



Development of an application based on a Machine Learning model
to identify potentially dangerous asteroids.

AGENDA

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- **introduction**
 - about the project
 - definitions
- **Exploratory Data Analysis, EDA**
 - dataset
 - study of correlations between variables
 - challenges of model development:
- **models**
 - XGBoost
 - choosing the best model
- **deployment**
 - building a mock-up
 - haas application

INTRODUCTION

ABOUT THE PROJECT**ABOUT THE PROJECT**

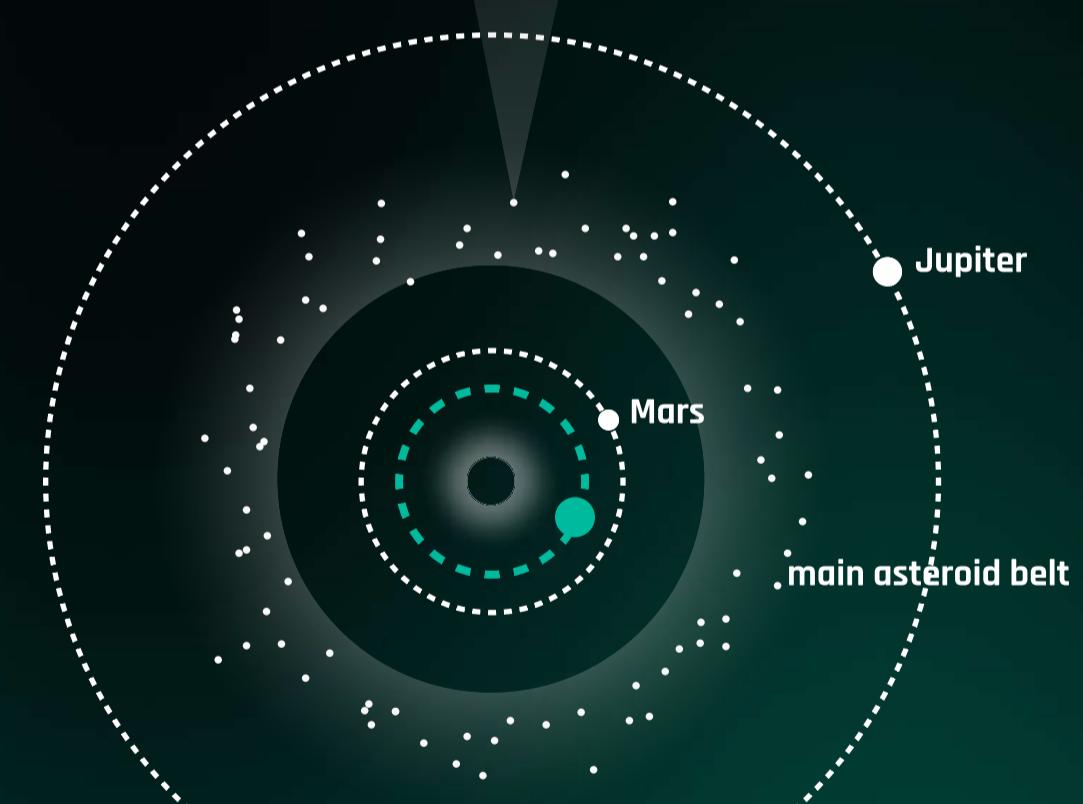
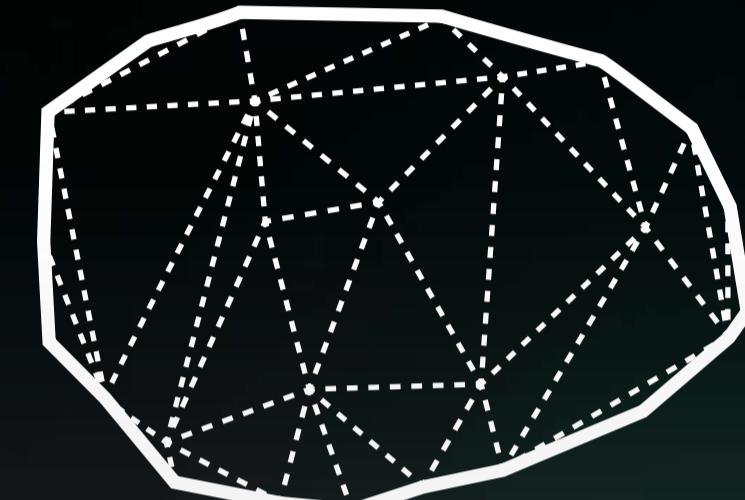
In the era of rapid technological advancement, monitoring threats from outer space has become crucial. One significant danger comes from asteroids that could potentially collide with Earth, causing catastrophic consequences. Our presentation, titled "Development of an Application Based on a Machine Learning Model to Identify Potentially Dangerous Asteroids," focuses on using advanced machine learning techniques to identify and assess the risk posed by these objects.

In this presentation, we will discuss the steps to create an effective predictive model, starting with NASA data analysis. This includes data processing, feature selection, and model training. We will present our results, covering evaluation metrics like precision and recall, and the challenges faced. Finally, we will showcase our application, featuring visualizations and interactive elements, enabling users to understand asteroid risks and make informed decisions.

DEFINITIONS**ASTEROID (PLANETOID)**

Asteroid (planetoid) is a celestial body, which is not a planet, moon, comet or meteoroid, with a size from 1m to more than 1000 km.

Most of the recorded 1.3 million asteroids are located between the orbits of Mars and Jupiter - in the so-called main asteroid belt.



