

**REPORT**  
on  
“Metadata extraction & Hotspot Detection”

by

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## GITHUB REPOSITORY

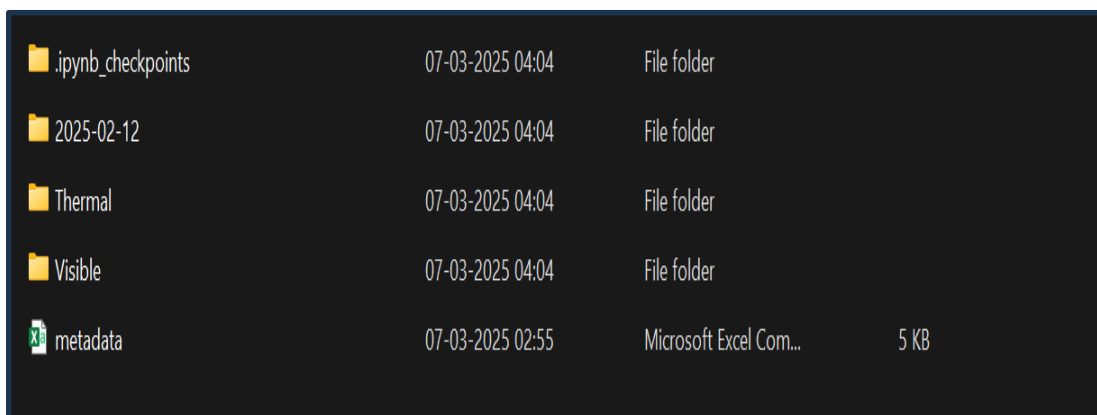
[https://github.com/YathinGK/Hornbill\\_Transformer.git](https://github.com/YathinGK/Hornbill_Transformer.git)

## Solution

### Systematic Image Organization

To ensure efficient access and retrieval, images are systematically organized into:

- Classified Images/ → Categorized into:
  - Thermal/ (Infrared images for hotspot detection)
  - Visible/ (Standard RGB images for reference)
- Date\_Wise\_Images/ → Groups images by capture date for easy tracking.
- Geolocation-Based Naming → Images are renamed using latitude and longitude for quick identification.
- Metadata Storage (metadata.csv) → A structured file storing extracted metadata for reference.

A screenshot of a file explorer window with a dark background. It displays a list of files and folders. The first four items are folders: '.ipynb\_checkpoints', '2025-02-12', 'Thermal', and 'Visible'. The last item is a file named 'metadata' with an Excel icon. Each item shows its name, a timestamp, and its type or size.

.ipynb_checkpoints	07-03-2025 04:04	File folder	
2025-02-12	07-03-2025 04:04	File folder	
Thermal	07-03-2025 04:04	File folder	
Visible	07-03-2025 04:04	File folder	
metadata	07-03-2025 02:55	Microsoft Excel Com...	5 KB

Fig 1 Output File Structure

### Metadata Extraction Process

Metadata is extracted using **PIL (Pillow)** and **ExifTags**, retrieving:

- **Date & Time** → Extracted from EXIF data.
- **GPS Coordinates** → Stored for location-based organization.
- **Filenames** → Renamed using metadata for consistency.
- **Image Category** → Classified as **Thermal or Visible** using AI.
- **Storage Paths** → Tracks image locations in structured directories.

Extracted metadata is stored in **metadata.csv** for easy tracking. **GPS coordinates** are converted to **decimal format** for accuracy, ensuring precise geolocation tagging.

