

Review Article

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Exploring Spatial Statistics for Cluster and Outlier Analysis in Crime Hotspots: A Geospatial Study of Coastal Zones in Kollam and Thiruvananthapuram

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Abstract

This study investigates crime patterns within the coastal zones of Kollam and Thiruvananthapuram districts in Kerala, India. Utilizing crime data from over 60 police stations for the year 2020, the research employs advanced spatial statistical techniques to conduct a rigorous analysis. A comprehensive geospatial approach is adopted to identify spatial clusters and outliers of criminal activity, offering valuable insights into the spatial dynamics of crime in these regions. The findings contribute to a deeper understanding of the geographical distribution of crime, providing a scientific foundation for informed decision-making in both law enforcement and urban planning initiatives.

Keywords: Geospatial Information Systems (GIS), Crime, Hotspots, Geography, Cluster, Cartography

1. Introduction

A comprehensive understanding of crime patterns and their spatial distribution is fundamental to formulating effective law enforcement strategies and urban planning initiatives. This research delves into this critical domain by investigating crime patterns within the coastal zones of Kollam and Thiruvananthapuram districts in Kerala, India. Employing a rigorous geospatial approach, the study leverages crime data encompassing over 60 police stations for the year 2020. By utilizing advanced spatial statistical techniques, the research not only identifies spatial clusters of criminal activity but also uncovers potential outliers. This comprehensive analysis sheds light on the spatial dynamics of crime within these regions, offering valuable insights for informed decision-making in both law enforcement and urban planning endeavors. Ultimately, the findings contribute to a deeper understanding of the geographical distribution of crime, providing a scientific foundation for the development of targeted interventions and preventative measures.

2. Materials and Methods

2.1 Data Acquisition: Crime data encompassing various crime categories for the year 2020 was meticulously procured from over 60 police stations situated within the coastal regions of Kollam and Thiruvananthapuram districts, Kerala, India. To ensure data integrity and consistency, a standardized data collection protocol was strictly adhered to throughout the acquisition process.

2.2 Data Preprocessing: The acquired quantitative crime data underwent a rigorous geospatial conversion process. Meticulous

geocoding procedures were implemented to associate each crime record with its precise geographic coordinates, thereby enabling subsequent spatial analysis.

Spatial Data Analysis: A robust geospatial approach was employed to elucidate the spatial distribution of crime incidents within the study area. This involved the application of advanced spatial statistical techniques, including:

• **Cluster and Outlier Analysis:** To identify statistically significant spatial clusters and outliers of criminal activity, Anselin's Local Moran's I statistic was utilized. The analysis yielded comprehensive results encompassing the Local Moran's I (Lmi) index, Lmi Z-score, Lmi p-value, cluster/outlier (CO) type, number of neighbors, Z-transform, and spatial lag for each location.

• Hotspot Visualization: Following the cluster and outlier analysis, the Inverse Distance Weighted (IDW) interpolation technique was employed to effectively visualize the identified hotspots of criminal activity. This technique facilitated the creation of a spatial heatmap, which visually depicted areas exhibiting high and low concentrations of crime incidents.

This research methodology provides a rigorous foundation for understanding crime patterns in the coastal zones of Kollam and Thiruvananthapuram districts, offering valuable insights for informed decision-making in both law enforcement and urban planning endeavors. **2.3 Data Analysis and Interpretation:** The results obtained from the cluster and outlier analysis, as well as the IDW visualization, were subjected to rigorous scrutiny and interpretation. This comprehensive analysis aimed to gain a deeper understanding of the underlying spatial dynamics of crime within the investigated coastal zones, and to glean valuable insights that can inform future crime prevention strategies and urban planning initiatives.

Analysis of Spatial Patterns: This analysis reveals distinct spatial patterns of crime distribution across the studied police stations:

1. High-High (HH) Cluster:

• Stations: Valiyamala, Aryanad, and Aruvikkara.

• Characteristics:

• Positive Lmi index (above 1) indicating spatial clustering of high crime values.

