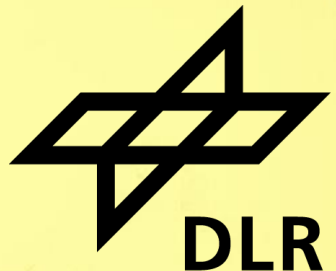


SMART CSP

HOW ARTIFICIAL INTELLIGENCE CAN SUPPORT CONCENTRATING SOLAR TECHNOLOGIES

Robert Pitz-Paal, Institute of Solar Research

30.11.2022



1. Characteristics of CSP

2. Status and Perspectives

3. Advanced methods to operate a CSP plant efficiently and autonomously

- Airborne condition monitoring of solar collector fields
- Data driven flux density prediction
- Dispatch optimization of power production considering weather forecast uncertainties

4. Conclusions

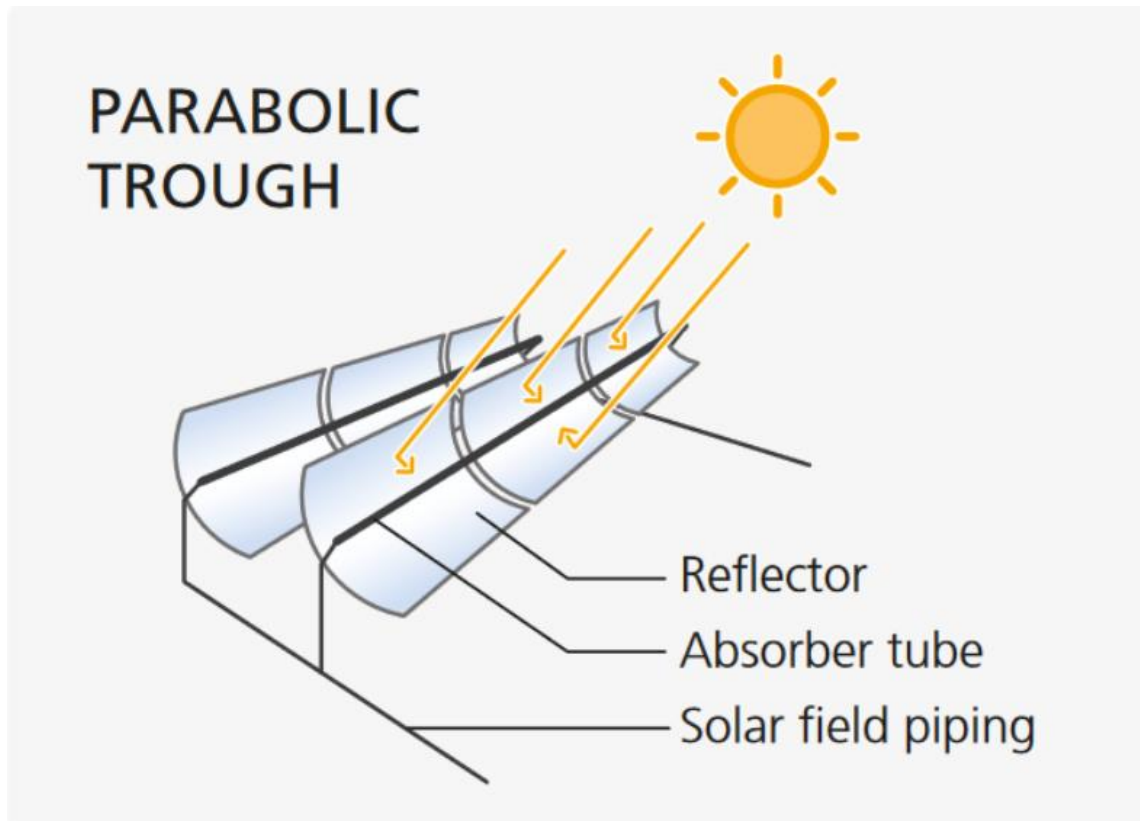
A wide-angle photograph of a Concentrated Solar Power (CSP) field at sunset. The field is composed of thousands of heliostats (mirrors) arranged in concentric circular patterns, reflecting the bright orange and yellow light of the setting sun. In the center of the field, a tall, slender receiver tower stands with a crane on top. The sky is a mix of orange, yellow, and dark blue, with some clouds visible. The overall scene is a vast, open landscape under a dramatic sky.

CHARATERISTICS OF CSP

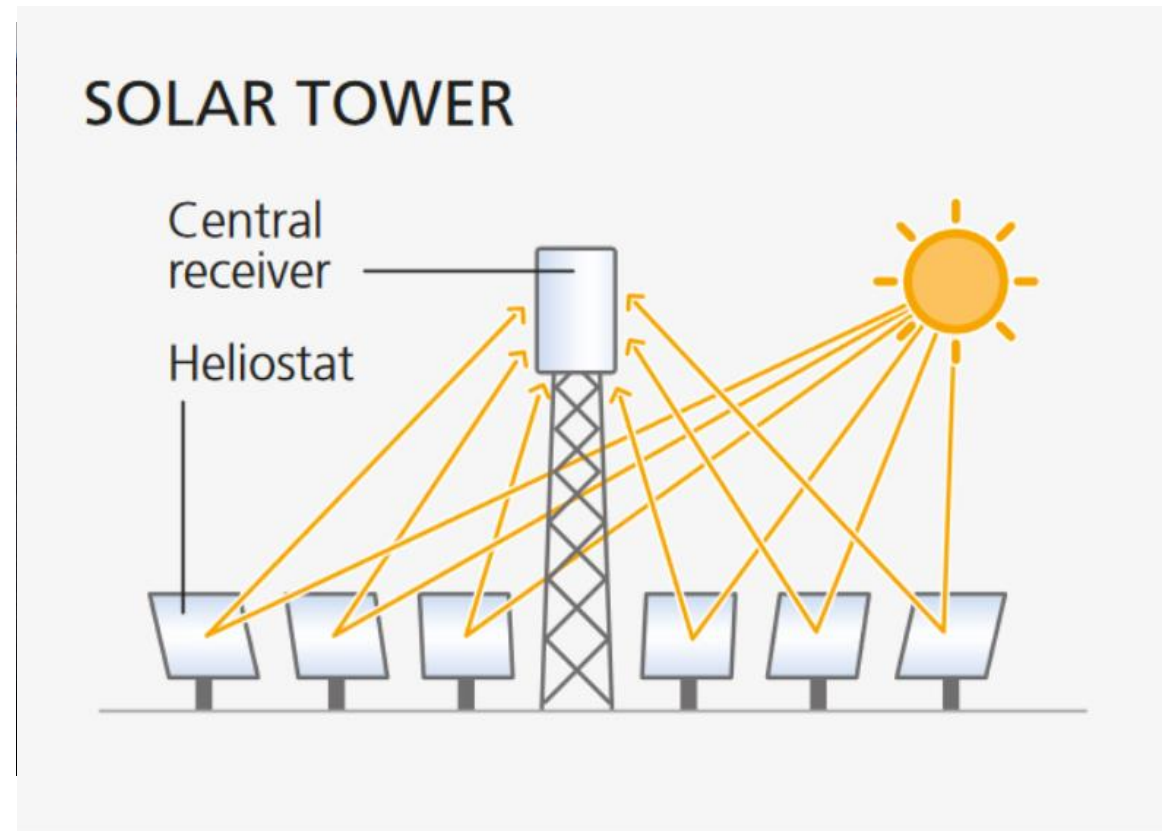
In a nutshell: How does a solar thermal power plant work?

Special mirrors focus direct sunlight and reflect it onto...

- ... a linear receiver

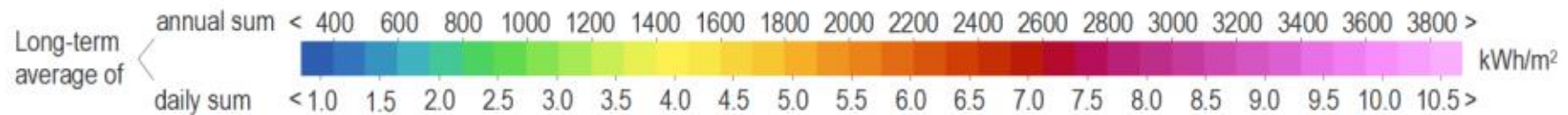
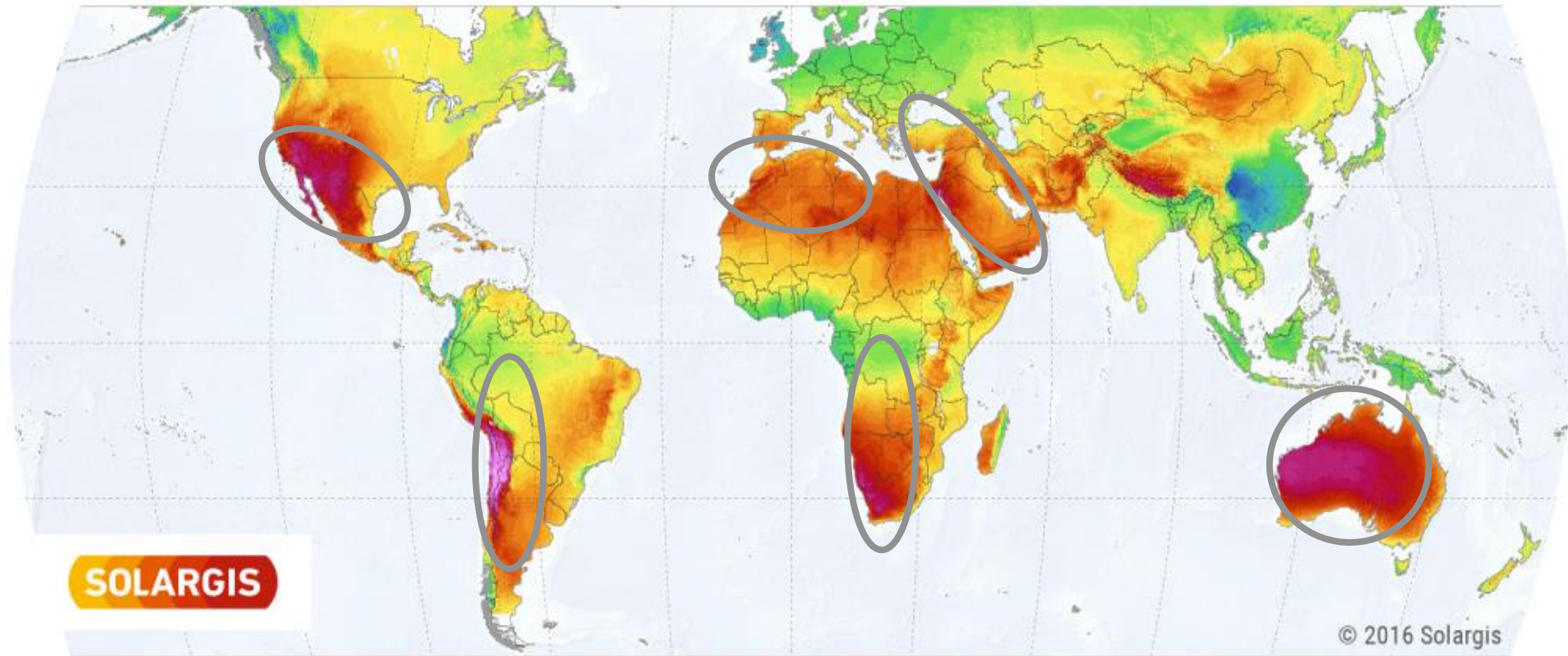


- ... a central receiver

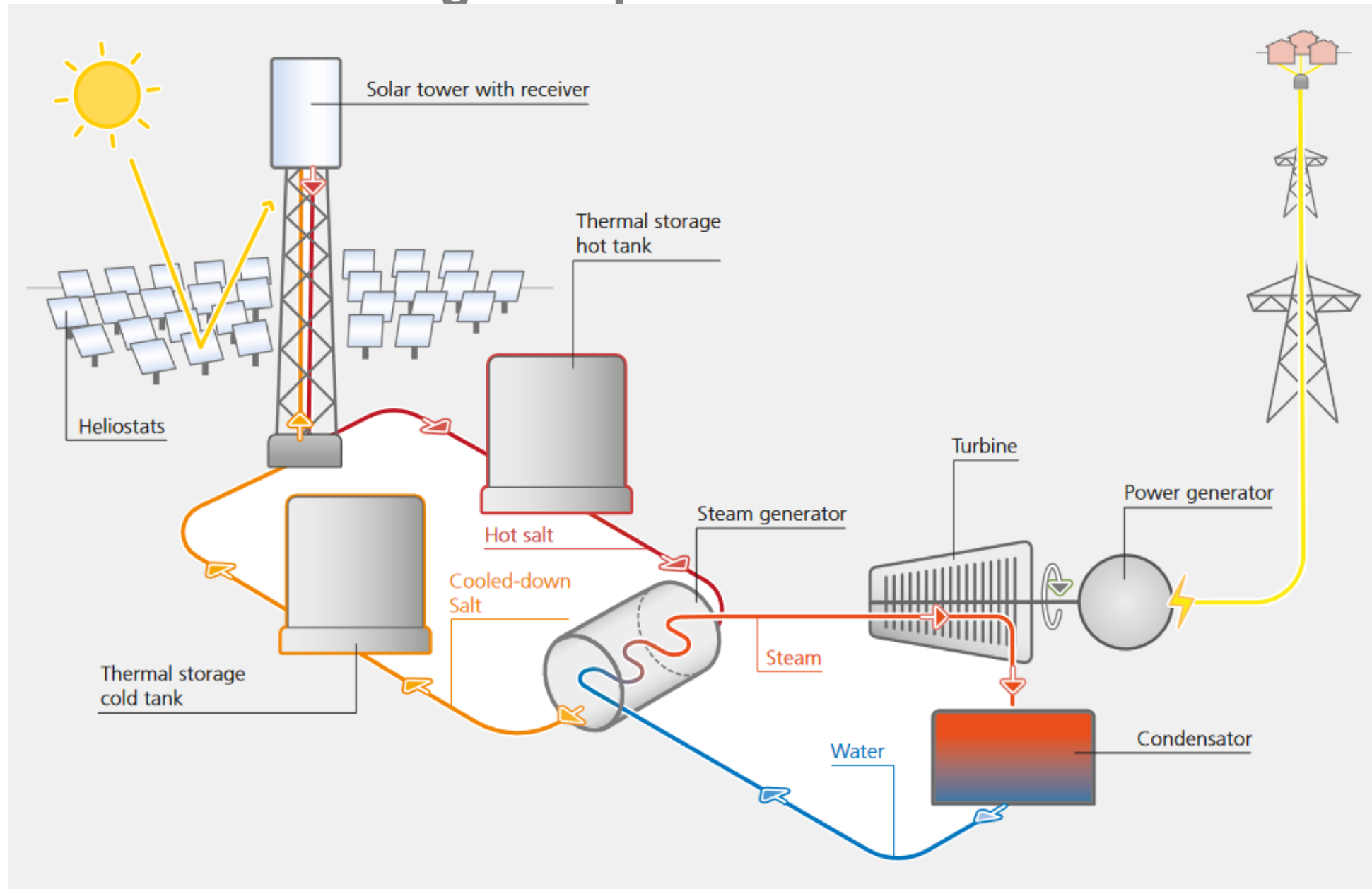


CSP best suitable in areas with high direct normal radiation

DIRECT NORMAL IRRADIATION

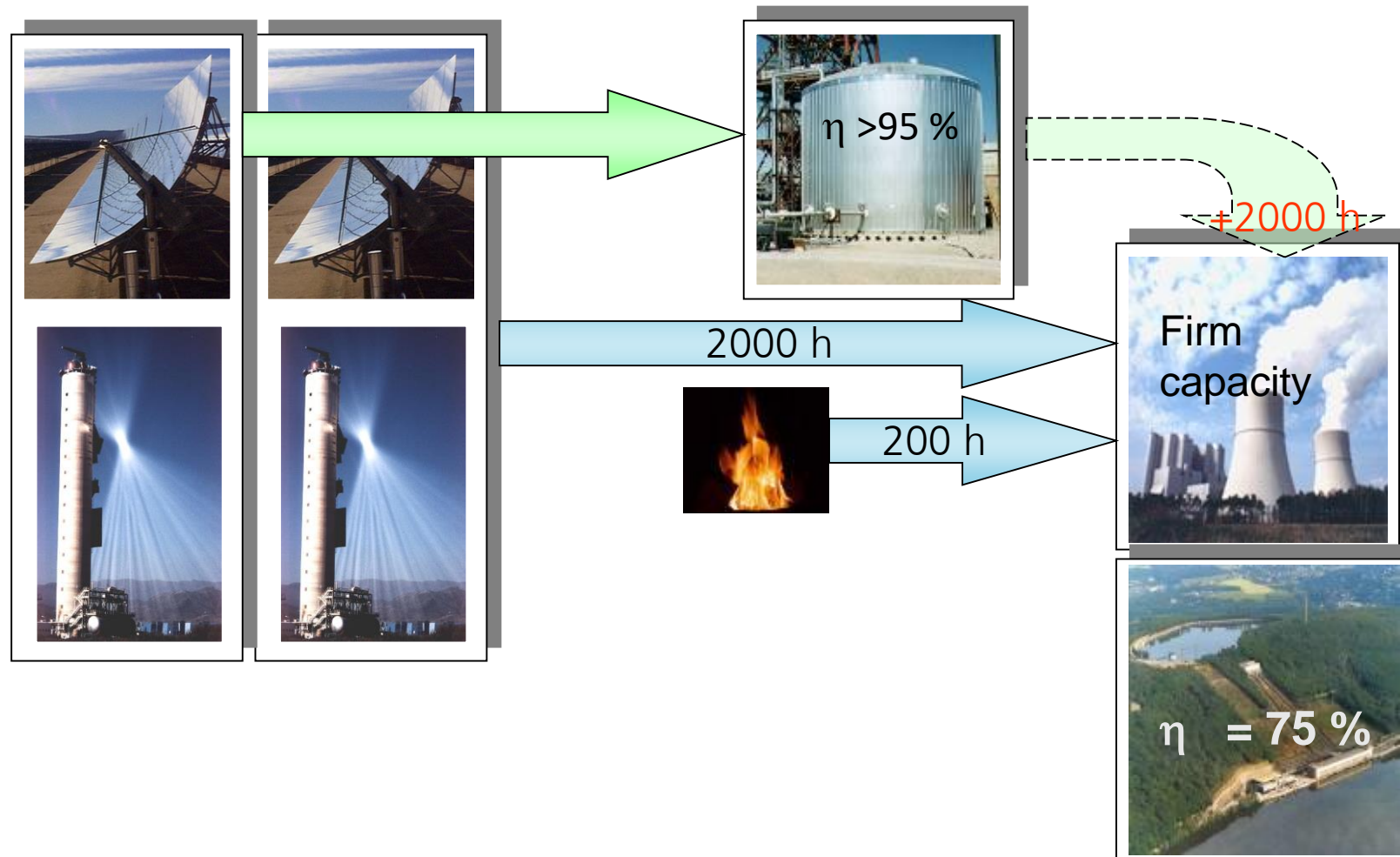


In a nutshell: From high temperatures to electric current



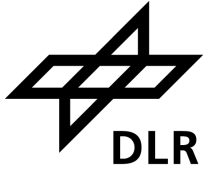
Thermal Storage vs. Electric Storage

CSP with thermal storage and fossil back provides reliable dispatchable power
at no additional cost

