

SFERA-III

Solar Facilities for the European Research Area



“Towards a fully automated flux density prediction using
data driven models”

Max Pargmann, German Aerospace Center (DLR)

NETWORKING

Summer School: “Smart CSP: How Smart Tools, Devices, and Software can help improve the Design and Operation of Concentrating Solar Power Technologies” - WP1 Capacity building and training activities - Cologne, Germany, September 14th-15th 2023



THIS PROJECT HAS RECEIVED FUNDING FROM THE EUROPEAN UNION'S HORIZON 2020 RESEARCH AND INNOVATION PROGRAMME UNDER GRANT AGREEMENT NO **823802**



- Motivation
- Methods: Differentiable Raytracing
- Three Foundations of Automation
 - Diagnosis
 - Control
 - Prediction
- Discussion

SFERA-III

Solar Facilities for the European Research Area

Motivation

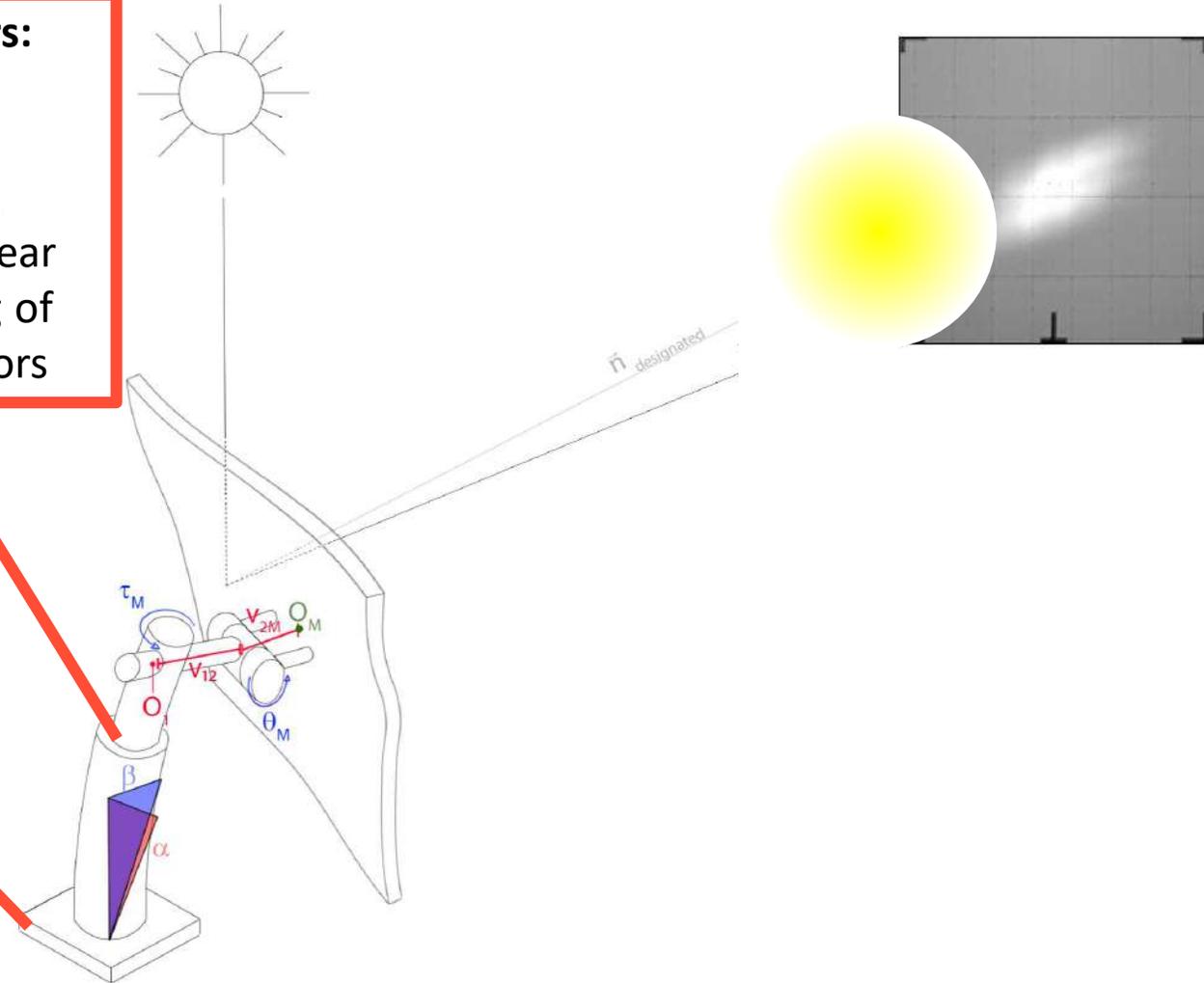






Alignment errors:

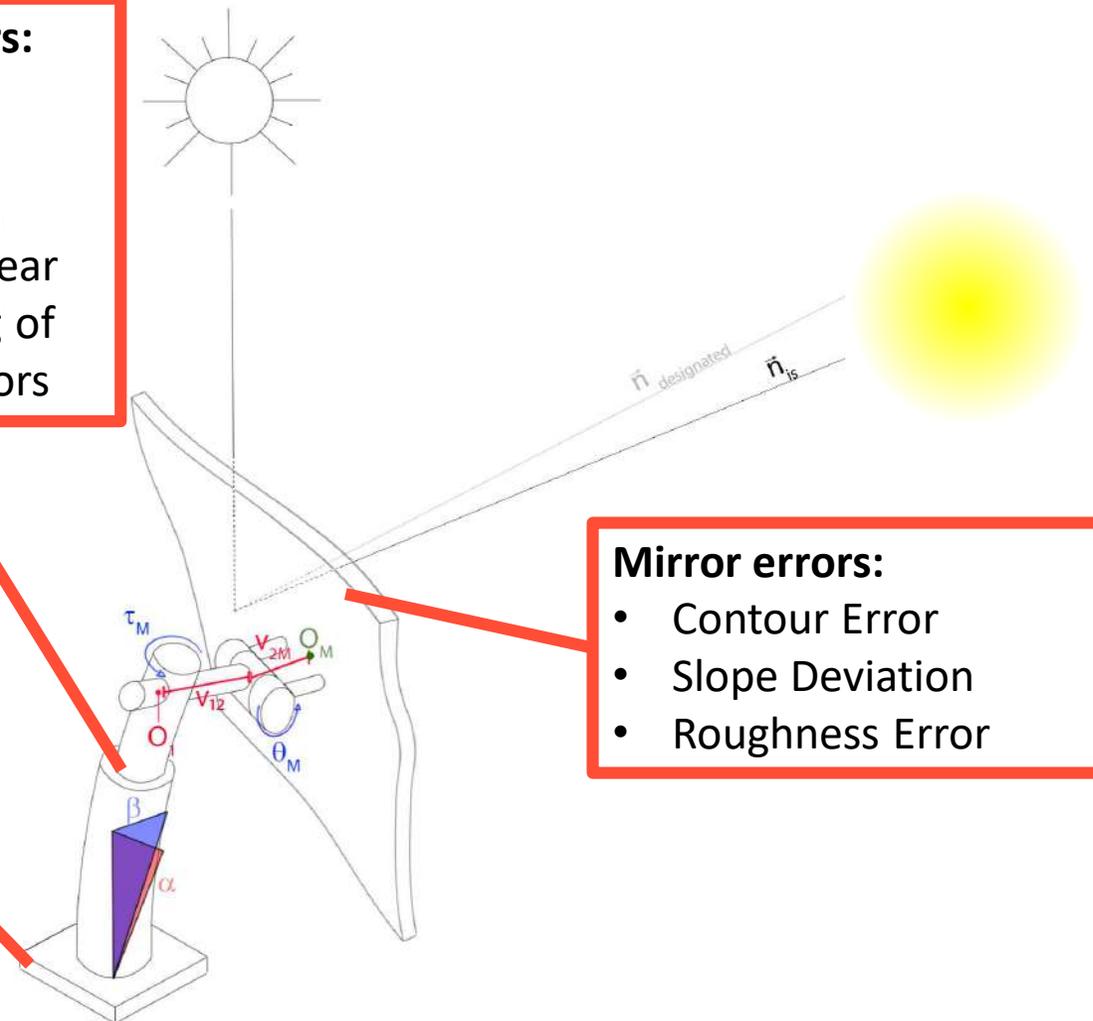
- Wind load
- Rigidity
- Sag
- Backlash / Wear
- Misscounting of Stepper Motors





Alignment errors:

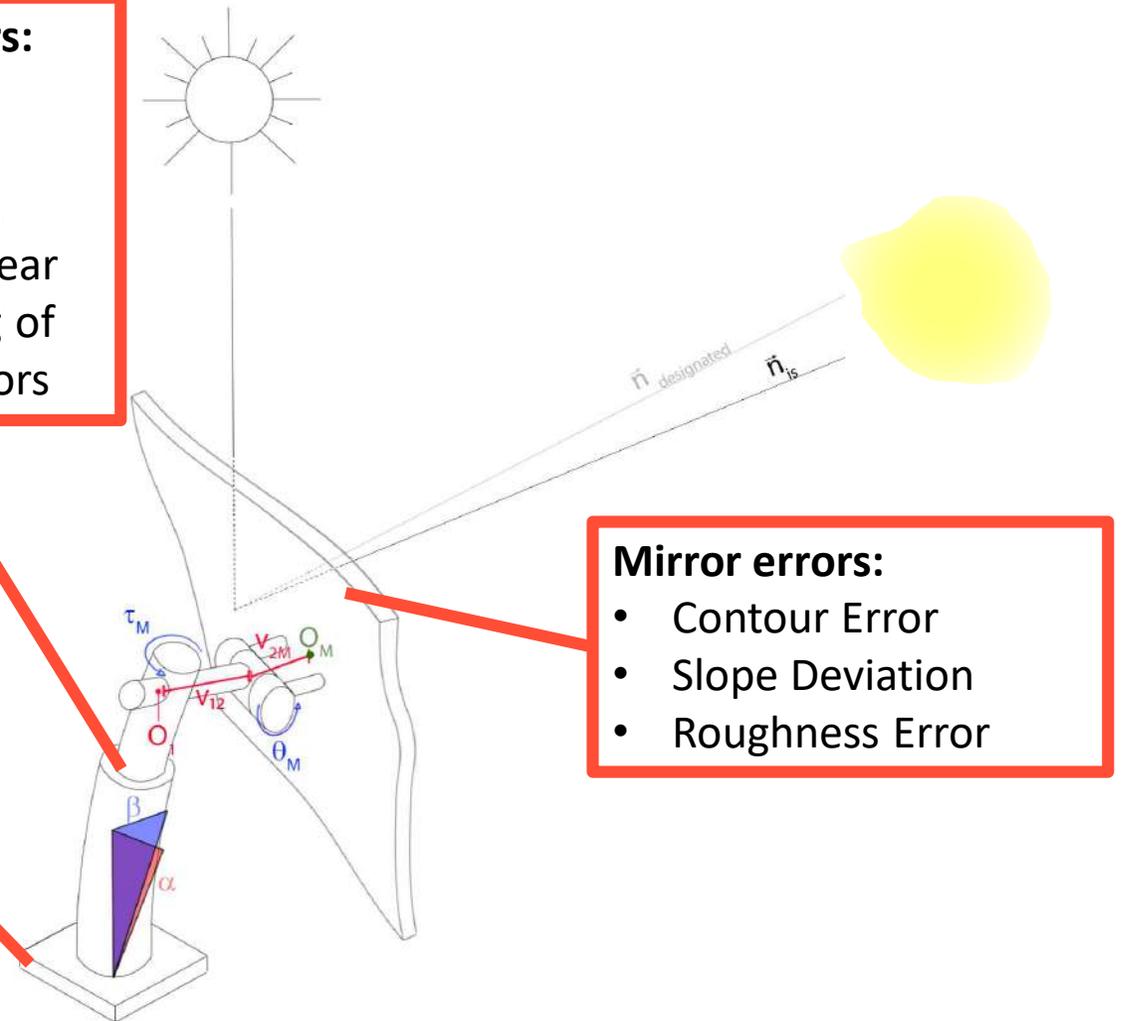
- Wind load
- Rigidity
- Sag
- Backlash / Wear
- Misscounting of Stepper Motors





Alignment errors:

- Wind load
- Rigidity
- Sag
- Backlash / Wear
- Misscounting of Stepper Motors



Mirror errors:

- Contour Error
- Slope Deviation
- Roughness Error



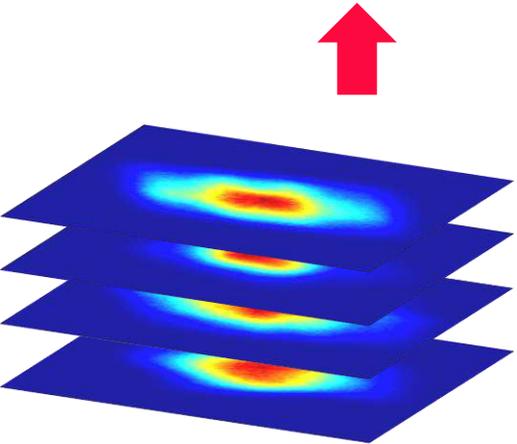
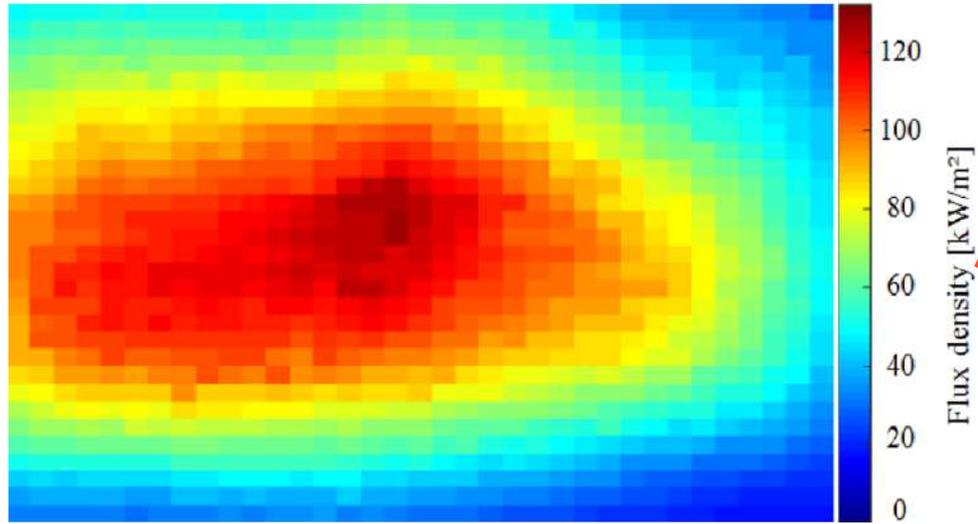
Alignment errors

- Wind load
- Rigidity
- Sag
- Backlash / V
- Misscounting
Stepper Mo



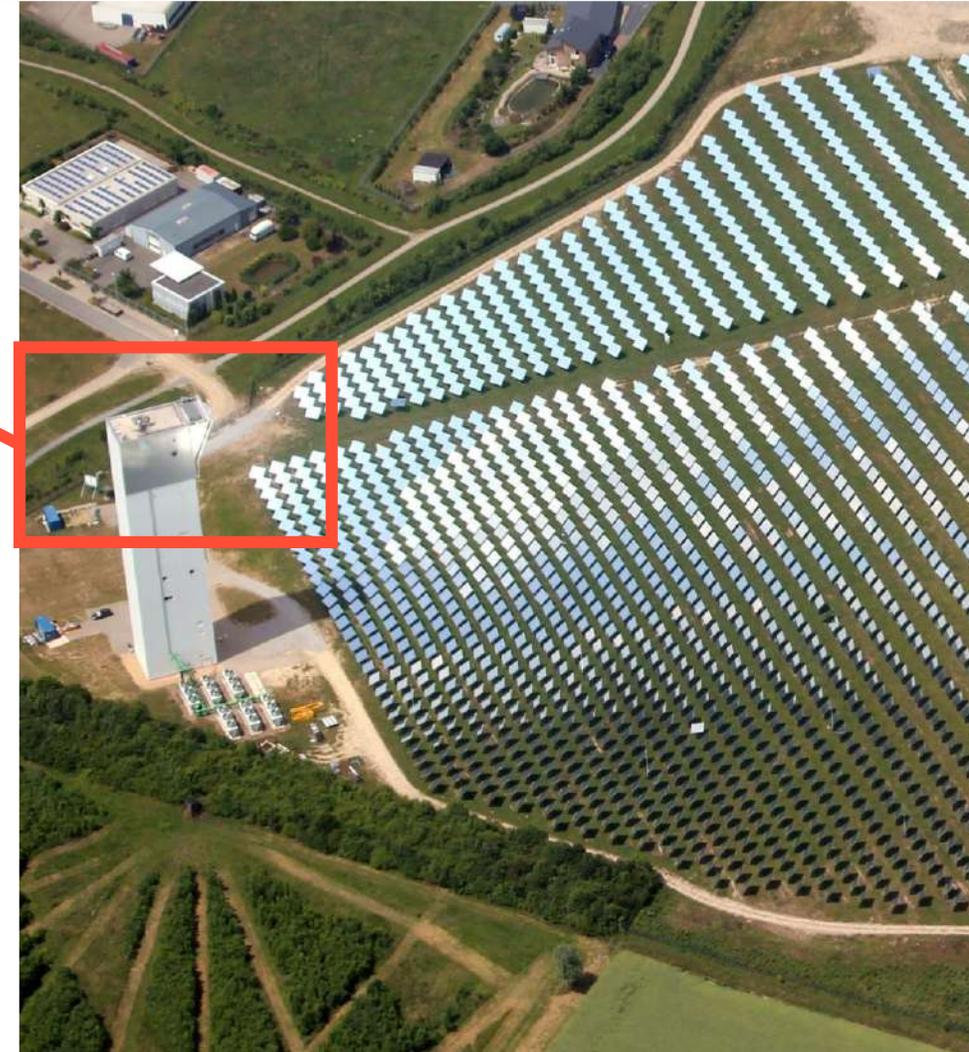
Alignment errors

- Wind load
- Rigidity
- Sag
- Backlash / V
- Misscounting
Stepper Mo



Difficulties:

- Unique Pointing Command
- Quality of Calibration
- Command Delay
- No Sensors
- No Measurement





Diagnosis

Heliostat soiling;
changes in
alignment and
mirror curvature;
Receiver
heat spikes



Control

Aimpoint
Assignment/Opti-
mization; in
dependency of;
Lucky Calibration,
Mirror
Defomrations



Prediction

Optimized control
strategies for
upcoming
weather
condtions



Diagnosis

Heliostat soiling;
changes in
alignment and
mirror curvature;
Receiver
heat spikes



Control

Aimpoint
Assignment/Opti-
mization; in
dependency of;
Lucky Calibration,
Mirror
Defomrations



Prediction

Optimized control
strategies for
upcoming
weather
condtions



Diagnosis

Heliostat soiling;
changes in
alignment and
mirror curvature;
Receiver
heat spikes



Control

Aimpoint
Assignment/Opti-
mization; in
dependency of;
Lucky Calibration,
Mirror
Defomrations



Prediction

Optimized control
strategies for
upcoming
weather
condtions

Difficulties:

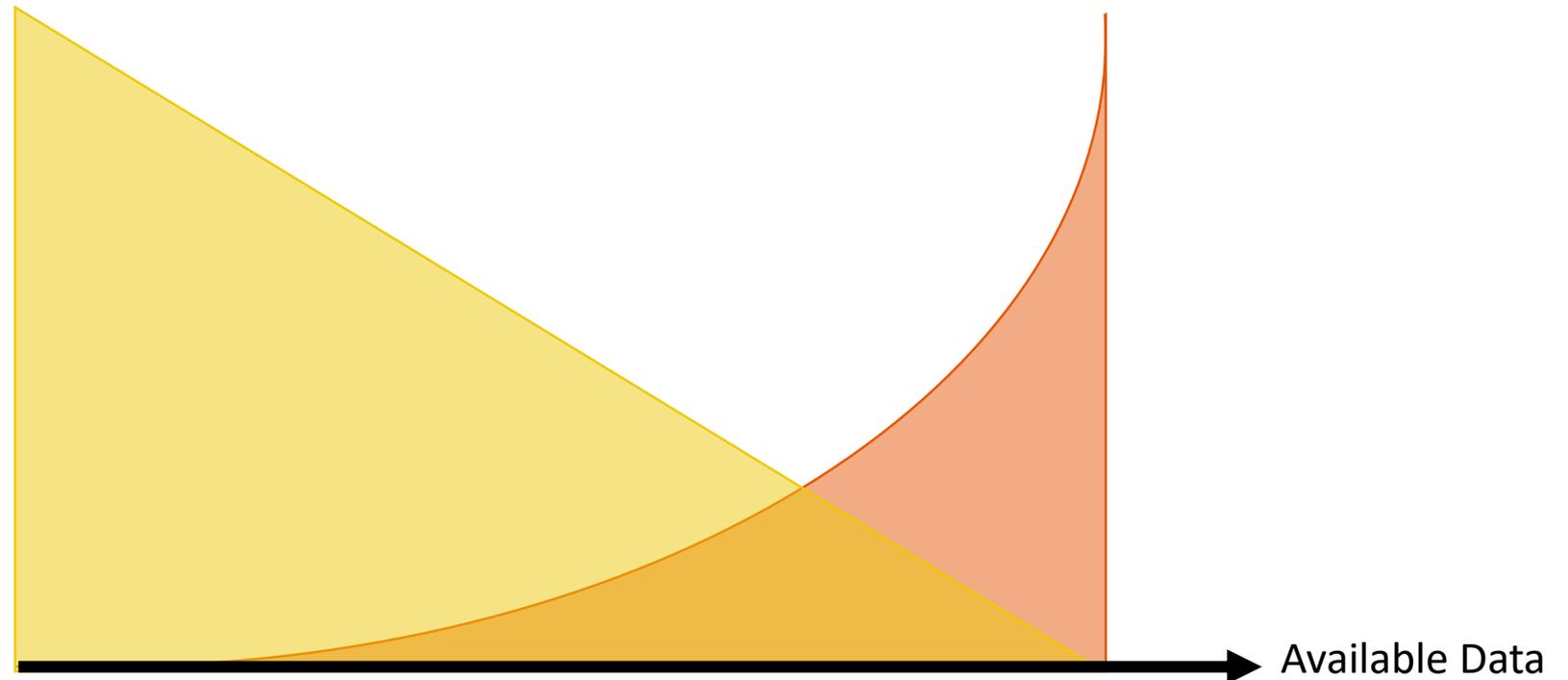
- Open Loop Control
- Measurements for
 - Calibration are slow
 - Mirror deformations are unreliable and cost intensive
 - Flux measurements are experimental[1] or cost intensive

Solution:

- Closed Loop Control
- Install More Sensors

Model Error

Cost



[1] Offergeld, Matthias, et al. "Flux density measurement for industrial-scale solar power towers using the reflection off the absorber." *AIP Conference Proceedings*. Vol. 2126. No. 1. AIP Publishing, 2019.

