

# Optimizing Optical Satellite Data Analysis: Methods and Strategies

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## Agenda

□ Introduction and Overview

□ Importance of Optimizing Satellite Data Analysis

Data Challenges in Optical Satellite Imagery

Workflow for Optical Data Analysis

Preprocessing Techniques

Advanced Feature Extraction Techniques

□ MBRSC Products: Use Cases

□ Future Trends in Satellite Data Analysis





## Introduction and Overview

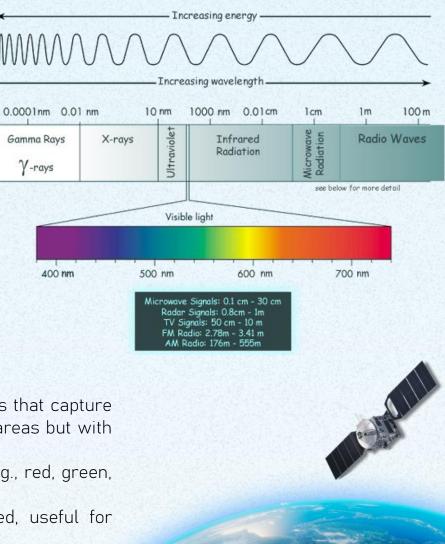
Optical satellite images are taken by sensors that capture visible light, as well as parts of the ultraviolet and infrared spectrums, by detecting sunlight reflected from Earth's surface.

Optical satellite sensors capture sunlight reflected in visible and NIR wavelengths, allowing them to distinguish between materials like vegetation, water, soil, and urban structures. This imaging resembles human vision but extends to wavelengths beyond what we can see.

Optical satellite images vary in resolution:

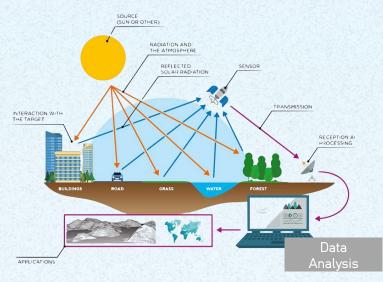
- Spatial Resolution: The detail of the image, ranging from high-resolution images that capture details as small as a few centimeters to lower resolutions that capture larger areas but with less detail.
- Spectral Resolution: The ability to capture data in specific wavelength bands (e.g., red, green, blue, and infrared).
- Temporal Resolution: How frequently images of the same area are captured, useful for monitoring changes over time.







# Importance of Optimizing Satellite Data Analysis



• Enhanced Accuracy: Improved algorithms lead to better accuracy in interpreting land cover, vegetation health, water quality, etc.

• Faster Processing: Optimization reduces processing times, allowing near real-time data analysis, crucial for disaster response and time-sensitive applications.

• Cost Efficiency: Efficient data processing minimizes computing costs, especially for high-resolution or largescale datasets.

• Scalability: Optimized processes allow analysis of growing data volumes as satellite networks and data collection expand.



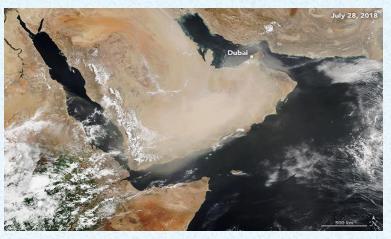


# Data Challenges in Optical Satellite Imagery (1/3)

## . Atmospheric Interference

- Cloud Cover: Clouds obscure large areas of imagery, limiting data collection, especially in tropical and high-latitude regions. Persistent cloud cover can hinder long-term monitoring efforts.
- Aerosols and Dust: Particles in the atmosphere, such as dust, smoke, and pollutants, can distort the captured images, affecting data accuracy, especially in urban and desert areas.

**Solution**: Atmospheric correction algorithms and multi-temporal imaging can help mitigate these effects, though complete elimination remains challenging.



## 2. Lighting Conditions



- Sun Angle Variability: Variations in sunlight angle due to time of day or season affect the brightness and contrast in images, making it difficult to maintain consistency across datasets.
- Shadows: Shadows cast by natural or man-made features can obscure underlying details, complicating image interpretation, especially in mountainous or densely built areas.

**Solution**: Standardizing image acquisition times and employing shadow removal techniques can improve data quality.





# Data Challenges in Optical Satellite Imagery (2/3)

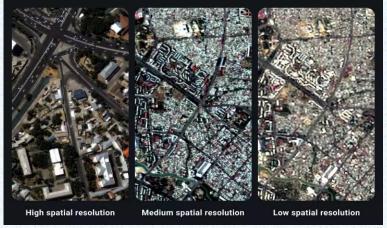
## 3. Limited Temporal Coverage

- Daylight Dependency: Optical sensors rely on sunlight, so they cannot capture data at night, limiting temporal coverage for certain types of monitoring.
- Sparse Revisits: Many satellites only capture images of the same location every few days or weeks, which can be insufficient for fast-changing events, such as wildfires or floods.

