

# MACHINE LEARNING FOR THE CLASSIFICATION OF PLANETS

**THE BRIDGE**

Presented by: Paris Martínez  
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# OBJECTIVE

Search for habitable planets:

- Habitable Zone
- Planet Type
- Geology
- Temperature
- Atmosphere
- Orbit and Rotation





# Problem

## Machine Learning

- Supervised
  - Regression
  - Classification
- Unsupervised

# DATA ACQUISITION



# KEPLER MISSION



During just over nine and a half years in orbit, the Kepler space telescope observed more than half a million stars and discovered more than 2,600 planets.



# WEB SCRAPING

## Data:

- NASA Exoplanet Archive

## Modules:

- Selenium
- BeautifulSoup

## URL:

- <https://exoplanets.nasa.gov/discovery/exoplanet-catalog/>

# DATA EXPLORATION

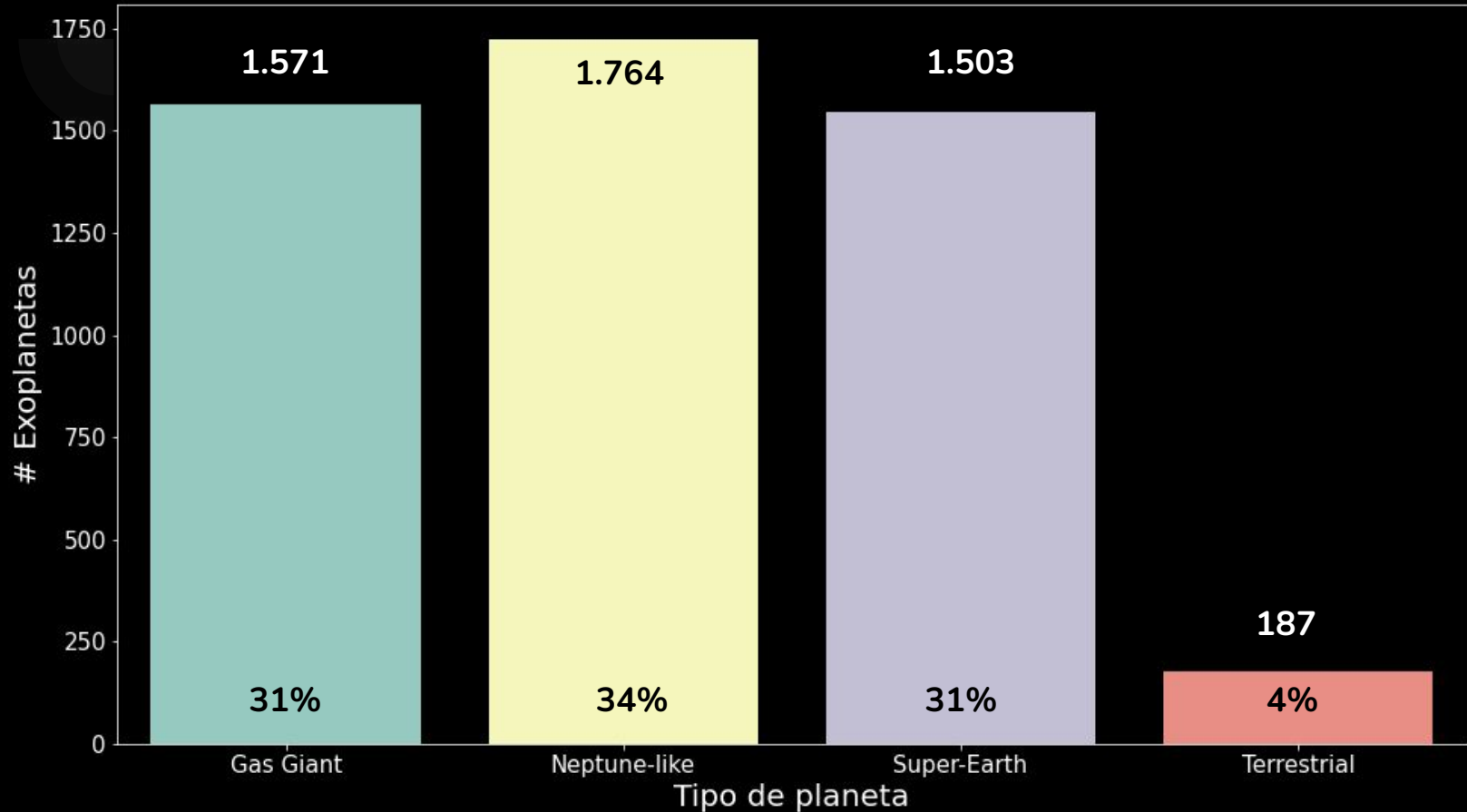




# CUMMULATIVE DISCOVERY PLANETS PER YEAR



# DISCOVERED EXOPLANETS



# DATA PREPARATION



- Transform Jupiter mass, radius to Earth
- Drop Categorical columns: **names, detection\_method**
- Drop Unknowns targets: **2 stars, 5 planets**
- Encode target column
- Change Infinite with mode in orbital days: **1 planet**
- Fill NaN with KNNImpute: **337 NaN**

**177 eccentricity, 117 stellar\_magnitude, 18 mass, 16 distance, 9 radius**



# FEATURE ENGINEERING



# CREATE NEW FEATURES

## Density and Log Density

Masa de la Tierra:  $5,972 \times 10^{24}$  kg

Radio de la Tierra: 6.371 km

Density (Kg/m<sup>3</sup>) =  $(\text{mass} * (5.972 * 10^{24})) / (4/3 * \pi * ((\text{radius} * 6371000) ** 3))$

Log transform features (mass, radius, density, distance and orbital days)

# Features      Log Features

distance

log\_distance

mass\_E

log\_mass

radius\_E

log\_radius

density

log\_density

orbital\_days

log\_orbital

discovery\_year

stellar\_magnitude

eccentricity



# Target

Planet Encode:

- Gas Giant: 3
- Neptune-like: 2
- Super Earth: 1
- Terrestrial: 0





# SELECTION MODEL

Random Forest

Cat Boost

Decision Tree Classifier

Lightgbm

Multi Layer Perceptron

Voting(RF+XGB)

