

An aerial photograph of a large, irregularly shaped island covered in dense, vibrant green forest. The island is surrounded by calm water that reflects the warm, golden light of a sunset or sunrise. The sky is not visible, but the water's surface is a mix of deep blue and warm orange tones. The forest appears thick and healthy, with many small trees visible from above.

# Amazon Fires

Mid-project  
DA Bootcamp Ironhack

OCT, 2022 | Adela & Bruno

# Index

---

1 / Amazon Context

2 / Problem

3 / Dataset Information

4 / Analysis

5 / Predictive Model

6 / Conclusions

7 / Recommendations for future analysis

# 1/ Amazon Forest Context

---

# **6.7 million square kilometers among 9 countries**

or 28 times the size of the UK

# **80K species of flora**

and an estimated 400 billion trees standing

# **6% of the world's oxygen production**

act as a carbon sink, absorbing large amounts of carbon dioxide from the atmosphere (Leman J., 2021),

# **home for 30 million people**

including almost 3 million indigenous people



## 2/ Problem

---

The rate at which the Amazon is deforested is a global concern, as it plays a key role in the environment and the atmosphere, not only in South America but also at a global scale.

Different studies indicate that deforestation has meant the **loss of 20% of the forest**, and if it continued at this rate by 2050, 40% of it would have disappeared (H Ter Steege, 2019). Between 36% and 57% of the Amazonian tree species could be in danger.

### 3/ Dataset information

---

# Characteristics of the fires in the Amazon forest between 1998 and 2013,

# Shape 1513059 rows × 15 columns,

# Some of the main features:

- **C:** Carbon emissions in g C/m<sup>2</sup>,
- **NPP:** Net primary production, carbon retained in g C/m<sup>2</sup>,
- **Fire spots:** number of spot fires ignited outside the perimeter of the main fire.
- **Burned area:** area affected by each fire in ha.
- **Latitude/Longitude:** geographical coordinates of the fires

## 4/ Analysis

---

# Did the number of fires increased?

# Do carbon emissions correlate with the size of the burned areas?

# What is the capacity of the affected areas to retain Carbon?

# What are the most affected areas and how are they distributed in the territory?

# How the Amazon fires behave throughout the months of the years?

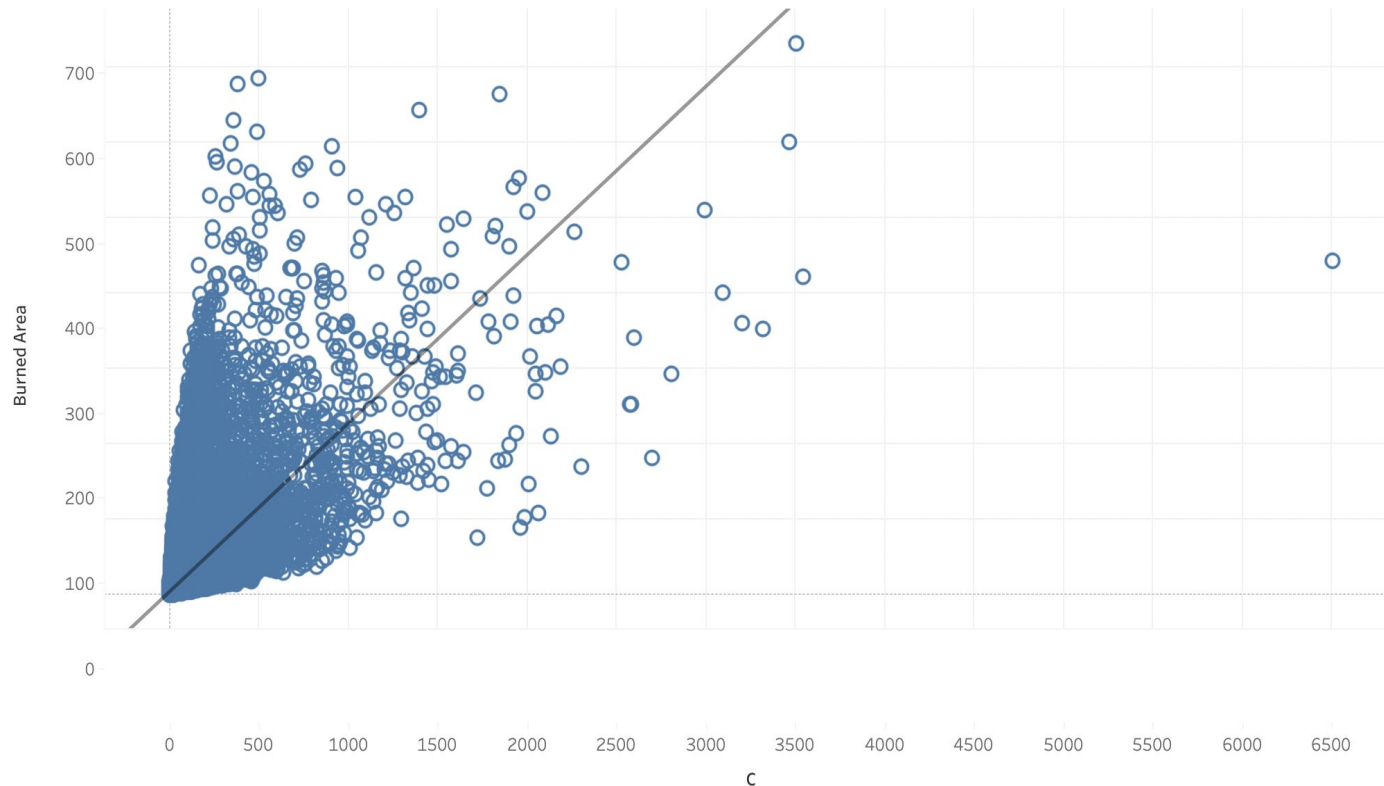
# Number of fires rapidly decreased and then stabilized