DEFINITIONS

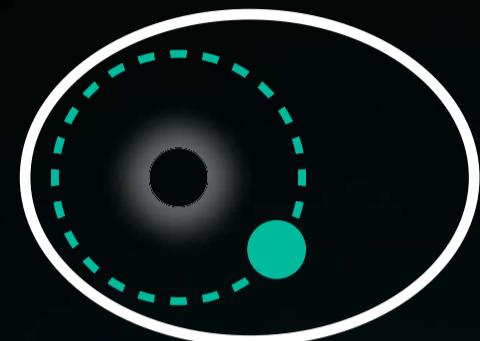
POTENTIALLY HAZARDOUS ASTEROIDS

Potentially Hazardous Asteroids

(PHA for short) - asteroids that move around the Sun in orbits that intersect the Earth's orbit or are close to it.

grupa**Amora**

orbita znajduje się całkowicie poza orbitą Ziemi

**grupa****Apolla**

półosi wielka większa niż orbita Ziemi
orbity przecinają się

**grupa****Atena**

półosi wielka mniejsza niż orbita Ziemi
orbity przecinają się

**grupa****Atiry**

orbita zawiera się całkowicie wewnętrznie
orbity Ziemi



DEFINITIONS**TORINO SCALE**

Torino Scale - an 11-point scale (from 0 to 10) by which NASA assesses the potential threat of space catastrophe posed by various objects approaching Earth from space.

For the purposes of the project, the scale has been reduced to a binary classification (0 - safe, 1 - hazardous).

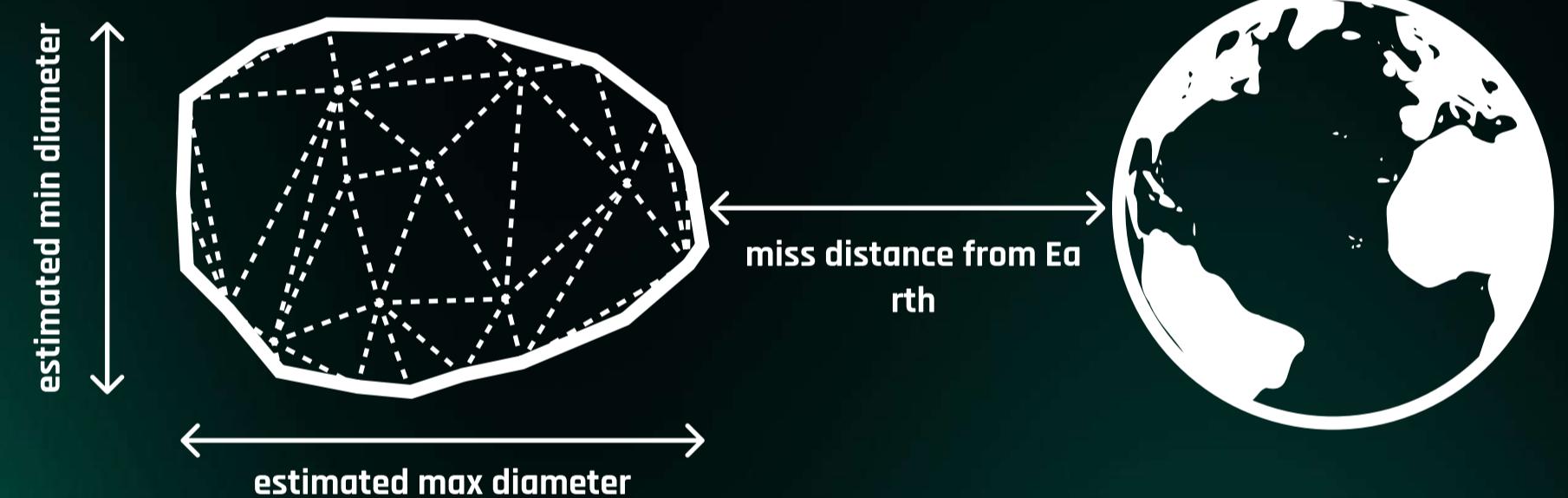
pewne zderzenie**zagrożenie****wymagana
uwaga
astronautów****norma****brak zagrożenia****10****9****8****7****6****5****4****3****2****1****0****dangerous****safe**

EXPLORATORY DATA ANALYSIS, EDA

DATASET**ANALYSED FEATURES OF THE ASTEROID**

estimated min diameter (est_diameter_min)
estimated max diameter (est_diameter_max)
relative velocity (relative_velocity)
miss distance from Earth (miss_distance)
absolute magnitude (absolute_magnitude)