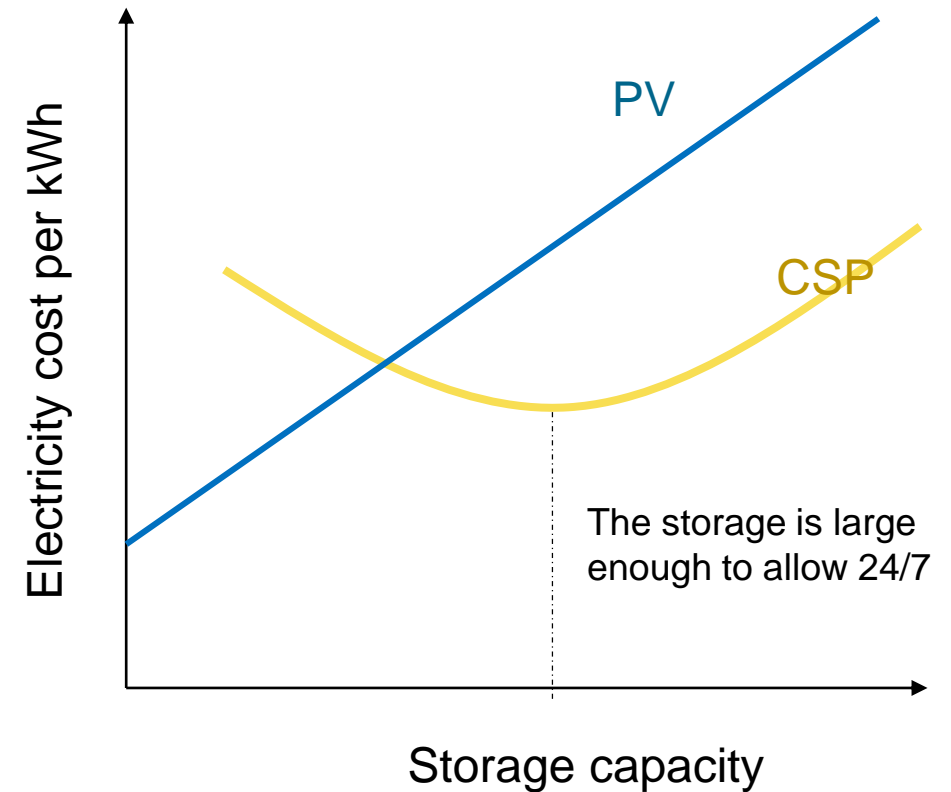


CSP / PV complement each other -

Security of supply around the clock cheaper than with oil, gas, and coal



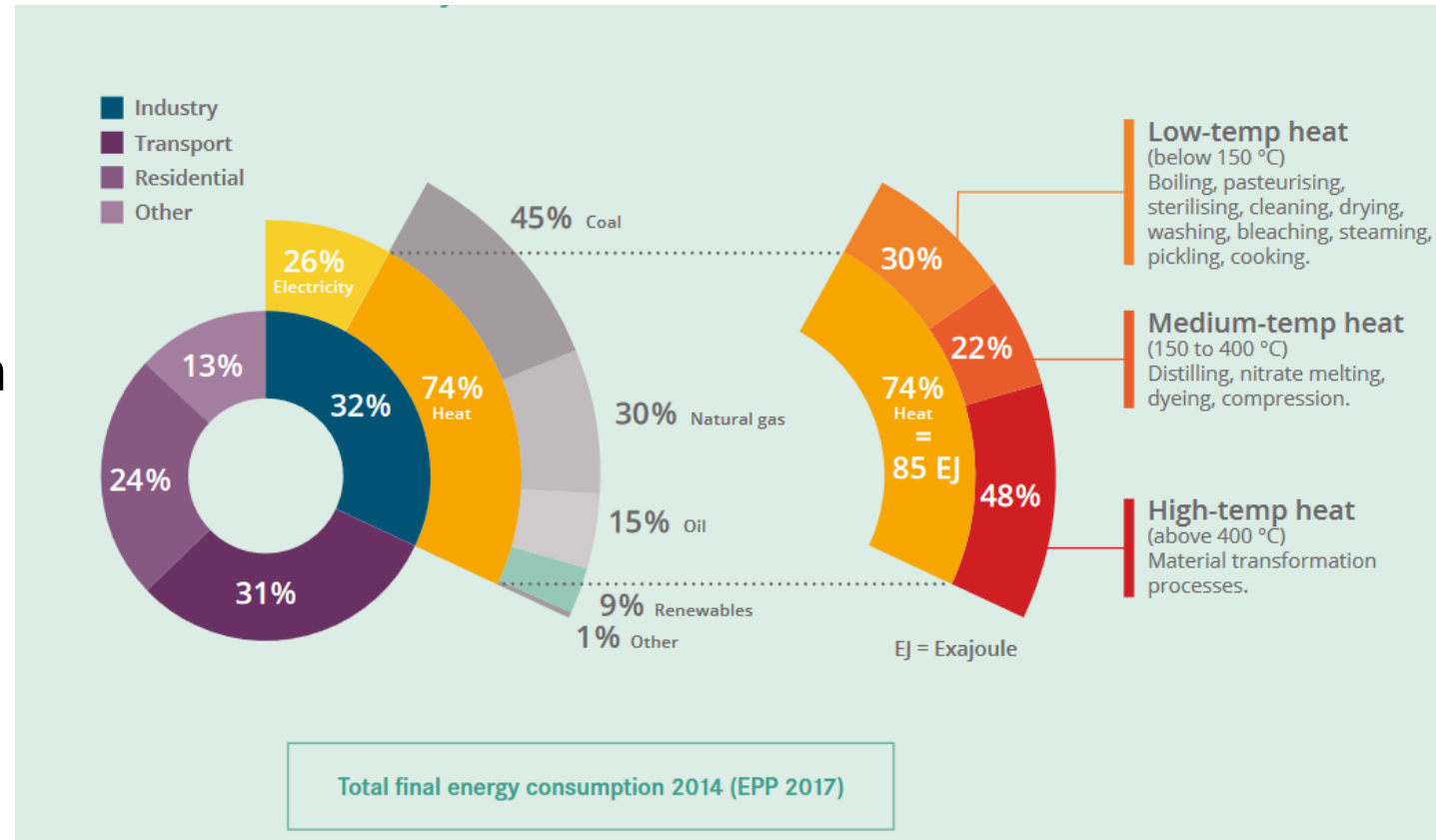
- CSP power plants with storage deliver **cheaper** power than CSP power plants without storage
- For PV systems, a system without storage always has the **lowest electricity costs**.
- **Hybrid systems (CSP + PV)** offer the lowest cost of electricity when **electricity is also needed at night**



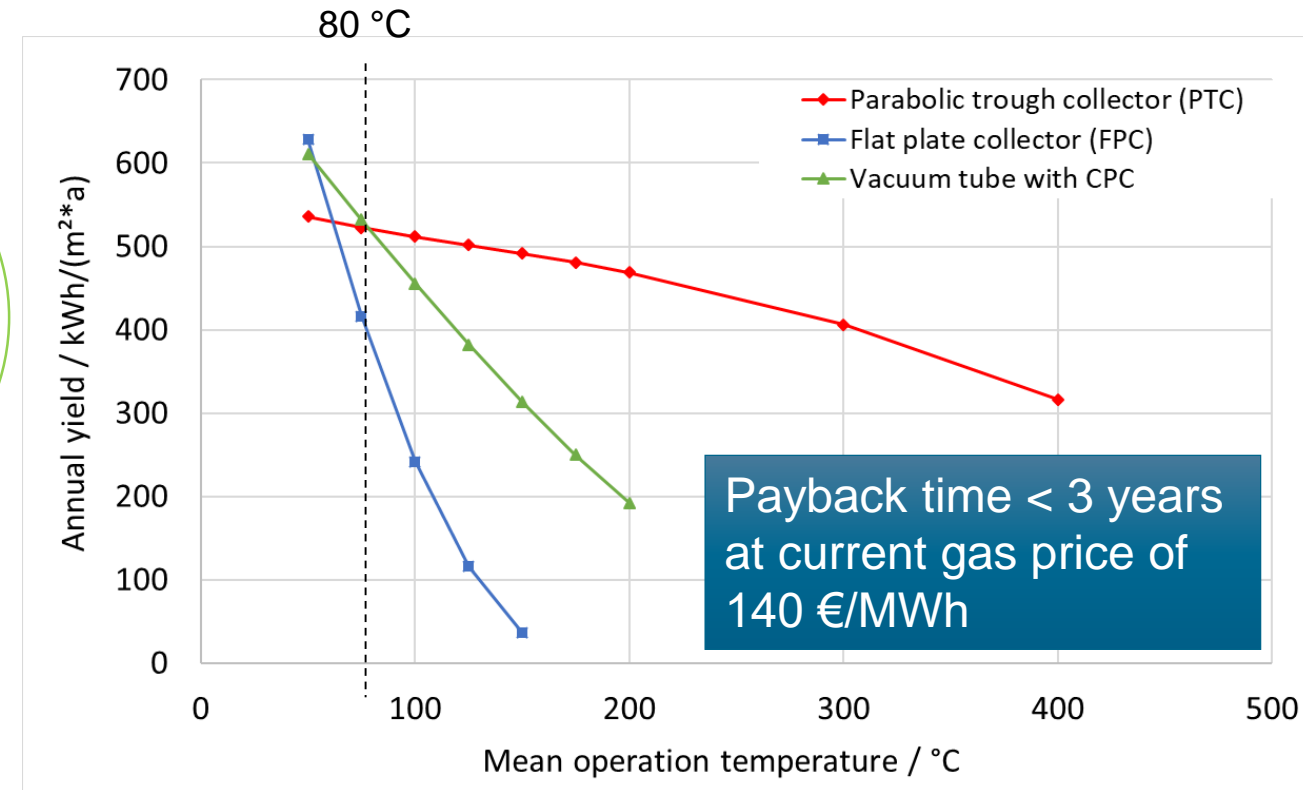
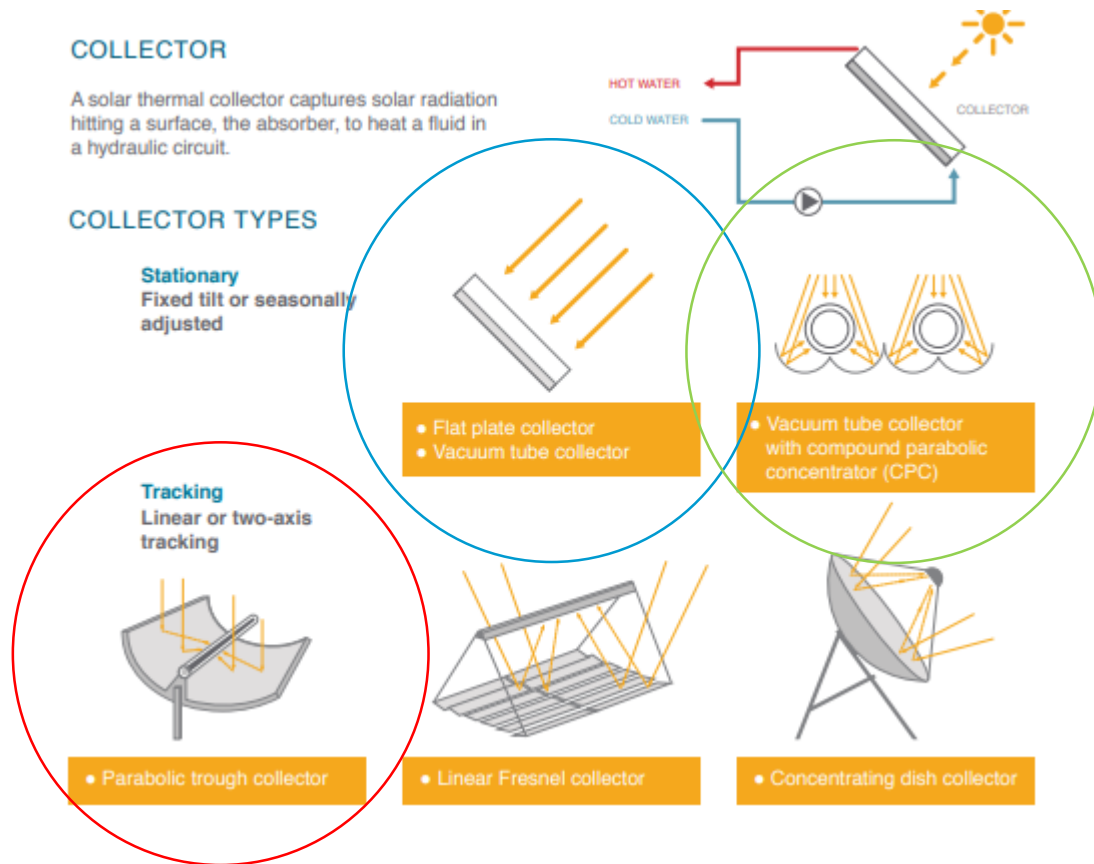
Market potential of CSP technologies for industrial process heat



- Mature technology with more than 800 MW installed worldwide
- Temperature range up to 400 °C fully commercial
- Round-the-clock operation through integrated thermal energy storage
- Low cost compared to fossil fuels
- Growing number of European companies committed to rapid deployment



Energy yield of solar collectors in Germany



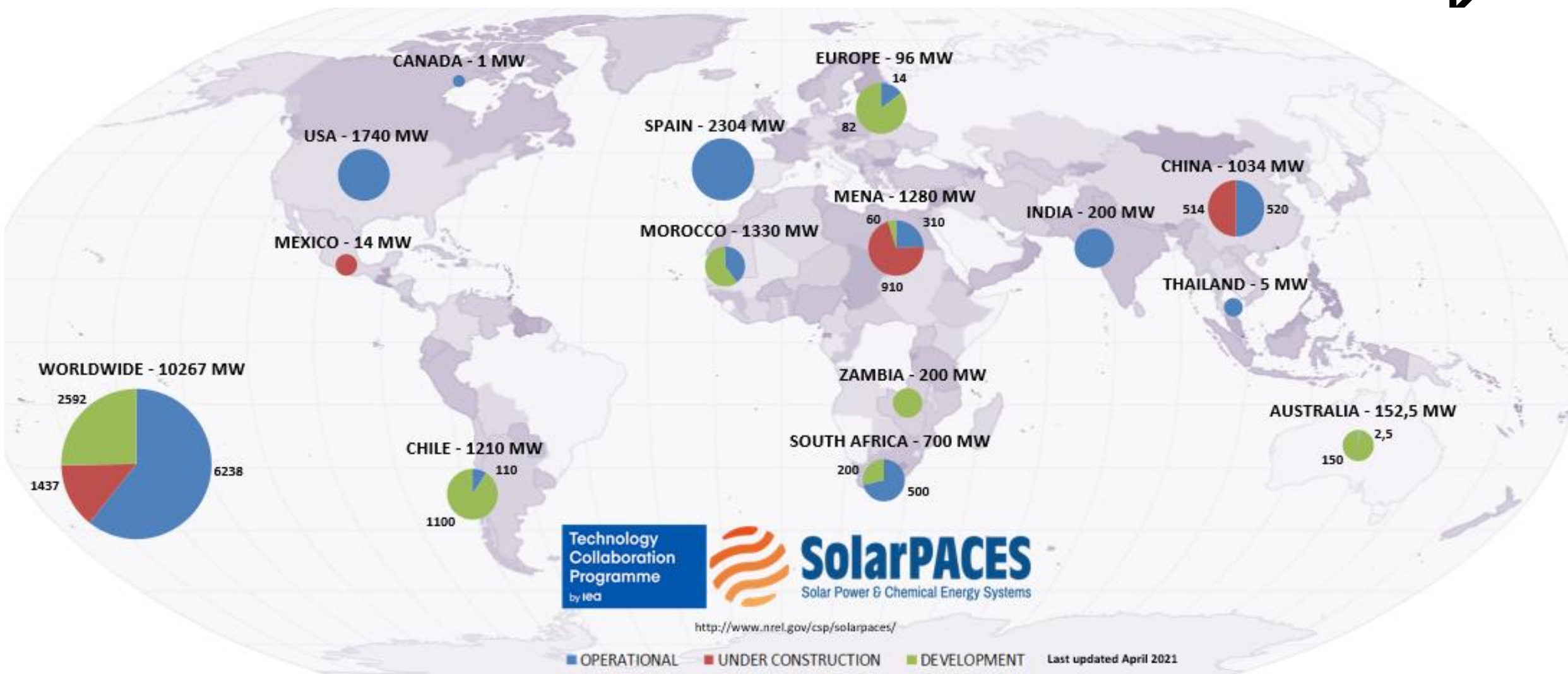
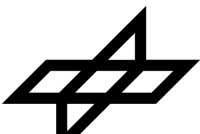
Advantages of concentrating collectors

- More yield from approx. 80 °C collector outlet temperature
- Reaches the desired storage temperature at any time of the year
- Operation also of heat networks with 130 °C flow temperature

The background image shows a large solar tower (CSP) in a desert landscape. The ground is covered with a vast field of solar collectors (heliostats) arranged in concentric circles around the tower. The sun is low on the horizon, creating a bright, hazy glow. The sky is a mix of blue and orange. A yellow banner is at the bottom.