**Solution**: Using constellations of satellites or combining data from multiple sources can increase temporal resolution.



## 4. Spatial and Spectral Resolution Trade-offs



- Resolution Constraints: High spatial resolution (fine detail) often requires lower spectral or temporal resolution due to technical limitations, making it challenging to capture highly detailed images with comprehensive spectral data.
- Spectral Bands Limitations: Most optical sensors capture a limited range of spectral bands, often excluding thermal or additional near-infrared bands that could provide useful data.

**Solution**: Combining multiple datasets can help balance these tradeoffs for more robust analyses.





# Data Challenges in Optical Satellite Imagery (3/3)

## 5. Data Volume and Storage Requirements

- Large File Sizes: High-resolution optical images produce massive amounts of data, requiring significant storage, bandwidth, and processing power to manage effectively.
- Storage and Management Costs: Storing and maintaining these large datasets can become costly, especially for organizations handling extensive archives.

**Solution**: Cloud-based storage solutions and compression techniques can ease storage demands, while data curation practices ensure efficient data retrieval.

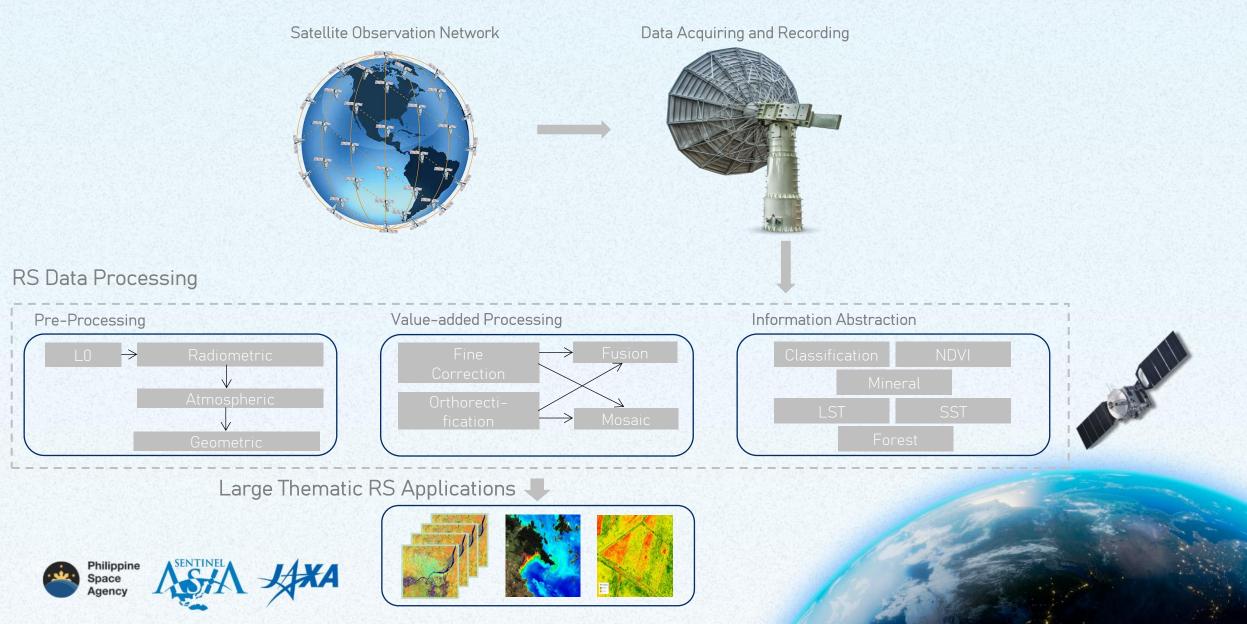
## 6. Variability in Data Quality

- Sensor Degradation: Over time, satellite sensors degrade, which can reduce image quality and lead to inconsistencies across datasets, particularly in older images.
- Calibration Issues: Differences in calibration between satellite missions and even individual sensors can introduce variations, complicating cross-sensor analysis.

**Solution**: Regular calibration, radiometric correction, and standardized processing methods help maintain data quality and consistency.



# Workflow for Optical Data Analysis



## **Preprocessing Techniques**

Pre-processing techniques for optical satellite images are essential for enhancing data quality and ensuring consistent analysis. Here's an ordered list of common pre-processing steps:

### **Radiometric Correction**

### **Atmospheric Correction**

Removes atmospheric effects

such as scattering and absorption

caused by aerosols, gases, and

Subtraction, Quick Atmospheric

Correction (QUAC), and radiative

transfer models (e.g., 6S model)

Dark

Object

## **Geometric Correction**

### Purpose:

Corrects geometric distortions caused by sensor angle, Earth's curvature, and terrain variations. aligning the image to a specific map projection.

### Techniques:

Involves ground control points (GCPs) and digital elevation models (DEMs) to adjust the image geometry.