- **relative velocity**
- **absolute magnitude**



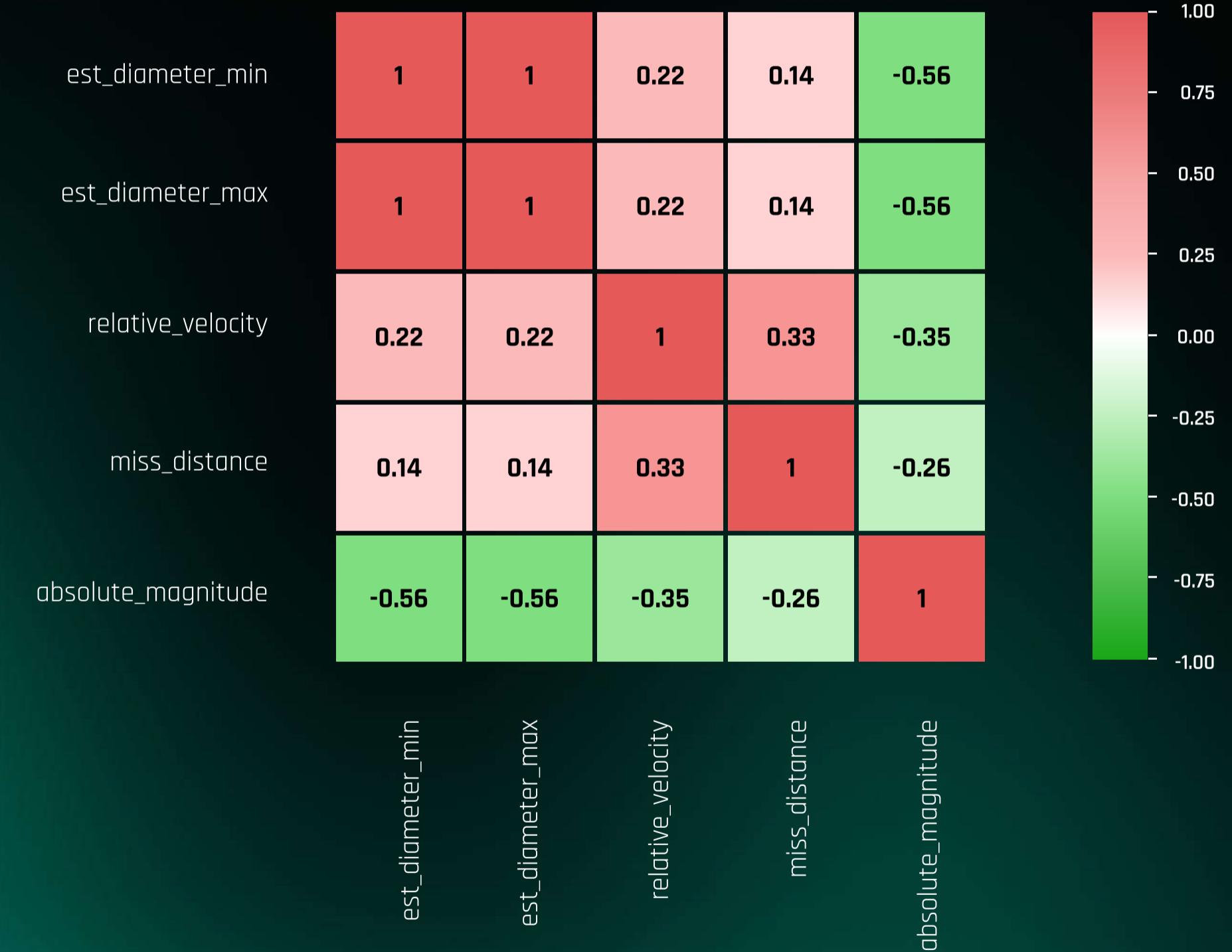
STUDY OF CORRELATIONS BETWEEN VARIABLES

CORRELATION MATRIX

Pearson correlation matrix

Both diameters (min and max) are correlated, so we will only use one of them to train the model. The max diameter will be used.

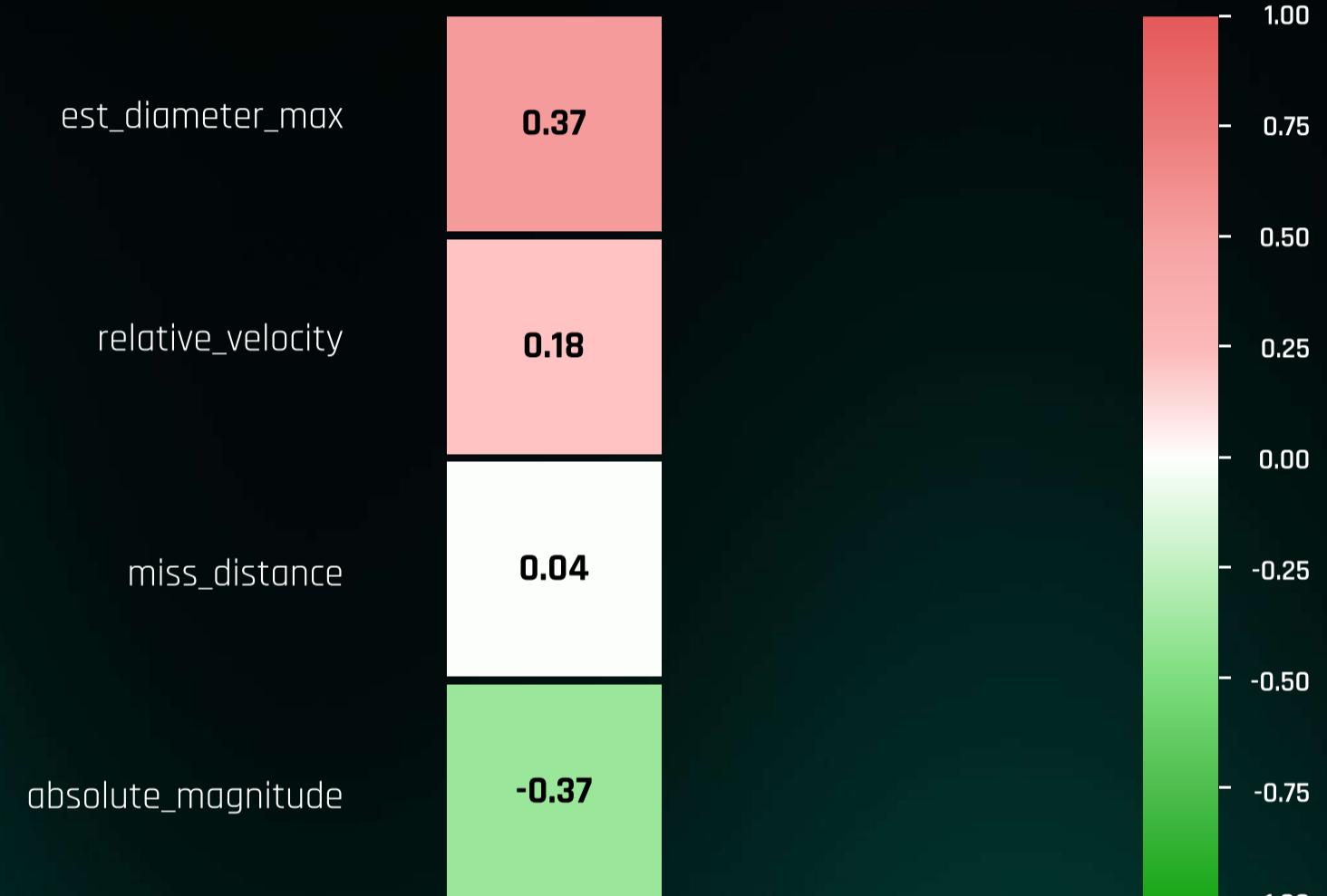
correlation matrix - all parameters



STUDY OF CORRELATIONS BETWEEN VARIABLES**CORRELATION MATRIX**

Spearman correlation matrix

Correlations between variables
and the potential asteroid
potentially hazardous to Earth.

