

#PVPMC2026

May 12<sup>th</sup>, 2026

# Probabilistic reconstruction of unknown tracker angles for sub-hourly loss quantification



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# PREVIOUS WORK AND GOAL(S) OF THIS STUDY

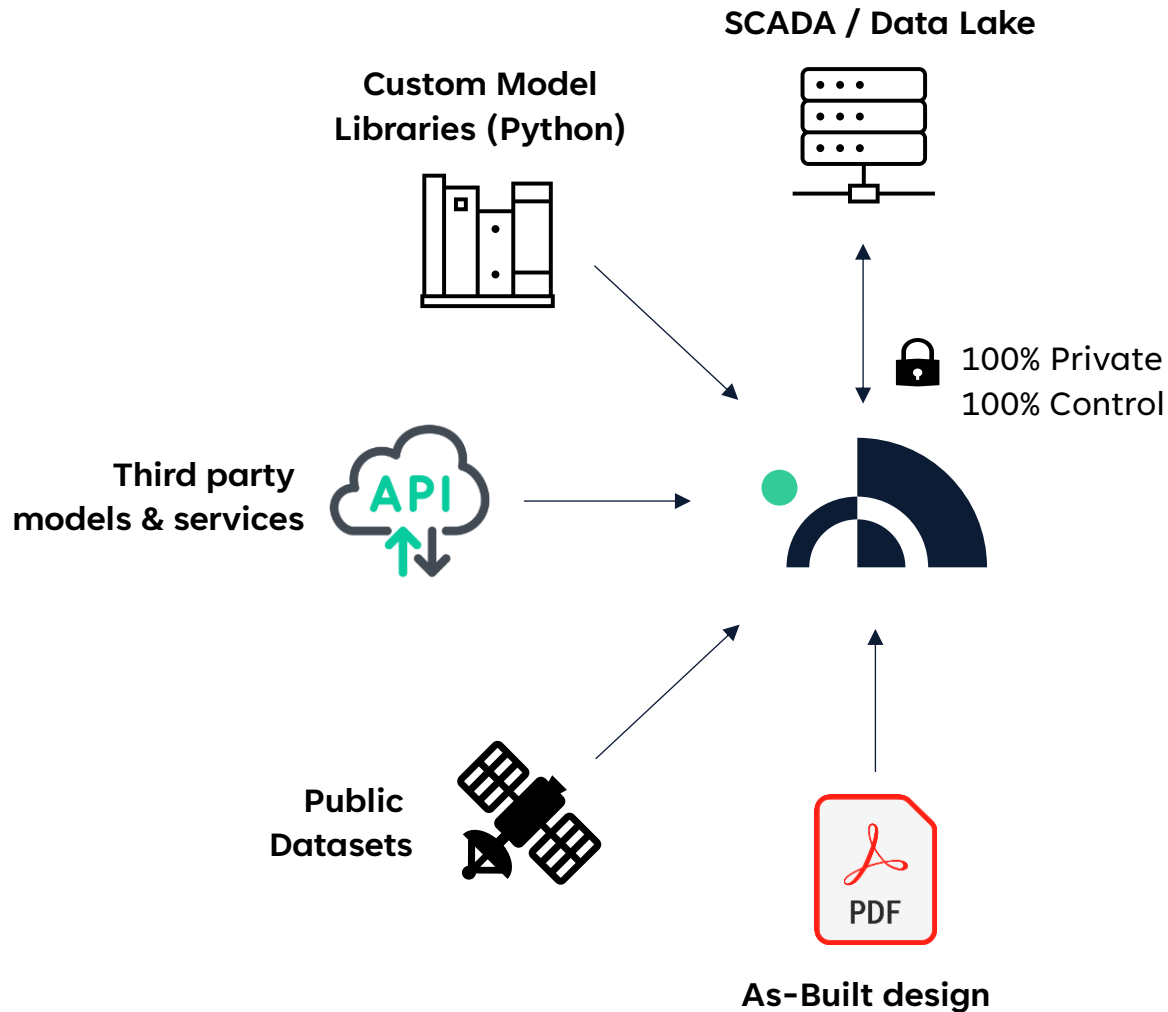
“A method for estimating time series PV production loss from solar tracking failures”, Anderson et al., *IEEE Journal of Photovoltaics*, 2021

1. Model power time series for:
    - Proper tracking
    - Fixed orientations (every 5°)
  2. Compare to power measurement
  3. Classify entire day as “Tracking” or “Stalled” (at angle)
- ✓ Fleet-level analysis
  - ✓ Daily resolution
  - × Performance analytics
  - × Sub-hourly resolution

**This work: Is it possible to ...**

- 1. Decide if a tracker is stuck at minute level?**
- 2. Reconstruct tracker angles without measurement?**
- 3. Apply methods at scale over thousands of trackers and strings?**

# PVRADAR MODELING FRAMEWORK



Our goal is to allow you to ...

