### Forecasting Solar Irradiance by looking at clouds from above and below

Helmholtz Imaging Conference, Hamburg, 14.-17. June 2023

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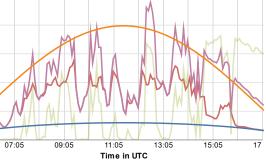


#### **Overview**

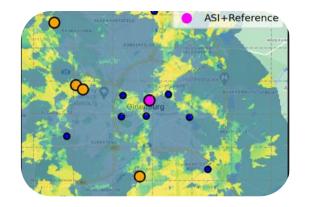




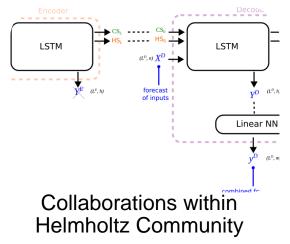
Tracking clouds from above and below – data sources **OLUOL:** Irradiance



Solar irradiance modelling in the presence of clouds



Towards a seamless forecast – from deterministic modelling to Al approaches

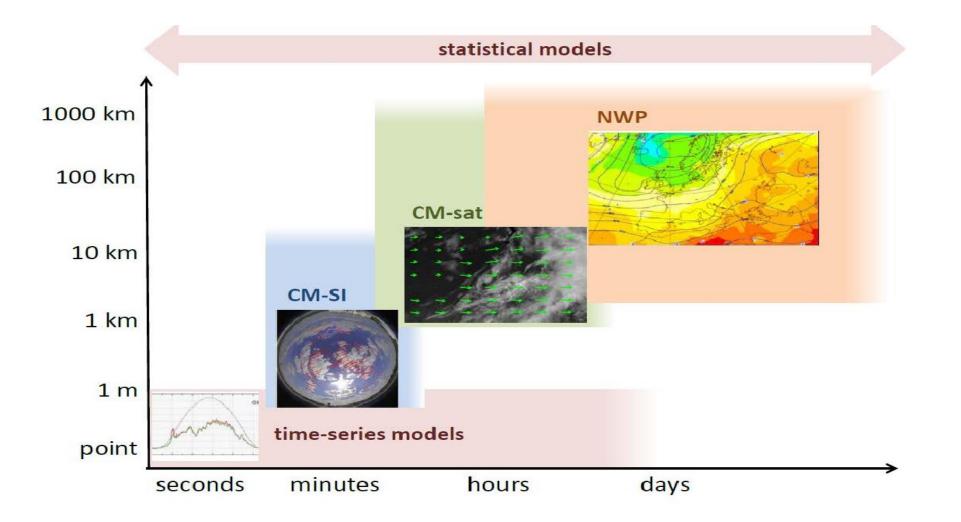




## DATA SOURCES

#### **Solar Power Forecasts**



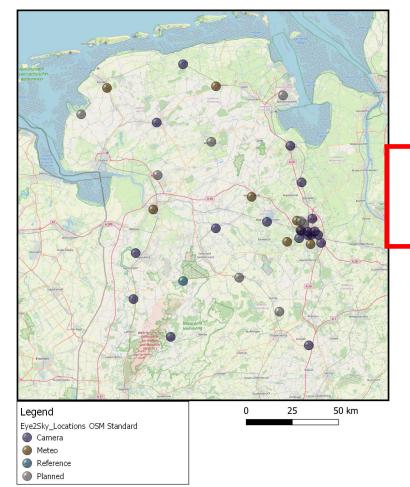


Sengupta et al.: Best Practices Handbook for the Collection and Use of Solar Resource Data for Solar Energy Applications, Technical Report NREL/TP-5D00-63112, Denver, 236pp, 2015

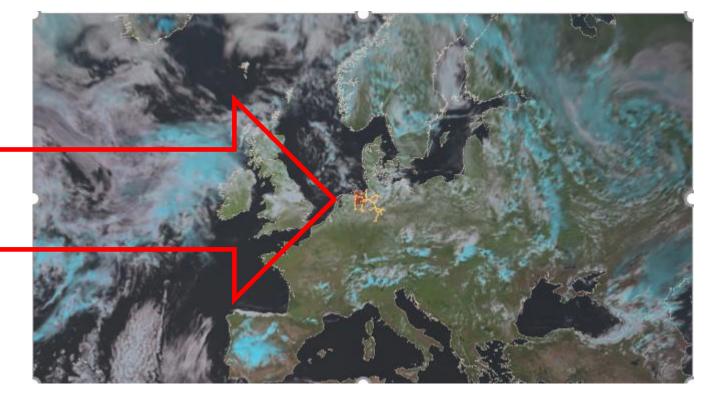
#### Data Sources for Cloud Detection and Solar Irradiance Measurements



