

Geospatial Crime Analytics: A GIS-Based Approach Towards Prediction of Crime Hotspots

Supriya Panigrahy^{1,*}

*Department of Computer Science and Engineering, College of Engineering and Technology,
Bhubaneswar, India, 751003*

Abstract

GIS has proven its proficiency in many fields but still require some more observations. The spatial interpolation which has seen widespread used in diverse domains with multiple applicability. Currently, some commercial GIS or analytics program packages now provide spatial interpolation technique including inverse distance weighting (IDW), kriging, triangulated irregular networks (TIN) interpolation, kernel density interpolation (KDE). Further, the two spatial autocorrelation techniques i.e., also known as LISA indicators that comprised of Moran's I and Getis-Ord G_i^* techniques. Amalgamation of spatial interpolation and spatial autocorrelation are being shown in this paper. For the implementation, related enabling technologies are summarized. Further, this paper entailed the literature survey of multiple papers comprising the associated works carried out in the domain of spatial interpolation and spatial autocorrelation areas of GIS. The proposed framework is a generalized framework with four tiers to realize geospatial systems. This paper scrutinized geospatial crime dataset for understanding the proposed framework realistically. Numerous results are produced to prove the efficiency of the proposed model. The paper entailed weighted-overlay analysis, symbological heatmap analysis, choropleth analysis, kernel density interpolation, and Getis-Ord G_i^* spatial autocorrelation analysis implemented to practically realize the essence of geospatial crime an-

*Corresponding Author

Email address: supriyapanigrahy1996.sp@gmail.com (Supriya Panigrahy)

alytics. Finally, future research directions are recapitulated, where the task of interpolation and autocorrelation might be merged with techniques of artificial intelligence, machine learning, and cloud computing paradigm, serverless computing framework and many such in order to enhance the proposed structure with lessened restrictions.

Keywords: Geographic Information Systems (GIS), Weighted-Overlay, Choropleth Analysis, Kernel Density interpolation, Getis-Ord G_i^* Autocorrelation.

1. Introduction

2 With the use of GIS (geographic information system), we can open up our
3 own custom maps on our personal computers, or add new data to existing maps,
4 each of which would give us the freedom to display several different ways of
5 seeing or create additional map features are just as important as critical as data
6 storage and maintenance, for the proper expansion of the amount of geographic
7 information and thus making it easier to use (Geographic data Systems). It
8 might be very recent, coming into being during the 1970s, but GIS has expanded
9 tremendously over the years will be available to only to certain businesses and
10 colleges who can afford expensive equipment to provide computerized geographic
11 information systems (GIS). Today, almost anyone who has a personal device
12 is using geo-coding tools, whether it's a desktop or a home computer. More
13 recently, GIS applications have been made simpler to use. There's no need for
14 a huge amounts of training; in fact, it is comparatively simple even for a novice
15 and beginner users. GIS is no longer just software: managing and processing
16 geographic information; it may also include managing and handling any and/all
17 geographic information. //

18 Geographic Information Systems (GIS) has emerged as a prime technology
19 is diverse domains such as disaster management, susceptibility assessment, epi-
20 demiological analysis and many such. In a cartographic viewpoint, GIS plays
21 a key role in designing maps and corresponding vector files [1, 2, 3]. The pro-

22 efficiency of spatial interpolation techniques as well as spatial autocorrelation
23 techniques in identifying heatmaps, hotspots, and coldspots has led to immense
24 rage these days [4, 5]. Interpolation is a widely used method to classify data
25 points in order to identify dense and sparse regions based on pseudo-color inten-
26 sities [6]. Further, spatial autocorrelation techniques are used to identify risk
27 zones (hotspots and coldspots) [7, 8].

28

29 As is critical in designing and making charts, it also has a crucial tasks when
30 it comes to making vector files. To make an interpolation, we have to presume
31 that the information is available, or as long as there is no available information
32 we can use some mathematical tool to make an approximation. A often used
33 strategy in expanding strings is to draw on distinct data to provide a more con-
34 tinuous form. And we switch from a vector to a raster, for interpolation. And,
35 thus, input data can be in the form of a point, line, or it can be in the form of a
36 waveforms. If we have identified these specific shapes on the computer graphics
37 programmers have open, it's simple to take those contour points and translate
38 them to continuous coordinates. We will have to expand on the definition of
39 contour because our point representation, whether discrete or continuous, is re-
40 quired.

41

42 In GIS, spatial interpolation can be used to measure any surface property at
43 a specified point, to create contours for viewing data graphically, and it's widely
44 used as a tool in spatial decision-making processes including terrain analysis, hy-
45 drology, resource prospecting, hydrocarbon discovery, and crime analysis. Until
46 going through various interpolation strategies, it's important to go through the
47 variations in surface representation methods; each representation is useful in
48 a particular circumstance. Furthermore, A grid representation of a surface is
49 called a functional surface since it holds only a single z value rather than mul-
50 tiple z values for every given x, y position, and the surface values are constant
51 rather than discrete.

52

53 The global expanders define a single feature which is applied to the entire
54 field, for example, for trend surface illustration and example, for drawing an
55 ascending or descending line trend. For example, the use of local interpolator's
56 sample-weighted vectors is done repeatedly on a smaller subset of the total num-
57 ber of points, for instance, IDW. Exact interpolators preserve the input data
58 while approximates allow for approximation. The more random approaches use
59 the principle of non-fit surfaces (in several ways close to linear regression, in-
60 cluding a best-fit plane) and the less random ones do not use probability theory.
61 Other expanders allow for faults (such as those), and to appear inside the data,
62 whereas the Smooth interpolators provide a smooth surface. Mathematical and
63 numerical (and geometric) interpolation is two kinds of interpolation, one of
64 which is linear and the other is non-linear. More importantly, interpolation
65 strategies can be divided into the following five groups:

66

67 • Kriging is a geostatistical method for computing the magnitude of a ran-
68 dom field (e.e. elevation, z) at an unknown position by interpolating
69 between observed values. Daniel Krige's Master's thesis helped refine and
70 improve the technique of interpolation and extrapolation, and was used
71 by the French mathematician Georges Matheron in the analysis of their
72 ideas. Global expansion is gradual, smooth, or sudden, or local; Kriging
73 is stochastic, exact, or precise, or stochastic.

74

75 • Shortened inverse distance Point-weighted interpolation: This type of
76 weighted average includes the weighting effect on the interpolation method:
77 calculates cell values by combining the values in the neighborhood of the
78 samples. To and subject cell value, an equal distance from the closer a
79 point has in the cell, the formula, the more power or weight it has in the
80 averaging.

81

82 • Spline Interpolation: This technique uses a mathematical function that

83 avoids the effect of total curvature, and ensures that the input points
84 lie on a smooth surface, allowing the calculated values to flow precisely
85 through the points.

86

87 • Triangulated Irregular Network: The TIN (3D geodatabase) is a data
88 structure that is used in GIS to illustrate a surface, meaning it's a dis-
89 torted in shape and uses a geometric, equal-area Cartesian grid. or A,
90 TINs can be produced by either by conversion of a functional surface data
91 or by the application of a pattern recognition technique to another func-
92 tional surface.

93

94 • Kernel Density Interpolation: Instead of using a parametric function for
95 calculation of likelihood density and using it, K-nearest is an alternative
96 that doesn't create a function for data smoothing. Spatial autocorrelation
97 tests the association with its constituent values across the magnitude of
98 the region under consideration.

99

100 The result of spatial autocorrelation can be either helpful or detrimental.
101 While people are seeing the same experiences in the same locations, they ap-
102 pear to be in agreement; yet when circumstances are different, they tend to
103 disagree.

104

105 • Moran's I: I believe that the most frequently employed strategy in spatial
106 autocorrelation is I is ordinary world autocor (i.e. a unique combina-
107 tions of events, each world, and native spatial autocor), Individuals of the
108 American Indian descent, Native Americans may consider their quality
109 of life to be good, may have higher cultural and financial standards than
110 other people of the United States. Correlation coefficients in Moran arises
111 from the whole universe of contrasts, which results in variances which are

112 greater than those of the interaction between the 2 conjoint variables is
113 measured in terms of their respective utility.

114

- 115 • Getis-Ord: In general Getis-Ord G can have a large value, which indicates
116 a correlation for prime averages of the strength, while an occasional value
117 indicates a correlation for low intensity values. It is important to provide
118 spatial autocorrelation.

119

120 Since it's often interpreted as suggesting that there's anything of concern
121 inside the distribution of map values that necessitates more analysis in order to
122 comprehend the reasons for the discovered spatial variance, Furthermore, the
123 existence of spatial autocorrelation means data redundancy, which has impor-
124 tant repercussions for spatial knowledge processing methodology.

125

126 The rage of GIS technology in the domain of computer applications has led to
127 immense applicability in sectors of disaster management, epidemiological anal-
128 ysis, susceptibility mapping and many such. This paper entails a novel strategy
129 to converge spatial interpolation techniques with spatial autocorrelation meth-
130 ods in order to perform proper risk as well as zonal analysis. The proficiency of
131 GIS technology has been recently proved in risk and zonal analysis of various
132 prevailing epidemics as well as disasters and crime analysis. But, some lacu-
133 nae still persists in providing predictive competency to the model. This paper
134 tries to overcome such hurdles through a proposed framework on convergence of
135 spatial interpolation and spatial autocorrelation. The outcome of the proposed
136 framework is to provide appropriate risk analysis along with zonal analysis by
137 identification of hotspots as well as coldspots.

