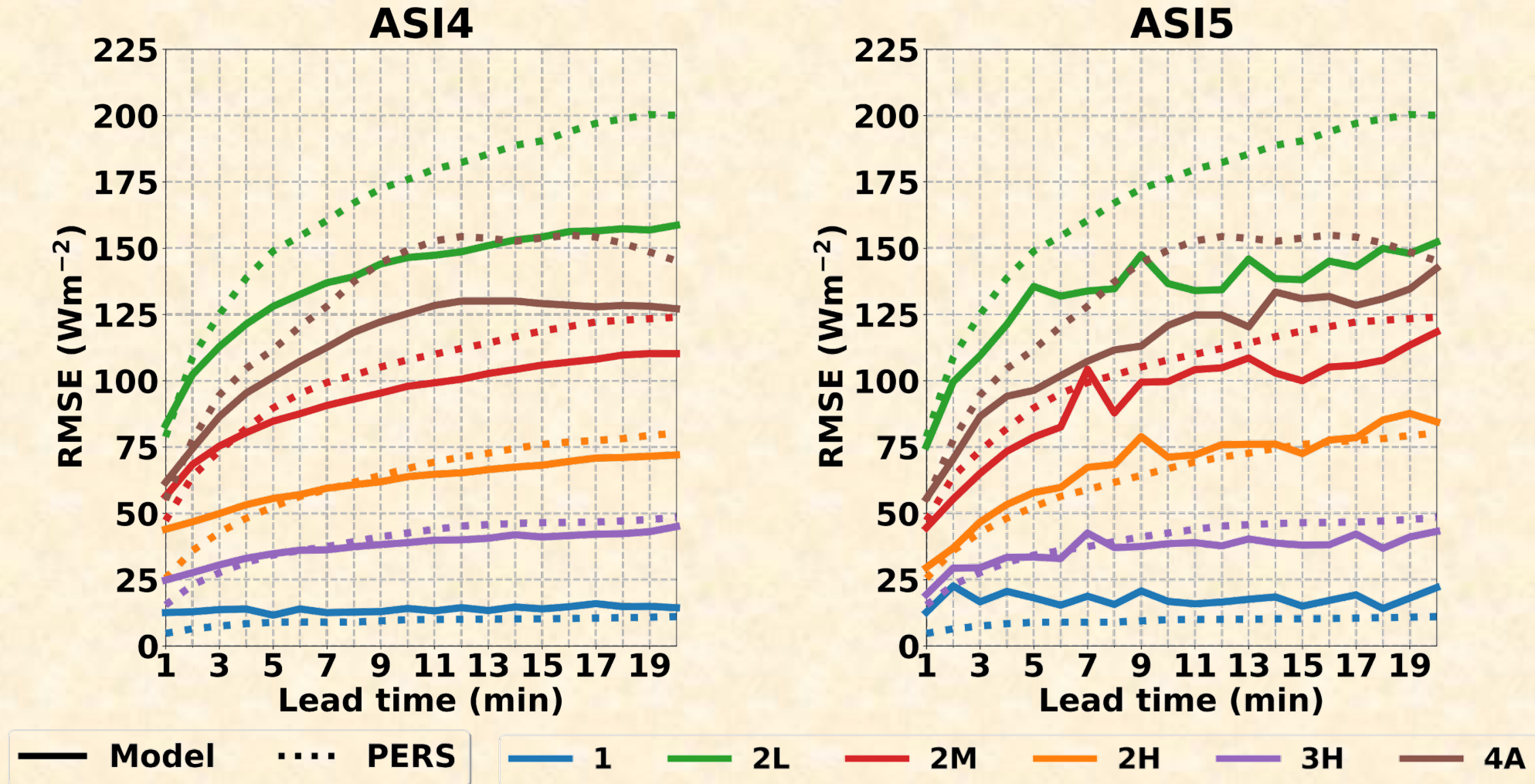
4A: Overcast cloud conditions during half of the day, scattered/broken cloudiness during the other half