---



- Peak with more than 40K fires/year in 1998/99
- From 2000 to 2012, average of to 18.5K fires/year

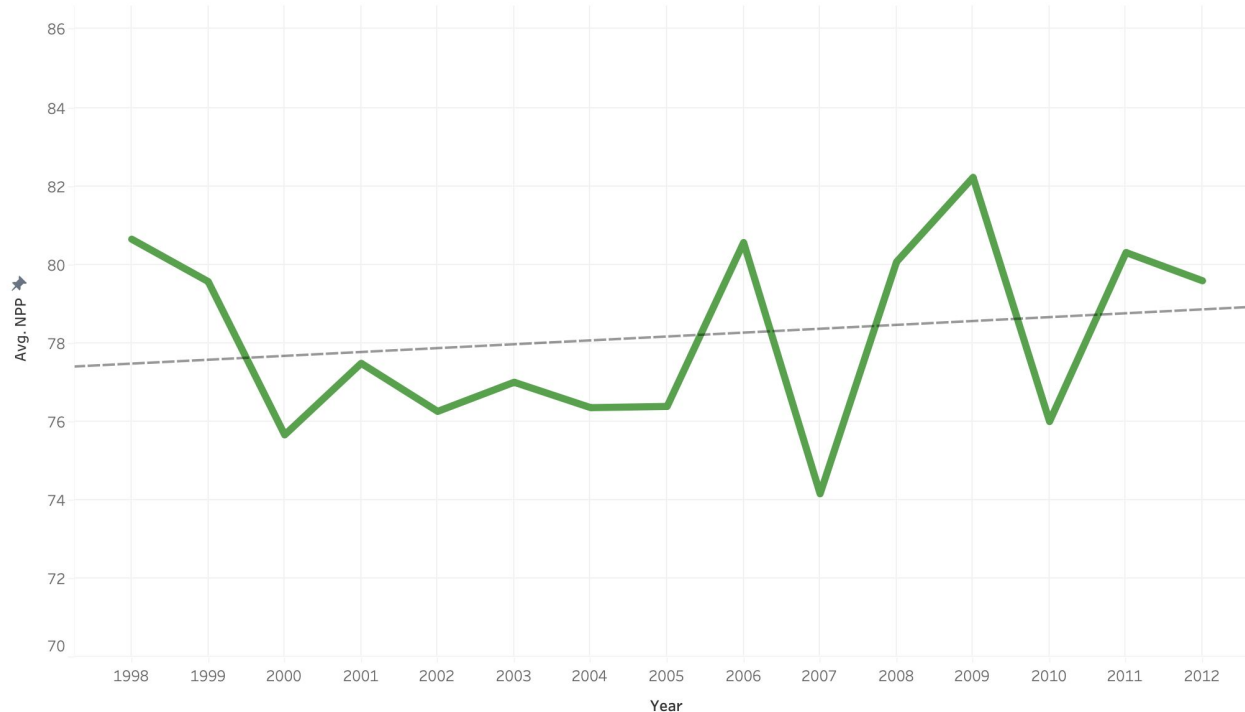
# Moderate correlation between Carbon Emission and Burned Area



- Correlation of 0.6 between both features
- Carbon emissions don't necessarily increase in the same rate if burned area is bigger



