



MACHINE LEARNING WORKSHOP

DAY 03

TOPIC NAME: EXOPLANET HUNTING USING ML
INSTRUCTORS NAME: PRAGYA AND SAHAS



ROADMAP OF WHAT WE WILL LEARN TODAY:

What is Machine
Learning

How to process data

Train and make your
predictions !!

How to hunt for exoplanets

How to visualize
the data



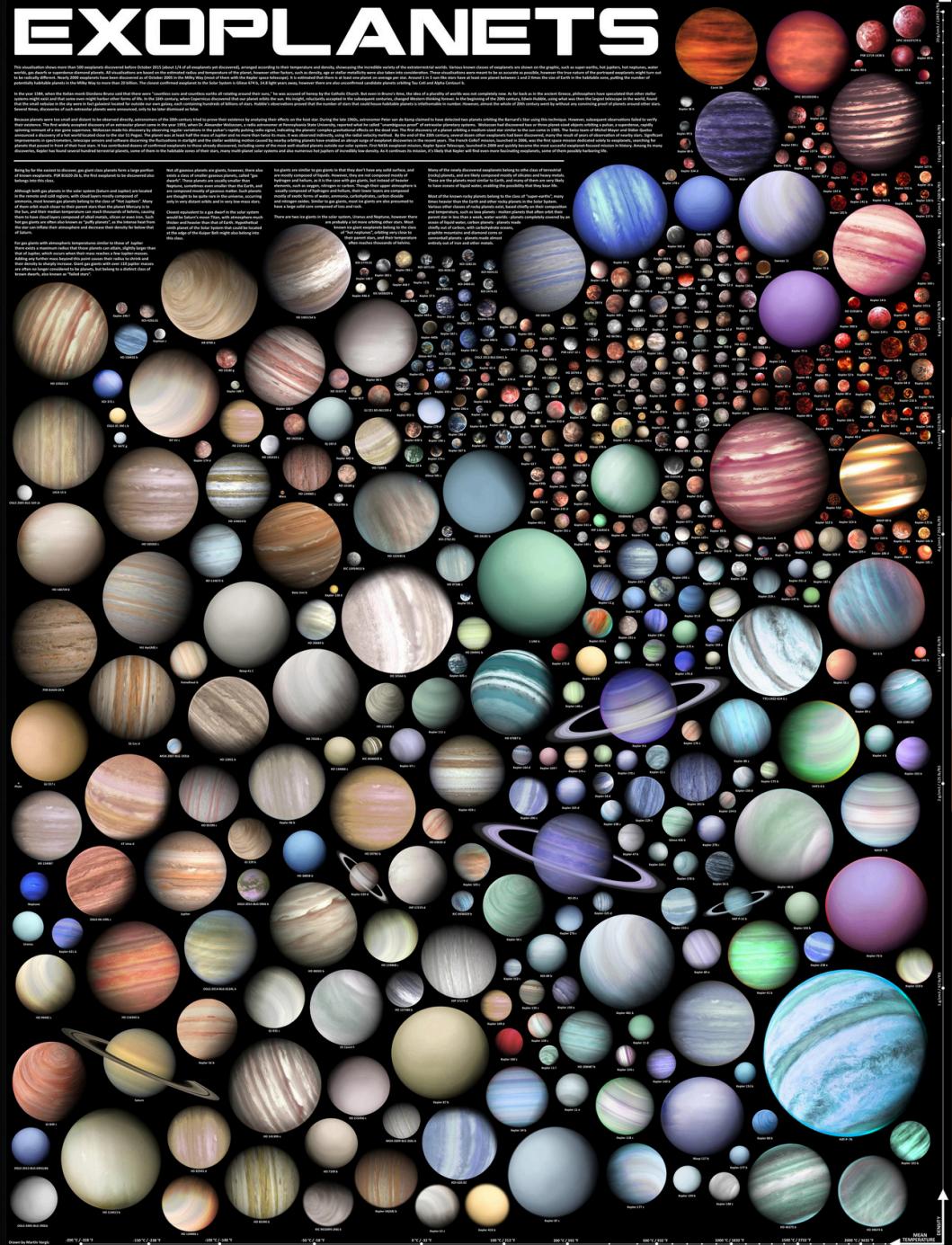
LIBRARIES WE NEED,

- Numpy
- Pandas
- Matplotlib
- Scikit-learn
- Lightkurve
- Scipy
- Imlearn
- XGBoost

WHY WE USE ML IN EXOPLANET DETECTION:



