amazon-rain-forest-fire-analysis

November 15, 2024

1 AMAZON RAIN FOREST FIRE ANALYSIS

The Amazon rainforest is the world's largest rainforest and river system, and it is the home to millions of species. It's also vital to life on Earth, as it: 1. Stabilizes the climate 2. Influences global weather 3. Is home to millions of species 4. Is home to over 400 indigenous groups

The Amazon Rainforest has been experiencing increasingly frequent and intense forest fires in recent years. Understanding the patterns of these fires is high priority for addressing the environmental and ecological impacts. In this project, we analyze the data related to forest fires in Brazil, focusing on the Amazon region. The dataset consists of the number of forest fires in different states of Brazil, recorded from 1998 to 2017. This analysis aims to uncover trends, identify the months, years, and states most affected by fires, and gain insights that can inform future conservation efforts.

The dataset includes the following columns:

year: The year in which the fire data was recorded.

state: The state of Brazil where the fires occurred.

month: The month of the year when the fires were recorded.

number: The number of forest fires reported.

date: The exact date of the reported fire.

1.1 Install and Import Required Libraries

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

1.1.1 Read the csv file

```
[]: Path='location of file'
[2]: df=pd.read_csv(r'Path',encoding='ISO-8859-1')
```

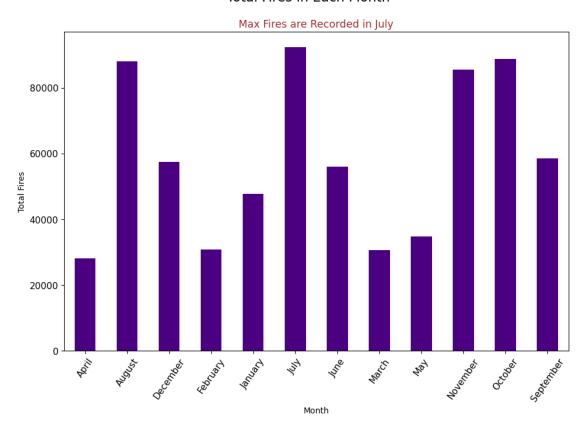
Display the first few rows and last rows of the dataset

```
[6]: df.head()
 [6]:
         year state
                       month number
                                            date
         1998
                                 0.0
                                      1998-01-01
               Acre
                     Janeiro
        1999 Acre
      1
                     Janeiro
                                 0.0
                                       1999-01-01
         2000
               Acre
                     Janeiro
                                 0.0
                                       2000-01-01
      3
         2001 Acre
                     Janeiro
                                 0.0
                                       2001-01-01
        2002 Acre
                     Janeiro
                                 0.0
                                       2002-01-01
 [7]: df.tail()
 [7]:
                                month
                                       number
            year
                      state
                                                      date
            2012
                                         128.0
      6449
                  Tocantins
                             Dezembro
                                                2012-01-01
      6450
            2013
                  Tocantins
                             Dezembro
                                         85.0
                                                2013-01-01
      6451
            2014
                             Dezembro
                                         223.0
                  Tocantins
                                                2014-01-01
      6452 2015
                  Tocantins Dezembro
                                        373.0
                                                2015-01-01
      6453
           2016 Tocantins Dezembro
                                         119.0
                                                2016-01-01
     Basic Info and Data Processing
 [8]:
      df.shape
 [8]: (6454, 5)
 [9]:
      df.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 6454 entries, 0 to 6453
     Data columns (total 5 columns):
                                  Dtype
          Column
                  Non-Null Count
                  -----
      0
                  6454 non-null
                                   int64
          year
                  6454 non-null
      1
          state
                                   object
      2
                  6454 non-null
          month
                                   object
      3
          number
                  6454 non-null
                                   float64
          date
                  6454 non-null
                                   object
     dtypes: float64(1), int64(1), object(3)
     memory usage: 252.2+ KB
     Check for duplicate rows and remove them
[10]: df.drop_duplicates(inplace=True)
     Check for null values in dataset
[13]: df.isnull().sum()
```