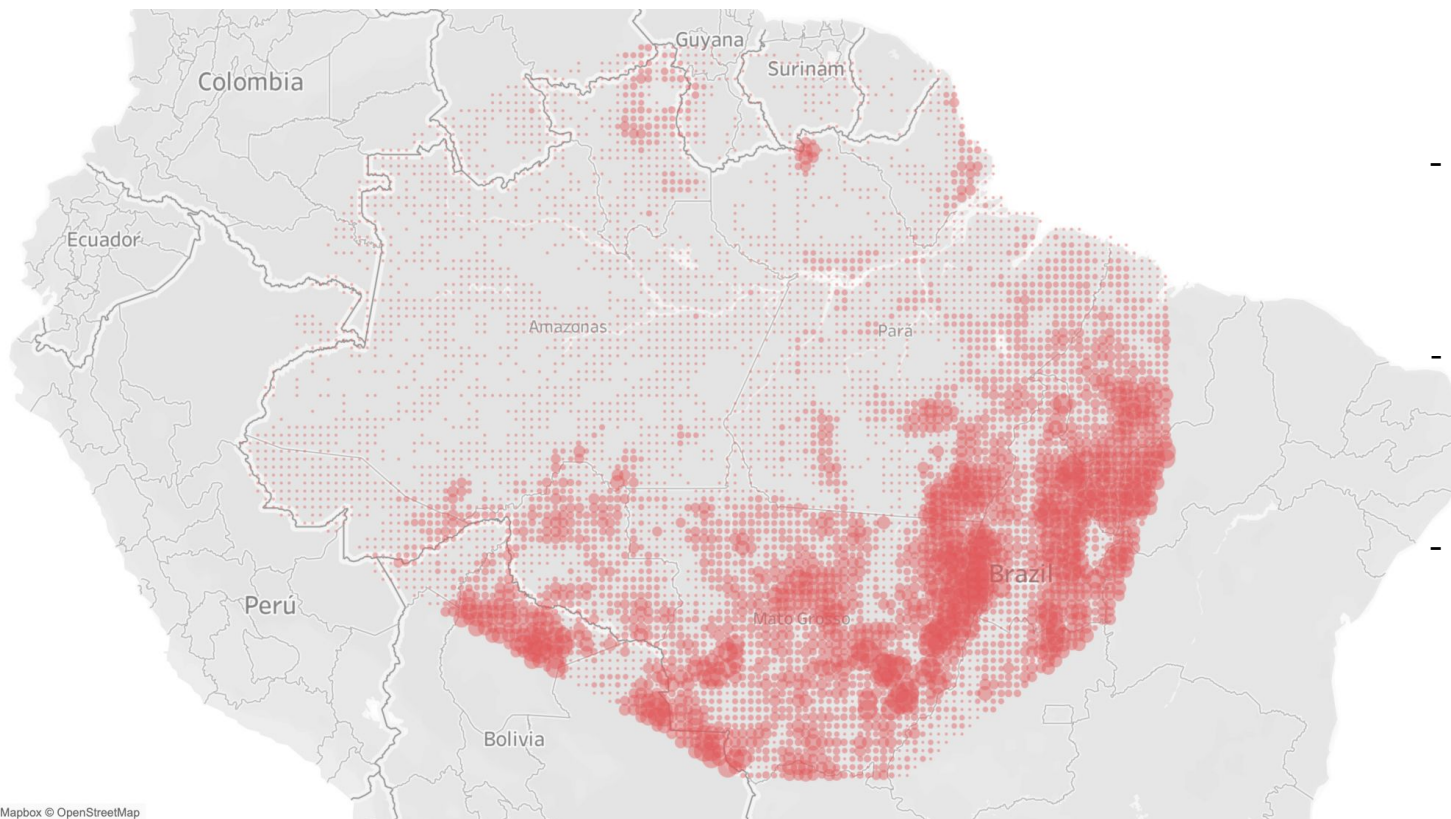
# Average capacity to retain carbon slightly increased



- Average NPP increased among the years reaching a peak of 82k g/C m2 in 2009
- Capacity of the affected areas to retain carbon from the atmosphere slightly improved

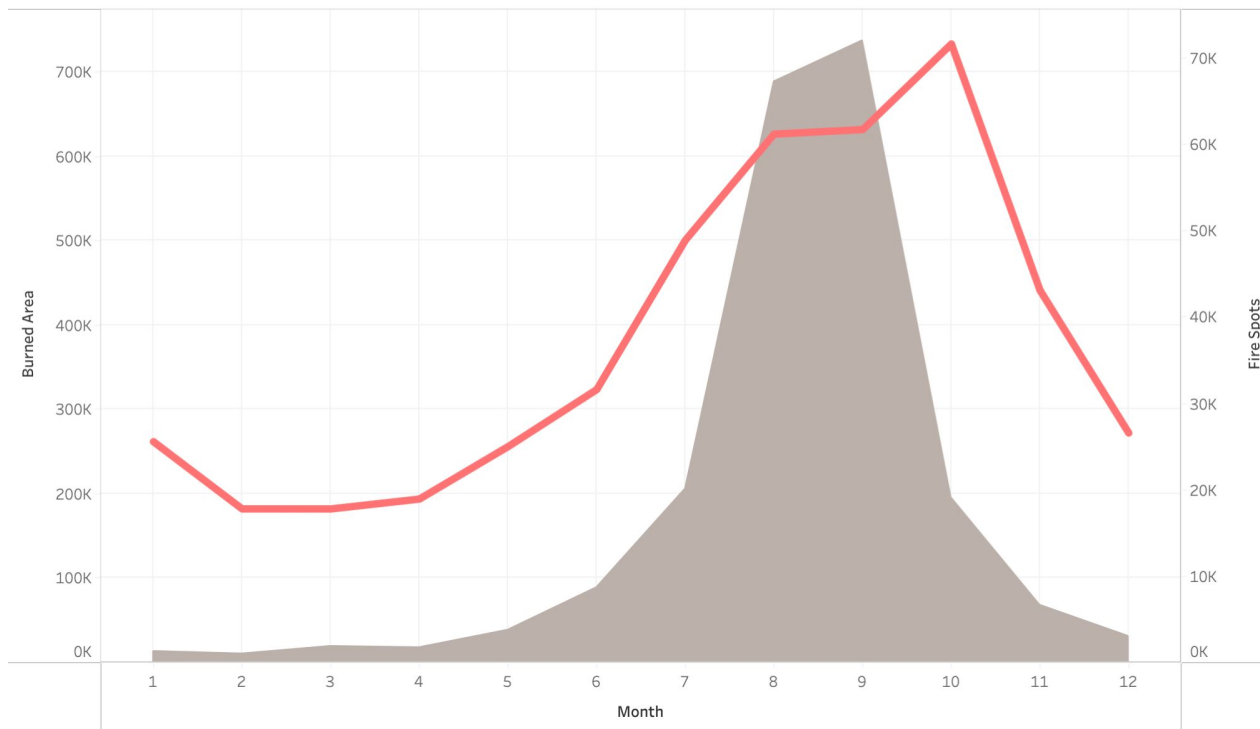
# Burned Areas: mostly concentrated in the surroundings of the forest