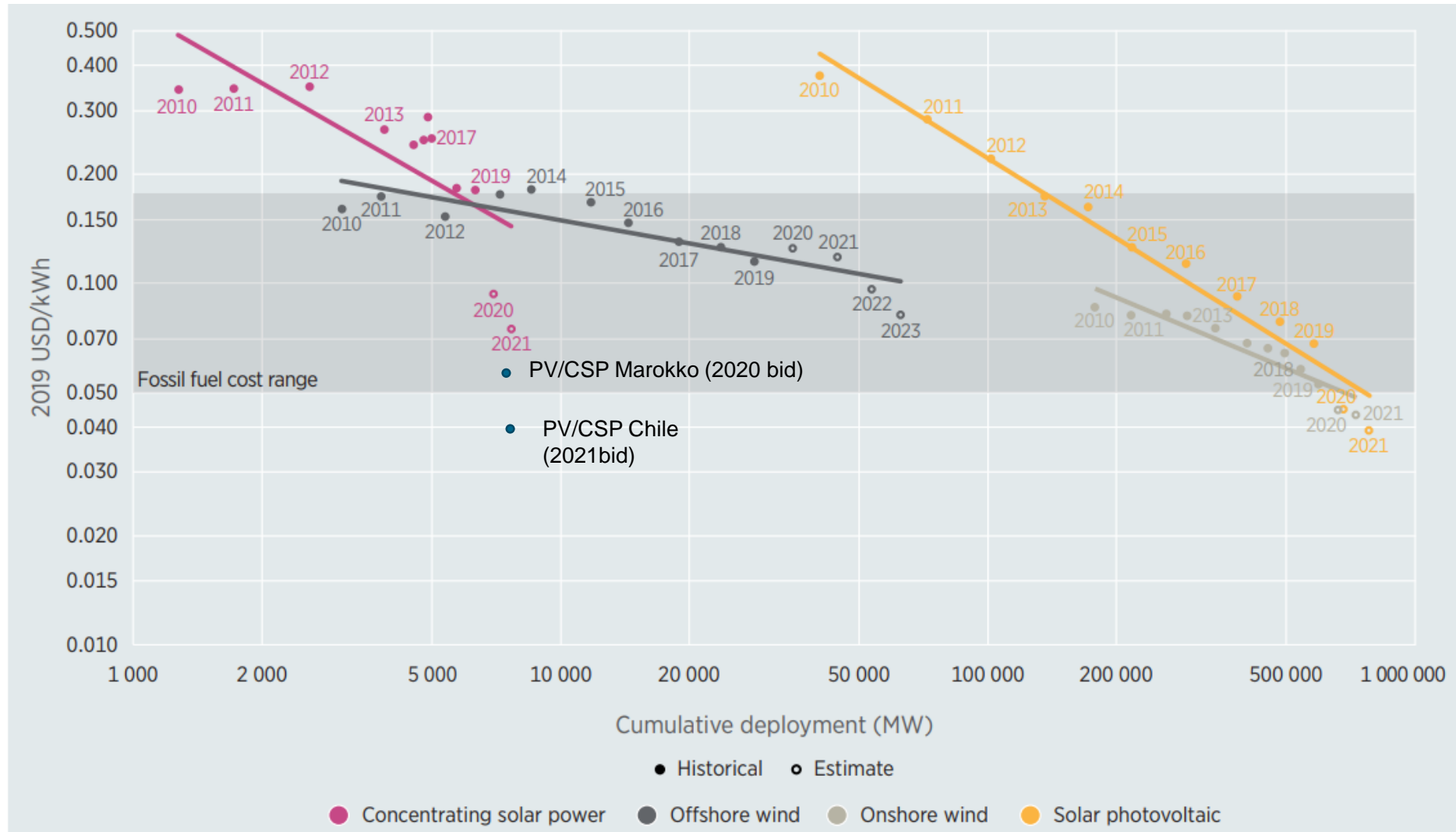
STATUS AND PERSPECTIVES

Current Market Overview CSP: 6.2 GW operational around the world



<https://www.solarpaces.org/csp-technologies/csp-projects-around-the-world>

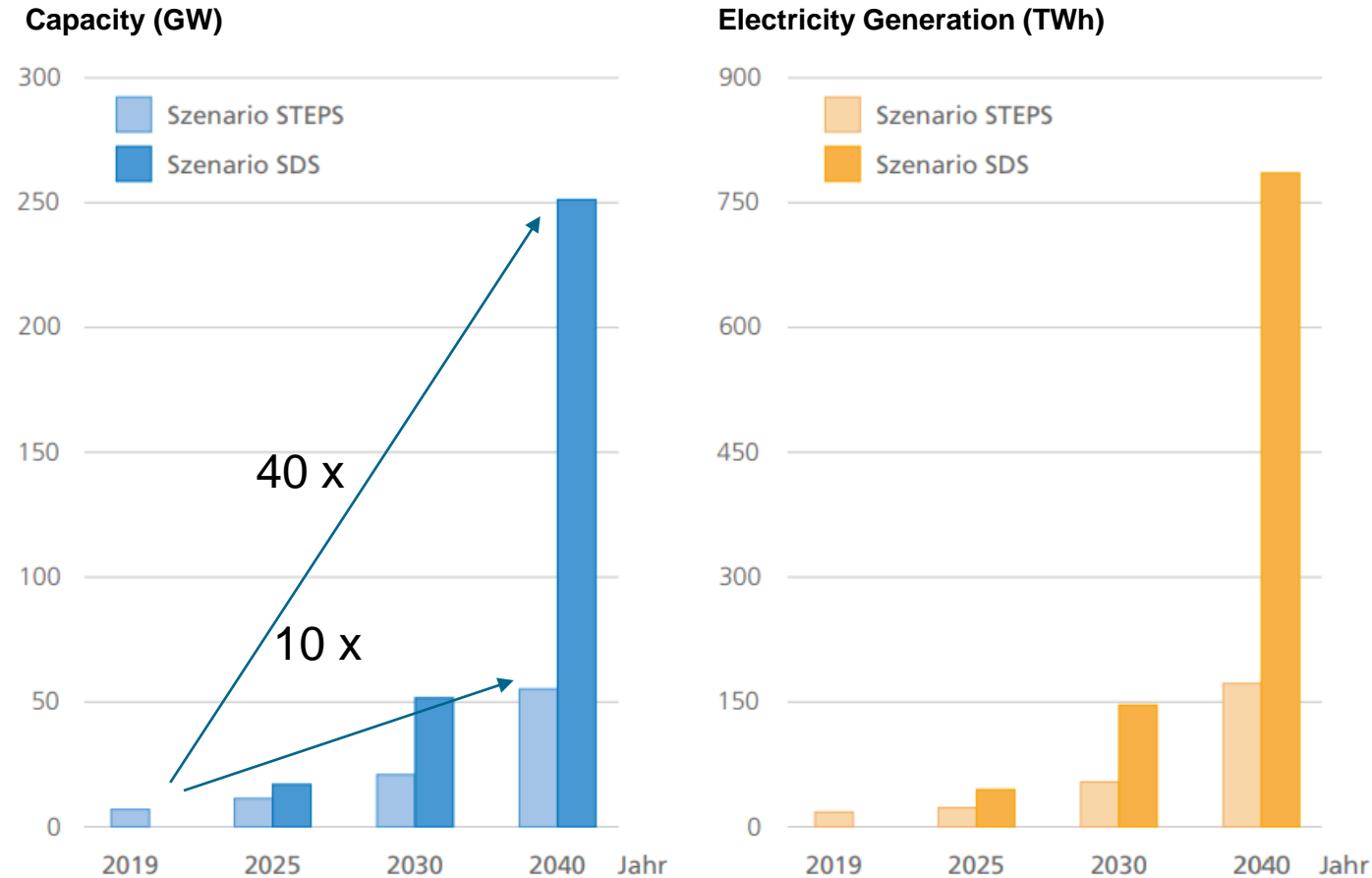
Strong cost degradation in CSP at relatively low total deployment



Source: IRENA, RENEWABLE POWER GENERATION COSTS IN 2019, Figure 1.11 The global weighted-average LCOE and Auction/PPA price learning curve trends for solar PV, CSP, onshore and offshore wind, 2010 – 2021/23

Possible CSP growth scenarios of IEA 2020-2040 (in conjunction with growing capacities of PV and Wind)

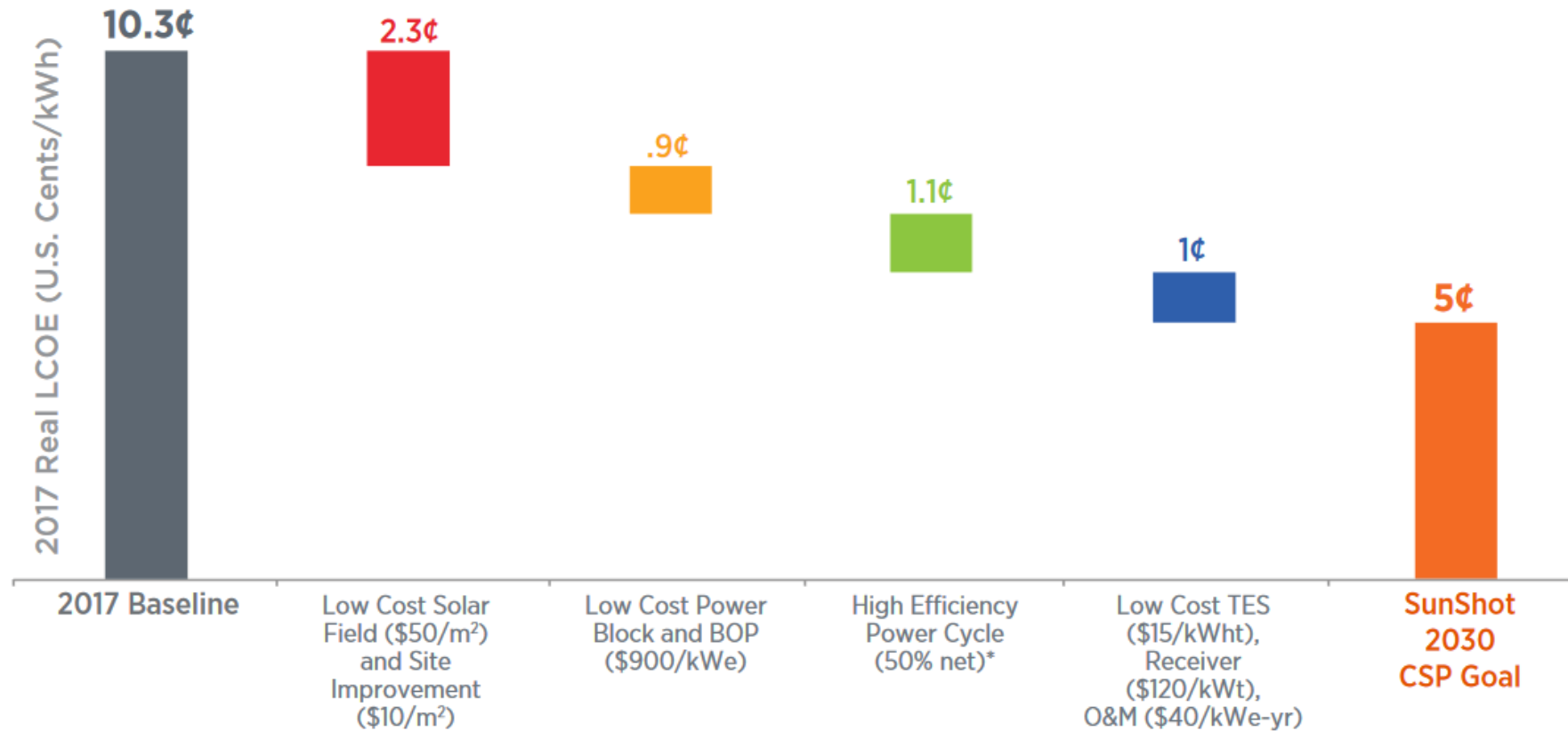
15 - 25 % annual growth rate estimated



STEPS: Stated Policies; SDS: Sustainable Development (<1,5 °C)

Source Data from IEA-WEO 2020, Table A.3

Cost reduction scenario of DOE



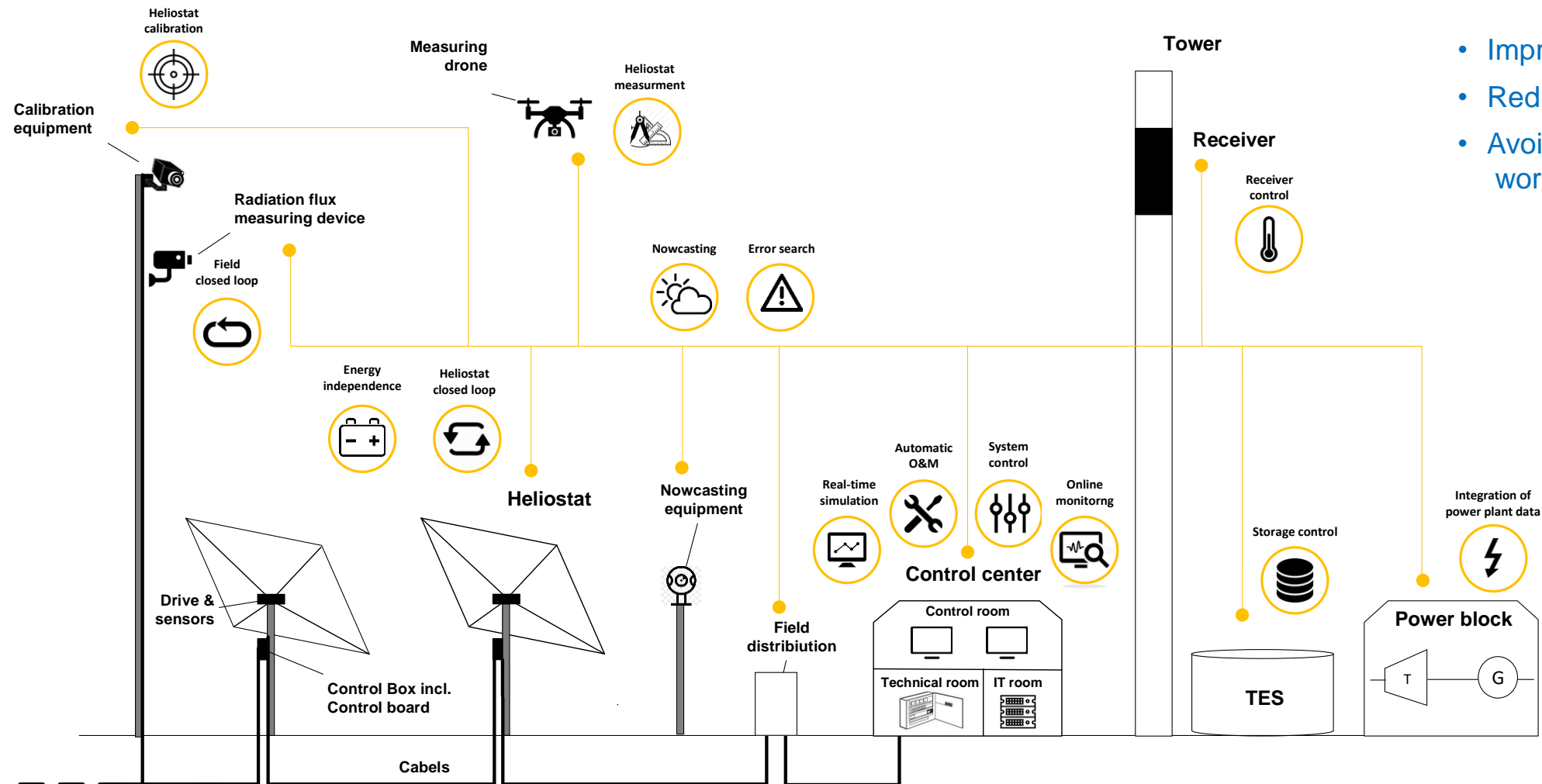
*Assumes a gross to net conversion factor of 0.9

Source: DoE

An aerial photograph of a Concentrated Solar Power (CSP) plant. The foreground and middle ground are filled with thousands of heliostats (mirrors) arranged in neat rows, reflecting sunlight. In the background, a tall, white receiver tower stands prominently. The plant is situated in a dry, open landscape with some trees and a body of water visible in the distance. A yellow banner with white text is overlaid at the bottom.

ADVANCED METHODS TO OPERATE A CSP PLANT EFFICIENTLY AND AUTONOMOUSLY

Autonomous CSP Plant Control based on Artificial Intelligence



- Improved energy yield
- Reduced OPEX
- Avoidance of harsh working conditions

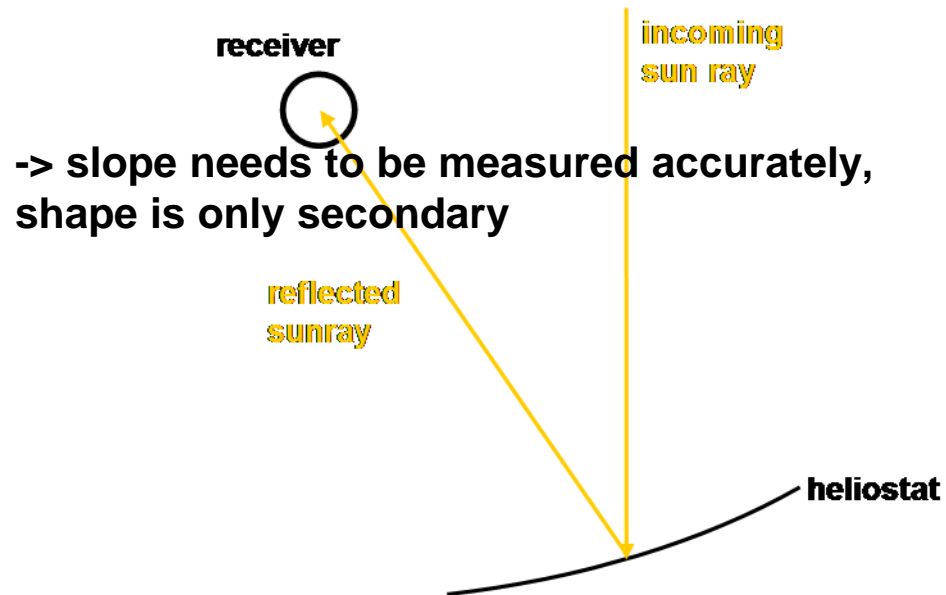
A white quadcopter drone is shown in flight, hovering over a vast field of solar panels. The drone has a camera mounted underneath and is equipped with various sensors and antennas. The solar panels below are arranged in neat rows, and the background is a clear blue sky with some light clouds.

AIRBORNE CONDITION MONITORING OF SOLAR FIELDS

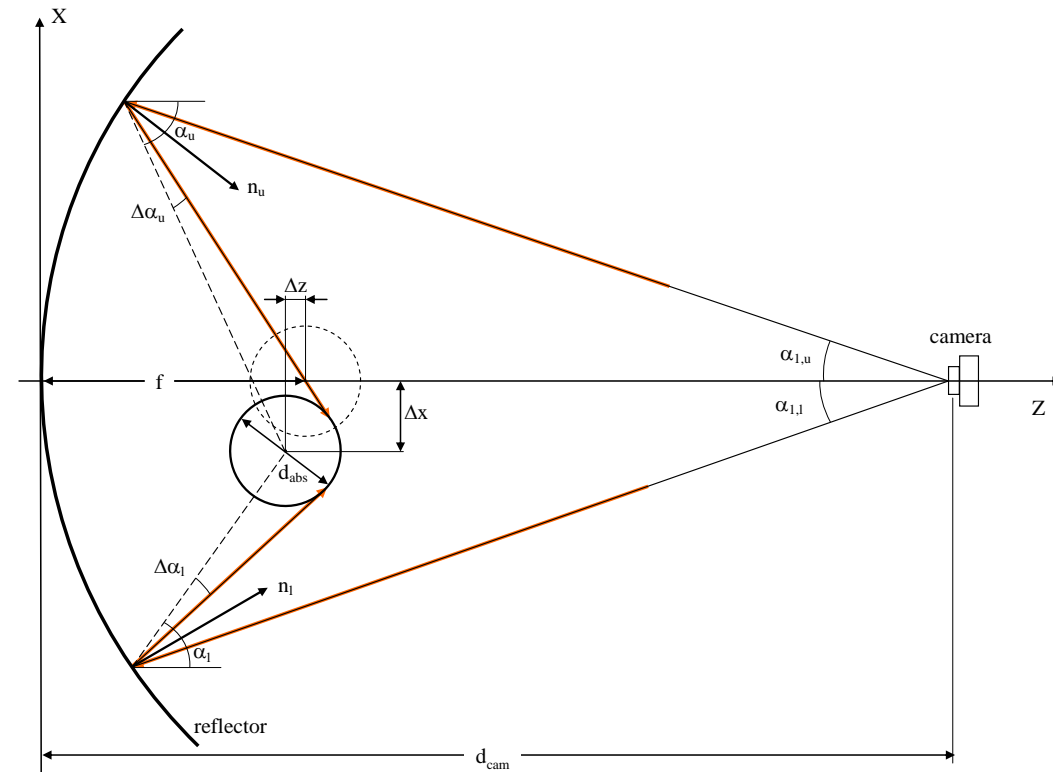
Introduction: Shape and Slope Deviations

Deviations of the ideal shape of curved mirrors for CSP applications can have a significant impact on the optical efficiency and thus the performance of the power plant.