- ✓ Create internal tools
- ✓ Apply at scale
- ✓ Automate manual tasks

*Much easier and faster.*

Use-cases PV + BESS:

1. performance analytics
2. design optimization

# METHODOLOGY

Hypothetical 2.7 MWp / 2.25 MW power plant  
placed in Nye county, Nevada



Image: Google Earth

## 1. Meteo data for exact location

- GHI from BSRN “Desert Rock”
- Wind speed and air temp. from ERA5
- Soiling from PVRADAR soiling model

## 2. Tracker angles

- Ideal angles from pvlib
- Small chance tracker stops communicating (angles are NaN)
- Small chance to be stuck

## 3. Calc. component current, voltage, power

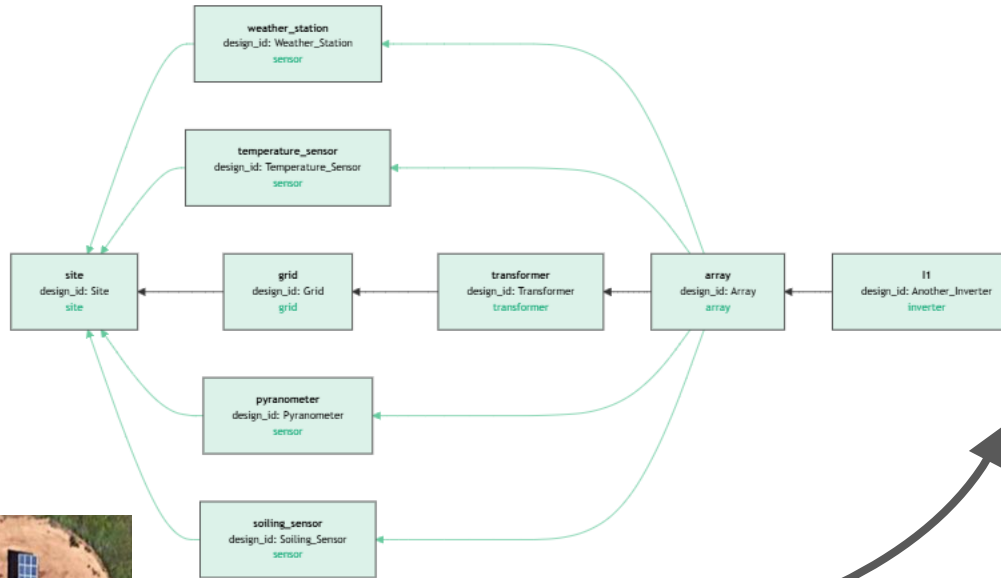
## 4. Add measurement error as noise

## Example: Module design

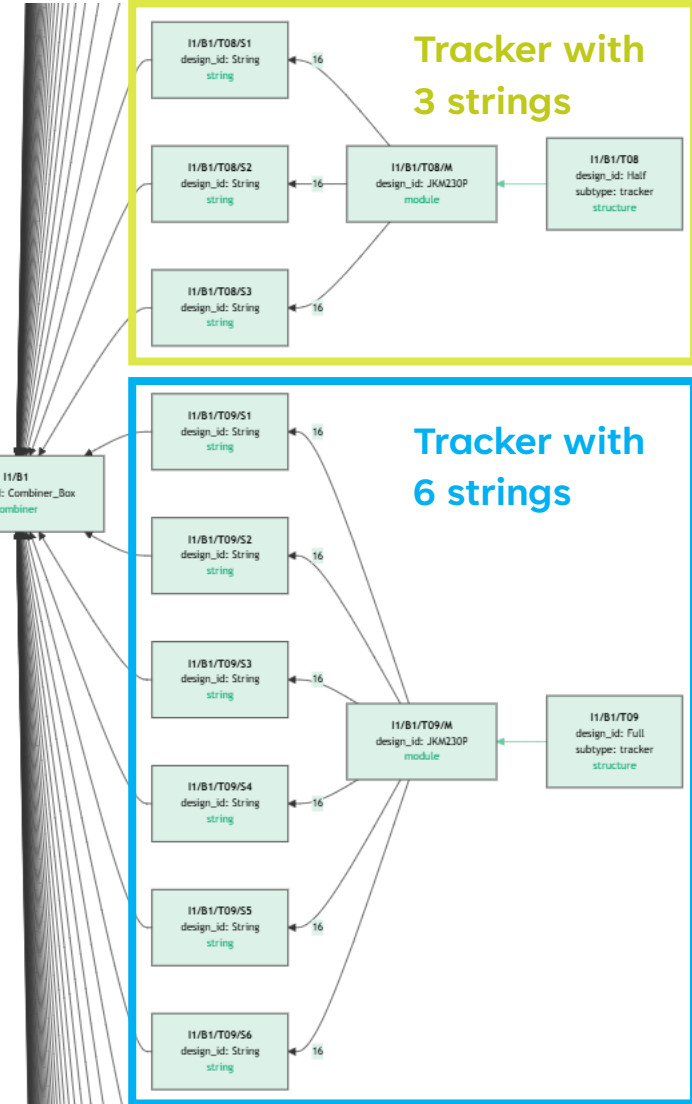
```

JKM235P
manufacturer: Jinko Solar
model: JKM235P-60
rated_module_power: 235
module_eff_value: 0.1435
bifaciality_factor: 0
degradation_rate: 0.7
light_induced_degradation: 2.5
temperature_coefficient_power: -0.0045
temperature_coefficient_isc: 0.0005
temperature_coefficient_voc: -0.0027
cell_string_count: 3
half_cell: 0
module_construction: glass_polymer
short_side: 0.992
long_side: 1.65
nominal_operating_cell_temp: 45
cells_per_cell_string: 20
mpp_voltage_stc: 29.8
mpp_current_stc: 7.89
open_circuit_voltage_stc: 36.9
short_circuit_current_stc: 8.47
bandgap_energy: 1.121
bandgap_energy_temp_coef: -0.0002677
surface_area: 1.6368
    
```

