

solar panel defects using drone imagery

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## WHO WE ARE

#### **CERTH:**

Founded in 2000 - one of the leading R&D centres in Greece

#### Includes 5 institutes:

- Information Technologies Institute (ITI)
- Chemical Process & Energy Resources Institute (CPERI)
- Hellenic Institute of Transport (HIT)
- Institute of Applied Bioscience (INAB)
- Institute of Bio-Economy and Agri-Technology (IBO)

## Information Technologies Institute:

- Part of CERTH since 2000
- Leading Institution of Greece in the fields of Informatics, Telematics and Telecommunications, etc.
- A total budget of 135 M€
- ~15 M€ funding per year (last 3 years)



## **PVGNOSIS**



**PVgnosis** "DiaGNOSIS, maintenance and operation of PV plants" is a **SOLAR-ERA.NET** Cofund 2 project implemented by **CERTH/ITI**, University of Cyprus, ENGAIA Renewable Energy Systems S.A. and Checkwatt AB

**PVgnosis** aims to create an ICT Platform integrating all the necessary tools for delivering advanced diagnosis, predictive maintenance and intelligent visual inspection on installed PV plants.



### UAVs AND DEEP LEARNING

Thermal images of solar panels collected from UAVs can provide valuable insights about panel condition / faults. This process, when supported by deep learning techniques, can significantly reduce maintenance cost and prevent energy / turnover loss.