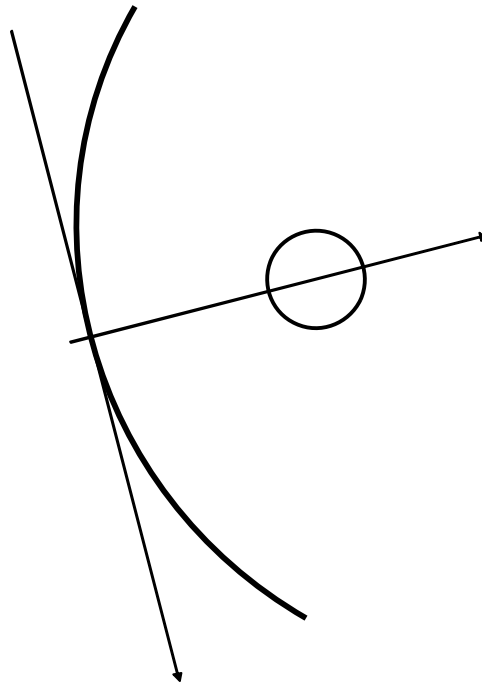
Critical measure is slope deviation, not shape deviation:



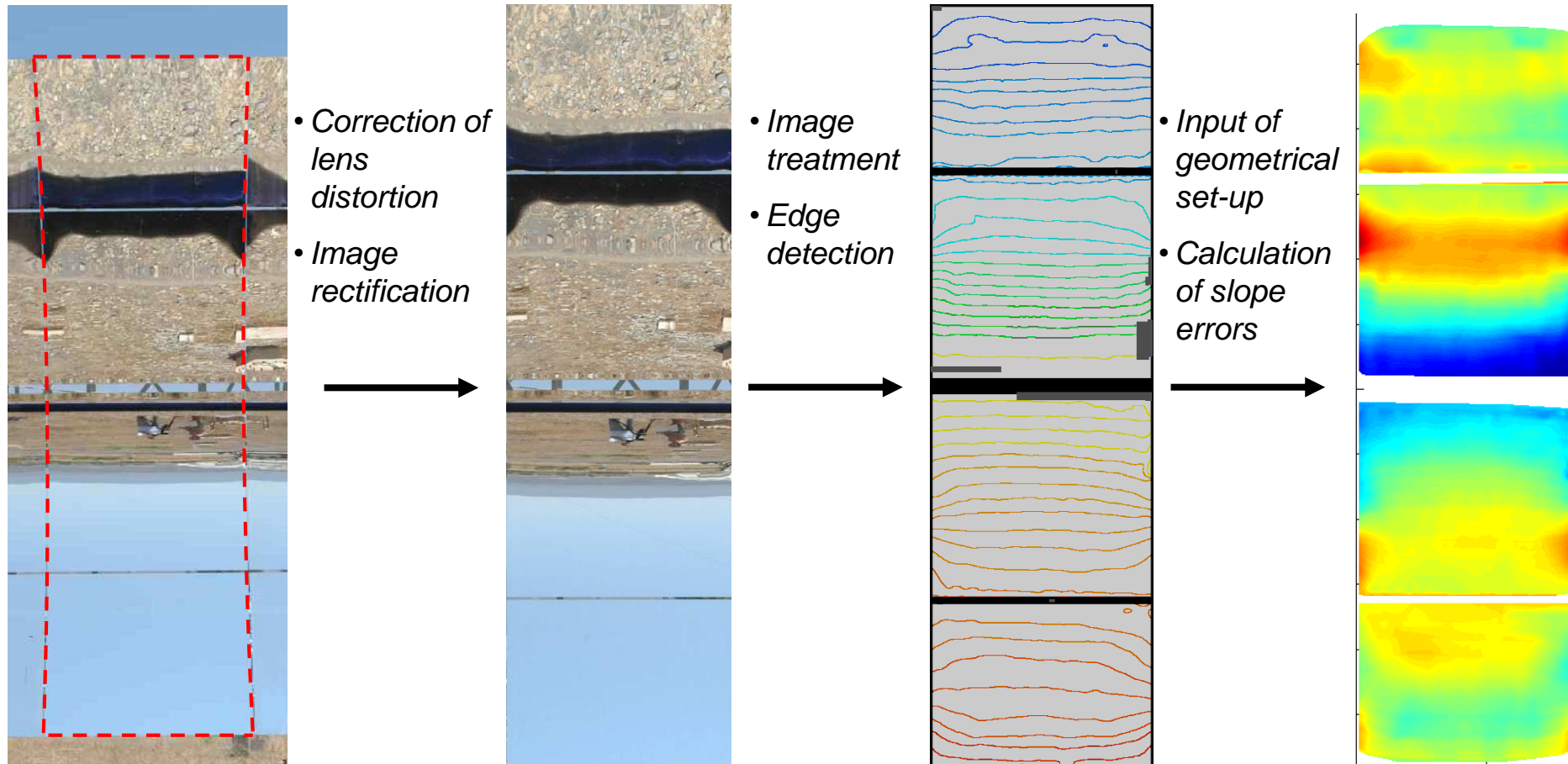
20



Measurement: Turning of collector with camera at close distance (~17 m)



Evaluation



Example ① Parabolic Trough Shape Accuracy

QFly Data Acquisition

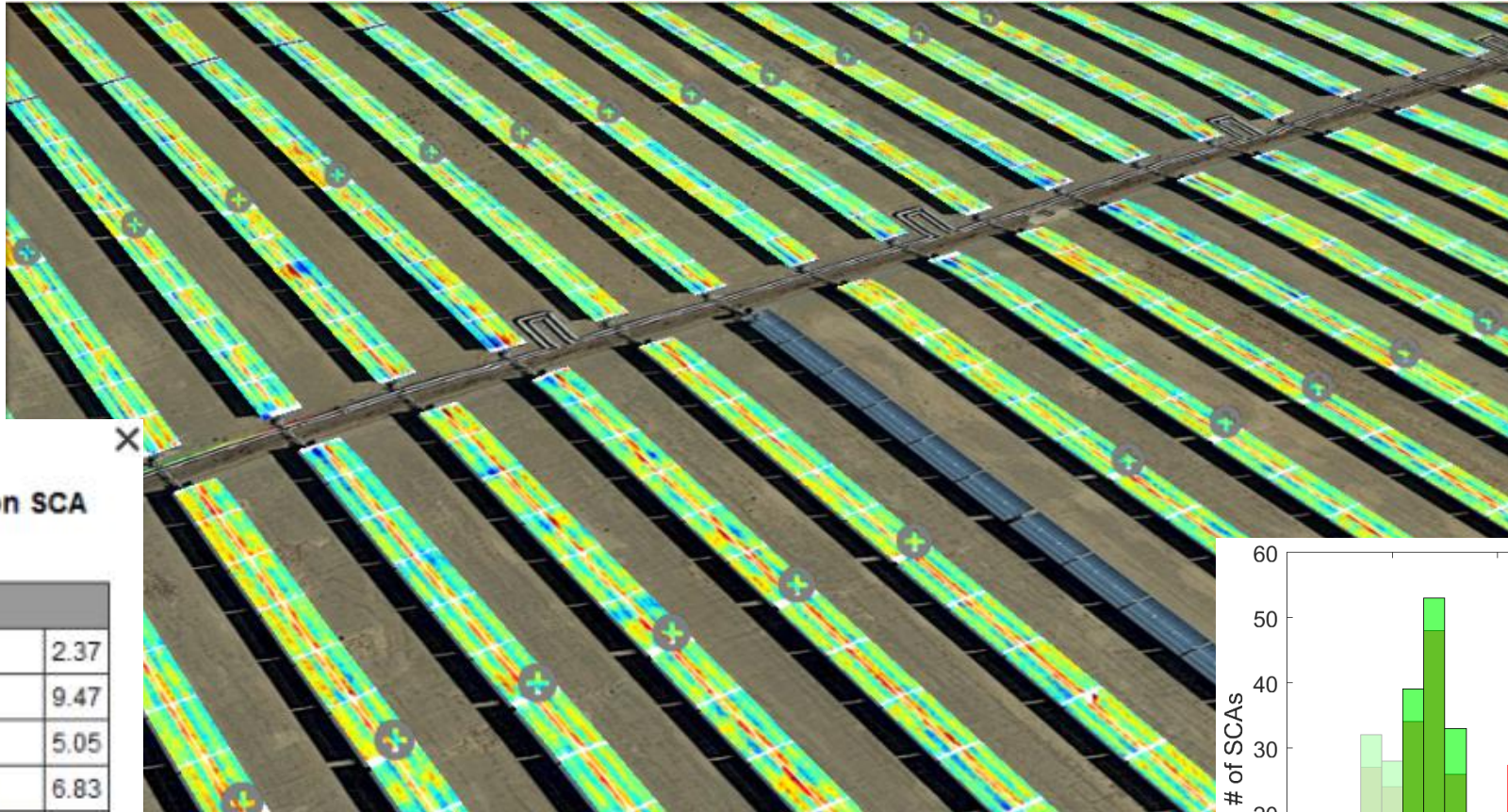


Example 1 Parabolic Trough Shape Accuracy

QFly Results: Solar Field



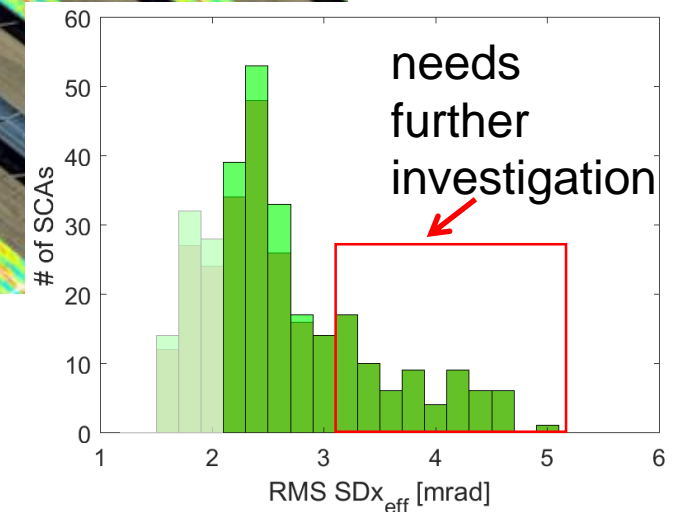
Slope Deviation $SD_{x,eff}$ for the whole collector (SCA) in mrad



Additional Information on SCA 22

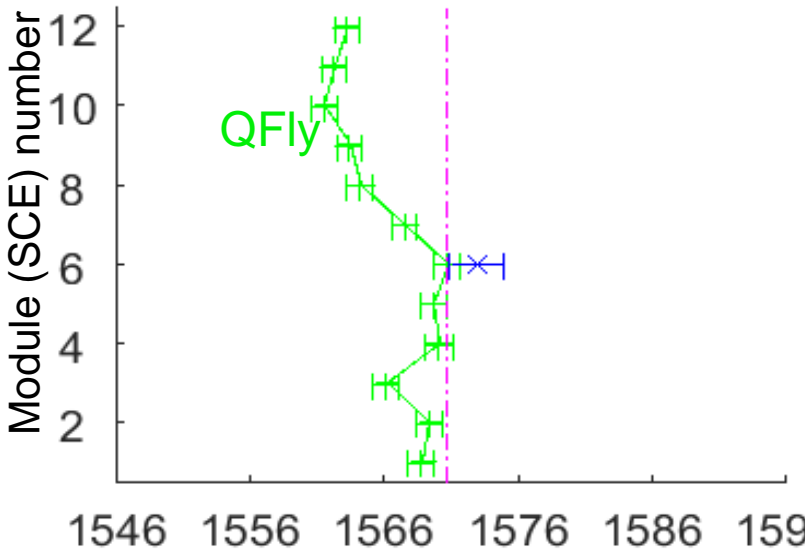
Additional Information	
RMS_SDx	2.37
RMS_FDx	9.47
dtrack_SurveyMinusLOC	5.05
dtrack_SurveyMinusTarget	6.83
dtrack_RMSWithinSCArel2Drive	1.65

Directions: [To here](#) - [From here](#)



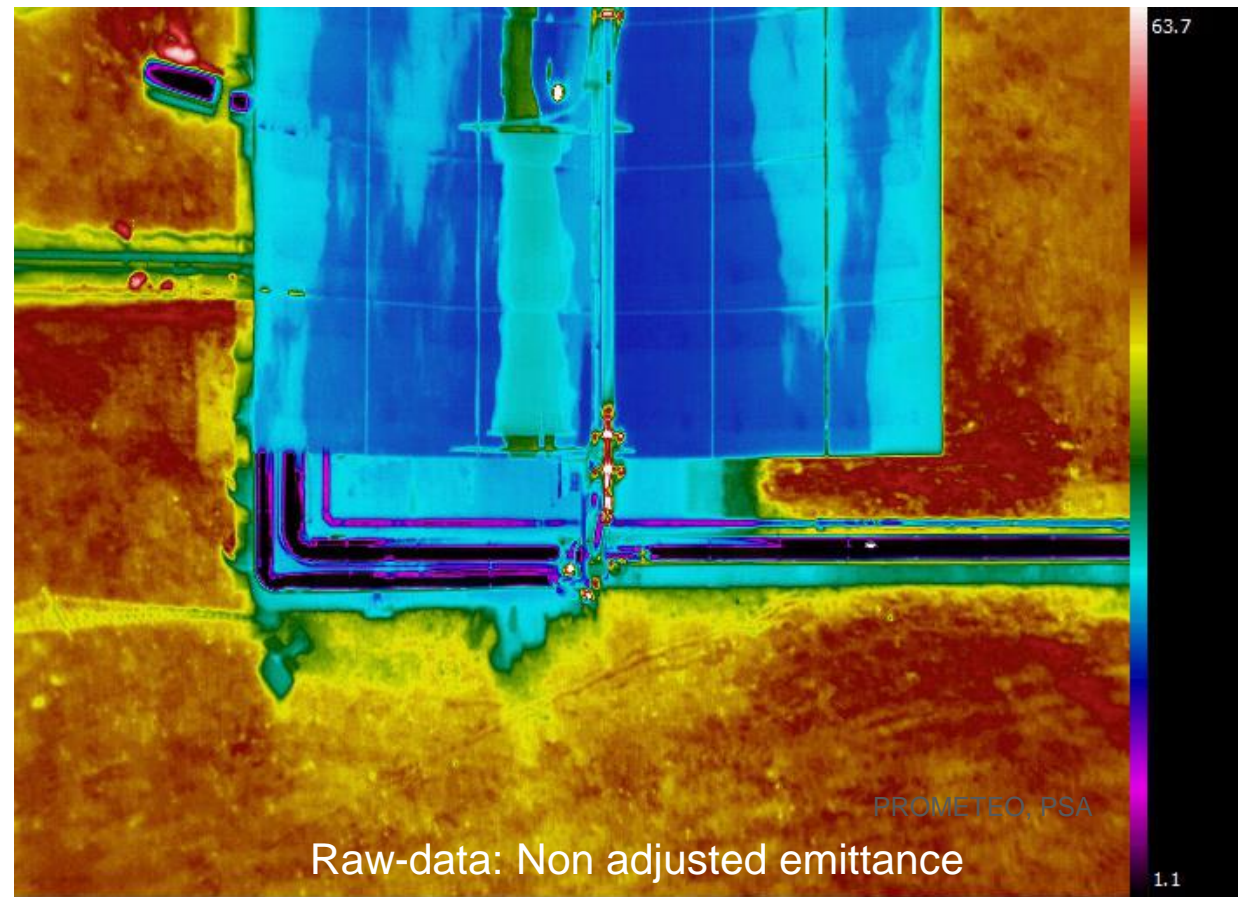
Example 2 Parabolic Trough Torsion/Tracking

QFly Results: Collector (SCA)



Example 2 Parabolic Trough HCE Quality Screening

- Measurement of glass envelope temperature by IR camera
- Measurement accuracy ~ 2 K

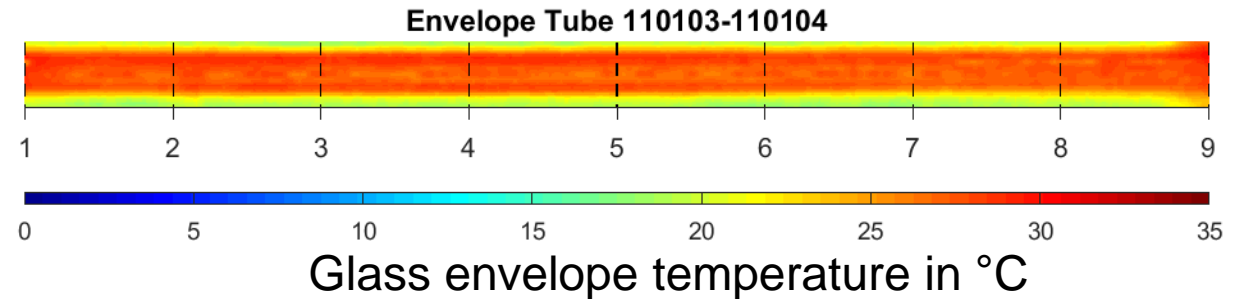


Example 3 Parabolic Trough HCE Quality Screening

QFly Results: Collector (SCA)