### Outcome:

Creates a spatially accurate, map-aligned image, essential for accurate spatial analysis.

## (Optional) **Spatial Enhancement**

### Purpose:

Enhances spatial resolution or sharpness to improve feature visibility, particularly useful for high-resolution analysis.

## Techniques:

Includes sharpening filters. resampling, and pan-sharpening (fusing high-resolution panchromatic with lowerresolution multispectral images). Outcome:

Produces clearer images with enhanced details, though it may introduce artifacts if not done carefully.



### Purpose:

Adjusts for sensor-related errors and atmospheric conditions that affect pixel values.

### Techniques:

Sensor calibration

### Outcome:

Ensures pixel values accurately represent surface reflectance, improving data consistency.

## Outcome:

Purpose:

water vapor.

Techniques:

Methods like

are commonly used.

Reduces atmospheric distortions, providing a more accurate reflection of surface properties.

## **Preprocessing Techniques**

Pre-processing techniques for optical satellite images are essential for enhancing data quality and ensuring consistent analysis. Here's an ordered list of common pre-processing steps:

### **Noise Reduction**

### Cloud Masking

### Purpose:

Reduces sensor noise and other image artifacts, especially in images captured under low-light conditions or in specific spectral bands.

### Techniques:

Techniques like median filtering, Gaussian filtering, or other smoothing algorithms are applied.

### Outcome:

Creates a cleaner image by reducing random pixel variations that could interfere with analysis.

### Purpose:

Identifies and masks clouds and their shadows, as these can obscure the surface and affect data accuracy

### Techniques:

Uses algorithms like Fmask (Function of Mask), cloud probability assessments, or spectral thresholds in visible and infrared bands.

### Outcome:

Provides cloud-free images, crucial for applications like land cover analysis where surface visibility is essential.

## Image Resampling

### Purpose:

Adjusts pixel size to match other images in the dataset or to a desired resolution, aiding in data consistency and alignment.

## Techniques:

Common methods include nearest neighbor, bilinear, and cubic convolution resampling. Outcome:

Ensures uniform pixel sizes across images, useful for multitemporal analysis and data fusion.

### Data Normalization

### Purpose:

Ensures uniformity in pixel values across images taken at different times or by different sensors, facilitating comparison.

### Techniques:

Normalization methods include relative radiometric normalization, histogram matching, and transforming data to reflectance values. Outcome:

Produces consistent datasets, ideal for change detection and multi-temporal analysis.







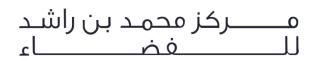
## Advanced Feature Extraction Techniques

- 1. Object-Based Image Analysis (OBIA)
- 2. Machine Learning Classification
- 3. Deep Learning and Convolutional Neural Networks (CNNs)
- 4. Spectral Unmixing
- 5. Change Detection Analysis
- 6. Texture Analysis
- 7. Data Fusion Techniques









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## Urban Planning & Development













Land Use/ Land Cover Mapping

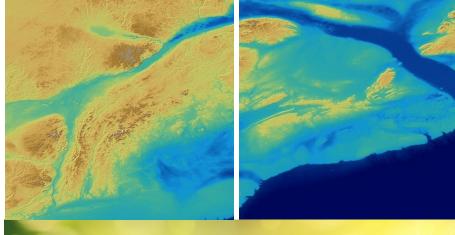








## **Environmental Studies**





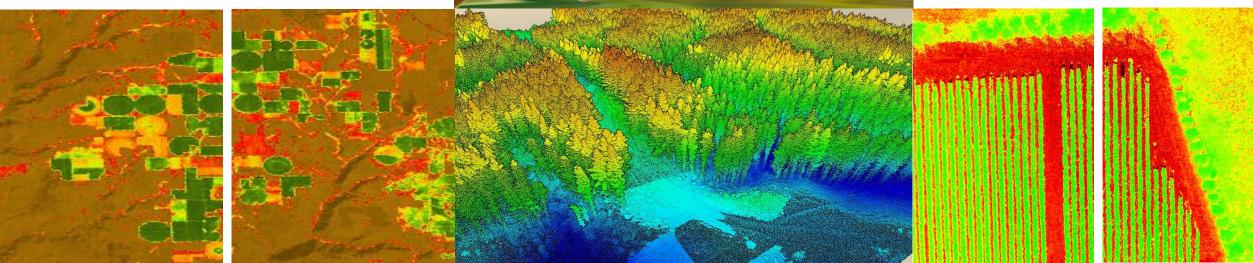


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Estimate Water Level in Dams: Numerical Analysis

