

Project Report

on

CRIME RATE PREDICTION USING MACHINE LEARNING

*Submitted in partial fulfillment of the
requirement for the award of the degree of*

Bachelor of technology

Computer science and engineering



(Established under Galgotias University Uttar Pradesh Act No. 14 of 2011)

**Under The Supervision of
Mr. Mukesh Kumar Jha**

Submitted By

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**SCHOOL OF COMPUTING SCIENCE AND ENGINEERING
DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING
GALGOTIAS UNIVERSITY, GREATER NOIDA
INDIA
DECEMBER, 2021**



**SCHOOL OF COMPUTING SCIENCE AND
ENGINEERING
GALGOTIAS UNIVERSITY, GREATER NOIDA**

CANDIDATE'S DECLARATION

I/We hereby certify that the work which is being presented in the project, entitled “**CRIME RATE PREDICTION USING MACHINE LEARNING**” in partial fulfillment of the requirements for the award of the B.Tech submitted in the School of Computing Science and Engineering of Galgotias University, Greater Noida, is an original work carried out during the Period of october 2021 to December 2021 , under the supervision of Mr. Mukesh Kr. Jha(Assistant Professor), Department of Computer Science and Engineering/Computer Application and Information and Science, of School of Computing Science and Engineering , Galgotias University, Greater Noida

The matter presented in the project has not been submitted by me/us for the award of any other degree of this or any other places.

Ayush Verma
Devansh Saini

This is to certify that the above statement made by the candidates is correct to the best of my knowledge.

Mr. Mukesh Jha
Assistant
Professor

CERTIFICATE

The Final Thesis/Project/ Dissertation Viva-Voce examination of ~~Name: Admission No~~ has been held on _____ and his/her work is recommended for the award of ~~Name of Degree~~.

Signature of Examiner(s)

Signature of Supervisor(s)

Signature of Project Coordinator

Signature of Dean

Date: November, 2013

PLACE: GREATER NOIDA
Fall 2021 - 2022

TABLE OF CONTENTS

S.No	Particulars	Page No
1	Abstract	3
2	Introduction	4
3	Literature Reviews	5
4	Design	6-11
5	Requirements of Project	12
6	Implementation	13-22
7	Result	24
8	Conclusion	25
9	Reference	26

ABSTRACT

Crime investigation and prediction is a precise methodology for breaking down and recognizing various examples, connections and patterns in crime. The locales with high likelihood of event of crime is predicted by the framework. The framework created will assist with accelerating the most common way of settling crime for the law authorization offices. The utilization of AI and machine learning to identify crime by means of sound or cameras presently exists, is demonstrated to work, and expected to keep on growing. The utilization of AI/ML in anticipating crime or a singular's probability for committing a crime has guarantee yet is even a greater amount of an unknown. The current information from the police is utilized which subsequent to utilizing diverse prediction and clustering algorithms gives an understanding that will assist with foreseeing the probability of incidents, track crimes and help the law authorization specialists to convey assets and furthermore settle crime cases at a quicker rate.

Upgrades in crime prevention innovation will probably prod expanded all out spending on this innovation. We likewise attempt to make our characterization task more significant by consolidating different classes into bigger classes. At last, we report and think about our outcomes with various classifiers, and well on roads for future work.

Introduction

Crimes are the significant threat to the humankind. There are many crimes that happen in regular intervals of time. Perhaps it is increasing and spreading at a fast and vast rate. Crimes happen from small village, town to big cities. Crimes are of different type – robbery, murder, rape, assault, battery, false imprisonment, kidnapping, homicide. Since crimes are increasing there is a need to solve the cases in a much faster way. The crime activities have been increased at a faster rate and it is the responsibility of police department to control and reduce the crime activities. Crime prediction and criminal identification are the major problems to the police department as there are tremendous amount of crime data that exist. There is a need of technology through which the case solving could be faster.

Through many documentation and cases, it came out that machine learning and data science can make the work easier and faster. The inputs to our algorithms are time (hour, day, month, and year), place (latitude and longitude), and class of crime:

- Act 379 - Robbery
 - Act 13 - Gambling
 - Act 279 - Accident
 - Act 323 - Violence
 - Act 302 - Murder
 - Act 363 - Kidnapping
-

Literature Survey

The principle issue is that everyday the population will be expanded and by that the wrongdoings are likewise going to be Increased in various regions by this the crime percentage can't be precisely anticipated by the authorities. The authorities as they center around many issues may not foresee the wrongdoings to be occurred later on. The authorities/cops in spite of the fact that they attempt to diminish the crime percentage they may not lessen in undeniable way. The crime percentage forecast in future might be hard for them.

There has been incalculable of work done identified with wrongdoings. Enormous datasets have been looked into, and data, for example, area and the kind of violations have been removed to assist individuals with observing law requirements. Existing strategies have utilized these information bases to recognize wrongdoing areas of interest dependent on areas.