138

139 This is a realistic representation of the system with geospatial analysis, or
140 weighted overlay, weighted factors present the main challenge of measuring crime
141 locations in the United States, particularly in the initial stage. In a region-

142 specific way, a mapped weighted average shows crime per capita. To distinguish
143 crime hotspots based on their severity, a symbological graded heat map was cre-
144 ated where crime rate varies geographically few estimates are given. In addition,
145 a choropleth map, which is simply a collection of evenly-spaced points joined
146 to the vector dataset is shown. An obvious goal of geospatial crime analytics is
147 to use kernel density to calculate the interpolation. The radius is set to 5 units
148 and an interpolated map is created. Additional, Getis-Ord G_i^* , a local measure
149 of spatial interaction, denotes spatial autocorrelation Spatially auto-correlated
150 can be classified as hotspots and coldspots based on assigned zone scores. The
151 findings reveal a well-grounded study of crime data on law enforcement proce-
152 dures and other policies implemented in the U.S. in different areas. Often, the
153 findings may be charted on a zone-based study of U.S. criminal statistics.

154 **2. Related Works**

155 *2.1. Spatial Interpolation and Autocorrelation in Crime Hotspot Mapping*

156 Ansari and Kale [9] discussed that Crime detection has been an important
157 feature of GIS, where analysts utilized the geographical and temporal aspects of
158 crime data for analysis and forecasting utilizing the Hotspot analysis method.
159 A hotspot is a region where there is a high incidence of crime. There are many
160 types of Hotspot detection techniques, including Spatial Analysis, Interpolation,
161 and Spatial Autocorrelation, that can be used to locate a crime hotspot. The
162 methods of Kernel Density Estimation, Inverse Distance Weighted, and Getis-
163 Ord G_i^* from each class are explored in this article. These techniques were used
164 in Aurangabad, Maharashtra, India, to identify hotspots for crime. Recent
165 crimes indicate that the requirement for a massive expansion of facilities to de-
166 tain criminals. Cops have an obligation to stay in a step ahead of the suspects,
167 who are perpetually re-veining the field with police by trying to keep ahead of
168 the lawbreakers, who always want to surpass their efforts. This paper proved
169 that KDE (Kernel Density Estimation) and IDW (Inverse Distance Weighting)
170 for spatial interpolation and Getis-Ord G_i^* for spatial autocorrelation are more

171 precise than other techniques.

172

173 Kennao et al. [10] summed up that Policing authorities aren't good at fore-
174 casting where specific crimes will arise in the future; As a result, geography
175 plays an important part in law enforcement and criminal justice. The author
176 significantly highlighted the most identified crime in the police stations of Alla-
177 habad city via a thematic chart that police departments may use in this analysis.
178 Author and the group decided by an interpolation process that a crime hotspot
179 analysis chart would include information on offences. The police force must
180 identify and describe sources of crimes for the information they want in order
181 to stay ahead of criminal behavior patterns. The interpolation process (IDW)
182 was used to create a hotspot chart to show the concentration of crime (both
183 crimes against life and crimes against property) in each police station. The
184 police department's crime problems necessitate the use of computerized crime
185 mapping for research. Crime analysis plays a crucial role in devising solutions
186 in formulating crime preventions strategies and crime problems. For the form of
187 crime investigated in this paper, the jurisdictions of Doomanganj Police Station
188 and Kernal Ganj Police Station need increased patrolling and successful steps.

189

190 Jakobi and podor [11] analysis aimed to determine whether human charac-
191 teristics such as gender or age have an impact on crime fear, suggesting, for
192 instances, that aged people are more fearful of things that young people are less
193 afraid of, or that women might experience fear in different places than men.
194 Using visual sketch maps and predictive GIS tools, this analysis analyses ap-
195 prehension of crime experience and social crime statistics in a spatial sense.
196 By analyzing the outcomes of selected big, medium, and small sized Hungarian
197 towns, the research aimed to assess causal reasons for fear of crime. There was a
198 web-application where people could identify locations on a chart where they felt
199 secure, allowing them to track crime information that was available. To make
200 these interactive maps comparable, GIS technology was used to process the re-
201 sults. Such items and others were mirrored by the grid-based model. According

202 to the findings, there were papers of both coincidences and reverse associations
203 between crime rates and perceptions of dangerous locations, demonstrating the
204 role of location in crime fear study. Furthermore, the findings revealed that
205 while the scale of the area or the gender of the respondent had no substantial
206 impact on total judgments of locations, the absolute amount of protected mark-
207 ings and the local number of recorded crime incidents can have an impact on
208 local outcomes.

209

210 Hu et al. [12] epitomized in the paper to track geographic distributions of
211 historical crime cases and forecast possible crime trends, various spatial analy-
212 sis approaches have been used. The first approach counts crime cases and then
213 measures statistics based on regional areas including census units. The multi-
214 variate regression approach is then used to investigate the association between
215 crime rates and a variety of crime attraction and inhibition factors, including
216 social circumstances, community populations, land use types, cultural beliefs,
217 and drug misuse records. In hotspot policing, predictive hotspot analysis is im-
218 portant. Existing techniques, such as the often used kernel density estimation
219 (KDE), do not take the temporal component of crime into account. The article
220 suggested a spatio-temporal system for predictive hotspot mapping and assess-
221 ment, building on previous studies in related fields. In comparison to previous
222 work in this area, the proposed structure has four significant characteristics:

223

- 224 • A spatio-temporal kernel density estimation (STKDE) technique is used
225 to incorporate the temporal component into predictive hotspot mapping,
- 226 • a data-driven optimization technique called likelihood cross-validation is
227 used to choose the most appropriate bandwidths,
- 228 • a statistical significance test is used to exclude false positives in the density
229 estimates, and
- 230 • a brand new benchmark was commenced to project precision inventory .

231

232 Ristea et al. [13] have shown that society's level of real and suspected vi-
233 olence in metropolitan environments is critical and that of crime is crucial in
234 maintaining a healthy community. It turns out the, that fear of crime occurs
235 at a greater pace than recorded incidents of crime. Urban decay describes the
236 issue where a neighbor-hood declines due to poverty and related issues. Geo-
237 graphic Information Science (GISc) and the use of geospatial technologies have
238 greatly benefited spatial crime detection and perceived (crime) protection anal-
239 ysis. This paper used an innovative analytical methodology to chart the usage
240 of new geospatial technology to investigate perceived community protection in
241 Baton Rouge, Louisiana, through the lenses of fear of crime and crime inter-
242 pretation. A survey, spatial video geonarrative (SVG) in the field with research
243 subjects, and the extraction of moments of tension (MOS) from bio sensing
244 wristbands are among the mixed techniques used. The participants in this re-
245 search were 46 people who completed geonarratives and MOS identification.
246 Here is a selection of ten of these geonarratives, Each participant was driven
247 along a predetermined route in a car fitted with audio recording and spatial
248 video while wearing Empatica E4 wristbands that measured three physiological
249 variables, all of which were connected by timestamp. The findings reveal gender-
250 based disparities in participants emotions (positive or negative) and MOS in the
251 area. This mixed-methods approaches are promising for establishing links be-
252 tween real crime occurrences and neighborhood perceptions of crime in urban
253 communities.

254

255 Yang [14] asked for four spatial trends in violent crime to travel to be ana-
256 lyzed:

257

- 258 • Analyses of street violence and learning environment patterns across schools
259 since the safe passage program was established,
- 260 • creating a web-enabled mobile application to allow researchers and resi-

- 261 dents to collect and share GIS data on the situation,
- 262 • enhancing a research network of shared information about it, and using
263 it to drive robotic vehicles to patrol these locations (to search for specific
264 patterns and locations),
 - 265 • undertaking a process to gather and analyze geo-referenced-maps of areas
266 that will provide them with these findings.

267 Via this study the author discovered that, while the amount of secure routes
268 have risen in these regions, hotspot rates for gun violence have also become
269 worse over the years. GIS has then serves as a great foundation for community-
270 based study initiatives. In his paper, Yang looked at gun-related shootings in
271 each of the four communities. To that purpose, ArcGIS Online, a cloud-based
272 visualization tool, was used to build maps for gun-related crime counts. It's im-
273 portant to communicate scientific findings or geospatial data with the general
274 public because it allows them to engage in current studies in their communities.
275 As a result of this study, an interactive map was created that included neigh-
276 borhood services as well as the results of the spatial analysis and statistics. The
277 author illustrated Numeric data with region details is transformed into spatial
278 data with geographic coordinates using GIS-based charts. GIS is an excellent
279 forum for both urban populations and academics, since it offers a geographic
280 overview of group partnerships and hotspots, along with local infrastructure
281 and crime statistics.

282

283 Ristea and Leitner [15] highlighted few research articles having devoted dis-
284 cussions to approaches that can address the question known police patrol route
285 problems, the two main ones being traffic enforcement (the issue here) and
286 population monitoring (the issue being ignored). Crime prevention and law en-
287 forcement are also needed because the police would make sure that no criminals
288 are found, but can tend to crimes if they happen. They laid out the points,
289 from the street (or neighborhood) down to the city of Milwaukee, arguing that
290 if broken windows are not cleaned up, crime would increase (US). It is proposed

291 that further study of this hypothesis is required for many factors that include:

292

- 293 • violent broken window enforcement against marginalized populations;
- 294 • disparities in crime intensity,
- 295 • disturbance styles and their dominance in crime; and ,
- 296 • the ethnic stratification in neighborhoods.

