



Team Name : LunarCraft

Team Leader Name : Arunaprabha K N

Problem Statement : Dual image super resolution for high-resolution optical satellite imagery and its Blind Evaluation

Team Members

Team Leader:

Name: Arunaprabha K N
College: M. Kumarasamy college of
Engineering

Team Member-1:

Name: Priyanka P
College: M. Kumarasamy college of
Engineering

Team Member-2:

Name: Maneesha K
College: M. Kumarasamy college of
Engineering

Brief about the Idea:

We propose a deep learning model that takes two low-resolution satellite images of the same scene and intelligently fuses them to generate a high-resolution output.

This is achieved using:

- Dual-Branch CNN for extracting complementary features
- Attention-Based Fusion to align and merge fine details
- Super-Resolution Decoder to upscale and enhance clarity
- Streamlit Web App for real-time testing & visualization

No need for ground truth at runtime — works on raw dual inputs!

Trained on patch-level satellite data and deployable offline.



Opportunity should be able to explain the following:

Difference from existing ideas:

Uses dual low-res images with attention-based fusion, unlike typical single-image super-resolution.

How it solves the problem:

Combines complementary info from two LR inputs to generate a clearer, higher-res image without needing HR references.

Unique Selling Point (USP):

- No ground truth needed at runtime.
- Attention-driven fusion for better detail recovery.
- Tailored for real-world satellite image enhancement.

List of features offered by the solution :

- Dual low-res image fusion for detail enhancement
- Attention-based deep learning for smart focusing
- No ground truth needed at runtime
- Auto-adjusts to various input sizes
- Delivers sharp & natural super-resolved output
- Web-based interface with GPU-powered inference



LR1 + LR2



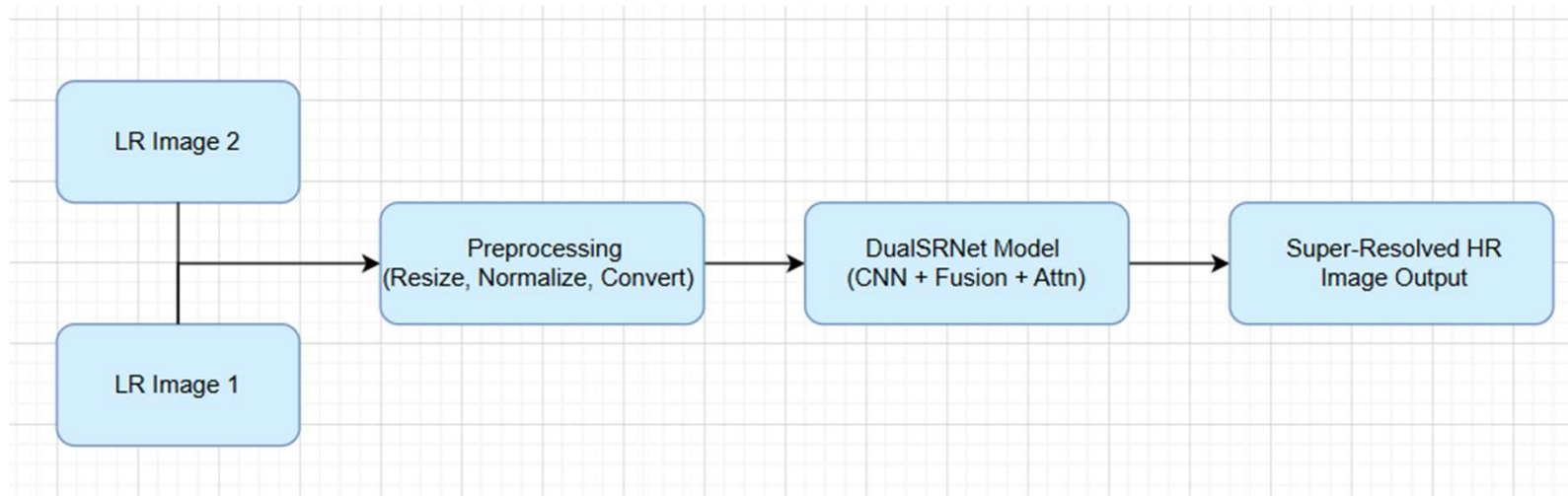
Model:
DualSRNet
CNN + Feature
Fusion + Attention



Super-Resolved Output

Enhancing satellite images through a dual-input super-resolution model built with CNN and attention.

Process flow diagram or Use-case diagram



User Interface – Web App Preview (Built with Streamlit)

Deploy


How It Works

- Upload two complementary LR images (e.g. different angles, slight shifts).
- (Optional) Upload a Ground Truth HR image for visual comparison.
- Click **Generate** to create a super-resolved image.
- Download the output or compare with original inputs.