Difficulties:

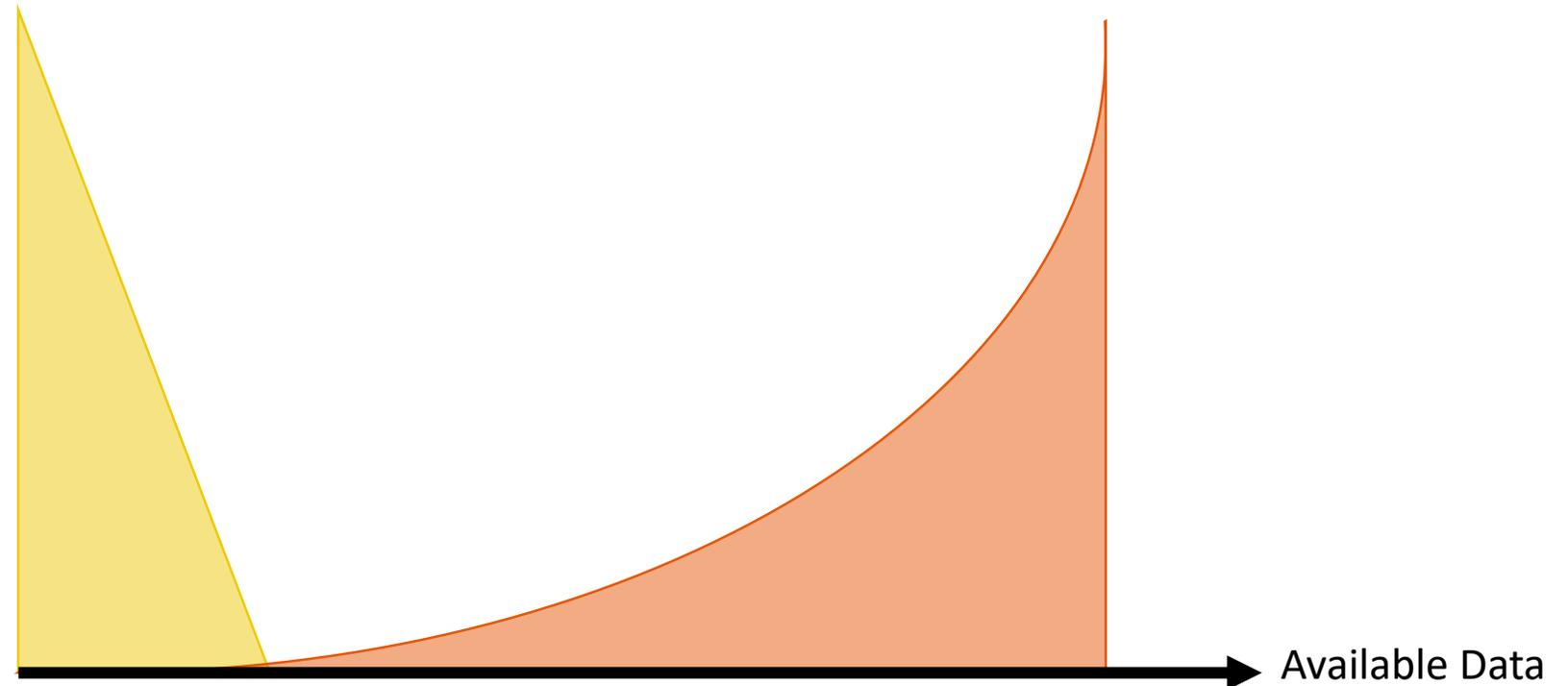
- Open Loop Control
- Measurements for
 - Calibration are slow
 - Mirror deformations are unreliable and cost intensive
 - Flux measurements are experimental[1] or cost intensive

Solution:

- Closed Loop Control
- Install More Sensors
- **Use Modern Algorithms**

Model Error

Cost



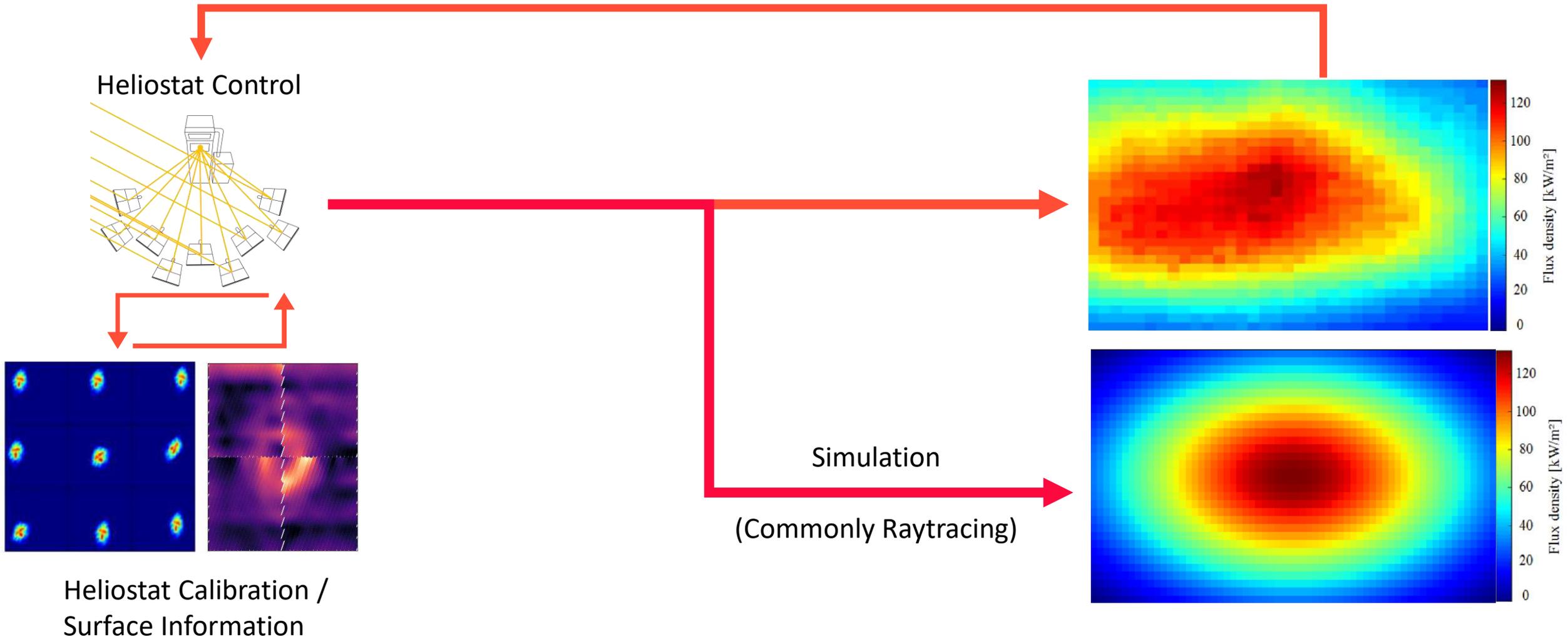
[1] Offergeld, Matthias, et al. "Flux density measurement for industrial-scale solar power towers using the reflection off the absorber." *AIP Conference Proceedings*. Vol. 2126. No. 1. AIP Publishing, 2019.

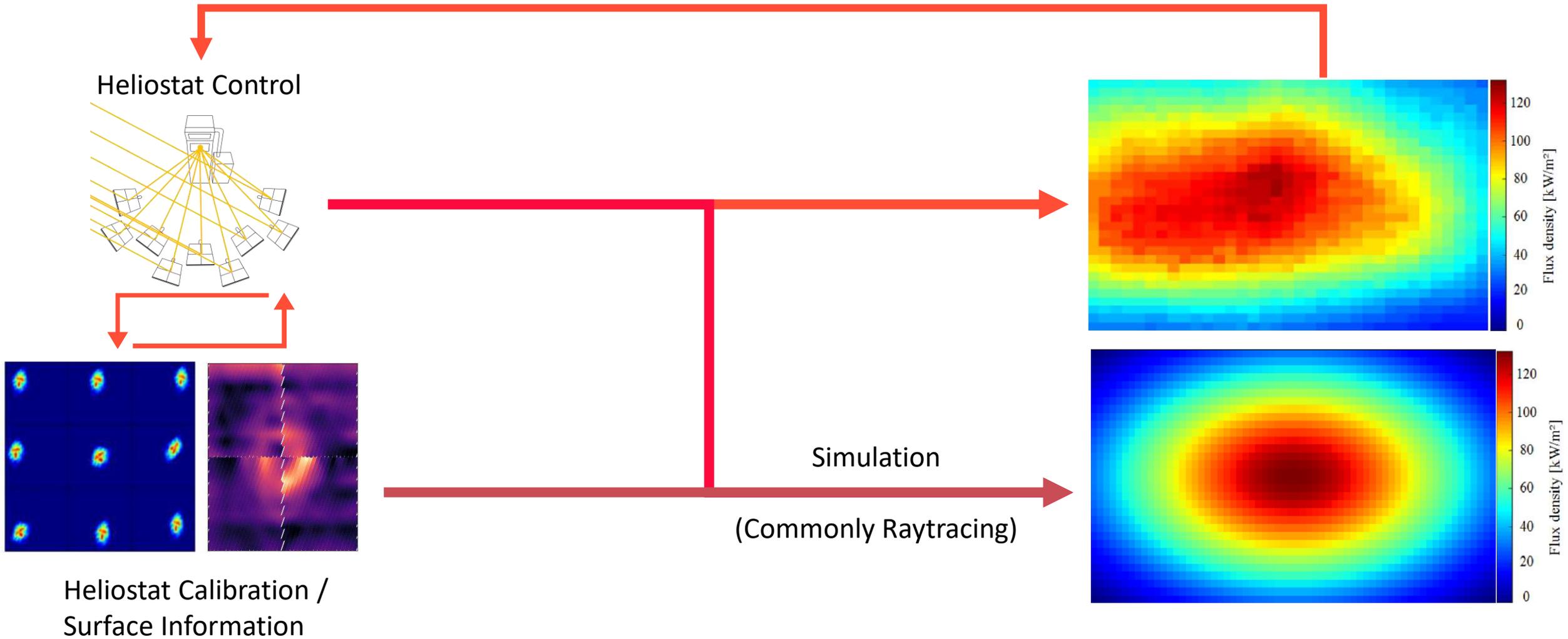
SFERA-III

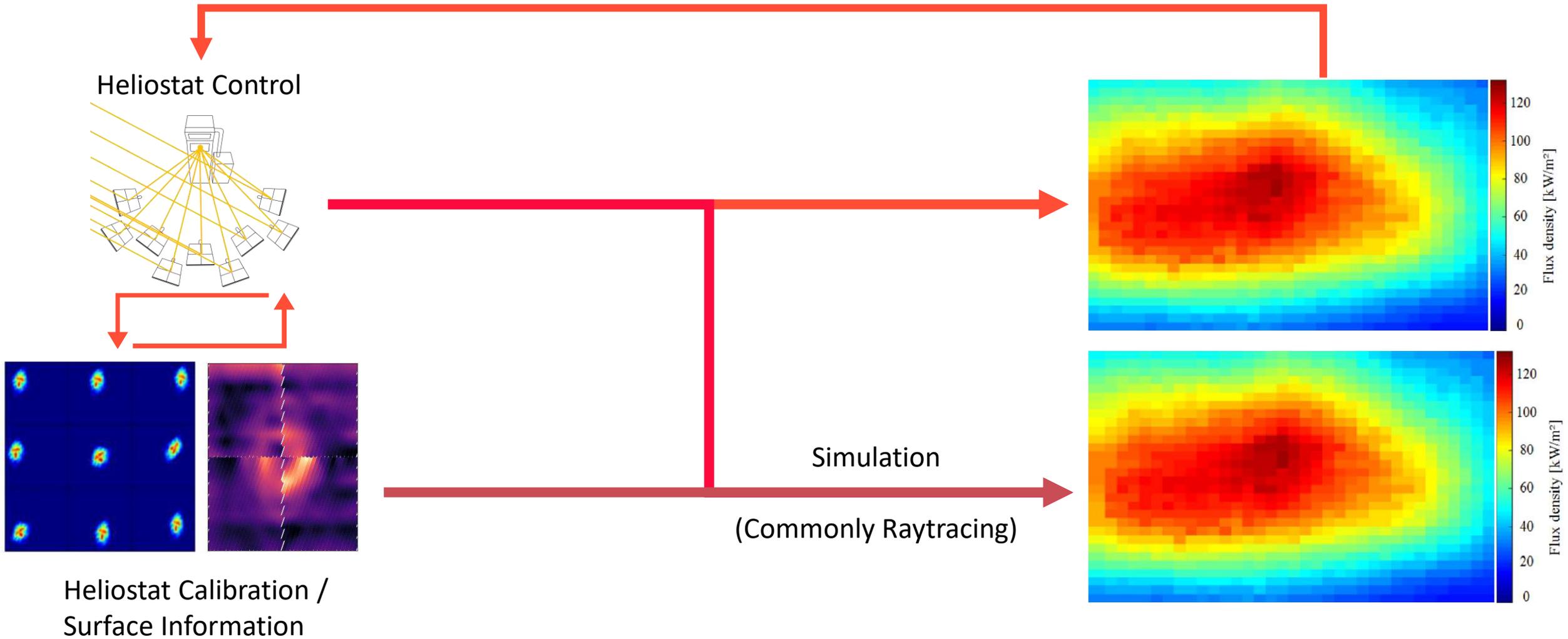
Solar Facilities for the European Research Area

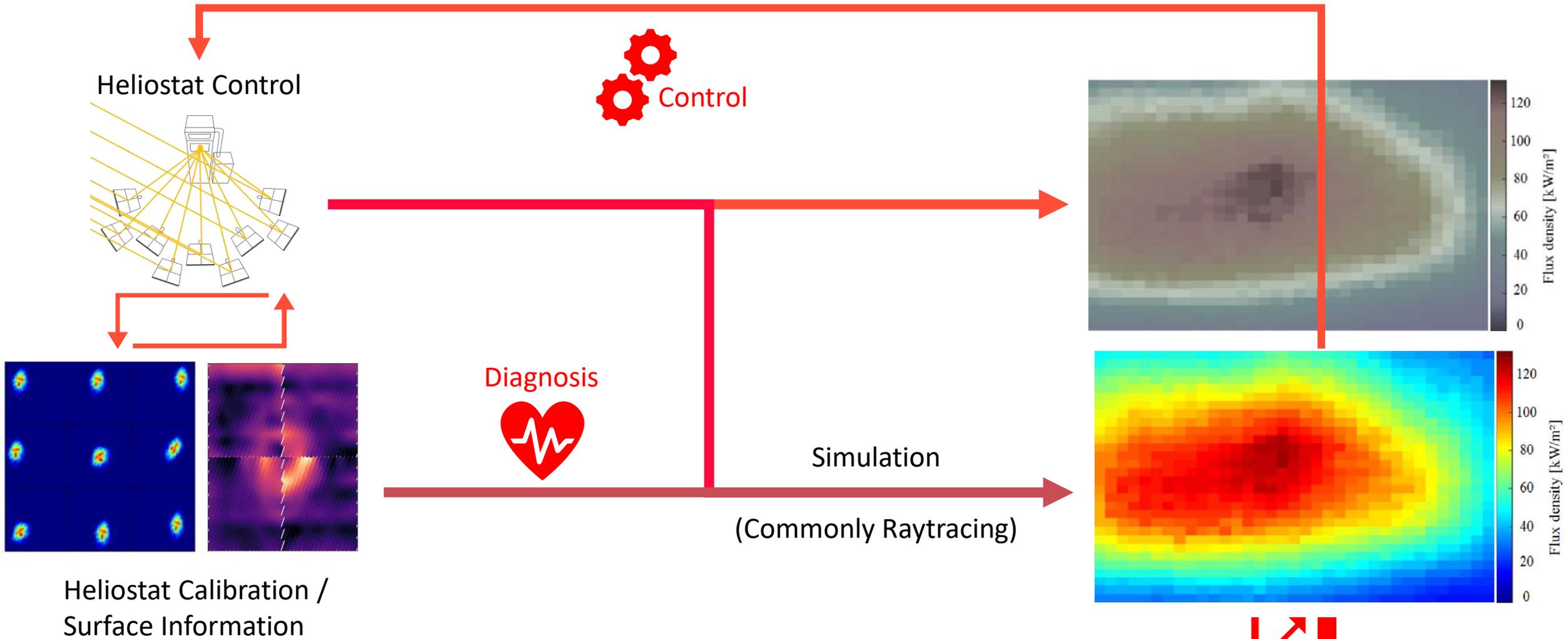
Methods





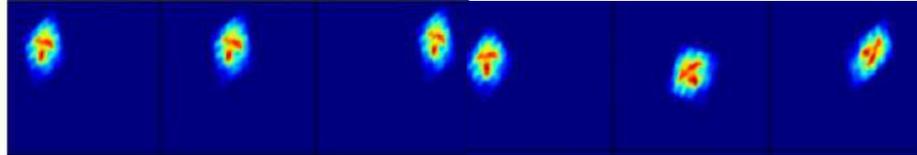




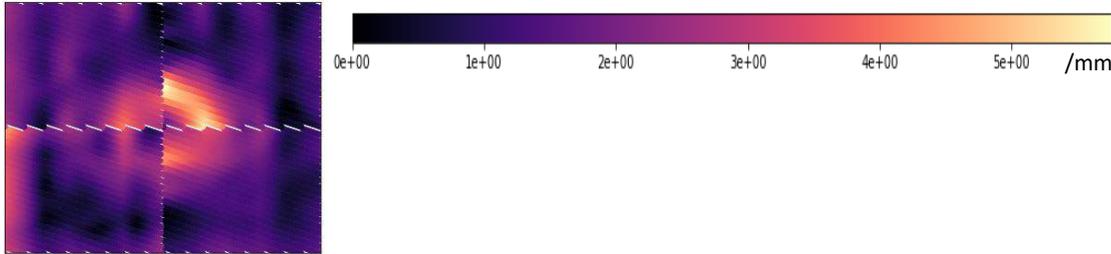


In a commercial power plant, we have access to a maximum of:

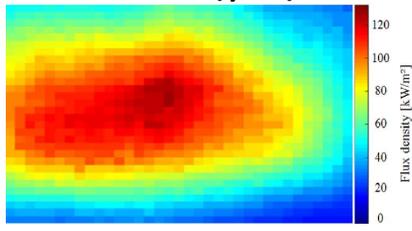
Calibration Data → ca 10-100 per Heliostat



Surface Information → A few dozen

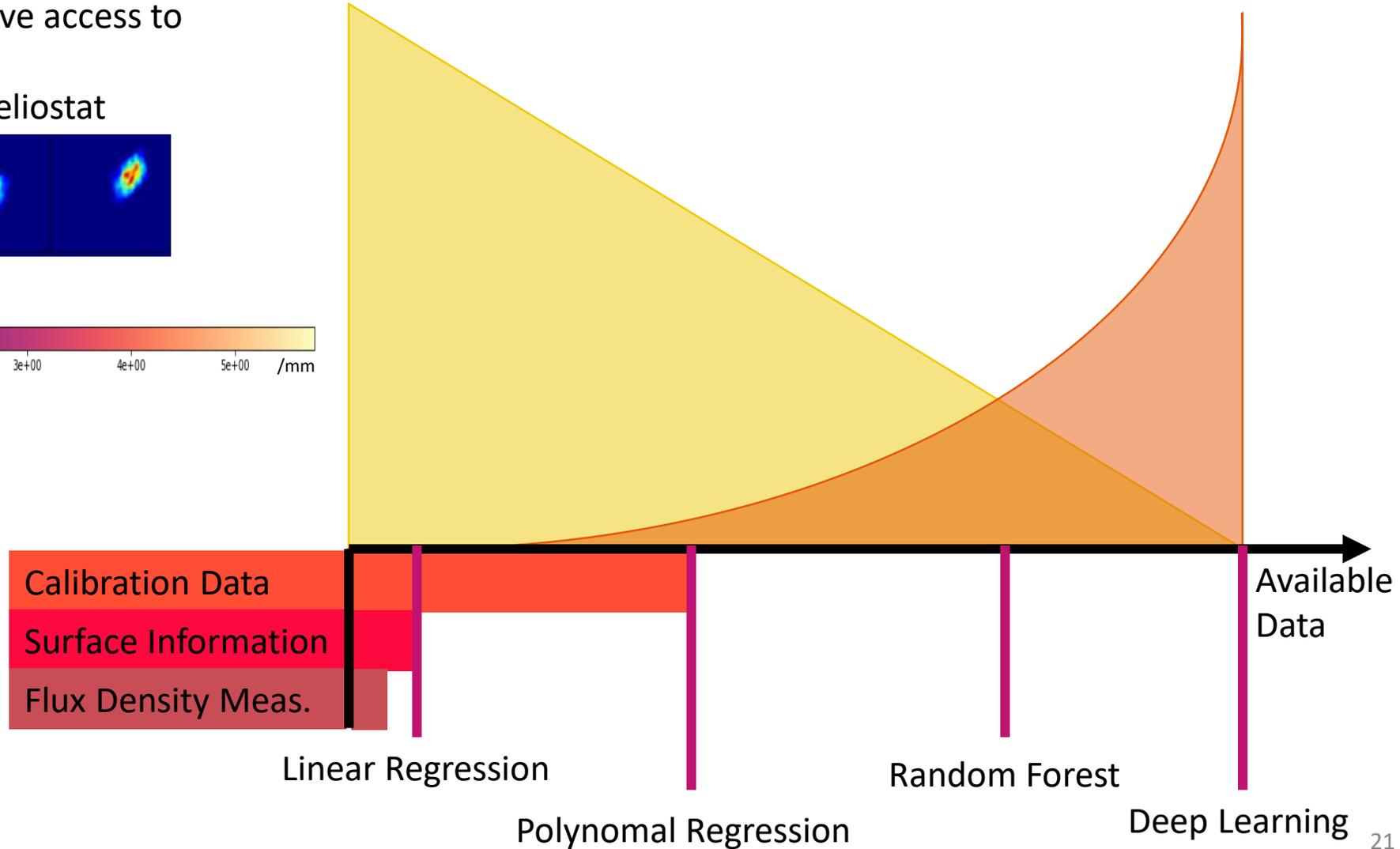


Flux Density Measurements → None (yet)



Model Error

Cost



Calibration Data

Surface Information

Flux Density Meas.