---



- Southern and eastern regions of the amazon have the greater burned areas
- Most affected areas are in the states of Tocantins and Mato Grosso (Brazil)
- 2.115.433ha of area burned according to this dataset (0.3% of the whole territory)

## Evidence of “fire season”, where both spot fires and burned area increase



- Season of fires in the amazon occurs mainly between july and october
- Amount of spot fires increase significantly during fire season
- Both burned area and spot fires follow a similar pattern

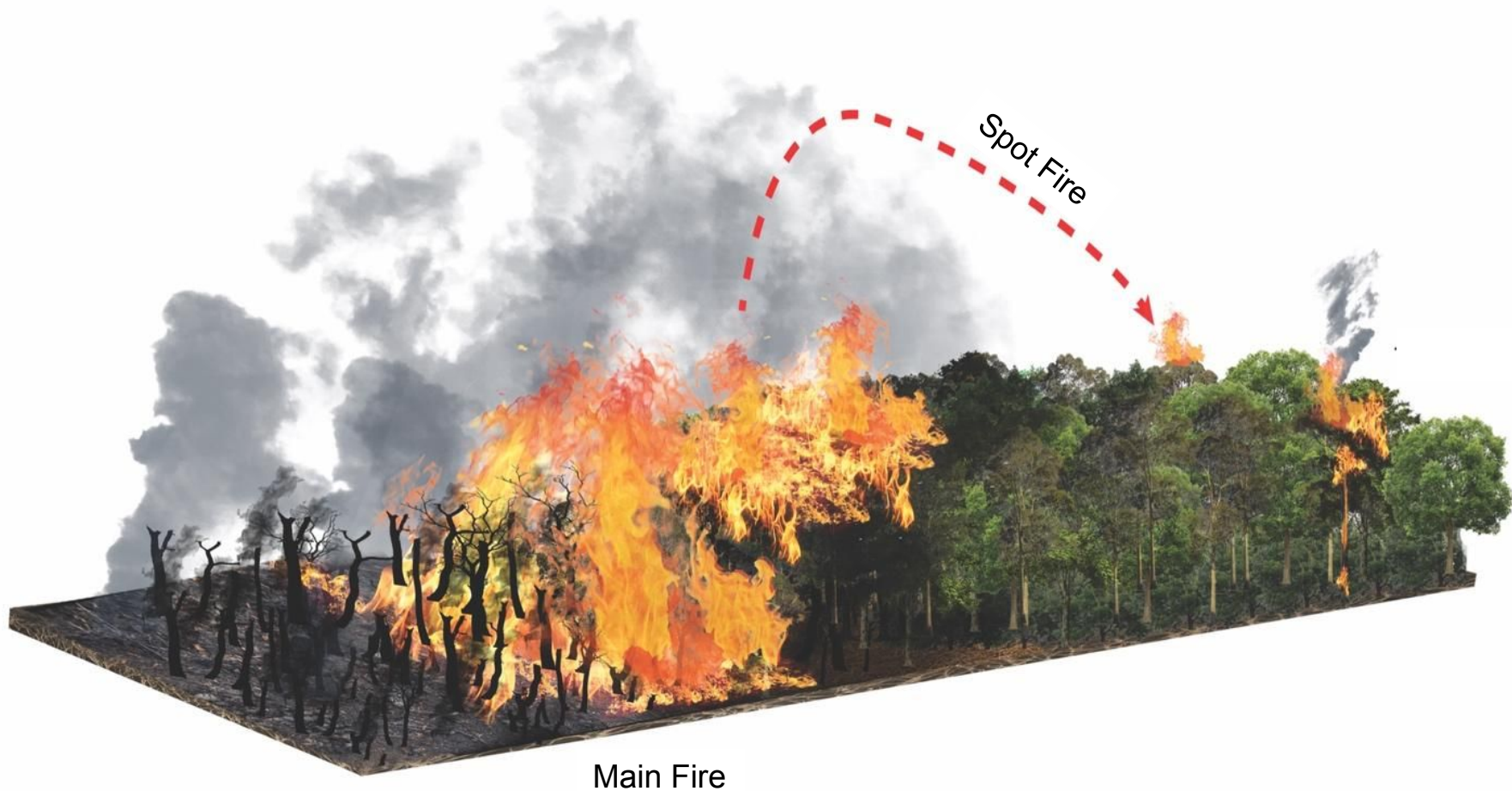
An aerial photograph of a lush green forest island in the Amazon, surrounded by calm water. The forest is dense with various shades of green, and the water reflects the surrounding landscape. The text is overlaid on the image.