May 2015 – May 2016

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Sep 2016 – Nov 2016

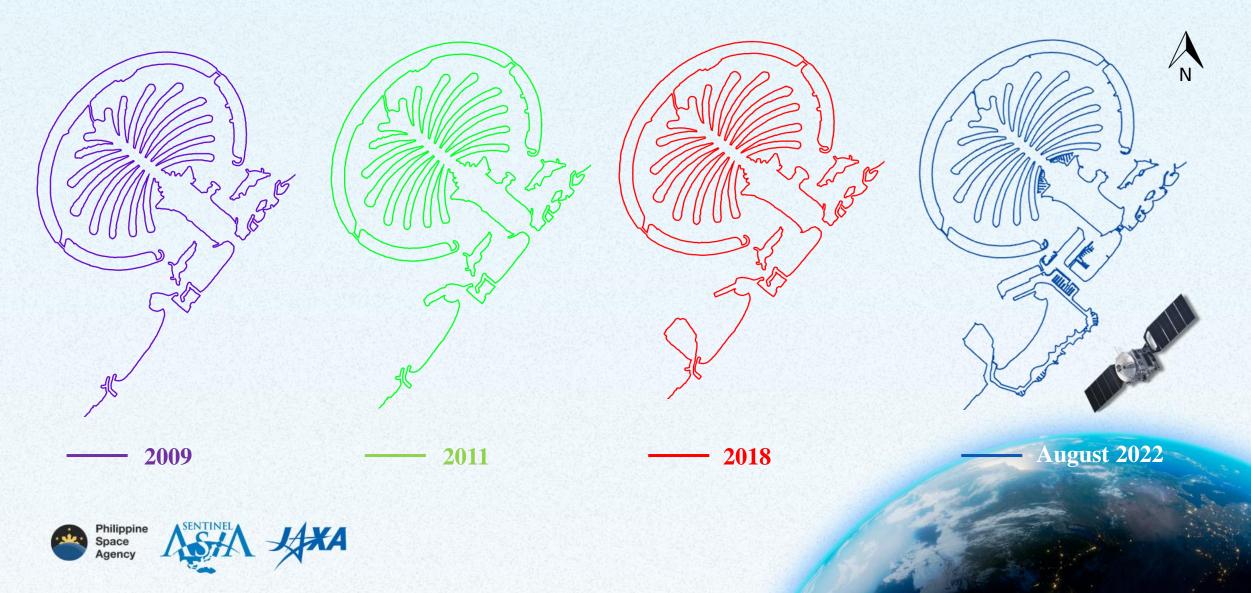
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Philippine Space Agency May 2016 - Sep 2016

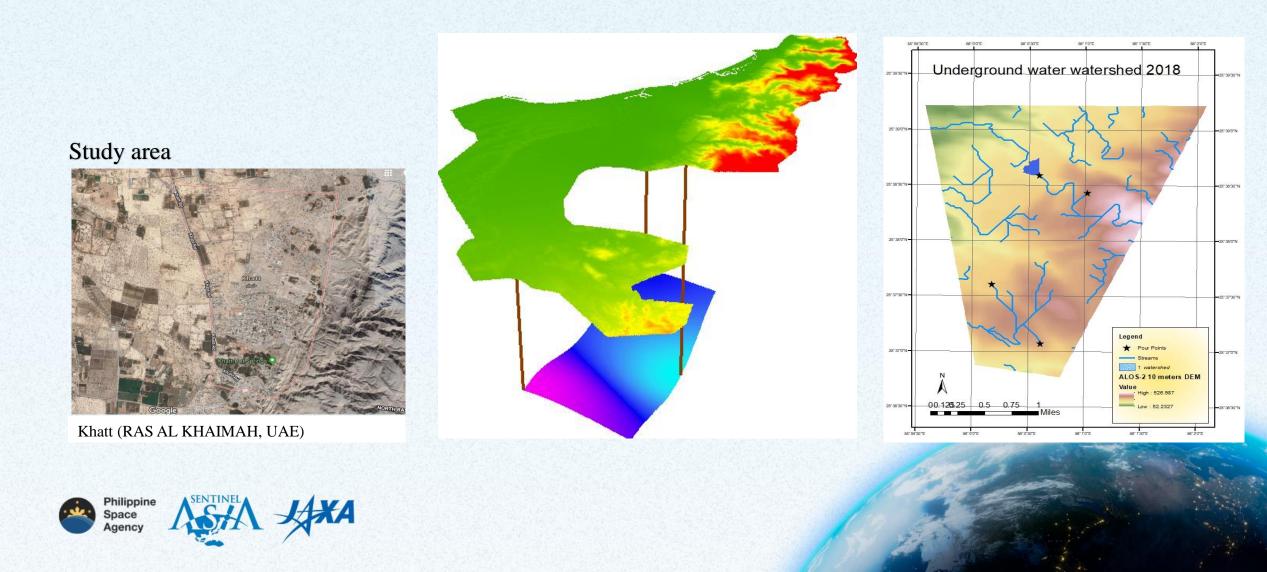
Nov 2016 – Jan 2017

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Coastline Monitoring – Dubai shoreline: Case Study



