Crime Pattern Analysis Group No 13

Kusum Bunkar|19111047 Nitesh Trivedi|19211266 Renuka Ahirwar|19111071 Virendra Nishad|19111099 Yash Kumar|19211274

Index

- Problem Statement
- Introduction and Motivation
- Oataset Used
- Pre-processing
- 6 Methodology
- Results and Analysis
- Future Directions

Problem Statement

- We aim to analyse hot-spot and cold-spot districts/states in crimes across India for years 2001 to 2010. We have considered following four major types of crimes:
 - 1 Crimes on SC
 - 2 Crimes on ST
 - Crimes on Children
 - Crimes on Women
- We also aim to back our analysis by finding relation with the census data.

Introduction and Motivation

- Crime pattern analysis plays an important role in construction of policies to overcome crimes committed in a region.
- The crime pattern analysis focuses on detection of hot-spot areas; the areas with high density of crime.
- Studies state that 10% people commit about 50% crimes [1].
- India ranks 65th in crimes with crime index of 44.2 https://www.numbeo.com/crime/rankings_by_country.jsp.
- The efficient utilization of resources can overcome crimes in India in a cost effective manner

Dataset Used

- District wise crime data record of India for the year 2001-2014.
- Data has been shared on Open Govt Data Platform India portal under Govt.
 Open Data License India
- Census Data for the period 2001 to 2010.
- Census Data released (and owned by) the Registrar General and Census Commissioner of India under the Ministry of Home Affairs, Government of India.

6/12/20

Pre-Processing

Following steps were followed to pre-process the data:

- The missing state/districts crime count were made zero by adding a new entry for that particular year.
- Several district's names were changed with common name if they had any other alias.
- Districts of union territories were combined together as one single district named by the corresponding union territory's name.
- Spelling mistakes in district's name were handled with correct name.
- The number of districts in the crime files were made equal in number for every year(2001-2010) using above mentioned steps.
- Census Data state/district names were made similar to the pre-processed crime files.

Methodology

We define our methodology in four steps:

- **Finding Extremum:** We found top five hotsspots/coldspots districts/states using following two metrics:
 - Population Normalized(Density)
 - ZScore(Volume)
- **Growth Rate:** We have calculated growth rate for two consecutive years of the extracted hotspot/coldspot districts/states for all crime files.
- Contribution of sub category crimes: Each subcategory contribution in percentages.
- Correlation: We found correlation of census data with crime files to extract relevant attributes from census data.
- **Inferences:** Here we discuss insights developed by analyzing inference bar plots.

Results and Analysis

Our analysis is centered towards four types of crimes:

- Crime against STs
- 2 Crime against Children
- 3 Crime against SCs
- 4 Crime against Women

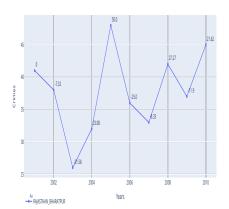
STs

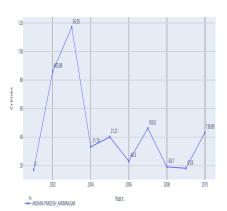
Population					
Districts		States			
Тор	Low	Тор	Low		
GWALIOR	JHABUA	RAJASTHAN	WEST BENGAL		
BHARATPUR	AMRAVATI	MADHYA PRADESH	ASSAM		
TONK	SRIKAKULAM	ANDHRA PRADESH	NAGALAND		
KARIMNAGAR	VALSAD	CHHATTISGARH	MEGHALAYA		
SIKAR	SUNDARGARH	ODISHA	BIHAR		

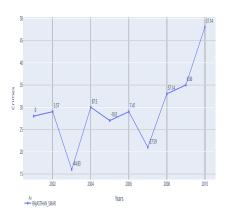
Zscore					
Districts		States			
Тор	Low	Тор	Low		
JAIPUR	TAWANG	MADHYA PRADESH	LAKSHADWEEP		
NALGONDA	GOPALGANJ	RAJASTHAN	MEGHALAYA		
BETUL	NORTH GOA	ANDHRA PRADESH	GOA		
MAYURBHANJ	SOUTH GOA	ODISHA	DAMAN & DIU		
SEONI	MANDI	CHHATTISGARH	J&K		