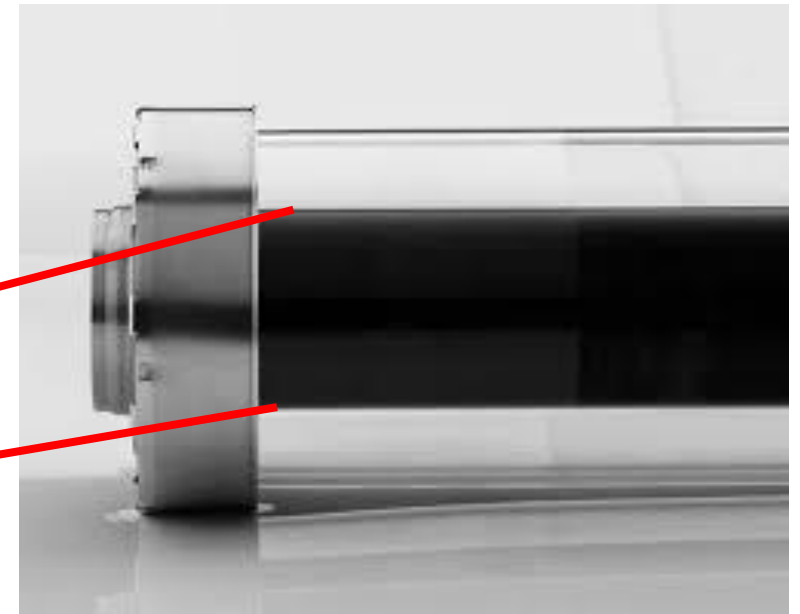
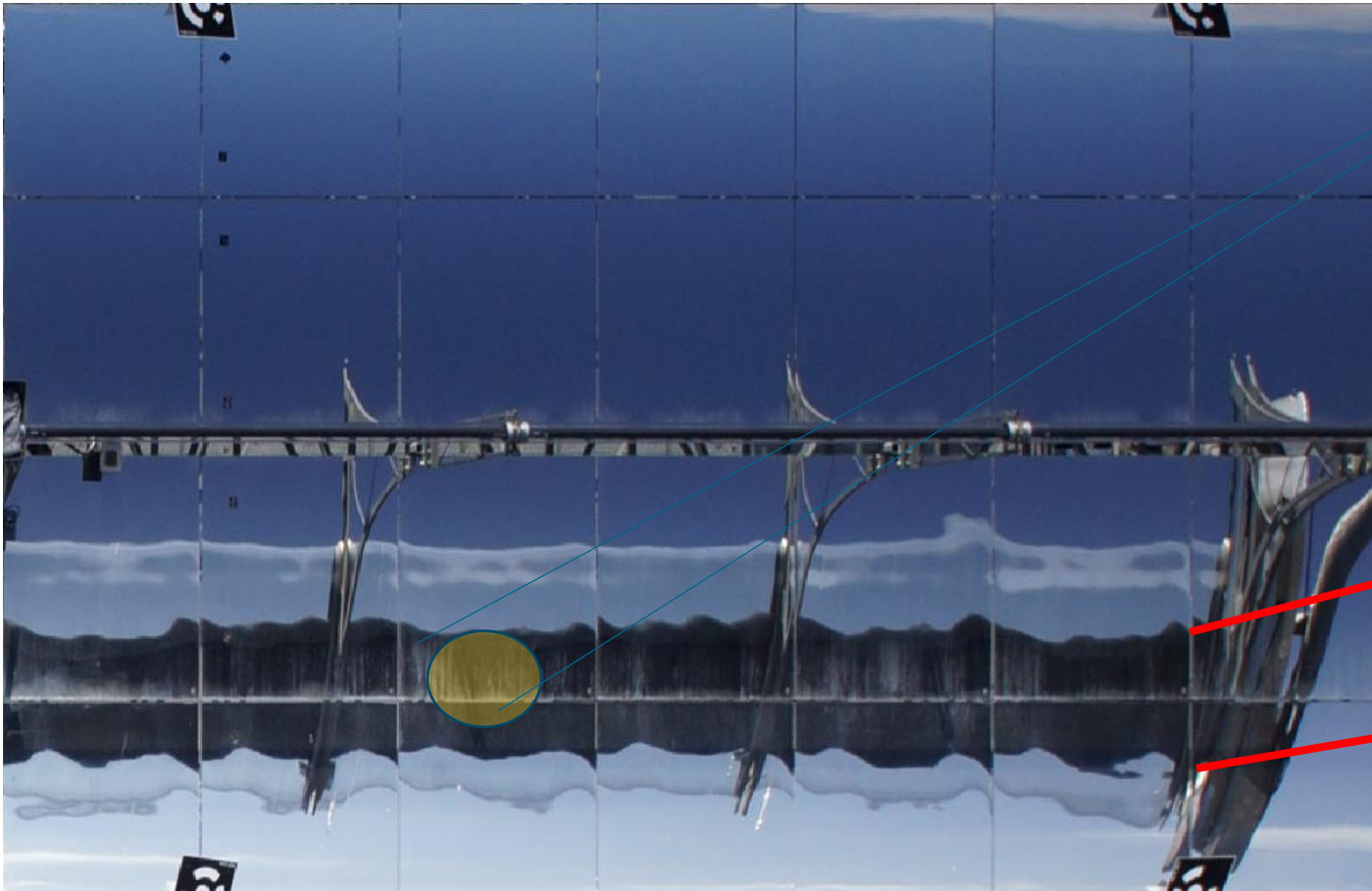
- Measurement of glass envelope temperature by IR camera
- Measurement accuracy ~2 K

- Automatic evaluation for glass temperature
- Automatic location of receivers in solar field and reporting
- Also applicable to other piping

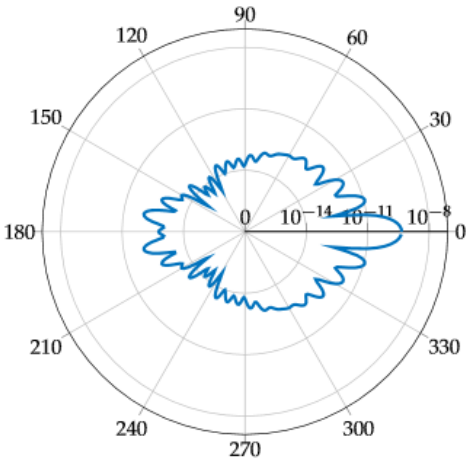
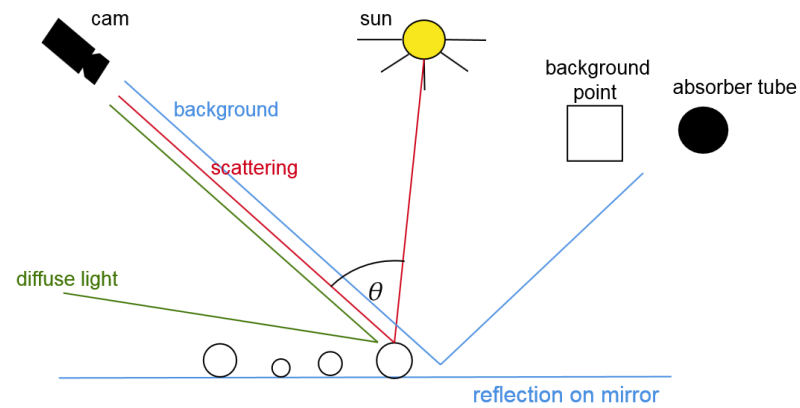
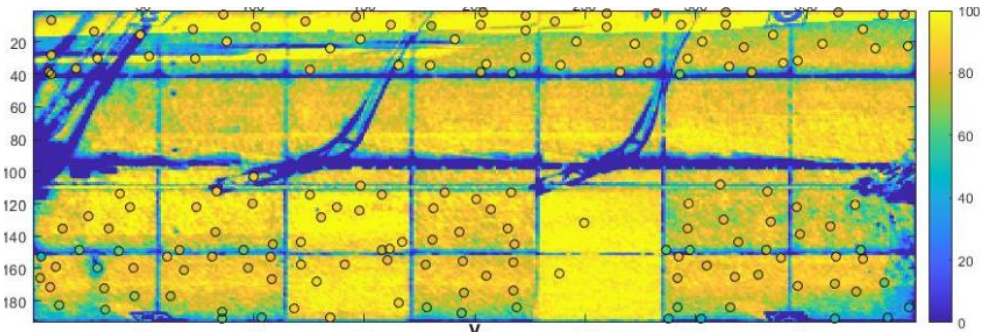
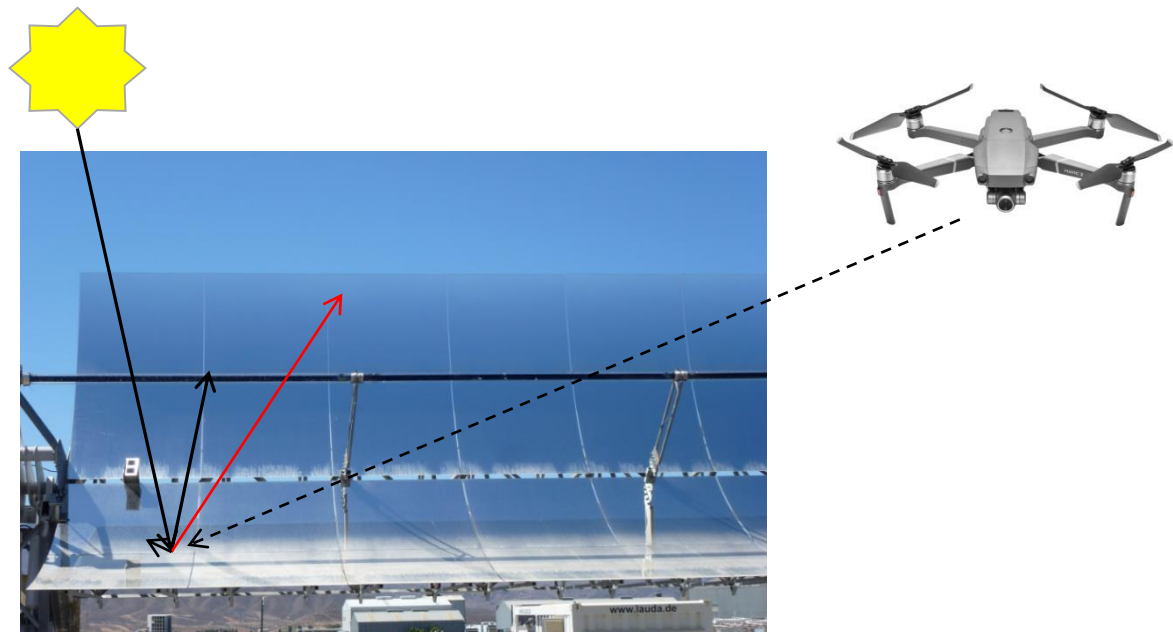
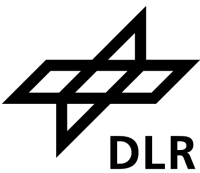


Example 4 Soiling Maps

Soiled regions are brighter



Creating soiling maps

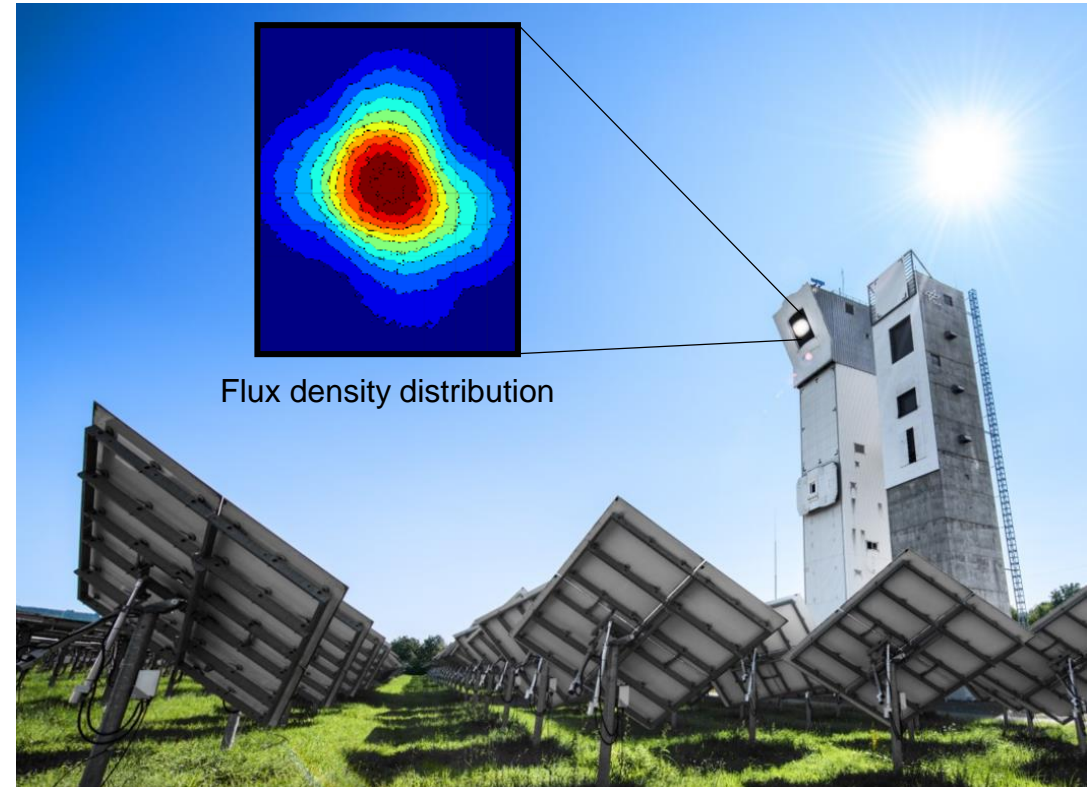


A wide-angle photograph of a solar tower (CSP) in a desert. The ground is covered with a vast field of solar collectors (heliostats) that reflect the bright sun, creating a shimmering, concentric pattern. The sun is low on the horizon, casting a long, golden glow across the sky and the landscape. The tower itself is a tall, slender structure with a crane on top, standing prominently in the center of the field.

DATA DRIVEN FLUX DENSITY PREDICTION

Motivation - flux density distribution

- The flux density distribution on the receiver is the power distribution caused by the concentration of the solar radiation
- This flux density should maintain requirements for materials and homogeneity



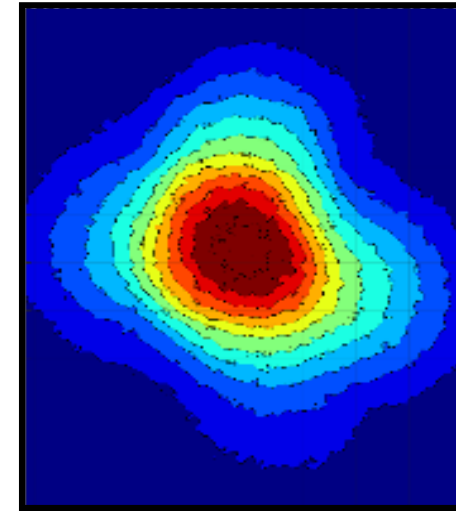
Therefore a more precise knowledge of the flux density allows

- a more efficient control
- longer durability of the components

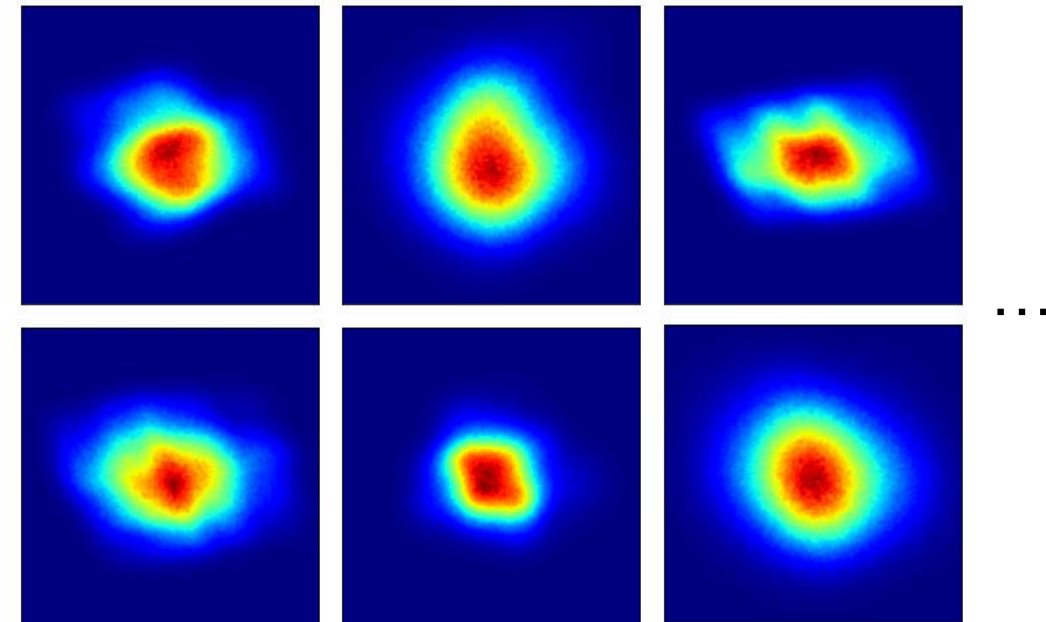
Flux spot of a single heliostat

- The flux density consists of the superposition of the focal spots of the individual heliostats of the field
- The focal spots can have different sizes and shapes and vary with the position of the sun
- We want to predict the focal spots of the individual heliostats at any sun position

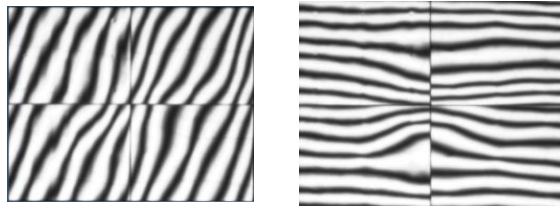
Flux density distribution



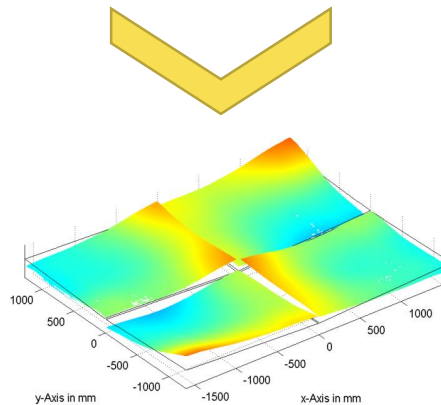
Focal spots of individual heliostats



State of the art: Stripe deflectometry

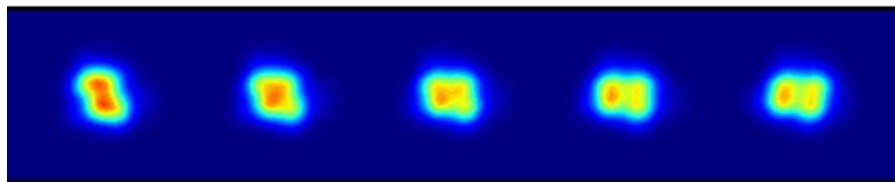


Pictures of heliostat from stripe deflectometry measurement



calculate surface deformations

Predictions



Sun position \vec{s}

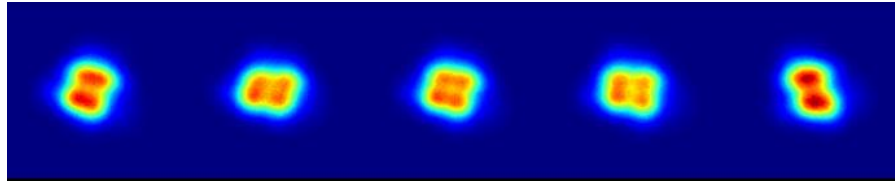
Use surface within a raytracer to get flux predictions



[Automated high resolution measurement of heliostat slope errors, *Ulmer et al.*]