module



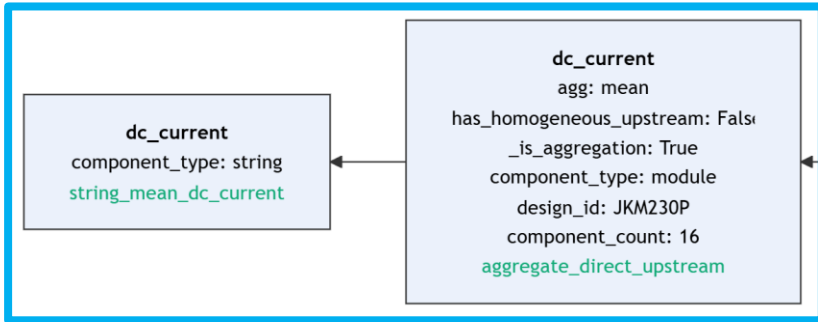
# FROM LAYOUT TO COMPONENT GRAPH



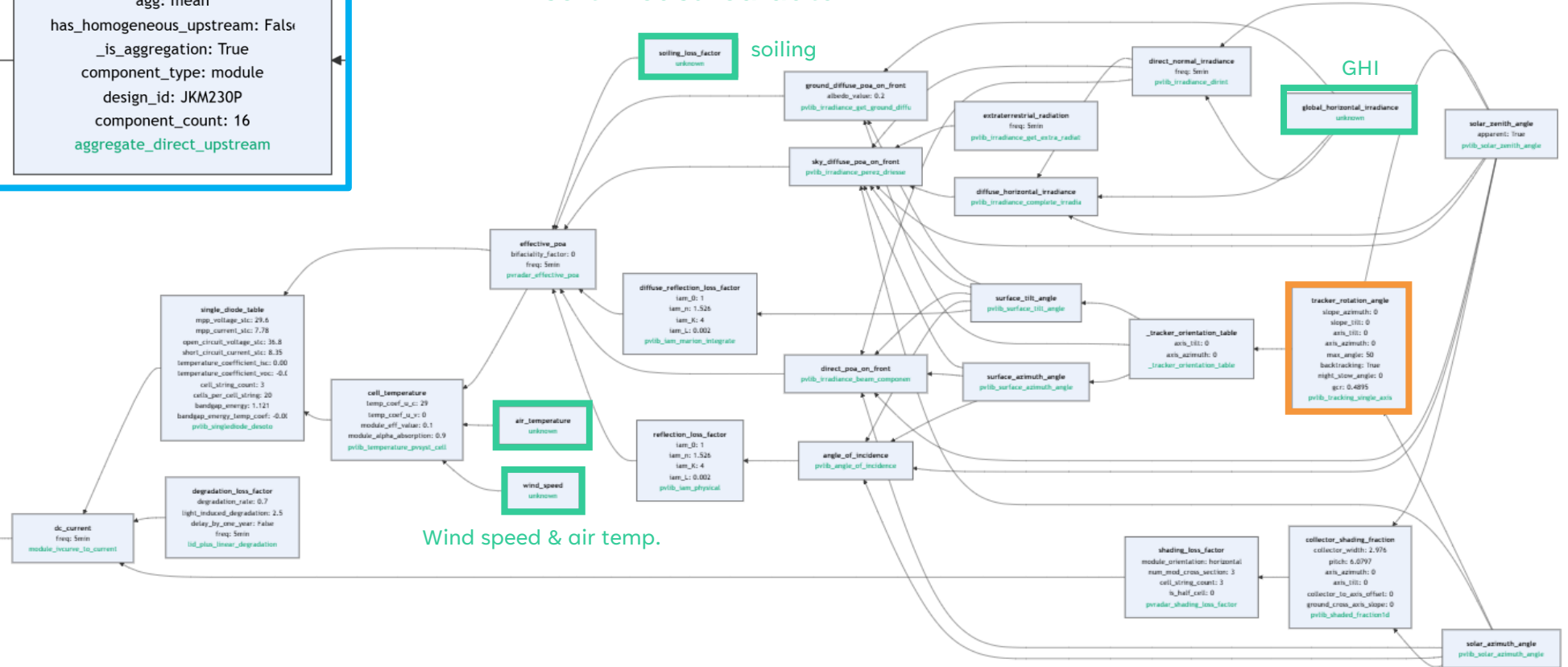
Here visible part of Block 1 (of 8).  
Full graph has 1000 nodes.