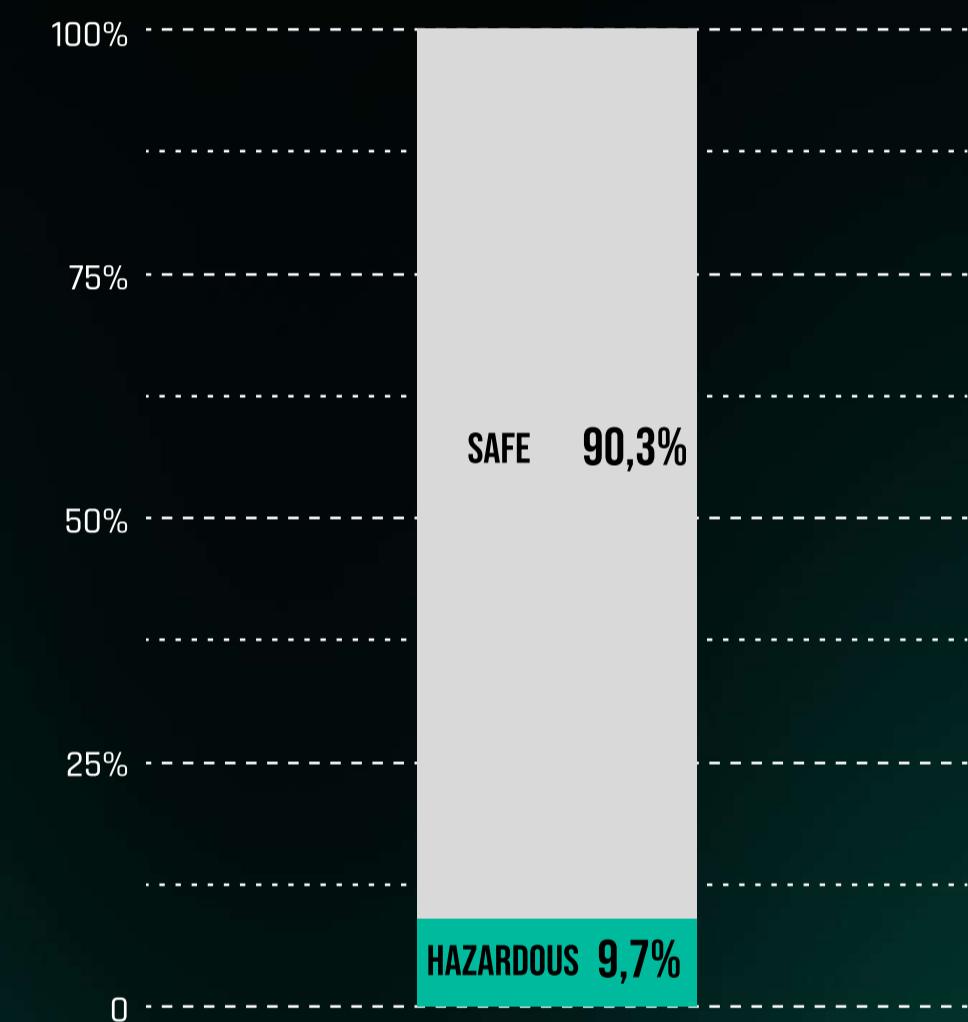
correlation matrix - selected parameters

CHALLENGES OF MODEL DEVELOPMENT**UNBALANCED DATA SET**

In the dataset, the number of records is unbalanced. Hazardous asteroids account for 9.7% of the cases.

This can be a challenge when training machine learning models.

is the asteroid hazardous?

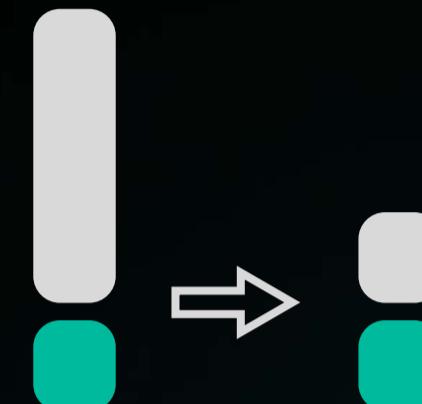


CHALLENGES OF MODEL DEVELOPMENT**BALANCING METHODS****undersampling**

- Near miss - deleting data between both groups
- Random sampler

oversampling

- Random sampler
- SMOTE - synthetic data for minority group

increase/decrease weighting**undersampling****oversampling****scaling weights**

MODELS

XGBOOST

XGBOOST INTRODUCTION

XGBoost is an optimized distributed gradient boosting library designed to be highly efficient, flexible and portable. It implements machine learning algorithms under the Gradient Boosting framework. XGBoost provides a parallel tree boosting (also known as GBDT, GBM) that solve many data science problems in a fast and accurate way.