Linear Regression

Polynomial Regression

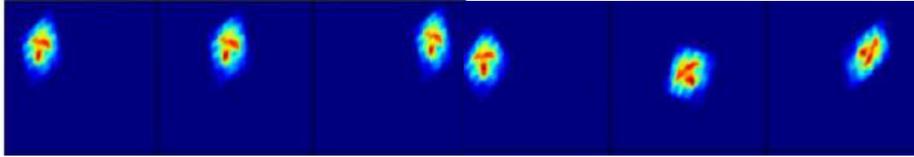
Random Forest

Deep Learning

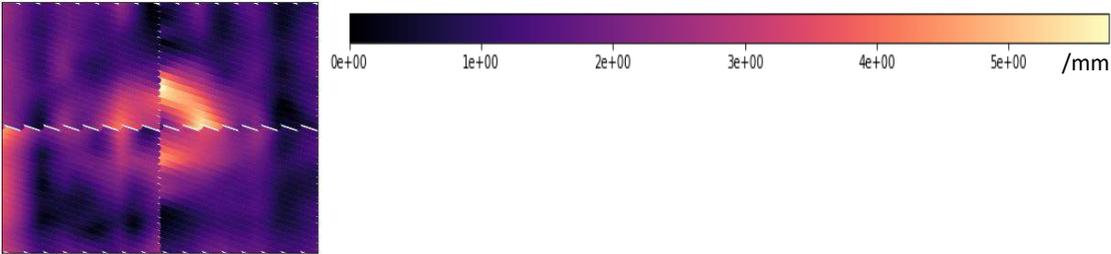
Available Data

In a commercial power plant, we have access to a maximum of:

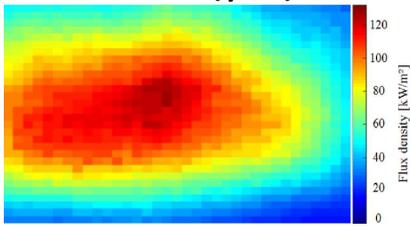
Calibration Data → ca 10-100 per Heliostat



Surface Information → A few dozen

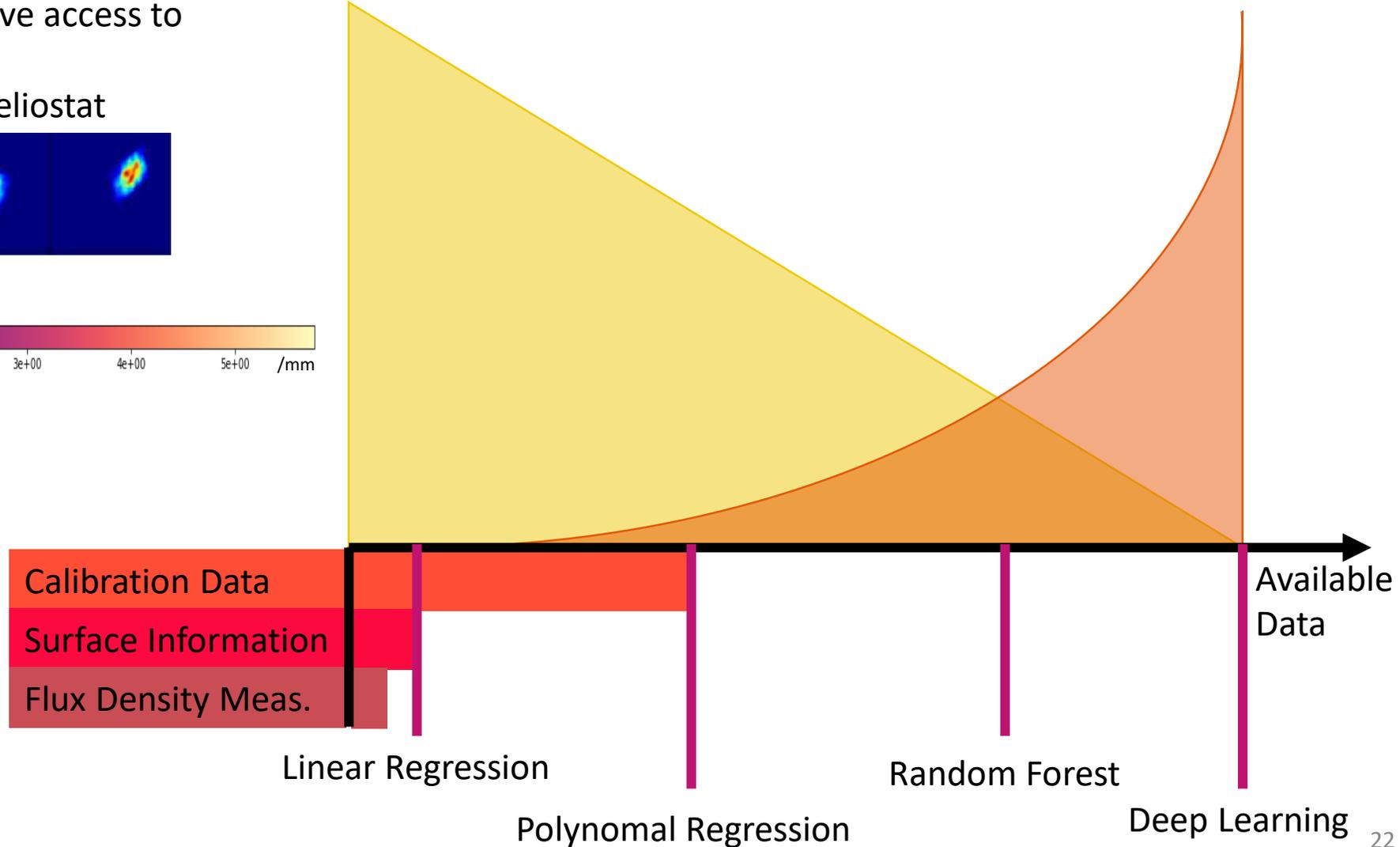


Flux Density Measurements → None (yet)



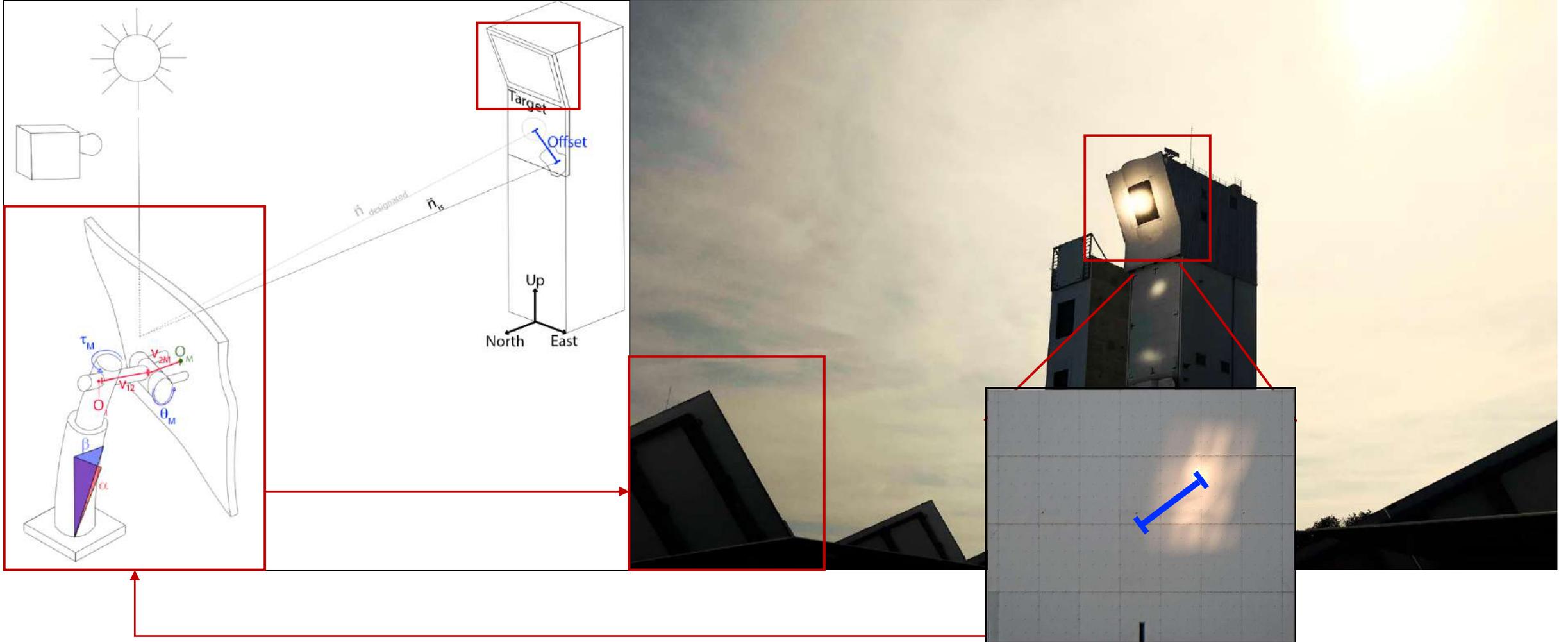
Model Error

Cost



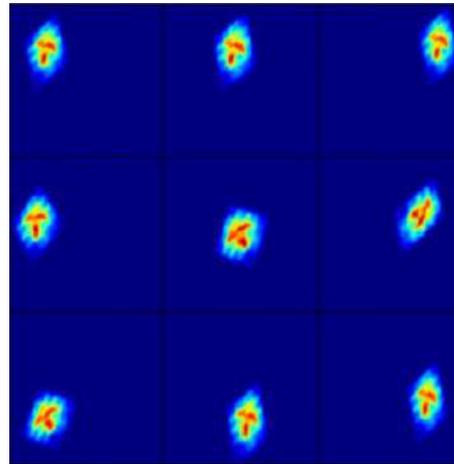
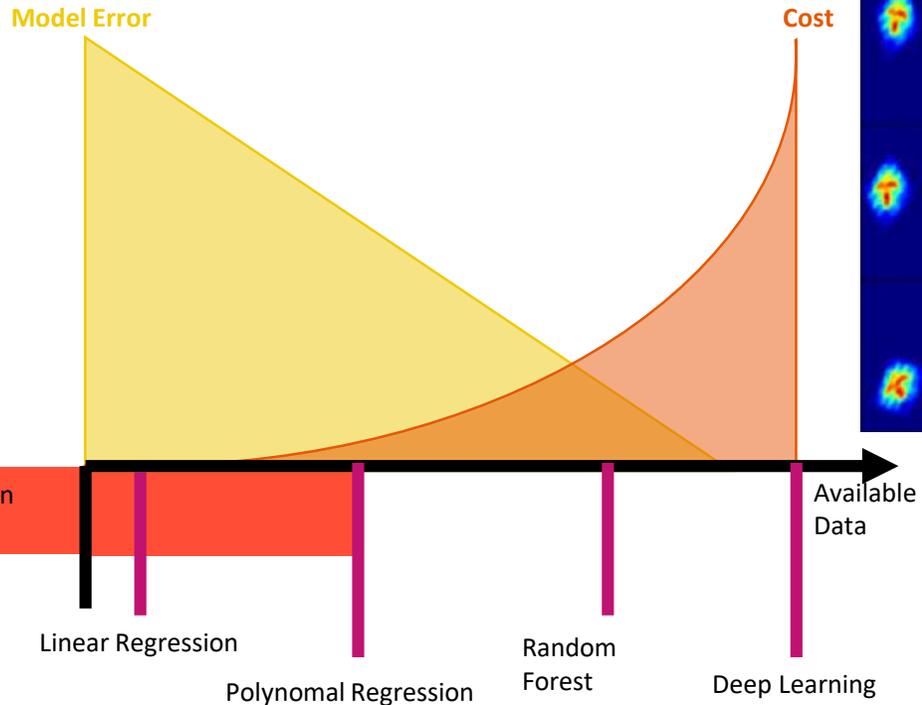
Digital Instance

Real Object



How much information about each heliostat can be deduced from field data?

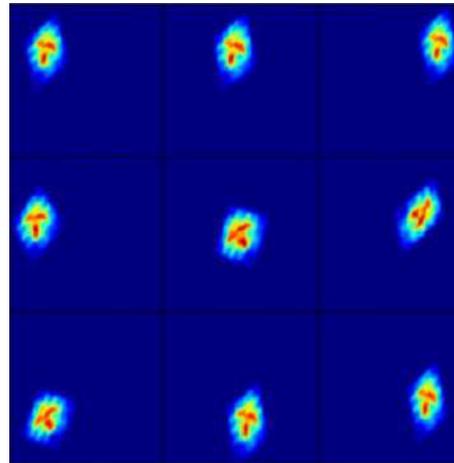
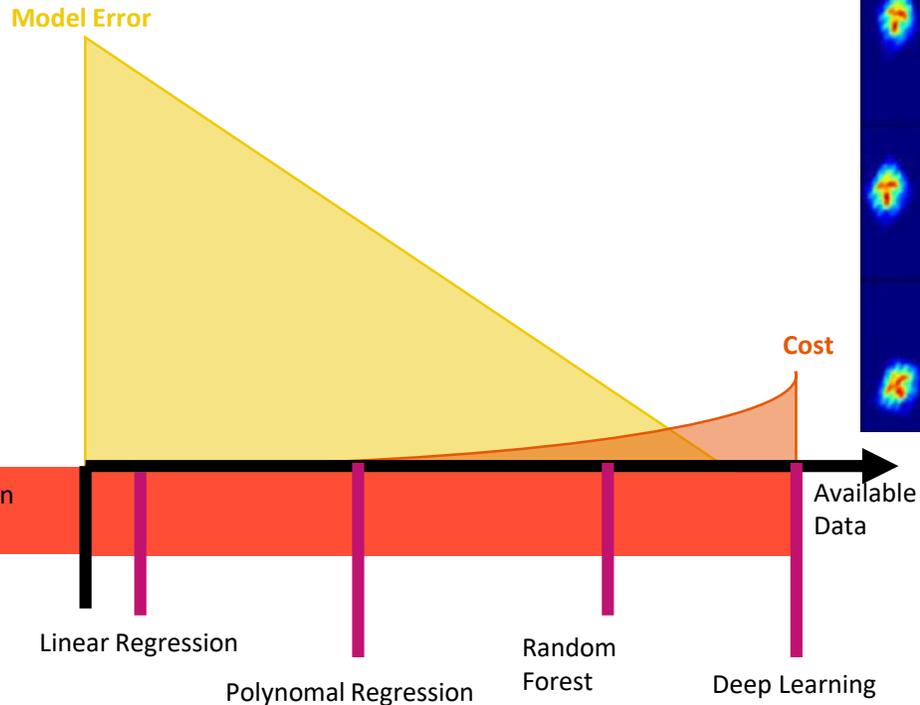
GANCSTR | ARTIST



How much information about each heliostat can be deduced from field data?

GANCSTR | ARTIST

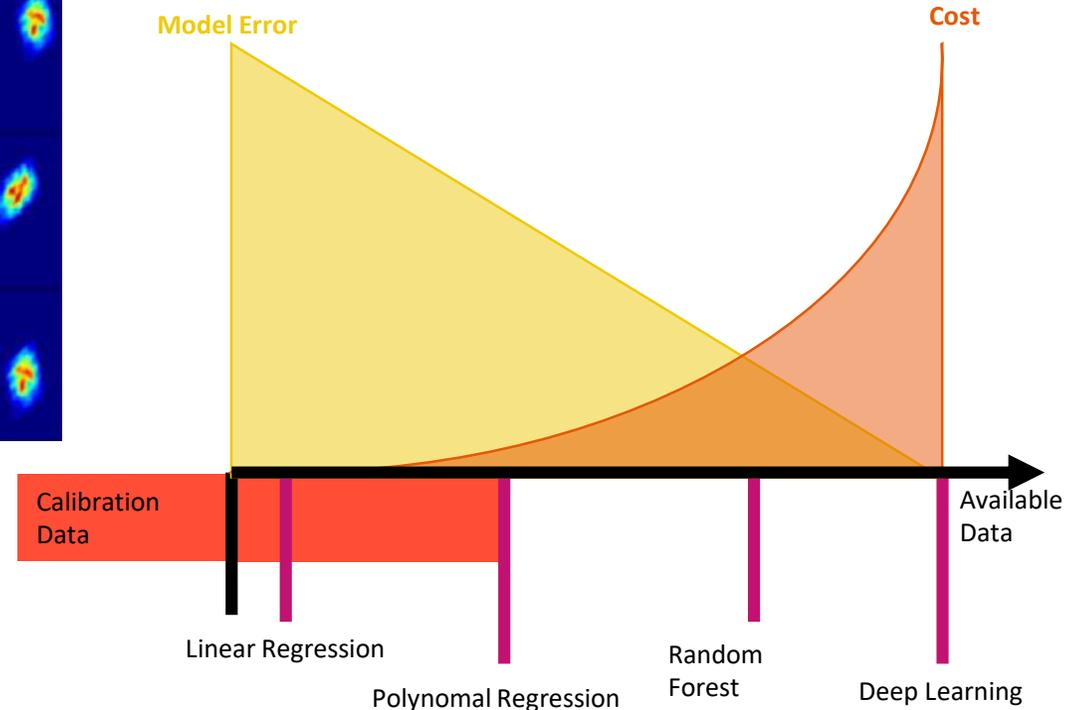
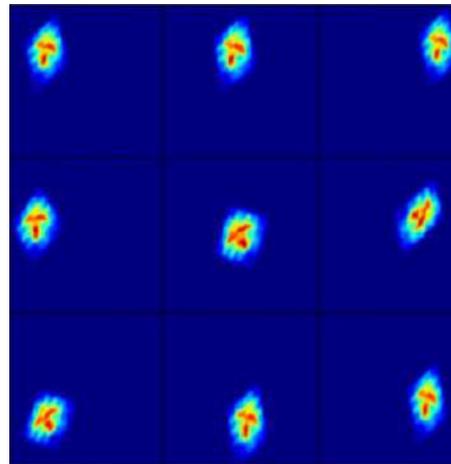
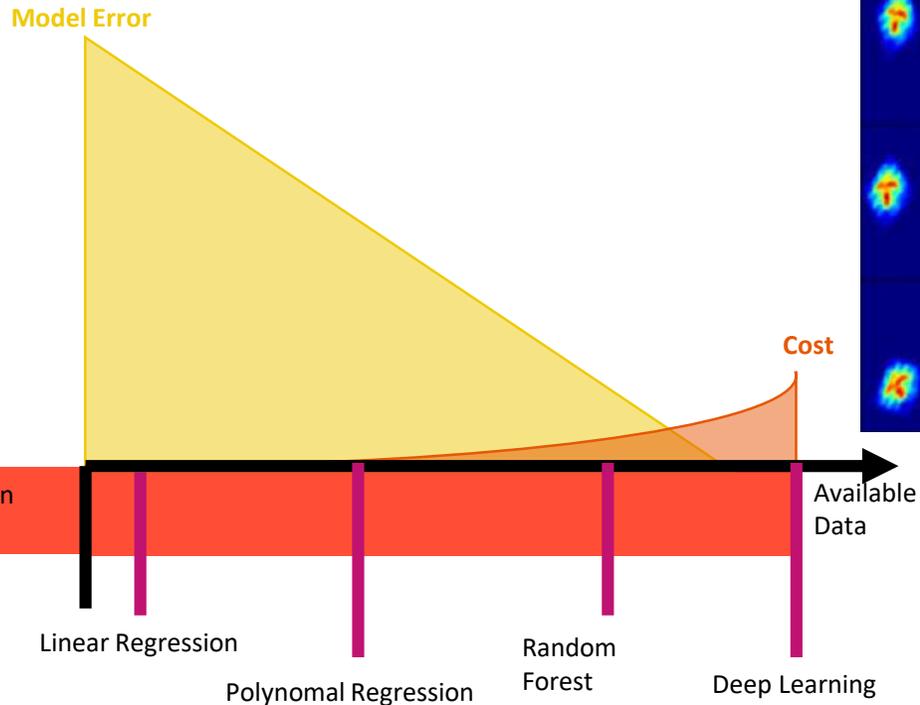
How much information can be deduced with as few data as possible?