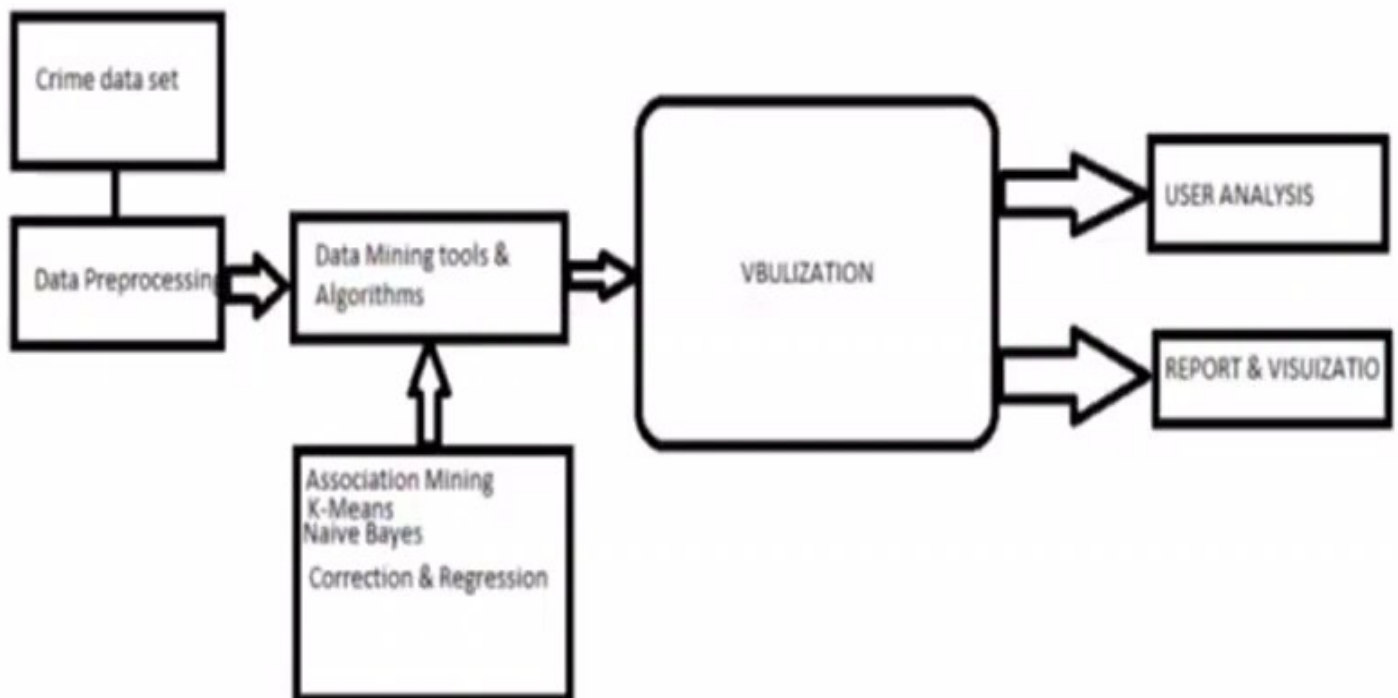
Despite the fact that wrongdoing areas have been distinguished, there is no data accessible that incorporates the wrongdoing event date and time alongside strategies that can precisely foresee what violations will happen later on.

Our review expects to discover spatial and transient criminal areas of interest utilizing a bunch of genuine world datasets of wrongdoings. We will attempt to find the most probable wrongdoing areas and their incessant event time. Moreover, we will foresee what sort of wrongdoing may happen next in a particular area inside a specific time. At long last,

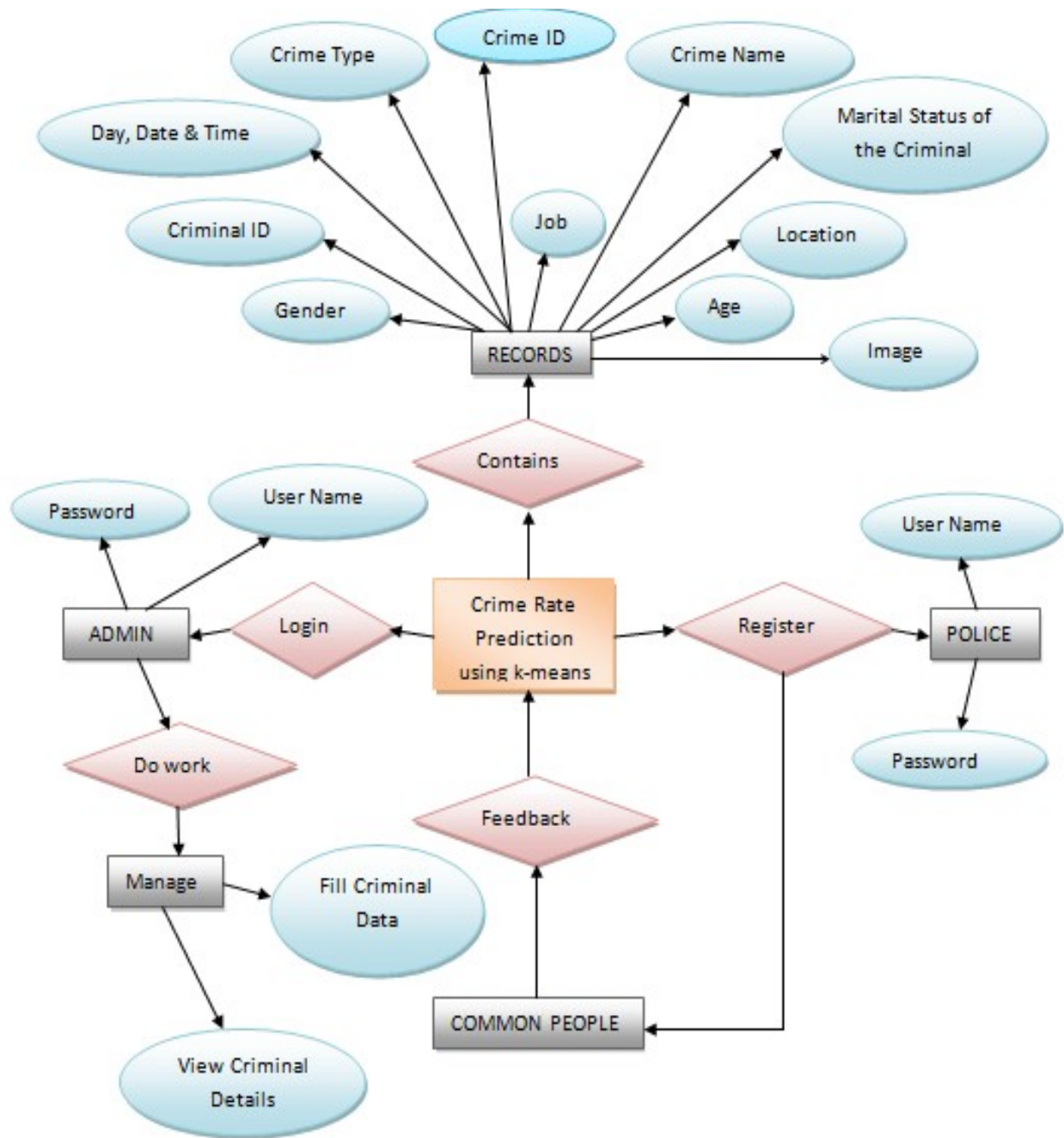
we plan to give an examination study by joining our discoveries of a specific violations datasets with its demographics data.