• Significant Z-scores (above 1.96) indicating statistically significant clustering at the 95% confidence level.

Low p-values (below 0.05) confirming the statistical significance.
Interpretation: These stations exhibit a concentrated area of high crime activity.

2. Low-Low (LL) Cluster:

o Stations: Vellarada and Kadakkavoor.

o Characteristics:

• Negative Lmi index (below -1) indicating spatial clustering of low crime values.

• Significant Z-scores (below -1.96) indicating statistically significant clustering at the 95% confidence level.

• Low p-values (below 0.05) confirming the statistical significance. o Interpretation: These stations demonstrate a concentrated area of low crime activity.

3. Low-High (LH) Outlier:

• Station: Vattappara.

• Characteristics:

• Negative Lmi index (below -1) suggesting lower crime than its surroundings.

o Interpretation: Vattappara exhibits a relatively low crime rate compared to the surrounding areas, making it an outlier within the broader context.

4. Potential Transitional Zones:

o Stations: Chirayinkeezhu and Kollam East.

o Characteristics:

• Lmi indices close to 0 and non-significant Z-scores.

o Interpretation: These stations might represent transitional zones between high and low crime areas, as they lack statistically significant clustering.

Overall, the spatial pattern highlights distinct clusters of high and low crime activity, with Vattappara emerging as an outlier with a low crime rate surrounded by higher crime areas. Stations in Chirayinkeezhu and Kollam East might be undergoing transitions in their crime patterns, requiring further investigation.

This analysis provides valuable insights into the localized nature of crime within the study area. Further exploration of the underlying factors contributing to these patterns, such as socio-economic characteristics, environmental factors, and police deployment strategies, could offer deeper understanding and inform targeted interventions for crime prevention.

Analysis of Spatial Relationships: The provided data offers insights into the spatial relationships between crime patterns across the studied police stations, as indicated by the following:

1. Spatial Lag:

o This column reflects the average crime rate in the neighboring areas for each station.

o Positive values: Suggest the station is surrounded by areas with similar crime rates.

• This is evident in stations like Valiyamala (1.407154), Aryanad (1.228076), and Aruvikkara (0.969504), which belong to the HH cluster and are surrounded by other stations with high crime rates. o Negative values: Suggest the station is surrounded by areas with contrasting crime rates.

• This is observed in Vattappara (-1.192847), which has a relatively low crime rate compared to its LH outlier classification and is surrounded by stations with higher crime rates.

2. Z-Transform:

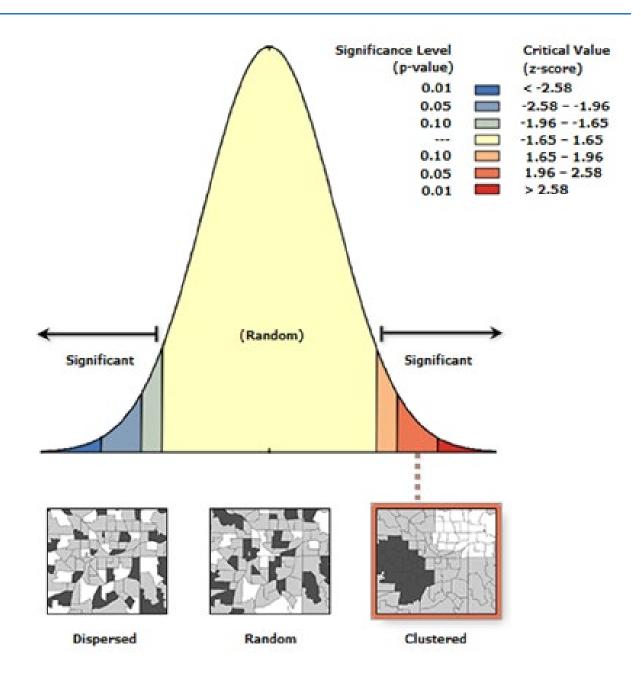
o This column standardizes the spatial lag values, allowing for easier comparison across stations with varying numbers of neighbors.