# MODEL CHAIN

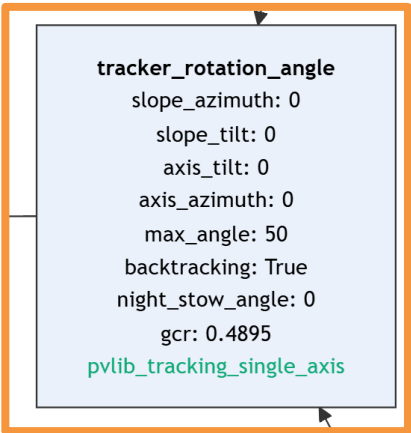
Aggregation: from 16 Module currents to 1 string



Insert measured data

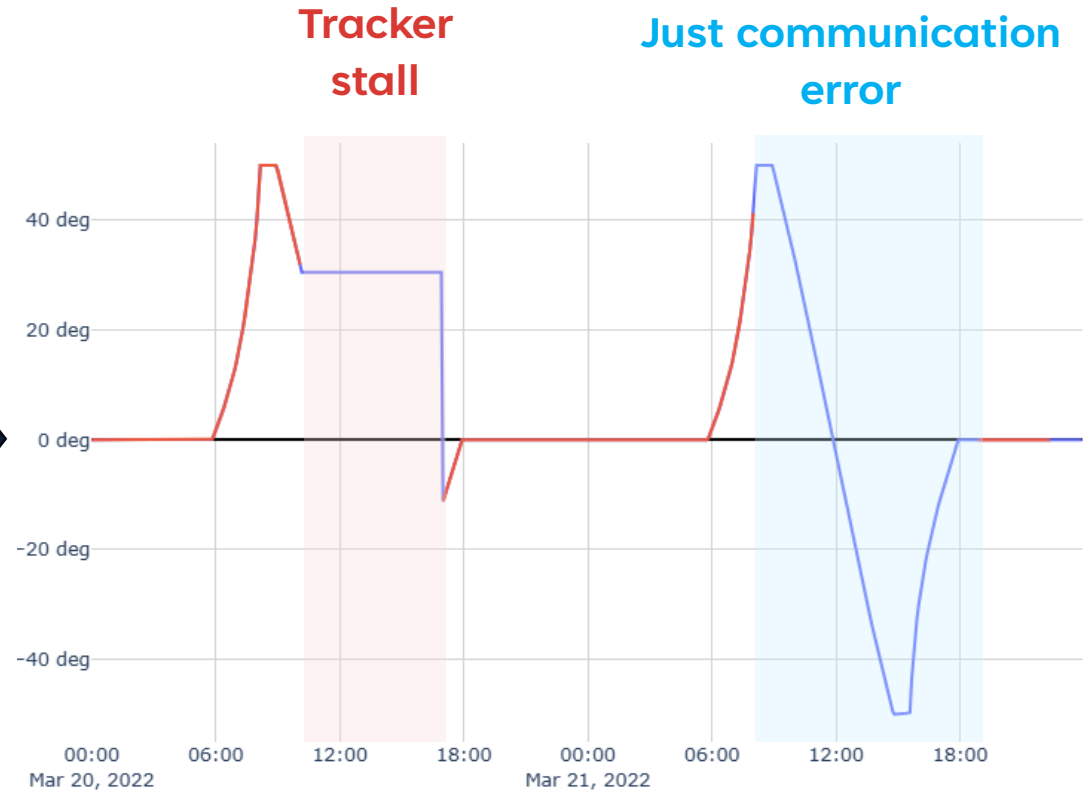
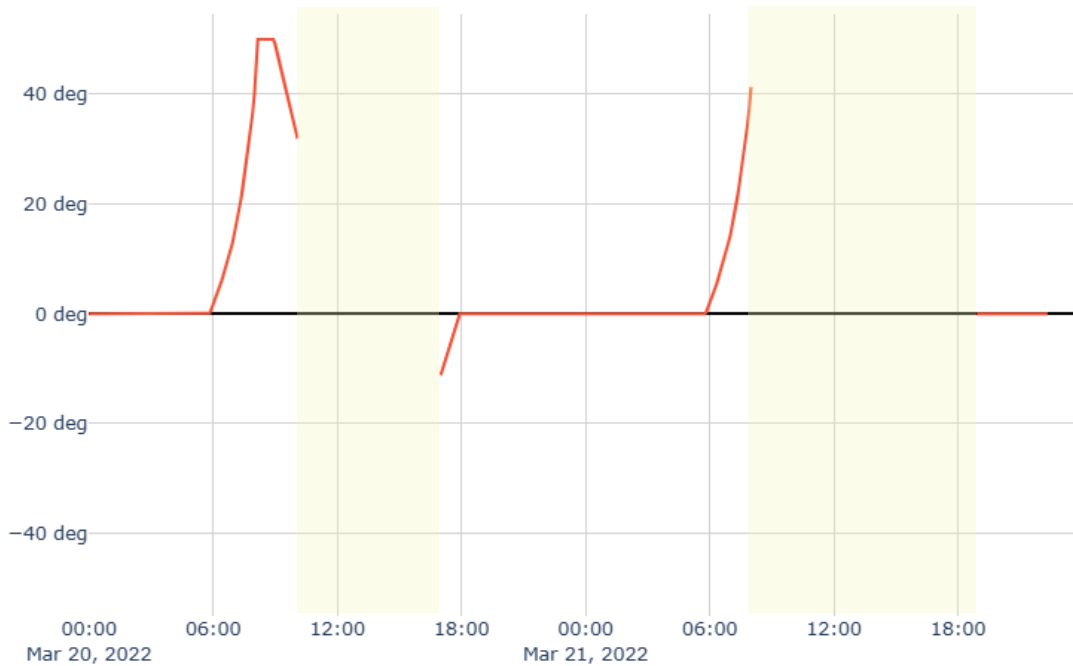