297 This study explored the question of whether there is a relationship between
298 crime and disorder, as other studies in the authors have studied in the near
299 past, but has focused on the different details, and found it to be only partly
300 applicable. However, it seems that the result is to be restricted to public area
301 disturbance, with a more powerful bearing on the second category of offences,
302 which correlates with less severe violence in the community as a result, rela-
303 tionships with the city should resolve problems with neighborhood incivility are
304 commonly practiced. The authors developed online and interactive maps, that
305 be used to survey a variety of the communities that include these medium and
306 small cities in Hungary, as well as Hungary's larger and smaller municipali-
307 ties. The aim of this analysis was to find out why people perceive the safety
308 in lab studies as important. Overall, the results are only quite mixed because
309 of the similarities and disparities within studied metropolitan environments. It
310 appears to be the residents opinions about safety and how often criminal activ-
311 ities happenings are happening which shape their beliefs about the security of
312 a particular area. The cities and the scale do not seem to affect the judgment
313 is insignificant in relation to the population size; on the other hand, gender is
314 considered important for the total number of respondents Expanding on this,
315 the findings show the significance of the local view in interpretation of protec-
316 tion testing. The author were studying the improvement in spatial precision for
317 predicting crime in the case of burglaries in Wuhan Houkao involves analyzing
318 an extension of related concepts. They put forward a more multidimensional
319 crime model, as the standard one only accounts for temporal and spatial aspects,

320 anisotropic diffusion (AnisDM), since it contains external variables, which are
321 able to vary in location.

322

323 Mohammed and Baiee [16] explored that there are so many places where
324 crime must be researched and studied for someone to examine and analyze it
325 carefully in just one place. For the past few years, there has been an increased
326 social, socioeconomic, and person dysfunction, which is having a measurable
327 impact on crime rates. This has made the collection of information regarding
328 crime more challenging than it was used to be as a result of modern practices
329 and procedures being brought in to bear. Crime prevention can be aided by a
330 great advancements in our day, including GIS and can be a helpful in providing
331 alternative solutions to locate better solutions. The paper illustated an increase
332 in crime rates in developed countries is mostly attributable to the uneven distri-
333 bution of psychological and economic hardship. This study seeks to classify the
334 crime-probation areas and examine the historical crime data to ascertain where
335 criminal activity may concentrate during the next four years, with the goal of
336 gauging their tendency to create crime patterns and predicting them spatially.
337 Both of the author have put forward their efforts in Examination of crime data
338 using GIS enables law enforcement to recognize and monitor trends that al-
339 low for flexibility in implementing spatial crime control strategies and makes
340 it simple to visualize crime clusters and move the crimes from one area to an-
341 other. The most helpful method of localizing the location of criminal incidents
342 for crime trend identification is through means of GIS; a viable approaches to
343 finding criminal hotspots include spatially expanding one's boundaries in order
344 to criminal pattern details and crime correlation; detection of potential patterns
345 is done by examining criminal incidents from a geographic perspective. Using
346 spatial correlation, the Getis-Ord G_i^* resulted in the conclusion that the crime
347 data are used to plot statistically relevant crime hotspots to find locations with
348 high and low concentrations as well as identified less dense areas of criminal
349 activity for increased detail with more effectiveness using a GIS. This study,
350 with the help of the Grid network, was used to identify the danger locations

351 of violence like Shooting, Homicide, and Assault as well as crime hotspots in
352 the state of Baltimore, Maryland. The article explored that GIS is capable of
353 visualizing and temporal relations including violence, such as incidence, in ad-
354 dition to spatial features, is often capable of being used to forecast events based
355 on the method by which the areas are used. Ensconced with delivering un-
356 derstandable, real-life-world data in the decisions that enable multi-application
357 comprehension, even those inside Defense, and in disaster scenarios, the foun-
358 dation provides users with focused, practical applications. While dynamic or
359 expandable Hotspot and zone maps can also be represented using Getis-Ord G_i^*
360 these are generally considered dangerous.

361
362 Through this research Hashim et al. [17] attempted to identify high-risk
363 areas for opioid activity hotspots to avoid accidents and improve emergency
364 response times. When we use "hotspot" to describe a location with a higher
365 density of information and it occurs in the geographic system, we use the term
366 "geo". The first phase in the analyst's search for problems associated with drug
367 distribution is locating these hotspots, when each area can or may be different
368 from others. Use of visualization and statistical analysis of illicit drug use pat-
369 terns can assist in locating the high concentration of users of illicit drugs in re-
370 gional areas. In this research authors emphasized the drug use is much stronger
371 and is affecting the population's well-being. people have a habit of concluding
372 that approximately 133,684 to 1/4 (or 20,684 individuals) of the population in
373 Malaysia is addicted to drugs. Furthermore, the federal prison population of
374 56% consists of individuals who are serving sentences for drug-connected crimes.
375 This research is intended to classify the parts of Selor, Malaysia that are par-
376 ticularly problematic for opioid users. Using three separate statistical methods,
377 kernel density (KDE), Getis-Ord G_i , and IDW mapping, the researchers aimed
378 to illustrate the concentrations of opioid usage among the various addicts. The
379 hotspot algorithm calculates density-depth curves using the Gaussian kernel
380 density estimate (KDE), and Getis-Ord G_i^* and IDW. The findings indicated
381 that there are eight statistically important areas with high rates of opioid use

382 in the subdistricts (99% certain and p-value is less than 0.001). Although the
383 investigation of drug hotspots utilising GIS methods has not yet concluded,
384 the preliminary data already indicates that this is a very promising solution.
385 The present research has implemented multi-dimensional GIS mapping on the
386 hotspots for opioid use in an in order to elucidate a more complex geographic
387 picture of them.

388

389 Achu and Rose [18] explored GIS analysis tells, the current level of crime
390 , as well as its history and likely future trends. By using this study both the
391 author enlightened that the incidence of crime is rising in developed countries
392 owing to poverty and inequality. This research would look at the temporal and
393 spatial variation in crime in Thiruvanthapuram over the time span of 2010 to
394 2014 to learn more about crime occurrence. Computer aided Geographic infor-
395 mation tools like GIS, in addition to providing an increased level of detail for
396 law enforcement, have the potential to draw up maps that enable them to find
397 areas of criminal activity and papering, geographic correlation, is applied to the
398 resulting hotspot patterns to determine the origin of crime hotspots, and inter-
399 polation is used to further pinpoint the location of criminal events among the
400 others. Spatial auto-correlation has been conducted before in order to discover
401 Getis-Ord G_i^* space anomalies and outlier phenomena previous to Getis-Ord
402 G_i^* detecting trend analysis of spatial auto-correlation. Crime analysis utilises
403 four key components, which include venue, timing, rates, factors, and types, in
404 identifying areas where statistically important crimes are more likely to occur,
405 then makes use of IDW interpolation to provide a stronger visualisation. One
406 of the uses of this methods on the data from Thiruvanthapuram in Kerala is for
407 the areas of high crime prevalence including manslaughter, robbery, stealing,
408 and petty theft cases in that jurisdiction.

409

410 Rastogi et al. [19] signified the applications of GIS in crime visualization
411 and crime detection and investigation are indispensable response capabilities
412 are more commonly implemented using a combination of data from many or-

413 ganizations and outlets. Via an internet search of cases in police records, both
414 through past observation, investigators may identify the suspects that have been
415 recently been detained in certain areas and determine the crime's hot spots. The
416 group explored that due to both a spike in violence and the social problems,
417 there is a need for new ways to be developed to deal with this many measures
418 have been put in place to lower the amount of crimes against women in towns,
419 but many of these cities are also considered dangerous. The proposed study
420 seeks to consider the socio-economic issues contributing to an increase in the
421 risk of violence against women. Inverse-Weighted (IDW), Kriging, and Spline,
422 and even, in order to explore the relationship between socio-economic charac-
423 teristics and socio-emotional parameters on these maps more comprehensively.
424 This is the best interpolation method: by using the best selected interpolation
425 methodology, the thematic maps are generated to forecast the trend of criminal
426 activity. Once the proposed architecture has been built using GIS and big data
427 technologies, it can be implemented as a real-time decision-making method for
428 helping those in need. True evidence is used to test the study's methods in
429 the Jhunjhunu district of Rajasthan, India, and a case study is conducted to
430 analyze it. It is captured by the group of authors that there is an increasing
431 need for an atmosphere of safety in Indian cities.

432

433 Zahrane et al. [20] signified in studying the use of three distinct types of
434 hotspot analysis to define and classify RTA hot spots and measure their impact
435 on road traffic congestion, in order to arrive at reliable metrics for the resolution
436 of congestion issues. We have three available approaches at our disposal: the
437 Density (KDE) approach, which was recently endorsed by the Getis-Ord G_i^* ,
438 and the recently introduced cost-based Getis-Ord G_i^* , system (STA) approach,
439 and the Getis- Ord G_i^* , procedure for calculating cost-related risk, as proposed
440 in an expanding, both of integrating social risks in the model. Until analyzing
441 the RTA data using any of these different approaches, they were first prepared
442 in digital form in ArcGIS, and then imported into ESRI ArcSDE programme.
443 Then, the actual safety measures (those listed above) are evaluated in relation

444 to the places, the number, region, and value of the locations where the RTA
445 is installed to find out the priority and cost/benefit of each measure has with
446 each resulting metric. This article identified that in order to examine and de-
447 termine the reason for these very long-term market fluctuations, a micro-level
448 study might be needed. Thus, the methods presented above, especially in regard
449 to stopping quickly with RTA expansion, have great value, but should not be
450 pursued if an RTA expansion is detected. Immediately once they have identified
451 a probable hotspot, the findings should be checked and subsequently, they can
452 map out all the possible dangers that will lead to further collisions, so they can
453 discover what will have the greatest impact.