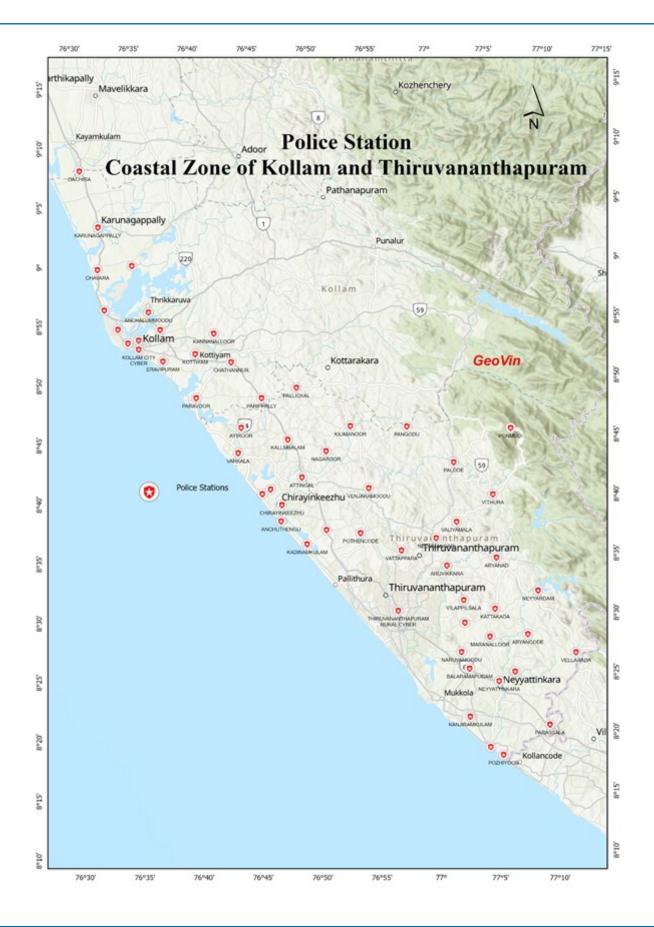
• Overall, the analysis of spatial relationships reveals that stations within the same cluster (HH or LL) tend to be surrounded by areas with similar crime rates, while outliers like Vattappara are situated in areas with contrasting crime patterns.

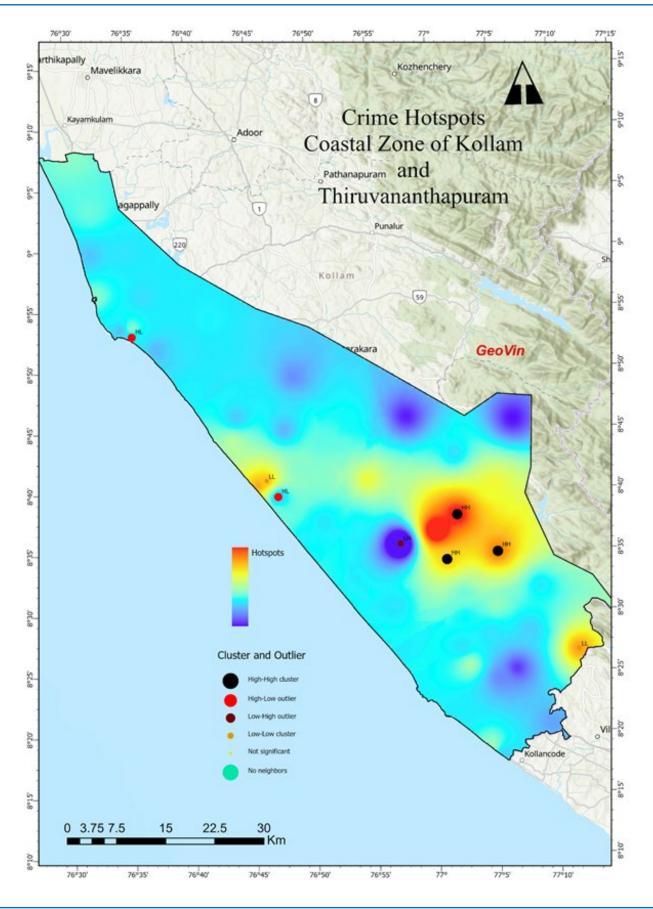
• It's crucial to remember that these are just preliminary observations based on the provided data. A more comprehensive understanding of spatial relationships could be achieved by:

• **Mapping:** Visualizing the spatial distribution of crime rates and stations using color gradients or symbols representing different crime categories. This could provide a clearer picture of how crime patterns cluster and how individual stations relate to their surroundings.