EXOPLANETS



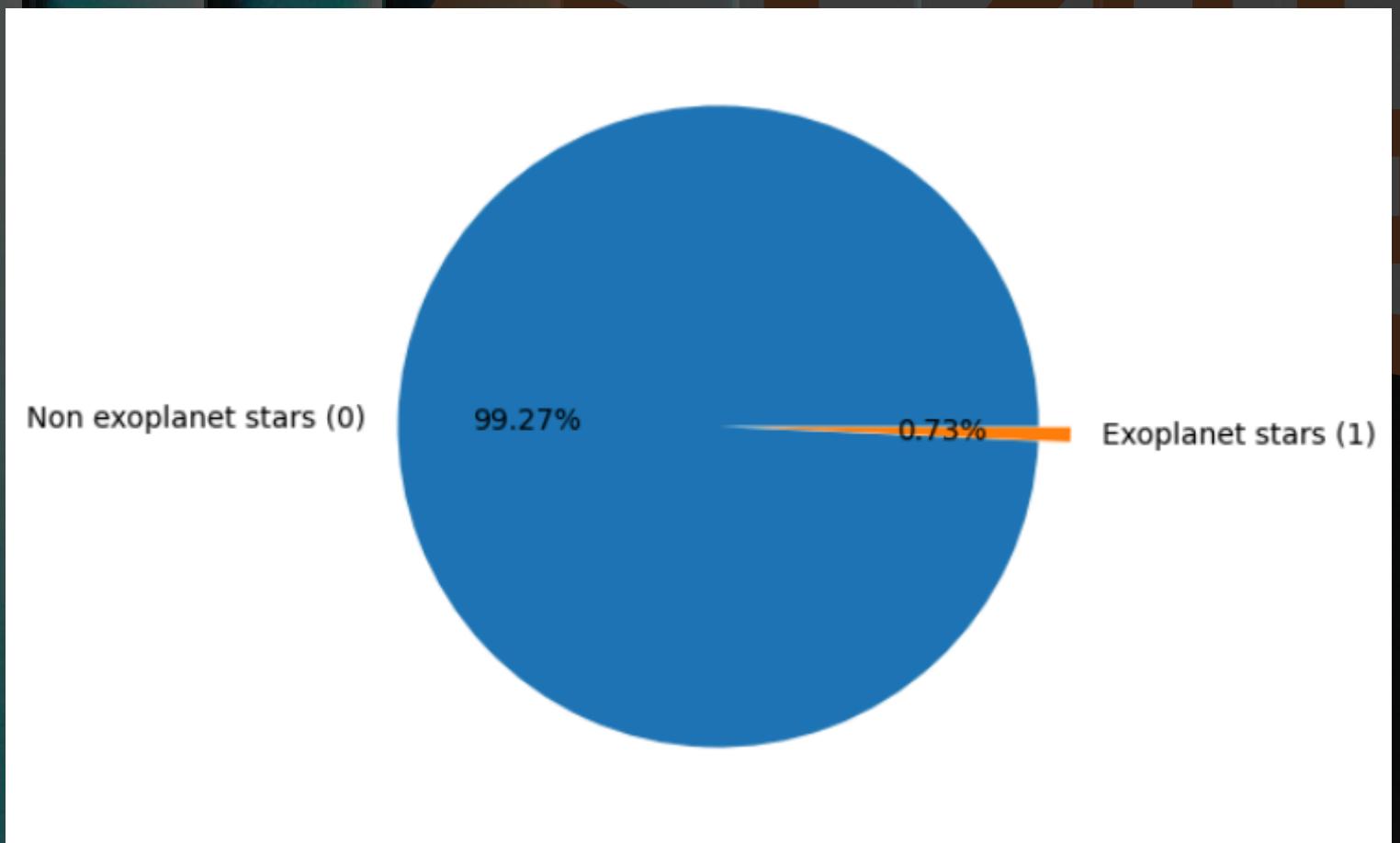
Machine learning helps us find new planets outside our solar system by quickly and accurately analyzing huge amounts of space data that would take humans a very long time to go through

CODE SNIPPETS

EXPLORATORY DATA ANALYSIS



```
[ ] label_counts = train_data['LABEL'].value_counts()  
labels = ['Non exoplanet stars (0)', 'Exoplanet stars (1)']  
explode = [0,0.1]  
plt.figure(figsize=(5,5))  
plt.pie(label_counts, labels=labels, autopct='%.1f%%', explode=explode)  
plt.show()
```