XGBoost

Ada Boost

Logistic Regression

Gradient Boosting

Deep Learning Model

# Deep Learning

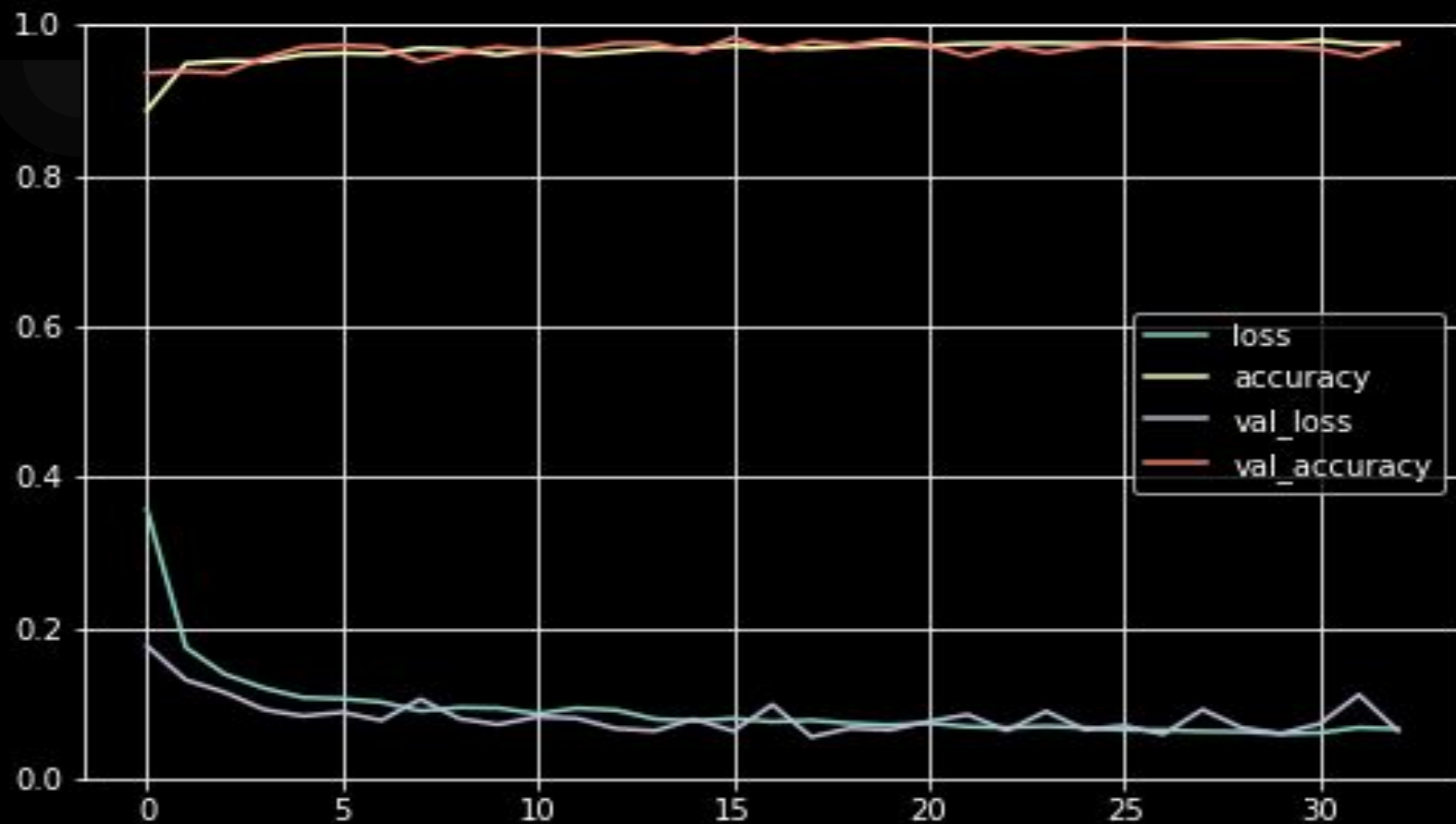
## Accuracy: 97,5%

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 300)	1800
dense_1 (Dense)	(None, 100)	30100
dense_2 (Dense)	(None, 4)	404
Total params: 32,304		
Trainable params: 32,304		
Non-trainable params: 0		

32/32 [=====] - 0s 1ms/step - loss: 0.0800 - accuracy: 0.9752  
[0.0800432339310646, 0.9751983880996704]

# TRAINING





# VOTING (RF+XGB)

## Random Forest

random\_state = 42

n\_estimators = 500

max\_leaf\_nodes = 16

## XGBoost

random\_state = 42



# Results

	Accuracy
Voting(RF+XGB)	0.993056
XGBoost	0.991071
Random Forest	0.990079
CatBoost	0.990079
Lightgbm	0.990079
Gradient Boosting	0.989087
Decision Tree	0.983135
AdaBoost	0.981151
MLP	0.957341
Logistic	0.739087