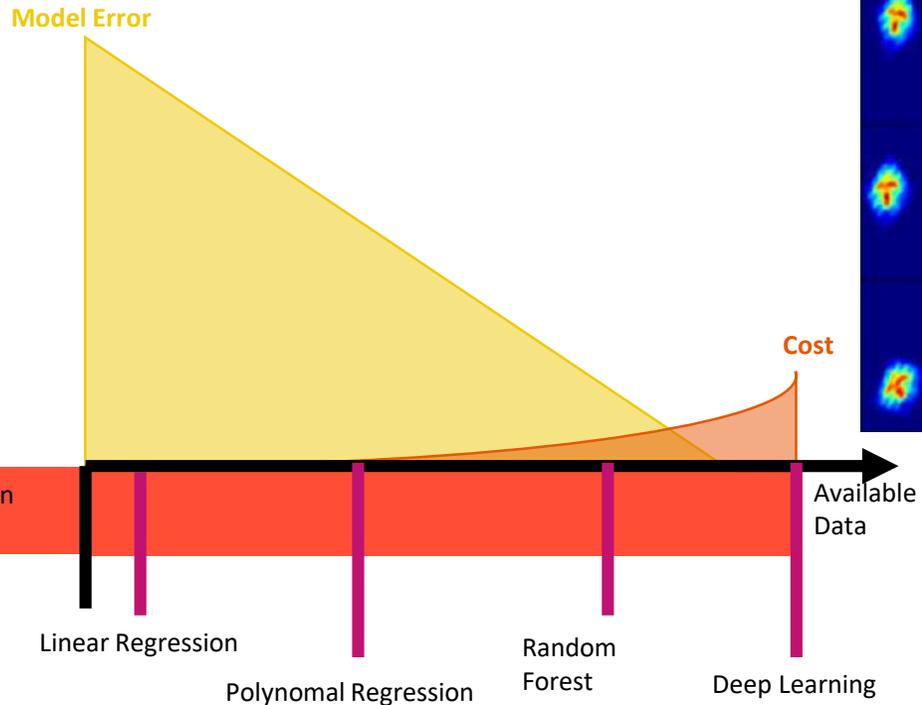
GANCSTR | ARTIST

How much information about each heliostat can be deduced from field data?

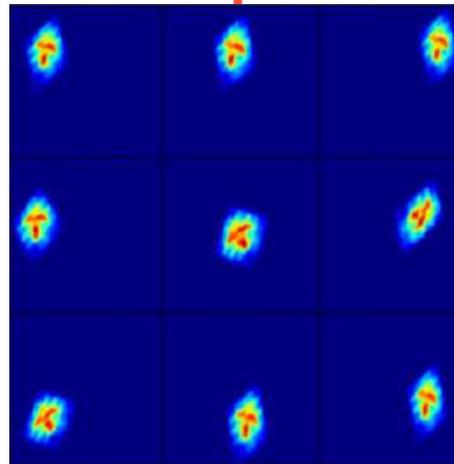
How much information can be deduced with as few data as possible?



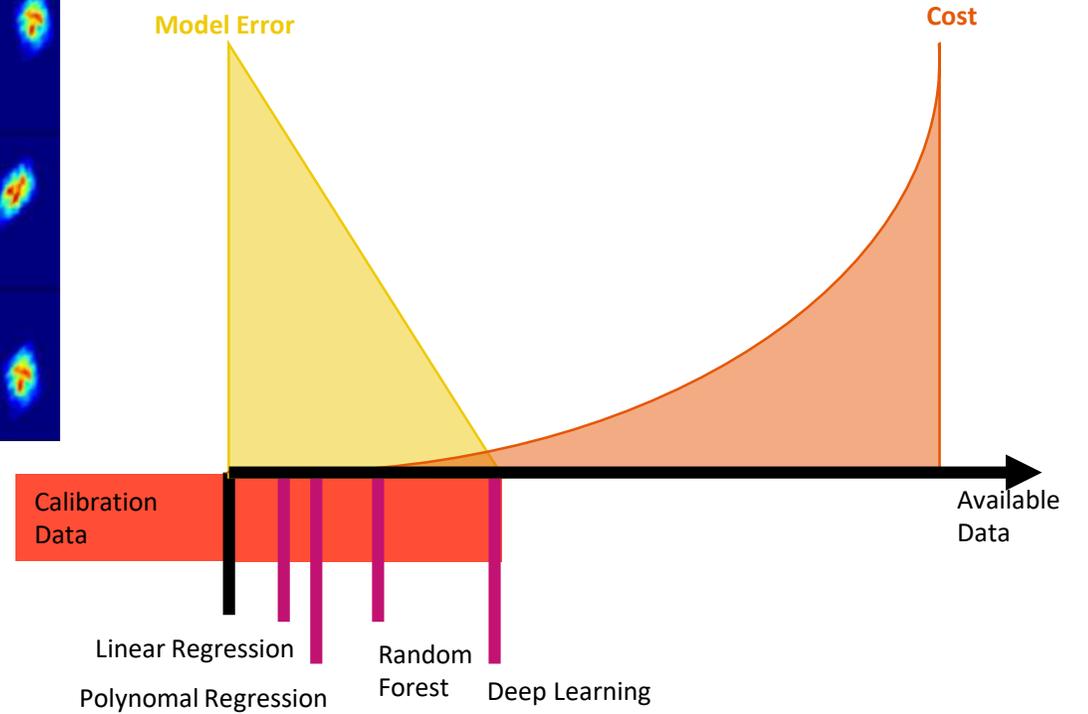
How much information about each heliostat can be deduced from field data?



GANCSTR ARTIST



How much information can be deduced with as few data as possible?

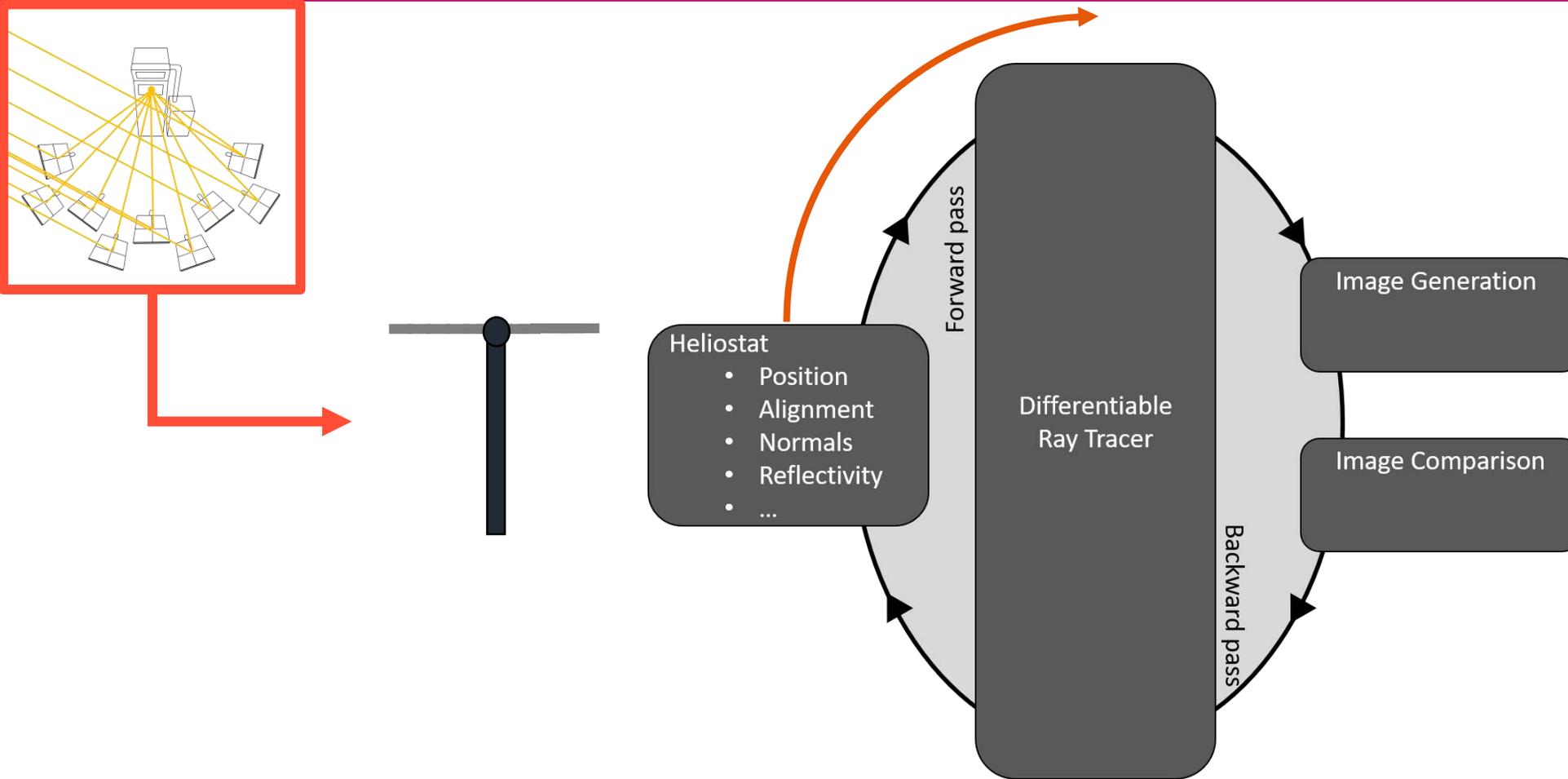


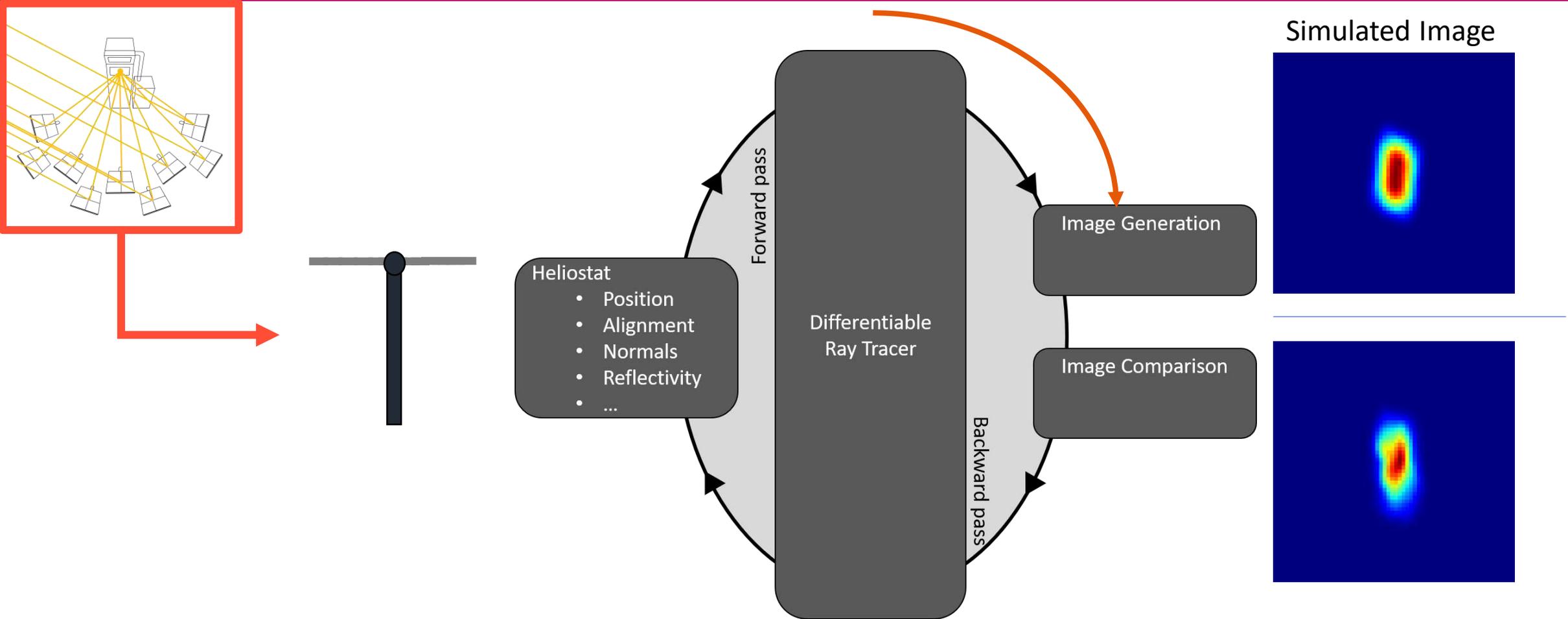
SFERA-III

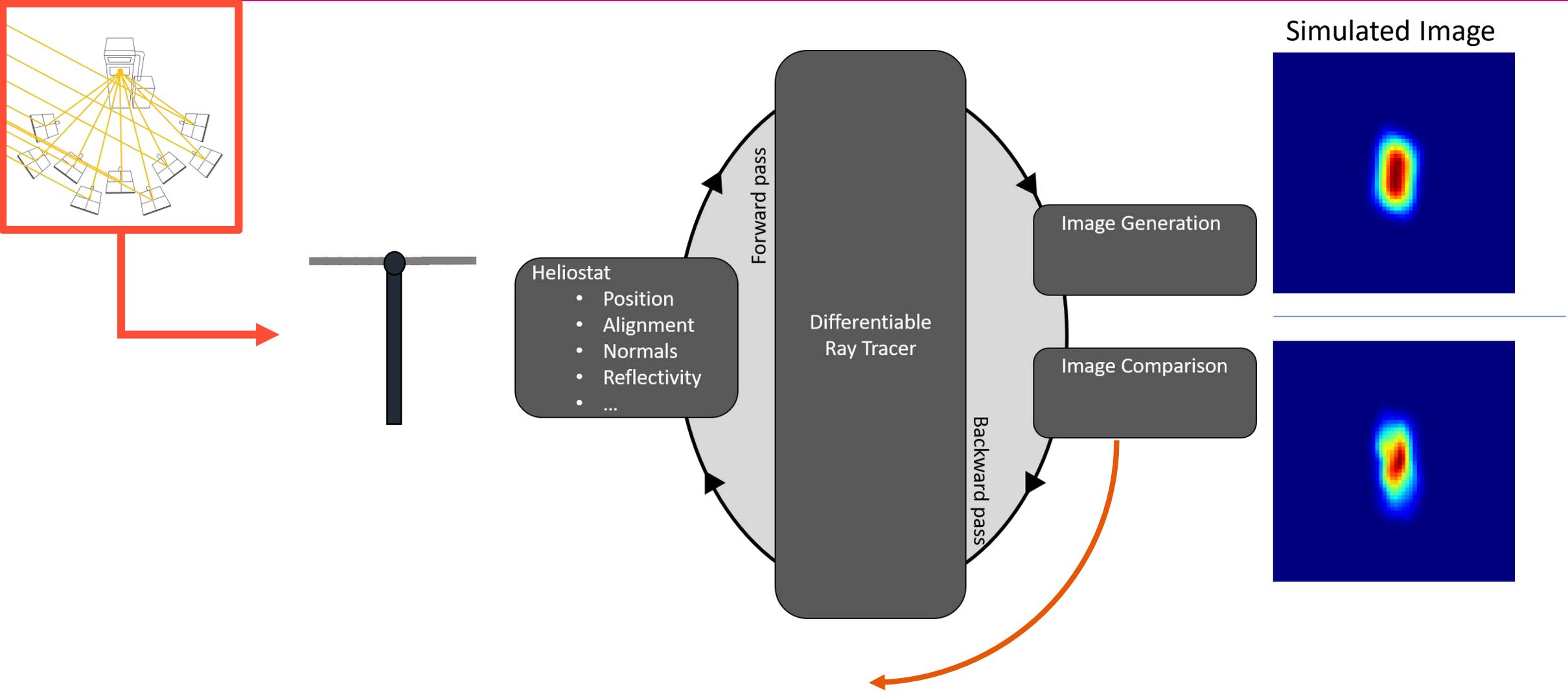
Solar Facilities for the European Research Area