	A	B	C	D	E	F	G	H	I
1	FileName	Latitude	Longitude	Date	Time	NewFileName			
2	DJI_20250	17.9735	76.4512	12-02-2025	11:41:21	2025-02-12_17.9735_76.4512.jpg			
3	DJI_20250	17.9735	76.4512	12-02-2025	11:41:21	2025-02-12_17.9735_76.4512.jpg			
4	DJI_20250	17.9734	76.4514	12-02-2025	11:41:38	2025-02-12_17.9734_76.4514.jpg			
5	DJI_20250	17.9734	76.4514	12-02-2025	11:41:38	2025-02-12_17.9734_76.4514.jpg			
6	DJI_20250	17.9735	76.4514	12-02-2025	11:41:51	2025-02-12_17.9735_76.4514.jpg			
7	DJI_20250	17.9735	76.4514	12-02-2025	11:41:51	2025-02-12_17.9735_76.4514.jpg			
8	DJI_20250	17.9735	76.4514	12-02-2025	11:41:56	2025-02-12_17.9735_76.4514.jpg			
9	DJI_20250	17.9735	76.4514	12-02-2025	11:41:56	2025-02-12_17.9735_76.4514.jpg			
10	DJI_20250	17.9735	76.4514	12-02-2025	11:42:03	2025-02-12_17.9735_76.4514.jpg			
11	DJI_20250	17.9735	76.4514	12-02-2025	11:42:03	2025-02-12_17.9735_76.4514.jpg			
12	DJI_20250	17.9735	76.4514	12-02-2025	11:42:08	2025-02-12_17.9735_76.4514.jpg			
13	DJI_20250	17.9735	76.4514	12-02-2025	11:42:08	2025-02-12_17.9735_76.4514.jpg			
14	DJI_20250	17.9735	76.4514	12-02-2025	11:42:19	2025-02-12_17.9735_76.4514.jpg			
15	DJI_20250	17.9735	76.4514	12-02-2025	11:42:19	2025-02-12_17.9735_76.4514.jpg			
16	DJI_20250	17.9735	76.4514	12-02-2025	11:42:32	2025-02-12_17.9735_76.4514.jpg			
17	DJI_20250	17.9735	76.4514	12-02-2025	11:42:32	2025-02-12_17.9735_76.4514.jpg			
18	DJI_20250	17.9735	76.4514	12-02-2025	11:42:41	2025-02-12_17.9735_76.4514.jpg			
19	DJI_20250	17.9735	76.4514	12-02-2025	11:42:41	2025-02-12_17.9735_76.4514.jpg			
20	DJI_20250	17.9735	76.4514	12-02-2025	11:42:44	2025-02-12_17.9735_76.4514.jpg			
21	DJI_20250	17.9735	76.4514	12-02-2025	11:42:44	2025-02-12_17.9735_76.4514.jpg			
22	DJI_20250	17.9735	76.4514	12-02-2025	11:42:55	2025-02-12_17.9735_76.4514.jpg			
23	DJI_20250	17.9735	76.4514	12-02-2025	11:42:55	2025-02-12_17.9735_76.4514.jpg			
24	DJI_20250	17.9735	76.4514	12-02-2025	11:43:07	2025-02-12_17.9735_76.4514.jpg			

Fig 2 Extracted Metadata in .CSV

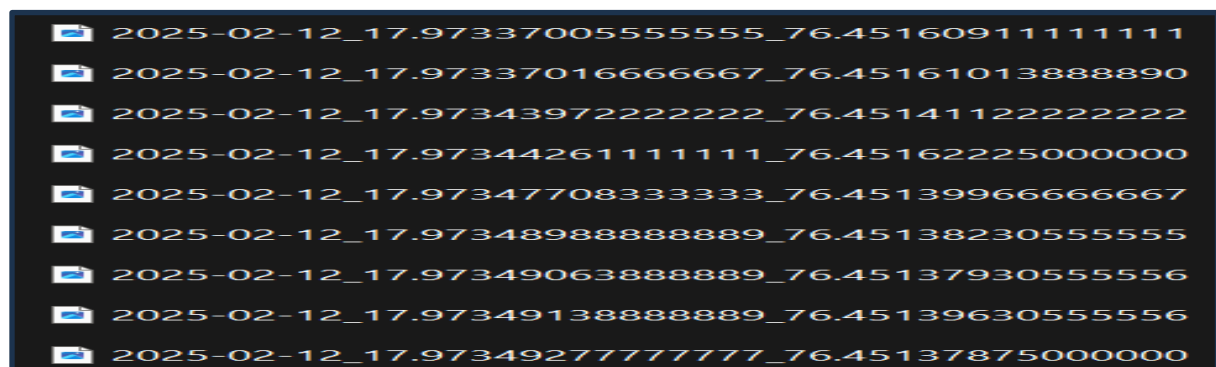


Fig 3 Image renames based on Date and GPS

## AI Model for Image Classification

### 1. Model Architecture

A CNN (Convolutional Neural Network) was trained using TensorFlow/Keras to classify images as Thermal or Visible. The architecture includes:

- Convolutional Layers → Extracts key image features.
- Pooling Layers → Reduces complexity while preserving information.

### 2. Training and Performance

- Training → 10 epochs using the Adam optimizer.
- Accuracy → Over 95% on validation data.
- Deployment → Classifies newly uploaded images, automatically sorting them into Thermal or Visible folders.