Project Design

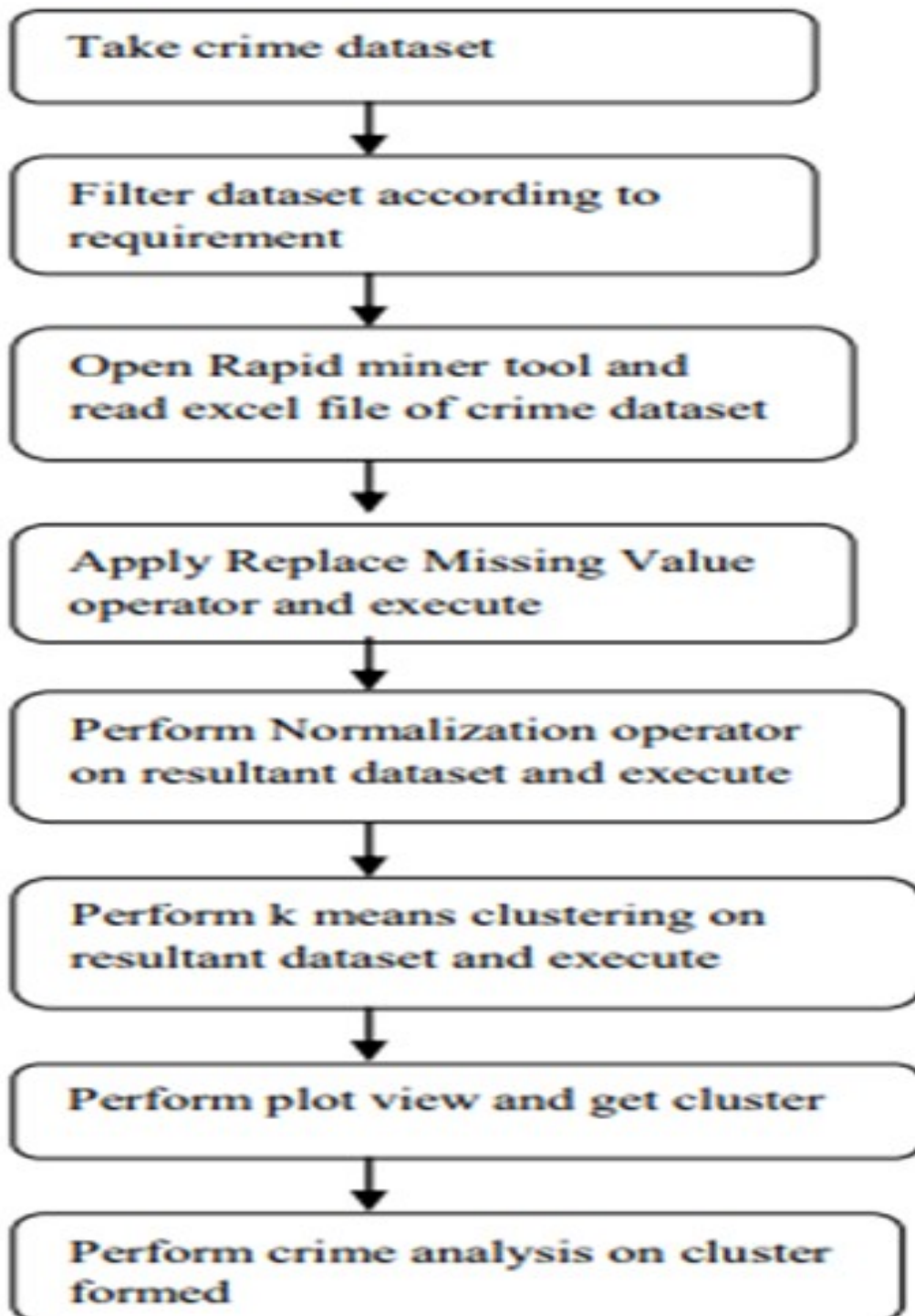
DATA FLOW DIAGRAM:-



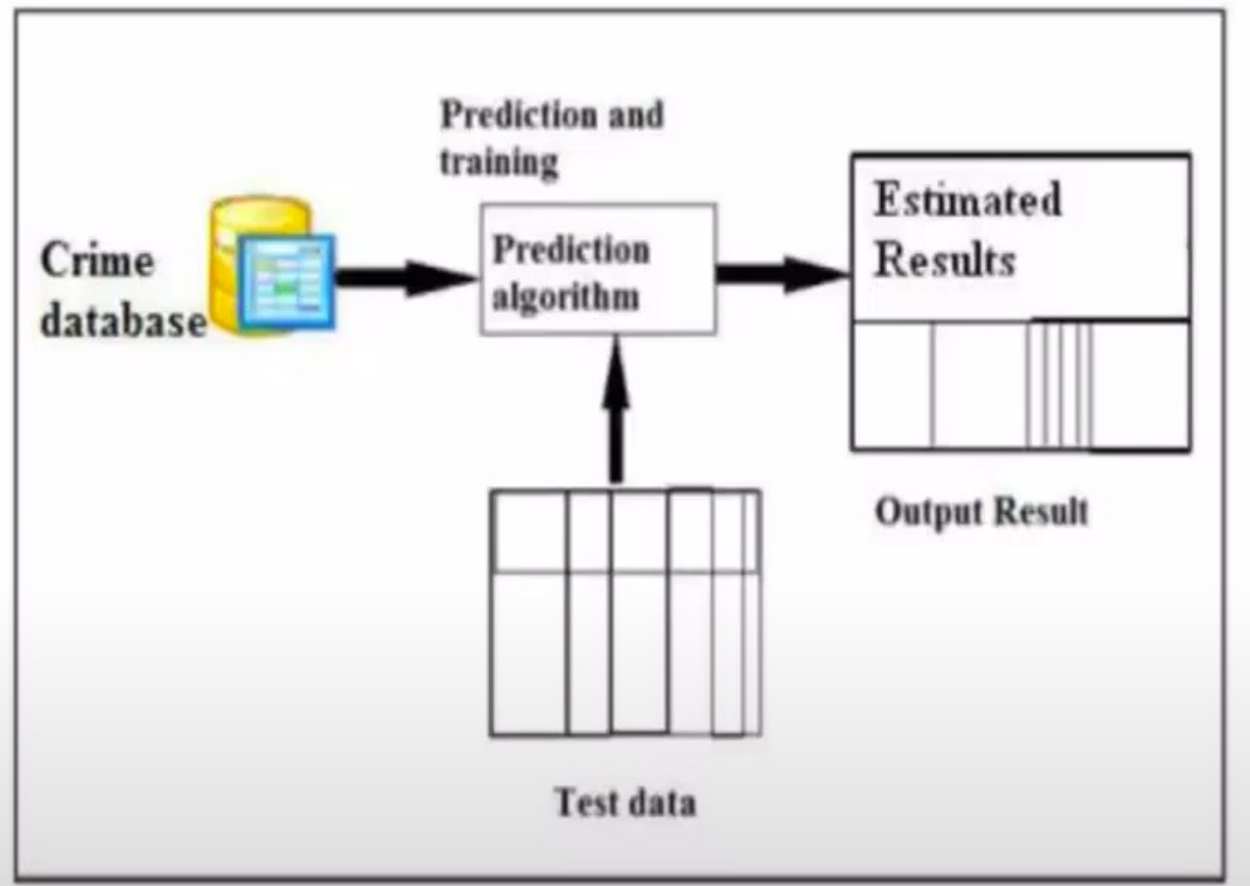
ER DIAGRAM:-



FLOWCHART:-



ARCHITECTURE DIAGRAM:-



Dataset

'/kaggle/input/up-crime-data-year-2019/MR data - Compiled Data Set.csv'