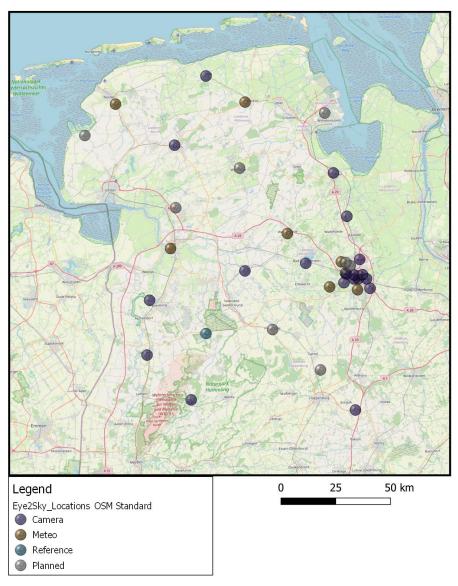
Eye2Sky - Cloud camera and meteorogical measurement network in Oldenburg

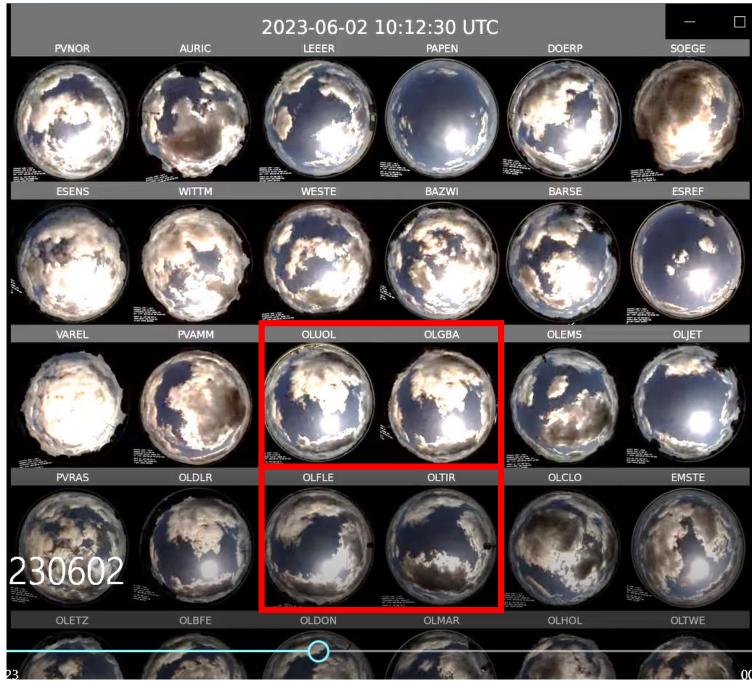


Meteosat Second Generation – geostationary satellite



Eye2Sky - Cloud camera and meteorogical measurement network in Oldenburg



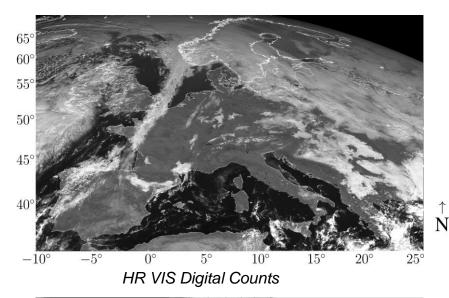


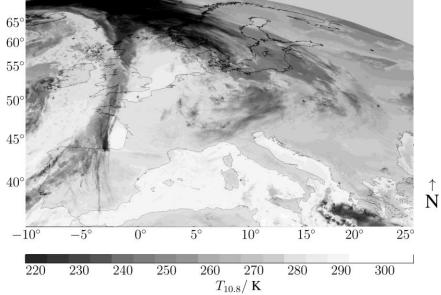
Annette Hammer, DLR - Institute of Networked Energy Systems

