

Department of Computer Science and Engineering

Satellite Image Classification Using CNNs

Dr.Rakesh Kumar M
Assistant Professor/CSE

TAMILINI D K
(220701299)
VELLANKI CHANDRA HARSHA
(220701314)

Problem Statement and Motivation

- ❑ traditional land classification methods are time-consuming, prone to human error, and lack scalability.
- ❑ Our project aims to automate land-type identification from satellite images using CNN to enhance both speed and accuracy.
- ❑ Manual classification lacks consistency and scalability across regions.

Existing System

- ❑ Current systems rely on manual interpretation or basic image processing techniques.
- ❑ These methods are inefficient for large datasets and fail with complex land-use patterns.
- ❑ Many traditional methods don't effectively capture spatial features.
- ❑ Real-time analysis is not feasible in most legacy systems.

Objectives

- ❑ Classify satellite images into predefined land use categories.
- ❑ Develop a CNN-based model with high accuracy and reliability.
- ❑ Deploy a simple interface for testing classification results.

Abstract

- ❑ We use CNN to classify satellite images into land types like forest, industrial, and residential.
- ❑ EuroSAT dataset is used to train and validate the model.
- ❑ CNNs learn hierarchical features from images directly.
- ❑ The system supports 10 land cover classes using EuroSAT.

Proposed System

- ❑ Satellite images are processed and passed into a CNN for classification.
- ❑ The system extracts visual features and classifies them into land types.
- ❑ Includes dropout and augmentation to improve robustness.
- ❑ Batch processing of multiple images is supported.

System Architecture

- ☐ **Dataset Acquisition (EuroSAT)**
- ☐ **Preprocessing (resize, normalize)**
- ☐ **CNN Model (training & feature extraction)**
- ☐ **Model Evaluation**
- ☐ **Output – Prediction of Land Type**
- ☐ **Training metrics monitored for performance optimization.**



List of Modules

1. Dataset Loader & Preprocessing
2. CNN Model Training
3. Evaluation & Metrics Display
4. Image Prediction Interface
5. Data Augmentation Module
6. Visualization Module

Functional Description

- ❑ **Preprocessing:** Resize, normalize images for model input.
- ❑ **Training:** CNN learns spatial and color features.
- ❑ **Prediction:** Uses softmax output to classify image.
- ❑ **Evaluation:** Accuracy, precision, confusion matrix.
- ❑ **Data Augmentation:** Adds variety (rotate/flip) for better generalization.
- ❑ **Visualization:** Displays predicted class with confidence score.

Implementation & Results of Module

- ❑ Implemented in Python using TensorFlow/Keras.
- ❑ Accuracy achieved: $\sim 90\%$ on validation dataset.
- ❑ Results shown through confusion matrix and sample predictions.
- ❑ Model trained with early stopping to prevent overfitting.
- ❑ Validation accuracy stabilized after 25–30 epochs.

Conclusion & Future Work

- ❑ CNNs effectively classify satellite land types with high accuracy.
- ❑ The model can help in city planning and disaster management.
- ❑ Extend to time-series data for environmental change detection.
- ❑ Optimize model for edge devices and real-time use cases.

References

- ❑ EuroSAT Dataset – <https://github.com/phelber/eurosat>
- ❑ TensorFlow CNN Guide – <https://www.tensorflow.org>
- ❑ IEEE Access: Deep Learning for Remote Sensing
- ❑ Elsevier: Comparative Review on Land Use Classification
- ❑ Research papers on Satellite Image Classification



Thank You