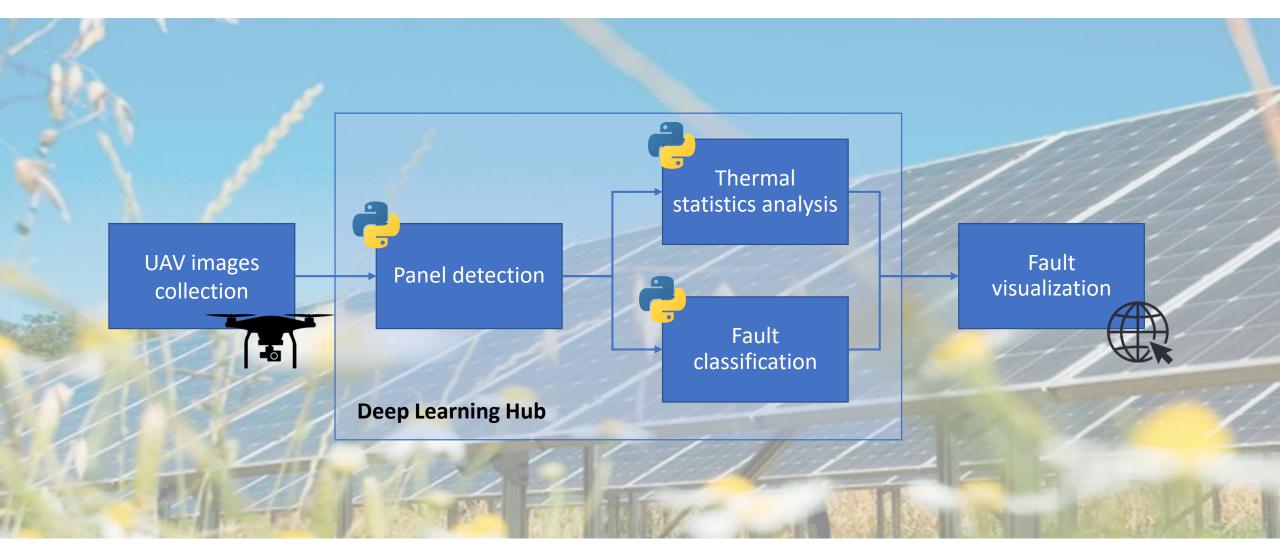






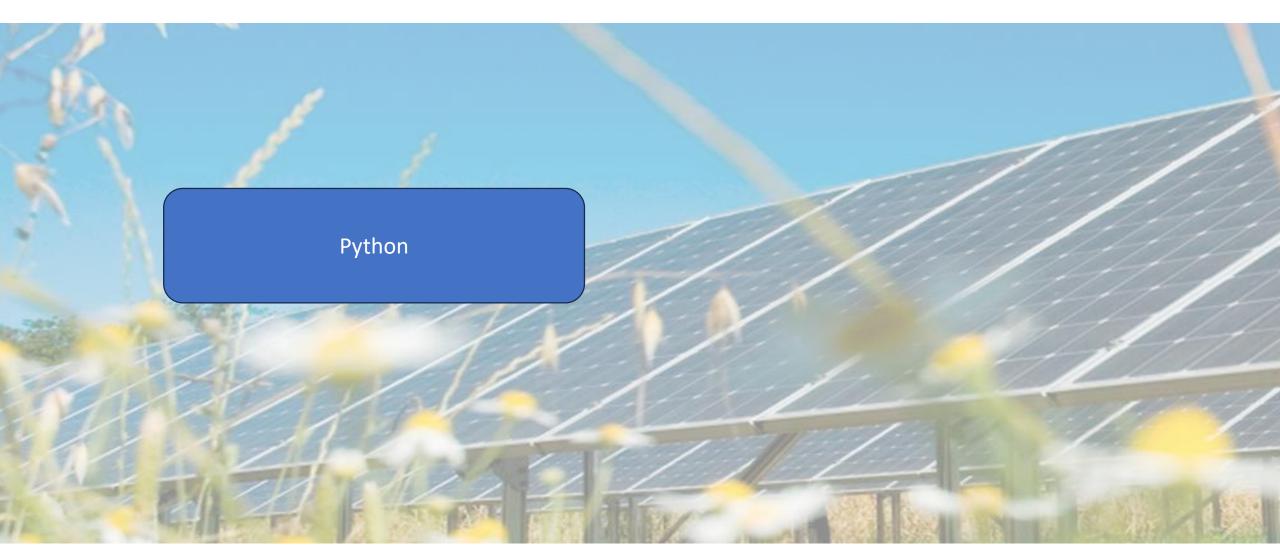


## **WORKFLOW PIPELINE**





## Convolutional Neural Networks (CNN)

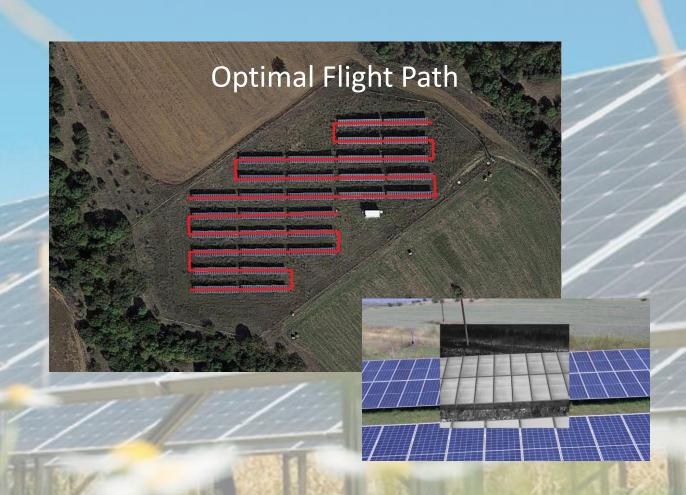




## **UAV IMAGES COLLECTION**

- Image collection is performed using drones equipped with optical and thermal cameras
- Images should be acquired under specific flight conditions (weather, height, speed, GSD, flight path)
- Currently supported images from AUTEL and DJI drones







## PANEL DETECTION

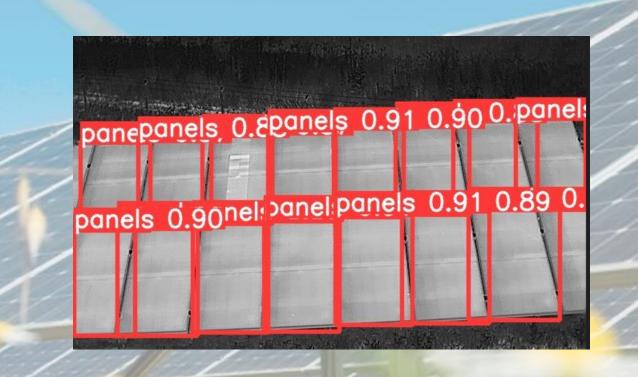
#### STEP 1: Initial Panel Detector

#### Goal:

 Rough detection of panels (Bounding box)

#### **Implementation:**

- CNN model based on Yolo architecture
- Training on three hundred images from Greek solar parks
- Insensitive in panel rotations
- Very High accuracy >90%





