

Frontend and Backend Documentation

ExoHabitAI: Technical Documentation

1. Project Overview

ExoHabitAI is an advanced machine learning platform designed to assess the habitability of exoplanets. It goes beyond simple prediction by providing a **Dynamic 3D Simulation** of the target planet based on its physical properties.

The system features a **React Frontend** for real-time visualization, a **Three.js Engine** for 3D rendering, and a **Flask Backend** that serves the AI model.

2. Key Features

A. Dynamic 3D Planet Simulation

Unlike static images, the 3D planet **mutates** based on user input:

- **Magma World:** If Temperature > 330K, the planet glows red with volcanic activity.
- **Ice World:** If Temperature < 200K, the surface becomes reflective white/cyan (frozen).
- **Gas Giant:** If Radius > 1.6 Earths, the planet turns into a thick, hazy Jovian world (purple/gas texture).
- **Terra Class:** If conditions are habitable, it renders as a blue, ocean-covered world with white clouds.

B. Commander's Mission Log

- **Auto-Save:** Every scan is automatically saved to the browser's local storage.
- **Persistence:** The history remains even if the page is refreshed or the browser is closed.
- **Quick Recall:** Users can click on past logs to instantly reload that planet's data and 3D model.

3. Technology Stack

Component	Technology	Purpose
Frontend	React 18 (Vite)	User Interface & State Management
3D Engine	Three.js / React-Three-Fiber	Real-time 3D rendering of planets
Styling	Tailwind CSS	Glassmorphism UI & Responsive Design
Backend	Python (Flask)	API Server & AI Logic
AI Model	Scikit-Learn	Random Forest Classifier
Data Handling	Pandas / NumPy	Data processing & Scientific calculations

4. System Architecture

Data Flow (The "Brain" of the App)

- Input:** User enters **Radius, Mass, Temperature** in **Analyzer.jsx**.
- API Call:** React sends a **POST** request to Flask (**/predict**).
- AI Inference:** Flask loads the Random Forest model and predicts:
 - Habitable (1):** High probability of liquid water.
 - Non-Habitable (0):** Hostile conditions.
- 3D Rendering Logic:**

- React receives the prediction AND the raw input data.
 - **Planet3D.jsx** analyzes the data locally to decide the **texture** and **color** of the sphere (e.g., "Is it hot? Make it red.").
5. **History:** The result is appended to the **localStorage** array for the "Mission Log."

5. Frontend Documentation (/frontend)

A. **Analyzer.jsx** (The Core Tool)

- **Role:** The main interactive dashboard.
- **State:** Manages **formData** (inputs), **result** (prediction), and **history** (saved logs).
- **Integration:**
 - Fetches data from Backend.
 - Passes data to **<Planet3D />** for rendering.
 - Saves scans to **localStorage**.

B. **Planet3D.jsx** (The Visualization Engine)

- **Role:** Renders the interactive 3D sphere.

Logic (Procedural Generation):

JavaScript

```
const biome = useMemo(() => {
  if (radius > 1.6) return { type: "GAS GIANT", color: "purple" };
  if (temp > 330) return { type: "LAVA", color: "red", emissive: true };
  if (temp < 200) return { type: "ICE", color: "white", roughness: 0.1 };
  return { type: "TERRA", color: "blue" }; // Earth-like
}, [radius, temp]);
```

-
- **Effects:** Uses **Sparkles** (for stars/atmosphere) and **MeshStandardMaterial** for realistic lighting.

C. **Dashboard.jsx** (The Overview)

- **Role:** Displays global statistics (Total Planets, Habitable Count).
- **Charts:** Uses **Recharts** to show Feature Importance (which variables the AI cares about most).

6. Backend Documentation (/backend)

app.py (The API)

- **Endpoints:**
 - **POST /predict:** Handles the core logic. Includes a **safety fallback** (if **Period** is empty, it assumes Earth-like 365 days).
 - **GET /dashboard-data:** Returns dataset statistics.
- **Error Handling:** Wraps predictions in **try/except** blocks to prevent 500 crashes on bad input.

Tech Stack: Python, Flask, Pandas, Scikit-Learn

System Architecture

The backend is designed as a lightweight **REST API** using the Flask framework. Due to the development environment (Google Colab), the API is instantiated and tested interactively within the notebook runtime using a **test_client** rather than a persistent web server.

Core Components

1. **app (Flask Instance):** Handles HTTP routing and error management.
2. **utils.preprocess_input:** A dedicated preprocessing module that mirrors the training phase. It transforms raw JSON data into the specific **16-feature vector** required by the model.
3. **model (Random Forest):** The pre-trained **.pkl** artifact loaded into memory for real-time inference.

1. API Overview

The ExoHabitAI backend is built on **Flask**. It exposes the trained Random Forest model to perform real-time inference. It accepts raw planetary data, performs real-time feature engineering (matching the training pipeline), and returns a probabilistic classification.

2. Endpoints

Endpoint	Method	Description
/	GET	Health check. Returns API status.
/predict	POST	Accepts exoplanet JSON, returns habitability prediction
/rank	GET	Returns the top 10 pre-ranked habitable candidates

3. Request Format (/predict)

JSON:

```
{ "Name": "Earth 2.0", "Radius": 1.2, "Mass": 1.5, "Period": 200,  
  "SemiMajorAxis": 0.8, "EqTemp": 270, "Density": 5.5, "StarTemp": 5000,  
  "StarLum": 0.8, "StarMet": 0.0, "Insolation": 0.9, "StarType": "K" }
```

4. Response Format

JSON:

```
{  
  "confidence_score": 0.87,  
  "habitable_flag": 1,  
  "input_planet": "Earth 2.0",  
  "prediction": "Potentially Habitable"  
}
```

5)Endpoint: Predict Habitability

- **URL:** /predict
- **Method:** POST

- **Description:** Accepts raw planetary parameters, applies scaling/feature engineering, and returns a binary classification with a confidence score.