Idea: Use existing data to improve flux predictions

Measured train data: Flux $X(\vec{s})$

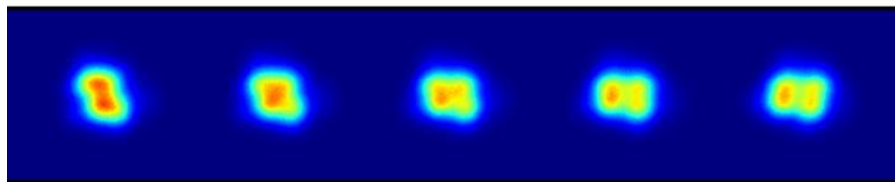


Measured data from target of single heliostats during calibration process

heliostat model

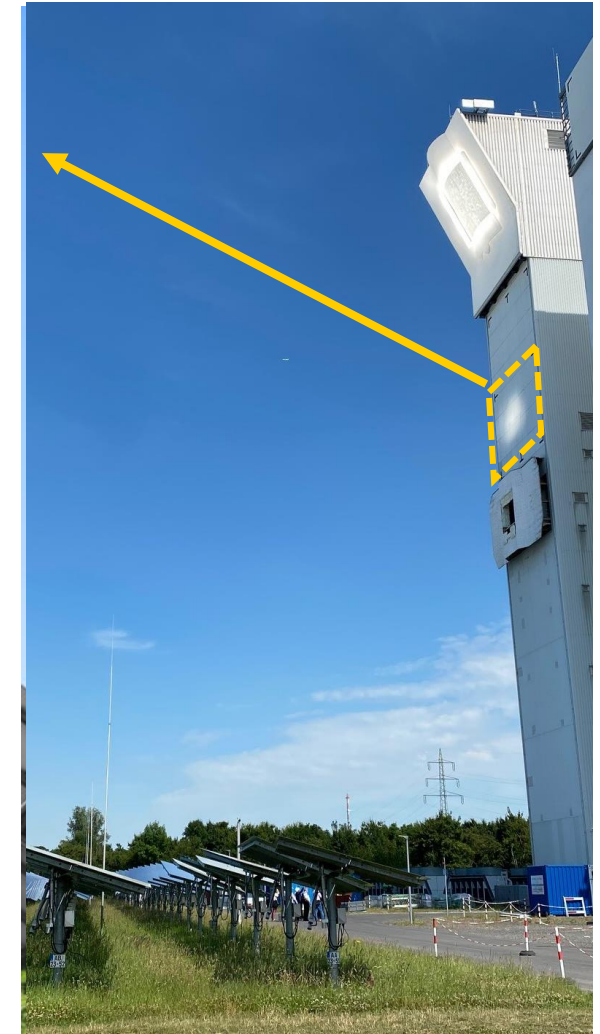
Use data to train a heliostat model

Predictions

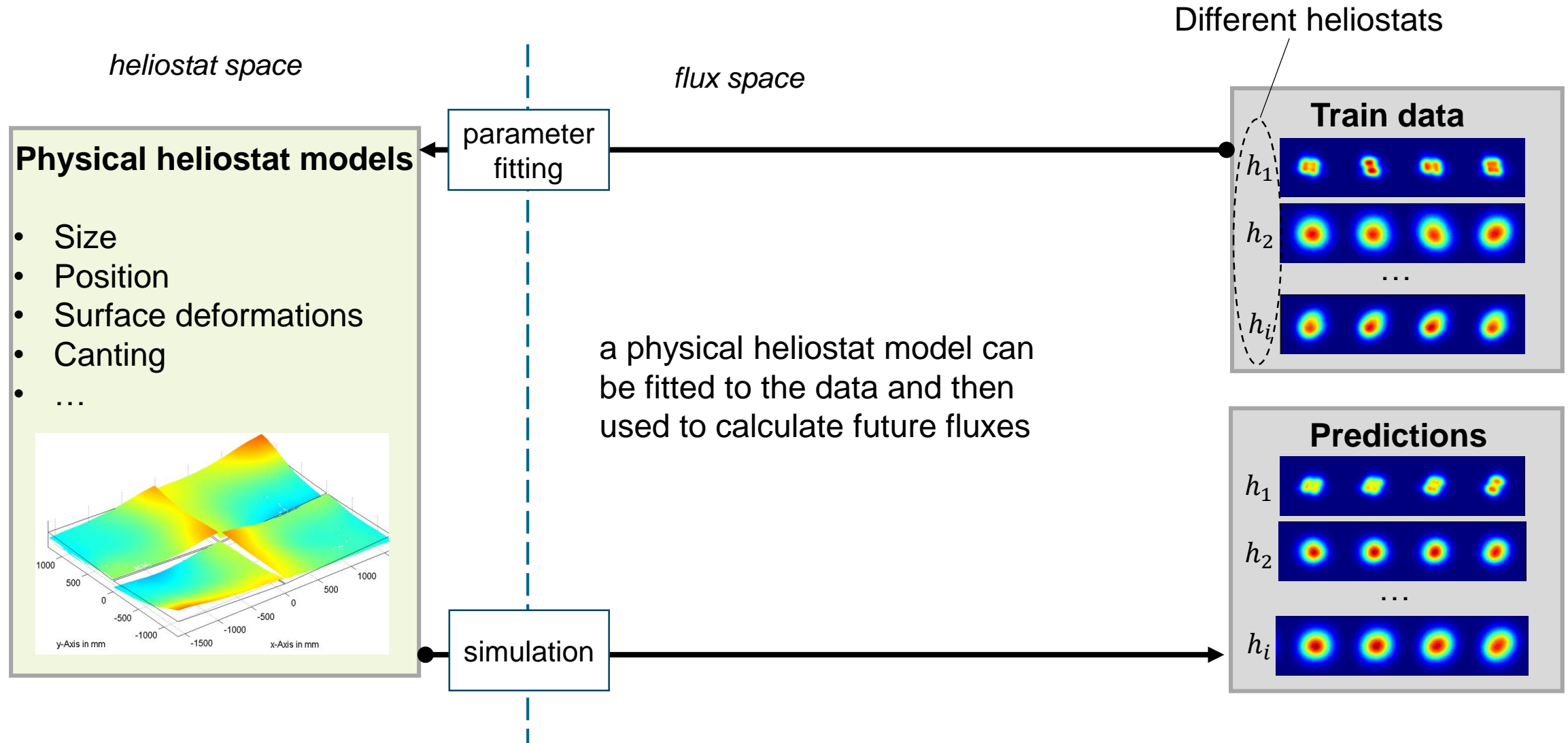


Use trained model to predict any future flux density distribution

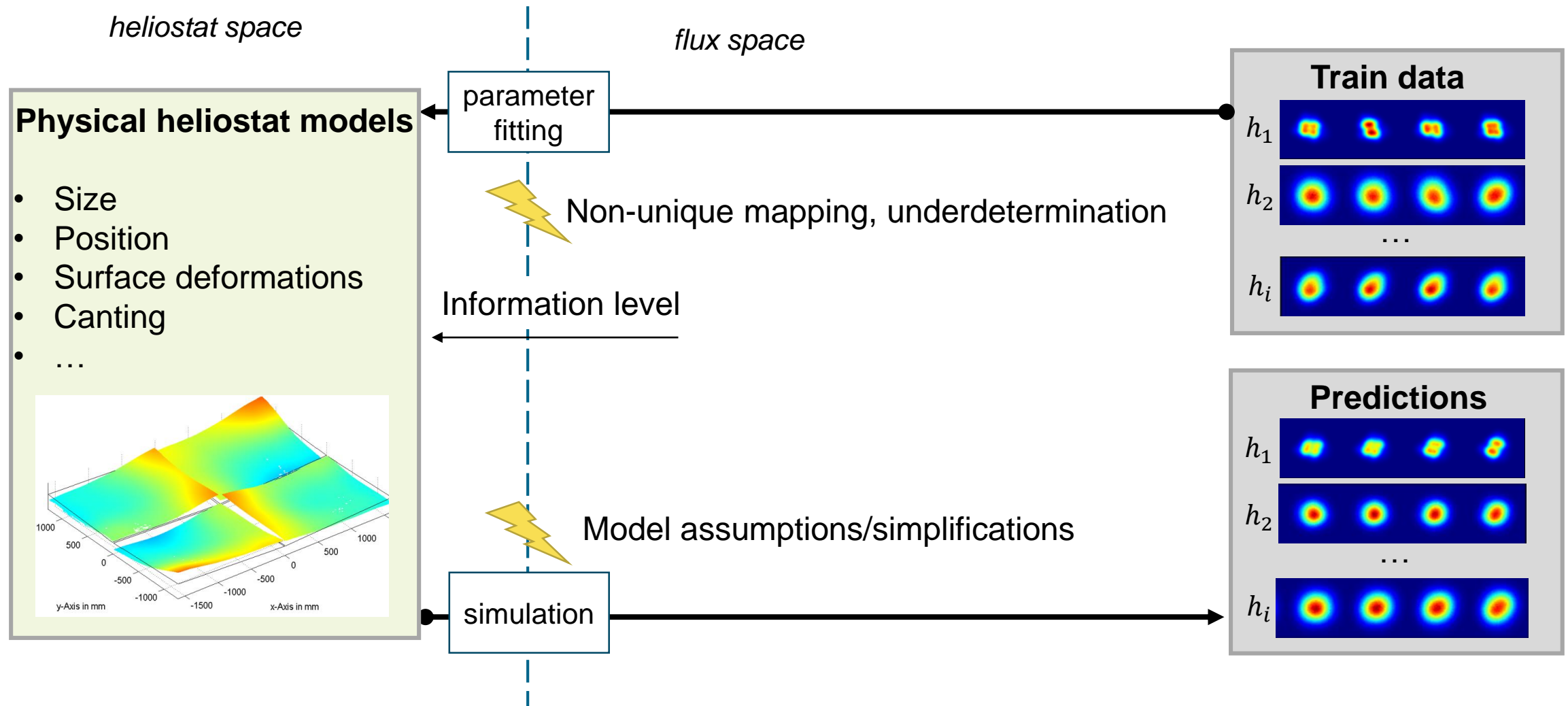
Sun position \vec{s}



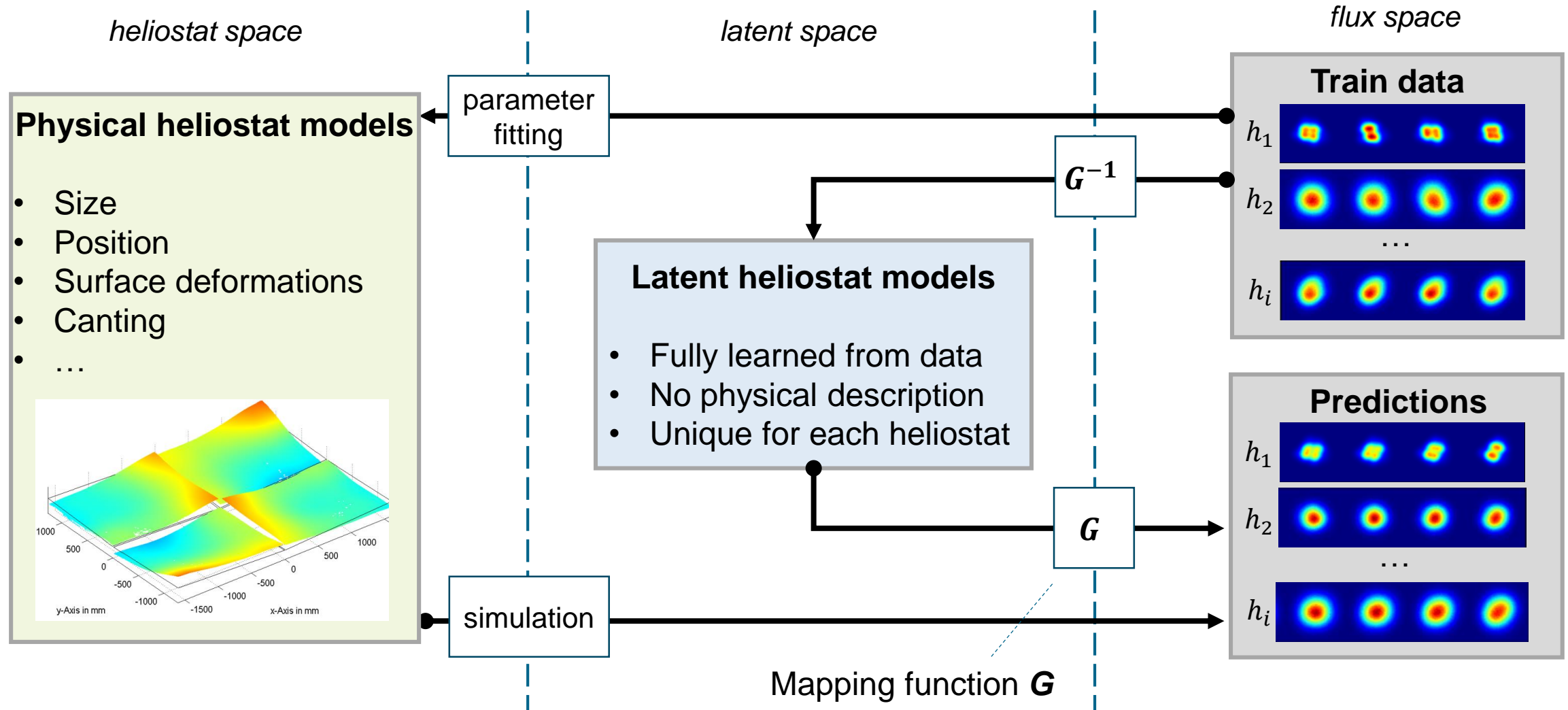
Physical vs. data driven heliostat model



Physical vs. data driven heliostat model

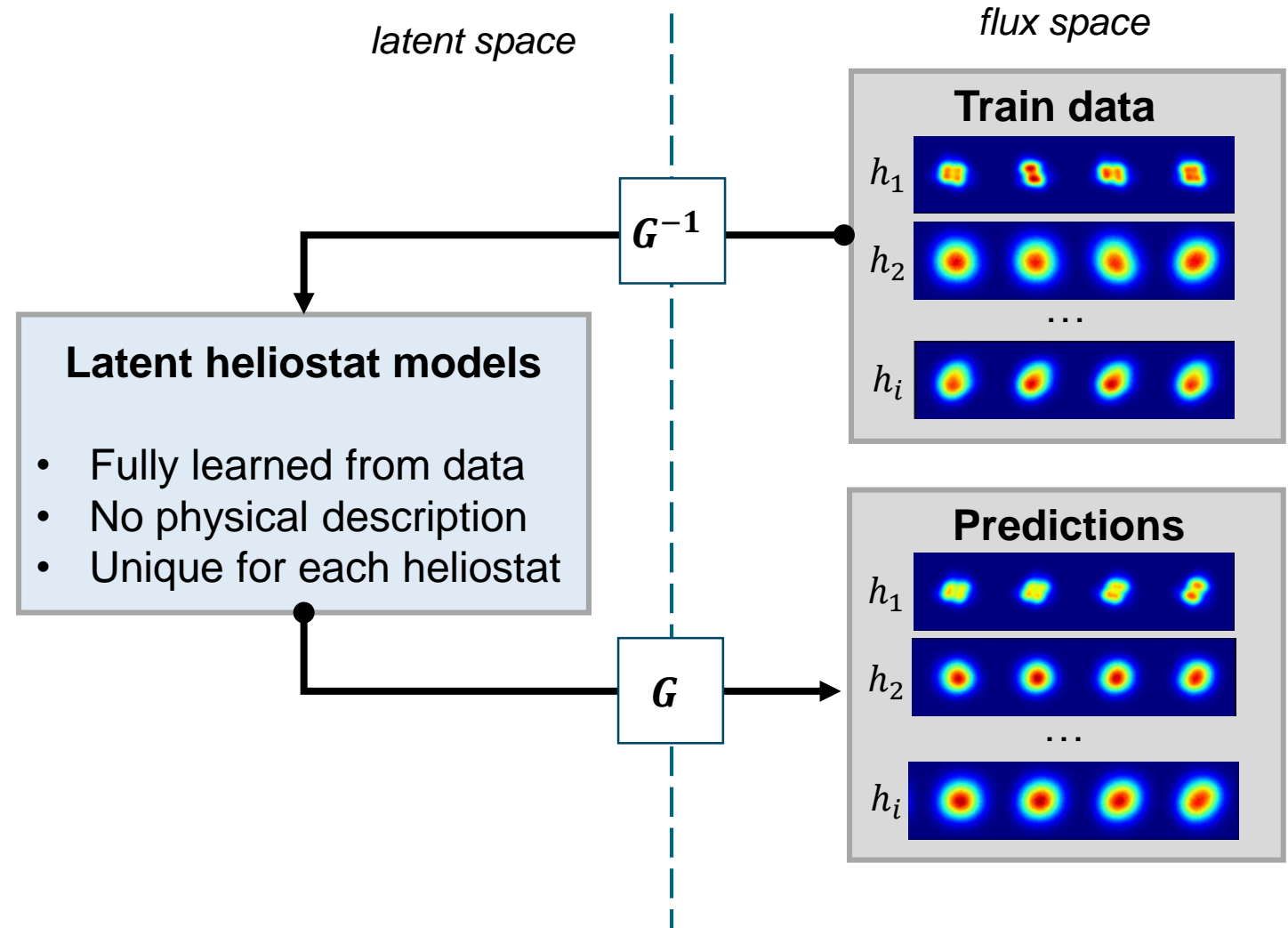


Physical vs. data driven heliostat model

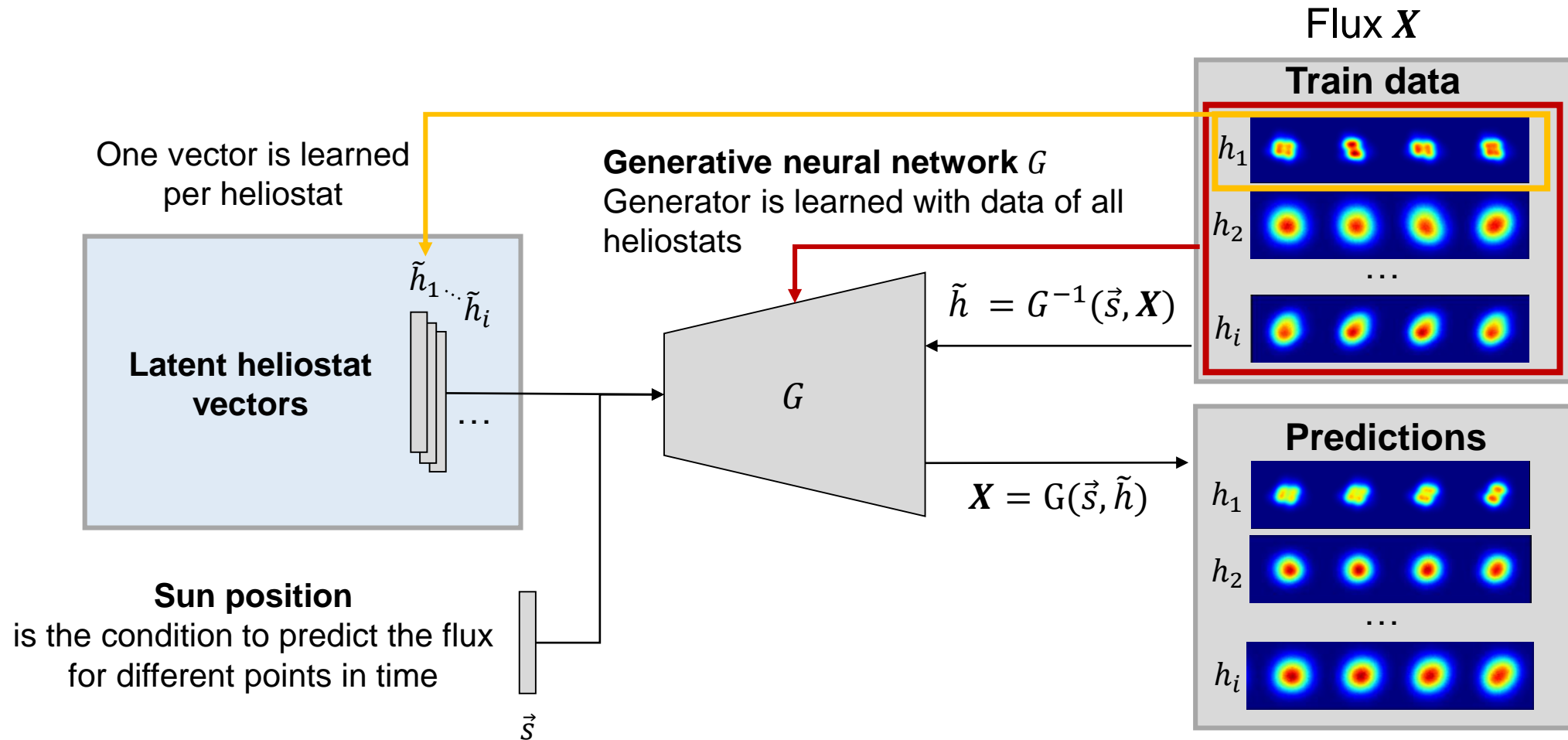


Physical vs. data driven heliostat model

- Instead of a physical model a unique abstract latent model for each heliostat is learned
- without the physical regularization, there is less information lost when mapping to the heliostat model
- The mapping function \mathbf{G} is also learned from the data
- Because both \mathbf{G} and the latent models are learned from data, we are able to find a more suitable representation for the flux space