5/ Predictive Model

# Spot Fires in the Amazon Forest

Target feature:  
fire\_spot\_ranges

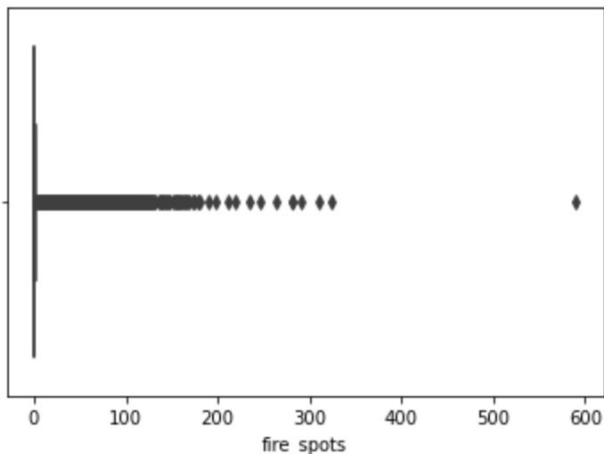




# Data Cleaning

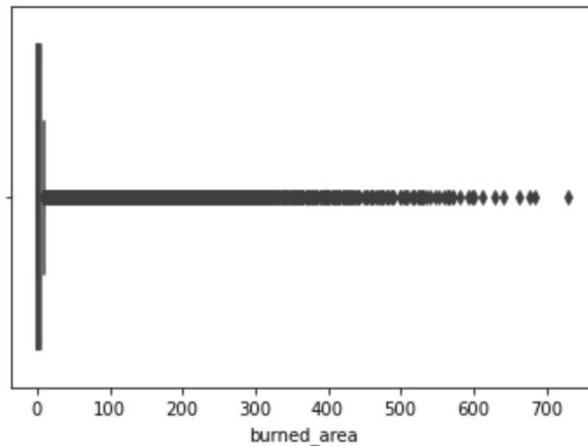
---

*Features with high presence of outliers, such as fire spots and burned area, were kept*



***Fire Spots feature***

*Most of the fires had 0 spot fires, but 24% of them had 1 or more*



***Burned Areas feature***

*Burned areas varies from less than 1 ha to more than 700 ha*

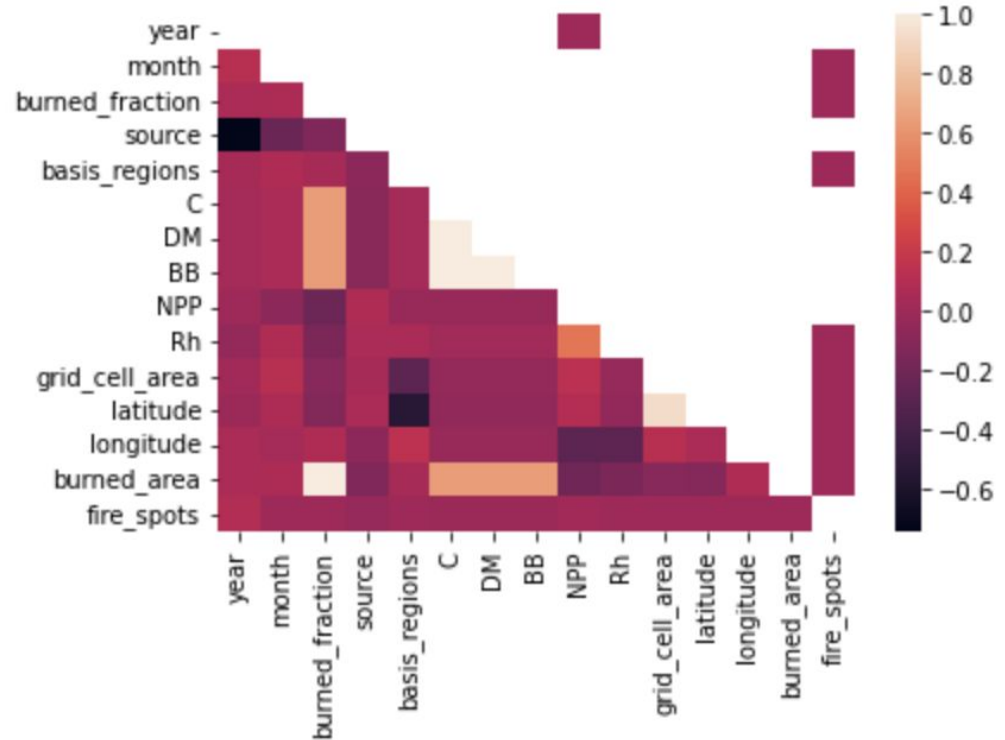
# Feature Selection

*Features with correlations higher than 0.9 were dropped*

## Multicollinearity analysis

Explanatory features dropped:

- burned\_fraction
- DM (Dry matter)
- BB (Biomass burning)
- grilled\_cel\_area



# Multi Classification: KNN Model

Predicting different ranges of spot fires in the Amazon



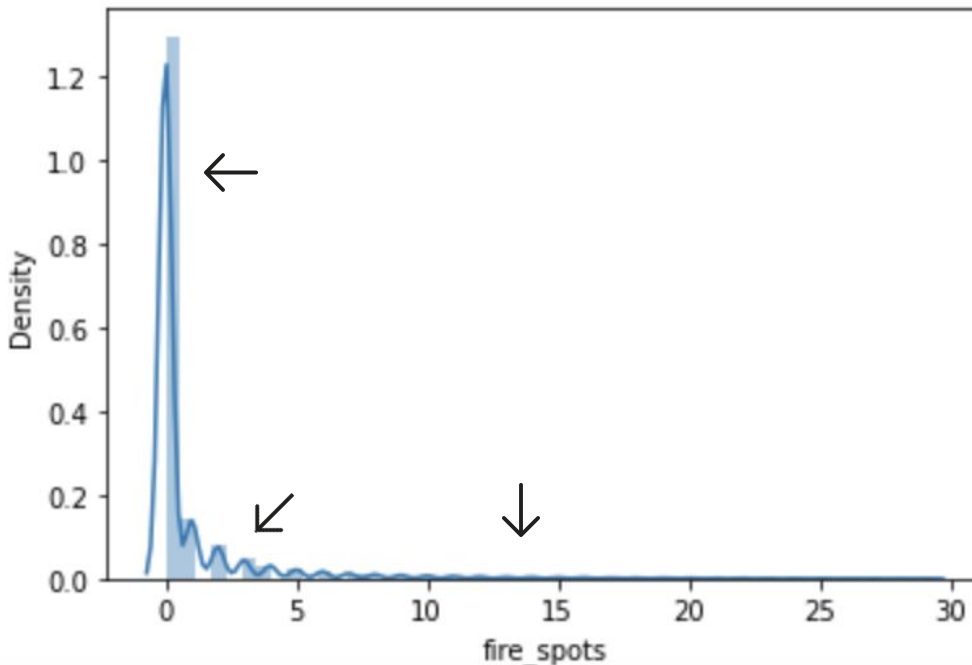


# Classification decision based on fire\_spots distribution

---

## Multi class decision:

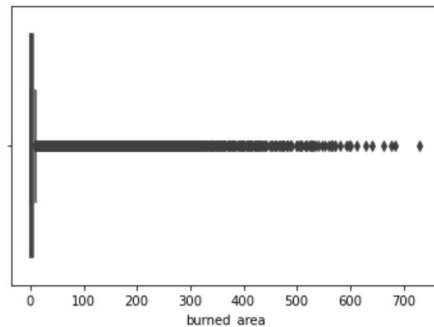
- 0: no spot fires detected
- 1: from 1 to 5 spot fires detected
- 2: more than 6 spot fires detected



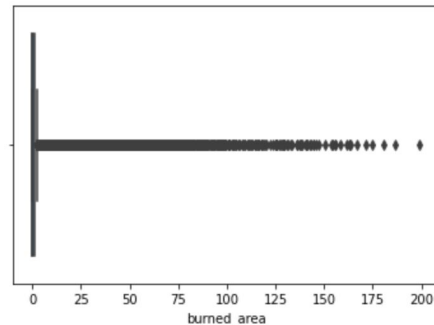
# Feature scaling: scaler selected to deal with outliers

## Scaler

- Robust Scaler
  - Removes the median and scales the data according to the quantile range