Requirements of Project

- Linux operating System
- Python 2.7
- Flask Framework
- Flask wtForms
- Flask Mysqldb
- Numpy
- Flask Mail
- SciPy
- Scikit-learn

Implementation

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image:
https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load
```

```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
```

```
# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all
files under the input directory
```

```
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
```

```
# You can write up to 20GB to the current directory (/kaggle/working/) that gets
preserved as output when you create a version using "Save & Run All"
# You can also write temporary files to /kaggle/temp/, but they won't be saved
outside of the current session
```

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load

import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory

import os
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    for filename in filenames:
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# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Save & Run All"
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session

/kaggle/input/up-crime-data-year-2019/MR data - Compiled Data Set.csv
+ Code + Markdown

[15]:
file_path = '../input/up-crime-data-year-2019/MR data - Compiled Data Set.csv'
df = pd.read_csv(file_path)

#dropping title and text columns
x_full = df.drop(axis = 1, labels = ['Title','Text'])

[16]:
#Renaming big column names(cleaning)

x_full = x_full.rename(columns = {'Murder Happened with reason:\n1:Property/Land Disputes,\n2:Family Dispute\n3:Petty Quarrels,\n4:Money Disputes, \n5:Personal Ven
x_full.columns
```

```
file_path = '../input/up-crime-data-year-2019/MR data - Compiled Data Set.csv'
df = pd.read_csv(file_path)
```

```
#dropping title and text columns
x_full = df.drop(axis = 1, labels = ['Title','Text'])
```

```
#Renaming big column names(cleaning)
```

```
x_full = x_full.rename(columns = {'Murder Happened with
reason:\n1:Property/Land Disputes,\n2:Family Dispute\n3:Petty Quarrels,\
n4:Money Disputes, \n5:Personal Vendetta, \n6:Love Affairs, \n7:Casteism\n8:
Unknown/other':'Murder Reason', 'Crime Against Women ( Combined ):\n1.
Murder with Rape \n2. Dowry Deaths(Sec. 3048)\n3. Suicide(sec 305/306)\n4.
Kidnapping(All)\n5. Acid Attack(Sec. 326A IPC)\n6. Cruelty by Husband/in-
laws((Sec.498 A IPC)\n7. Rape only(Sec. 376 or 511 IPC)\n8. Assault on Women
with Intent to Outrage her Modesty (Sec. 354 IPC)\n9. Cyber Crimes against
Women\n10. Protection of Children from Sexual Offences Act':'crime against
women(Act)'})
```

```
x_full.columns
```

```
#drop rows with all null values apart from 'City '( cleaning )
```

```
x_full = x_full.dropna(axis = 0, subset= ['City '], how = 'all')
```

Crime Rate Prediction

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Draft Session (26m)

```
df = pd.read_csv(file_path)

#dropping title and text columns
x_full = df.drop(axis = 1, labels = ['Title','Text'])
```

```
#Renaming big column names(cleaning)

x_full = x_full.rename(columns = {'Murder Happened with reason:\n1:Property/Land Disputes,\n2:Family Dispute\n3:Petty Quarrels,\n4:Money Disputes, \n5:Personal Ve
x_full.columns
```

```
[16]: Index(['City ', 'Murder Reason', 'Number of child victims',
       'Number of male victims(adult)', 'Number of female victims(adult)',
       'Kidnapping:\nNumber of child victims', 'Number of male(adult) ',
       'Number of female(adult)', 'crime against women(Act)',
       'Number of adult victims', 'Number of child involved'],
      dtype='object')
```

+ Code + Markdown

```
#drop rows with all null values apart from 'City '( cleaning )
x_full = x_full.dropna(axis = 0, subset = ['City '], how = 'all')
```

+ Code + Markdown

```
[19]: #checking Null Values
x_full.isnull().sum()
```

```
[19]: City          0
Murder Reason    319
Number of child victims    453
Number of male victims(adult)    371
Number of female victims(adult)    419
Kidnapping:\nNumber of child victims    455
Number of male(adult)    458
Number of female(adult)    463
crime against women(Act)    266
Number of adult victims    350
Number of child involved    381
dtype: int64
```

Console

#checking Null Values
 x_full.isnull().sum()

Crime Rate Prediction

File Edit View Run Add-ons Help

Share

Draft Session (31m)

```
#checking Null Values
x_full.isnull().sum()
```

```
[19]: City          0
Murder Reason    319
Number of child victims    453
Number of male victims(adult)    371
Number of female victims(adult)    419
Kidnapping:\nNumber of child victims    455
Number of male(adult)    458
Number of female(adult)    463
crime against women(Act)    266
Number of adult victims    350
Number of child involved    381
dtype: int64
```

+ Code + Markdown

```
[20]: #replacing NaN values in "number of" type and NaN values in crime agaisnt women column with 0
x_full = x_full.fillna(0)
```

```
[21]: #seperating acts

crime_acts = x_full['crime against women(Act)'].astype('string').values
for i in range(1,11):
    x_full.insert(8+i,"Act {}".format(i),0)
for j in range(len(crime_acts)):
    splitted = crime_acts[j].split(',')
    for x in splitted:
        if x=='0':
            continue
        x_full.loc[int(j),"Act {}".format(x.strip())]=1
x_full
```

```
[21]: City Murder Reason Number of child victims Number of male victims(adult) Number of female victims(adult) Kidnapping:\nNumber of child victims Number of male(adult) Number of female(adult) crime against women(Act) Act 1 ... Act 4 Act 5 Act 6 Act 7 Act 8 Act 9 Act 10 Number of adult victims Number of child involved Act 11
```