• Geospatial analysis: Investigating the correlation between crime patterns and other geographic factors (e.g., proximity to main roads, distance to police stations, socio-economic indicators of surrounding areas) could reveal potential contributing factors influencing the observed spatial relationships.







| | | ture the shore | 1 | Luci D Value | 00 T | | 7 7 | Custial Las |
|---------------------------------|-----------------|-----------------------|-------------|--------------|---------|-------------|-------------|-------------|
| Police Station ANCHALUMMOODU | Number of Crime | Lmi Index | Lmi Z Score | Lmi P Value | СО Туре | N Neighbors | Z Transform | Spatial Lag |
| | 1916 | -0.053282 | -0.998583 | 0.146 | | 13 | 0.207317 | -0.26129 |
| ANCHUTHENGU | 1297 | 0.161386 | 1.115806 | 0.138 | | 6 | -0.319863 | -0.512957 |
| ANCHUTHENGU COASTAL | 9 | 1.051335 | 1.568716 | 0.054 | | 6 | -1.416807 | -0.754413 |
| ARUVIKKARA | 2779 | 0.898591 | 2.869429 | 0.006 | HH | 8 | 0.942304 | 0.969504 |
| ARYANAD | 2840 | 1.201005 | 3.606974 | 0.004 | HH | 7 | 0.994256 | 1.228076 |
| ARYANCODE | 952 | -0.098918 | -0.440737 | 0.326 | | 7 | -0.613687 | 0.163873 |
| ATTINGAL | 601 | 0.223951 | 0.733114 | 0.244 | | 7 | -0.912621 | -0.249483 |
| AYIROOR | 2050 | -0.110737 | -0.765763 | 0.218 | | 5 | 0.321441 | -0.350243 |
| BALARAMAPURAM | 3020 | 0.332832 | 0.725334 | 0.218 | | 7 | 1.147555 | 0.29487 |
| CHATHANNUR | 1600 | -0.009258 | -0.349331 | 0.356 | | 5 | -0.061809 | 0.152283 |
| CHAVARA | 2351 | -0.124567 | -0.477558 | 0.31 | | 5 | 0.577792 | -0.219185 |
| CHAVARA THEKKUMBHAGAM | 1237 | -0.031425 | -0.146973 | 0.42 | | 5 | -0.370963 | 0.086123 |
| CHIRAYINKEEZHU | 1994 | -0.201677 | -1.880838 | 0.016 | HL | 7 | 0.273747 | -0.749005 |
| ERAVIPURAM | 2598 | -0.169723 | -0.764316 | 0.232 | | 12 | 0.788153 | -0.218931 |
| KADAKKAVOOR | 89 | 0.988472 | 1.721649 | 0.04 | LL | 7 | -1.348674 | -0.745137 |
| KADINAMKULAM | 833 | 0.016395 | 0.064503 | 0.482 | | 5 | -0.715035 | -0.023311 |
| KALLMBALAM | 2169 | -0.179508 | -1.311312 | 0.092 | | 10 | 0.422789 | -0.431658 |
| KANJIRAMKULAM | 1686 | 0.000113 | -0.008348 | 0.468 | | 7 | 0.011435 | 0.010042 |
| KANNANALLOOR | 1835 | 0.02898 | 0.38163 | 0.322 | | 4 | 0.138333 | 0.212988 |
| KARUNAGAPPALLY | 2684 | 0.15902 | 0.372218 | 0.352 | | 3 | 0.861396 | 0.187684 |
| KATTAKADA | 4027 | 0.15902 | 0.372218 | 0.32 | | 10 | 2.005181 | 0.134046 |
| KILIKOLLOOR | | | | | | 10 | | -0.213236 |
| KILIKOLLOOR | 1676 2425 | -0.000612 0.051429 | -0.79446 | 0.208 | | 5 | 0.002918 | |
| KOLLAM CITY CYBER | | | 0.18536 | | | | | 0.081594 |
| KOLLAM EAST | 0 | 0.0757 | 0.163893 | 0.452 | | 11 | -1.424472 | -0.054028 |
| | 3125 | -1.144307 | -1.607536 | 0.042 | HL | 11 | 1.23698 | -0.940499 |
| kollam traffic | 222 | 0.23902 | 0.672299 | 0.24 | | 11 | -1.235402 | -0.1967 |