454

455 Butt et al. [21] explored that many traffic-related incidents, the issue of
456 present times, are the result of human error, rather than bad weather or haz-
457 arduous conditions. The author Butt and group manifested in knowing where
458 the worst locations for car accidents and taking preventative measures to ensure
459 they do not occur is vital for traffic protection in the following year. The RTA
460 wanted to create a list of places with increased traffic risk and then monitor
461 it over time in order to identify areas that were more prone to accidents for
462 five years in Rawindi, in this case the GIS was utilized to research the extent
463 of the potential areas with greater danger of road mishaps. The method was
464 utilized for the aforementioned reason was the one established by Morans I and
465 Getis-Ord G_i^* method on the data gathered from the Punjab Emergency De-
466 partment. Most places marked for expansion and significant to the analysis
467 in terms of spatial clusters were concentrated in the northern and northeastern
468 portions of the city neighborhoods, which often included both industrial and res-
469 idential areas, although most hotspots representing an accident were found in
470 locations near colleges, airports and highways. This paper pronounced that an
471 Early warning system (or early detection system) using geographic information
472 systems can help to reduce the amount of resources needed to respond to a crisis.

473

474 *2.2. Geo-spatial Operations For Crime Visualization*

475 Dede et al. [22] conducted a research project with the researchers and com-
476 puter programmers in Bandung hopes of gaining different knowledge on crime
477 location and then exploring the possibilities of identifying danger distribution
478 by means of GIS. A number of statistical methods are used to quantitatively de-
479 termine the factor values and/ from that, under which of the above quantitative
480 tests, a set of deciles, the law is drawn to figure out the value distribution By
481 applying GIS program, a formula needs to calculate crime rate (or score) and
482 meaning (or weight) to various factors, such as the dispersion and population
483 density, Eastman would be able to obtain overlay results for Bandung. The au-
484 thor and group analyzed that any of the factors contributing to Bandung's crime
485 rates include history, public services, the availability of property, the amount
486 of roads, as well as the density of the community, along with issues concerning
487 the inequality of the poverty distribution using of GIS means that the location
488 of most criminal activity. This puts a crime-risk at the agglomeration in the
489 centre and western Bandung, because the region has been prioritized as a police
490 concern.

491

492 Lee et al. [23] the aim of the research was to understand the existing crimi-
493 nal population's habits and behavior patterns in order to allow new solutions to
494 be proposed; it commenced by studying and understanding the criminological
495 patterns that had already been established in other cities, later on conducted
496 some research in this area to determine where the root causes of crime can
497 be found; it developed methods to classify the current CCTV-based causes of
498 crime for solution; and at the same time, provided diagnostic recommendations.
499 Looking at the crime features, aquatic and stationary populations, as well as
500 crime locations, it was discovered that the floating population had the greatest
501 density of all in and of the places where it occurs. This paper explored that
502 the need for Korean people to feel secure and protected has increased, resulting
503 in the nation pursuing long-long-term strategies for constructing safe cities as
504 a matter of priority. In order to deter urban crime, it is essential to learn as

505 much as possible about the everyday life of urban citizens, as well as about
506 the locations and features of the city where they live. According to this pa-
507 per this result supports the idea that floating population data should be used
508 in lieu of actual population data, as the difficulties in collecting such data has
509 previously been cited as a problem. One big arterial roadway expansion after
510 another and City A's new residential developments focused on them, resulting
511 in the rapid growth of a floating community along these arteries. these places,
512 as a hotspot for violence, are the crime areas in these cities being associated
513 with people who stay by the water or in flotation's, as well as the frequent
514 place "anchor" points for criminals, which could be reason for installing CCTVs
515 and other crime-prevention tactics including increased police patrols Instead,
516 however, places prone to violence seem to concentrate further such where it is
517 most visible as a direct results of individual events, with most of the changes in
518 the people's everyday lives (i.e. over the course of time while they're working,
519 socializing, or shopping).

520

521 Faisal and Shaker [24] turn inside out that socially speaking, urban environ-
522 mental quality can be described as a description of how urban, environmentally,
523 the climate, and in terms of socio-economic factors are doing. A few partial at-
524 tempts have been tried with a multi-temporal and multi-resolution data in the
525 past, as these datasets give a clear picture of both visual and understand land
526 cover (to) water coverage and vegetation patterns in urban areas which helps
527 identify further patterns and patterns of usage and develop more extensive de-
528 tail in the Urban EQ assessment does so, as well, being thus more beneficial
529 to sustainable development and urban planning. They signified into that the
530 United Nations has predicted that the world population would double in the
531 planet within the next forty years, and have an effect on the atmosphere and
532 human well-being. The goal of this research was to use GIS and remote sensing
533 to analyze the UEQ with an in the case of Toronto, to gain a better under-
534 standing of the impact of various natural, urban, and socioeconomic conditions
535 on data-related trends. There was then a need to gather and monitor natural,

536 urban, and socioeconomic data, as well as geography information using remote
537 sensing, and GIS technology. To measure the effects, they collected comparison
538 data and examined the UEQ relationship to see how that it affects the quality
539 of the test's outcomes. More precisions and accuracies, at 71% and 65%, are
540 recorded by the GIS technique. This project would provide the municipality
541 with a generic metric to assist with preparing with regard to all educational,
542 environmental, and urban requirements where you might have a few other kinds
543 of results to monitor. Both the authors explored that via the use of two different
544 methods:

545

- 546 • GIS overlay (on a map) and,
- 547 • PCA, the results of this research aimed to evaluate the correlation of UEQ
548 with the study location in the city of Toronto, Canada.

549 Srivas and Khot [25] discussed a variety of tools and methods for multi-
550 dimensional database and visualizations that are available for complex data
551 sets. In this article they mainly covered the popular open source Quantum GIS
552 (QGIS) visualization tool which works well with several different visualization
553 functions, as well as complex data types and data models. They are mostly used
554 for obtaining several representations of records, unions and combination of num-
555 bers, but can also be used for data extraction and numerous other processes. A
556 bonus aspect that the consumer may take advantage of is the greater precision
557 of the software is Edit Spatial Information and exporting maps to and diagrams.
558 The software application of GIS is well suited for creating geospatial decision
559 support systems (DSS). The process of collecting a vast volume of spatial data
560 takes a lot of time and resources. Research has resulted in different techniques
561 for spatially enhanced research which are quicker and provide improved analysis
562 outcomes in lessening the amount of time spent on its completion. The Spatial
563 Data Management box offers a wide range of resources that are made available
564 for generating, managing, and managing, and fixing problems, as well as for
565 growth, design, management, and advancement, by using function datasets.

566

567 Jefferson [26] made use of a mixed-methods methodology, taking into ac-
568 count work with the Predictive Analytics, as well as analytical documentation
569 pertaining to city-planning documents from the police, to examine the role of
570 predictive technology in co-prediction on complex problem-solving with the so-
571 cialized policing. Indeed, there is a rising number of geospatial intelligence
572 specialists and law enforcement agencies as well as well as criminal justice de-
573 partments that see GIS and predictive analytics playing a role in improving
574 police targeting to "to placing a focus on the arrest of criminals at particular
575 places rather than on "groups of persons or geographies. The author signified
576 with the added consideration of temporal factors, these Chicago researchers are
577 investigating the dynamics of department partnerships, including geographical
578 intelligence scientists and police experts who are presently collecting time-based
579 data and spatial information in GIS maps to determine where potential criminal
580 activity will arise. The author focused on important police-distant policing as
581 well as well as on societal spatial dynamics to learn more about the extent to
582 which predictive crime mapping has on shared data strategies.

583

584 Jayakrishnan and Shook [27] demonstrated how spatial simulation methods
585 can be used to extend spatiotemporal analysis methods to the general case on
586 massive datasets. A separate parallel calculator was created to help our team
587 keep track of their work for ensuring its continuous improvement, which was then
588 compared to an existing one to track of the same purpose, real crime data anal-
589 ysis was used as well as established tools to help in understanding how scalable
590 it is, and an assessment of performance gains occurred in this latter function.
591 Both the author highlighted Crimes, by their nature, are spatial phenomena,
592 and in terms of its locations, are temporal phenomena repetitive, descriptive,
593 describing, and objectively measuring the locations and time intervals of crim-
594 inal activity are used to classify, characterize, and document spatial-temporal
595 trends of criminal activity. Similar speech about criminal behavior improves
596 comprehension of commutability as well as spatial and temporal relationships.

597 This article explored this as something that is well suited to temporal stud-
598 ies because it provides spatially referenced crime data, or spatial correlations,
599 enabling crime data to be mapped out in time and determining the actual rela-
600 tionship between them.