Console

```
#replacing NaN values in "number of" type and NaN values in crime against
women column with 0
x_full = x_full.fillna(0)
```

```
#seperating acts
```

```
crime_acts = x_full['crime against women(Act)'].astype('string').values
for i in range(1,11):
    x_full.insert(8+i,"Act {}".format(i),0)
for j in range(len(crime_acts)):
    splitted = crime_acts[j].split(',')
    for x in splitted:
        if x=='0':
            continue
        x_full.loc[int(j),'Act {}'.format(x.strip())]=1
x_full
```

Crime Rate Prediction Draft saved

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+ Draft Session (32m)

```

continue
x_full.loc[int(j),'Act {}'.format(x.strip())]=1
x_full

```

City	Murder Reason	Number of child victims	Number of male victims(adult)	Number of female victims(adult)	Kidnapping:\nNumber of child victims	Number of male(adult)	Number of female(adult)	crime against women(Act)	Act 1	Act 4	Act 5	Act 6	Act 7	Act 8	Act 9	Act 10	Number of adult victims	Number of childs Involved	Act 11
0	Ghaziabad	0	0.0	0.0	0.0	0.0	0.0	8,10	0	0	0	0	0	1	0	1	0.0	3.0	NaN
1	Ghaziabad	0	0.0	0.0	0.0	0.0	0.0	0	0	0	0	0	0	0	0	0	0.0	0.0	NaN
2	Ghaziabad	0	0.0	0.0	0.0	0.0	0.0	0	0	0	0	0	0	0	0	0	0.0	0.0	NaN
3	Ghaziabad	0	0.0	0.0	0.0	0.0	0.0	0	0	0	0	0	0	0	0	0	0.0	0.0	NaN
4	Ghaziabad	0	0.0	0.0	0.0	0.0	0.0	0	0	0	0	0	0	0	0	0	0.0	0.0	NaN
...
461	Lucknow	0	0.0	0.0	0.0	0.0	0.0	7	0	0	0	0	1	0	0	0	1.0	0.0	NaN
462	Lucknow	0	0.0	0.0	0.0	0.0	0.0	1	1	0	0	0	0	0	0	0	1.0	0.0	NaN
463	Lucknow	0	0.0	0.0	0.0	0.0	0.0	3,8	0	0	0	0	0	1	0	0	0.0	1.0	NaN
464	Lucknow	0	0.0	0.0	0.0	0.0	0.0	8	0	0	0	0	0	1	0	0	1.0	0.0	NaN
465	Lucknow	Petty Quarrels	0.0	1.0	0.0	0.0	0.0	0	0	0	0	0	0	0	0	0	0.0	0.0	NaN

466 rows x 22 columns

```

import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set_style('whitegrid')
sns.set()

```

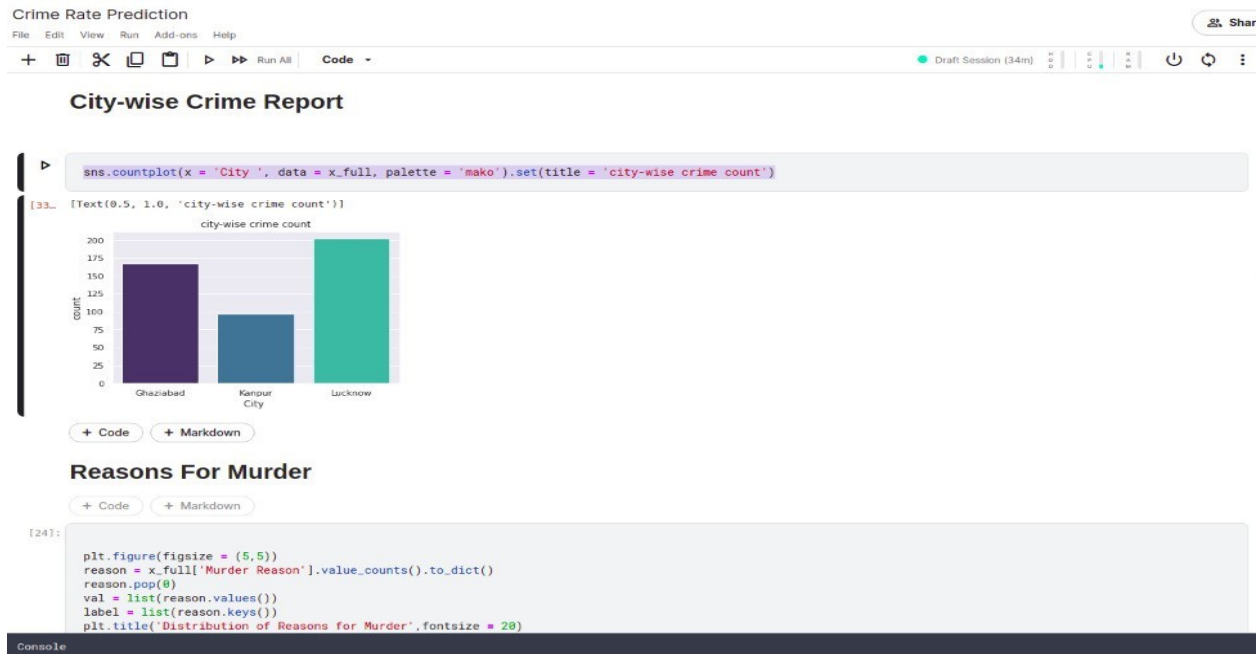
+ Code + Markdown

City-wise Crime Report

Console

```
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set_style('whitegrid')
sns.set()
```

```
sns.countplot(x = 'City ', data = x_full, palette = 'mako').set(title = 'city-wise crime count')
```



```
plt.figure(figsize = (5,5))
reason = x_full['Murder Reason'].value_counts().to_dict()
reason.pop(0)
val = list(reason.values())
label = list(reason.keys())
plt.title('Distribution of Reasons for Murder', fontsize = 20)
plt.pie(x = val, labels = label)
plt.show()
```

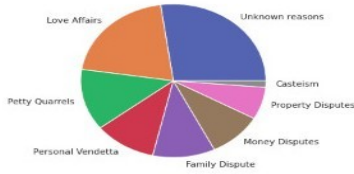


```

Crime Rate Prediction Draft saved
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+ [Icons] Run All Markdown
reason_pos(7)
val = list(reason.values())
label = list(reason.keys())
plt.title('Distribution of Reasons for Murder',fontsize = 20)
plt.pie(x = val,labels = label)
plt.show()

```

Distribution of Reasons for Murder



+ Code + Markdown

Type Markdown and LaTeX: α^2

Most of the known murder cases are due to Love Affairs

Though murder cases due to casteism is least but its still there, stating caste system still prevails in UP