## Terrestrial Planets

Total: 42

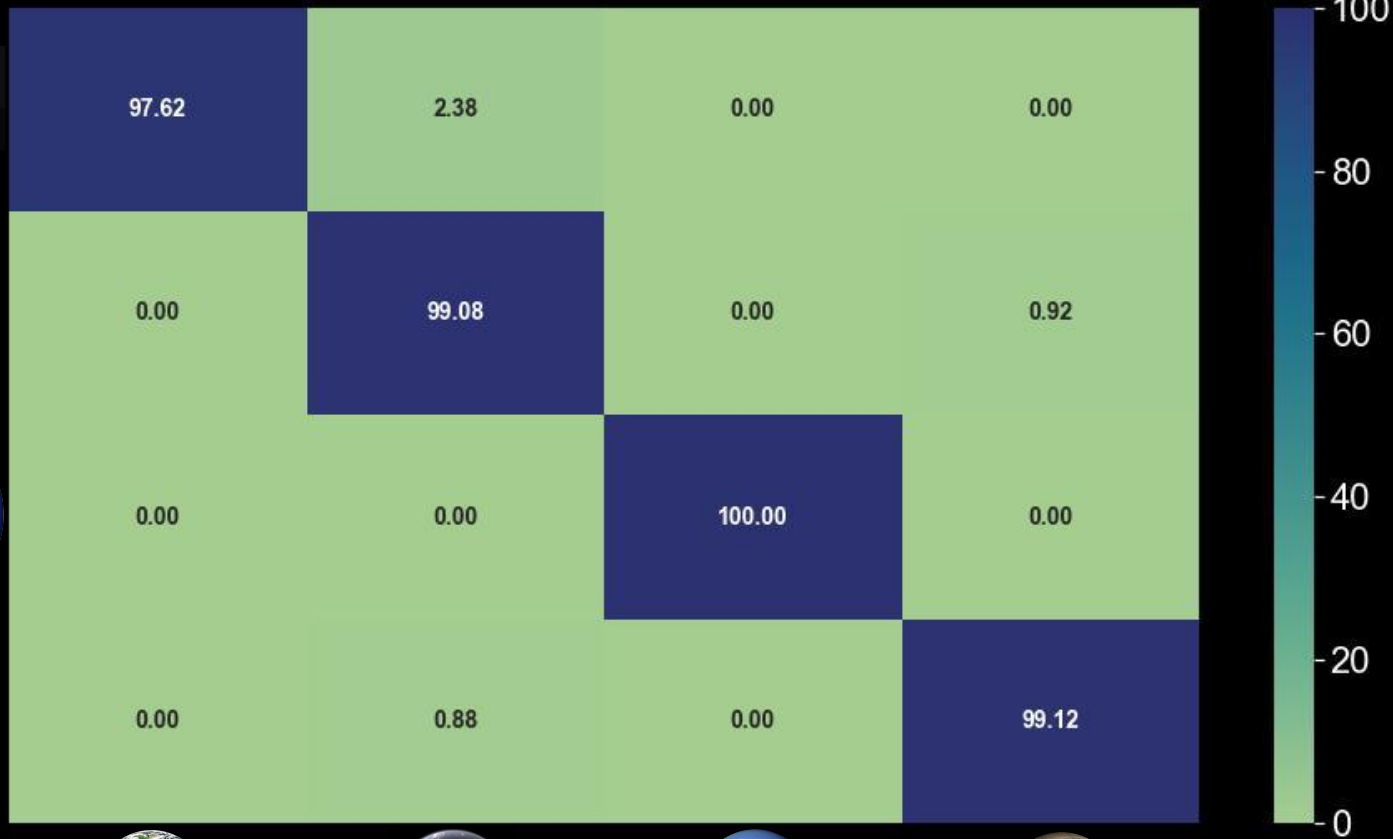
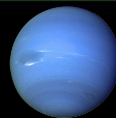
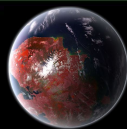
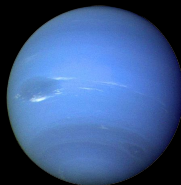
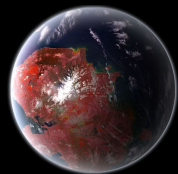
Classified as Terrestrial: 41