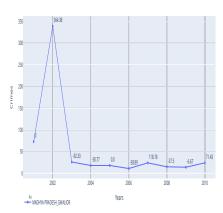
Growth Rate of STs

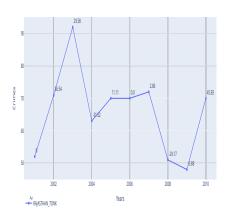
ST Crime Growth Rate



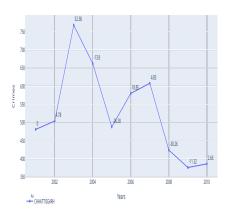


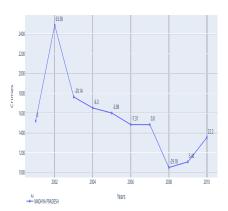


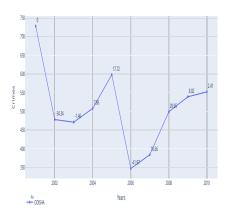


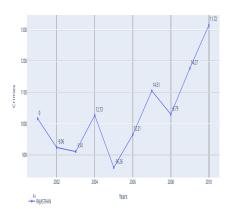




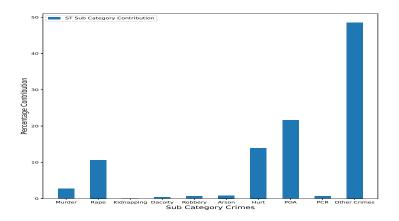








Contribution of sub category in ST crimes



Inference





- 0.4

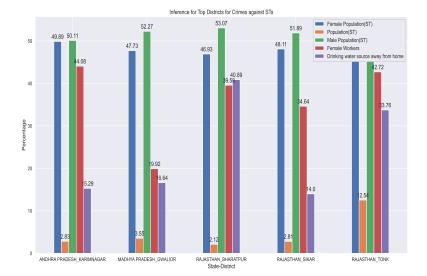
-0.3

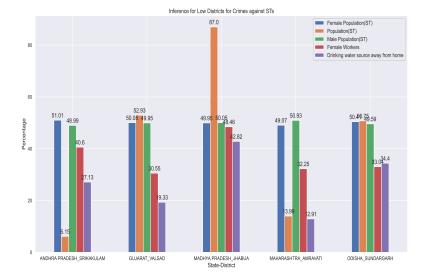
-0.2

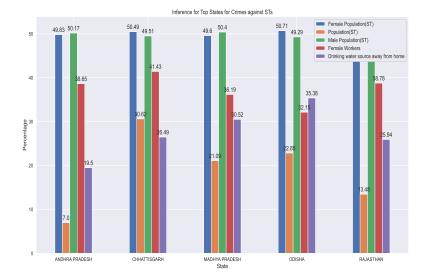
- 0.1

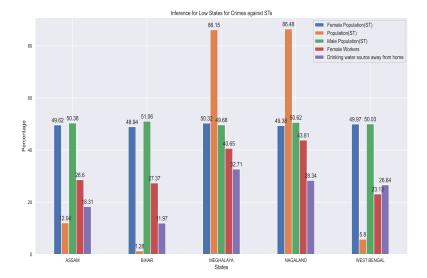
The attributes extracted from the heatmap are :

- No latrine facility within the premises.
- Cultivator Workers.
- Below Primary Education roof
- Bathing facility enclosure without roof.
- Illiterate.









Inferences drawn from the plots are:

- Comparative look at plots of cold-spot and hot-spot district shows that districts with more ST population have less crimes than districts with less ST population.
- hot-spot districts/states have more male ST population than cold-spot districts.
- Majorly hot-spot districts have high female workers than the cold-spot districts.

