

### AI/ML BASED MAPPING OF PLANETARY MORPHOLOGY

Exploring advanced techniques for analyzing planetary surface features using AI and ML.



1

#### Project AIM

To develop an Al/ML-based system that accurately identifies and analyzes planetary objects in the universe.

2

#### Importance of Project

This project is significant for advancing space research, as it provides tools to explore planetary surfaces remotely, reducing the need for physical exploration and paving the way for automated planetary studies.

3

#### Impact of Project

The Project assist space agencies in planning exploration missions, facilitate real-time monitoring of planetary changes, and inspire further advancements in astronomical AI applications.

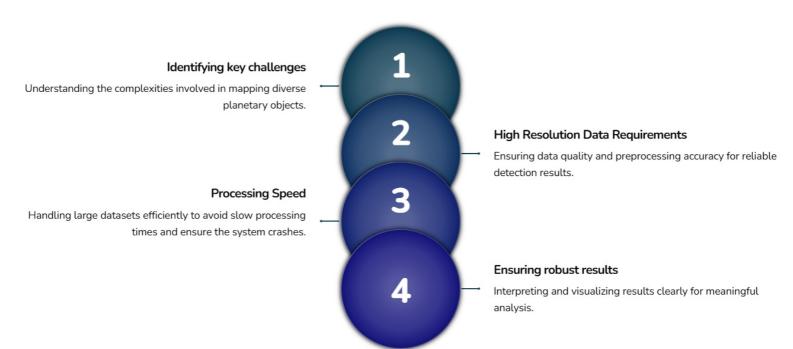
## INTRODUCTION

Identifying Diverse planetary objects for Various

Applications

## **CHALLENGES**

Identifying and addressing key issues in planetary morphology mapping



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Data Requirements	Hardware Requirements	Software Requirements
High-resolution satellite imagery, multispectral data.	Processor: 6-core or higher (e.g., Intel i7/Ryzen 7 or better) RAM: 16 GB or more GPU:( NVIDIA Tesla V100 better for deep learning)	TensorFlow, Colab Jupyter Notebook, python, keras, matplotlib other tools

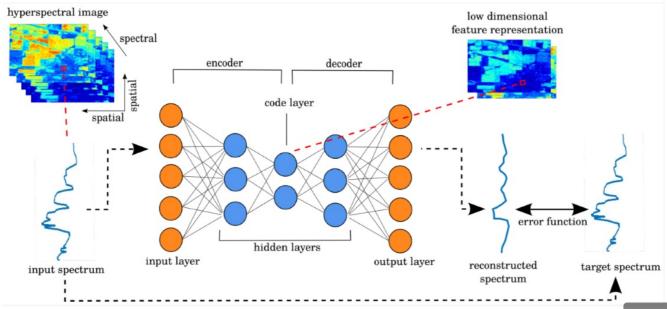
# **REQUIREMENTS**

Essential tools and resources for effective mapping techniques



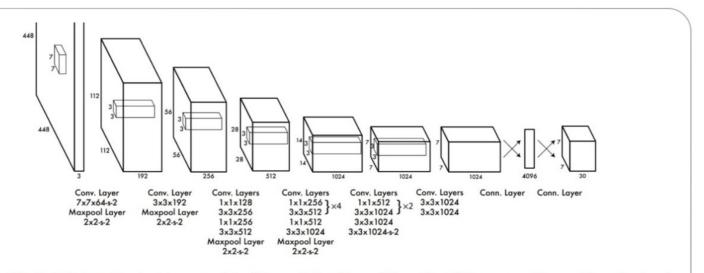
# **CNN MODEL ARCHITECTURE**

Overview of Al/ML techniques (e.g., CNNs, U-Net) applied for mapping.



## YOLO MODEL ARCHITECTURE

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The Architecture. Our detection network has 24 convolutional layers followed by 2 fully connected layers. Alternating  $1 \times 1$  convolutional layers reduce the features space from preceding layers. We pretrain the convolutional layers on the ImageNet classification task at half the resolution ( $224 \times 224$  input image) and then double the resolution for detection.

## **METHODOLOGY**

1 2 3 4

#### **Data Collection**

Gathering relevant datasets from various sources to form the foundation for model training.

#### **Data Preprocessing**

Cleaning and transforming raw data into a suitable format for analysis, enhancing model performance.

#### **Model Training**

Applying algorithms to the preprocessed data to train the Al/ML model, learning from the patterns within the data.

#### Model Validation

Evaluating the trained model using validation datasets to assess its performance and generalizability.

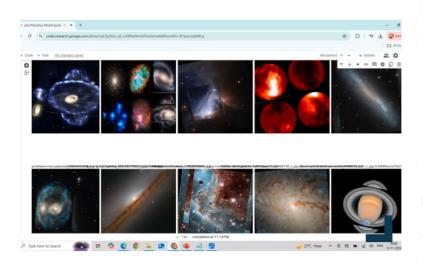
## DIFFERENCE BETWEEN CNN AND YOLO

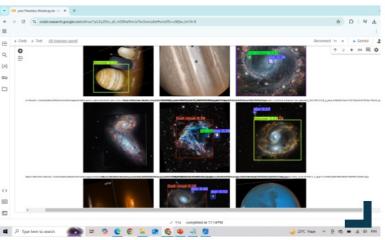
	CNN	YOLO
Basic Concept	Multi-stage detection with separate proposal.	Single-stage detection in one forward pass.
Speed	Slower, less suited for real-time use.	Very fast, ideal for real-time applications.
Detection Process	Region proposals followed by classification.	Directly predicts bounding boxes and classes per grid cell.
Emerging Use	Static image analysis, segmentation, and complex object detection.	Text to formula, data wrangling, synthetic data

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## **RESULTS AND ANALYSIS**

Effectiveness of AI/ML Models in Terrain Mapping





#### **Output Images**

Description of a primary heading

#### Input Images

Description of a primary heading



## CONCLUSION AND FUTURE DIRECTIONS

Exploring the Potential of AI/ML in Planetary Morphology Mapping

Real-time Data Processing

responsiveness.

data from planetary missions, improving

Implementing AI/ML allows for real-time processing of

#### Integration of AI/ML Techniques

Leveraging advanced AI/ML methods enhances the accuracy of planetary morphology mapping.

# **Future Research Opportunities** Identifying further areas for research can lead to significant discoveries in planetary science. Interdisciplinary Collaboration Improvement in Data Analysis Collaboration between different scientific disciplines will AI/ML can streamline data analysis processes, enabling foster innovation in planetary mapping. quicker interpretations of geological features.

