

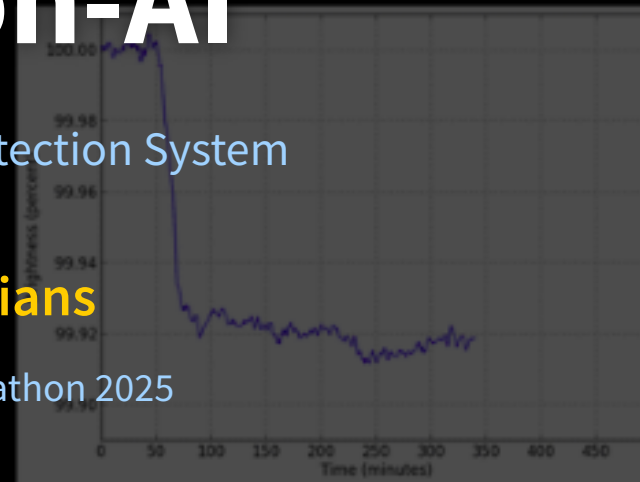
# Step 1: find exoplanets

## ExoVision-AI

AI-Powered Exoplanet Detection System

Team **Nephelians**

NASA Space Apps Hackathon 2025



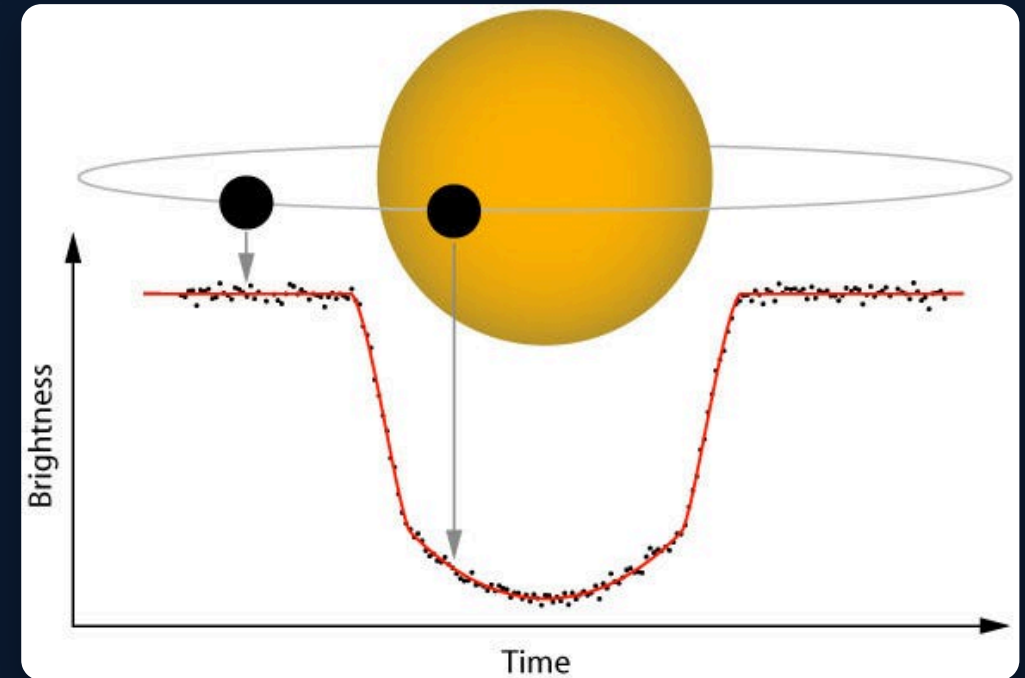
# Project Overview & Problem Statement

## 🌐 The Challenge

- Thousands of **exoplanet candidates** identified by Kepler/K2 missions
- Many remain **unconfirmed** due to false positives
- Causes: binary stars, stellar variability, instrumental noise

## 🧠 Our Solution

- **AI-powered classification** of exoplanet candidates
- Accelerates discovery by reducing manual verification time
- Provides confidence scores for candidate validation



# Data & Methodology

## Dataset

- ▶ NASA **K2 Planets and Candidates Catalog**
- ▶ **4,004 entries** with 28 carefully selected features