Original feature  
burned\_area: data  
distributed between 0  
and 700 ha



Robust Scaled feature  
burned\_area: data  
distributed between 0  
and 200 ha

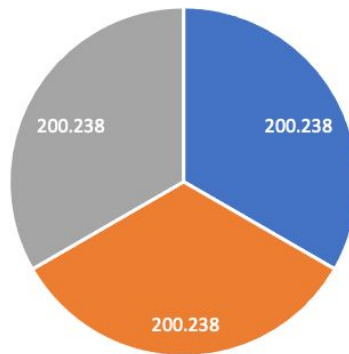
# Balancing the data

---

## SMOTE oversampling method

- Class 0: 74.8% -> 33.3%
- Class 1: 19.9% -> 33.3%
- Class 2: 5.2% -> 33.3%

Multi Class: balanced y feature distribution

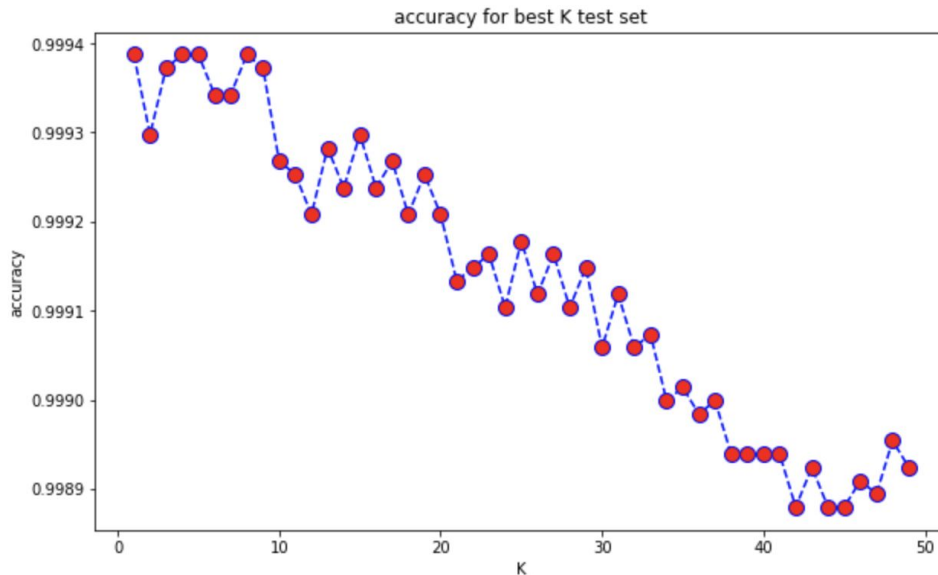


■ 0 Spot Fires ■ 1 to 5 Spot Fires ■ More than 6 Spot Fires

# Running the KNN model

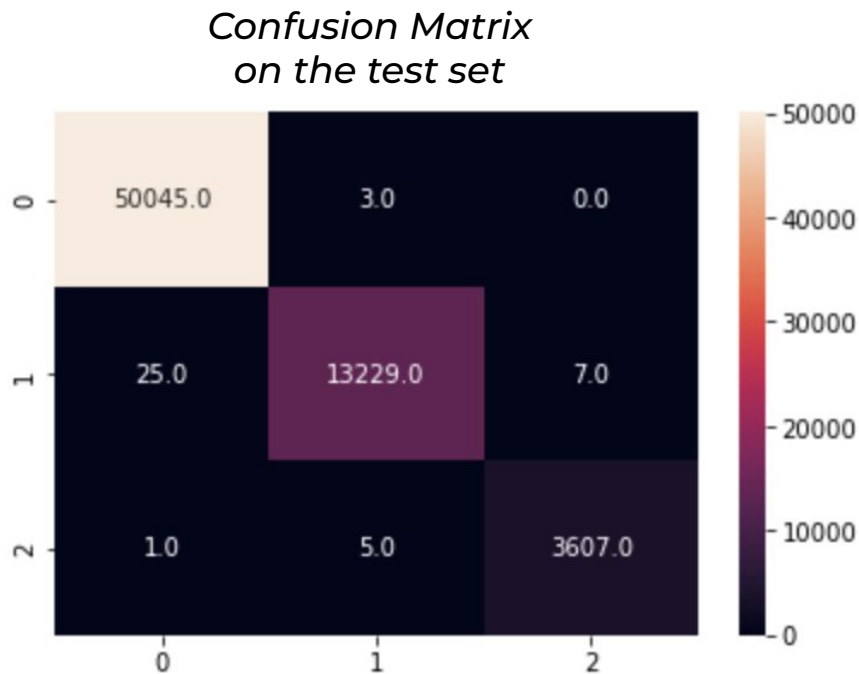
## KNN Parameters

- weight = 'distance'
  - greater influence from closer neighbors slightly improved the model
- $K = 1$ 
  - best number of neighbours according to the plot, showing accuracy result for each K value between 0 and 50



# Results: high accuracy prediction for the classes of spot fires

---



**99.6%**  
test set accuracy

**100%**  
train set accuracy

**0.3032 - 0.3064**  
conf interval for fire\_spots  
(98% conf level)

## 6/ Model Conclusions

---

- The model is able to predict the range of spot fires in the Amazon with an accuracy of more than 99% on the test set
- Overfits when the data is imbalanced and depending on the scalar adopted
- Highly sensitive to outliers, as the results improve when applying a scalar that reduces significantly the influence of them.

## 6/ General Conclusions

---

- From 1998 to 2000 the number of the fires in the Amazon rapidly decreased, but stabilized for the following 13 years, showing no clear trend for the future,
- There is an evident season of fires in the Amazon and the spot fires increase significantly in that period, a similar pattern followed by the burned areas,
- Predicting the spot fires could be relevant to control the spread of the fire, specially during the fire season,
- The multi class model high-performs with a 99% of accuracy on the test set to predict the 3 different ranges of spot fires in the Amazon forest,

## 7/ Recommendations for further studies

---

- Increase the number of classes to understand if the model is capable to predict with high accuracy more ranges of spot fires
- Explore the types of fire and fire methods to try to explain the presence of several outliers in the dataset (such as the relation between low carbon emissions in big burned areas)
- Study the impact of agriculture and livestock in the Amazon to better understand the presence of fires in the outskirts of the forest



An aerial photograph of a large, irregularly shaped island covered in dense green forest. The island is surrounded by a calm body of water, likely a lake or a wide river. The water reflects the warm, golden light of the setting or rising sun, which is visible on the right side of the frame. The sky is not visible, but the overall atmosphere is peaceful and serene. The text "Thank you" is overlaid in the center of the image in a white, sans-serif font.

Thank you

# Annexes

# Prediction of spot fires in the Amazon

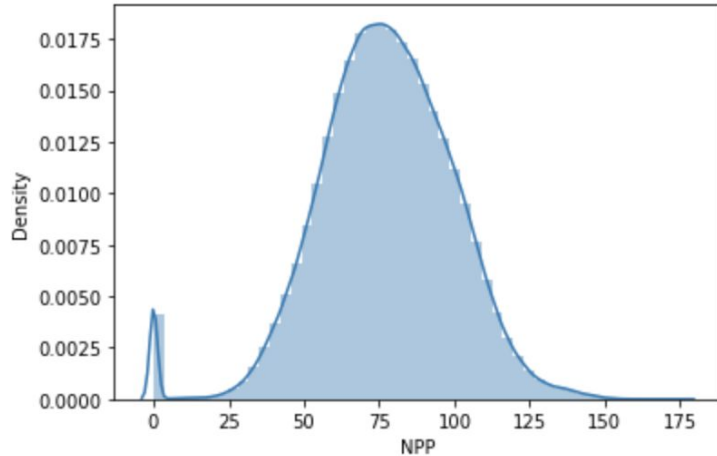
---

The main goal of this study is to understand some of the main characteristics of the fires in the Amazon and develop a model to predict the amount of spot fires for each main fire to address the problem of fire spreads in the forest.