| KOLLAM VANITHA | 0 | 0.0757 | 0.132867 | 0.494 | | 11 | -1.424472 | -0.054028 |
| KOLLAM WEST | 2100 | -0.158683 | -1.330155 | 0.096 | | 10 | 0.364024 | -0.443179 |
| κοττιγαμ | 1994 | -0.020047 | -0.257721 | 0.42 | | 11 | 0.273747 | -0.074451 |
| MALAYINKEEZHU | 1755 | 0.035796 | 1.275681 | 0.112 | | 7 | 0.070199 | 0.518421 |
| MANGALAPURAM | 2052 | -0.019952 | -0.169342 | 0.426 | | 6 | 0.323144 | -0.062771 |
| MARANALLOOR | 1578 | -0.03794 | -1.395662 | 0.07 | | 8 | -0.080545 | 0.478894 |
| MARAYAMUTTOM | 462 | -0.555944 | -1.452159 | 0.082 | | 9 | -1.031003 | 0.548213 |
| NAGAROOR | 2199 | 0.135666 | 0.623715 | 0.262 | | 4 | 0.448339 | 0.307641 |
| NARUVAMOODU | 1596 | -0.037473 | -1.580618 | 0.068 | | 8 | -0.065215 | 0.584188 |
| NEDUMANGAD | 6010 | 1.828709 | 1.38169 | 0.09 | | 4 | 3.694032 | 0.503295 |
| NEENDAKARA COASTA | 13 | 0.234212 | 0.692262 | 0.256 | | 11 | -1.4134 | -0.16847 |
| NEYYARDAM | 1666 | -0.004226 | -1.47732 | 0.074 | | 3 | -0.005599 | 0.76748 |
| NEYYATTINKARA | 3639 | -0.329062 | -0.463766 | 0.322 | | 6 | 1.674735 | -0.199761 |
| OACHIRA | 2012 | 0.244928 | 0.80093 | 0.184 | | 1 | 0.289077 | 0.861396 |
| PALLICKAL | 449 | -0.300903 | -0.383504 | 0.328 | | 3 | -1.042074 | 0.293566 |
| PALLITHOTTAM | 1703 | -0.01369 | -0.945717 | 0.17 | | 11 | 0.025913 | -0.537119 |
| PALODE | 2214 | -0.036762 | -0.082065 | 0.486 | | 4 | 0.461114 | -0.081053 |
| PANGODU | 356 | -0.609919 | -0.859768 | 0.19 | | 2 | -1.121279 | 0.553015 |
| PARASSALA | 2145 | -0.239739 | -1.241999 | 0.1 | | 4 | 0.402349 | -0.60578 |
| PARAVOOR | | | | | | | | |
| PARAVOOR | 1705 | 0.00761 | 0.675273 | 0.244 | | 5 | 0.027616 | 0.280158 |
| | 1689 | -0.003621 | -0.61024 | 0.29 | | | 0.01399 | -0.263169 |
| PONMUDI | 20 | -0.63835 | -0.399164 | 0.242 | | 1 | -1.407438 | 0.461114 |
| POOVAR | 1420 | 0.300439 | 1.311318 | 0.068 | | 4 | -0.215108 | -1.419966 |
| POOVAR COASTAL | 4 | 0.300319 | 0.263611 | 0.31 | | 4 | -1.421065 | -0.214856 |
| POTHENCODE | 2150 | 0.019176 | 0.123856 | 0.444 | | 4 | 0.406607 | 0.047947 |
| POZHIYOOR | 1834 | -0.080381 | -1.085299 | 0.132 | | 4 | 0.137481 | -0.594417 |
| SAKTHIKULANGARA | 1532 | 0.03763 | 1.18487 | 0.09 | | 12 | -0.119722 | -0.319548 |
| THIRUVANANTHAPURAM RURAL CYBEI | | -0.024725 | 0.000589 | 0.49 | | 3 | -1.424472 | 0.017646 |
| VALIYAMALA | 2940 | 1.494012 | 3.317372 | 0.004 | HH | 6 | 1.079422 | 1.407154 |
| VARKALA | 137 | 0.399799 | 0.711426 | 0.236 | | 5 | -1.307794 | -0.3108 |
| | 341 | -1.330577 | -2.535053 | 0.02 | LH | 5 | -1.134054 | 1.192847 |
| VATTAPPARA | | | | 0.05 | | 2 | -1.368262 | -0.799831 |
| VATTAPPARA VELLARADA | 66 | 1.076437 | 1.332071 | 0.05 | LL | 2 | -1.308202 | |
| | 66 3480 | 1.076437 0.677721 | 1.332071 | 0.05 | LL | 4 | 1.539321 | 0.44761 |
| VELLARADA | | | | | | | | |