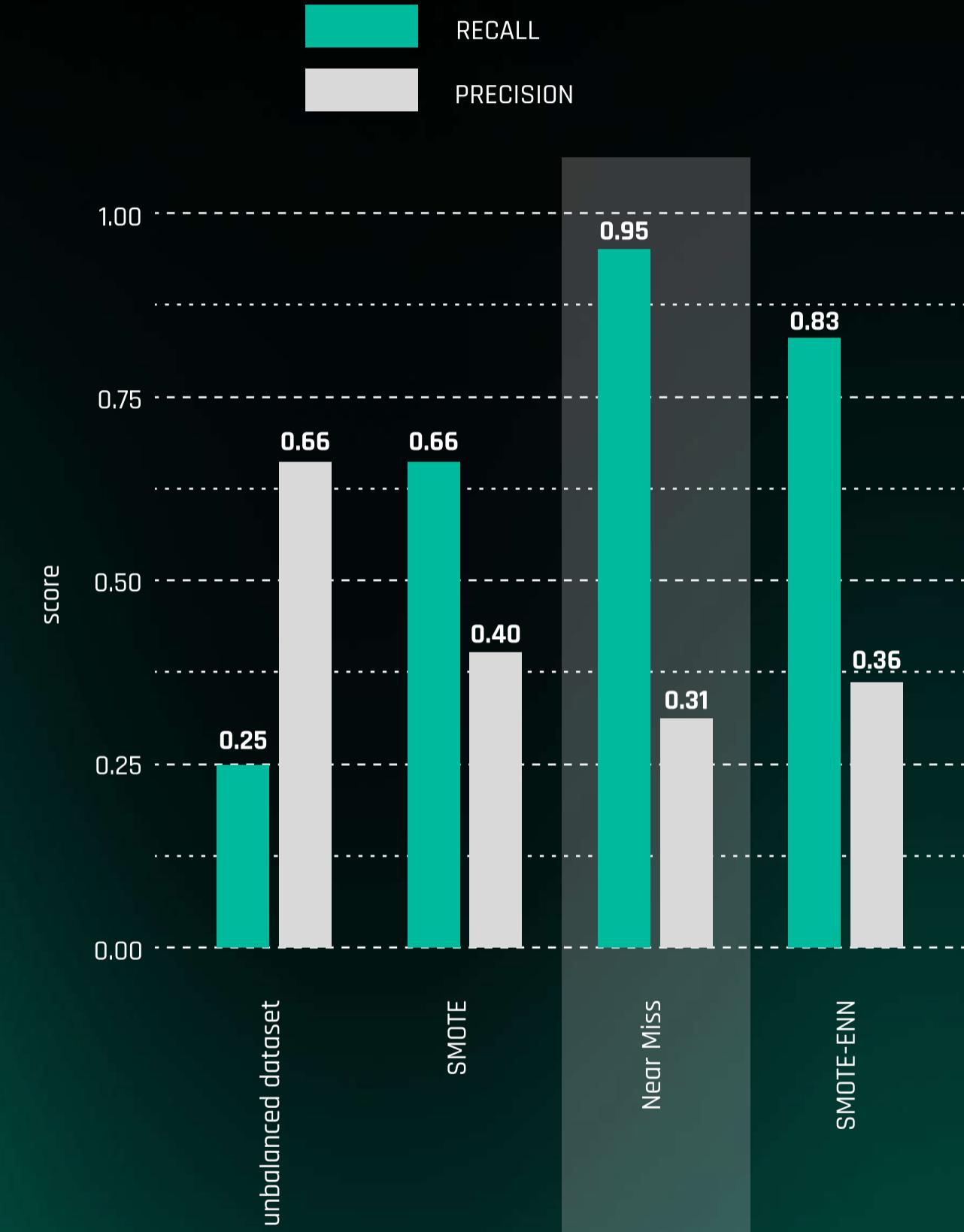


XGBOOST

IMPACT OF BALANCING TRAINING DATA ON THE XGBOOST MODEL

The effect of balancing training data on XGBoost model performance.

The model's task is not to miss any potentially hazardous asteroid. In order to meet this criterion, the decisive parameter for model evaluation will be the value of the recall parameter.