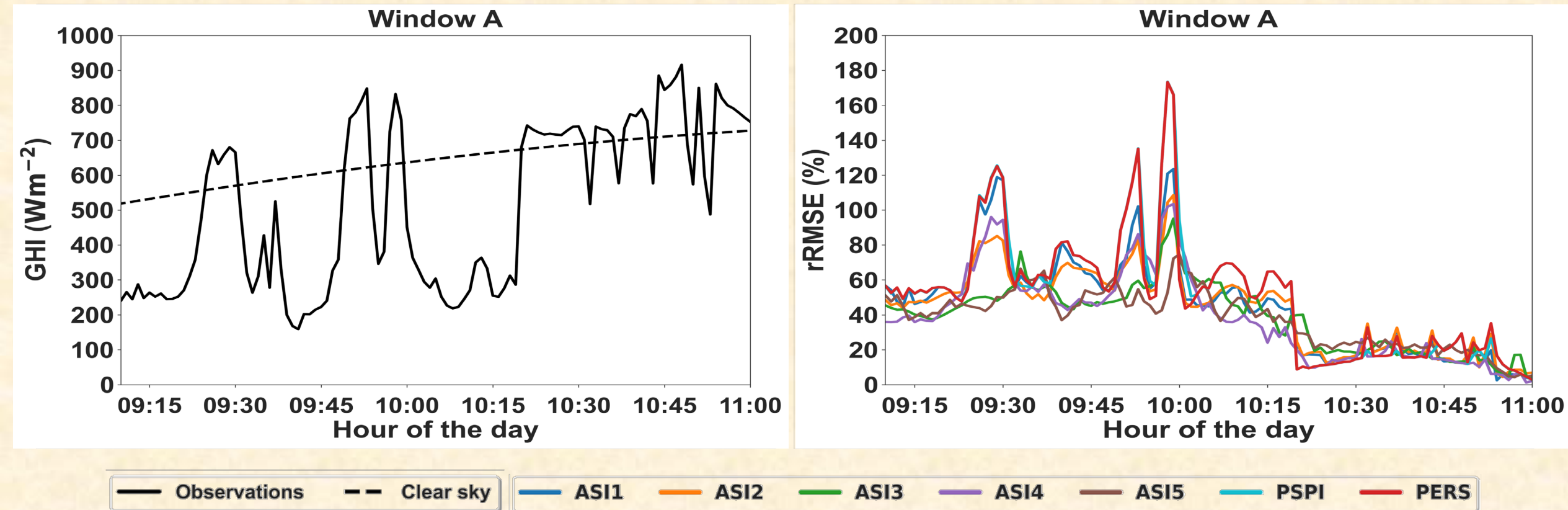
Nowcast performance at different cloud conditions and lead times against persistence