Love Affair victims Gender Distribution

Console

```

Crime Rate Prediction
File Edit View Run Add-ons Help
+ [Icons] Run All Markdown
Draft Session (35m)

```

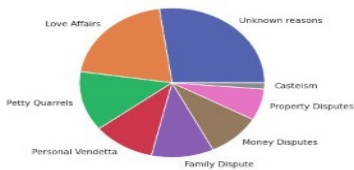
Reasons For Murder

```

[38]:
plt.figure(figsize = (5,5))
reason = x_full['Murder Reason'].value_counts().to_dict()
reason_pos(8)
val = list(reason.values())
label = list(reason.keys())
plt.title('Distribution of Reasons for Murder',fontsize = 20)
plt.pie(x = val,labels = label)
plt.show()

```

Distribution of Reasons for Murder



Console

```

#love affairs victims
mv = x_full.loc[x_full['Murder Reason'] == 'Love Affairs', "Number of male victims(adult)"].sum()
fv = x_full.loc[x_full['Murder Reason'] == 'Love Affairs', "Number of female victims(adult)"].sum()
plt.figure(figsize = (5,5))
plt.title('Love Affair victims',fontsize = 20)
plt.pie(x = [mv,fv],labels = ['females','males'],colors = ['pink','blue'])
plt.show()

```

Love Affair victims Gender Distribution

```
#love affairs victims
mv = x_full.loc[x_full['Murder Reason'] == 'Love Affairs', 'Number of male victims(adult)'].sum()
fv = x_full.loc[x_full['Murder Reason'] == 'Love Affairs', 'Number of female victims(adult)'].sum()
plt.figure(figsize = (5,5))
plt.title('Love Affair victims', fontsize = 20)
plt.pie(x = [mv, fv], labels = ['males', 'females'], colors = ['pink', 'blue'])
plt.show()
```

Love Affair victims



+ Code + Markdown

We see almost equal distribution, which shows both groups are equally victimized

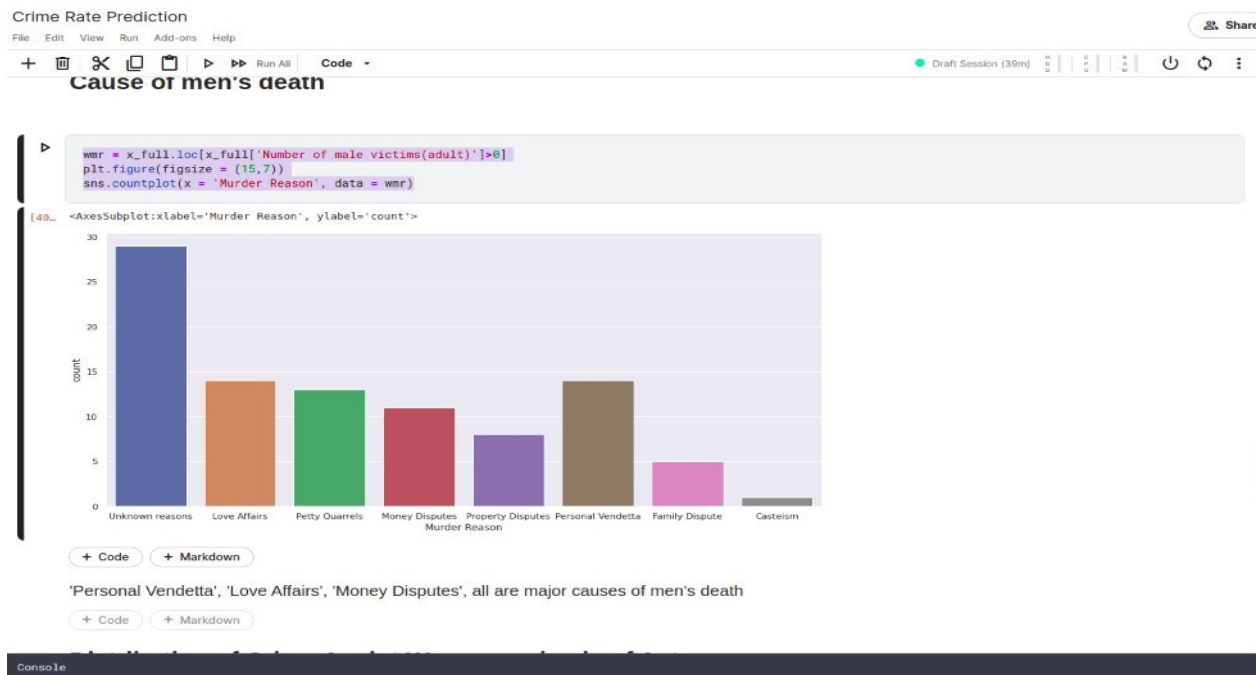
What crimes are causing women's death?

Console

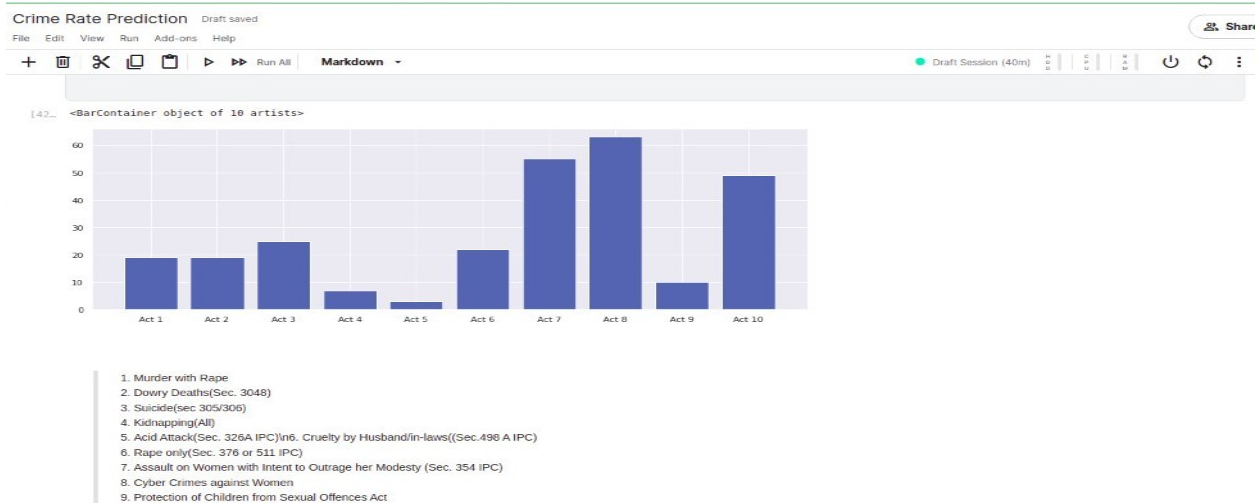
```
#what is the biggest cause of women's death
wmr = x_full.loc[x_full['Number of female victims(adult)']>0]
plt.figure(figsize = (15,7))
sns.countplot(x = 'Murder Reason', data = wmr)
```



```
wmr = x_full.loc[x_full['Number of male victims(adult)']>0]
plt.figure(figsize = (15,7))
sns.countplot(x = 'Murder Reason', data = wmr)
```



```
s = x_full.iloc[:,9:(9+10)].sum()
plt.figure(figsize = (15,5))
plt.bar(s.index,s.values)
```