Data driven heliostat model and generator



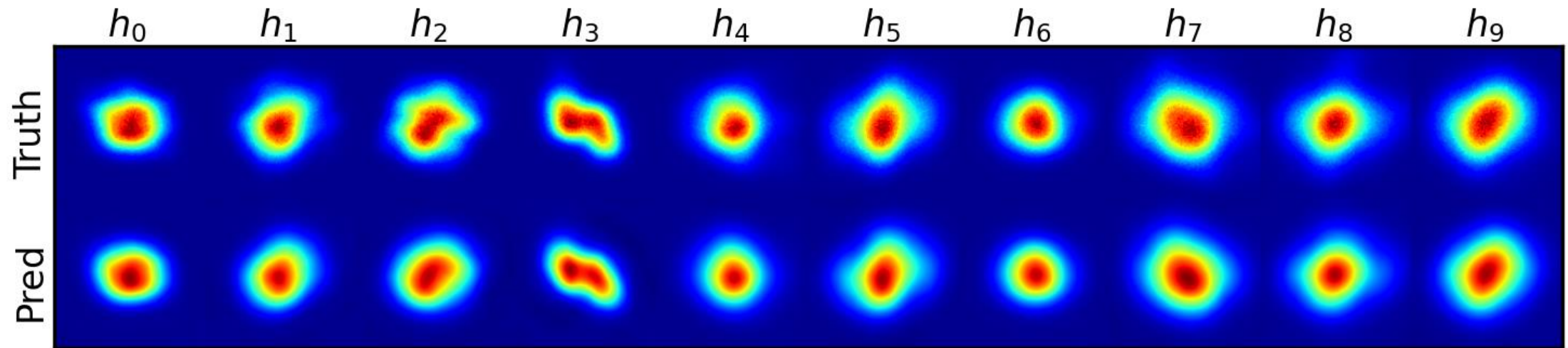
- Generator learns generalizing flux prediction for all heliostats
 - Heliostats profit from each others data
 - Heliostat specific features get extracted to latent vector

Results:

Predictions for new unseen sun positions for different heliostats

Model is able to

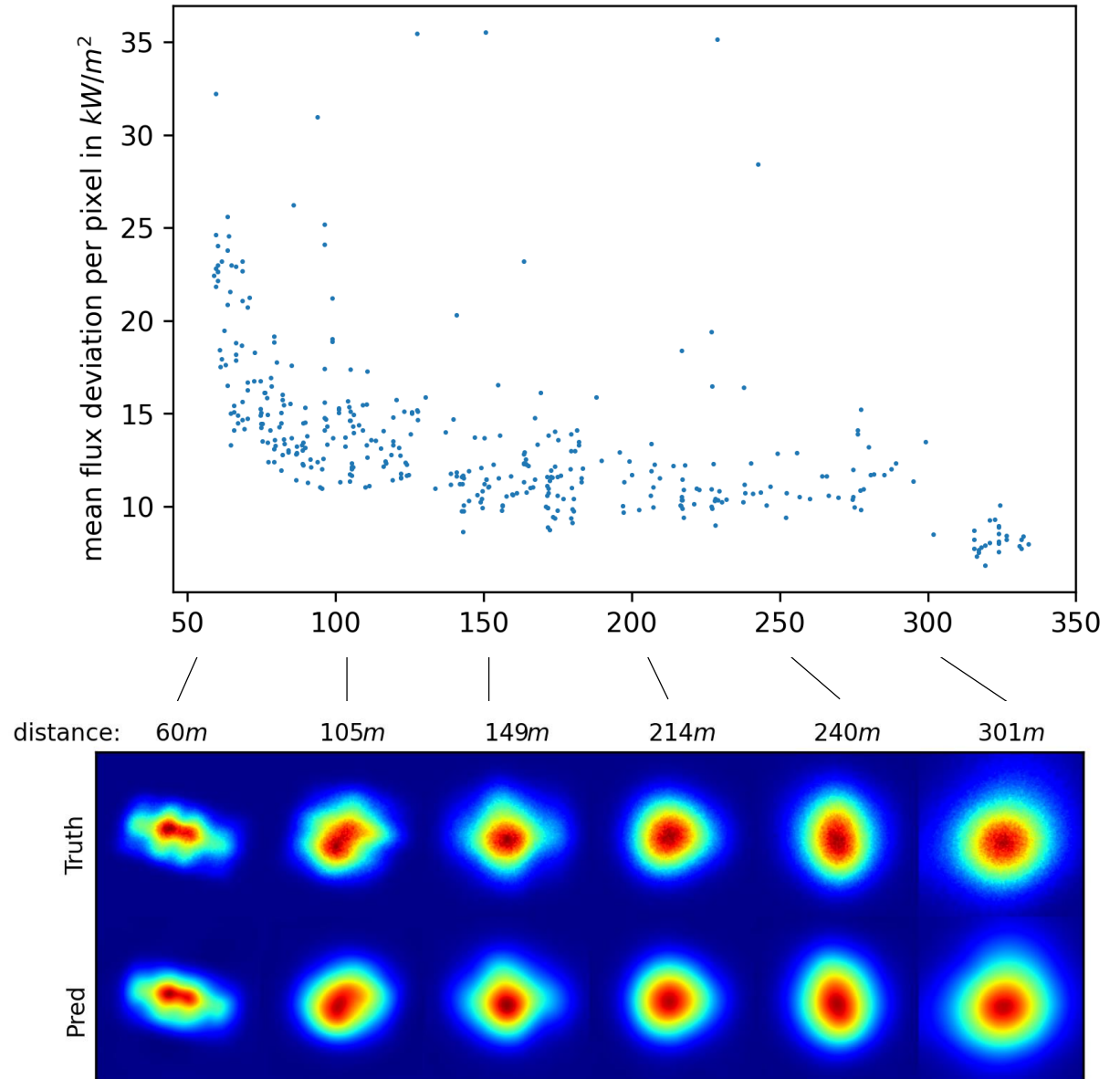
- extract heliostat specific features to latent vector
- predict flux for unseen future sun positions



Predicted flux by model (lower row) compared to groundtruth (upper row) for different heliostats h_i

Accuracy of flux predictions as a function of distance

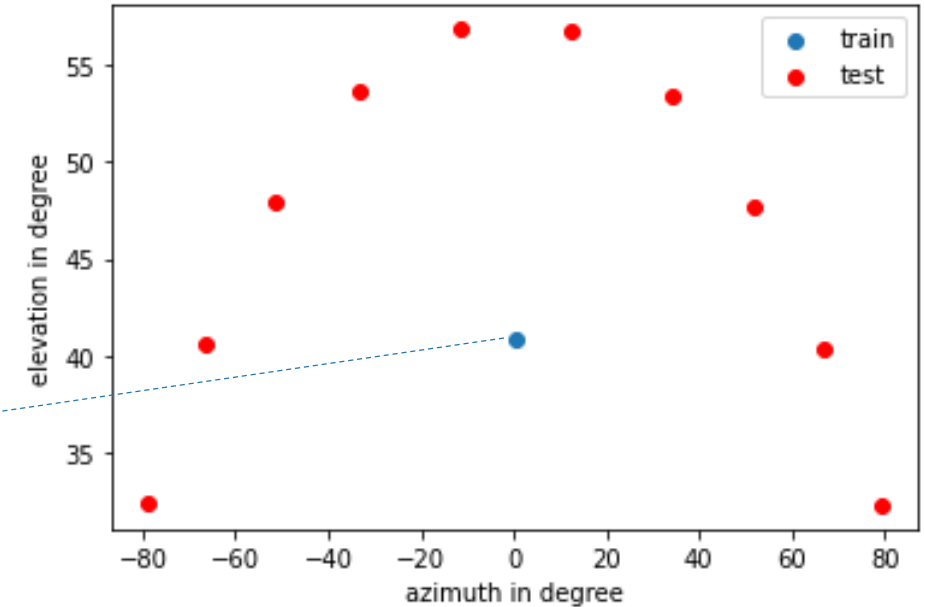
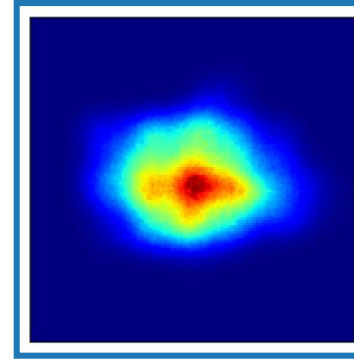
- Further distances lead to less information about the heliostat in the flux density
- the model is able to correctly predict the flux density for these heliostats
- Error ~5 %
- The error becomes smaller even for increasing distance
- which is due to the fact that those focal spots are easier to predict and better conditioned for pixelwise loss



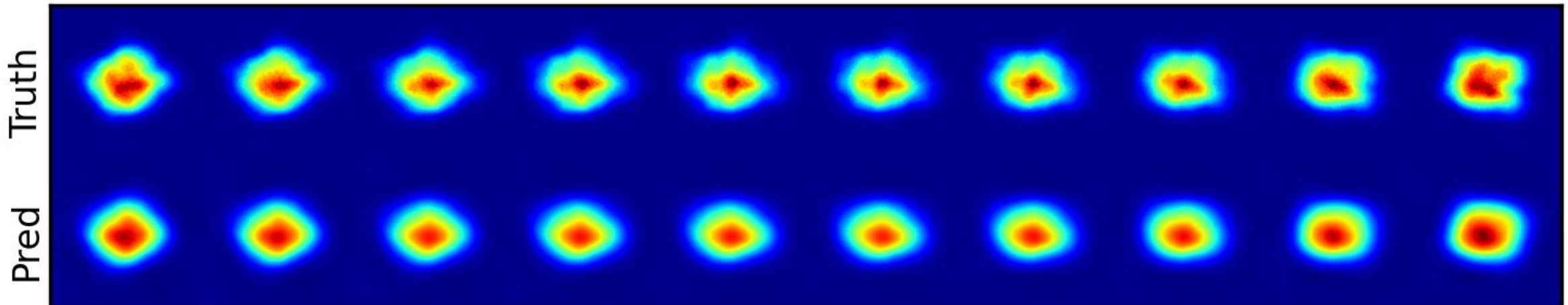
Predictions of heliostat vector trained with 1 picture

- Even from 1 training picture the model is able to predict other focal spots with different shapes
- Generator is able to transfer knowledge between heliostats

Training picture



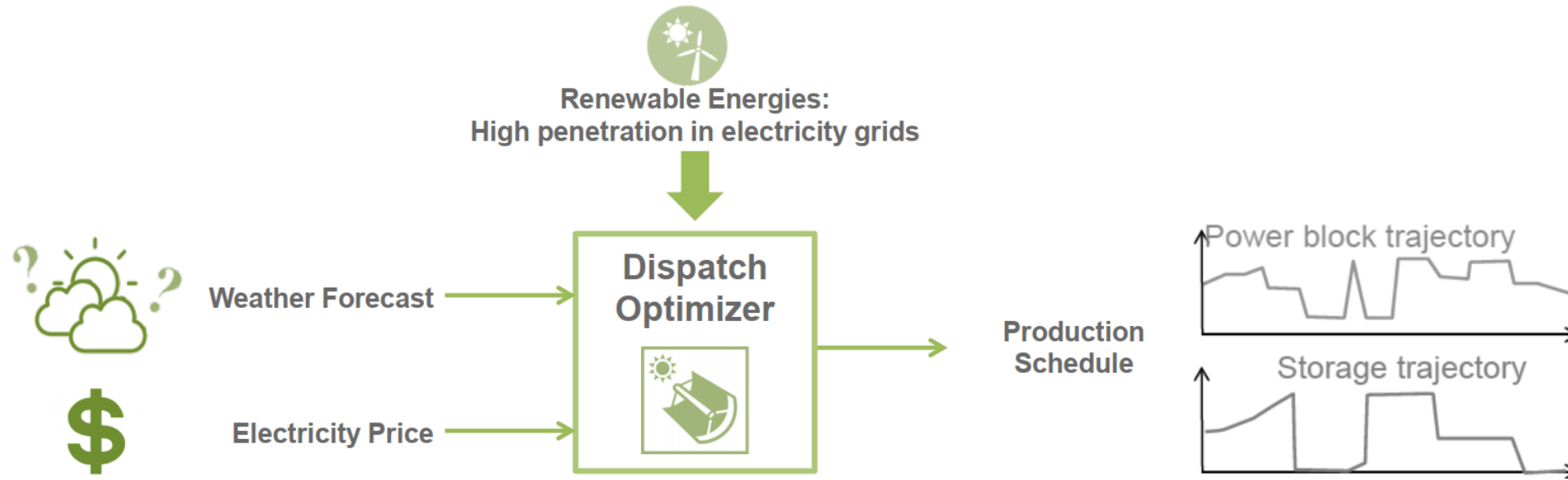
Predictions



The background of the slide is a photograph of a solar tower power plant. In the foreground, several large, rectangular heliostats (mirrors) are tilted at an angle, reflecting light. In the background, a tall, slender concrete tower stands against a clear blue sky. The sun is visible in the upper right corner, creating a bright lens flare. The ground is covered in green grass.

DISPATCH OPTIMIZATION OF POWER PRODUCTION CONSIDERING WEATHER FORECAST UNCERTAINTIES

How to operate a CSP power plant?

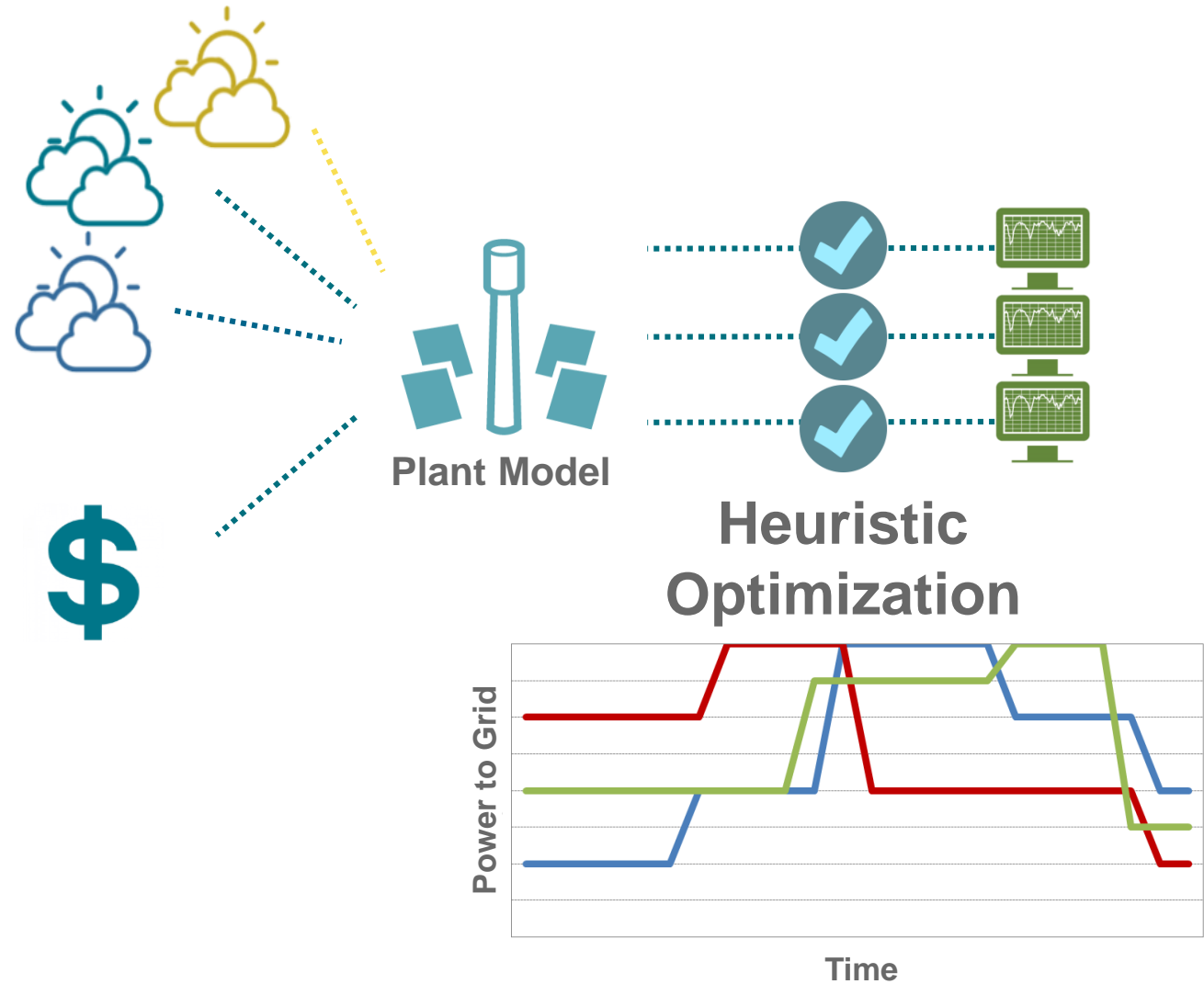


dispatch optimization algorithm that derives a plant operation schedule for the upcoming 48 hours