601

602 *2.3. Crime Hotspot Prediction*

603 Yang et al. [28] delve into a large percentage of people, such as the general
604 public, are affected by crime in a negative way. Previously, hotspot prediction
605 has been proposed as a means to expand law enforcement jurisdiction. There
606 are different ways of predicting which metropolitan regions are crime magnets,
607 some of which depend on data analysis of historical incidents and events, and
608 others which don't. It is only available through the current methods, though,
609 that have shown to have the ability to classify hotspots based on the past his-
610 tory of crime. None, however, have shown the predictive potential of the more
611 recent analyses of social and urban evidence. The author and group presented
612 this study in which we outline Crime Telescope, a visual platform combining
613 various kinds of data which employs both prediction and visualization to ex-
614 amine crime hotspots. If the initial set of data is obtained, it searches for core
615 properties in the data through mathematical and linguistic study. Lastly, it
616 gathers data from various places and allows users to zoom in on them on a
617 digital map to find specific locations with highlighted risk areas. The paper
618 prospected that urban councils ought to do not only to enhance the protection
619 of city neighborhoods but also to help the environment of the towns and cities
620 where they live in. They turned inside out that a picture used to find locations
621 with a significant rise in crime incidents, and then a dot or line indicates an
622 area's relationship to other dots in the chart and their general location on the
623 image, as criminologists use it in this context (i.e., crime hotspots). If police
624 and city governments want to find out where the violence occurs in cities, they
625 need to use a high-predictive "hot spots" chart that more closely matches actual
626 areas where crime already occurs with risk factors.

627

628 Chainey [29] signified that a hotspot mapping strategy is also employed for
629 assisting police with enforcement activities including directing patrols to cer-
630 tain areas where they are needed and lowering crime. Histogram approximation
631 is still has long been used, but in modern times there are several alternative
632 methods that have emerged, which seem to be more common is KDE. Known
633 because of the unique graphical representation of crime visualization and is the
634 most closely followed by government crime analysts, which uses the density of
635 land mapping is that KDE also incorporates is considered with other hotspot
636 mapping techniques. Entering values for two key parameters in the command
637 produces two kinds of output: cell size and bandwidth must be done to achieve
638 KDE results. The few studies that have been done have looked at these to
639 date on their influences have so far have only looked into their general hotspot-
640 mapping purposes. No study has yet looked at the function of these criteria on a
641 hotspot chart, although it can be assumed that these may be used to see where
642 crime occurs in the future. This studies helped to fill this void by utilizing infor-
643 mation on various cell sizes and statistics on crimes committed by people with
644 separate from the population on residential robbery and abuse. This study es-
645 tablished that a null influence relationship between cell size and crime hotspots
646 on spatial trends of crime, but wireless capacity does impact crime. In the
647 end, we will be addressing how the findings of this research can enable the po-
648 lice and researchers to improve the way we use KDE for criminal investigation.
649 The author explored that Not only has the introduction of heat mapping, but
650 even the use of hot-spot policing is credited with the birth of this new approach.

651

652 Araujo et al. [30] highlighted the use of an algorithmically an infrastruc-
653 ture architecture, which can then be implemented to service current and future
654 public protection networks and channels. The system is based on the ROTA
655 strategy, with public safety initiatives as its background, which was devised
656 for the greater purpose of improving urban services in Brazil's new African
657 metropolis, Natal. The results of an experiment gave valuable information re-

658 guarding the various estimators. Furthermore, the outcomes of the algorithm
659 selection test on the hotspot detection challenge shown by executing a simu-
660 lation. The author suggested a statistical method for detecting hotspots that
661 addresses geographical, temporal, and categorical configurations of crime utiliz-
662 ing machine learning algorithms that are more specifically calibrated for average
663 danger levels. This architecture is applied as a service using the algorithm as
664 an API. The structure is able to accommodate many data streams, has several
665 different ways of seeing, offers a wide range of specifications, which allows for
666 machine learning to be included, and includes the use of an algorithmic decision
667 making. The system produces an efficient artifact for the process of developing
668 and maintaining patrols, called a program card. Author and the group explored
669 that a significant amount of research has shown that the possibility of predictive
670 analyses leading to higher accuracy in identifying patterns and predictions of
671 crime patterns has been explored through using quantitative experiments. The
672 use of a novel portfolio of machine learning algorithms offers a mechanism that
673 more specifically identifies locations and periods when law enforcement officers
674 might be at greater risk of being targeted, allowing them to adjust their sched-
675 ules to certain instances with more exactitude.

676

677 Kang & Kang [31] signified to increase the accuracy of criminal activity
678 modeling; they investigated the design of protective factors, for example, the
679 principle of broken windows and the use of environmental protection measures.
680 To present this article, they have come up with a data fusion approach that
681 takes the text and contextual information into account, which utilizes a deep
682 neural network (DNN). They did a comparative analysis of the interactions to
683 see whether crime occurrences were correlated with the results of the study. This
684 process allowed for an excellent and successful forecast of crime patterns to be
685 made, as deviants harm the prediction. It is a multi-modal prediction model
686 and suggested here that they shall use to make more reliable estimates of crimi-
687 nal activities. The dataset of this article was compiled from numerous Chicago,
688 Illinois web datasets, such as violence, demographic, and meteorological and

689 photographic/archaeological information, as well as photos and photographs of
690 the city. First, Authors did some statistical analysis on crime-making patterns
691 before training the model with crime-related evidence. With spatial, tempo-
692 ral, environmental, and function dependencies, our DNN additionally consists
693 of four layers: spatial, temporal, environmental, meaning, and joint feature
694 representation. Coupled with the pertinent data from different contexts, our
695 analytics is an efficient decision-making process that employs data redundancy
696 analysis, our DNN integration is an effective way to glean critical insights.

697

698 Macbeth & Ariel [32] explored that the first phase of a proactive hot spots
699 policing approach was the effective detection of crimes that have been occurring
700 for a long time and can be expected to continue. The second stage is man-
701 aging any crime that is changing or appearing for the first time; such issues
702 cannot be defined as hot spots, since they are still volatile and inconsistent.
703 Based on research this document indicated where violence is likely to occur,
704 "recidive hot zones" is observed to focus more specifically. when using this lo-
705 cation feature, statistical forecasts will serve as a kind of professional judgment
706 as a "experts" about whether an idea has promise, through how well it would do
707 in various places. Although well-positioned predictive "hot spots" and "hurri-
708 cane damage" forecast possible events of coastal flooding, just 50% of the paths
709 are marked correctly" Business and political judgment will erroneously locates
710 future crimes, and is more suitable for handling incidents of a criminal nature.
711 Recidivism is a significant, however only one is relevant since it is characterized
712 as a "a hot spot on the map": it is the precise region in microdot locations.

713

714 Lin et al. [33] enlightened that grid-based crime modeling and spatio-
715 temporal details provided by the Google's Places API was used to show a va-
716 riety of temporal and spatial characteristics in Taoyuan City, Taiwan. In the
717 first place, the authors have examined two traditional approaches: a spatial-
718 temporal and an empirical model. The use of spatial and temporal models is
719 widely accepted as being effective for crime reduction, including Kernel Density

720 (KD) and Time Series. They expanded on sentence: These two approaches only
721 deal with time or space only, but crime is significantly influenced by a lot of
722 other variables. Due to the centralized data processing and grid-based reason-
723 ing, machine learning is particularly useful for grid-based crime prediction. The
724 most accurate model was learned to be a new Deep Neural Networks classifier,
725 which was found to outperform Support Vector Machines, Random Decision
726 Forests, and K-Nearest Neighbor algorithms. By comparing our F1 scores to
727 a base scores, the design' 100-baseline 100-month moving average, they found
728 that our performance increased by around 7% (meaning its top score improved).
729 This article proved that their design outperforms the standard by being more
730 effective at locating crime displacement in the designs. In this case, they de-
731 scribed "generally" as the process of creating new algorithms, or "typically" of
732 improving feature design model efficiency to raise model accuracy, whereas ker-
733 nel density estimation (high pattern-level estimation) models which often look
734 for blocks and distributions (predictions resulting from) diminishing data.

735

736 The extensive literature survey performed on diverse techniques of spatial in-
737 terpolation and autocorrelation techniques present that no convergence has been
738 observed till now. This paper provides a proposed framework which presents a
739 novelty in the domain of GIS which shall have immense applicability in diverse
740 domains. It can be observed that Kriging interpolation and kernel density in-
741 terpolation are widely used interpolation techniques in GIS. Further, Getis-Ord
742 G_i^* also is a popularly used tool in the domain of crime analysis since it provides
743 hotspots and coldspots in a very efficient manner.

744 **3. Materials and Methods**

745 *3.1. Proposed Framework*

746 The proposed framework is a four-tiered architecture represented in Figure
747 1 which epitomizes the implementation of spatial interpolation techniques con-
748 verged with spatial autocorrelation features. The primary tier i.e., the geospatial

749 data pre-processing tier serves the most inevitable part of the implementation
750 framework. Since, the data available have multiple lacunae such as missing val-
751 ues, noise and many such. To deal with such data we need to pre-process the
752 dataset. Since, we are dealing with spatial interpolation as well as autocorrela-
753 tion, we require the creation of vector files i.e., in ".shp" format (also called as
754 shape files). Further, missing values can be handled by either properly filling
755 the values or by instantiating the values with a common name like N/A, Null,
756 etc. Further, in cases where we are required to use non-geospatial data i.e.,
757 data with no coordinate specifications, we need to perform geo-coding opera-
758 tion where automatically the dataset is filled with latitude and longitude values
759 which makes the dataset geospatially compatible. The second tier comprises of
760 the vector analysis tier where initially we need to join a common attribute of
761 the vector file and the considered dataset in order to make the vector file suit-
762 able for further analysis. Then, we need to generate choropleth map in order to
763 classify diverse zones based on the specific classifiable attribute. A graduated
764 color map is generated where we are required to fix the range of the classifiable
765 attribute based on which color allocation will take place. After the choropleth
766 with proper rane fixation is done, the next layer needs to be implemented.