### Meteosat Second Generation vs Eye2Sky



	Eye2Sky	Meteosat Second Generation
Instrument and Provider	All Sky Imager: Webcam with fish eye lense; Mobotix	Spinning Enhanced Visible and InfraRed Imager (SEVIRI); EUMETSAT;
Data	RGB images	12 channels (600nm -12µm)
Type and Size	JPG 2112 x 2048 pixel	HRIT 10bit per pixel -> HDF5 3600 x 1800 pixel for Europe
Spatial resolution	50 m x 50 m	<ul> <li>1km x 1km (High Resolution Visible)</li> <li>/ 3km x 3km other solar and thermal channels (at sub satellite point)</li> </ul>
Update frequency	30 s	15 min; 5 min rapid scan



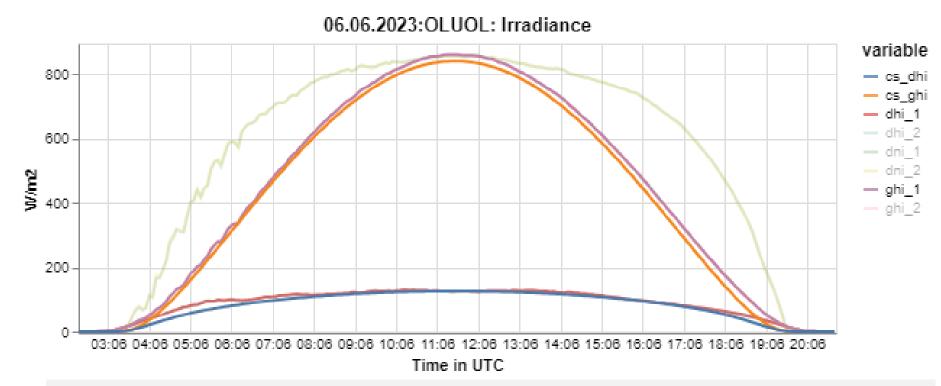




# SOLAR IRRADIANCE MODELING

#### Solar irradiance in a clear sky





Global Horizontal Irradiance (GHI); Clear Sky Model GHI; Measurement GHI Diffuse Horizontal Irradiance (DHI); Clear Sky Model DHI; Measurement DHI

### Clear Sky Model

Inputs: atmospheric turbidity and solar elevation

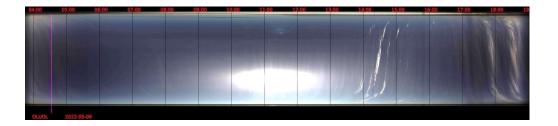
### Variability of solar irradiance mainly caused by clouds

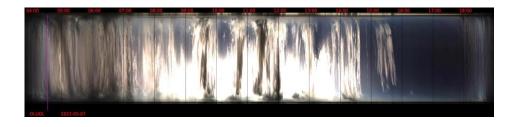


#### 2022-05-09

#### **OLUOL:** Irradiance **OLUOL:** Irradiance 1,000 1,000 variable — cs dhi – cs\_ghi 800 800 — dhi 1 — dhi 3 — dni 1 600 600 W/m2 W/m2 — dni 2 — ghi\_1 — ghi\_2 400 400 200 200 0\_\_\_\_\_ 0\_\_\_\_\_03:05 05:05 07:05 09:05 11:05 13:05 17:05 19:05 15:05 05:05 07:05 11:05 13:05 15:05 17:05 09:05 19:05 Time in UTC Time in UTC

2022-05-07



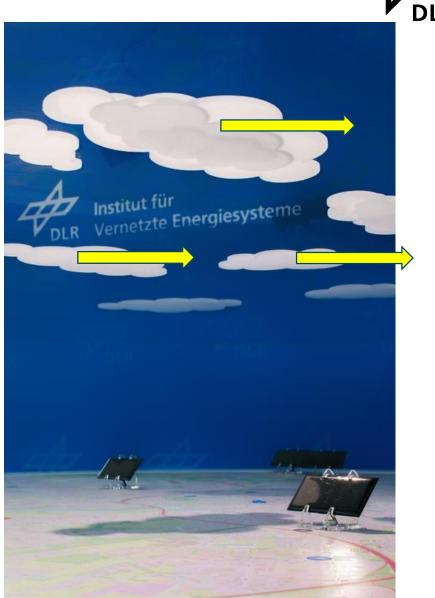




## SOLAR IRRADIANCE FORECASTING

#### What do we need to know for deterministic forecasting?

- Clear sky model
- Optical thickness of clouds / transmission
- Future positions of clouds (and sun)
- Cloud Motion Vectors