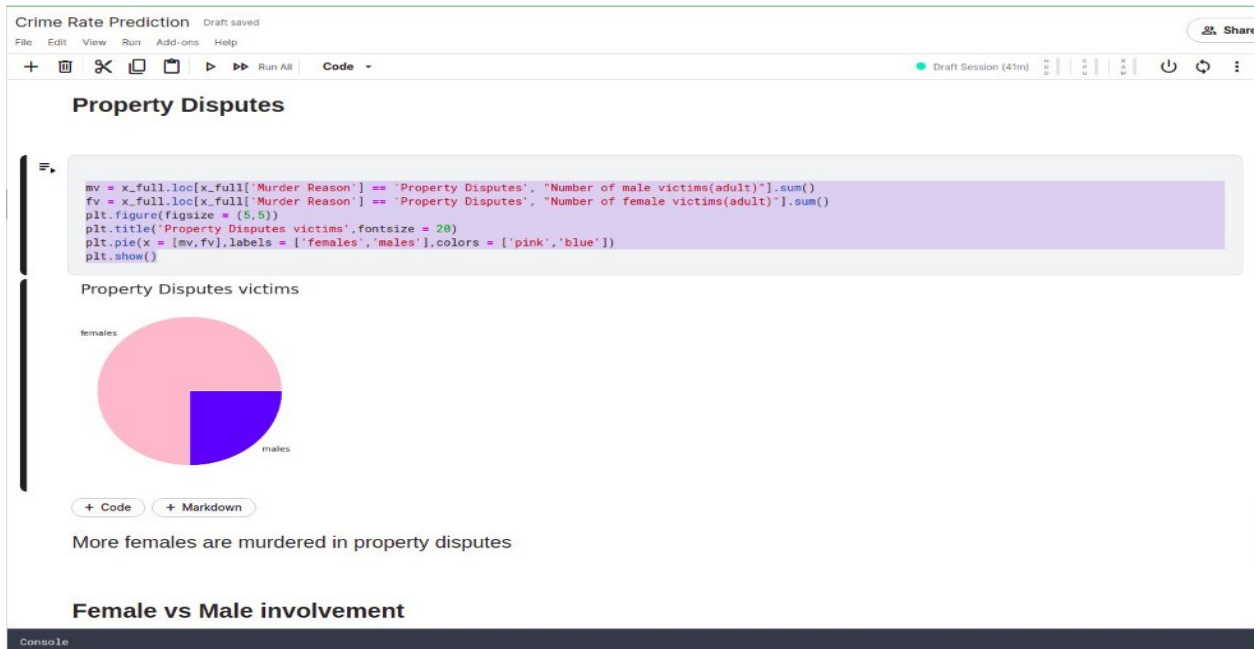


Property Disputes

+ Code + Markdown

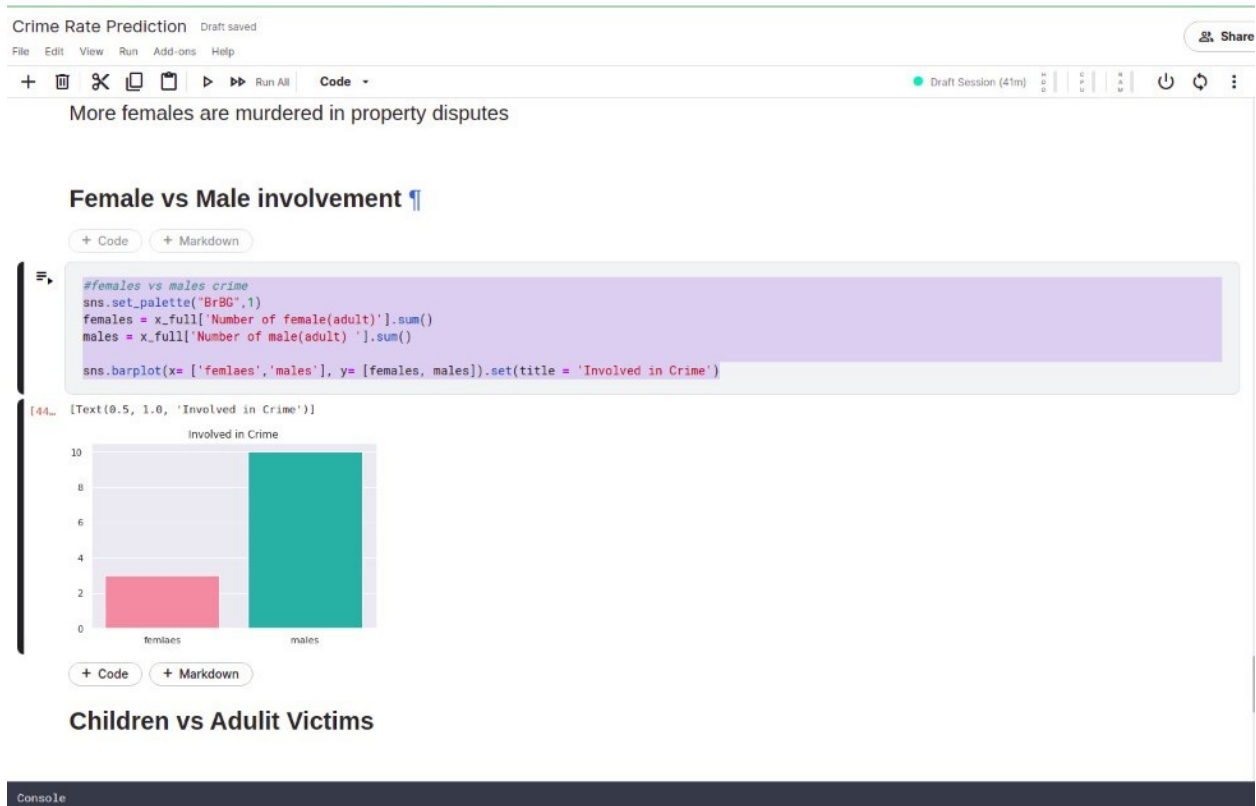
Console

```
mv = x_full.loc[x_full['Murder Reason'] == 'Property Disputes', "Number of male victims(adult)"].sum()
fv = x_full.loc[x_full['Murder Reason'] == 'Property Disputes', "Number of female victims(adult)"].sum()
plt.figure(figsize = (5,5))
plt.title('Property Disputes victims',fontsize = 20)
plt.pie(x = [mv,fv],labels = ['females','males'],colors = ['pink','blue'])
plt.show()
```



```
#females vs males crime
sns.set_palette("BrBG",1)
females = x_full['Number of female(adult)'].sum()
males = x_full['Number of male(adult) '].sum()
```

```
sns.barplot(x= ['femlaes','males'], y= [females, males]).set(title = 'Involved in Crime')
```



```
#Children vs Adults victims in crime city-wise
sns.set_palette("RdBu",)
```

```
females = x_full['Number of female victims(adult)'].sum()
males = x_full['Number of male victims(adult)'].sum()
child = x_full['Number of child victims'].sum()
```

```
sns.barplot(x= ['femlaes(adults)','males(adult)','children'], y= [females, males, child]).set(title = 'Victims of UP Crime')
```