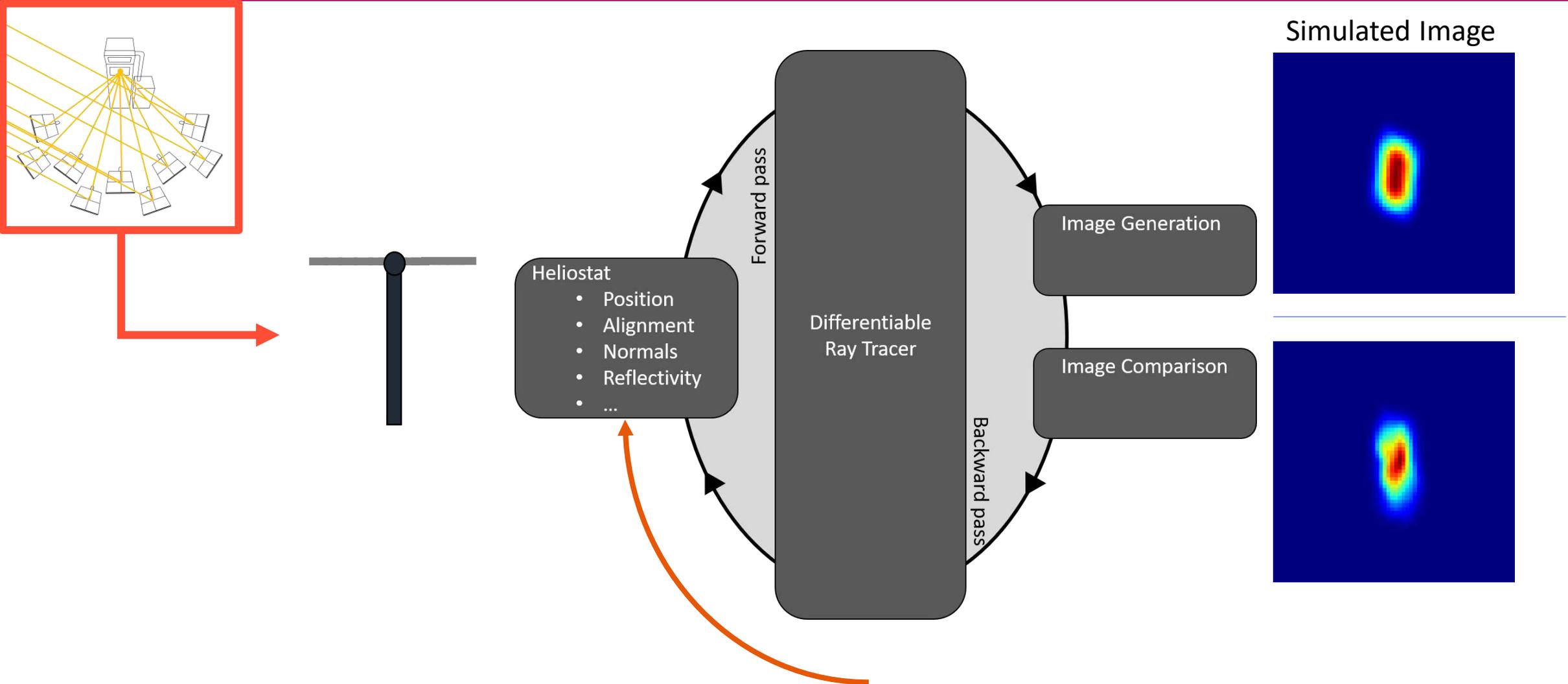
Methods: Differentiable Raytracing

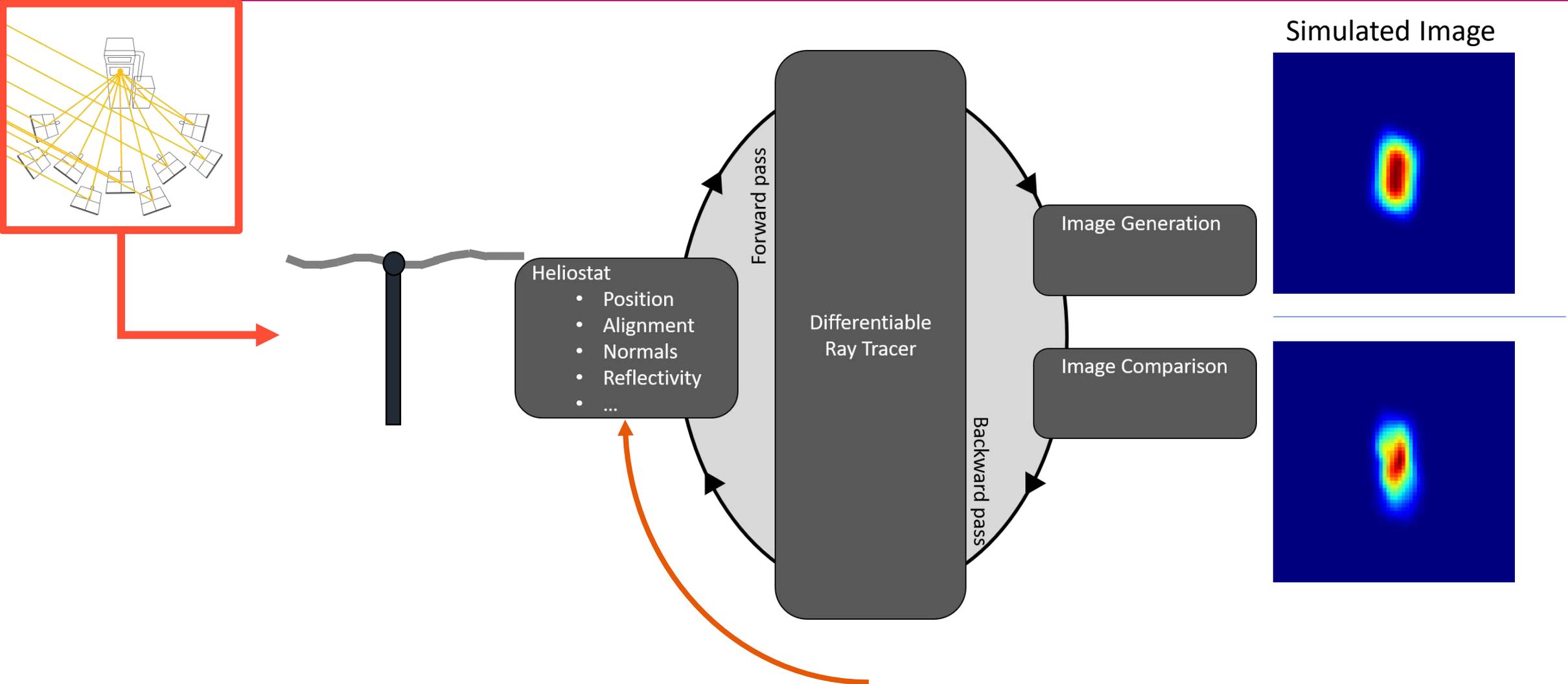


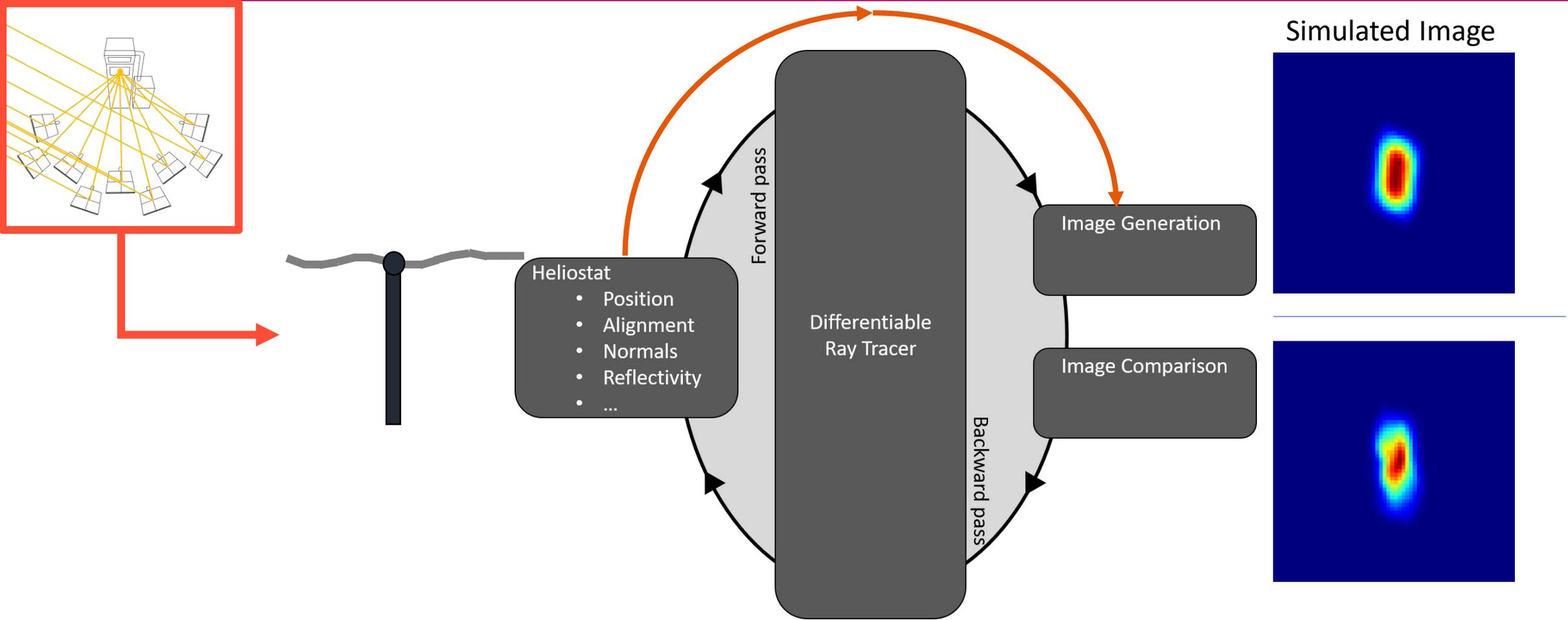


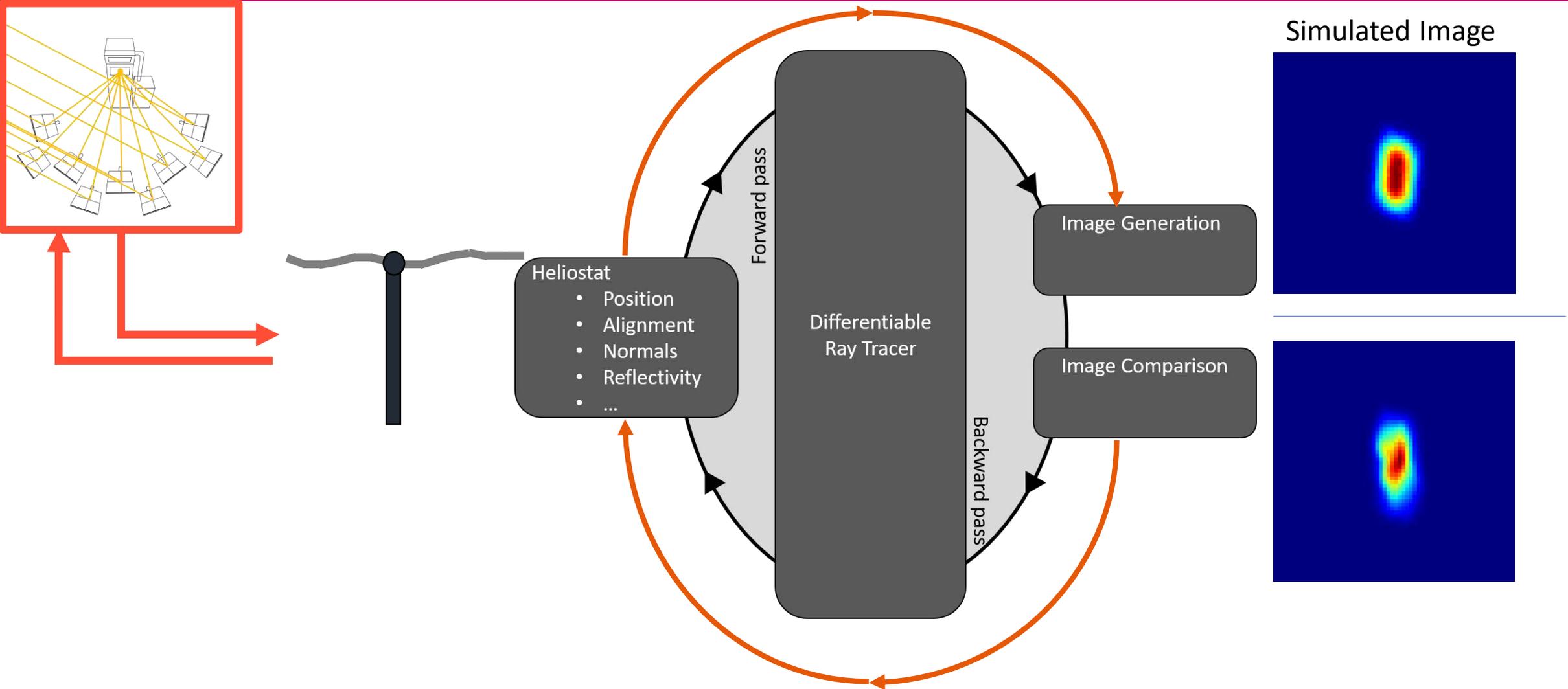












SFERA-III

Solar Facilities for the European Research Area



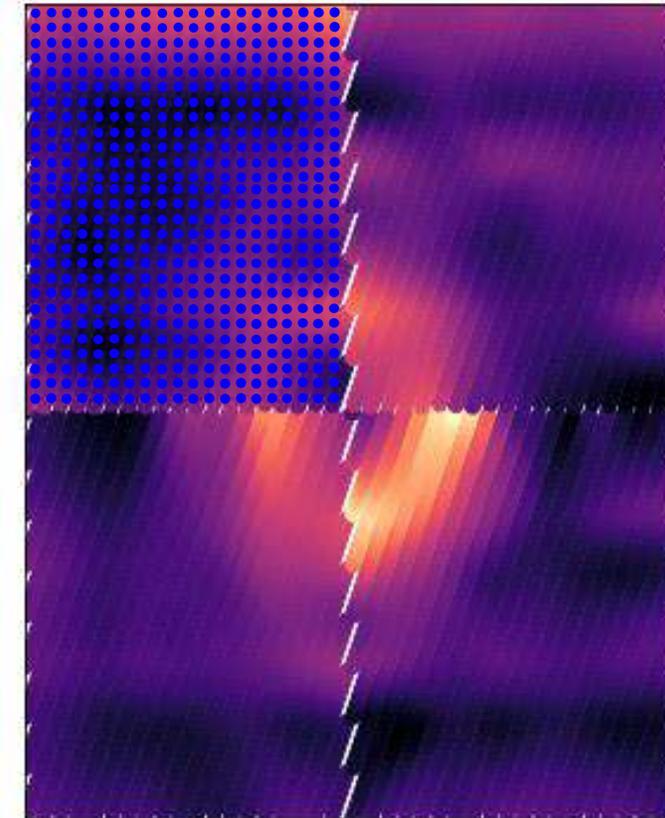
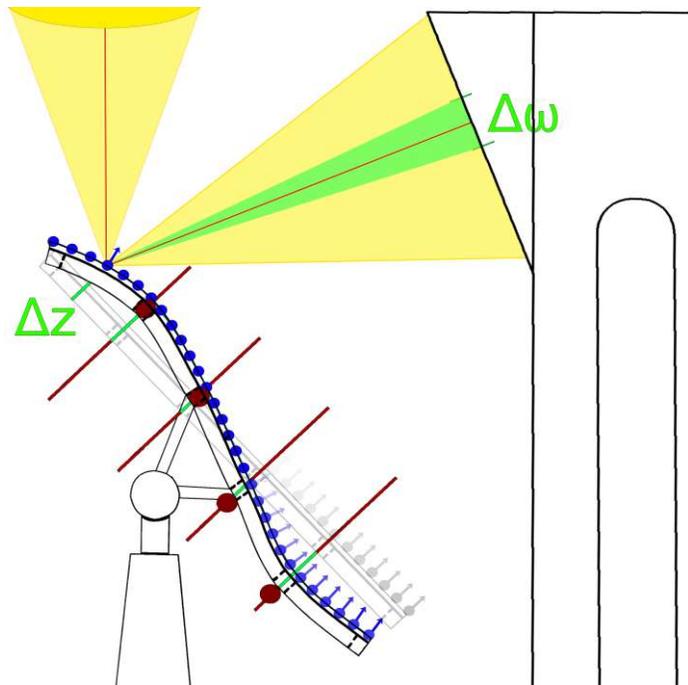
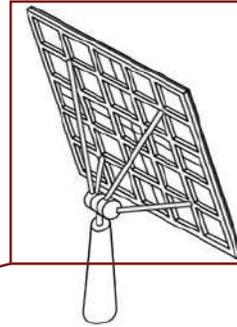
Diagnosis:

Detect Mirror Deformations

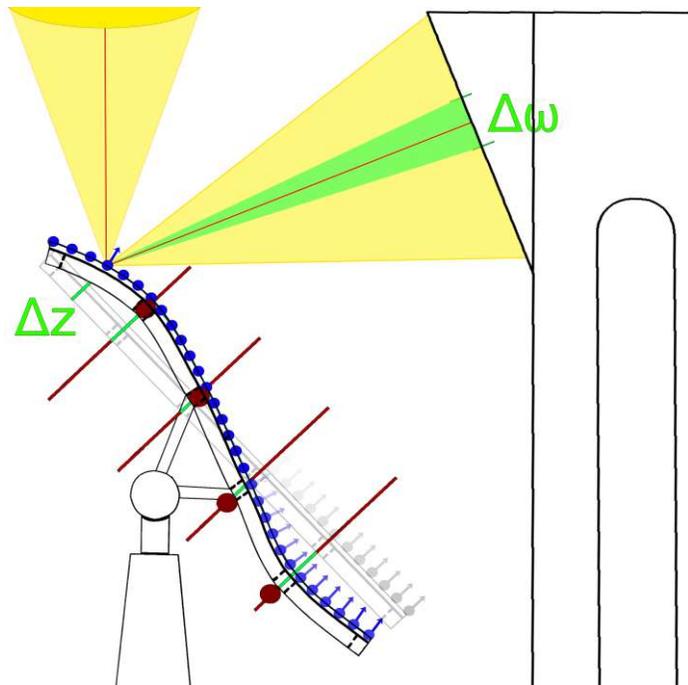
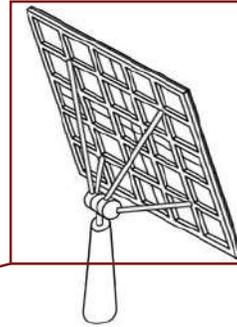


Ca. 25k points per facet

Continuous NURBS heliostat surface

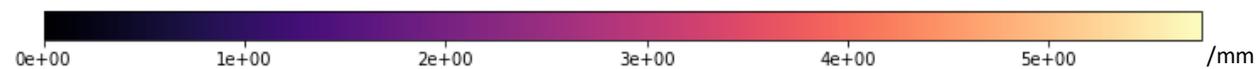
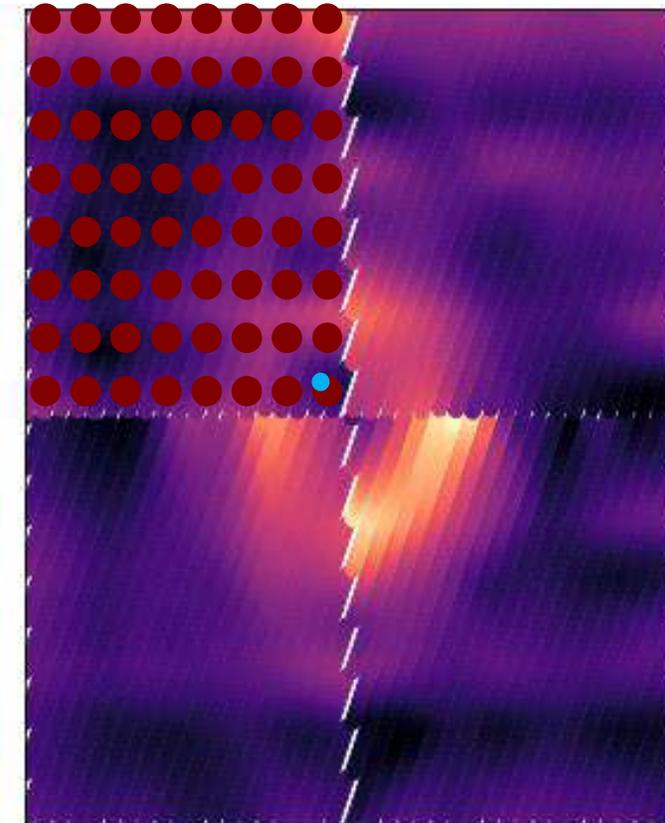