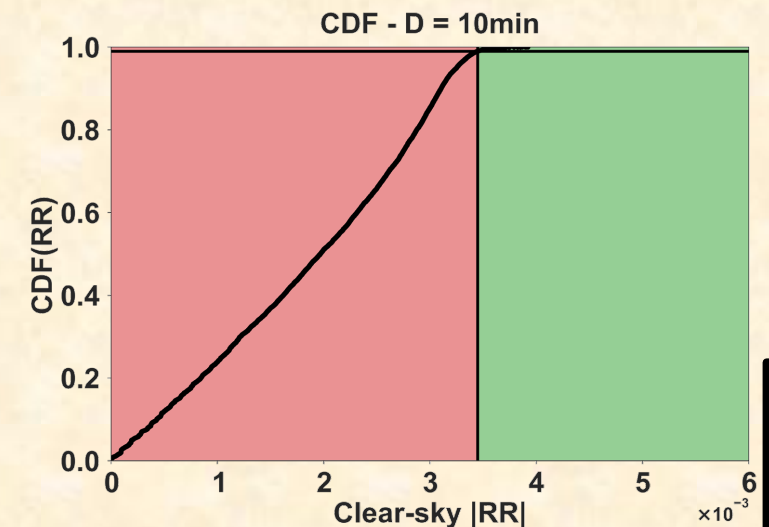
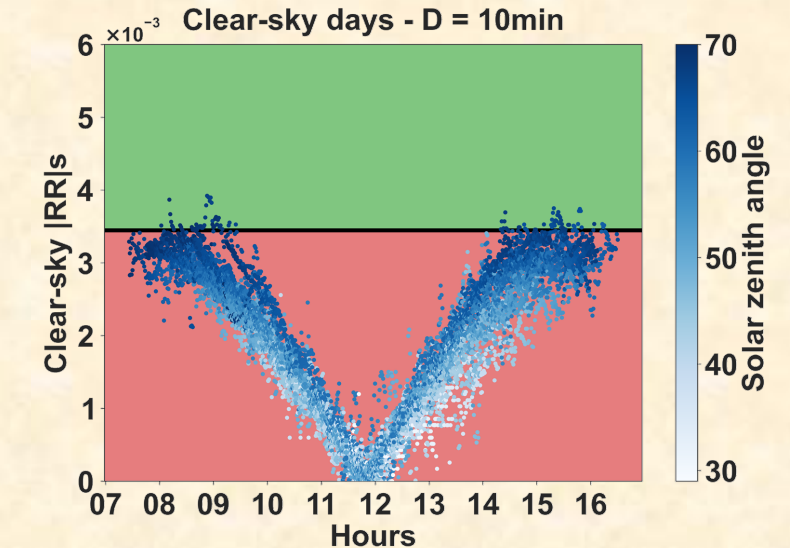
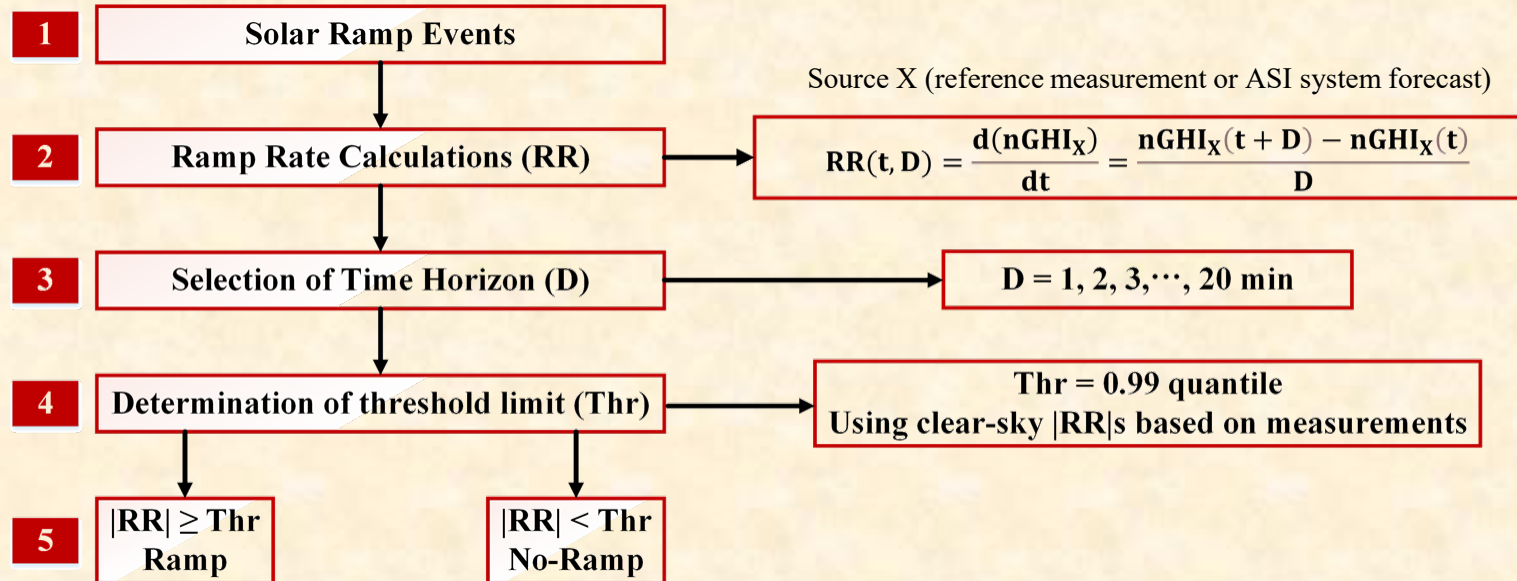
☒ Nowcast performance at different cloud conditions and lead times against persistence



☒ Nowcast performance at different cloud conditions and lead times against persistence