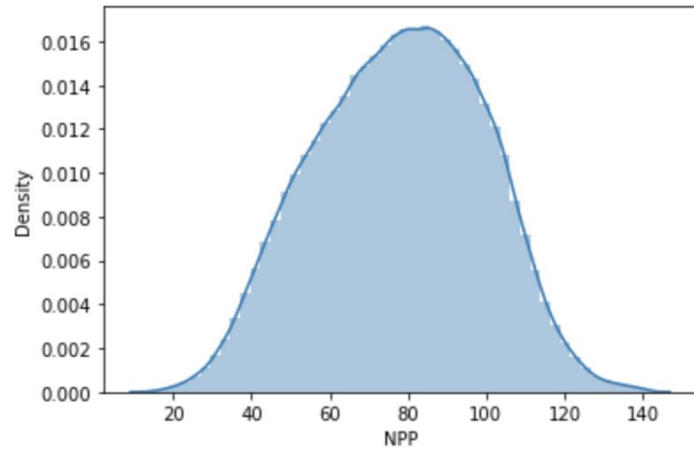
We carried out an analysis of a database with information of fires in the Amazon between the years of 1998 and 2013, following all the steps of the data science to analyze the data and develop the prediction model.

# Data Cleaning

*Features closer to a gaussian distribution (like Net Primary Production) had its outliers removed*



*Original NPP distribution*



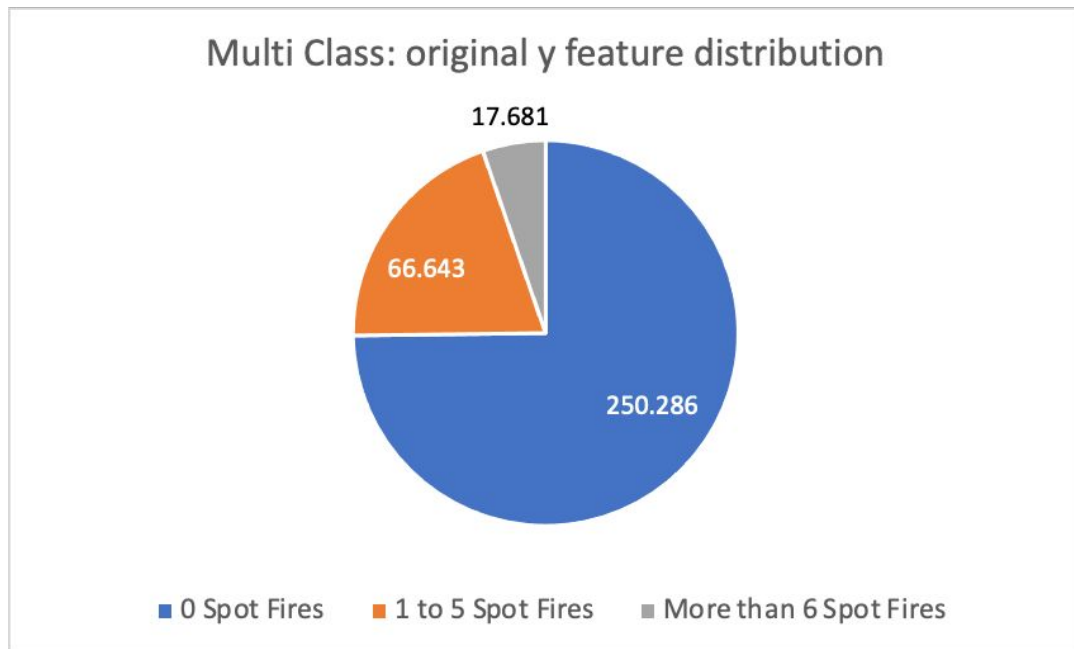
*NPP range from 15 to 140 g of C per m<sup>2</sup>  
after removing outliers*

# Classified data is imbalanced, mostly with 0 values

## Imbalanced data

### Original feature distribution

- 0: 74.8%
- 1: 19.9%
- 2: 5.3%



# KNN Model: classification report (train and test sets)

	precision	recall	f1-score	support
0	1.00	1.00	1.00	200238
1	1.00	1.00	1.00	200238
2	1.00	1.00	1.00	200238
accuracy			1.00	600714
macro avg	1.00	1.00	1.00	600714
weighted avg	1.00	1.00	1.00	600714

	precision	recall	f1-score	support
0	1.00	1.00	1.00	50048
1	0.99	0.99	0.99	13261
2	1.00	0.99	1.00	3613
accuracy			1.00	66922
macro avg	1.00	0.99	0.99	66922
weighted avg	1.00	1.00	1.00	66922