Key Findings

Reference

• **Spatial Patterns:** Distinct clusters and outliers emerged, highlighting concentrated areas of high (*Valiyamala, Aryanad, Aruvikkara*) and low (*Vellarada, Kadakkavoor*) crime activity. Vattappara emerged as an LH outlier, exhibiting a relatively low crime rate surrounded by higher-crime areas. *Chirayinkeezhu* and Kollam East might represent transitional zones between high and low crime areas.

• **Spatial Relationships:** Stations within the same cluster (HH or LL) tended to be surrounded by areas with similar crime rates, while outliers like *Vattappara* were situated in areas with contrasting crime patterns.

• **Trends:** While not explicitly evident from the provided data, further analysis incorporating temporal data could reveal trends in crime rates across the study area.

Implications: These findings offer valuable insights for informing crime prevention strategies within the study area. Focused interventions can be directed towards the identified hotspots of high crime activity (HH cluster). Additionally, understanding the unique context of *Vattappara* (LH outlier) and potential transitional zones (*Chirayinkeezhu* and Kollam East) necessitates further investigation to tailor appropriate crime prevention measures.

- 1. Anselin, L. (1995). Local indicators of spatial association— LISA. *Geographical analysis*, 27(2), 93-115.
- 2. Mitchell, A. (1999). *The ESRI guide to GIS analysis* (Vol. 3). Environmental Systems Research Institute.
- 3. Piacentino, D., Arbia, G., & Espa, G. (2021). Advances in spatial economic data analysis: methods and applications. *Spatial Economic Analysis, 16*(2), 121-125.
- Wang, D., Ding, W., Lo, H., Morabito, M., Chen, P., Salazar, J., & Stepinski, T. (2013). Understanding the spatial distribution of crime based on its related variables using geospatial discriminative patterns. *Computers, Environment and Urban Systems, 39*, 93-106.
- 5. "van Sleeuwen, S. E., Ruiter, S., & Steenbeek, W. (2021). Right place, right time? Making crime pattern theory timespecific. *Crime Science*, 10, 1-10.
- 6. Hipp, J. R., & Williams, S. A. (2020). Advances in spatial criminology: The spatial scale of crime. *Annual Review of Criminology*, *3*(1), 75-95.
- 7. Vineesh, V. (2024). Exploring Spatial Statistics for Cluster and Outlier Analysis in Crime Patterns: A Geospatial Study of Coastal Zones in Kollam and Thiruvananthapuram.

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