Pvlib wrappers



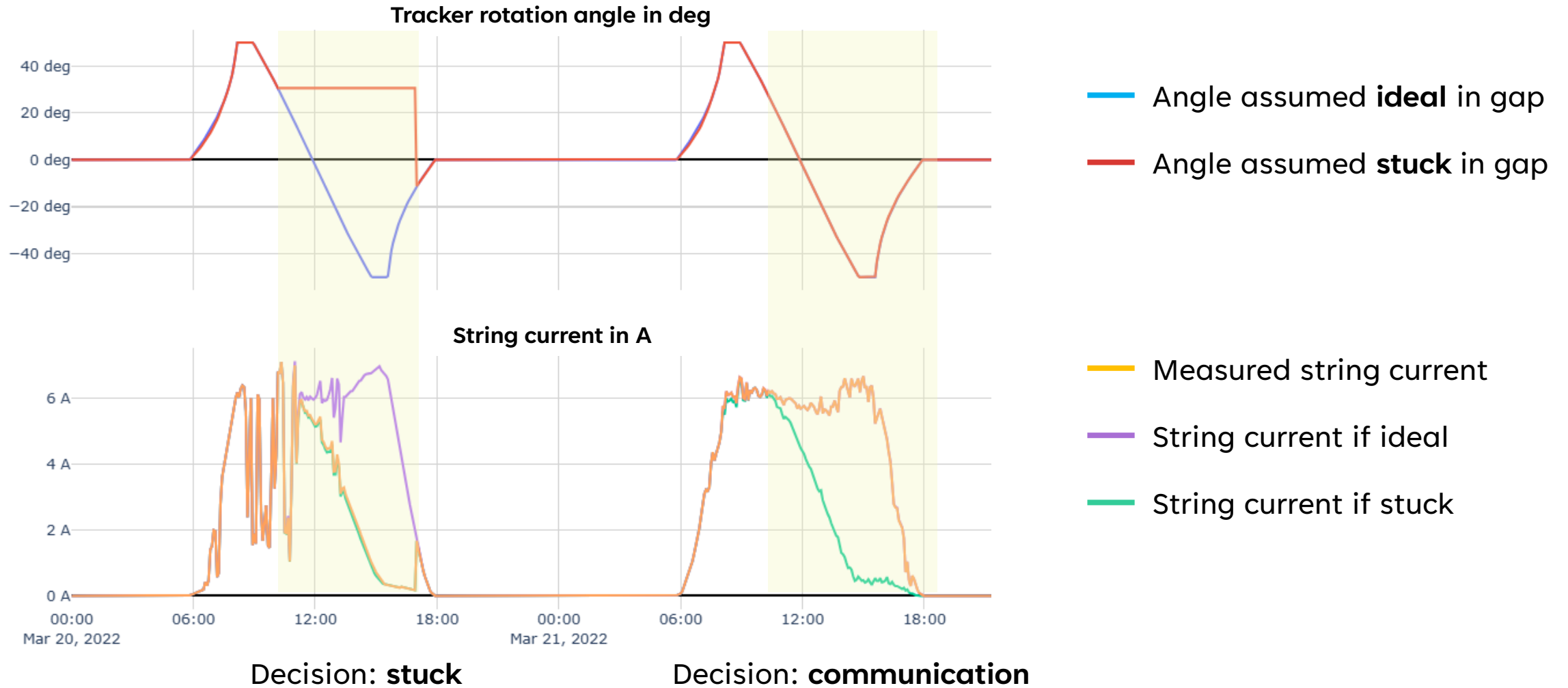
# CASE 1: DATA GAPS MAY RESULT FROM EITHER A TRACKER STALL OR A COMMUNICATION LOSS