L SEDS
class imbalance!
solve using SMOTE

MODEL CODE



▼ Model Training - Using XGBoost Classifier

```
[ ] from xgboost import XGBClassifier
from sklearn.metrics import classification_report
from imblearn.over_sampling import SMOTE
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA

[ ] # train
xgb = XGBClassifier(use_label_encoder=False, eval_metric='logloss')
xgb.fit(X_train_feat, y_train)

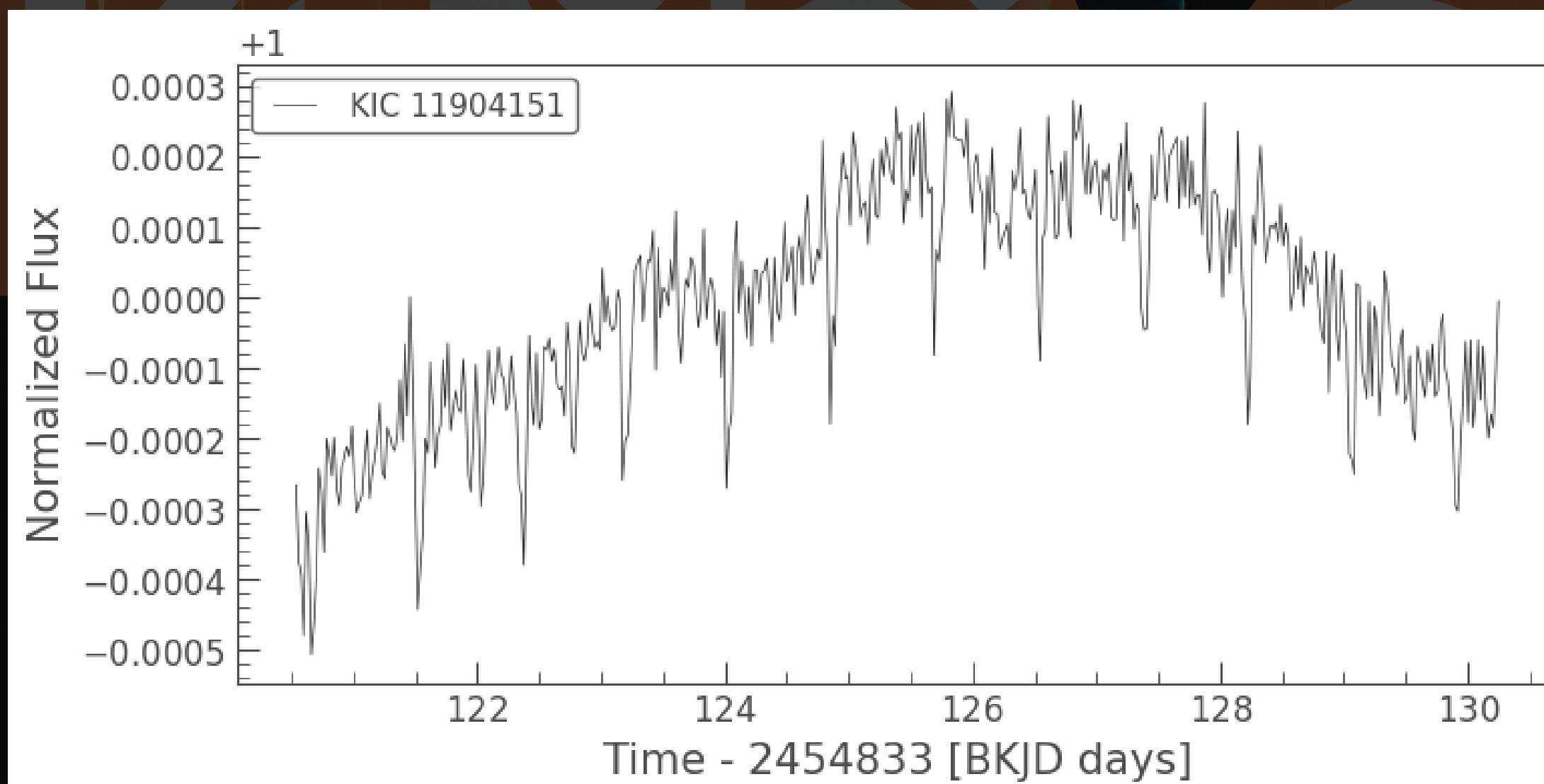
# eval
y_pred = xgb.predict(X_test_feat)
print(classification_report(y_test, y_pred))
```

IMPORTING LIGHTKURVE



```
[ ] !pip install lightkurve
from lightkurve import search_lightcurve
import matplotlib.pyplot as plt
```

```
▶ search_result = search_lightcurve("Kepler-10", author="Kepler", cadence="long")
lc = search_result.download()
lc = lc.normalize().remove_nans()
lc.plot()
```





PROBLEM STATEMENT 1:

Stellar Object Classification using Sloan Digital Sky Survey DR17

Dataset: <https://www.kaggle.com/datasets/fedesoriano/stellar-classification-dataset-sdss17>

Description: Develop a ML model to classify celestial objects into stars, galaxies, and quasars based on their spectral characteristics. Try to cover EDA, feature extraction, Model training, classification reports.

Bonus:

For objects classified as galaxies, develop a secondary model to distinguish between elliptical, spiral, and irregular galaxy types using photometric and structural parameters.



PROBLEM STATEMENT 2:

CLASSIFICATION OF TRANSIENT AND NON TRANSIENT OBJECTS BASED ON LIGHT CURVES

Dataset: <https://github.com/MachineLearningUniandes/MANTRA>

Description: Develop a ML model to classify transient object and non transient object light curves with any suitable algorithm. Try to cover EDA, feature extraction, Model training, classification reports.

Bonus:

The dataset can be used for both binary classification (above) and multi-class classification (8 classes - Some of the classes included in the dataset are: supernovae, cataclysmic variables, active galactic nuclei, high proper motion stars, blazars and flares.). Try developing a multi-class classification ML model that can classify transient astronomical events based on the light curves.

SUBMISSION GUIDELINES



- The codes must be shared as a Google Colab link.
- The notebook should be named according to the format - 'yourRegNo_yourName' Eg. 24BEC1234_JohnDoe
- Make sure you enable public access while sharing the notebook.
- Practices like adding comments or markdowns for better understanding is highly encouraged.