Underground Water Detection using Data Fusion



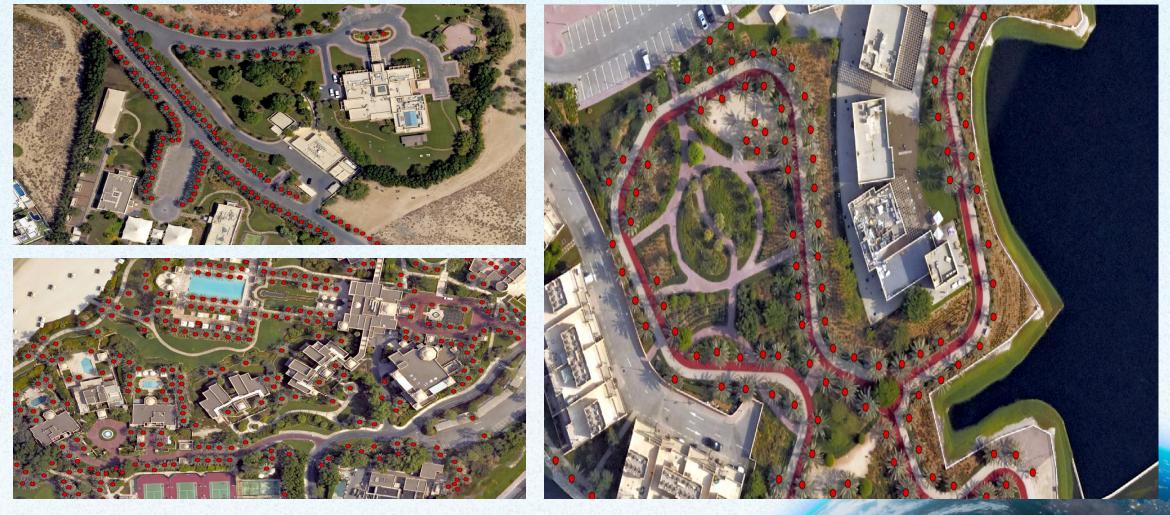
Vegetation Health Monitoring







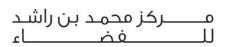
Palm Trees Detection











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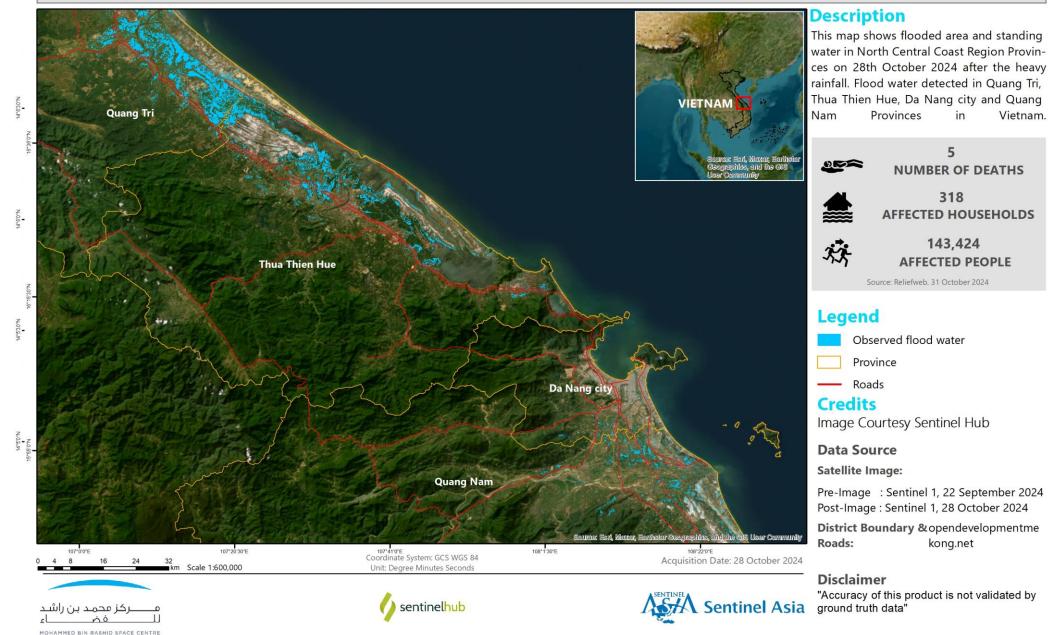