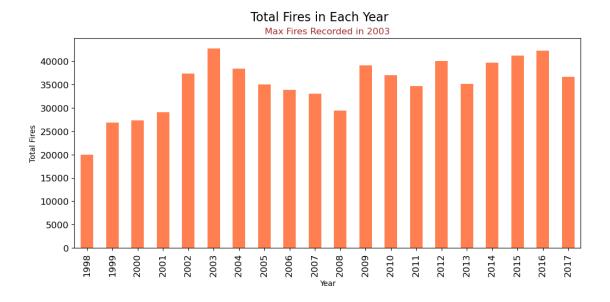
```
[13]: year
               0
     state
               0
     month
               0
     number
               0
     date
               0
     dtype: int64
[14]: df['month'].unique()
[14]: array(['Janeiro', 'Fevereiro', 'Março', 'Abril', 'Maio', 'Junho', 'Julho',
            'Agosto', 'Setembro', 'Outubro', 'Novembro', 'Dezembro'],
           dtype=object)
     Change Month names to English
[15]: month_names={'Janeiro':'January', 'Fevereiro':'February', 'Março':'March', |
       'Maio':'May', 'Junho':'June', 'Julho':'July','Agosto':'August',
                  'Setembro': 'September', 'Outubro': 'October', 'Novembro':
       [16]: df['month']=df['month'].map(month_names)
     1.2 Exploratory Data Analysis(EDA)
[17]: df.describe()
[17]:
                   year
                              number
            6422.000000 6422.000000
     count
     mean
            2007.490969
                          108.815178
     std
               5.731806
                          191.142482
                            0.000000
     min
            1998.000000
     25%
            2003.000000
                            3.000000
     50%
            2007.000000
                           24.497000
     75%
            2012.000000
                          114.000000
            2017.000000
     max
                          998.000000
     Monthly fire Analysis
[18]: #grouping with months and summation of fires in each month
     monthly_fire=df.groupby('month')['number'].sum()
[14]: monthly_fire
[14]: month
     April
                  28184.770
     August
                  88050.435
```

```
December
                  57535.480
      February
                  30839.050
      January
                  47681.844
                  92319.113
      July
      June
                  55997.675
     March
                  30709.405
     May
                  34725.363
     November
                  85508.054
      October
                  88681.579
      September
                  58578.305
     Name: number, dtype: float64
[19]: #to find the month with maximum fires
      max_fires_month=monthly_fire.idxmax()
[16]: max_fires_month
[16]: 'July'
[17]: max_fire=monthly_fire[max_fires_month]
[18]: max_fire
[18]: 92319.113
[19]: #visualize the monthly fires
      plt.figure(figsize=(11,7))
      monthly_fire.plot(kind='bar',color='indigo',xlabel="Month",ylabel="Total__
       ⇔Fires",fontsize=11)
      plt.xticks(rotation=55)
      plt.suptitle("Total Fires in Each Month",fontsize=16)
      plt.title(f"Max Fires are Recorded in {max fires month}",fontsize=12.
       plt.show()
      max_fires_month,max_fire
```

Total Fires in Each Month

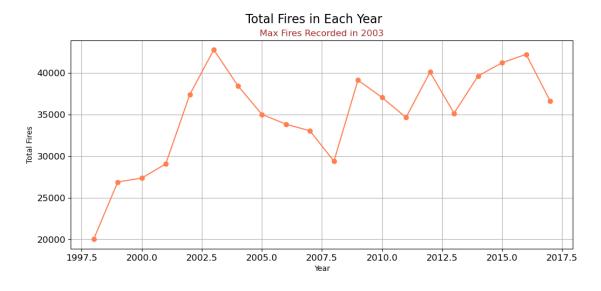


[19]: ('July', 92319.113)