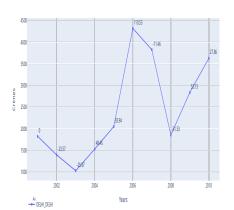
Children

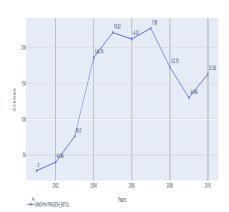
Population			
Districts		States	
Тор	Low	Тор	Low
DELHI	RAIGARH	MADHYA PRADESH	BIHAR
INDORE	CHITTOOR	DELHI	WEST BENGAL
KHANDWA	AMRELI	CHHATISGARH	TAMIL NADU
BETUL	CHENNAI	UTTAR PRADESH	KERLA
RAJGARH	SHIVPURI	MAHARASHTRA	KARNATAKA

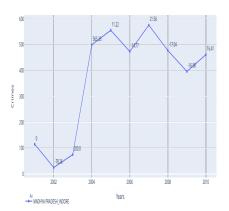
Zscore					
Districts		States			
Тор	Low	Тор	Low		
DELHI	KAMENG WEST	MADHYA PRADESH	LAKSHADWEEP		
PUNE	TAWANG	MAHARASHTRA	DAMAN & DIU		
MUMBAI	KAITHAL	UTTAR PRADESH	NAGALAND		
INDORE	JAMMU	DELHI	D & N HAVELI		
NAGPUR	KARGIL	ANDHRA PRADESH	PUDUCHERRY		

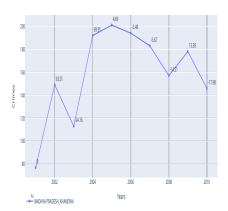
Growth Rate of Children

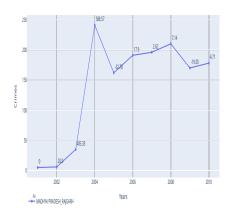
Children Crime Growth Rate

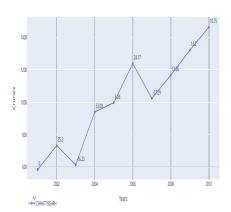


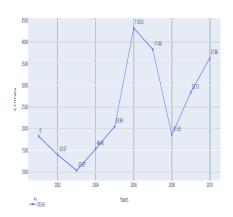


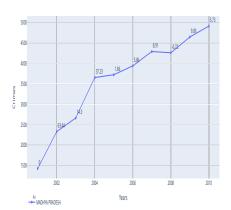


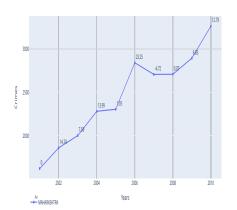


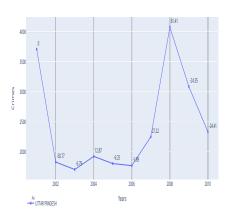




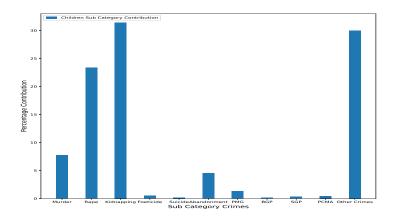




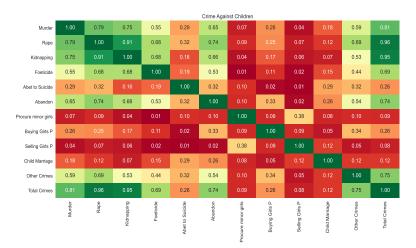




Contribution of sub category in Children crimes



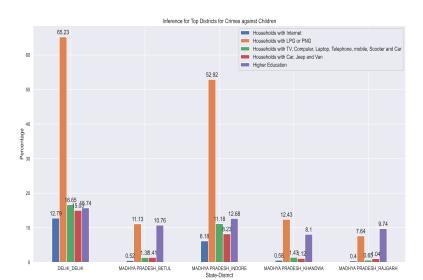
Inference

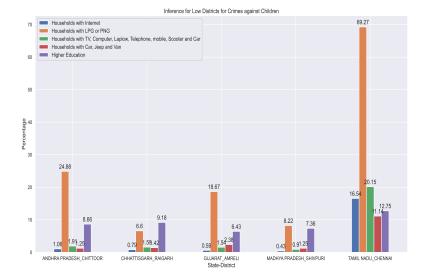




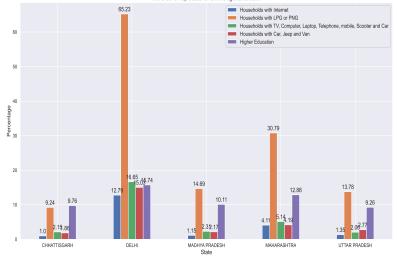
The attributes extracted from the heatmap are :