Request Format

JSON:

```
{
  "Name": "Kepler-186 f",
  "Radius": 1.17,
  "Mass": 1.4,
  "Period": 129.9,
  "EqTemp": 250,
  "Insolation": 0.45,
  "StarTemp": 3755,
  "StarType": "M"
}
```

Response Format

JSON:

```
{
  "input_planet": "Kepler-186 f",
  "prediction": "Non-Habitable",
  "habitable_flag": 0,
  "confidence_score": 0.0
}
```

3. Data Processing Logic

To ensure the API predictions match the model's training performance, the backend implements the following pipeline on every request:

1. Feature Engineering:

- Calculates **Habitability_Score** using the Earth Similarity Index (ESI) formula.

- Applies `Log(Period + 1)` transformation to normalize orbital periods.
- Performs One-Hot Encoding for `StarType` (G, K, M, Other).

2. Vectorization:

- Aligns input data to exactly **16 features** (Radius, Mass, Density, etc.).
- Fills missing physical values (e.g., `Density`) with training defaults (0) to prevent crashes.

3. Scaling:

- Applies the saved `StandardScaler` (`scaler.pkl`) to normalize inputs to the standard normal distribution expected by the model.

4. Validation & Testing

- **Methodology:** Integration testing was performed using the Flask `test_client()` to simulate HTTP POST requests within the notebook environment.
- **Result:** The API successfully processed a request for candidate **Kepler-186 f**, returning a valid JSON structure without runtime errors or dimension mismatches.
- **Error Handling:** The system includes try-catch blocks to return `500 Internal Server Error` messages with debug details if the input JSON is malformed or the model fails to load.

7. Setup & Run Guide

Step 1: Backend (The Brain)

Bash

```
cd backend
```

```
# 1. Install Libraries
```

```
pip install flask flask-cors pandas numpy scikit-learn joblib
```

```
# 2. Build the AI Model (Run Once)
```

```
python setup.py
```

```
# 3. Start Server
```

```
python app.py
```

Server is live at: <http://127.0.0.1:5000>

Step 2: Frontend (The Interface)

Open a **new terminal**:

Bash

cd frontend

1. Install Libraries (including 3D tools)

npm install

npm install three @react-three/fiber@8.16.8 @react-three/drei

--legacy-peer-deps

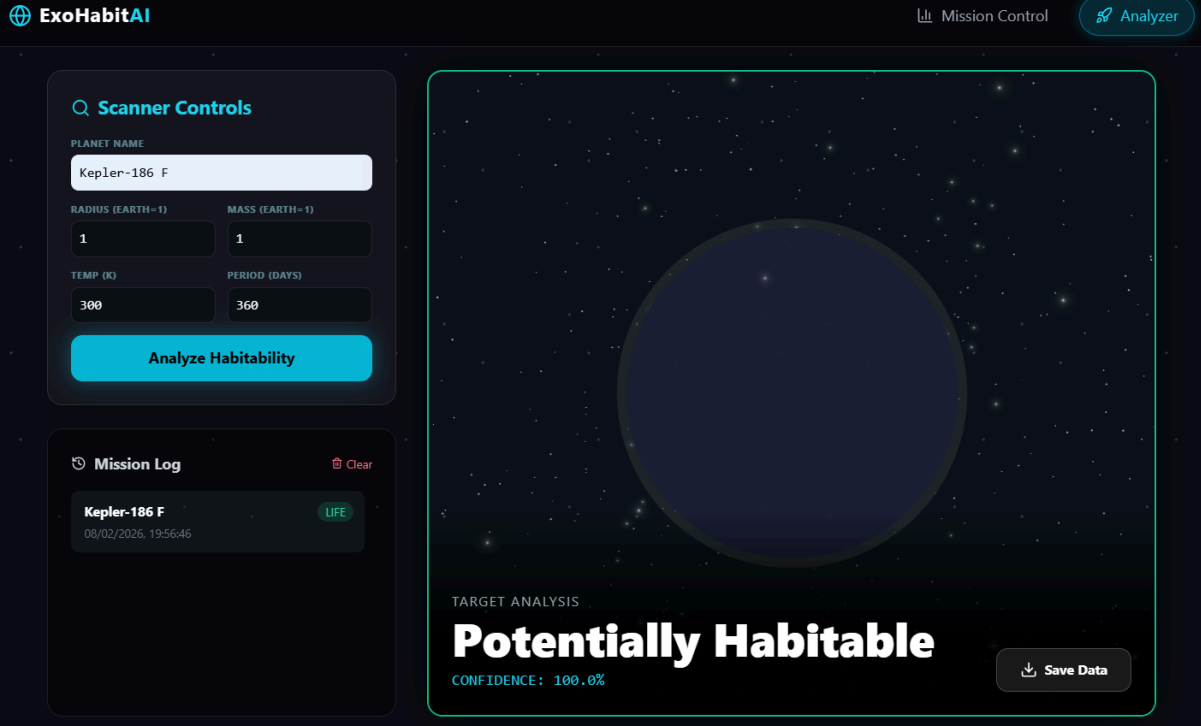
2. Launch UI

npm run dev

App is live at: <http://localhost:5173>

ScreenShorts:





Scanner Controls

Analyze Habitability

Mission Log

Clear

Kepler-186 F