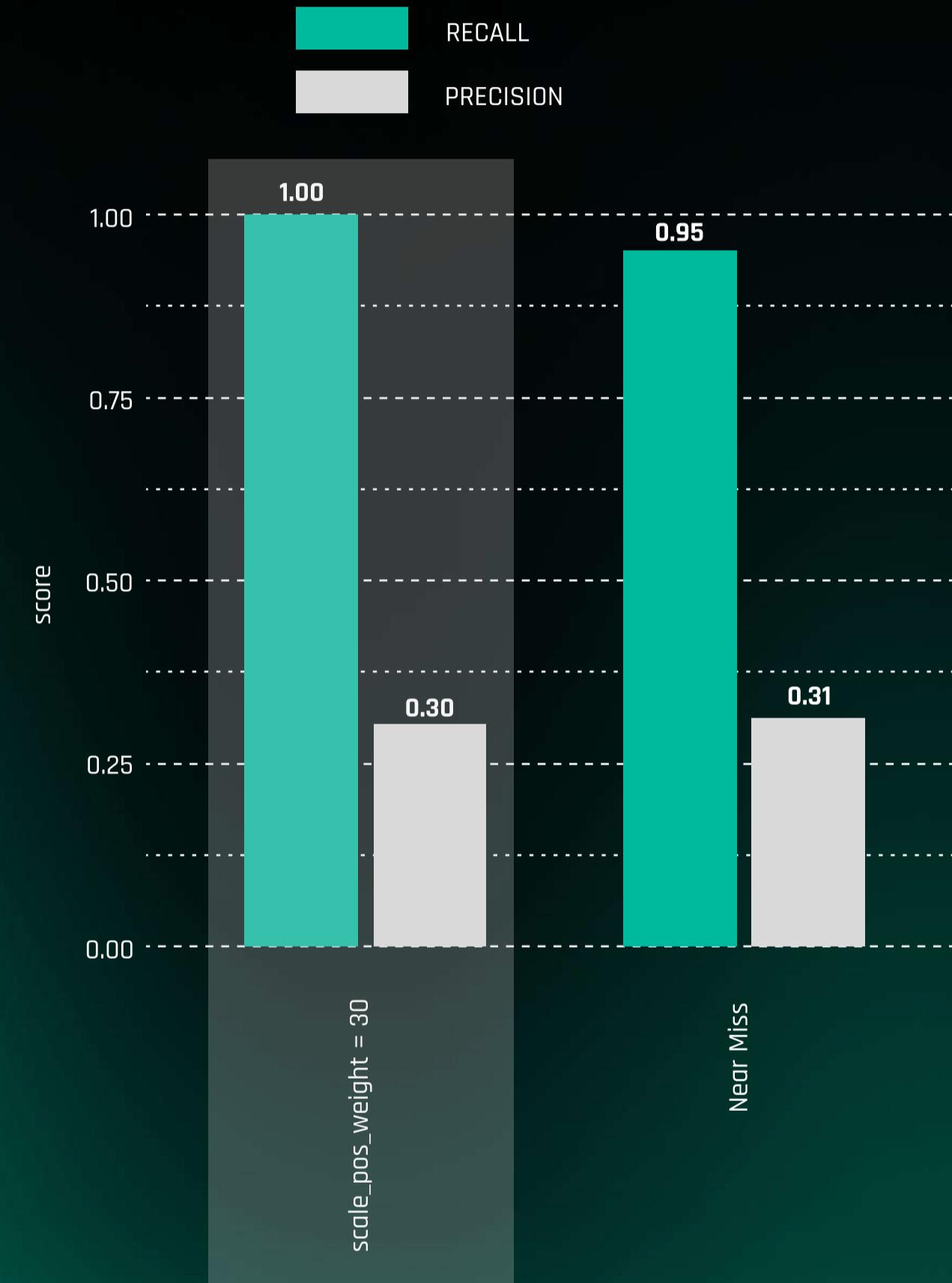
XGBOOST

OPTIMALISATION OF THE SCALE_POS_WEIGHT PARAMETER FOR THE XGBOOST MODEL

Optimisation of the scale_pos_weight parameter for the XGBoost model

The scale_pos_weight parameter:
Adjusts the error weights for the positive (minority) class to improve the prediction accuracy of this class.

$$\frac{\text{number of records classified as 0}}{\text{number of records classified as 1}} = 9.3$$



CHOOSING THE BEST MODEL

FINAL MODEL USED IN THE APPLICATION

XGBClassifier

n_estimators = 50,
max_depth = 3,
learning_rate = 0.1,
scale_pos_weight = 30

DEPLOYMENT

building a mock-up

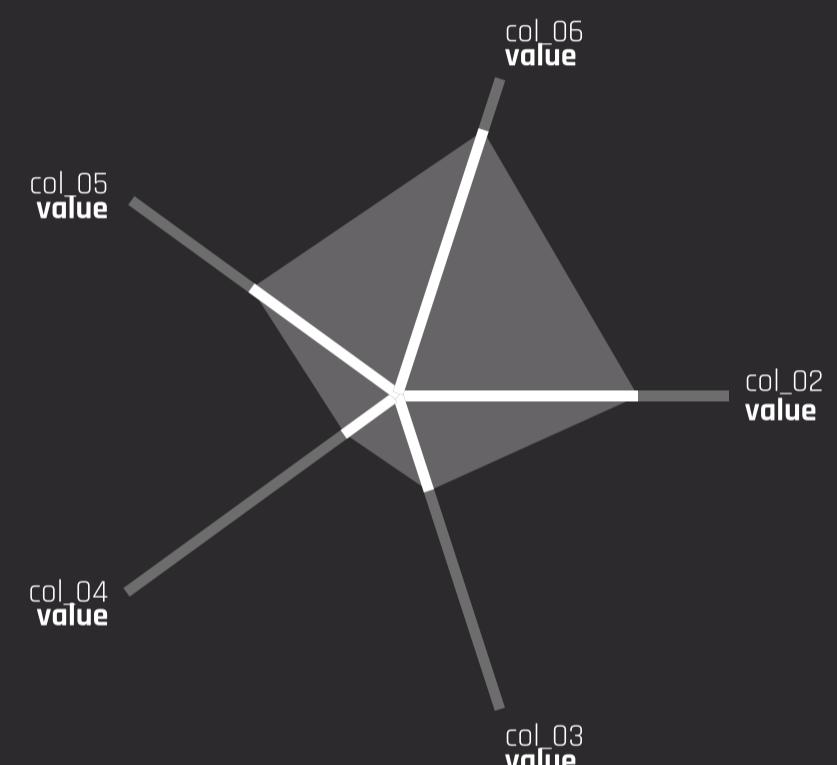
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model.py request.py app.py

```
51
52     id_letter = string.ascii_uppercase[len(database) % 26]
53     asteroid_id = f"2024 {id_letter}{len(database)+1}"
54
55     # Create a dictionary with the input data and the prediction
56     asteroid = {
57         'id': asteroid_id,
58         'est_diameter_min': est_diameter_min,
59         'est_diameter_max': est_diameter_max,
60         'relative_velocity': relative_velocity,
61         'miss_distance': miss_distance,
62         'absolute_magnitude': absolute_magnitude,
63         'hazardous': int(prediction)
64     }
65
66     # Add the dictionary to the database list at the first position
67     database.insert(0, asteroid)
68
69     # Keep only the last 5 entries
70     if len(database) > 5:
71         database.pop()
72
73
74
75     return render_template('index.html', prediction_text=f'{"0 / 1"}')
76
77 @app.route('/results', methods=['POST'])
78 def results():
79     data = request.get_json(force=True)
80     prediction = model.predict([np.array(list(data.values()))])
81     output = prediction[0]
82     return jsonify(output)
83
84 if __name__ == "__main__":
85     app.run(debug=True)
```