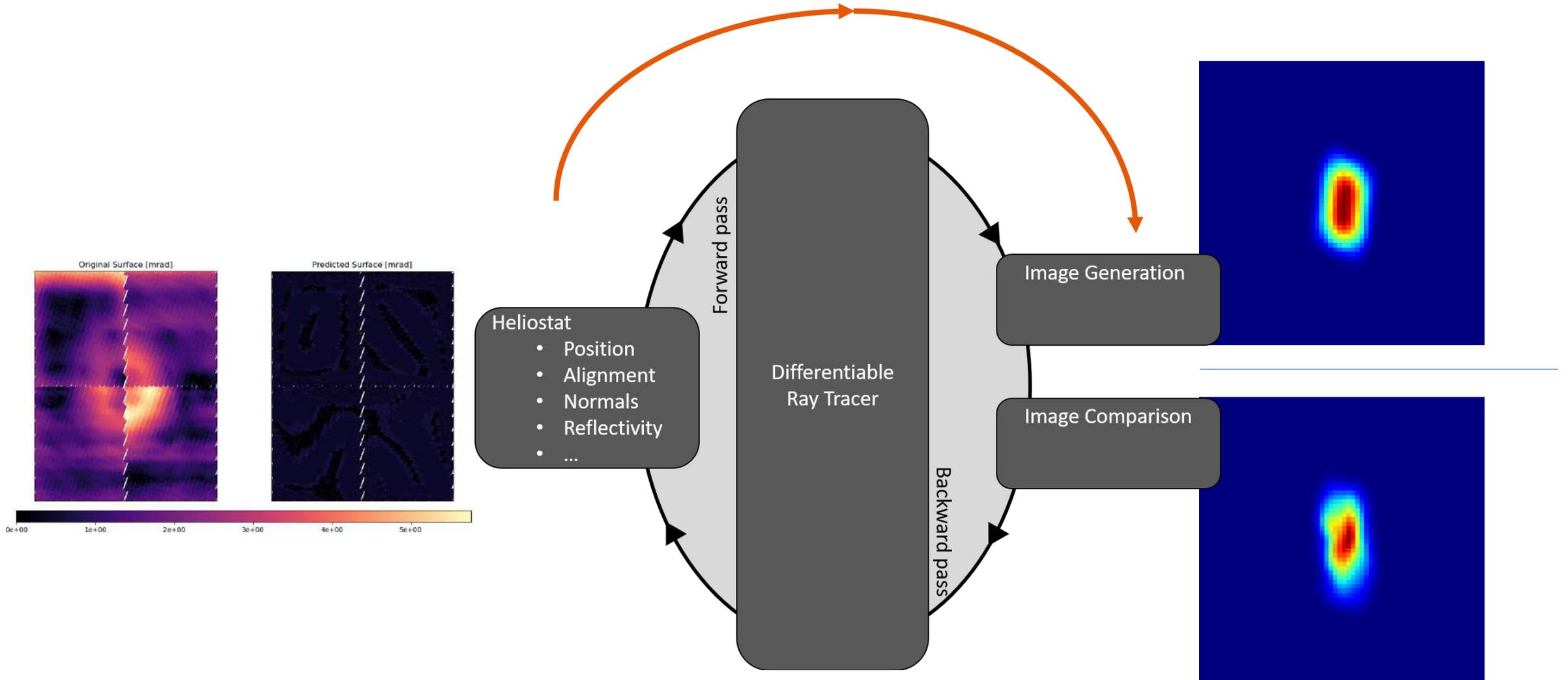
Continuous NURBS heliostat surface

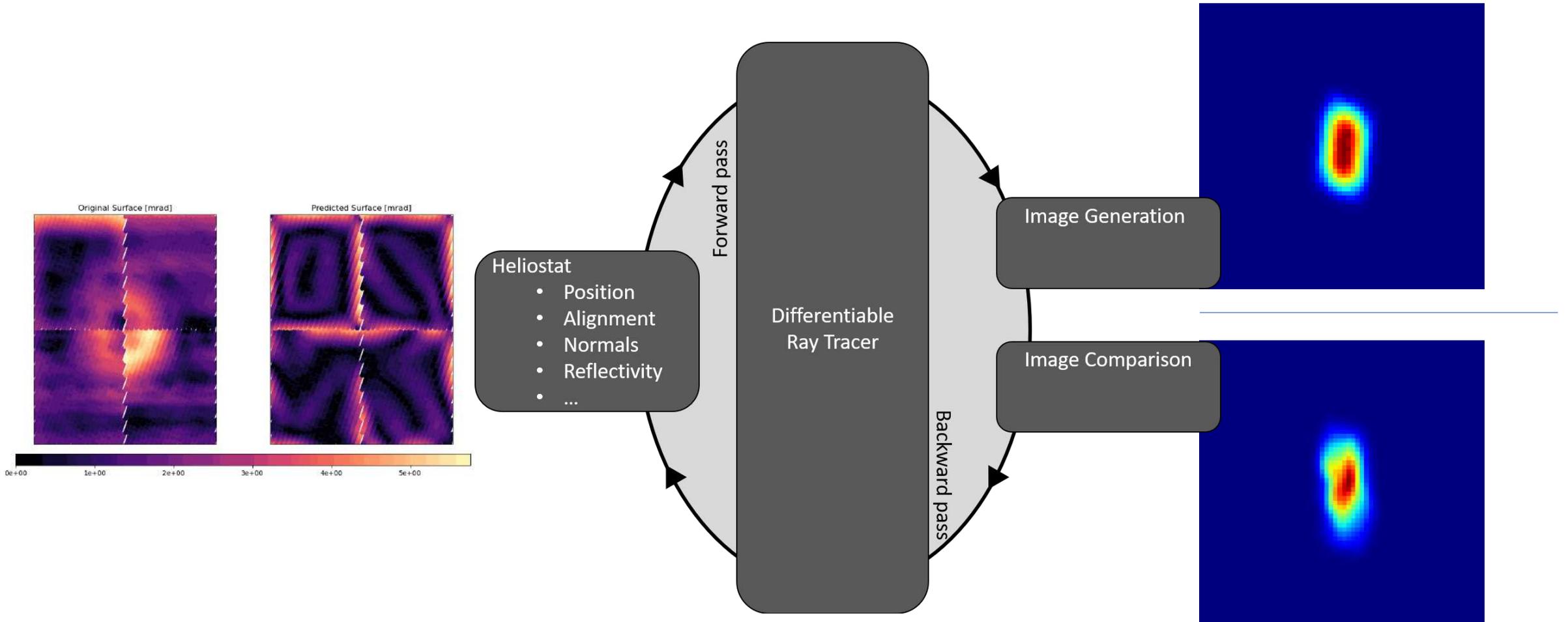


Ca. 25k points per facet

→ 64 points per facet with over 98% overlap







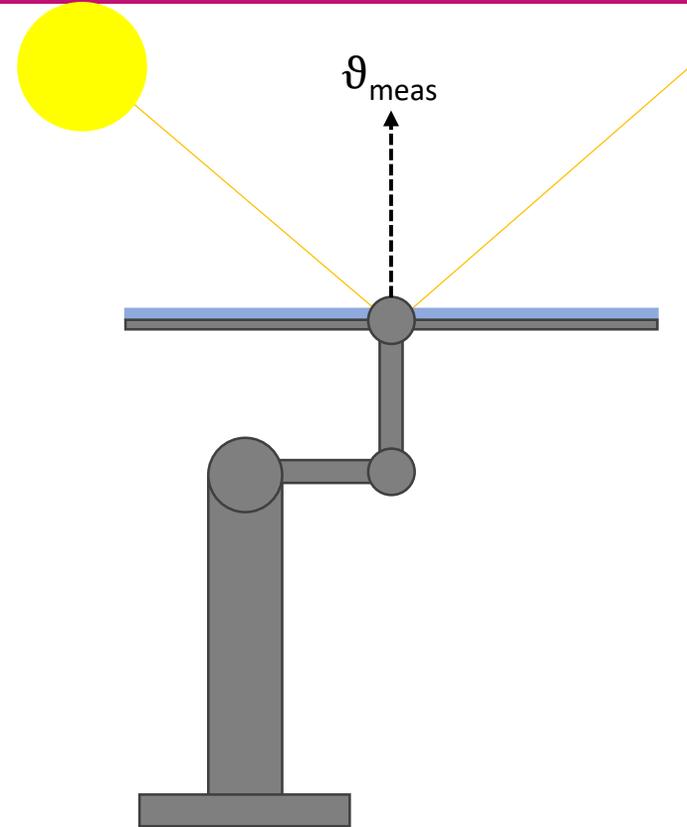
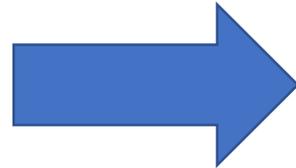
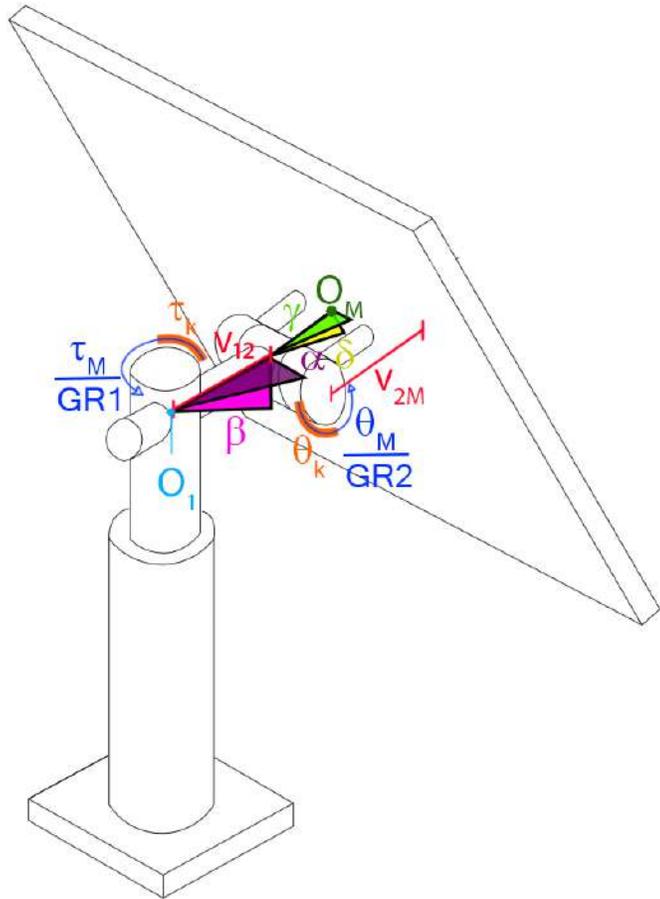
SFERA-III

Solar Facilities for the European Research Area

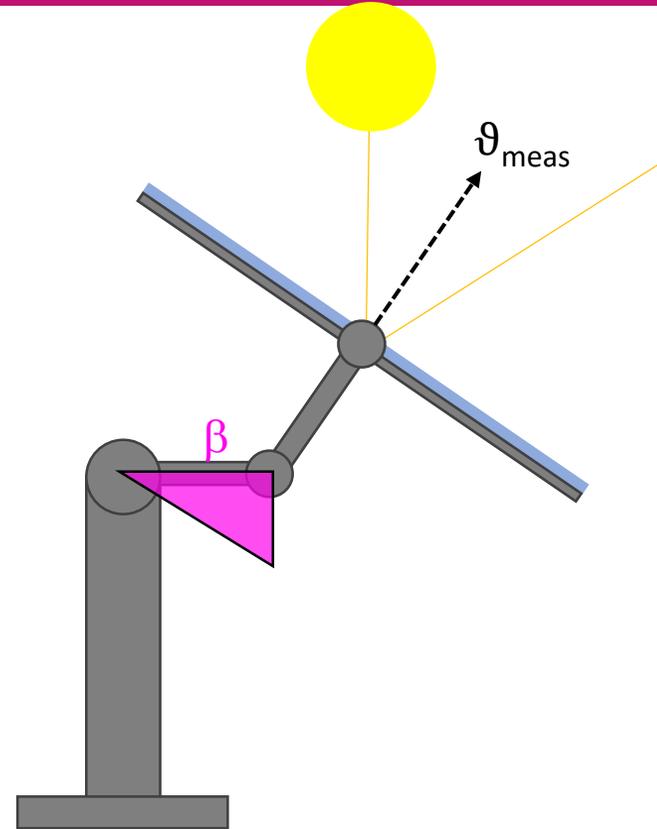
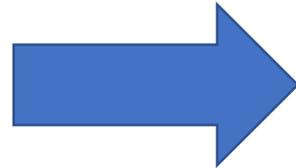
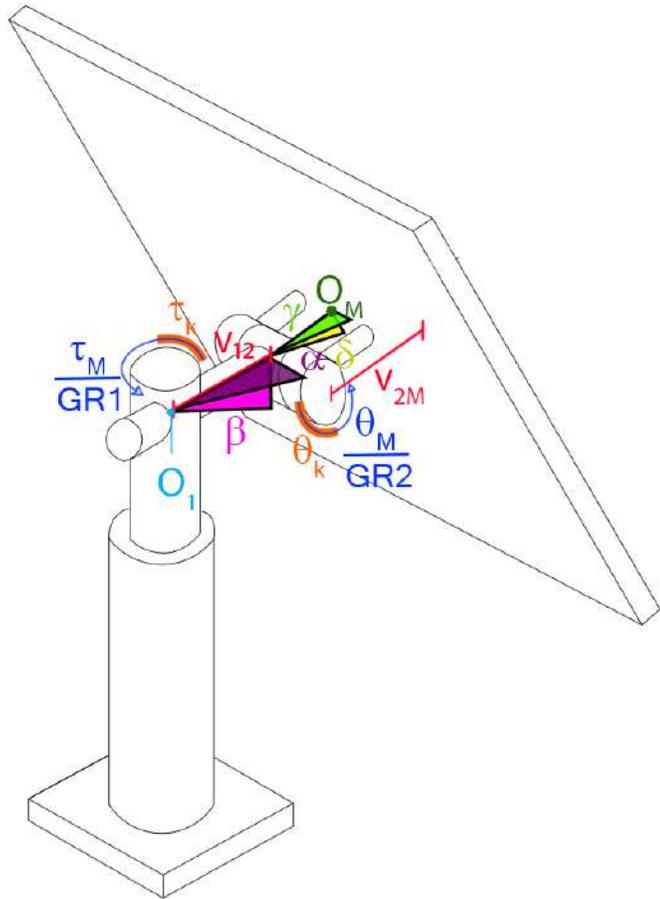


Control: *Heliostat Alignment*

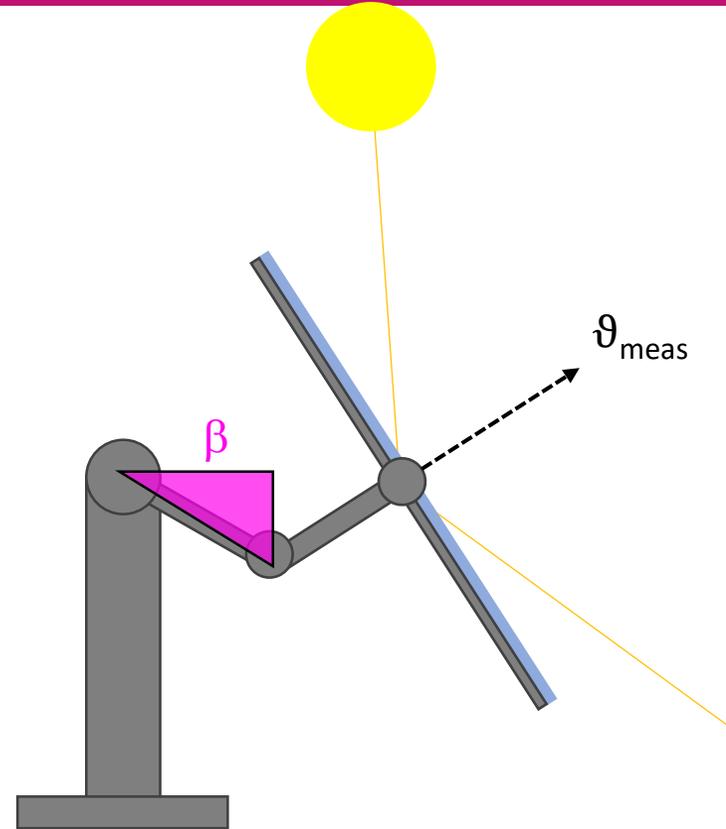
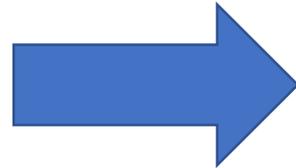
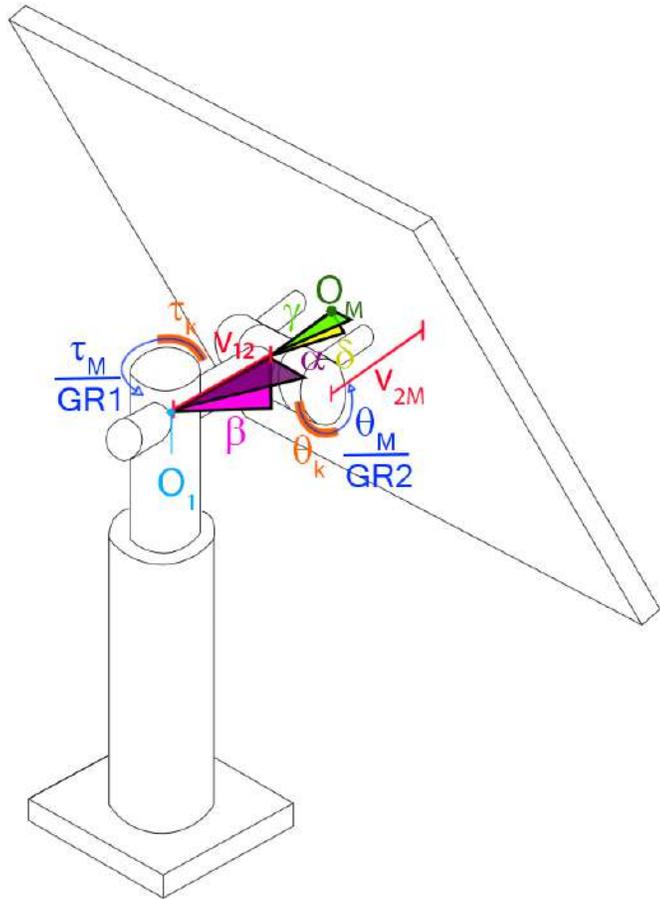




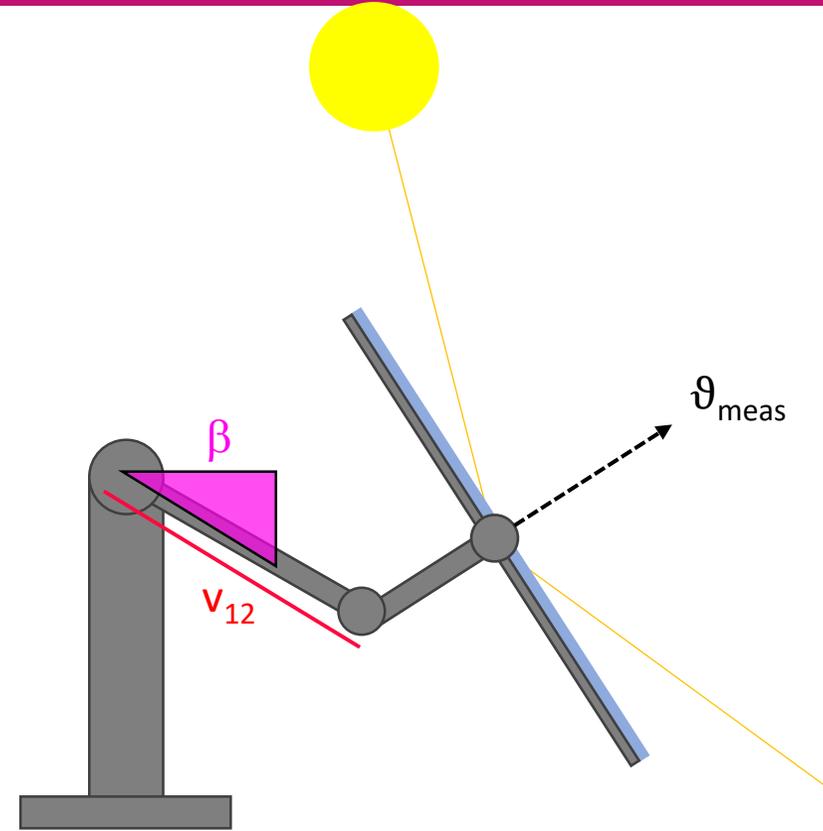
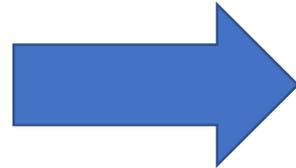
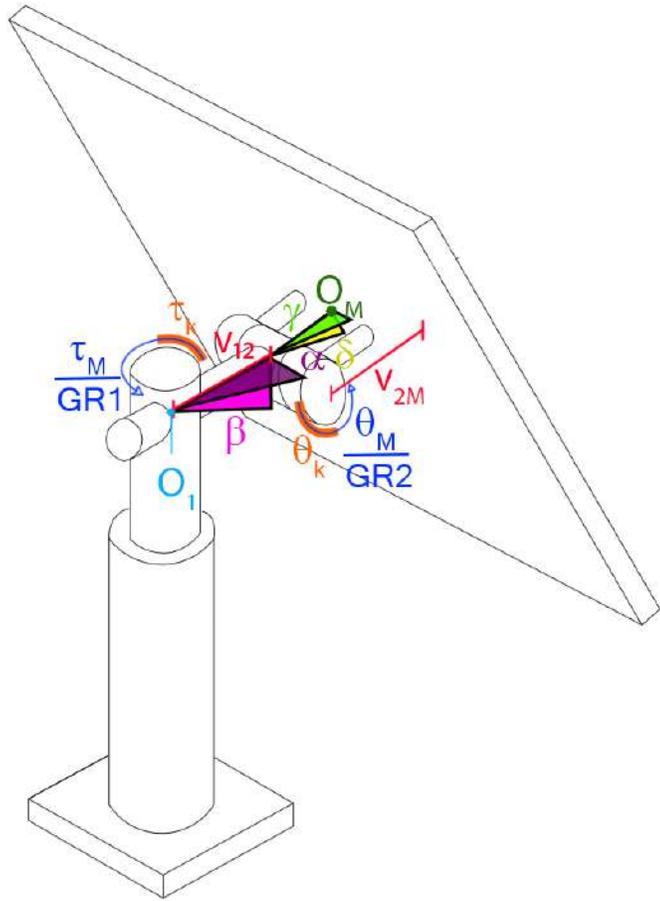
$$\vartheta_{meas} = f(\text{heliostat position, sun position, aimpoint})$$



$$\vartheta_{meas} = f(\text{heliostat position, sun position, aimpoint})$$



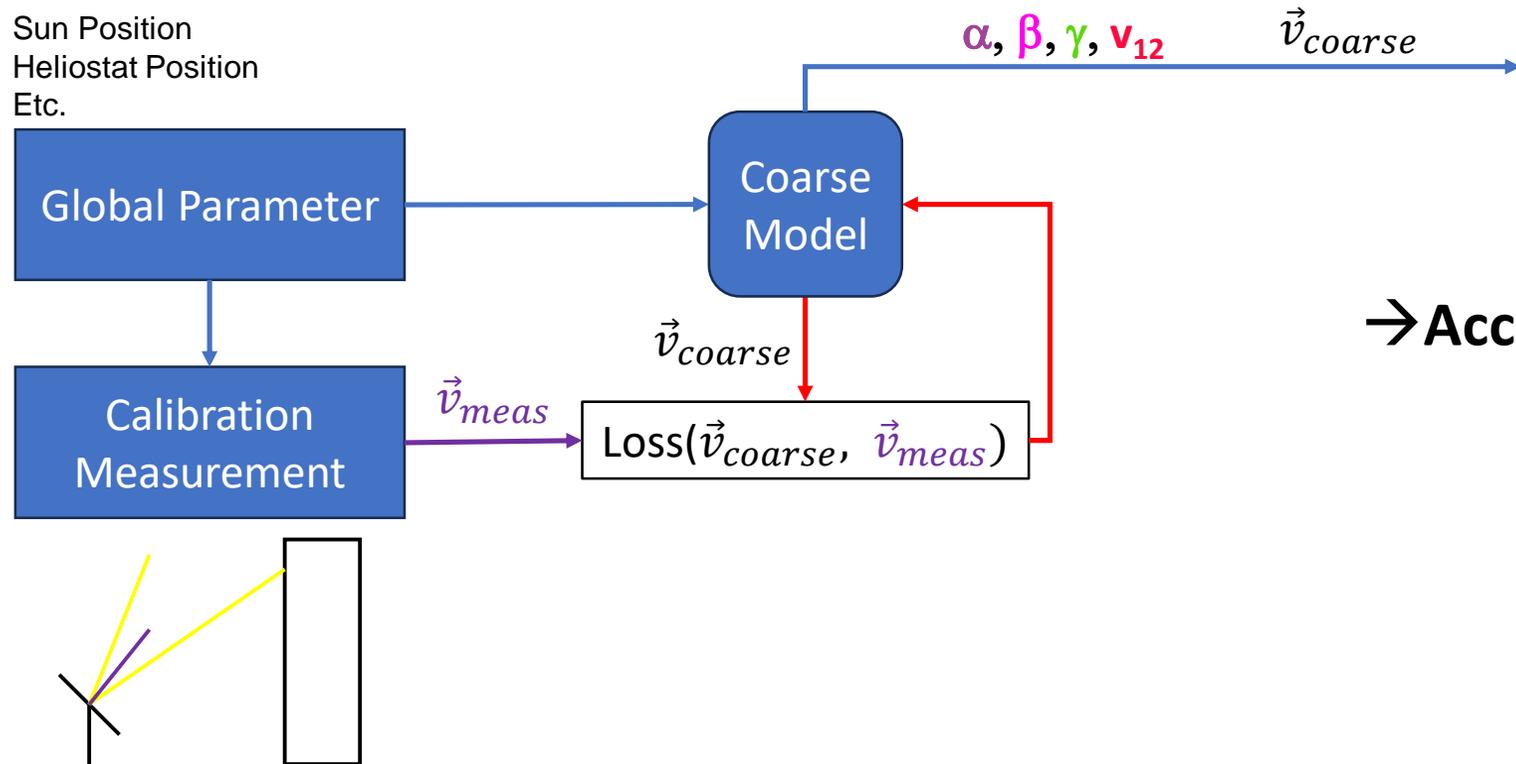
$$\vartheta_{meas} = f(\text{heliostat position, sun position, aimpoint})$$