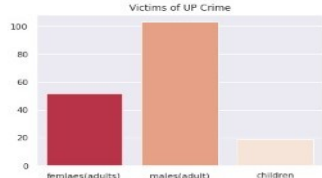
Children vs Adult Victims

```
#Children vs Adults victims in crime city-wise
sns.set_palette("RdBu",)

females = x_full['Number of female victims(adult)'].sum()
males = x_full['Number of male victims(adult)'].sum()
child = x_full['Number of child victims'].sum()

sns.barplot(x= ['females(adults)', 'males(adult)', 'children'], y= [females, males, child]).set(title = 'Victims of UP Crime')
```

[45.] [Text(0.5, 1.0, 'Victims of UP Crime')]



+ Code + Markdown

We see a higher number of adult male involvement, victimisation in UP crime report

City-wise Child victims

#child victims city-wise

sns.set_palette("YlOrBr", 1)

gbd = x_full.loc[x_full['City'] == 'Ghaziabad', 'Number of child victims'].sum()

lkw = x_full.loc[x_full['City'] == 'Lucknow', 'Number of child victims'].sum()

knp = x_full.loc[x_full['City'] == 'Kanpur', 'Number of child victims'].sum()

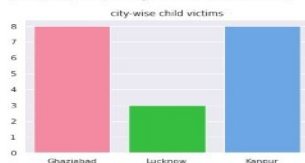
sns.barplot(y = [gbd,lkw,knp], x = ['Ghaziabad','Lucknow', 'Kanpur']).set(title = 'city-wise child victims')

City-wise Child victims

```
#child victims city-wise
sns.set_palette("YlOrBr", 1)
gbd = x_full.loc[x_full['City'] == 'Ghaziabad', 'Number of child victims'].sum()
lkw = x_full.loc[x_full['City'] == 'Lucknow', 'Number of child victims'].sum()
knp = x_full.loc[x_full['City'] == 'Kanpur', 'Number of child victims'].sum()

sns.barplot(y = [gbd,lkw,knp], x = ['Ghaziabad', 'Lucknow', 'Kanpur']).set(title = 'city-wise child victims')
```

[46.] [Text(0.5, 1.0, 'city-wise child victims')]



+ Code + Markdown

```
sns.set_palette("YlOrBr", 1)
gbd = x_full.loc[x_full['City'] == 'Ghaziabad', 'Kidnapping:\nNumber of child victims'].sum()
lkw = x_full.loc[x_full['City'] == 'Lucknow', 'Kidnapping:\nNumber of child victims'].sum()
knp = x_full.loc[x_full['City'] == 'Kanpur', 'Kidnapping:\nNumber of child victims'].sum()

sns.barplot(y = [gbd,lkw,knp], x = ['Ghaziabad', 'Lucknow', 'Kanpur']).set(title = 'city-wise child kidnappings')
```

```

sns.set_palette("YlOrBr", 1)
gbd = x_full.loc[x_full['City'] == 'Ghaziabad', 'Kidnaping:\nNumber of child victims' ].sum()
lkw = x_full.loc[x_full['City'] == 'Lucknow', 'Kidnaping:\nNumber of child victims' ].sum()
knp = x_full.loc[x_full['City'] == 'Kanpur', 'Kidnaping:\nNumber of child victims' ].sum()


sns.barplot(y = [gbd,lkw,knp], x = ['Ghaziabad','Lucknow', 'Kanpur']).set(title = 'city-wise child kidnappings')

```

Crime Rate Prediction

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Draft Session (43m)




```

sns.set_palette("YlOrBr", 1)
gbd = x_full.loc[x_full['City'] == 'Ghaziabad', 'Kidnaping:\nNumber of child victims' ].sum()
lkw = x_full.loc[x_full['City'] == 'Lucknow', 'Kidnaping:\nNumber of child victims' ].sum()
knp = x_full.loc[x_full['City'] == 'Kanpur', 'Kidnaping:\nNumber of child victims' ].sum()

sns.barplot(y = [gbd,lkw,knp], x = ['Ghaziabad','Lucknow', 'Kanpur']).set(title = 'city-wise child kidnappings')

```

[47.. [Text(0.5, 1.0, 'city-wise child kidnappings')]



+ Code + Markdown

Ghaziabad has high number of child victims and kidnappings, making the place unsafe for children

Console

Result

The output is the class of crime that is likely to have occurred. We try out multiple classification algorithms, such as KNN (K-Nearest Neighbors), Decision Trees, and Random Forests.

We also perform multiple classification tasks – we first try to predict which of 6 classes of crimes are likely to have occurred, and later try to differentiate between violent and non-violent crimes.

Conclusion

Crime Patterns can't be static since designs change over time, so unique bunching procedures like K-Means; Fuzzy C and so forth are utilized to deal with the changing crime patterns.

Much of the current work is focused in two major directions:

- Predicting surges and hotspots of crime, and
- Understanding patterns of criminal behavior that could help in solving criminal investigations.

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