web application mock-up

input



model

model_name
model_type
stat_1
stat_2
stat_3

prediction

0 / 1

new_db

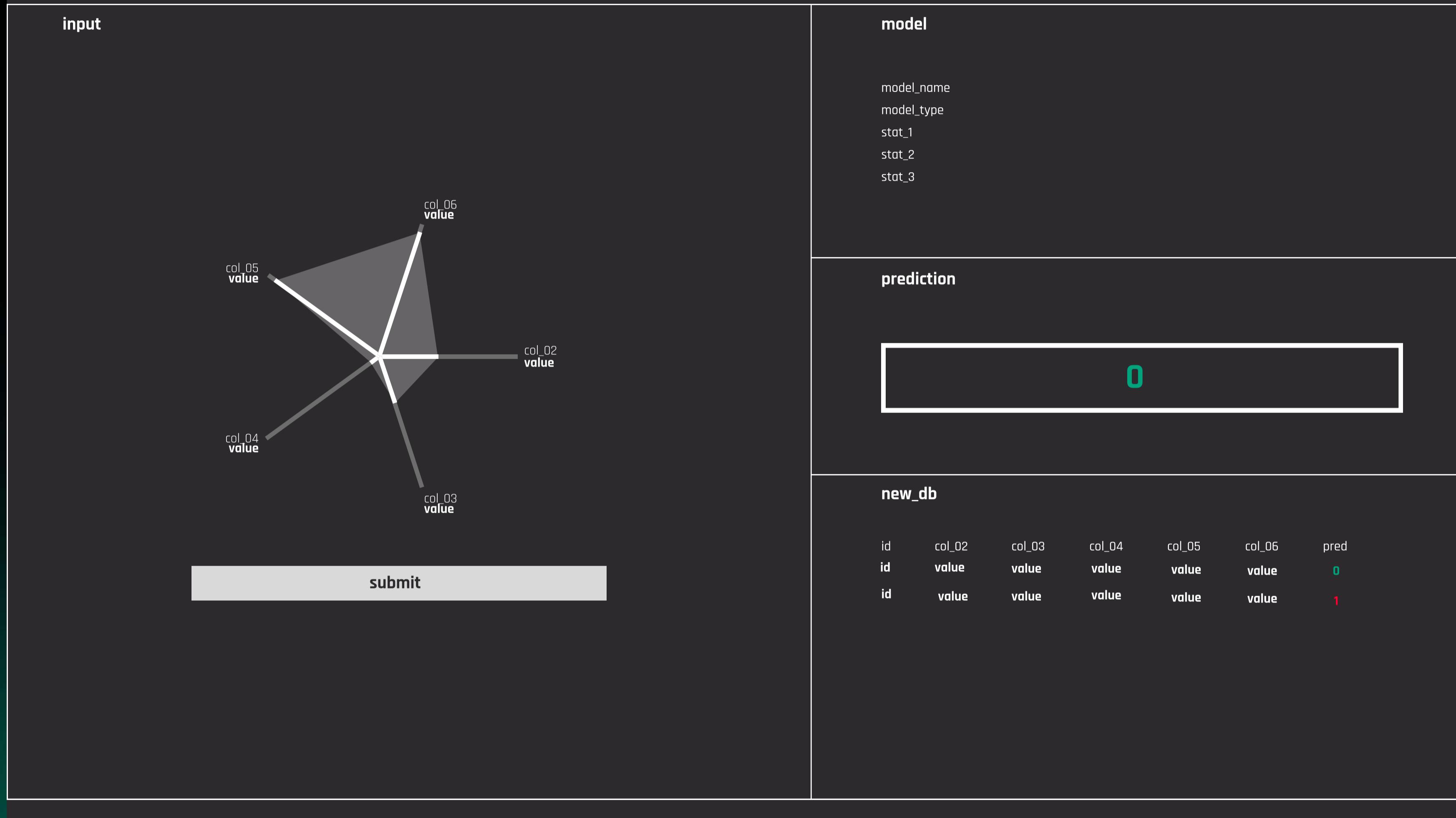
id col_02 col_03 col_04 col_05 col_06 pred