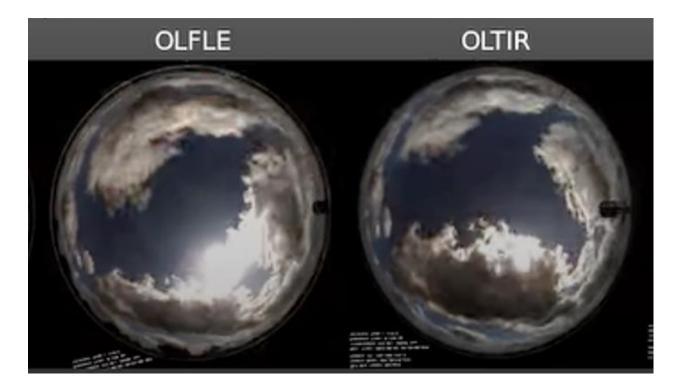




What do we need to know for deterministic forecasting?



Use two cameras for cloud height detection

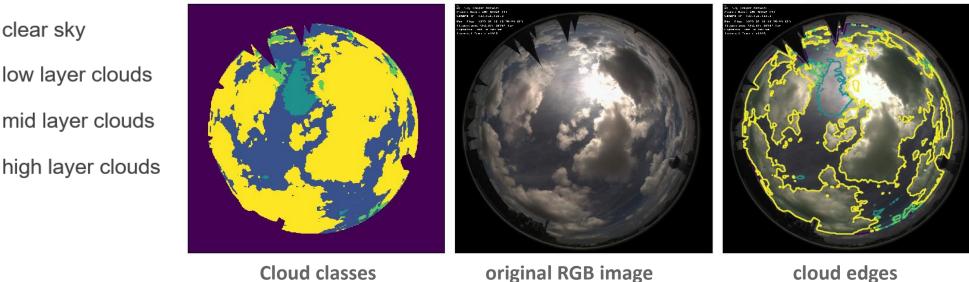


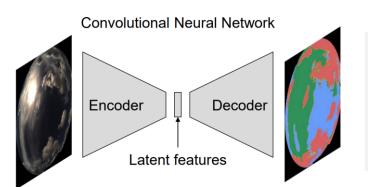


### **Cloud Detection – Semantic Segmentation**



22-06-26 10:36:00





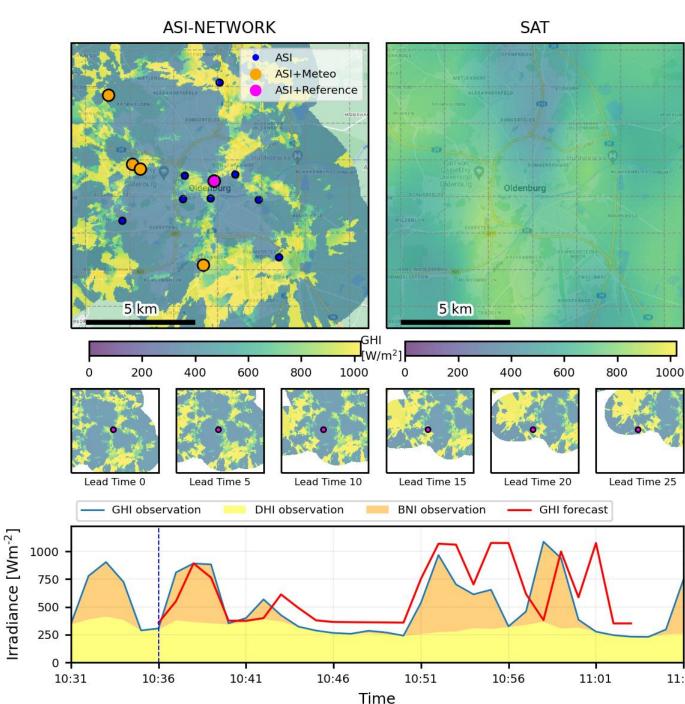
Fabel, Y.; Nouri, B.; Wilbert, S.; Blum, N.; Triebel, R.; Hasenbalg, M.; Kuhn, P.; Zarzalejo, L.F.; Pitz-Paal, R.
Applying self-supervised learning for semantic cloud segmentation of all-sky images.
Atmos. Meas. Tech. 2022, 15, 797–809.