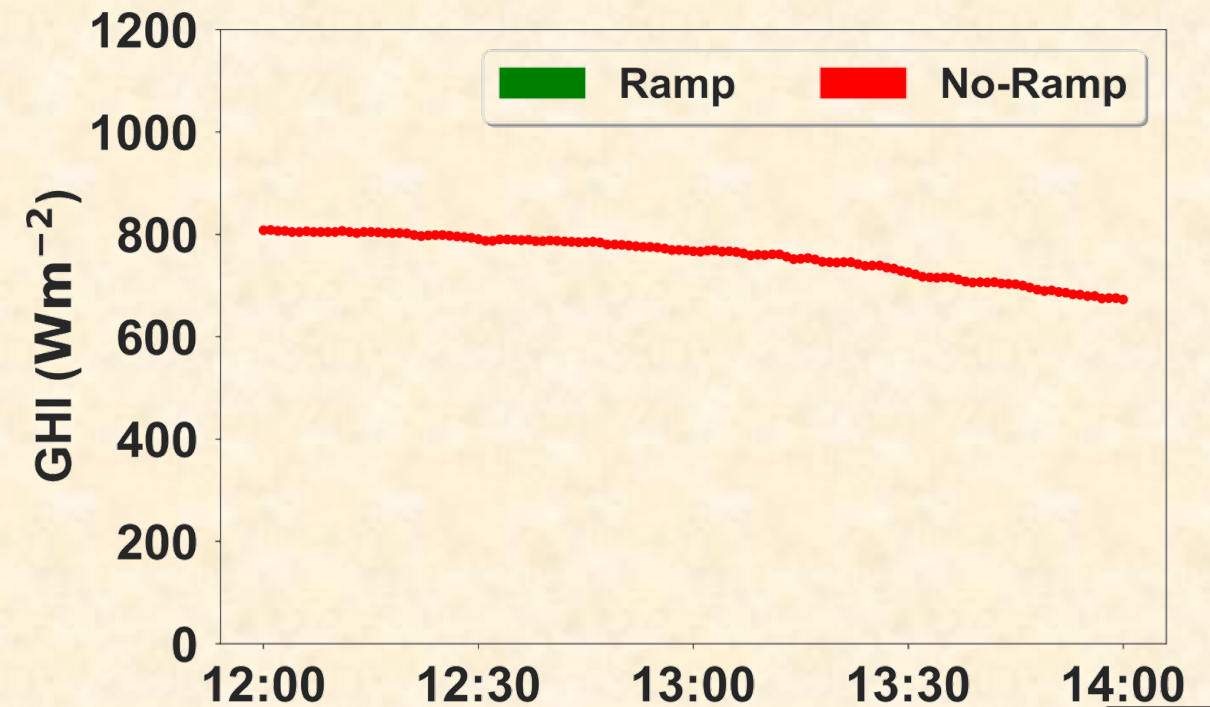
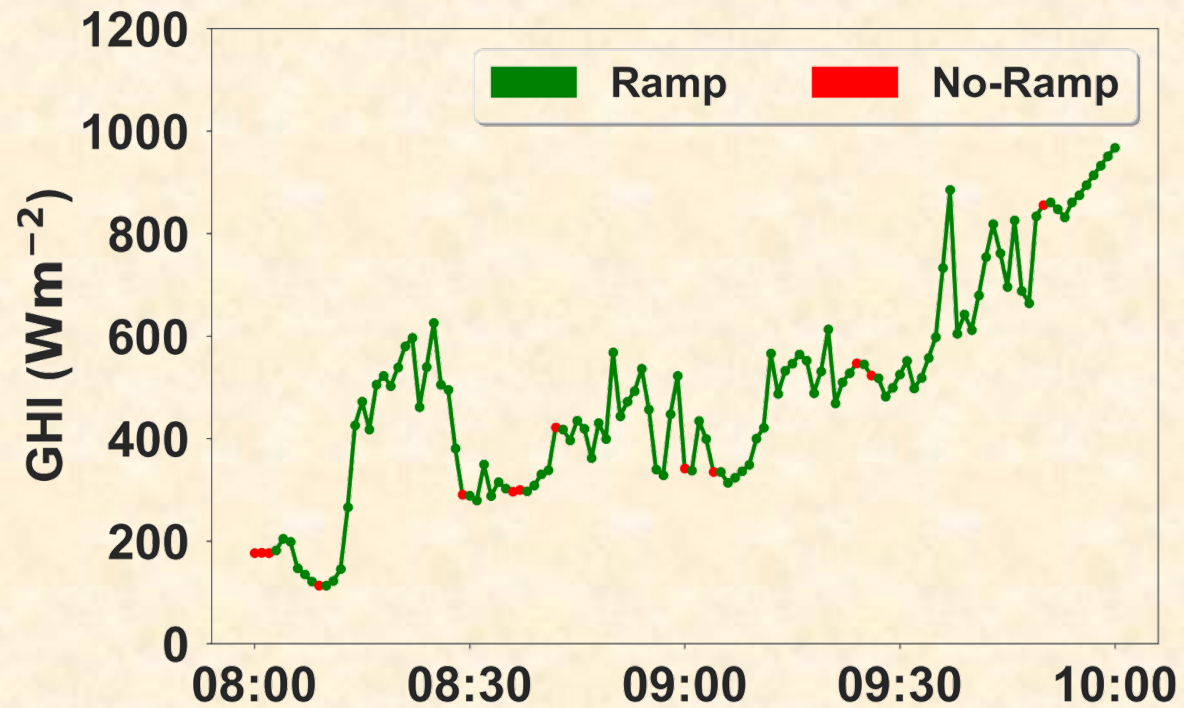


☒ Flowchart of the applied methodology



☐ Motivation ☐ Campaign ☐ Solar irradiance forecasting ☒ **Ramp Event forecasting** ☐ Conclusions

☒ **Verification of the applied methodology at a specific time horizon**



☒ Possible cases

		Observed Ramp Events	
		Ramp	No-Ramp
Predicted Ramp Events	Ramp	True Ramp (TR)	False Ramp (FR)
	No-Ramp	False No-Ramp (FNR)	True No-Ramp (TNR)