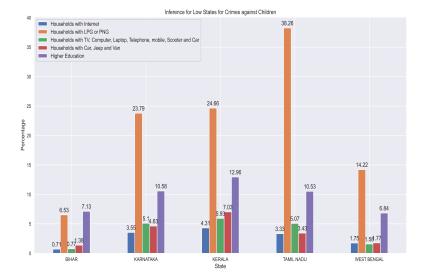
- Households with Internet.
- Households with LPG or PNG
- Households with TV, Computer, Laptop, Telephone, mobile, Scooter and Car.
- Households with Car, Jeep and Van.
- Higher Education.











Inferences drawn from the plots are:

- Delhi and Indore are the top hot-spot districts in case of crimes against children and the census data attributes like households with LPG or PNG are also high.
- However, there is exception in case of Betul, Khandwa and Rajgarh which are also in the hot-spot district but the attributes of census data like households with LPG or PNG, households with Internet are quite less.
- There is only one attribute that is Higher Education which can be accounted for majority of crimes as it is somewhat constant in all districts.

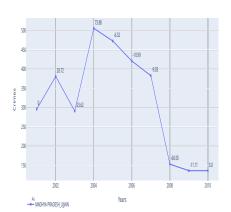
SCs

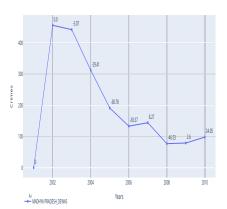
Population						
Districts		States				
Тор	Low	Тор	Low			
BHARATPUR	KAUSHAMBI	M.P	MAHARASHTRA			
UJJAIN	ADILABAD	RAJASTHAN	WEST BENGAL			
GUNA	PUNE	A.P	PUNJAB			
DEWAS	DAVANAGERE	U.P	HARYANA			
DHOLPUR	AMBEDKAR NAGAR	BIHAR	JHARKHAND			

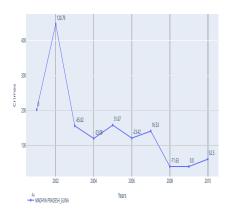
Zscore					
Districts		States			
Тор	Low	Тор	Low		
BHARATPUR	NALBARI	UTTAR PRADESH	MIZORAM		
UJJAIN	NARAYANPUR	RAJASTHAN	DAMAN & DIU		
KANPUR	JAMMU	MADHYA PRADESH	D & N HAVELI		
SITAPUR	RAJOURI	ANDHRA PRADESH	CHANDIGARH		
GANGANAGAR	RAMBAN	BIHAR	GOA		