[23]: (2003, 42760.674)

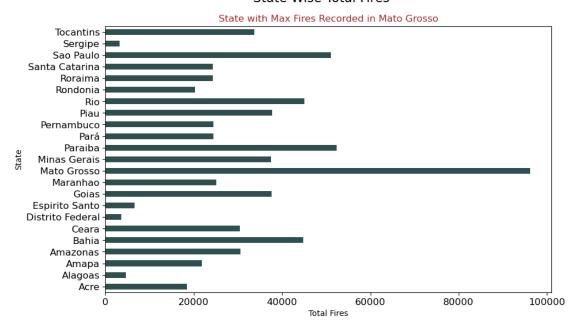
Similarly using line graph



[24]: (2003, 42760.674)

State Wise Forest Fire Analysis

State Wise Total Fires

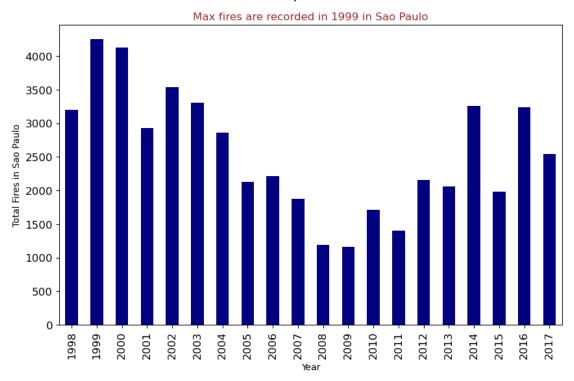


[27]: ('Mato Grosso', 96246.028)

1.2.1 STATE FIRE REPORT FOR SAO PAULO

Year-wise Analysis-Sao Paulo

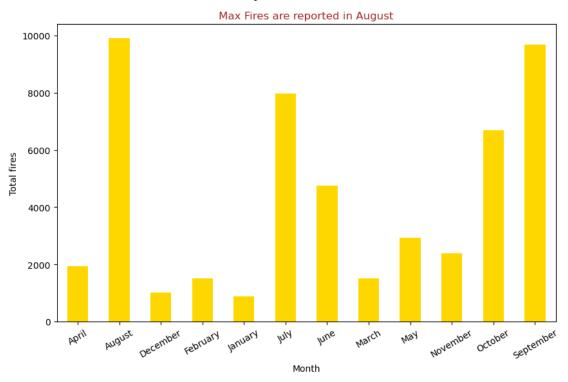
Year wise fires reported in Sau Paulo



Month-wise Fire Report(Sau Paulo)

[32]: sau_paulo_monthly_fires=sao_paulo_fire.groupby('month')['number'].sum()
max_fire_month_sau_paulo=sau_paulo_monthly_fires.idxmax()
max_sau_paulo_monthly_fires=sau_paulo_monthly_fires[max_fire_month_sau_paulo]

Monthly Fires in Sau Paulo



[33]:	month	
	April	1949.000
	August	9902.545
	December	1009.000
	February	1504.000
	January	872.000
	July	7961.000
	June	4746.000
	March	1507.000
	May	2924.000
	November	2394.000
	October	6681.000

September 9671.653

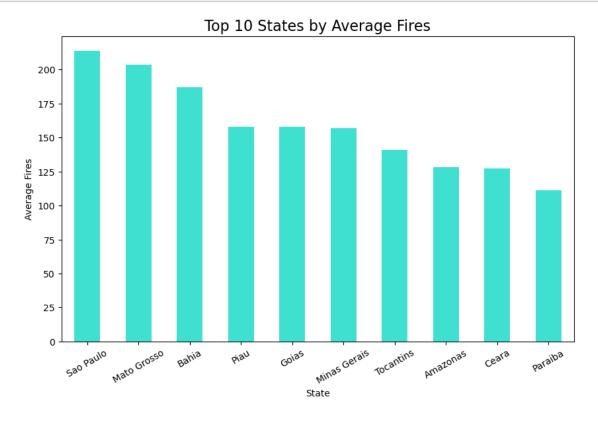
Name: number, dtype: float64

1.2.2 Special Analysis

State-wise Average Fire Reports

```
[34]: state_avg_fires= df.groupby('state')['number'].mean().

sort_values(ascending=False)
```