## PANEL DETECTION

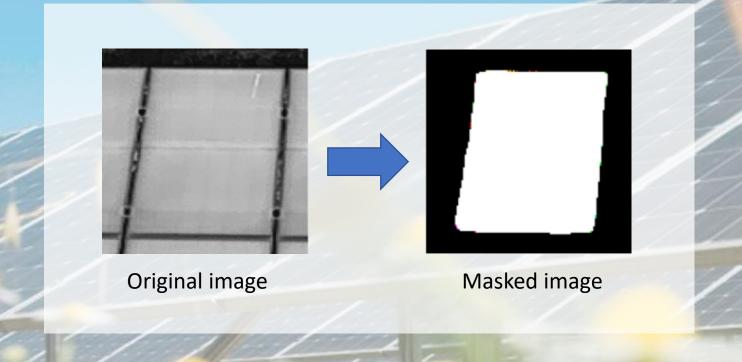
#### **STEP 2: Semantic Segmentation**

#### Goal:

- Isolate the region of interest
- Mask creation for every panel image

#### **Implementation:**

- CNN model based on Unet architecture
- Accuracy > 80%



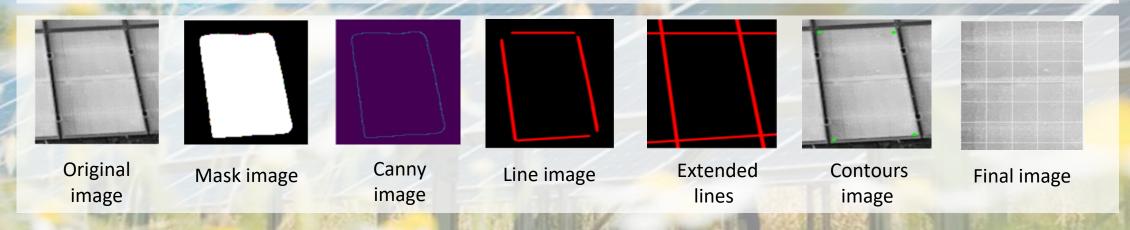


## PANEL DETECTION

#### STEP 3: Computer vision techniques

#### Goal:

- Create thermal panel image for CNN classifier
   Implementation:
- Using computer vision techniques from the OpenCV's python library (canny edge detection, houghLines, findContours, warpPerspective)





## PANEL CLASSIFICATION

Cell

# Training Dataset Trained classifier with limited online datasets of solar panel faults resulted in low accuracy with real data • Creation of a synthetic dataset with 4 fault categories (cell, diode, multi-cell, multi-diode)

Diode

Multi-Cell

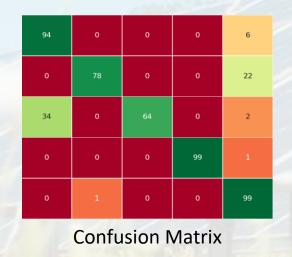


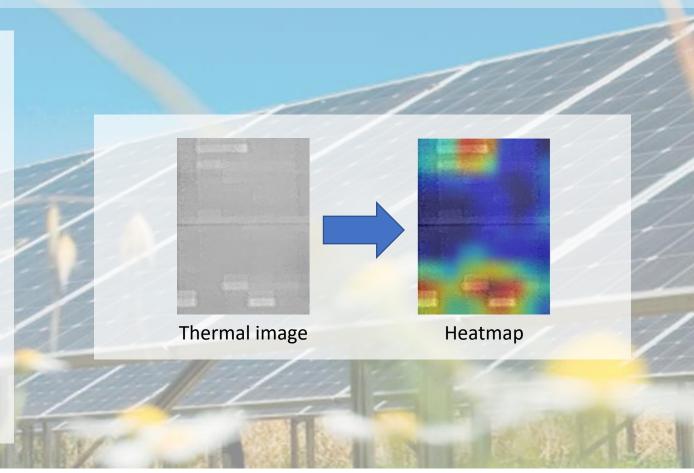
Multi-Diode

## PANEL CLASSIFICATION

#### **CNN Classifier**

- CNN based on EfficientNet
- Train on 7000 images for 25 epochs with batch size of 16
- Accuracy 89%





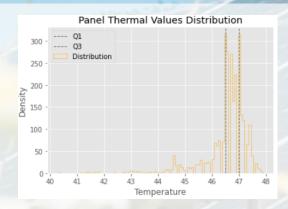


## THERMAL STATISTICS

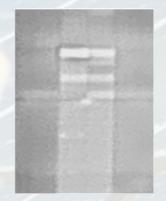
- Calculate thermal statistics for every panel (maximum value, minimum, mean, median, standard deviation, kurtosis, Skewness)
- Confirmation of anomaly (from image classification)
- Find problems that classifier is unable to detect (offline panels)
- Convolutional Neural Networks (CNN)



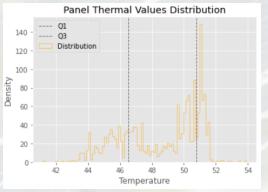
Panel without faults



No fault panel Histogram Standard deviation = 0.82



Multi-Diode anomaly panel



Multi-Diode panel Histogram
Standard deviation = 2.42



## More Information