Classified as Super-Earth: 1

Classified as Neptune-Like: 0

Classified as Gas Giant: 0

# CONFUSION MATRIX

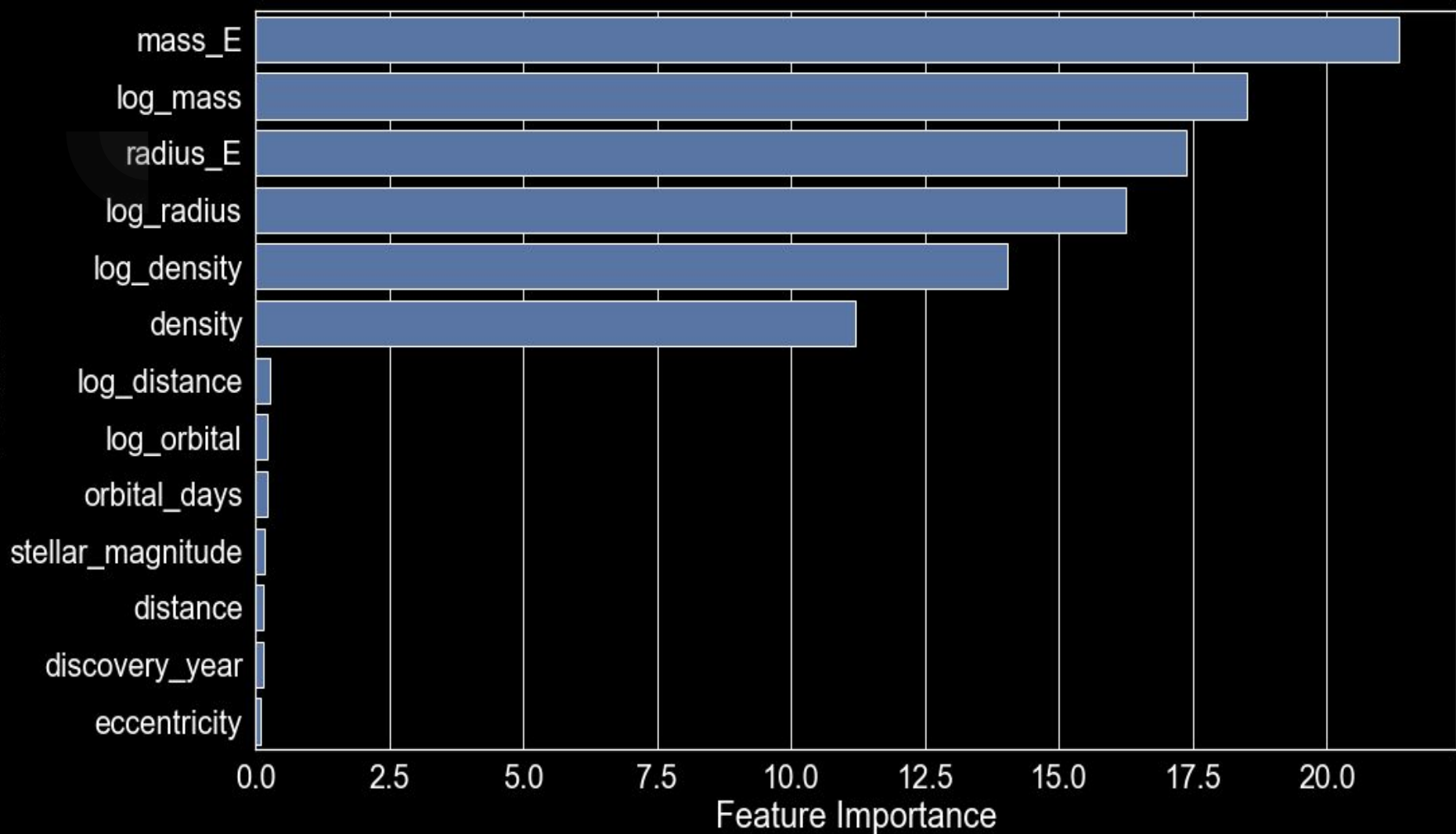


# Feature importance

feature_importance	
mass_E	21.357954
log_mass	18.516024
radius_E	17.386675
log_radius	16.235138
log_density	14.025148
density	11.203930
log_distance	0.270783
log_orbital	0.224094
orbital_days	0.209926
stellar_magnitude	0.181263
distance	0.143935
discovery_year	0.138453
eccentricity	0.106668