Tracking angles are available but **have gaps**.  
Is the tracker stuck or just not communicating?

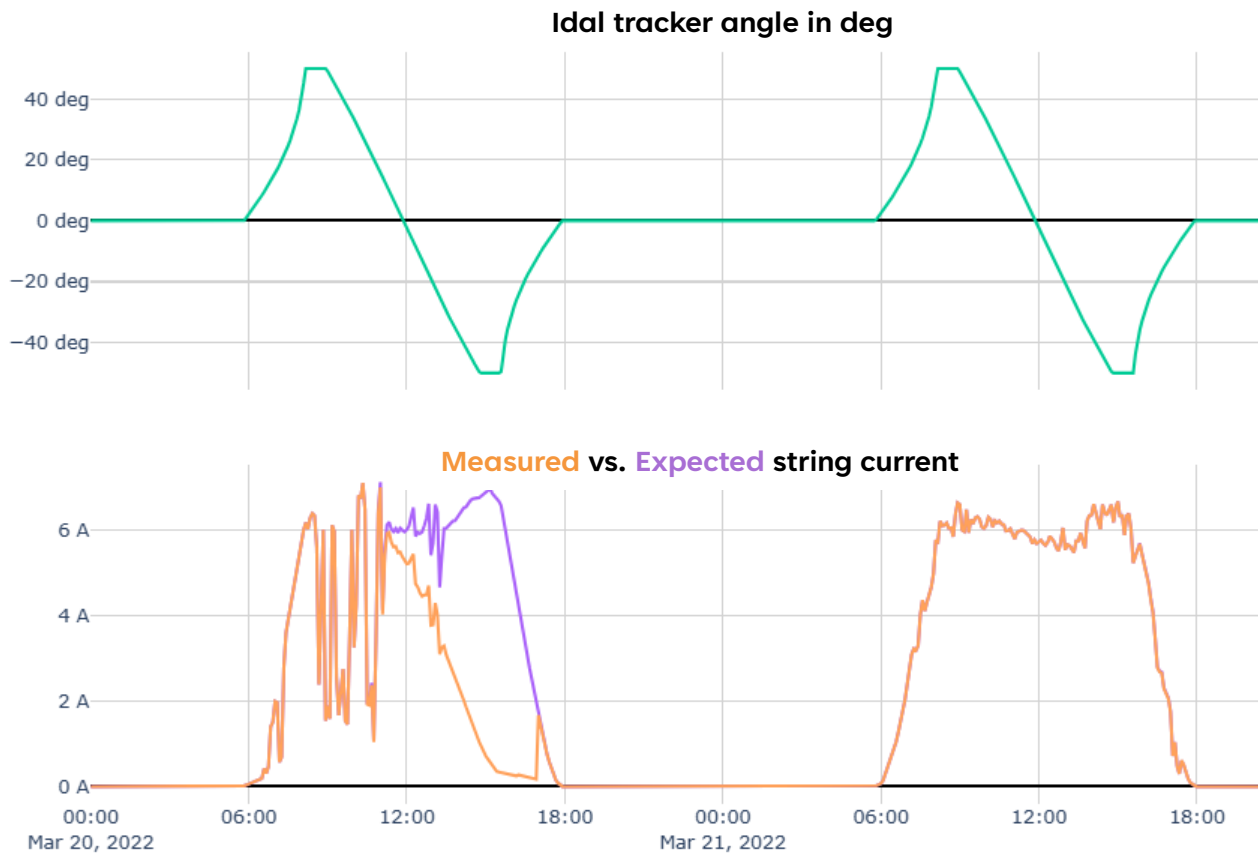


**Goal: Take decision for entire gap**

# CASE 1: COMPARE SCENARIOS WITH MEASUREMENT AND TAKE DECISION



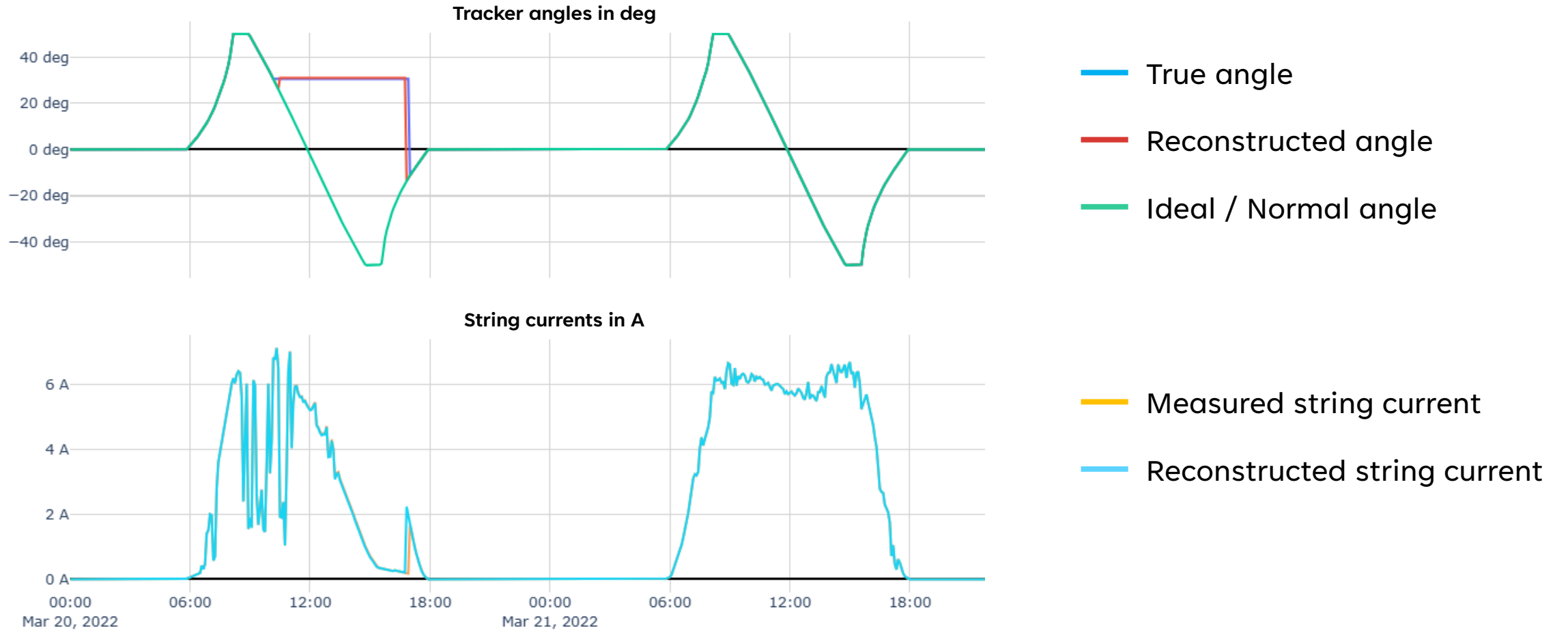
## CASE 2: IS IT POSSIBLE TO RECONSTRUCT A TRACKER ANGLE WITHOUT ANY TRACKER ANGLE MEASUREMENTS?



### Idea:

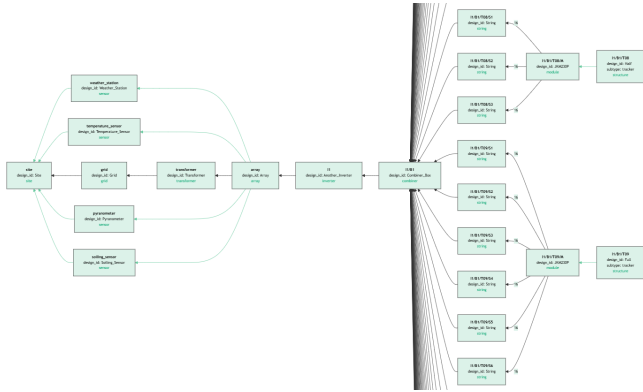
1. Go timestamp by timestamp
2. Compare measured and expected string current.
3. Detect persistent deviations
4. Calculate expected currents for all possible angles
5. For each stuck period, find best matching angle
6. Resume normal tracking once measured and expected behavior align again.