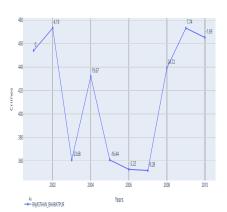
Growth Rate of SCs

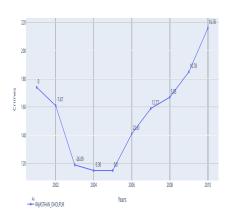
SC Crime Growth Rate

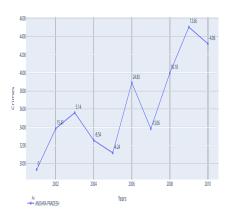


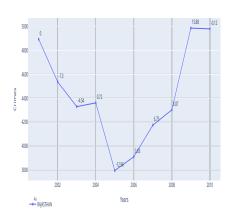




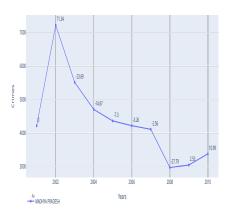


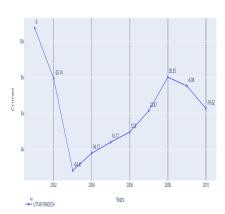




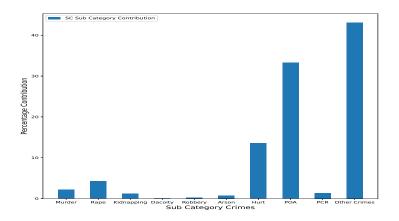








Contribution of sub category in SC crimes



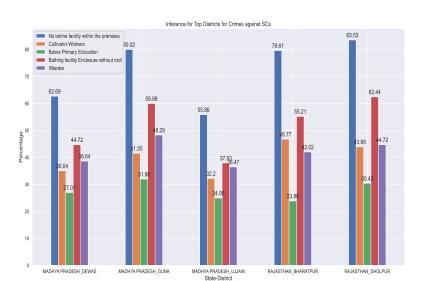
Inference

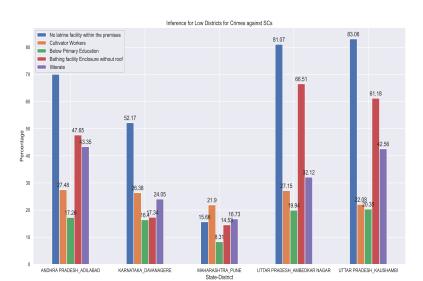
	Heatmap of Crime against SC with Census Data										
Agricultural_Workers -	0.23	0.23	0.14			0.19	0.30		0.30	0.22	0.40
Household_size_1_to_2_persons -		0.18	0.08				0.21	0.36	0.33		0.31
Type_of_bathing_facility_Enclosure_without_roof_Households -	0.49	0.37	0.40	0.21	0.20	0.38	0.29			0.37	
Married_couples_2_Households •	0.32	0.25	0.30	0.25	0.24	0.28	0.19	0.39		0.25	0.38
Female_Workers •	0.19	0.28	0.12	0.23		0.22	0.27	0.42	0.25	0.31	
Married_couples_4_Households -	0.38	0.22	0.33		0.23		0.12	0.41	0.04	0.27	0.38
Household_size_2_persons_Households •		0.18	0.08				0.22	0.37	0.34		0.31
Not_having_latrine_facility_within_the_premises_Alternative_source_Open_Households •	0.36	0.29	0.24		0.19	0.38	0.31	0.53	0.19	0.32	0.50
Illiterate_Education •	0.40	0.30	0.31		0.22	0.37	0.27	0.51	0.18	0.33	0.49
Below_Primary_Education •	0.39	0.28	0.32		0.23	0.36	0.28	0.50	0.17	0.29	
Household_size_6_8_persons_Households •	0.43	0.30	0.39	0.23	0.27	0.34	0.21			0.28	0.42
Cultivator_Workers -	0.37	0.37	0.27	0.25	0.24		0.23	0.32	0.05	0.43	
Married_couples_3_or_more_Households -	0.38	0.25	0.34	0.21	0.26	0.42		0.40	0.06	0.29	0.40
Household_size_9_persons_and_above_Households -	0.49	0.26	0.42	0.21	0.28	0.42		0.42	0.05	0.27	0.39
Hindus -	0.34	0.28	0.30	0.22	0.23	0.29	0.25	0.47	0.23	0.29	0.45
Married_couples_3_Households -	0.37	0.27	0.35	0.23	0.26	0.40	0.16	0.38	0.06	0.30	0.40
Household_size_1_person_Households -	0.13	0.19	0.08	0.11	0.06	0.07	0.18	0.31	0.31	0.16	0.28
Male_SC -		0.30	0.36	0.07	0.17	0.32	0.22	0.41	0.16	0.32	0.44
	At inter	n.	Middennia a	David.	Date land	4.5	100	nó.	non	00	Total Column