## Key Features

🕒 Orbital Period

📏 Planet Radius

🚀 Planet Mass

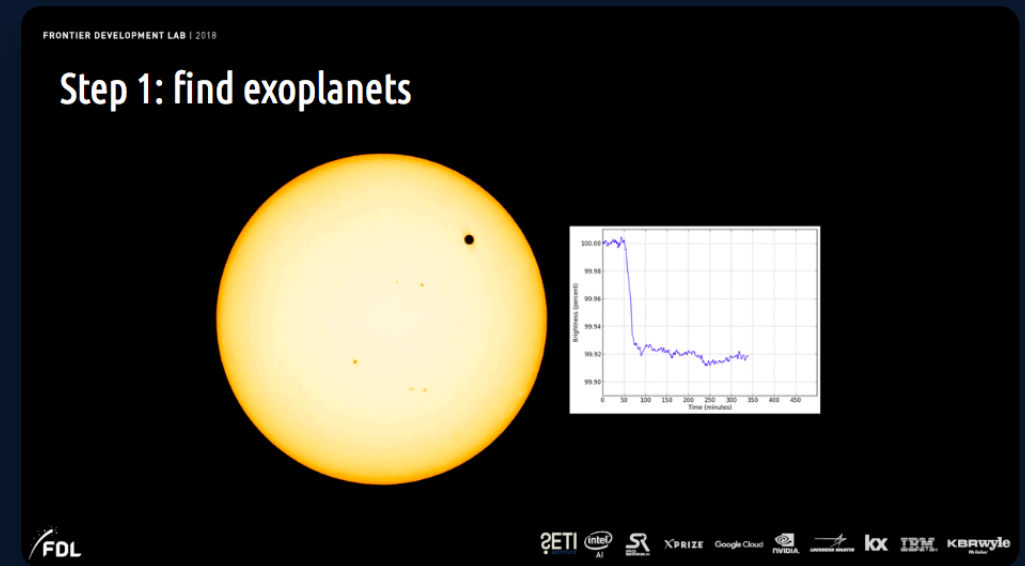
☀️ Stellar Temp

🌐 Stellar Radius

🌌 System Distance

## Preprocessing

- ▶ Missing values imputed with **median**
- ▶ Features **standardized** to zero mean and unit variance
- ▶ Created new features: star-planet size ratio, temperature difference



# Machine Learning Models

## ↗ XGBoost Classifier

n\_estimators: 300

learning\_rate: 0.05

max\_depth: 6

subsample: 0.8

✅ Best performing model

## 📊 Decision Tree Classifier

max\_depth: 4

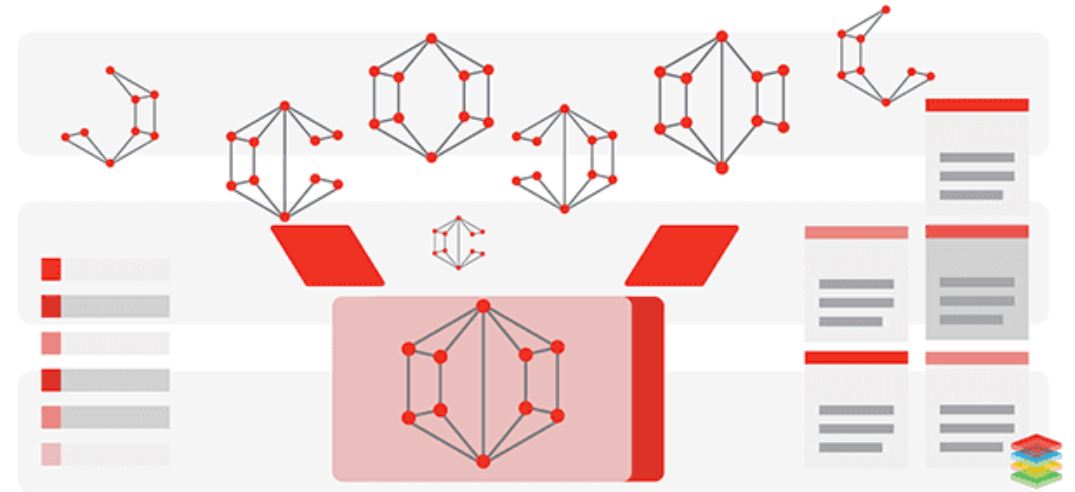
min\_samples\_split: 30

min\_samples\_leaf: 15

max\_features: sqrt

👁 Highly interpretable

## — ML / DL MODEL VISUALIZATION —



| Model         | Accuracy | Precision | Recall | F1 Score |
|---------------|----------|-----------|--------|----------|
| XGBoost       | 99.7%    | 99.5%     | 100%   | 99.7%    |
| Decision Tree | 99.6%    | 99.4%     | 100%   | 99.7%    |

# Web Application

## 📁 Flask-Based Interface

- **Two input methods** for exoplanet data
- **Real-time predictions** from both models
- Interactive **visualizations** and confidence scores

## 📁 Input Methods

- 📁 **CSV Upload** - Process multiple candidates at once
- ✎ **Manual Input** - Enter data for single candidate

## 🔗 Key Features

- 📊 Feature distribution charts
- 🔢 Correlation heatmaps
- ✓ ROC curves and performance metrics

### Exoplanet Manual Input

[DEMO](#)

Planet Radius (Earth radii)

1.2

Orbital Period (days)

10.5

Stellar Temperature (K)

5800

Discovery Method

Transit

Analyze and Predict

## 🔗 User Workflow

1

Input exoplanet data

2

AI model analysis

3

View predictions & visualizations

# Results & Impact

## Model Performance

**99.7%**

Accuracy

**99.5%**

Precision

**100%**

Recall

**99.7%**

F1 Score

## ⚠️ Limitations

- Catalog-level **information leakage**
- Performance may be **inflated** by dataset biases
- Limited to **K2 catalog** characteristics

## ↗️ Research Impact

- **Accelerates** exoplanet discovery process
- Reduces **manual verification** time
- Guides **follow-up observations**

## 💡 Future Improvements

- **Cross-validation** with different datasets
- More **rigorous scientific validation**
- Integration with **real-time telescope data**

# Team & Acknowledgments

## Team Nephelians



**Muhammad Ahmad**  
Team Leader & Data  
Scientist



**Syed Darain**  
ML Engineer



**Muhammad Ahsan**  
Backend Developer





**Muhammad Mohsin**  
Frontend Developer



**Ali Hassan**  
Research Lead

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## Technology Stack

 Python

 Scikit-learn

 XGBoost

 Flask

 Seaborn

 Pandas



### GitHub Repository

<https://github.com/c0llect0rr/ExoVision-AI>



### Contact Email

[mahmadshafi861@gmail.com](mailto:mahmadshafi861@gmail.com)