DETECTED FLOOD WATER IN NORTH CENTRAL COAST REGION OF

VIETNAM

As Observed by Sentinel-1 image on 28 October 2024

# 388 KM<sup>2</sup>



### DETECTED LANDSLIDE AND MUDSLIDE IN BATANGAS PROVINCE

## PHILIPPINES

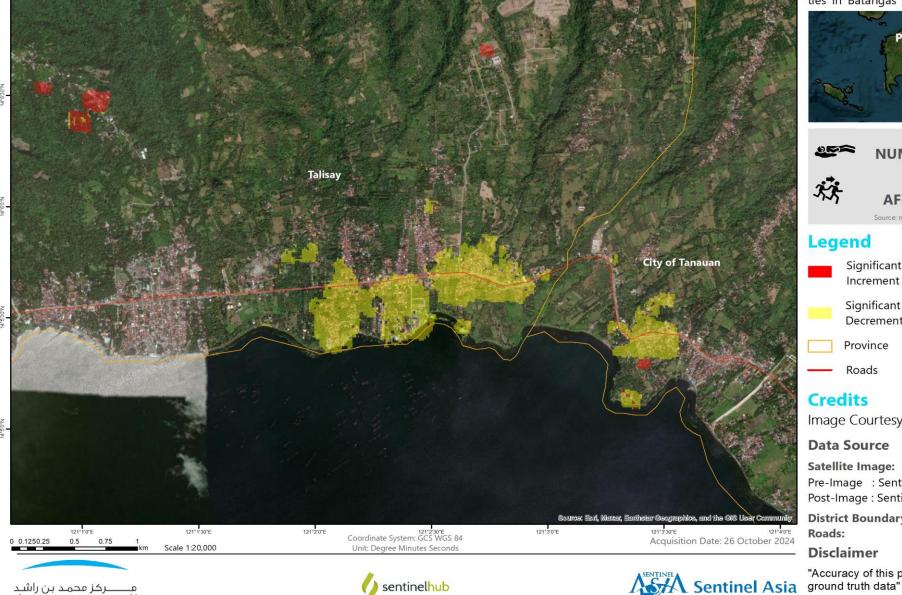
As Observed by Sentinel-1 image on 26 October 2024

## Description

This map shows Landslide and Mudslide area detected by Satellite on 26th October 2024 after Flood, Landslide and Mudslide in Talisay and City of Tanauan Municipalities in Batangas Province in Philippines.

PHILIPPINES

s, and the GIS Use



152 NUMBER OF DEATHS 40,000 AFFECTED PEOPLE Source: npr; 28 Oct: 2024

Significant Changes/Reflectivity Increment

Significant Changes/Reflectivity Decrement

Province

Roads

## Credits

Image Courtesy Sentinel Hub

**Data Source** 

Satellite Image:

Pre-Image : Sentinel 1, 03 August 2024 Post-Image : Sentinel 1, 26 October 2024

District Boundary & opendevelopmentme kong.net

Disclaimer

"Accuracy of this product is not validated by

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DETECTED FLOOD WATER ABU DHABI MUNICIPALITY SELECTED AREA



As Observed by Sentinel-1 image on 10 March 2024



## **Description**

This map shows flooded area and standing r water in the Abu Dhabi Municipality selected area on 10 March 2024 after the heavy rainfall |

## Legend

Observed flood water Selected Area Credits Image Courtesy Sentinel Hub

**Data Source** 

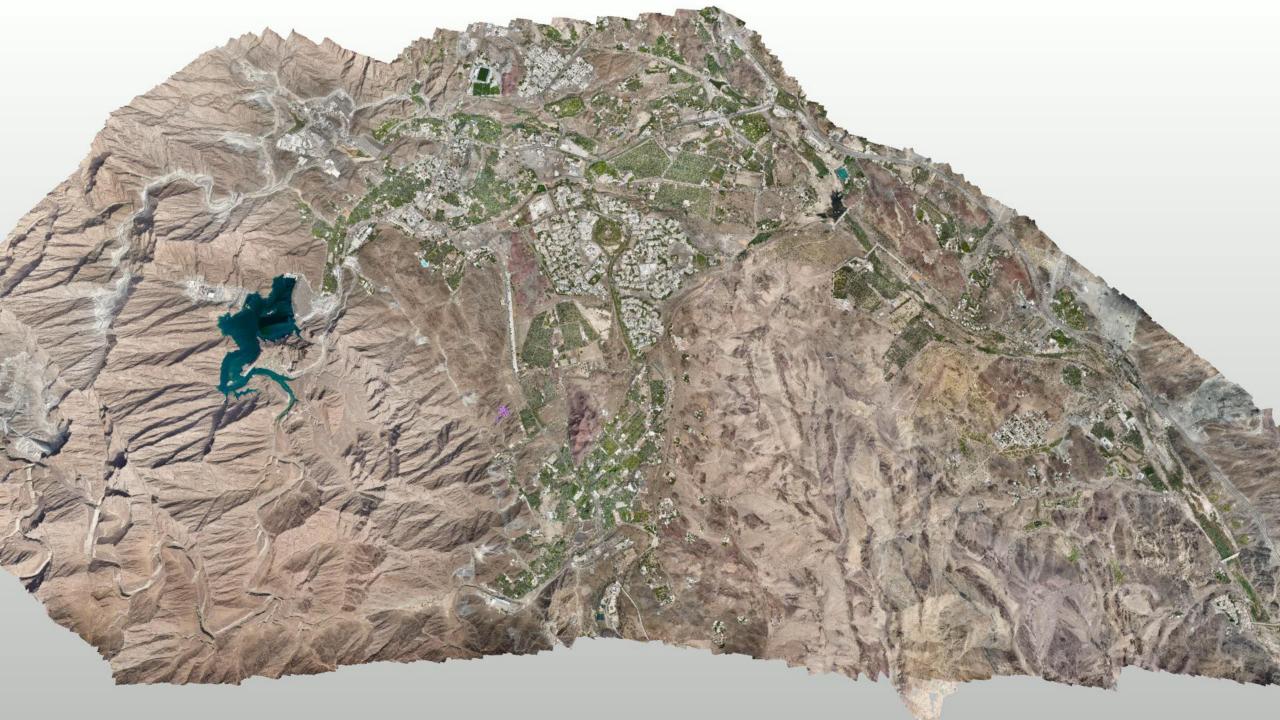
Satellite Image:

Pre-Image : Sentinel 1, 31 August 2023 Post-Image: Sentinel 1, 10 March 2024

District Boundary & opendevelopmentme Roads: kong.net

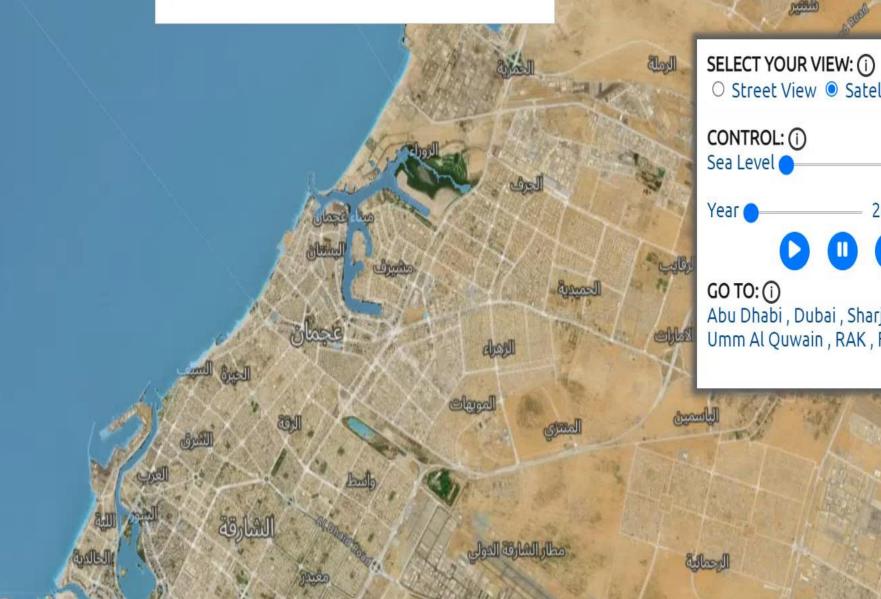
Disclaimer "Accuracy of this product is not validated by







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# **Artificial Intelligence**