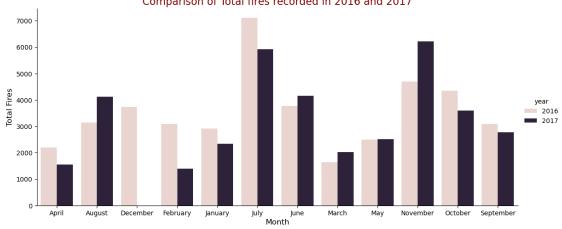


Monthly Fires Analysis in 2016 & 2017

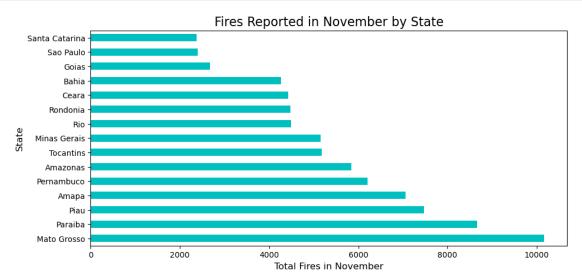
```
[21]: fires_2016_17=df[(df['year']==2016)|(df['year']==2017)]
```

[22]: monthly_fires_2016_17=fires_2016_17.groupby(['year','month'])['number'].sum()

```
[23]: monthly_fires_2016_17.info()
     <class 'pandas.core.series.Series'>
     MultiIndex: 23 entries, (2016, 'April') to (2017, 'September')
     Series name: number
     Non-Null Count Dtype
     23 non-null
                      float64
     dtypes: float64(1)
     memory usage: 433.0+ bytes
[24]: #convert the series to dataframe
      df2=monthly_fires_2016_17.to_frame()
[34]: sns.catplot(x = 'month', y='number',
                      hue = 'year', data=df2, kind='bar', aspect=16/7)
      plt.xlabel('Month',fontsize=12)
      plt.ylabel('Total Fires',fontsize=12)
      plt.title('Comparison of Total fires recorded in 2016 and
       ⇒2017',fontsize=16,color='maroon')
      plt.show()
                             Comparison of Total fires recorded in 2016 and 2017
```



```
plt.xlabel("Total Fires in November", fontsize=12)
plt.ylabel("State", fontsize=12)
plt.show()
```



2 Conclusion

The Amazon Rainforest fire analysis has provided some crucial insights into the occurrence of forest fires. The main findings include:

- 1. **Month with Maximum Fires:** Providing a valuable insight into seasonal fire trends, We have identified the month with the highest number of fires.
- 2. **Year with Maximum Fires:** We have identified the Peak fire seasons by the analysis of yearly trends.
- 3. State with Maximum Fires: With the help of state wise data we are able to find the state with highest fires and we identified Mato Grosso as the state with highest fires recorded which emphasizes the need of attention in this area.
- 4. Year-wise Analysis for Sau Paulo: The year-wise distribution of fires in Sau Paulo has revealed the significant trends in fire occurrences over the time period.
- 5. Monthly Comparision Analysis for 2016 and 2017: A breakdown of fires by month in 2016 and 2017 has provided insight into when fires were most concentrated during those years.
- 6. Average Fires by State: With the help of average fires analysis by state has revealed the top states with most frequently forest fires.

These insights are very crucial for understanding of the patterns of forest fires and their impacts. The visualization and analysis provides a base for further research, which may include investigating the causes behind the trends observed. The results can be helpful for controlling and reducing the

frequency of forest fires in the Amazon region, which is vital for preserving the rainforest and mitigating climate change.