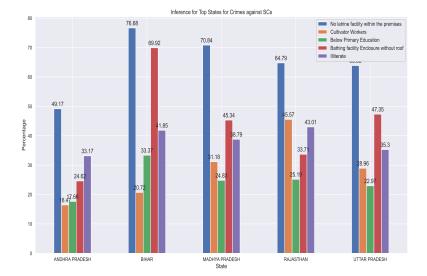
- 0.3

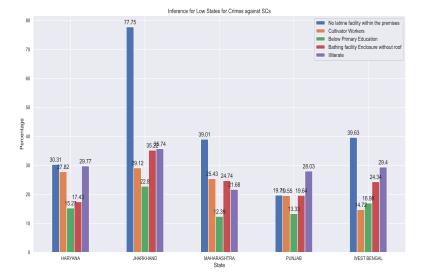
The attributes extracted from the heatmap are :

- No latrine facility within the premises.
- Cultivator Workers.
- Below Primary Education roof
- Bathing facility enclosure without roof.
- Illiterate.









Inference drawn from the plots are:

- Illiteracy as factor, low for state/district having low crime count against SCs. Example, Punjab(28 percent) a cold-spot but Bihar(41 percent) is a hot-spot. In districts, Pune(16 percent) a cold-spot and Bharatpur(42 percent) is a hot-spot for crimes against SCs.
- For Bathing Facility without roof percentage is more in Ujjain District(38 percent), a hot-spot but in Pune(14 percent) is a cold-spot. For state MP with 45 percent, a hot-spot but Maharashtra with 24 percent is a cold-spot for crimes against SCs.
- For Below Primary Education, Uttar Pradesh(23 percent), a hot-spot and West Bengal(16 percent) is a cold-spot for crime against SCs. In districts, Dewas district(27 percent) is hot-spot and district Adilabad(17 percent) is a cold-spot for crime against SCs.
- Cultivator Workers for district Dholpur(43 percent), a hot-spot whereas Kaushambi(22 percent) is a cold-spot. For state Rajasthan(45 percent) is a hot-spot and Jharkhand(29 percent) is a coldpot.

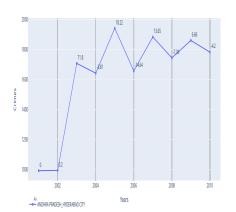
Women

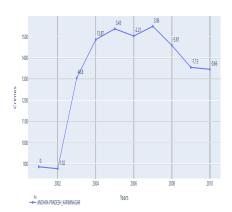
Population					
Districts		States			
Тор	Low	Тор	Low		
DELHI	DEHRADUN	DELHI	HARYANA		
HYDERABAD CITY	MADHUBANI	A.P	CHHATTISGARH		
CHITTORGARH	SIHORE	M.P	JHARKHAND		
KARIMNAGAR	ETAH	RAJASTHAN	PUNJAB		
KOTA	HISSAR	W.B	J & K		

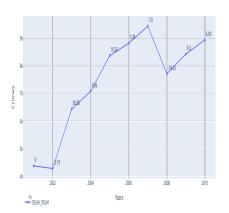
Zscore						
Districts		States				
Тор	Low	Тор	Low			
DELHI	TAMENGLONG	U.P	LAKSHADWEEP			
24 PARGANAS SOUTH	LONGLENG	A.P	DAMAN & DIU			
MUMBAI	ANJAW	M.P	NAGALAND			
24 PARGANAS NORTH	KIPHIRE	W.B	D & N HAVELI			
HYDERABAD CITY	CHANDEL	RAJASTHAN	PUDUCHERRY			