submit

building a mock-up

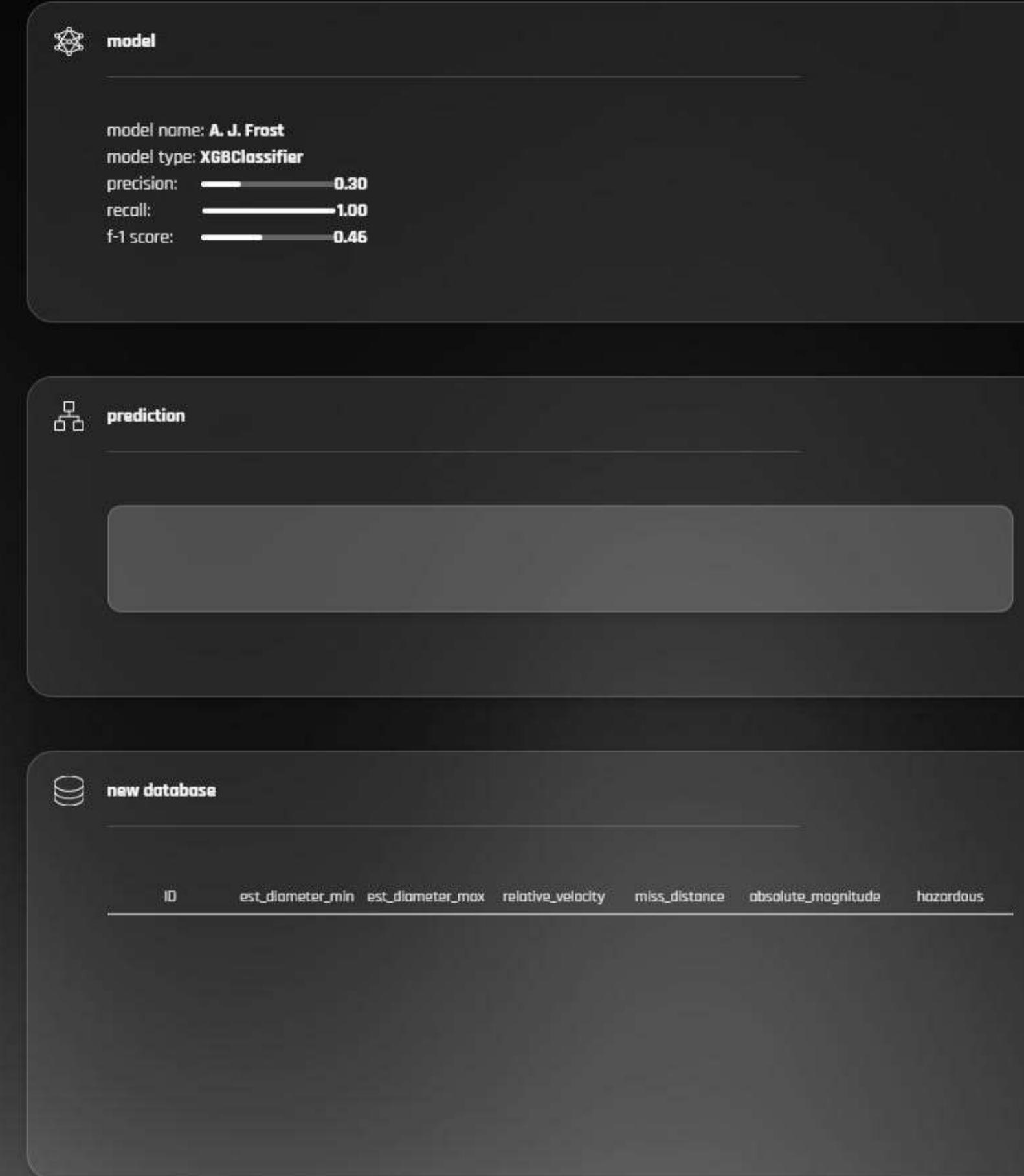
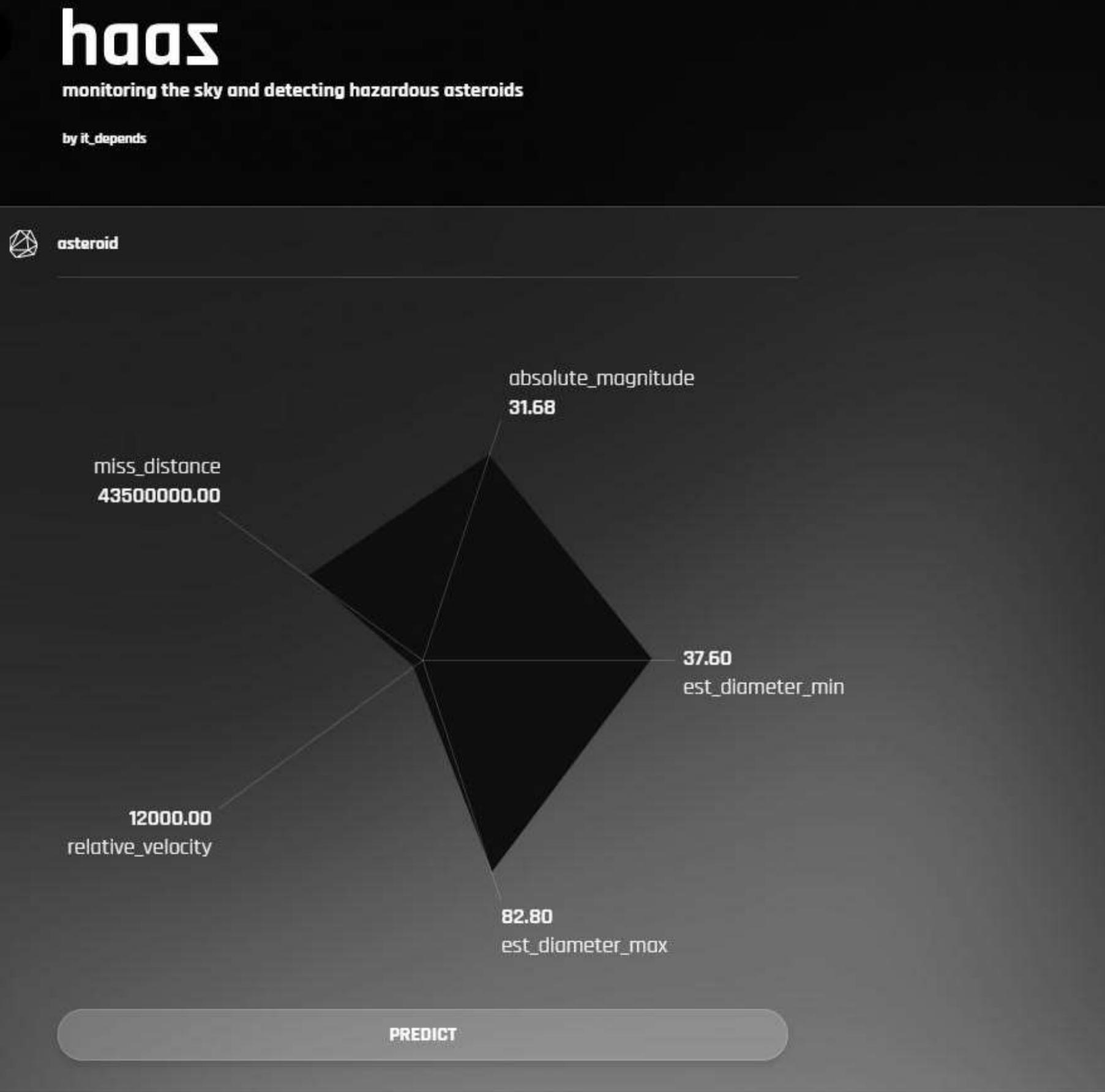
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web application mock-up



building a mock-up

it_depends



haas application



click the button below
to go to the github repository with the web application

REPO WITH WEB APPLICATION

bartłomiej brzostek

it_depends

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to identify potentially dangerous asteroids.