☒ Temporal analysis of ramp event detection

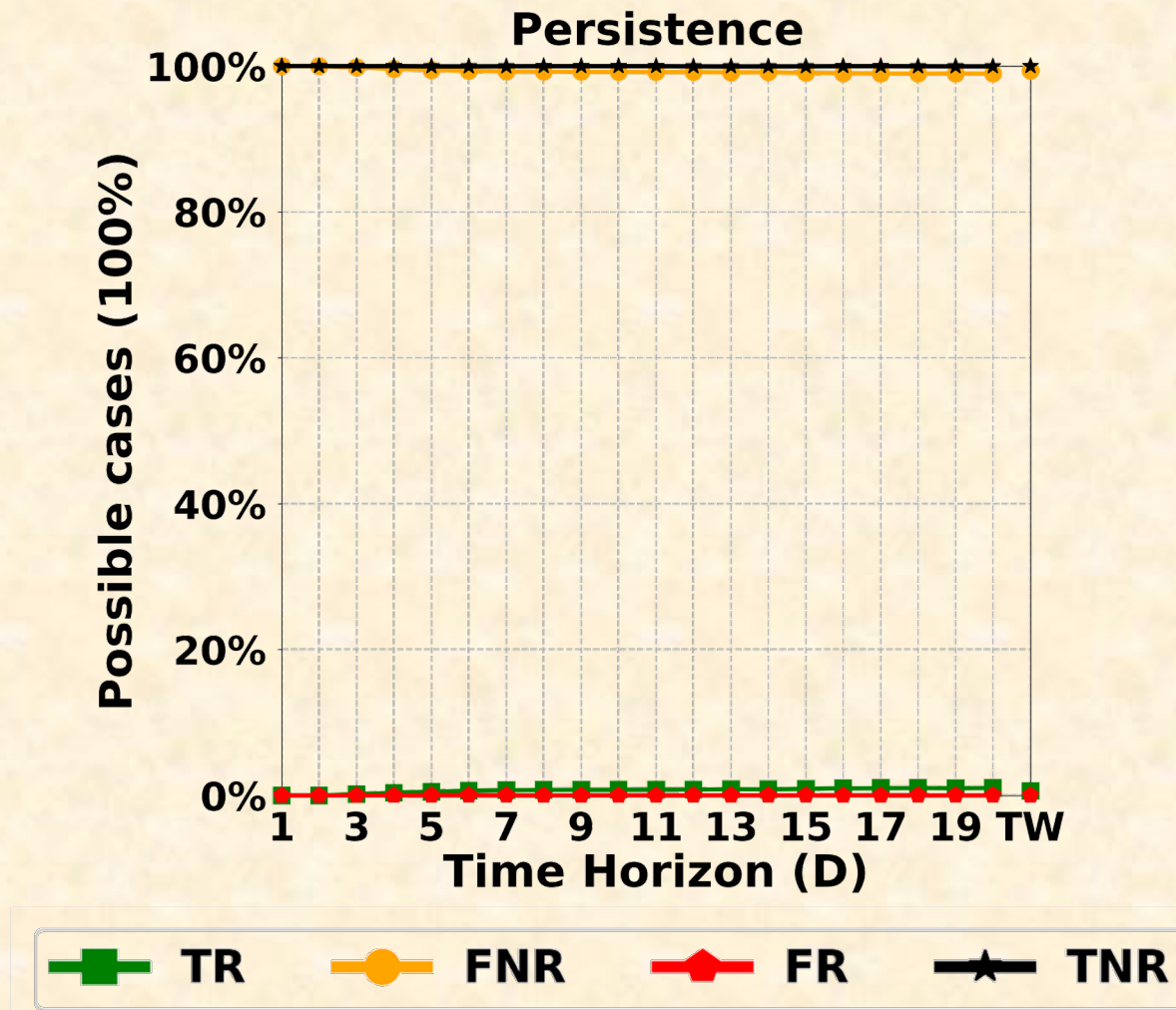
- 1) At each specific time horizon (from 1 to 20 min)
- 2) Over the whole 20-min time horizon (Time window, TW, analysis)

☒ Metric

$$\text{Total Accuracy} = \frac{\text{True cases}}{\text{Total cases}} = \frac{\text{TR} + \text{TNR}}{\text{TR} + \text{FR} + \text{FNR} + \text{TNR}}$$