767

768 The interpolation tier deals with diverse interpolation methods. We need
769 to specify radius at which interpolation shall be carried out. Further, shape
770 of interpolation also needs to be specified based on the application domain.
771 For, TIN interpolation there exists two interpolation attributes i.e, linear and
772 Clough-Toucher (cubic). Further, for IDW interpolation we require distance at-
773 tribute only. Also, for kernel density interpolation we need to specify the kernel
774 shape i.e., either quartic (bi-weight), triangular, tri-weight, uniform or Epanech-
775 nikov. After the interpolation is successfully implemented over the vector file,
776 we need to perform autocorrelation which is epitomized in the subsequent tier.

777

778 The fourth tier deals with spatial autocorrelation implementation. We need
779 to specify the LISA indicator which stands for local indicators of spatial asso-

780 ciation. There are two types of LISA indicators existing in GIS i.e., Moran's I
781 and Getis-Ord G_i^* . Moran's I is used in case of identifying correlation among
782 two data points whereas Getis-Ord G_i^* is used to find hotspots i.e., highly dense
783 regions as well as coldspots i.e, lowest dense regions. In order to perform the
784 task of autocorrelation we need to specify the spatial weight matrix which are
785 of types : queen, rook, bishop, distance and nearest neighbors. After, the au-
786 tocorrelation is performed we observe a graduated map providing p -values as
787 well as z -scores.

788 The output is a weighted-overlay map, heatmap, choropleth map classified
789 according to the considered attribute, an interpolated map, spatial autocorre-
790 lation map providing proper zonal analysis and corresponding p -values and
791 z -scores.

792 3.2. Dataset Specification

793 The data collected from [34] comprised of 15 columns and 2,829 rows i.e., a
794 total of 42,435 instances which comprised crime statistics from 1975 till 2015.
795 We have reduced the number of instances as per our implementation conve-
796 nience. The modified dataset comprised of 16 columns and 414 rows i.e., a total
797 of 6,624 instances that comprised of crime statistics from 2010 till 2015. Also,
798 the base data didn't have latitude and longitude values and thus, the modified
799 data has been geocoded to automatically allocate latitude and longitude values
800 to each row. Also, since it becomes difficult to join the dataset with the vector
801 shape file if joining attribute is not properly designed, this led towards adding
802 one more column to the modified dataset that entailed provinces to which the
803 regions belong to. Thus, the final dataset was made geospatially compatible in
804 order to perform geospatial crime analytics over it. The prime attributes i.e.,
805 selected during the implementation phase includes number of violent crimes,
806 latitude, longitude, crime per capita, and region. The term "crime per capita"
807 means the ratio of number of violent crimes to the total population in a region-
808 wise fashion.

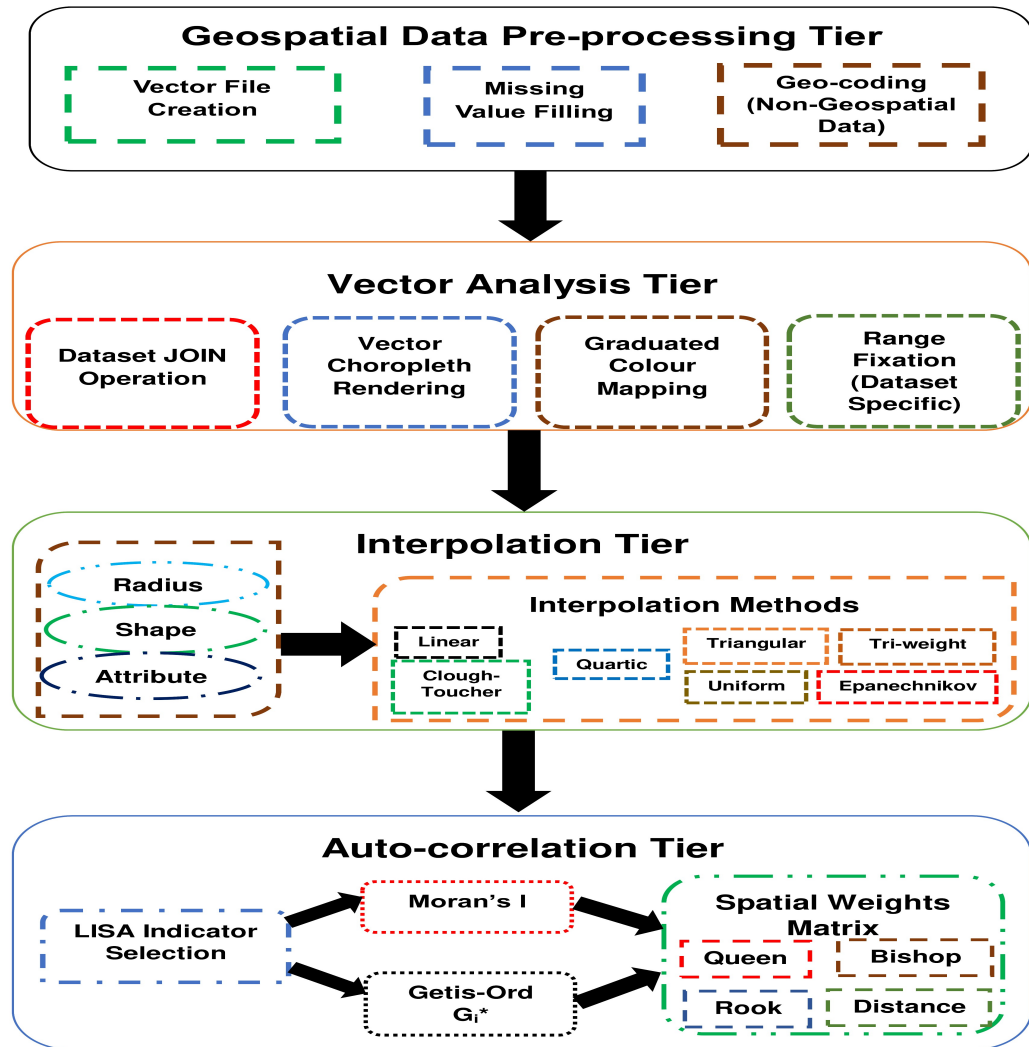


Figure 1: 4-tiered proposed framework for geospatial analytical implementation in context of spatial interpolation and autocorrelation.

809 *3.3. Experimental Setup*

810 The machine on which the experimentation has been performed includes
 811 hardware specifications as 1 TB hard disk drive (HDD), 4 GB random access
 812 memory (RAM), Intel Core i3 processor, and 2GB graphics card. The software

813 tools used for the desired implementation is QGIS 3.12 Bucuresti. For dataset
814 modification, the tool used is Microsoft Excel. Also, in order to perform the
815 geocoding operation on the dataset, Google Sheets were used with GeoCode
816 package so as to allocate geographical coordinates to each region automatically
817 by retrieving it from Google.

818 **4. Results and Discussion**

819 This chapter entails all the results obtained by practically realizing the pro-
820 posed framework. Figure 2 presents the initial task of geospatial crime ana-
821 lytics i.e., weighted-overlay analysis with labels signifying regions in differen-
822 t provinces of U.S.A. where crime has been reported as per the considered dataset.
823 Further, to visualize crime per capita in a region-wise fashion, a weighted-overlay
824 map has been labelled with the crime statistics in Figure 3. Further, to identify
825 crime hotspots based on crime per capita parameter, a symbological graduated
826 heatmap has been designed where based on color intensity the density of crime
827 can be visualized in diverse regions of U.S.A. The heatmap is presented in Fig-
828 ure 4. Further, a vector analysis, i.e., a choropleth map is designed after joining
829 the vector layer with the dataset and properly labelling it in equal intervals is
830 presented in Figure 5. Finally, the prime motto of geospatial crime analytics
831 is fulfilled by performing interpolation implementation with the assistance of
832 kernel density estimation. The radius is set to 5 units and kernel function set
833 to "quartic", an interpolated map is obtained which is represented in Figure 6.
834 Further, Getis-Ord G_i^* which is a local indicator of spatial association, used to
835 signify spatial autocorrelation is implemented. Based on the z -scores obtained,
836 the map is autocorrelated spatially. The regions can be identified as hotspot or
837 coldspot based on the color code allocated as per specific z -scores. The spatial
838 autocorrelation analysis is presented in Figure 7.

839

840 The results presented shows a detailed analysis of crime statistics in diverse
841 regions of U.S.A. based on which police patrolling, and other preventive mea-

842 sures can be taken in the regions which are identified as crime hotspots. Further,
 843 the results also can be visualized as a zonal analysis of crime parameters in the
 844 regions of U.S.A. based on which risk analysis can also be done.