#### **30min ahead forecasting** for domain Oldenburg

#### Method

Blum, Niklas, 2022: Nowcasting of Solar Irradiance and Photovoltaic Production Using a Network of All-Sky Imagers. Dissertation, RWTH Aachen

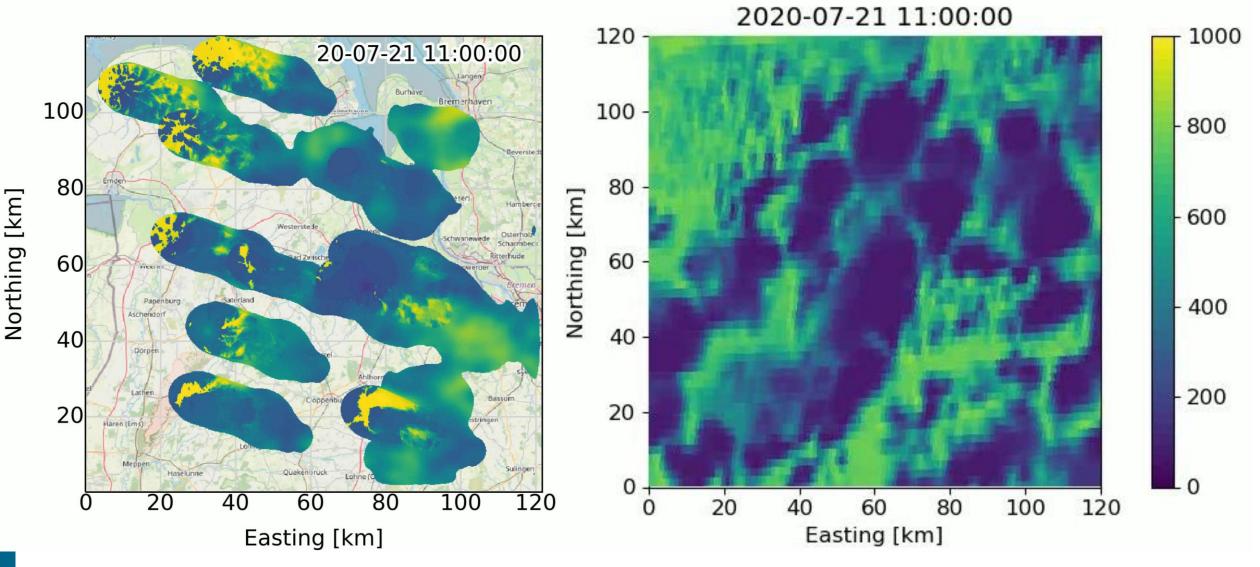
Visualization for 2022-06-26: Thomas Schmidt



11:06

#### **120min ahead forecasting for domain Weser Ems**





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#### **Hybrid Forecasts**



Combine both data sources to get a best estimate and a seamless forecast!

#### **End-to-End Nowcasting:**

# **INPUT:** Original images **OUTPUT:** GHI Timeseries

**Blending / Postprocessing:** 

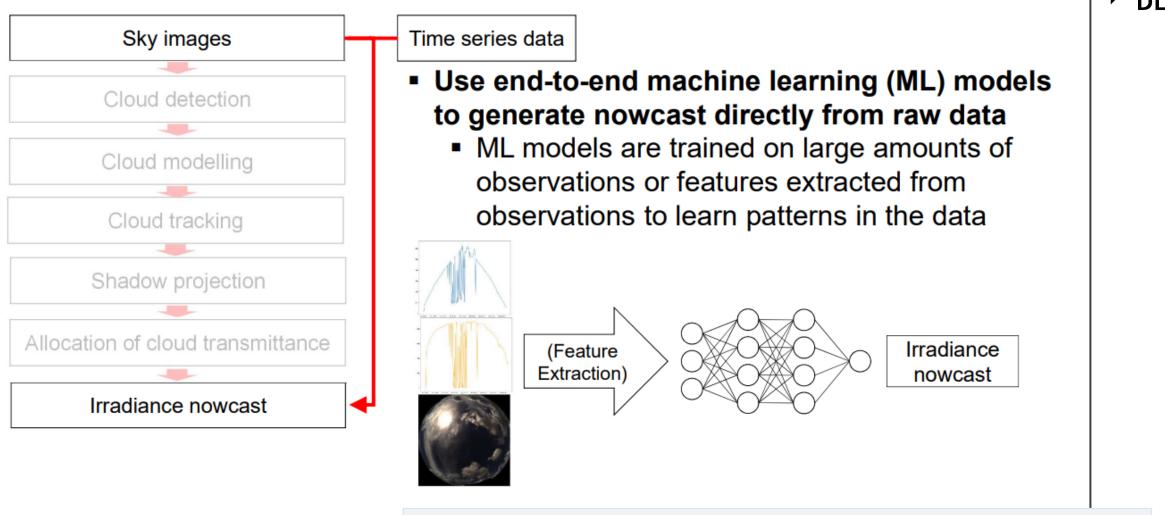
**INPUT:** SAT and ASI maps **OUTPUT**: GHI maps or timeseries