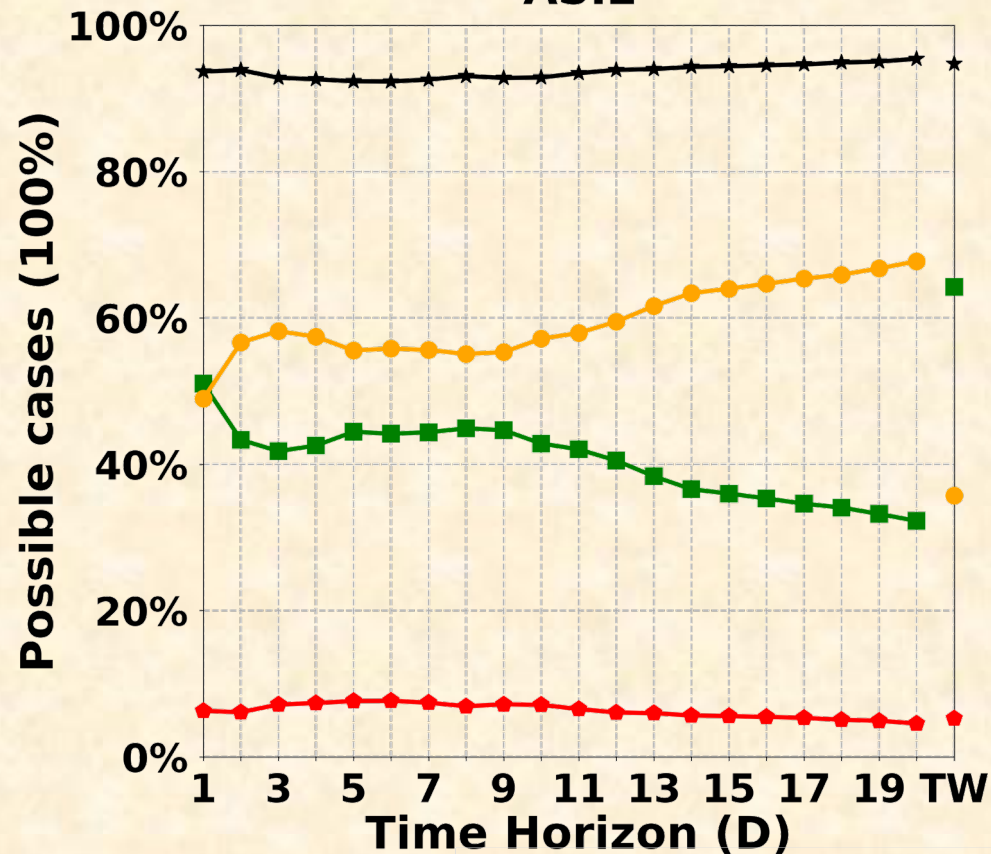
☒ ASIs performance at each time horizon D and for the TW



ASIs performance at each time horizon D and for the TW

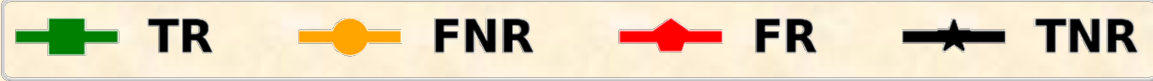
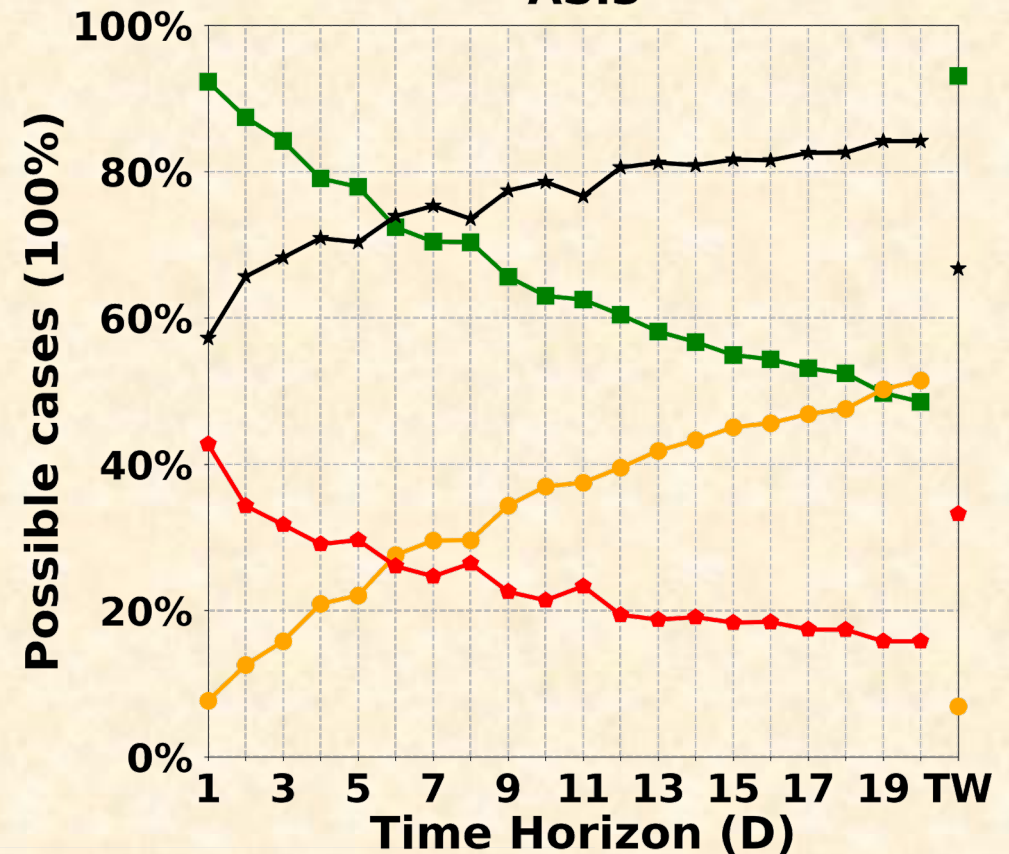
TW Accuracy=78.6%