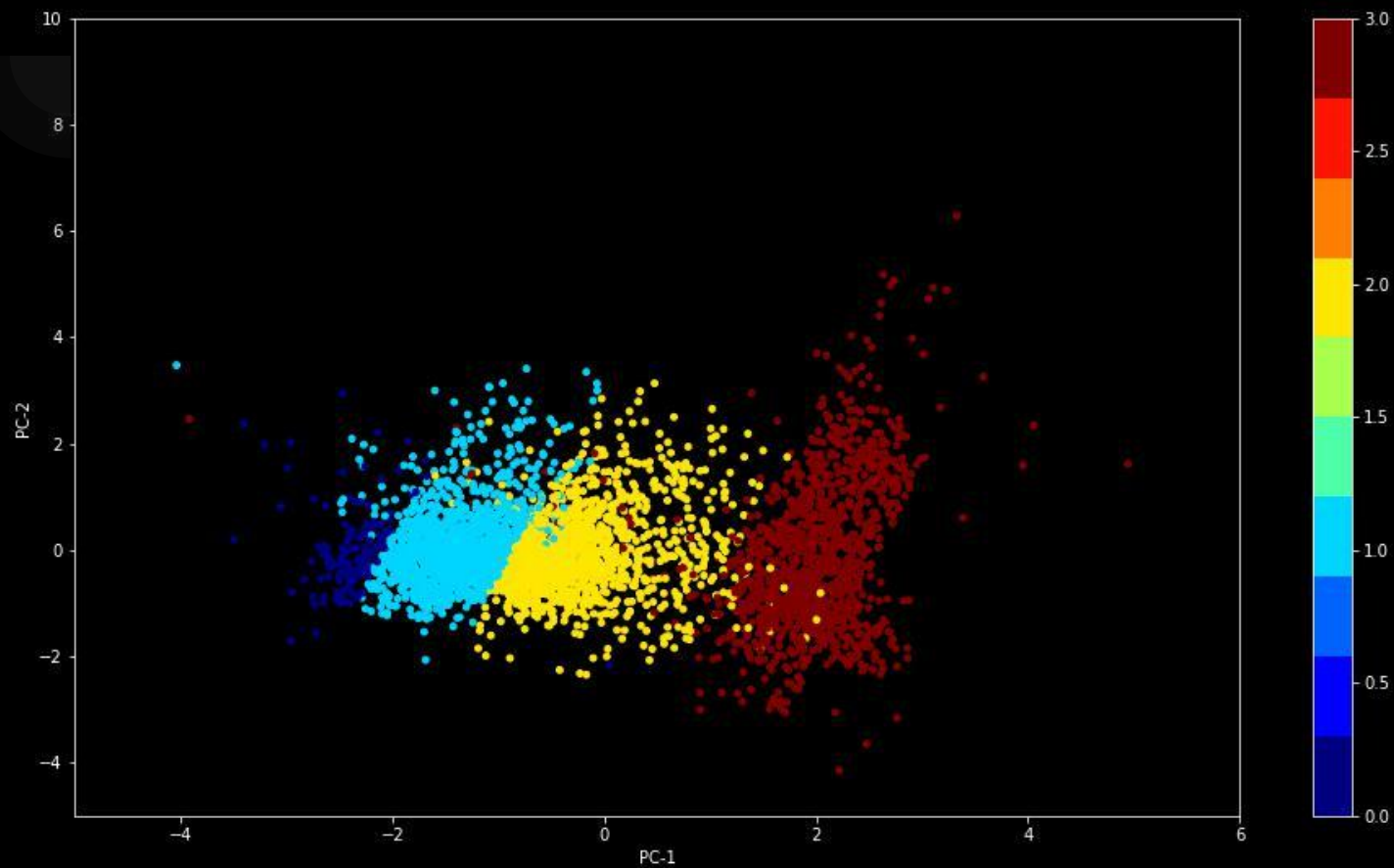
1. Mass
2. Radius
3. Density

Features





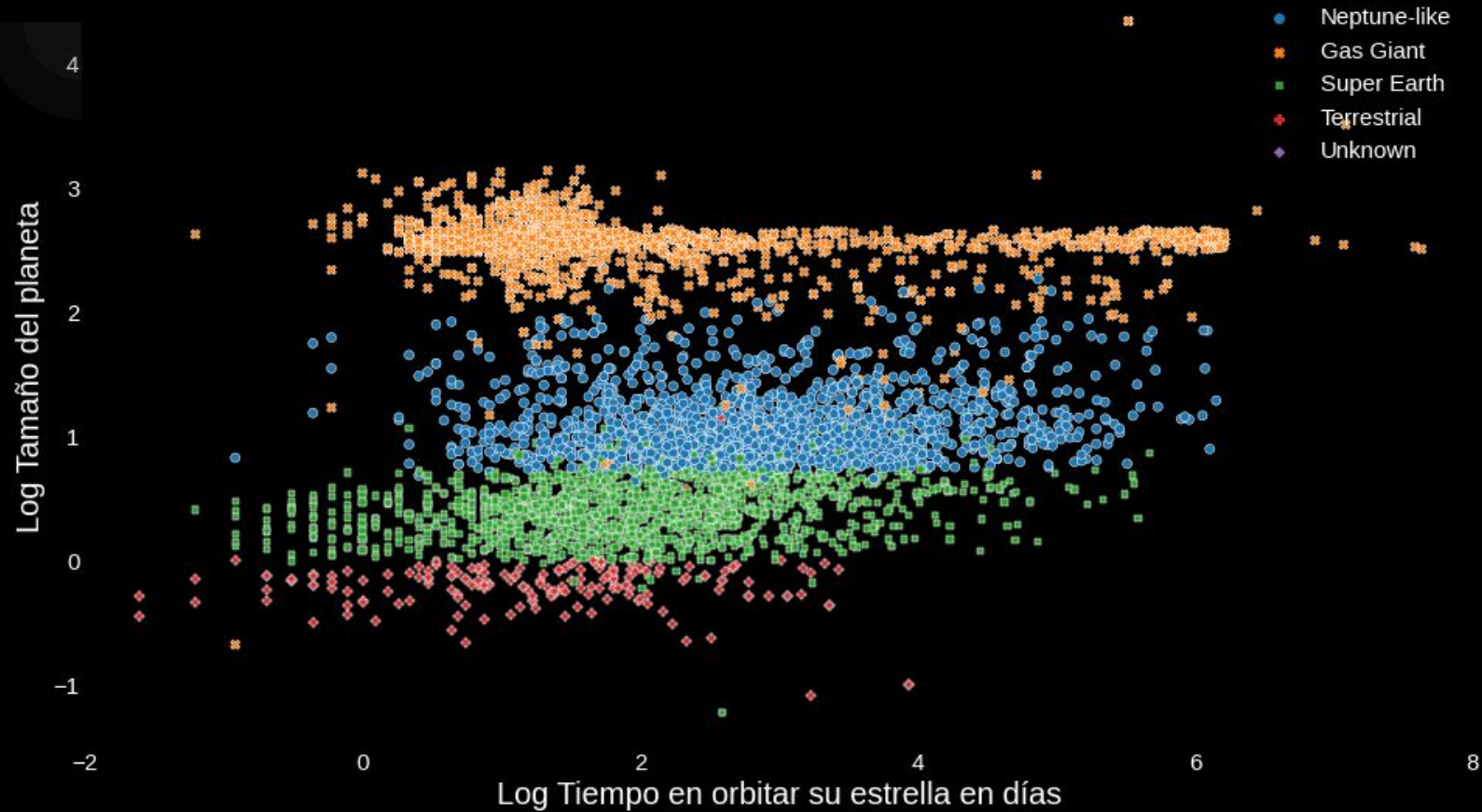
# PRINCIPAL COMPONENT ANALYSIS



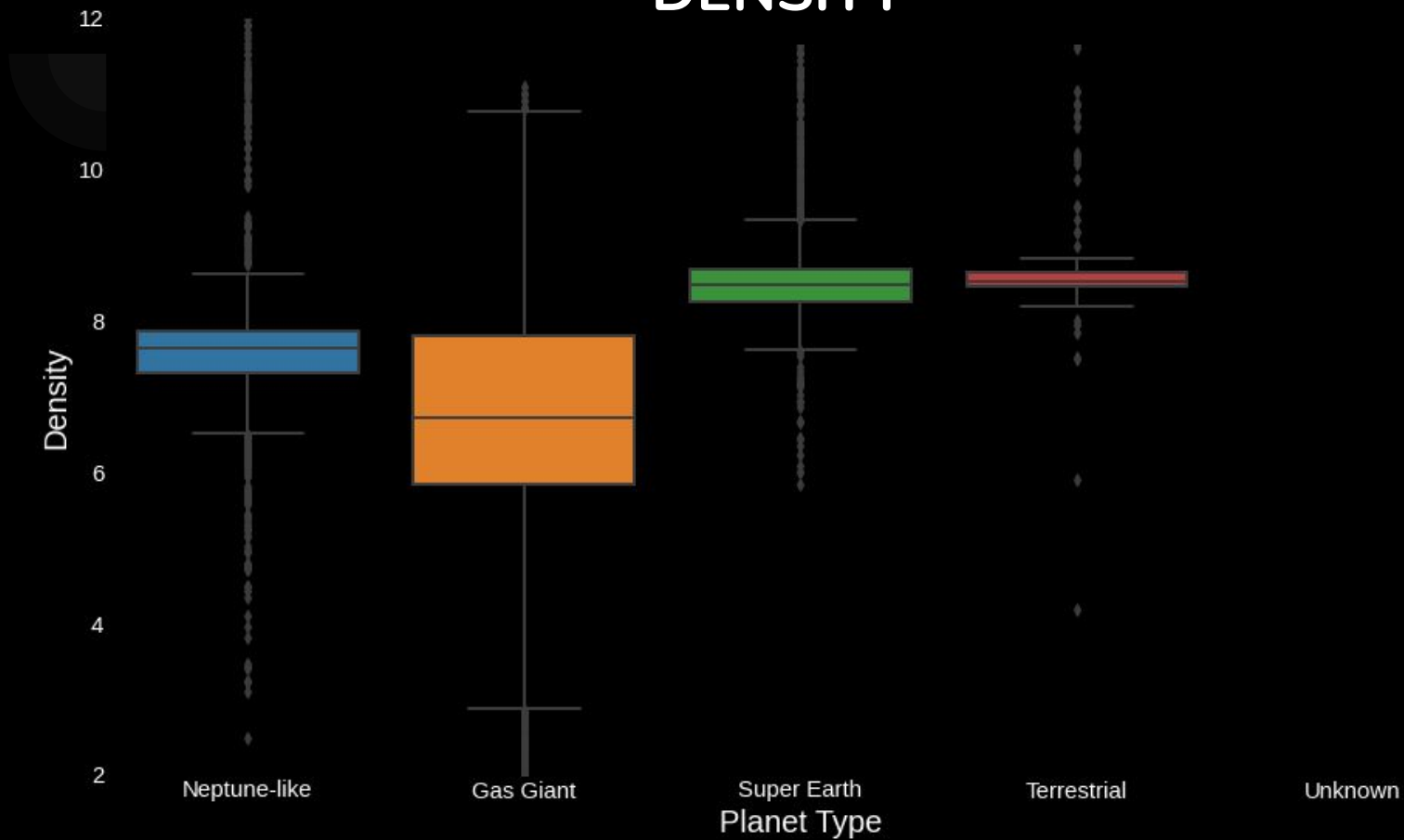
# MASS



# RADIUS



# DENSITY





# CONCLUSION

Starting from a few features (massa, radius, orbital days, distance, eccentricity, stellar magnitude and discovery year) obtained by **web scraping**

Preparing the data and doing some **Feature Engineering**,

We have solved the **Classification Problem** with **Supervised Machine Learning** and obtained an **Accuracy of 99.3%** in the prediction of planet types





# COMPARATIVE

Works of Rincón / Johans González:

- Models: Decision Tree
- Accuracy: 94,83% / 97,31%
- Observations: 3.286 exoplanets / 1.672 exoplanets

My results:

- Model: Voting(RF+XGB)
- Accuracy: 99,31%
- Observations: 5039 exoplanets



# FUTURE LINES OF DEVELOPMENT

- Search and detection of exoplanets.
- Classification of habitable and superhabitable planets

# SUPERHABITABLE PLANETS



Comparison between the size of Kepler-442b ( $1.34 R_{\oplus}$ ) and Earth (right).