- Linear Regression
- Random Forest
- LSTM



### AI APPROACHES AND COLLABORATIONS

#### End-to-end Nowcasting – A data-driven approach

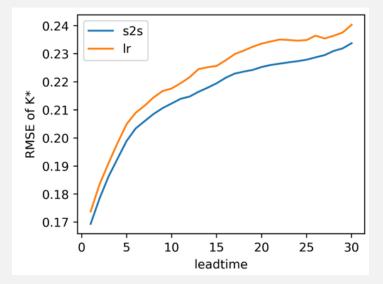


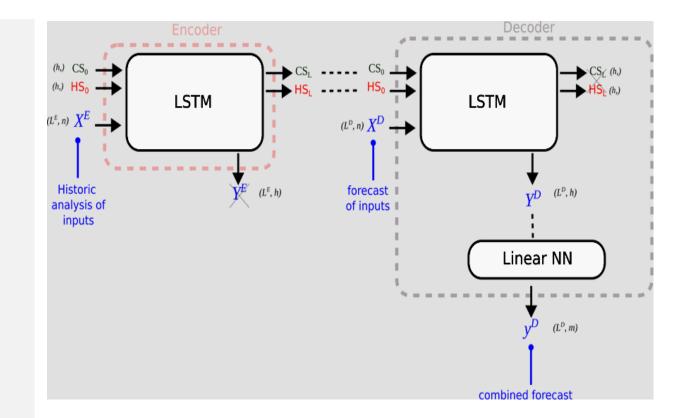
Fabel, Y.; et al: <u>Solar Nowcasting Systems Using AI Techniques.</u> 25th Cologne Solar Colloquium, 22nd of June 2022, Jülich, Germany.

### Sequence to Sequence (LSTM)



- INPUT: SAT and ASI maps of GHI
- OUTPUT: Sequences for all map pixels
- Better performance than Linear Regression





**Helmholtz Al Voucher:** Jorge Lezaca (DLR), Markus Götz, Charlotte Debus, and Arvid Weyrauch, Karlsruher Institut für Technologie (KIT), Steinbuch Centre for Computing (SCC)

#### **End-to-End-Combination**



 $\widehat{ci}_{ ext{t+20min}}$ 

PALETTA, Q., ARBOD, G., LASENBY, J., 2023: Omnivision forecasting: combining satellite observations with sky images for improved intra-hour solar energy predictions, Applied Energy, Volume 336, DOI:10.1016/j.apenergy.2023.120818.

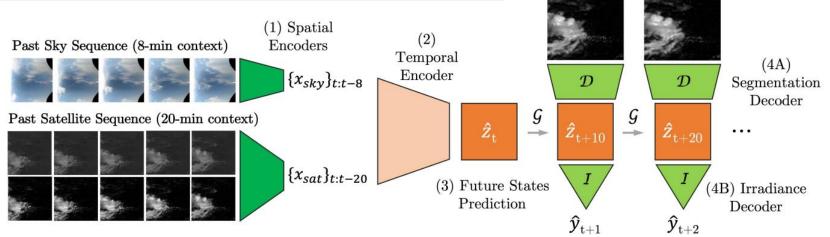


Figure 4: DL architecture predicting future cloud index maps and corresponding irradiance levels from past sky and satellite images. Parallel encoders are used to extract spatial features from the sequences of past sky and satellite observations (raw images and cloud index maps). Future states are iteratively predicted from the output of the temporal encoder. Each state can then be decoded into different representations (cloud index maps and irradiance levels in the figure).

Will be tested in Project DESYS by

- Keisuke Tanaka and Frank Dressel, DLR Institute of Software Methods for Product Virtualization
- Thomas Schmidt and Annette Hammer, DLR VE

#### Thank you!



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