✓ No need for HR image during actual usage

⚙ Works even without internet (locally)

⚡ GPU-accelerated if available



Dual Image Super-Resolution

Enhance two complementary low-resolution satellite images into one clear high-resolution output using deep learning.

Just upload, generate, and download.

📁 Upload LR Image 1

Drag and drop file here

Limit 200MB per file • JPG, JPEG, PNG

Browse files

📁 Upload LR Image 2

Drag and drop file here

Limit 200MB per file • JPG, JPEG, PNG

Browse files

📁 Upload HR Image (optional)

Drag and drop file here

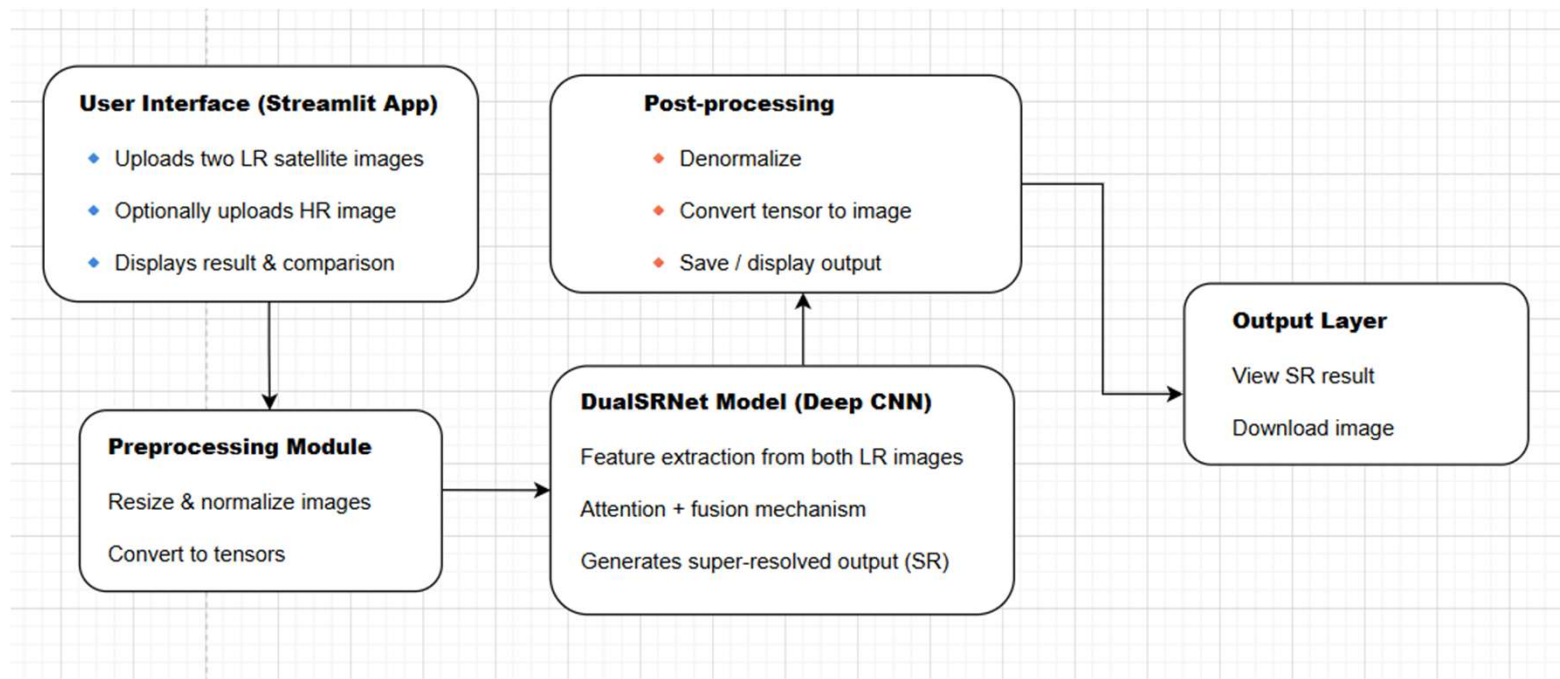
Limit 200MB per file • JPG, JPEG, PNG

Browse files

🚀 Generate Super-Resolved Image

📁 Upload inputs and click Generate to begin.

Architecture diagram of the proposed solution



Technologies to be used in the solution:



Deep learning
framework used to
train the SR model



Transforms, data loading, utility



Image preprocessing



Core programming language



Interactive web app
for testing the model

Estimated implementation cost :

Component	Description	Estimated Cost
Model Training	Local machine with CPU/GPU (academic use)	₹0
Dataset	Publicly available open-source datasets	₹0
Development Frameworks	PyTorch, Streamlit (open-source tools)	₹0
Deployment (UI)	Streamlit local app (no cloud needed)	₹0
Hardware Requirements	Standard system with Python environment	₹0
Total		₹0



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THANK YOU