

# INNOVATORS DAY 2025

## 24 HOURS HACKATHON

PS ID:1736

AI based frame interpolation, video generation and display system  
for WMS services

---

TEAM NAME : VISION TECHIES

TEAM MEMBERS:-

1. DHARANI K V
2. JANARTHANAN S
3. PAVITHRA V
4. RAJARAM P

## Table of Contents

**I Problem Statement**

**II Proposed Solution**

**III System I**

**IV System II**

**V Implementation and Feasibility**

**VI Conclusion**

# I Problem Statement

- **Title:** AI-Powered Frame Interpolation for Satellite Video Visualization on WebGL
- **Overview:** A novel solution for generating smooth, high-resolution satellite videos from sparse WMS imagery using frame interpolation techniques.
- **Objective:** To enable seamless visualization of satellite data on open-source WebGL libraries like OpenLayers/Leaflet, ensuring compatibility with OGC standards.

## Key Challenges:

- **Deformable Objects:** Interpolating complex, dynamic objects like clouds that deform, appear, and disappear.
- **Visualization Integration:** Overlaying generated videos on open-source WebGL platforms like OpenLayers.
- **Performance:** Ensuring smooth visualization with on-device GPU/NPUs for real-time processing.

## II Solution

### Key Features:

- Frame interpolation using AI to generate smooth videos from WMS imagery.
- Integration with OpenLayers for interactive WebGIS visualization.

### Two System Approaches:

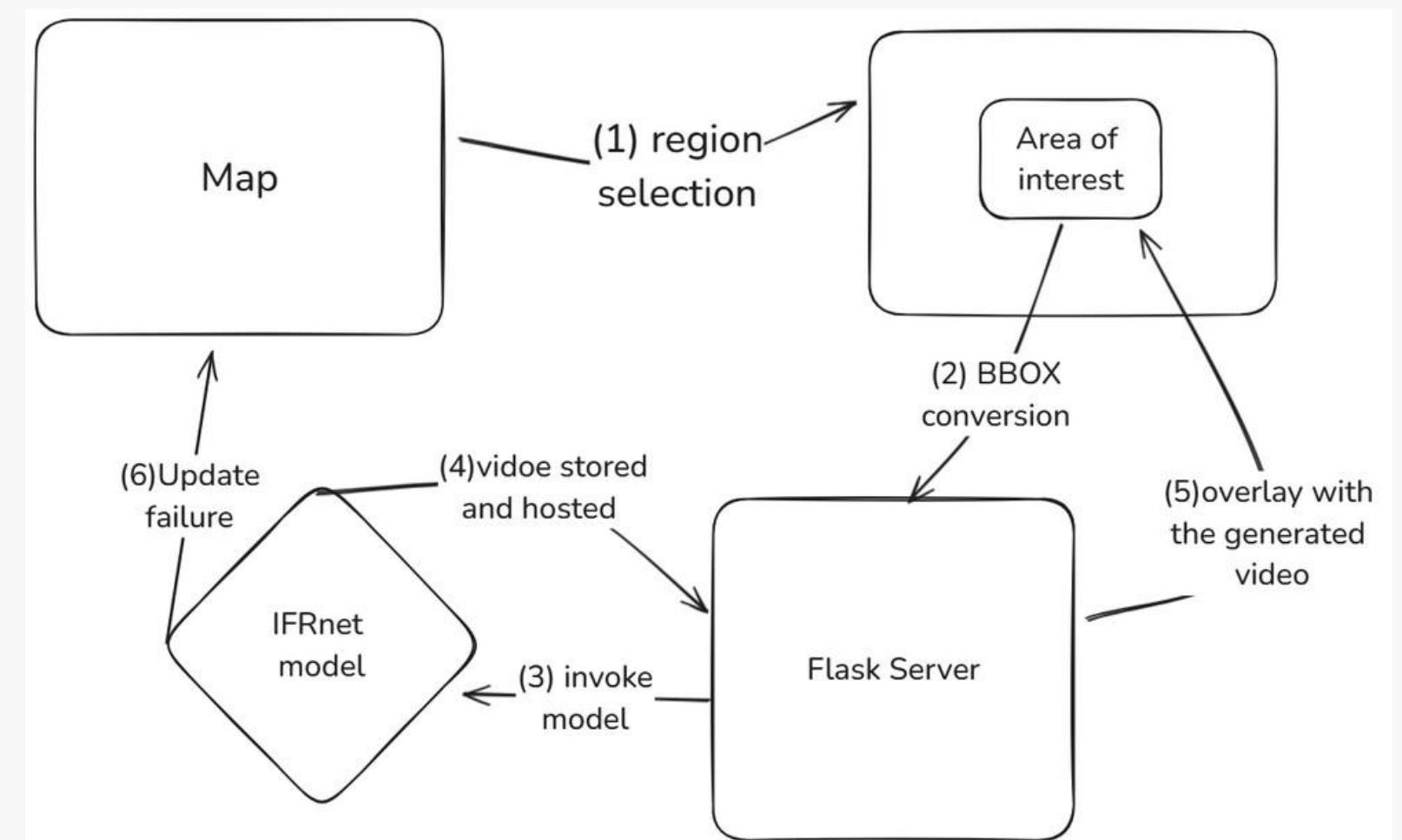
- System 1: Server-based processing with Flask backend.
- System 2: Client-side processing using TensorFlow.js.

### Usage Scenarios: Applications in platforms like

- VEDAS
- MOSDAC
- BHUVAN.