ASI2



TW Accuracy=80.7%

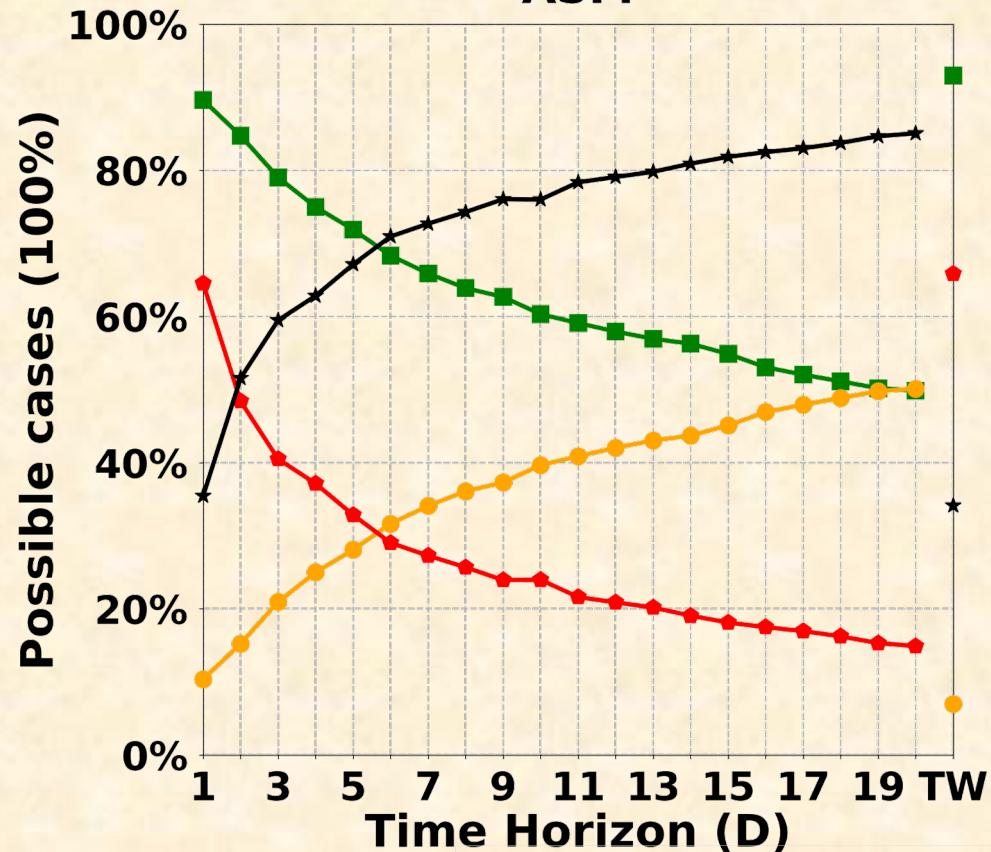
ASI3



ASIs performance at each time horizon D and for the TW

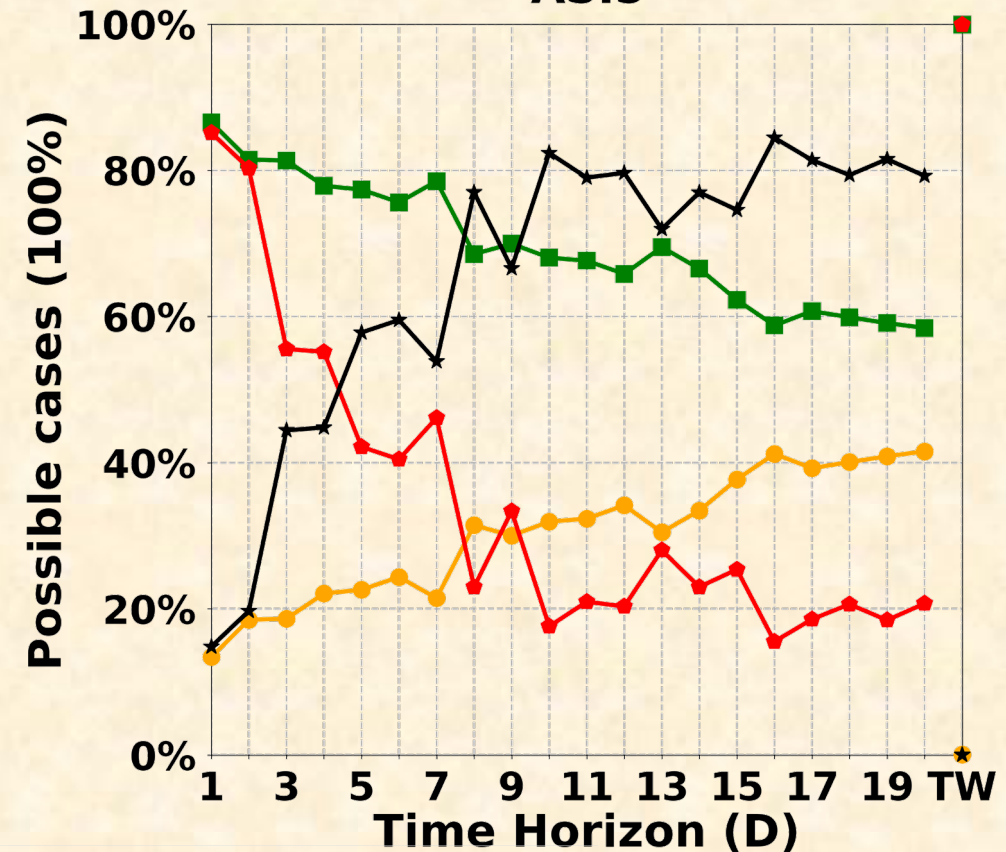
TW Accuracy=65.3%

ASI4



TW Accuracy=52.9%

ASI5



☒ TR
 ☒ FNR
 ☒ FR
 ☒ TNR



Within the whole validation period

- ASIs 1-2 reported the lowest deviations among the ASIs.
- In general, ASIs deviations follow an increasing trend as the time horizon increases, with ASI 1 and 2 revealing the highest forecast accuracy either at low (<5 min) or at distant (>15 min) lead times.

Under specific cloud conditions

- ASI1 and 2 outperform the persistence models at all cloud clusters and lead times.
- For time horizons longer than 5 min, the other 3 ASIs (3-5) are also efficient to outperform the persistence model under cloudy skies.

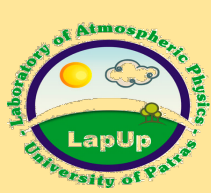
Under cloud-free conditions

- Three ASIs (1-3) proved to be capable of providing better results than persistence.

Overall conclusion