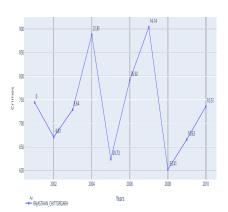
Growth Rate of Women

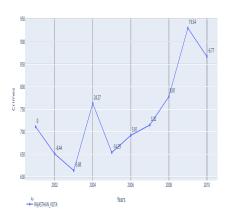
Women Crime Growth Rate

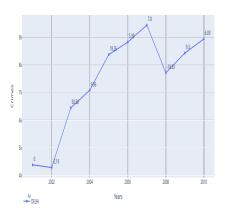


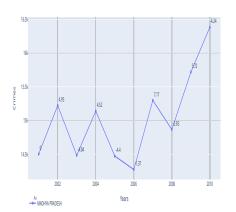






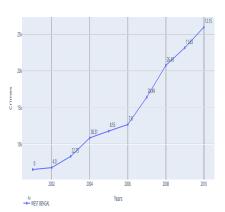




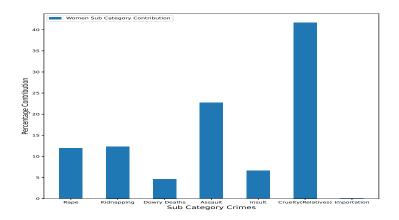




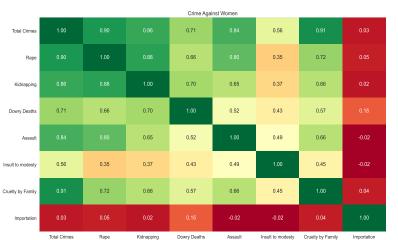




Contribution of sub category in Women crimes



Inference

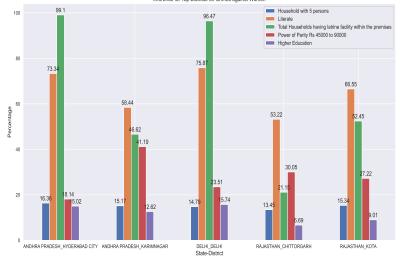


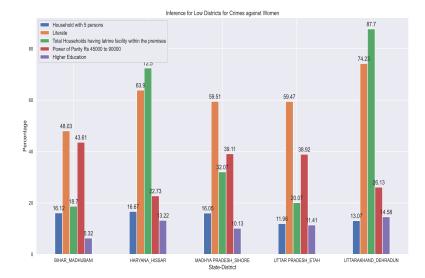


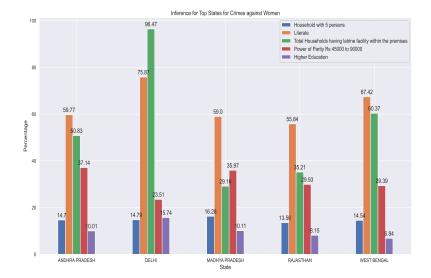
The attributes extracted from the heatmap are :

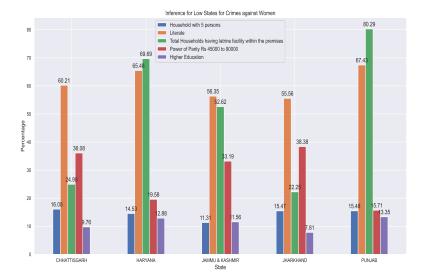
- Literates
- Households with 5 members.
- Households with latrine facility within the premises.
- Power of Parity Rs 45000 to 90000.
- Higher Education.











Inference drawn from the plots are:

- In major cases, districts appearing in top five hot-spot regions are cities.
- Power of Parity is on a higher side in low districts/states as compared to that in top districts/states. This can be considered as one the probable reason for crime against women.
- Higher Education is slightly high in major cases in low districts/states as compared to that in the top districts/states, clearly showing it's impact in total crimes against women.
- Other than the stated factors all other highly correlated factors stand in discriminating between hot-spots and cold-spot regions.

Future Directions

- We have only considered positively correlated census attributes while some negatively correlated attributes might help to explain analysis better.
- Comparative analysis can be improved by considering urban areas and rural areas separately.
- Prediction model could be trained to predict crimes in future.
- Other crimes might be considered which will expand breadth of the analysis.
- Data could be made more consistent using interpolation or regression model to predict missing values.
- Availability of more recent data may improve our results.

References I



S. V. Nath.

Crime pattern detection using data mining.

In 2006 IEEE/WIC/ACM International Conference on Web Intelligence and Intelligent Agent Technology Workshops, pages 41–44, 2006.