Goal : Include weather forecast uncertainty to find best schedule

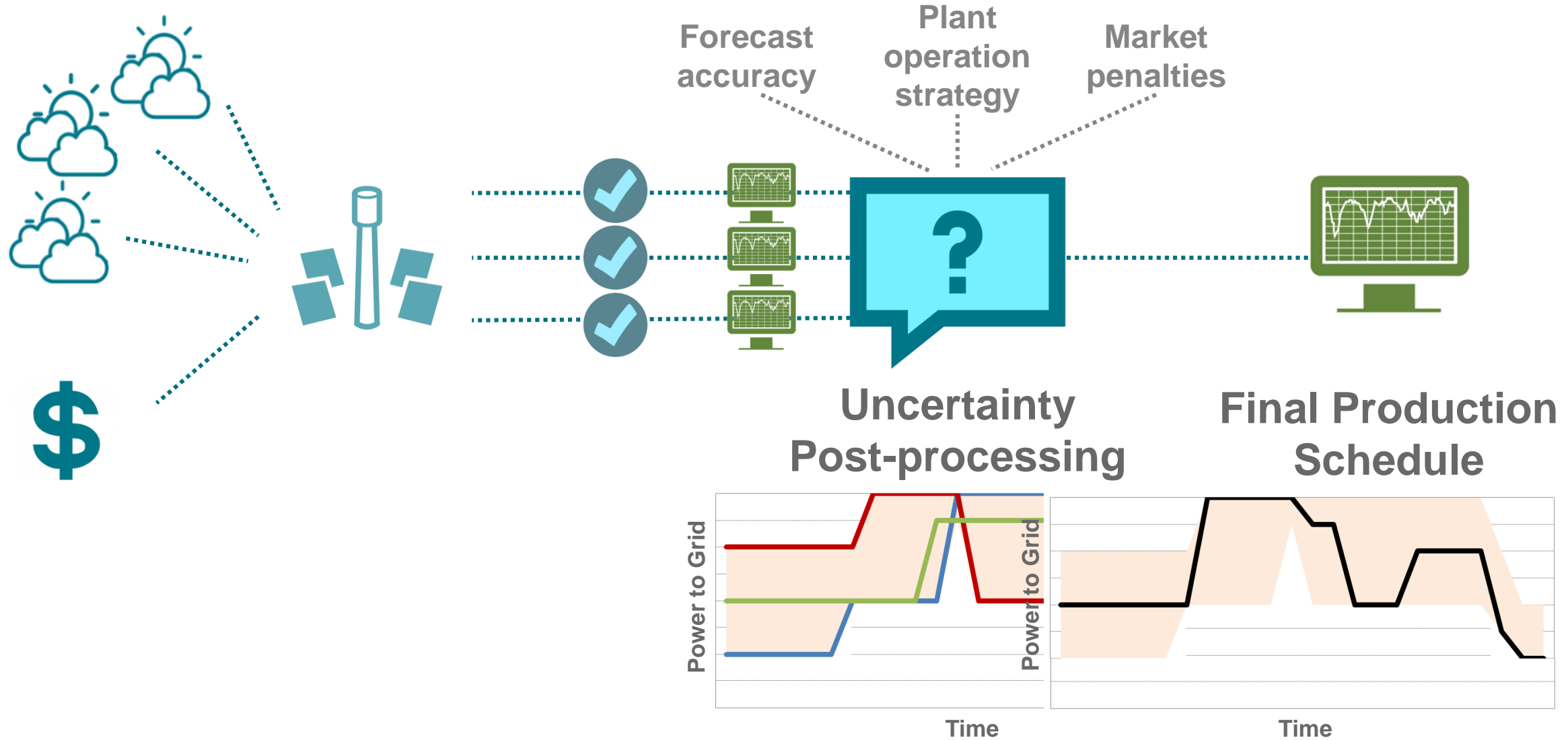
CSP Probabilistic Scheduling

The Dispatch Optimizer Tool



CSP Probabilistic Scheduling

The Dispatch Optimizer Tool



Machine learning uncertainty post-processing

Why?

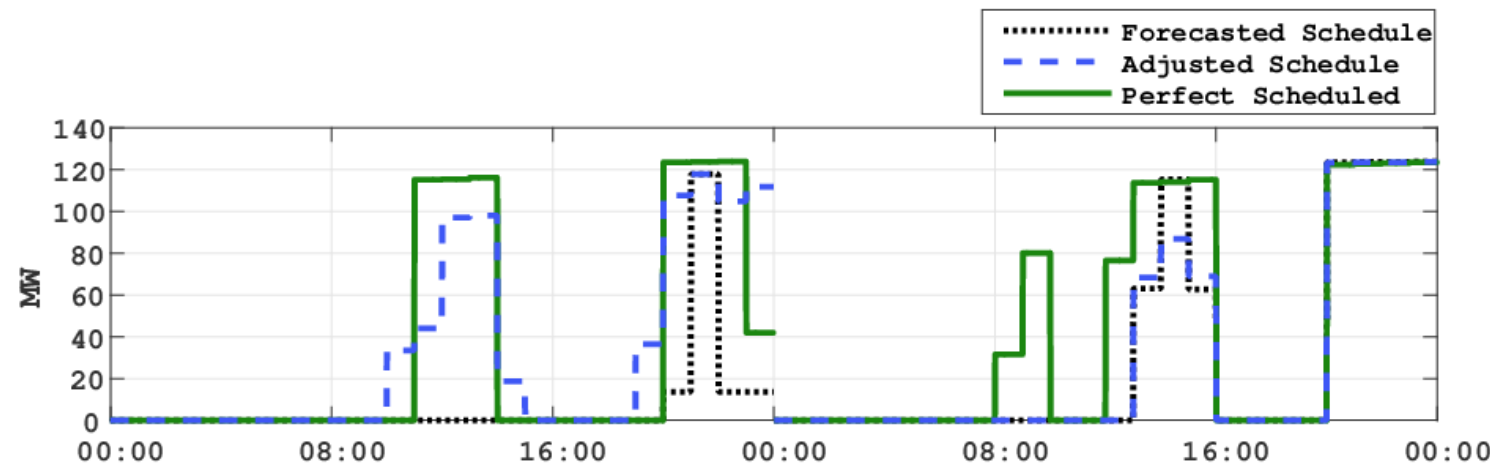
- Use of one or more possible schedule forecasts
- Use of historical data
- Use of several parameters related to the uncertainties

How?

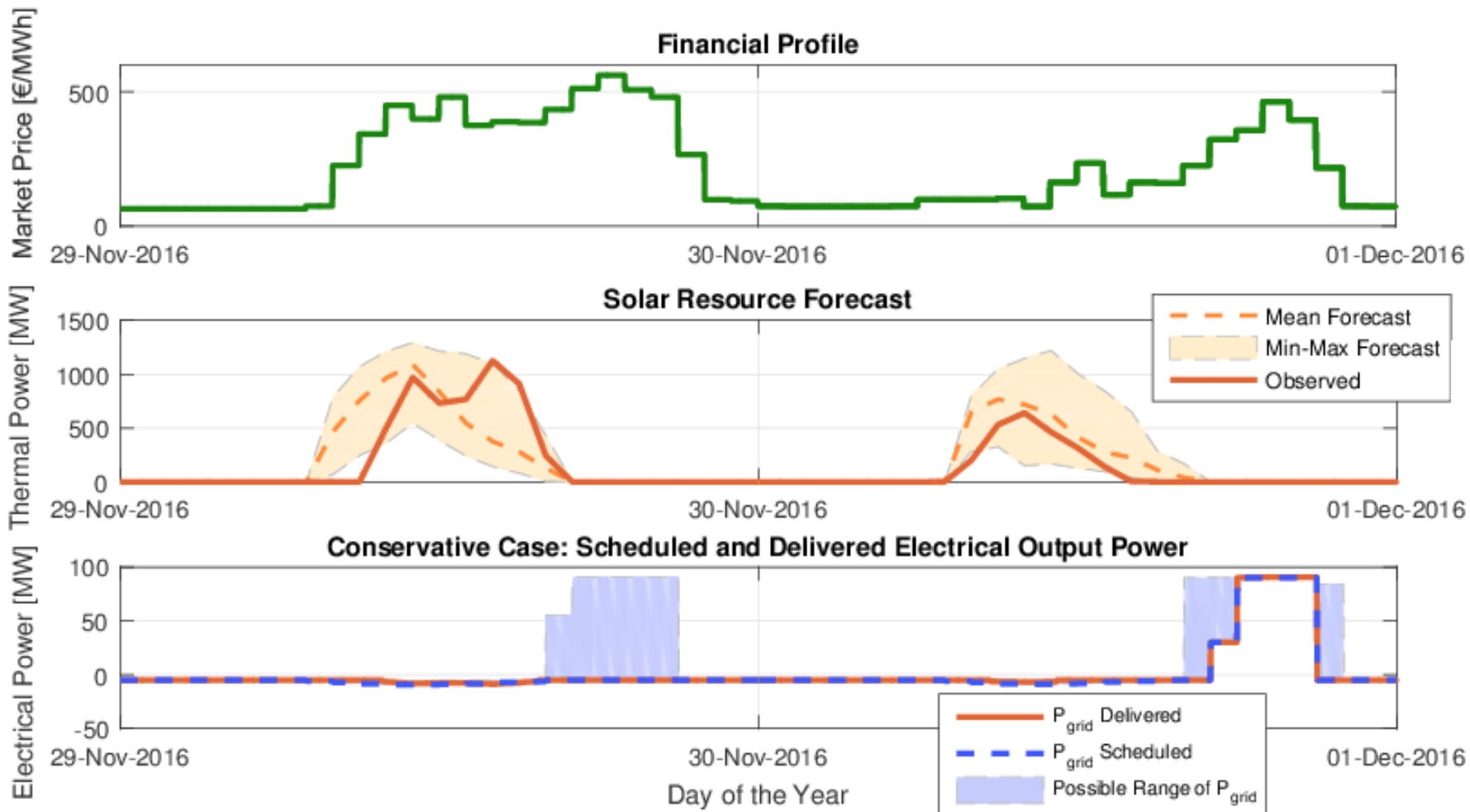
- Fuzzy Decision Tree approach

Results?

- Easy-to-understand
- Fast
- Improves the schedule quality?



CSP Probabilistic Scheduling Simulation Results



Price Input



Probabilistic Forecast



Output



Conservative Strategy \$

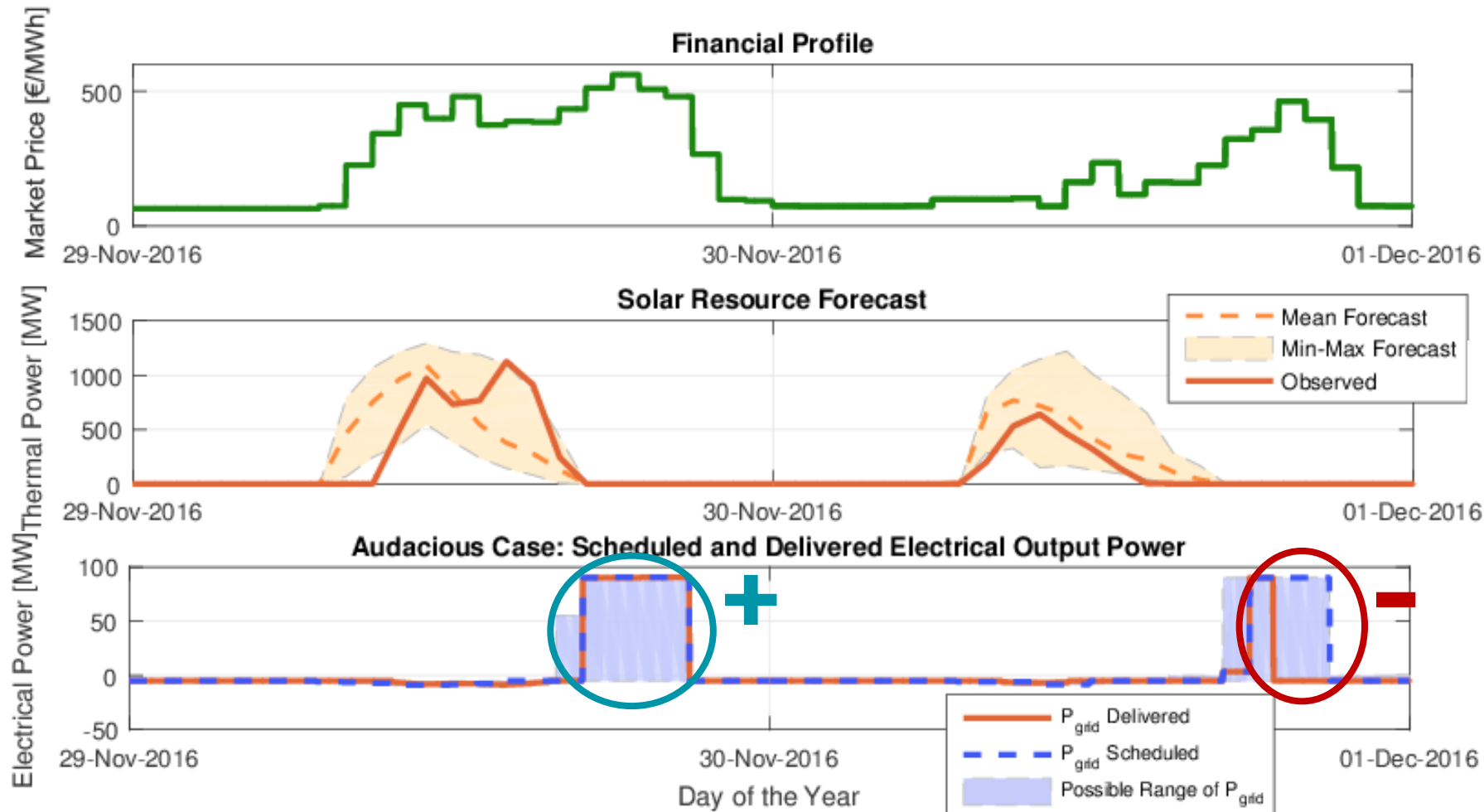
Change in Scheduling Strategy


Conservative




Audacious

CSP Probabilistic Scheduling Simulation Results



Price Input



Probabilistic Forecast



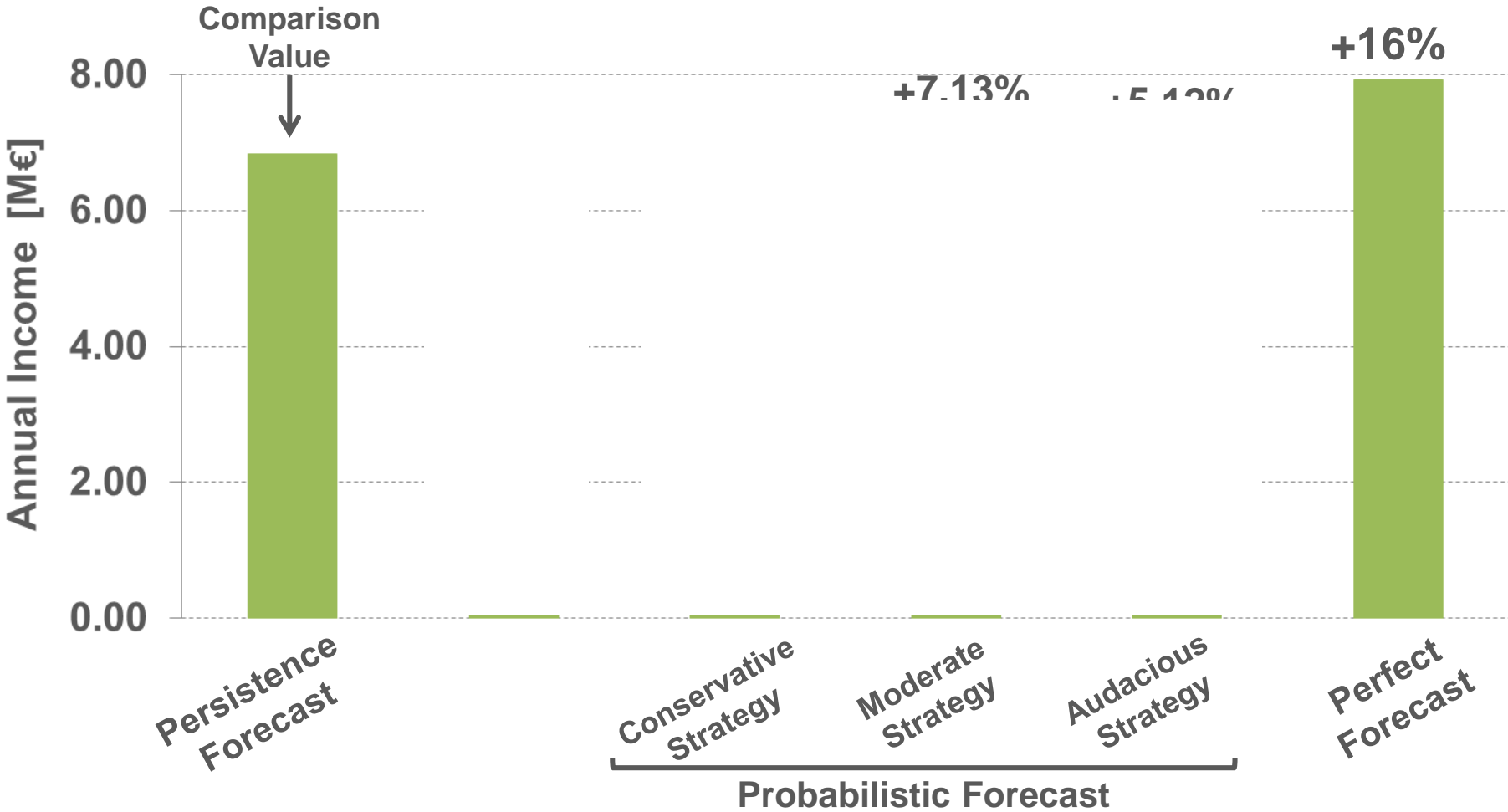
Output



Audacious Strategy

CSP Probabilistic Scheduling








Value of Forecast and Uncertainty Treatment



- CSP is a mature technology that complements PV electricity to enable reliable electricity supply around the clock.
- process heat supply @ $T > 100^{\circ}\text{C}$.
- Cost reduction until 2030
 - < 5 \$cents/kWh for dispatchable electricity
 - < 3 \$cents/kWh for 24/7 PV/CSP hybrids
 - < 1\$Cent/kWh for process heat
- Smart CSP is one key for cost reduction due to
 - performance increase
 - lifetime increase
 - maintenance cost reduction
- DLR is operating a full scale CSP power plant in Germany to develop, implement and prove the smart technology approaches in real scale to transfer it to the market.

Spin-off History



Spin-off	Business Case	Impact
2007 	Optimizing Solar Field Performance and maximizing plant lifetime	500 successful projects, resulting in a contribution to more than 90% of all installed CSP solar fields worldwide
2011 	Solar Water Treatment Plants	Demonstrated successfully at DLR Lampoldshausen, project for ESA in Kourou under preparation
2016 	Software, network and hardware: »Custom-designed mechatronic systems for sustainable technologies«	New heliostat design and control
2017 	Commercialization of Centrec® particle receiver technology developed by DLR	24/7 Low cost electricity and process heat based on high temperature technology
2019 	We automate building analyses through a measurement that takes only 2seconds per room	Energy efficient refurbishment of buildings worldwide
2020 	Fully automated condition monitoring of CSP and PV plants - Get actionable analytics for your solar field based on a digital twin	Improve performance and increase profitability of solar power plants
2022 	Digital materials research and development for fast & sustainable materials development	Speed-up the search for new materials, especially for chemical industry