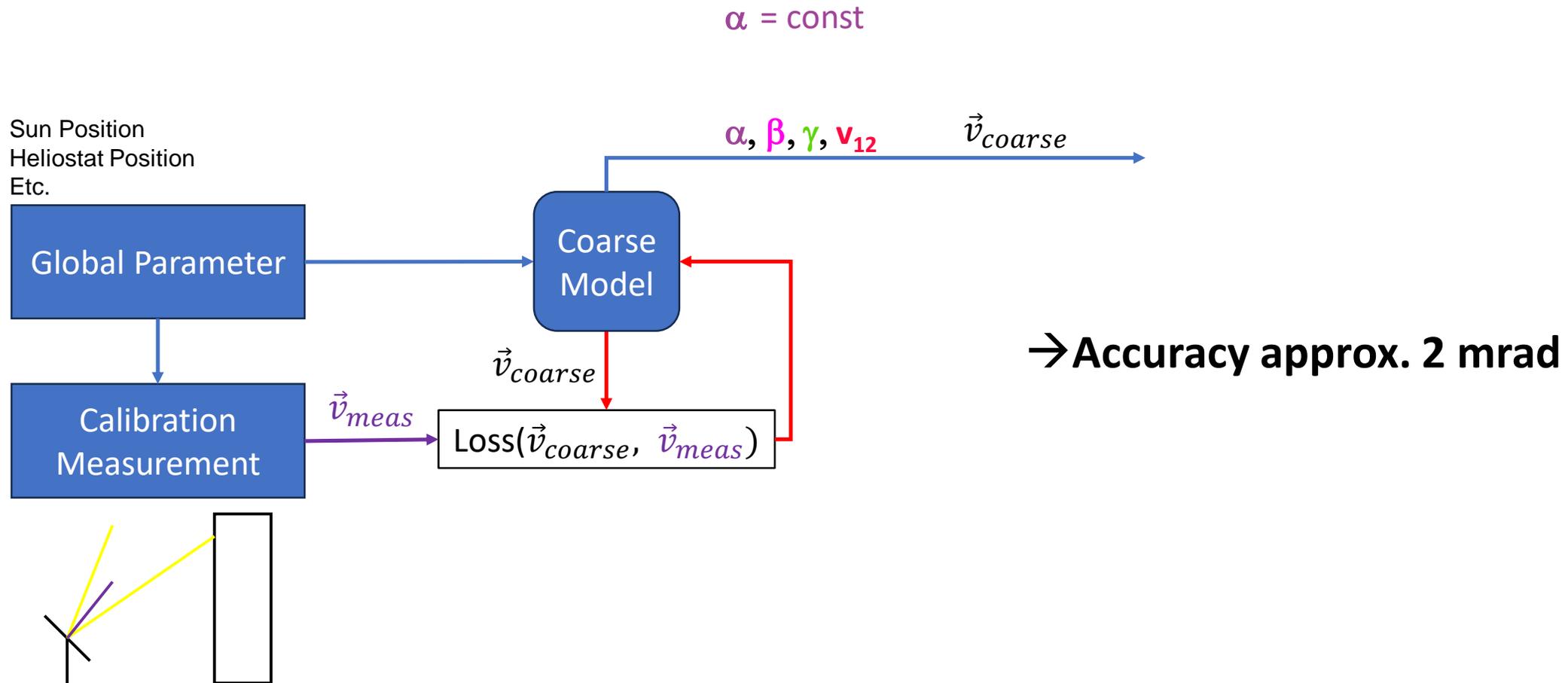
$$\vartheta_{\text{meas}} = f(\text{heliostat position, sun position, aimpoint}$$

$$\alpha, \beta, \gamma, \mathbf{v}_{12}, \dots) \rightarrow \mathbf{20 \text{ independent parameters}}$$

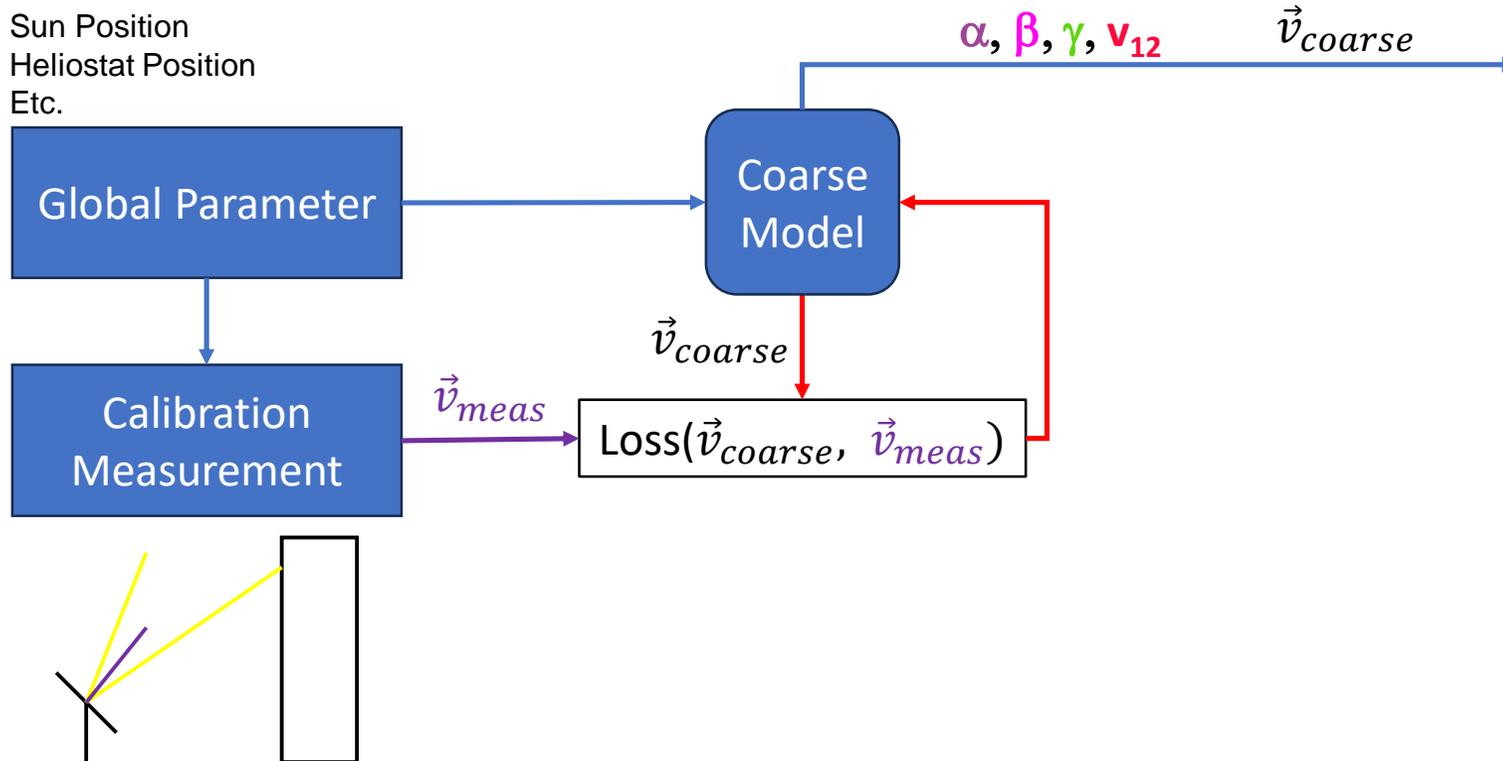
Sun Position
Heliostat Position
Etc.



→ Accuracy approx. 2 mrad

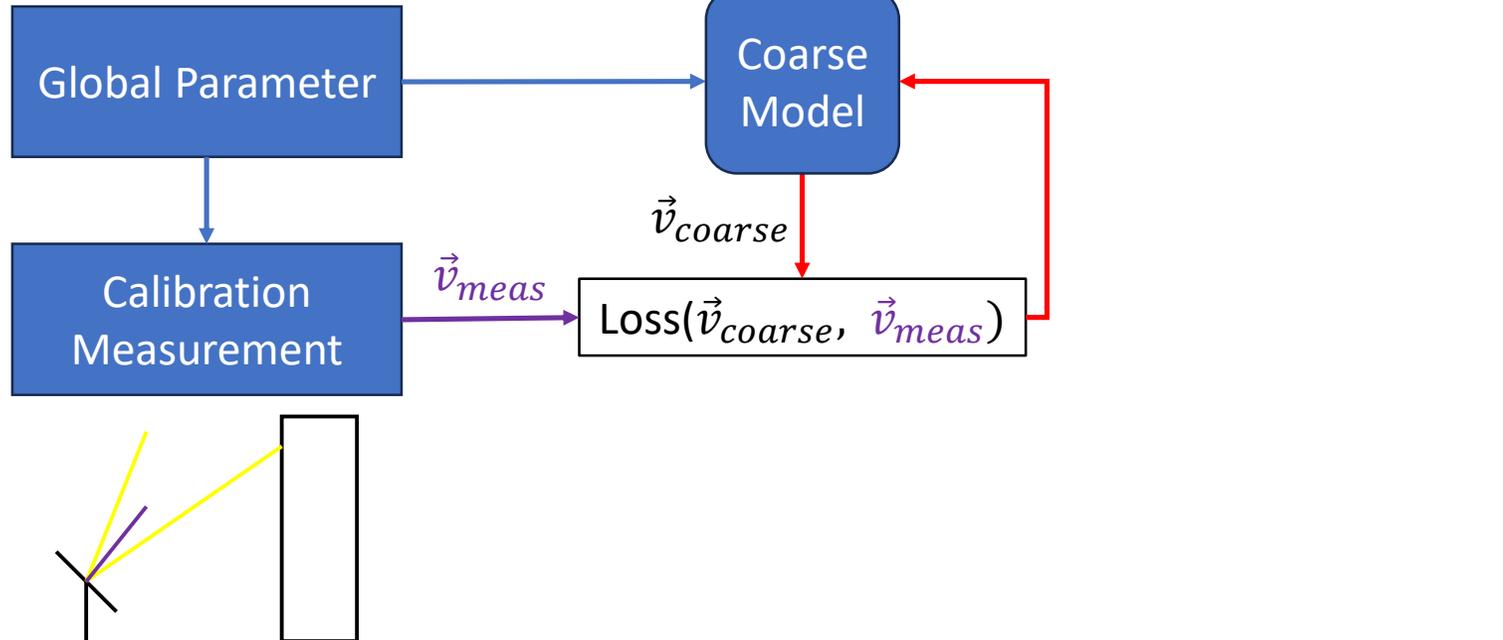


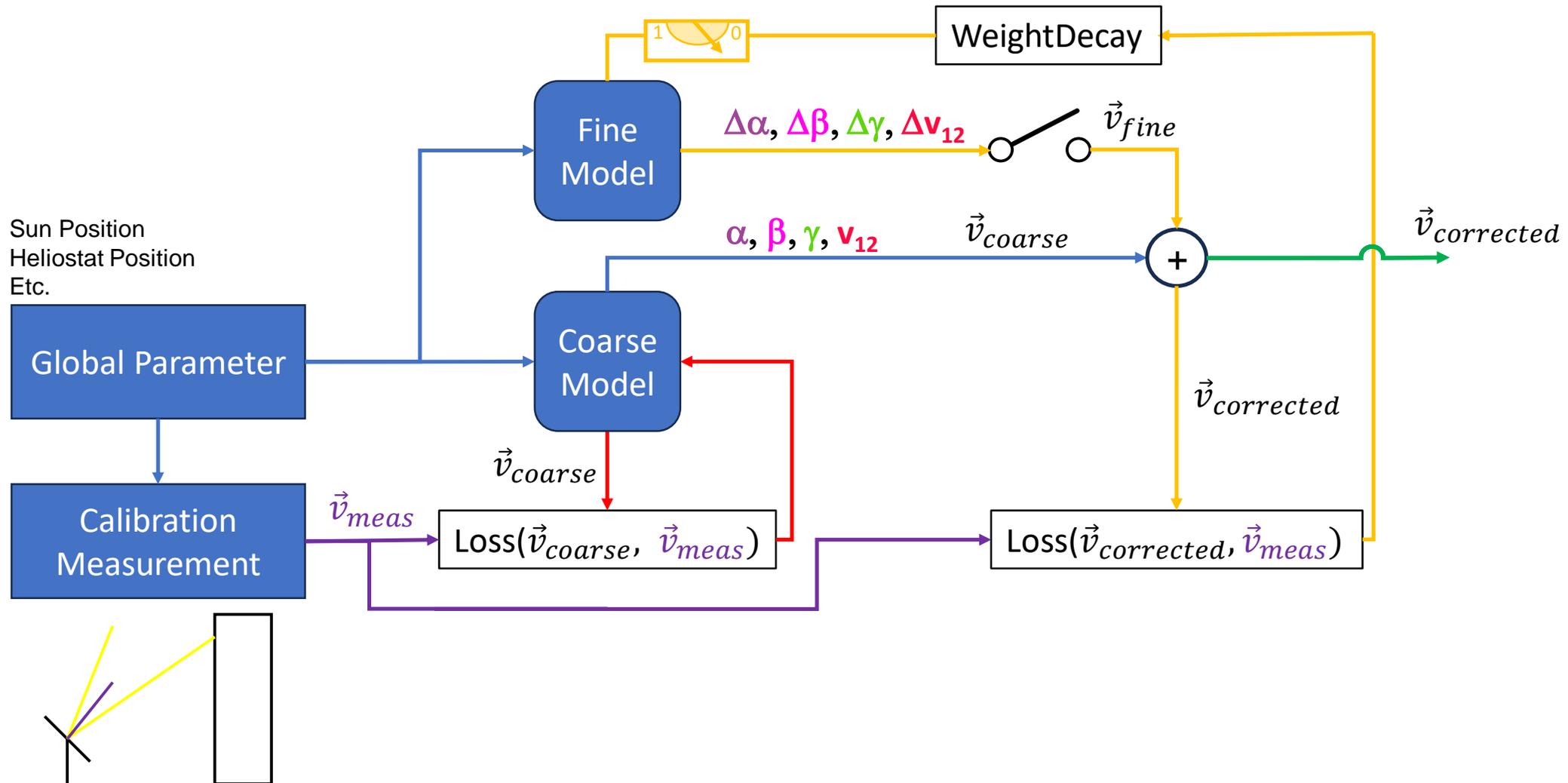
α (alignment, temp., time, wind, rigidity) = ???



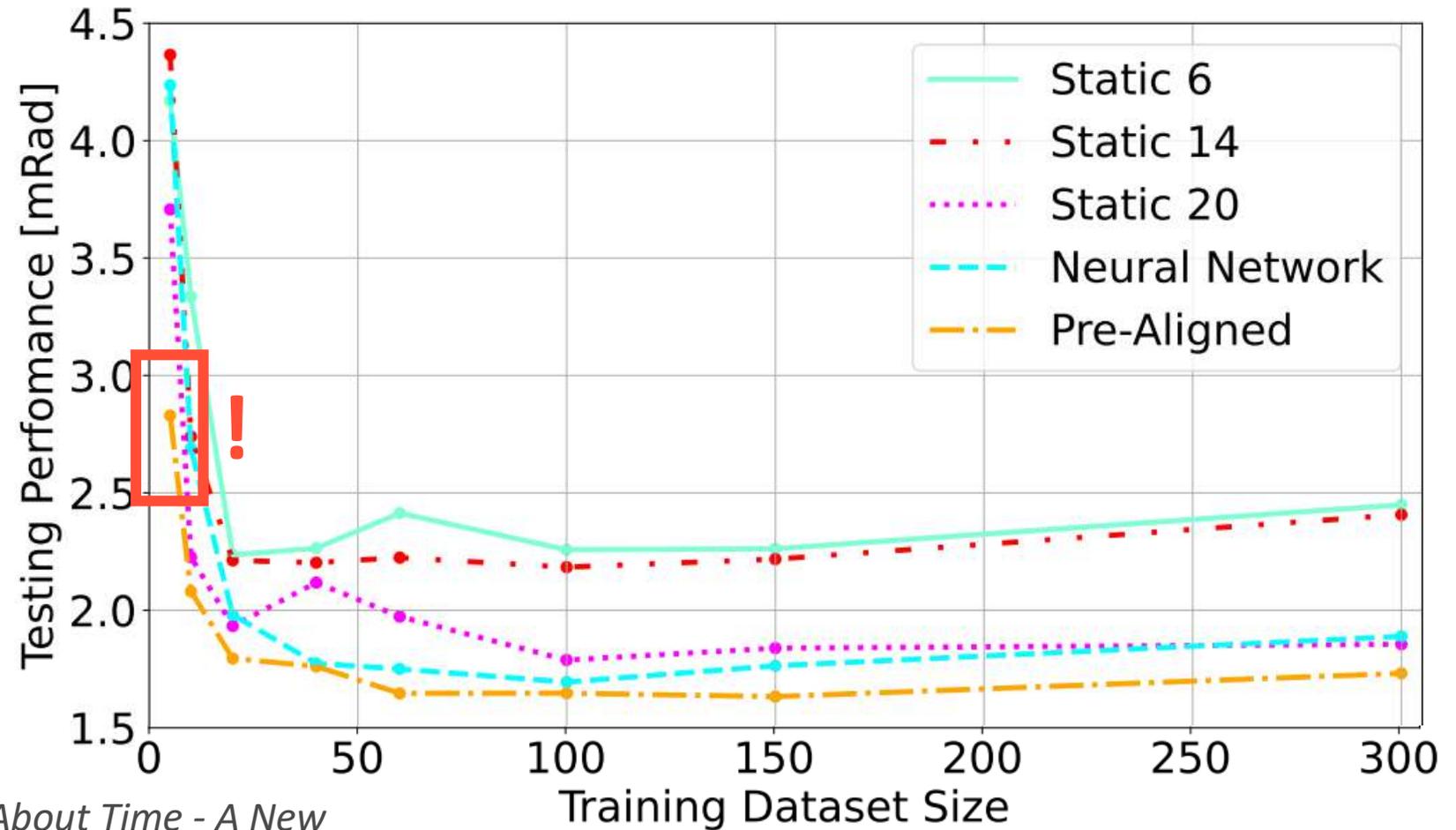
$$\alpha_{\text{complete}} = \alpha_{\text{const}} + \Delta\alpha_{\text{dynamic}} \text{ with } \Delta\alpha \ll \alpha$$

Sun Position
Heliostat Position
Etc.