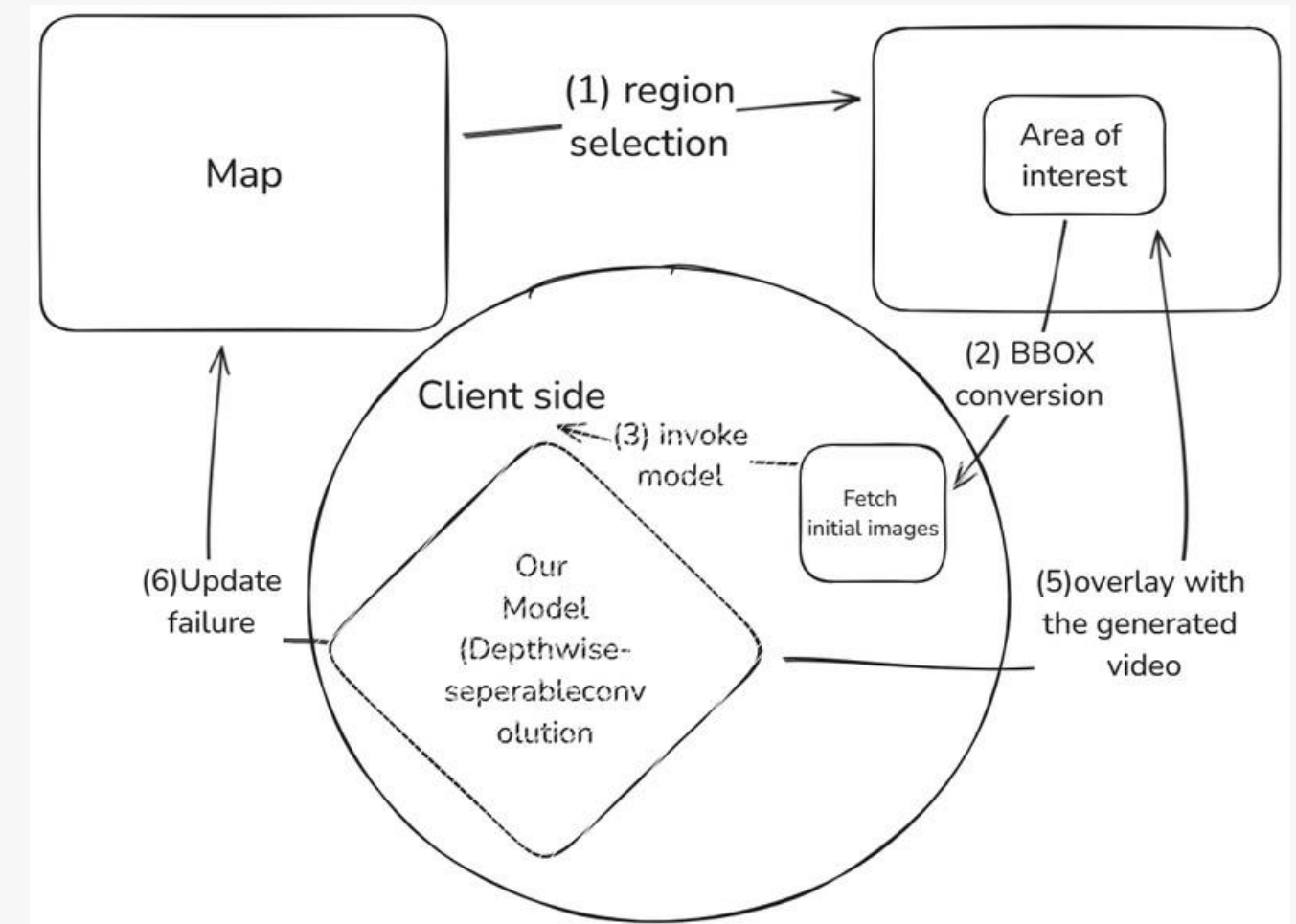
## III System - I

- This System uses the IFRnet model for image processing and video generation.
- The system handles bounding box conversion and error updates, ensuring seamless processing.
- The core components include the Map, IFRnet model, and Flask Server for video storage and hosting.



## IV System - II

- This system takes a user-selected region of interest from a map, processes it using a deep learning model, and overlays the generated video output back onto the map.
- It handles tasks like bounding box conversion and error/failure updates throughout the pipeline.
- The core system is a deep learning model, which uses efficient convolution Model to interpolate the image data.



# V Implementation and Feasibility

## Implementation Details

- Optimized Depthwise CNN model for mobile and web platforms
- Used TensorFlow JS for client-side inference

## Feasibility Considerations

- Computational resources: Ensured sufficient CPU, GPU, and RAM for efficient inference
- Memory constraints: Optimized model to minimize memory usage
- Power consumption: Optimized implementation to reduce power consumption
- Latency and throughput: Met required latency and throughput requirements for real-time applications
- Security and privacy: Implemented measures to protect sensitive user data and prevent model theft or tampering

## VI Conclusion

### Implementation Details

- Automatic video generation using frame interpolation techniques.
- Compatibility with OGC-compatible WMS services.
- Overlaying videos on an interactive browser-based map using Leaflet.
- Handling deformable objects like clouds

We plan to explore the following improvements when time permits:

- Optimizing our model for low-end devices to improve performance
- Enhancing accuracy by reducing noisy outputs and improving overall model reliability
- Implementing techniques to improve model efficiency and reduce computational requirements