# RECONSTRUCTED TRACKER ANGLES

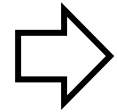


# APPLY TO ENTIRE ASSET MODEL: FULL LOSS WATERFALL

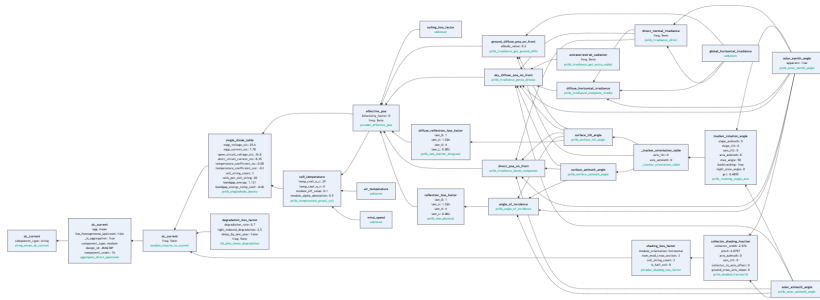
## Component Graph



+

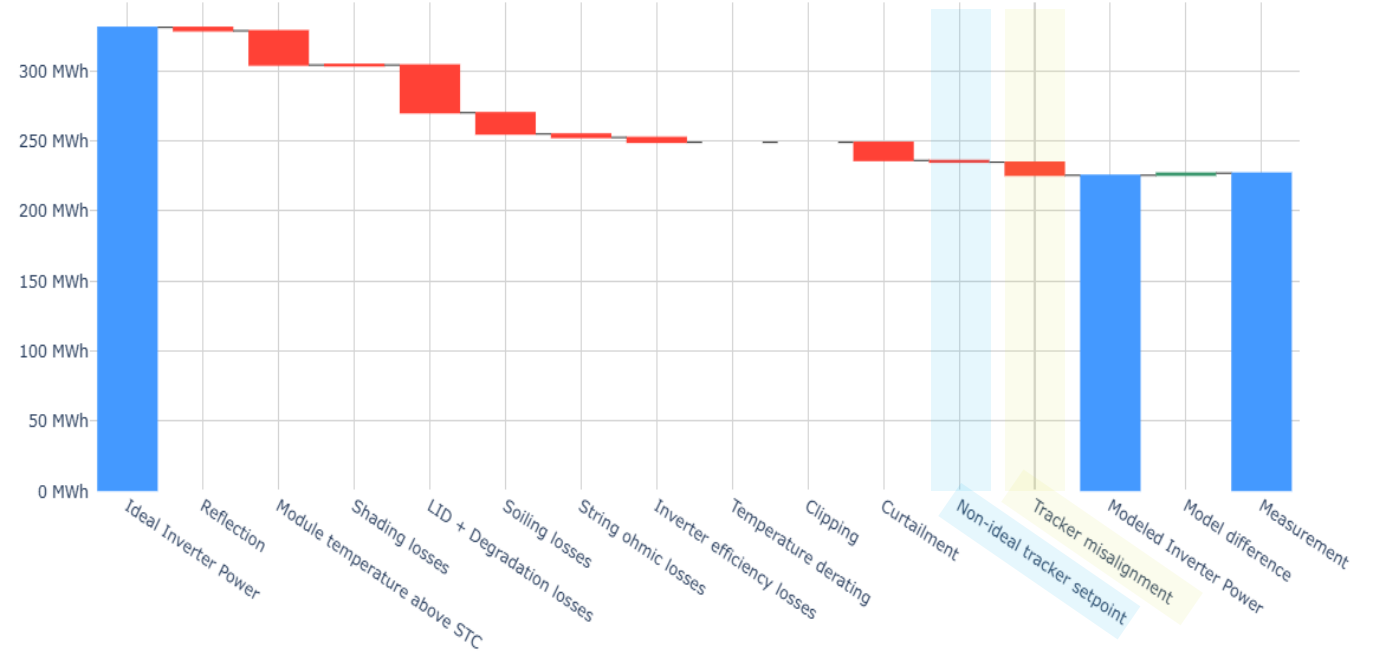


## Model Chain



## User defined loss waterfall

```
site.step_table(
    steps=my_custom_steps
).make_waterfall_table()
Python
```



# SUMMARY OF RESULTS FOR 1-YEAR OF SYNTHETIC DATA

		Case 1	Case 2
Accuracy	% classified correctly	96.8%	96.5%
Precision	% predicted stuck correct	93.6%	100.0%
Recall	% actual stuck correct	89.5%	70.5%
Specificity	% actual normal correct	98.6%	100.0%

**Case 1** Decision based on data gap beign either stuck or normal

**Case 2** Decision for every timestep based on reconstructed tracker angles

**Requirements:** Good quality high resolution data  
Good undestanding of other losses

Would you like to see this approach applied to your data? Contact us!

# START USING PVRADAR TODAY

## Build internal tools

- SDK available on PyPI  
*pip install pvradar-sdk*
- Request free API key  
from [pvradar.com](https://pvradar.com)
- Example notebooks  
available for download
- Component-based modeling  
approach requires additional  
packages, [contact us for a demo](#)

## Model soiling based on our proven soiling model

De-risking soiling from day 1 of development



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Thank you!



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