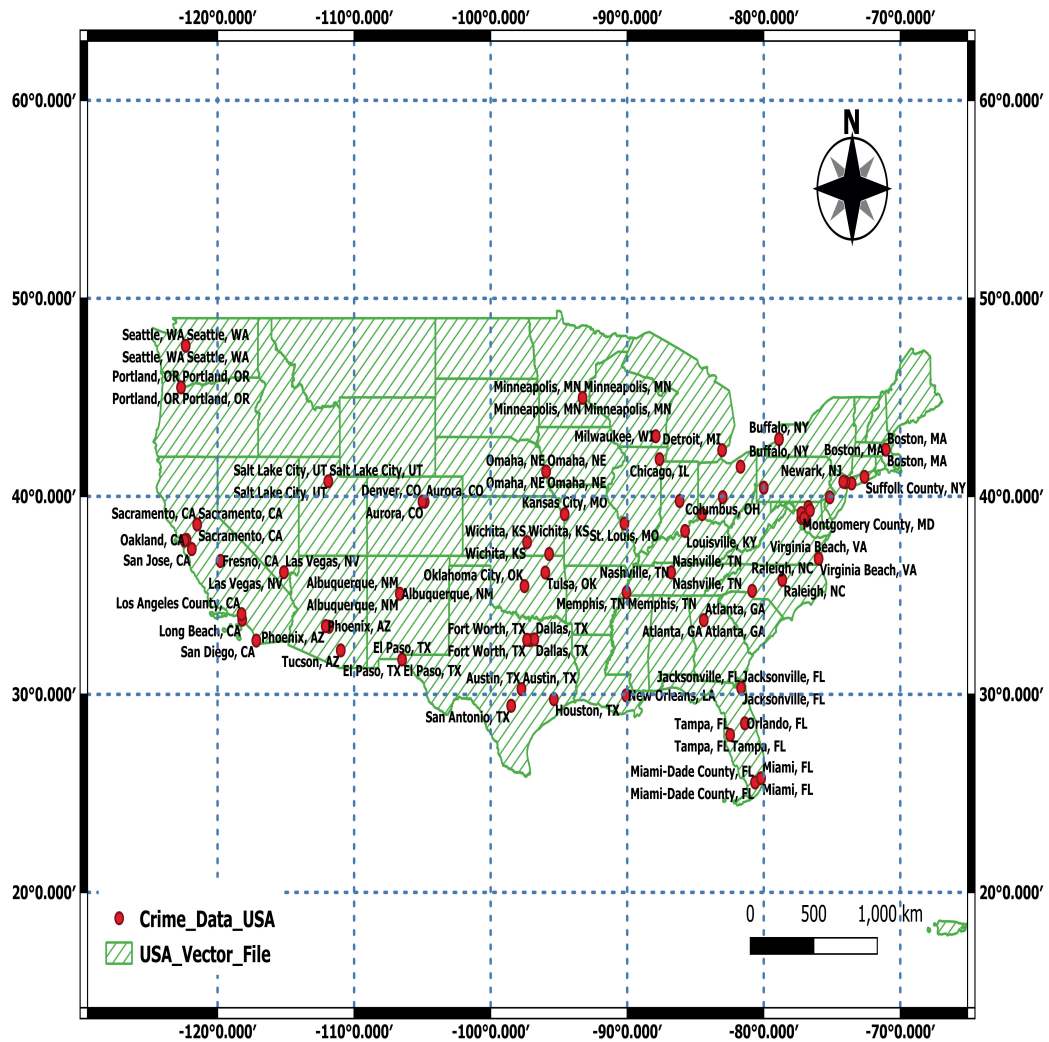


Figure 2: Weighted-Overlay analysis labelled with regions of crime in USA.

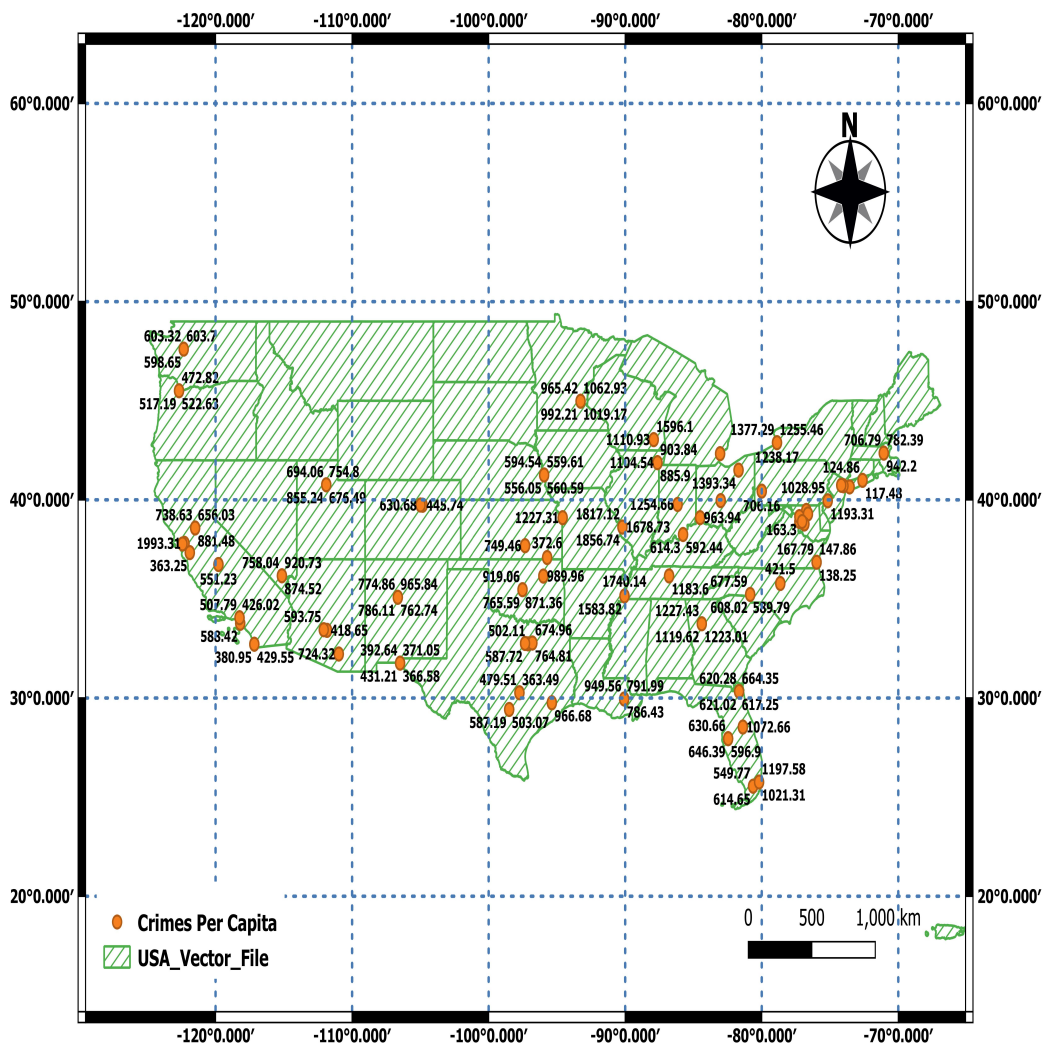


Figure 3: Weighted-Overlay analysis representing crime per capita in a region-wise fashion over USA vector file.

845 **5. Challenges**

846 There are few challenges which has been encountered during the phase of
 847 implementation which includes:

848

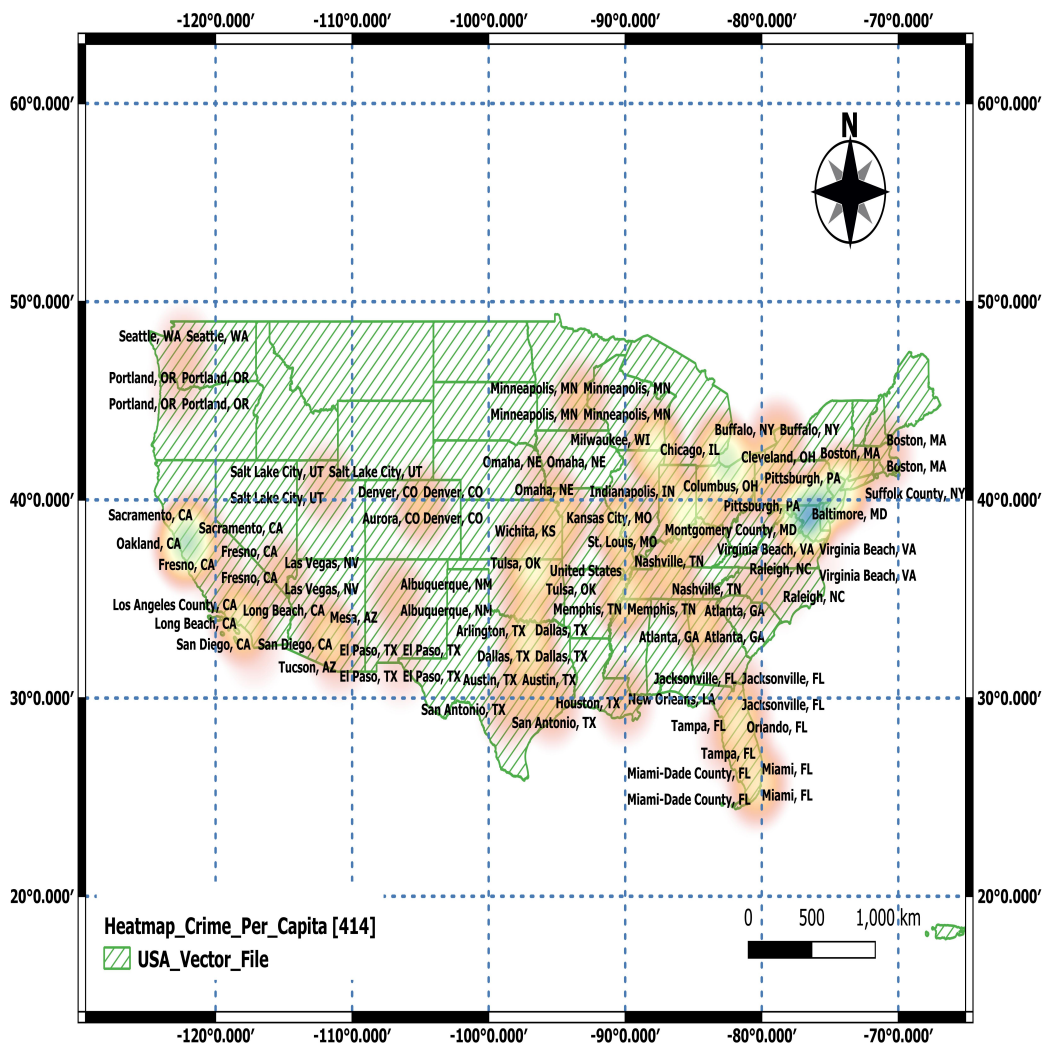


Figure 4: Symbological graduated heatmap analysis to identify crime hotspots in USA based on historical data.