- Specific ASIs **outperform** the persistence models even under clear, scattered, and overcast skies.





Solar irradiance and ramp events forecasting based on all-sky imagers



☐ Motivation ☐ Campaign ☐ Solar irradiance forecasting ☐ Ramp Event forecasting ☒ **Conclusions**

The performance of the detected ramp event is connected to the applied forecast algorithms.

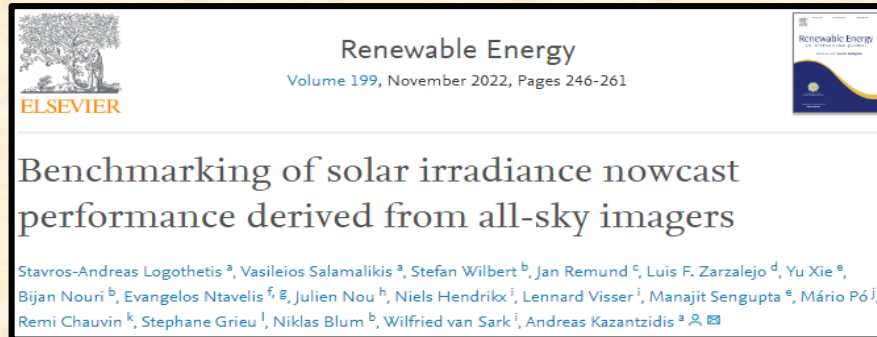
- The **true** predicted ramp event cases for all ASIs **decrease** as the time horizon increases.
- The ASI1 and ASI2, tend to **wrongly** record **no-ramp events** more often than ASI 3-5 systems.
- The ASI 3-5 show the tendency to **wrongly** predict **ramp events** which happens less often for ASI1 and ASI2.
- They can predict ramp events accurately from 30% to 95% of the cases.

Overall conclusion

- The selection of the most suitable ASI system for solar irradiance ramp event nowcasting **depends** on the **application**.



Thank you for your attention!!!



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