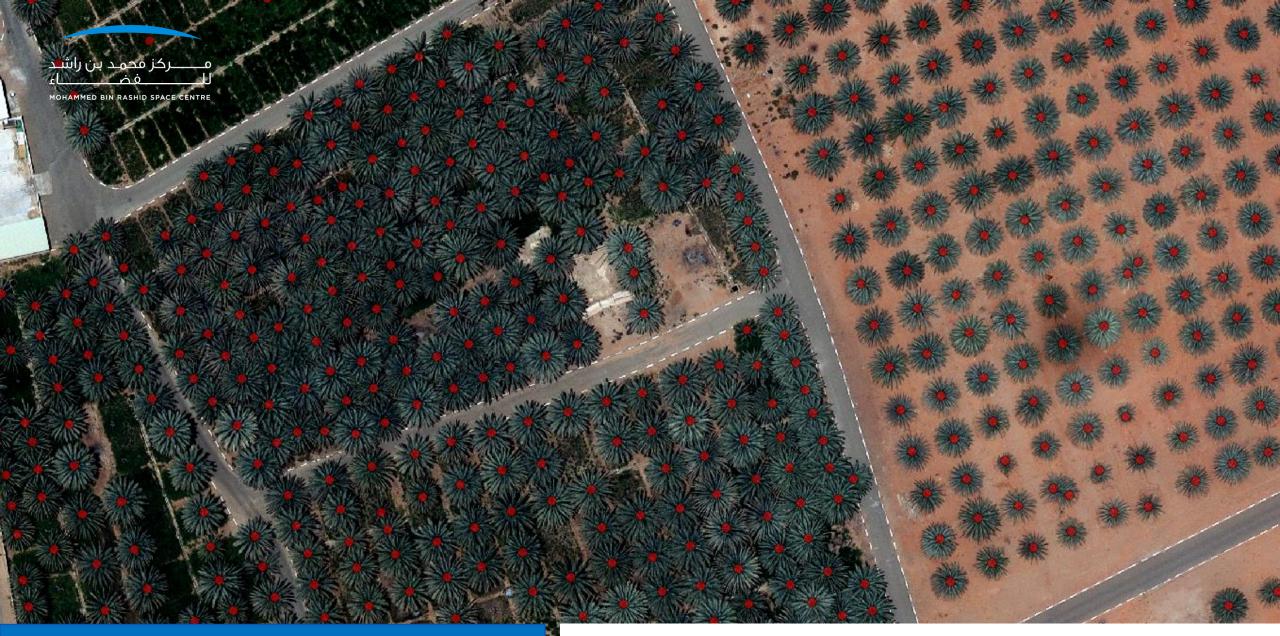












Automatic Palm trees Detection using Deep Learning

## **Detecting More than 4,000,000 Palm trees**

Automatic Ships Detection using Deep Learning







cloud : 47.160

cloud : 82.736



cloud

Ci

Cirrus

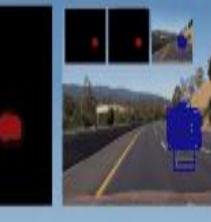
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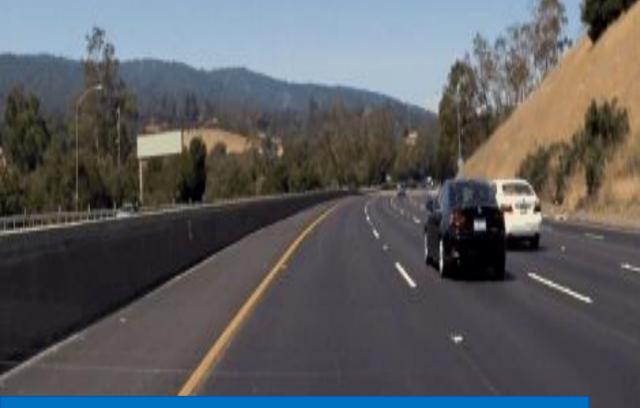
Nimbostratus

Image © 2020 Maxar Technologies

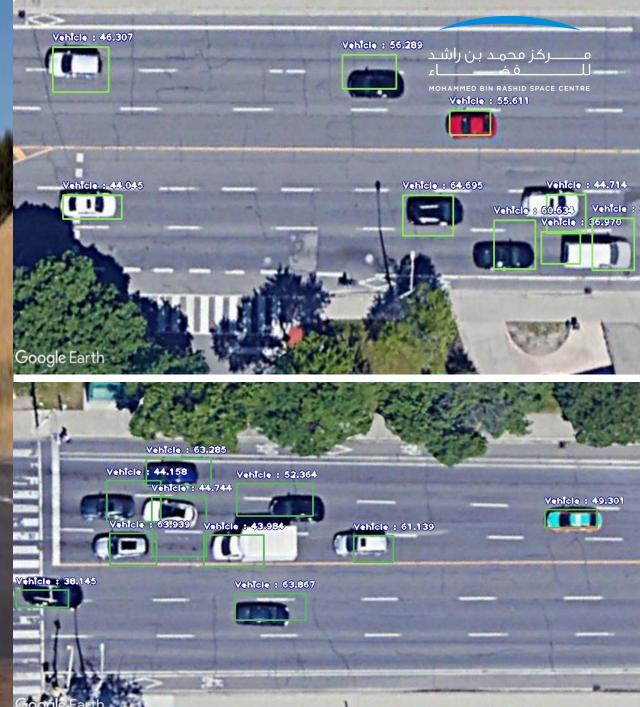
Google Earth







Automatic Cars Detection using Deep Learning



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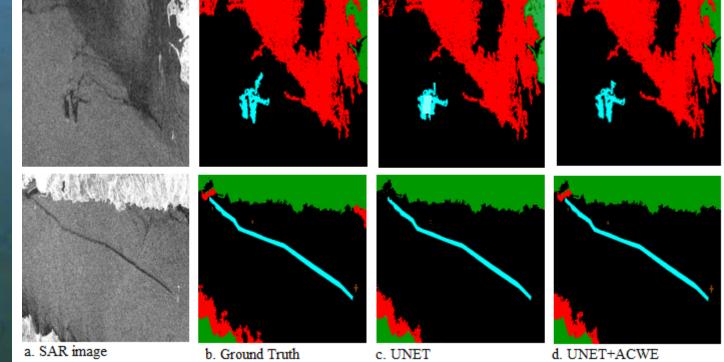


Automatic Change Detection using Deep Learning





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Automatic Oil Spill Detection using Deep Learning

# Thank you for your attention... Any Questions?