## Hotspot Detection in Thermal Images

### Hotspot Detection Algorithm

To detect overheating or faults in transmission towers, a hotspot detection algorithm was implemented using OpenCV:

- Grayscale Conversion → Enhances heat variations.
- Gaussian Blur → Reduces noise.
- Thresholding → Isolates bright (hot) regions.
- Contour Detection → Identifies hotspots and marking

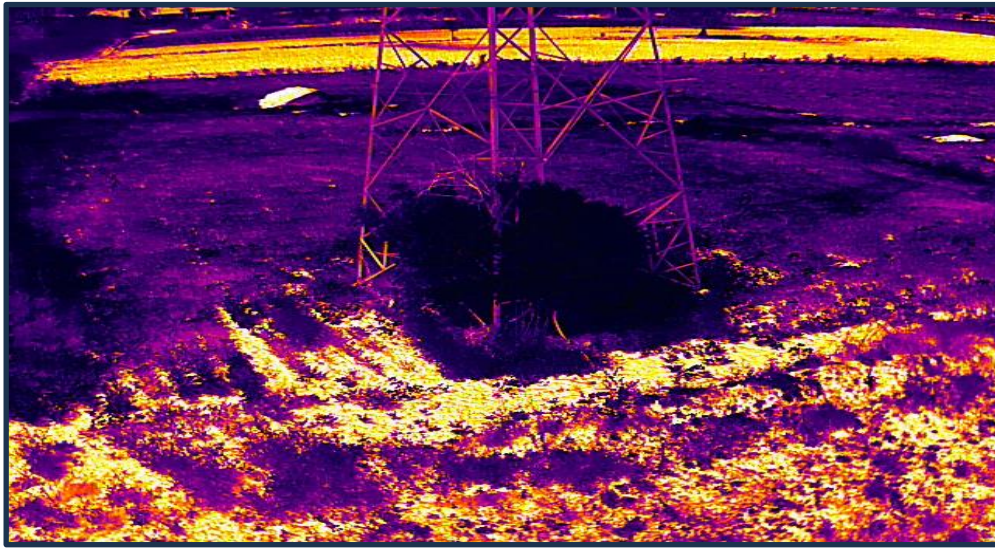


Fig 5 Image Before Hotspot Algorithm was applied

## OUTPUT

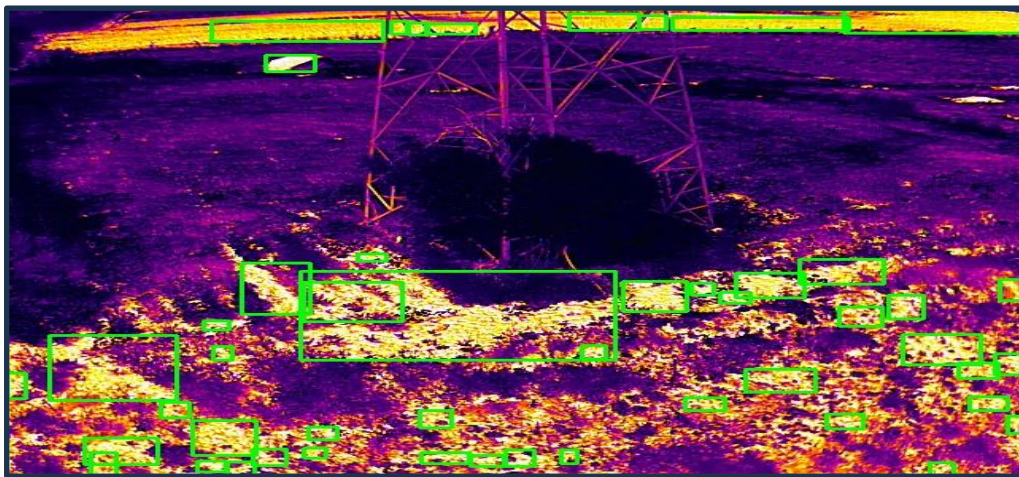


Fig 6 Hotspot Detection Using Algorithm

## Optimizing the Workflow

The workflow was fully automated, handling metadata extraction, image classification, organization, and hotspot detection without manual intervention. The script ensures seamless processing, accurately storing images in structured directories while minimizing errors. This automation significantly enhances efficiency, making it ideal for large-scale datasets.

To improve accuracy and scalability, the AI model was trained with optimized hyperparameters, and image preprocessing techniques like Gaussian blur, adaptive thresholding, and morphological operations were applied for better hotspot detection. Future enhancements include cloud integration (AWS, Google Cloud), real-time API processing, parallel computing, and GPU acceleration to further boost speed and scalability for industrial applications.