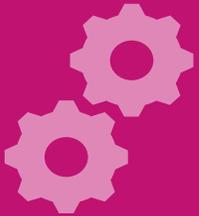
- The models are evaluated on the basis of a special data set training/test split, which provides information about the worst-case performance. Details in [1]
- The combination of rigid body and neural network model prediction is always best



[1] Pargmann and Leibauer et al. *It is Not About Time - A New Standard for Open-Loop Heliostat Calibration Methods*

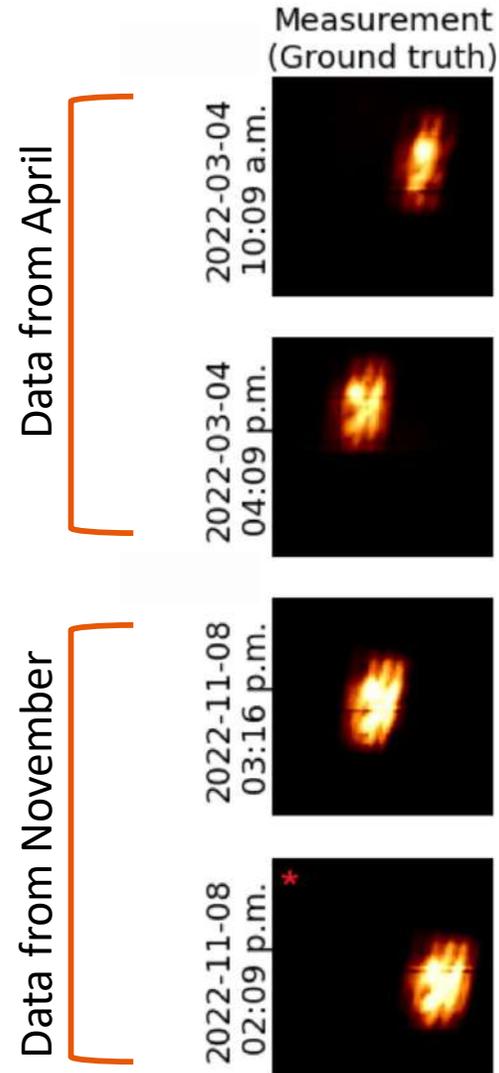
SFERA-III

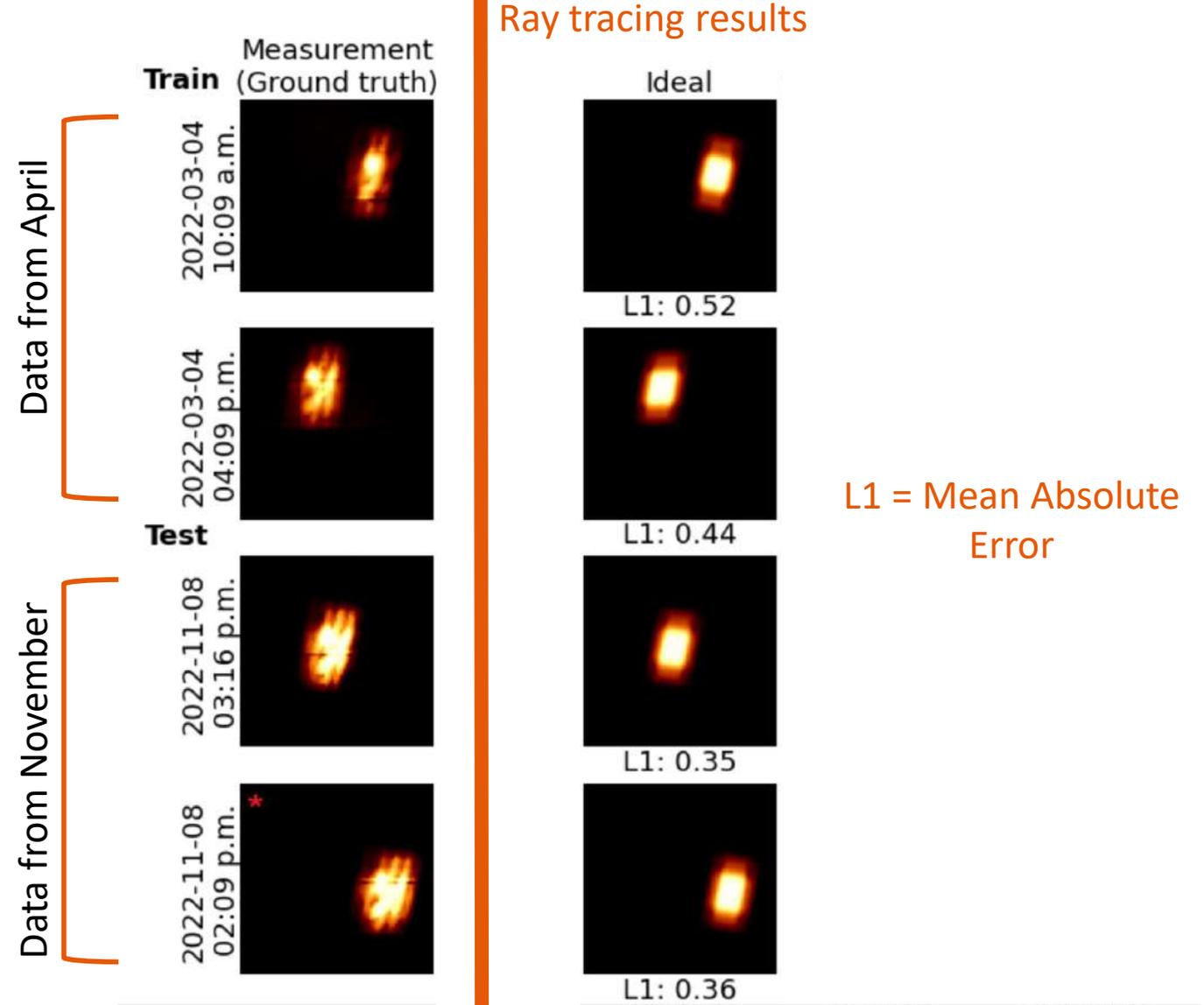
Solar Facilities for the European Research Area

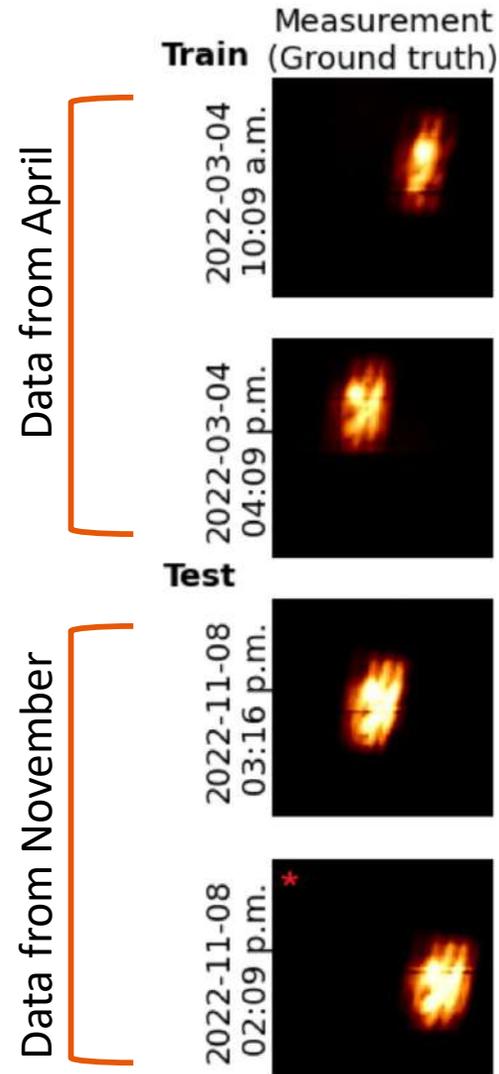


Prediction: *Single Heliostat Flux Density*

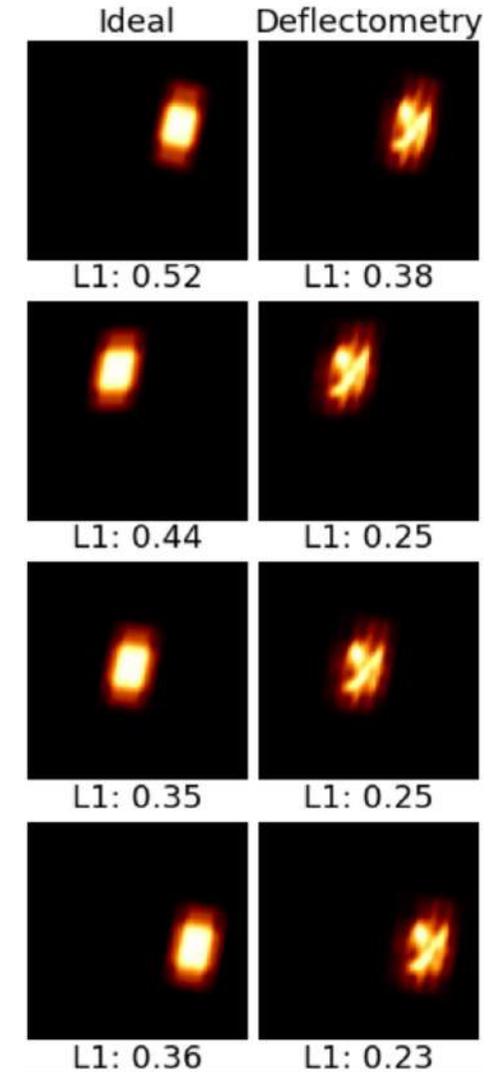


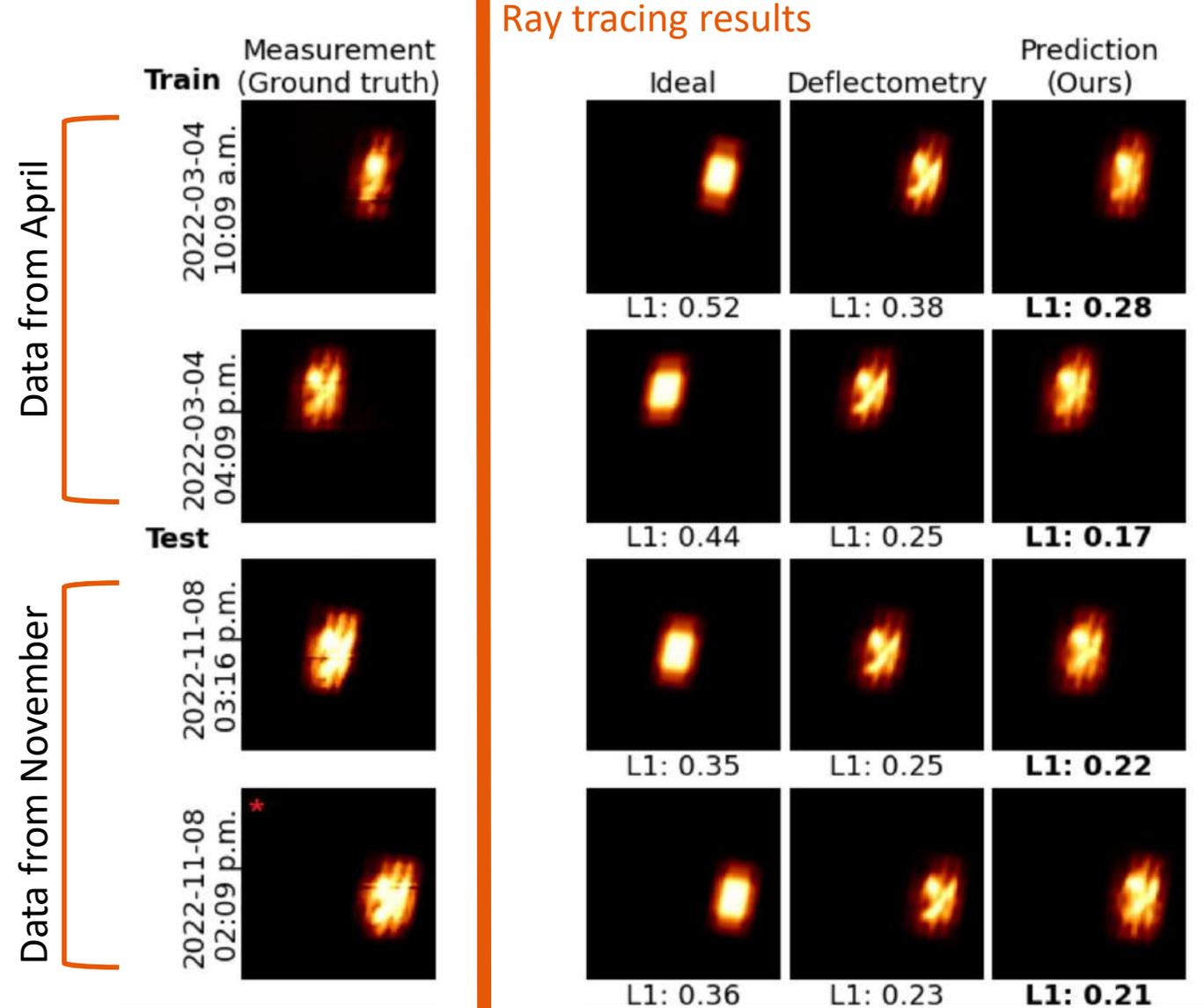






Ray tracing results





SFERA-III

Solar Facilities for the European Research Area

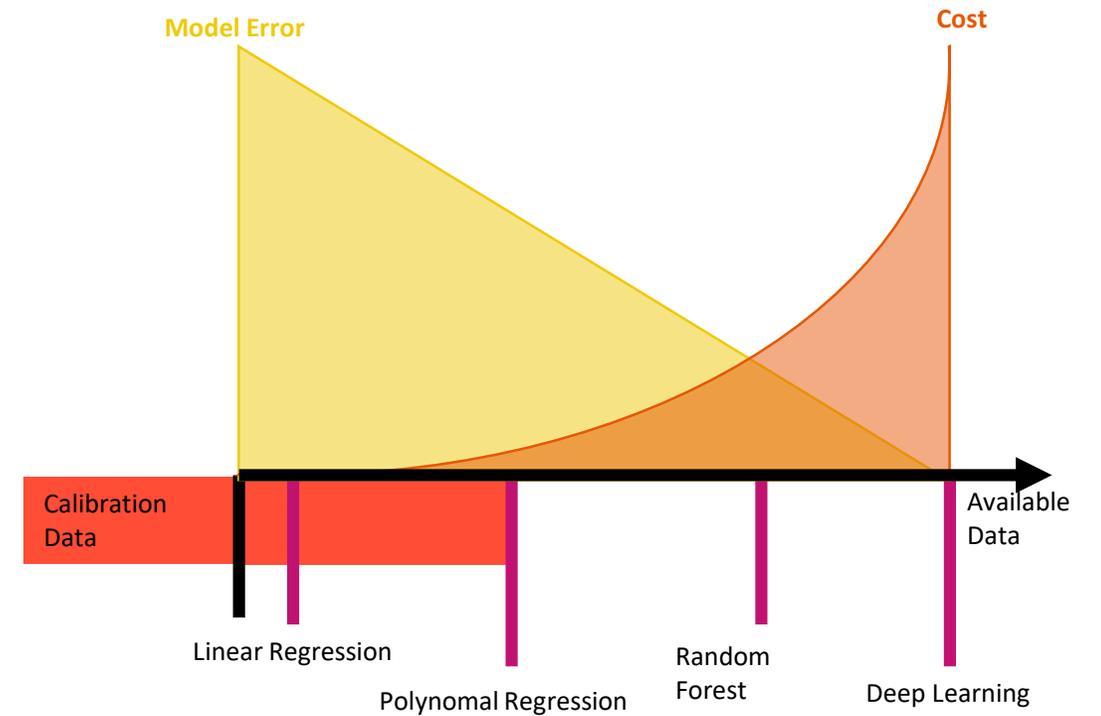


Outcomings



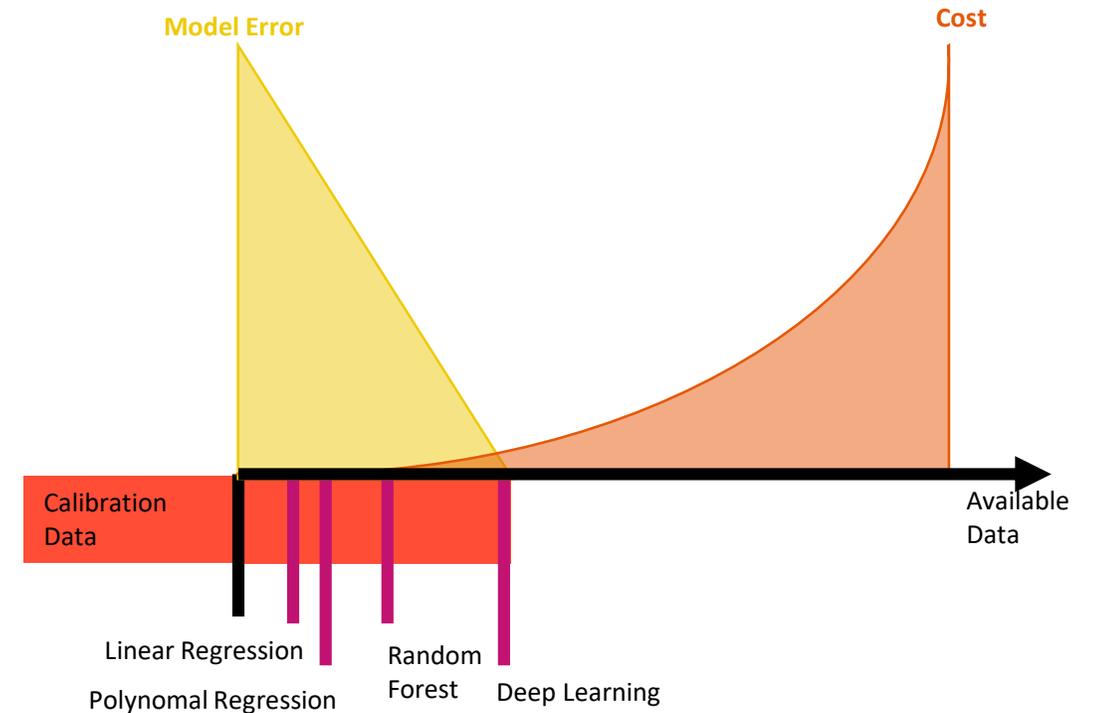
Solar Facilities for the European Research Area

How to apply Deep Learning on small data sets at solar towers?



How to apply Deep Learning on small data sets at solar towers?

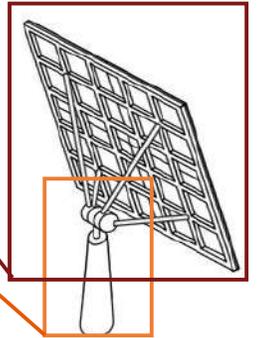
- Data is scarce, but physical Information isn't!
- Ways to reduce the data usage at the solar tower:
 - Use Physical Information for Initialization (eg. Heliostat Alignment)
 - Use Simulation/Augmentation data to pretrain Networks
 - Use a physics informed Loss function to smooth your optimization function
- Use images instead of tabular data



When to choose AI models over physic at solar towers (at small data sets)?

- Most processes at the solar tower are very good approximated by physical simulations
- AI can be used to close this gap when the error source is unknown
- Estimate unknown error:
 - Example Heliostat Alignment Model:
 - Potential optimal accuracy approx. 0.03mrad
 - Heliostatfield accuracy approx. 2.1mrad
 - Measurement and Hardware errors approx. 1mrad
 - Example Heliostat Surface Model
 - NURBS can fit Mirror to 100% (even with broken edges)

Continous NURBS heliostat surface
AI supported heliostat alignment model



➔ AI can reduce model error by half the absolute error

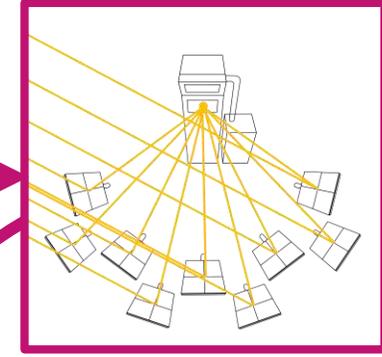
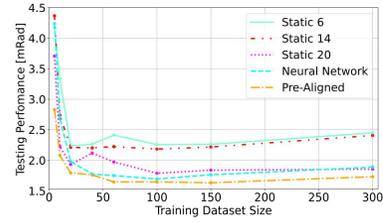
➔ Physical model's parameter space is sufficient

SFERA-III

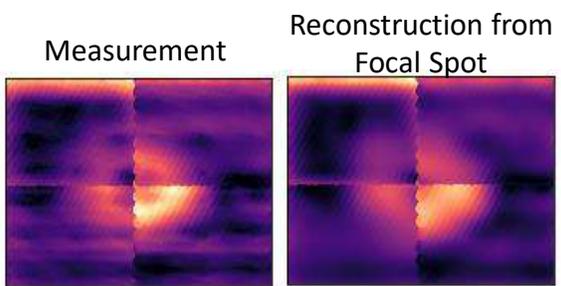
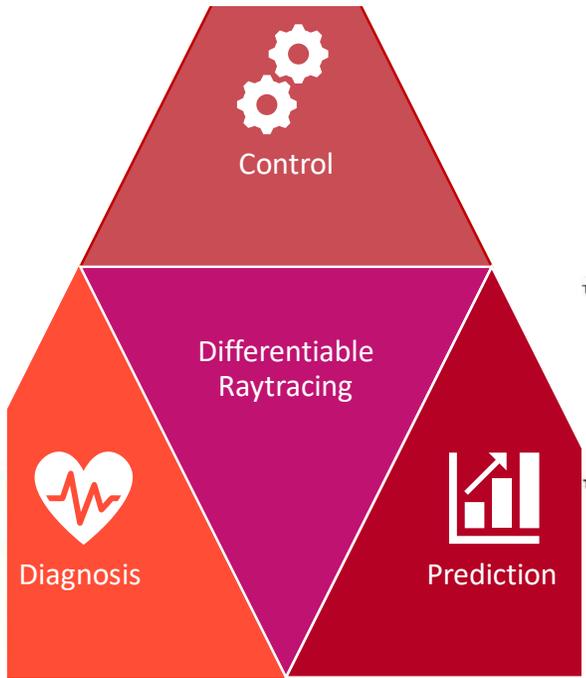
Solar Facilities for the European Research Area

Conclusion





Fully Automated Operation



	Measurement (Ground truth)	Ideal	Deflectometry	Prediction (Ours)
Train				
2022-03-04 10:09 a.m.				
		L1: 0.52	L1: 0.38	L1: 0.28
2022-03-04 04:09 p.m.				
		L1: 0.44	L1: 0.25	L1: 0.17
Test				
2022-11-08 03:16 p.m.				
		L1: 0.35	L1: 0.25	L1: 0.22
2022-11-08 07:09 p.m.				
		L1: 0.36	L1: 0.23	L1: 0.21