- 849 • Dealing with geospatial big data is still a major challenge with local engines which makes the geospatial big data analytics task a hectic one.
- 850
- 851 • Switching to cloud GIS for real-time implementation requires much resource to be allocated and its equivalent cost constraints. This challenge
- 852

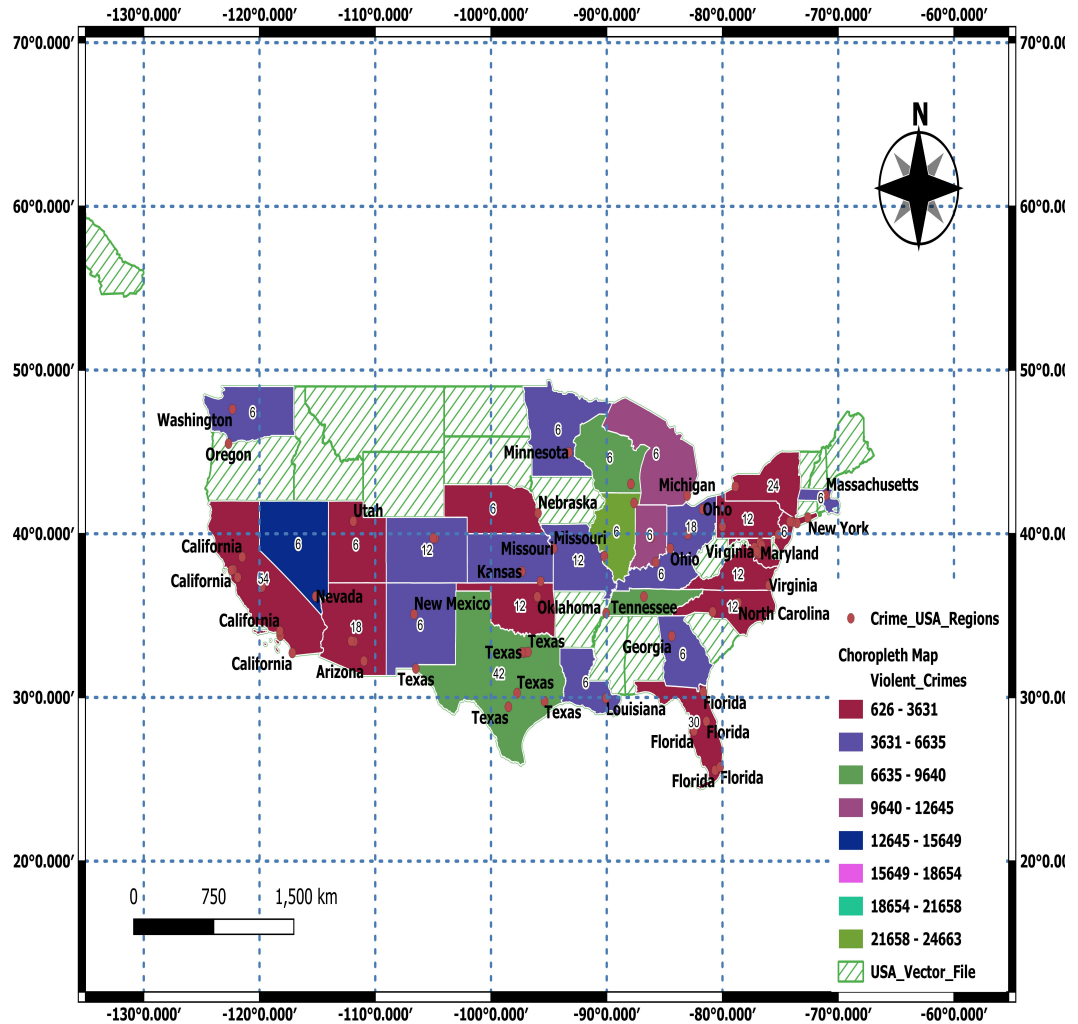


Figure 5: Choropleth analysis performed by graduating regions based on crime statistics.

853

can be overcome with serverless computing paradigm.

854

- Many machine learning tasks are used which are traditional like K-Means clustering, DBSCAN clustering, and many such which reduces the predictive competency of the geospatial model. It can be tackled with the

855

856

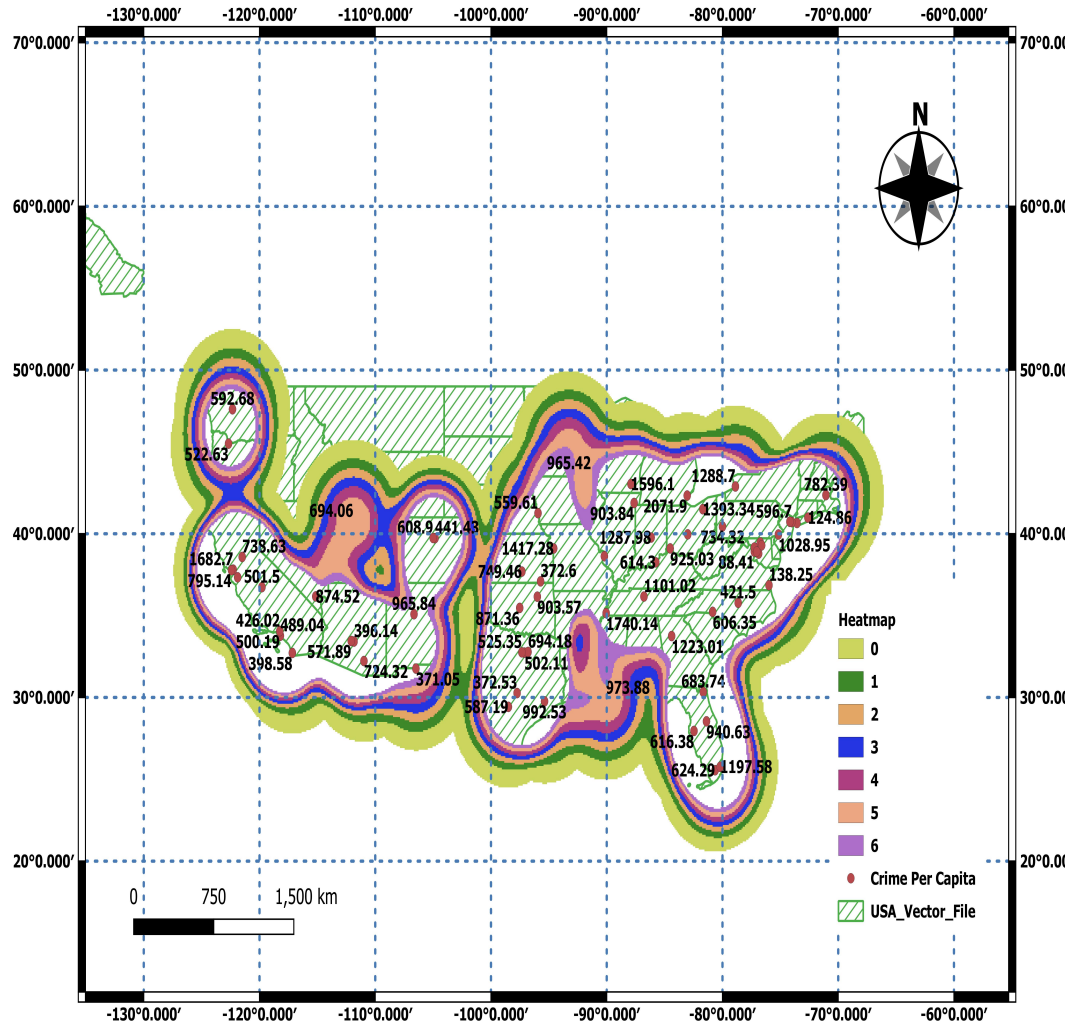


Figure 6: Kernel density estimation performed with radius 5 units and quartic functionality to visualize spatial interpolation.

857

advent of deep learning models.

858

- Crime hotspots could be predicted through this dissertation but to predict

859

where next crime is going to happen requires much more powerful model

860

since crime is an unprecedented event.

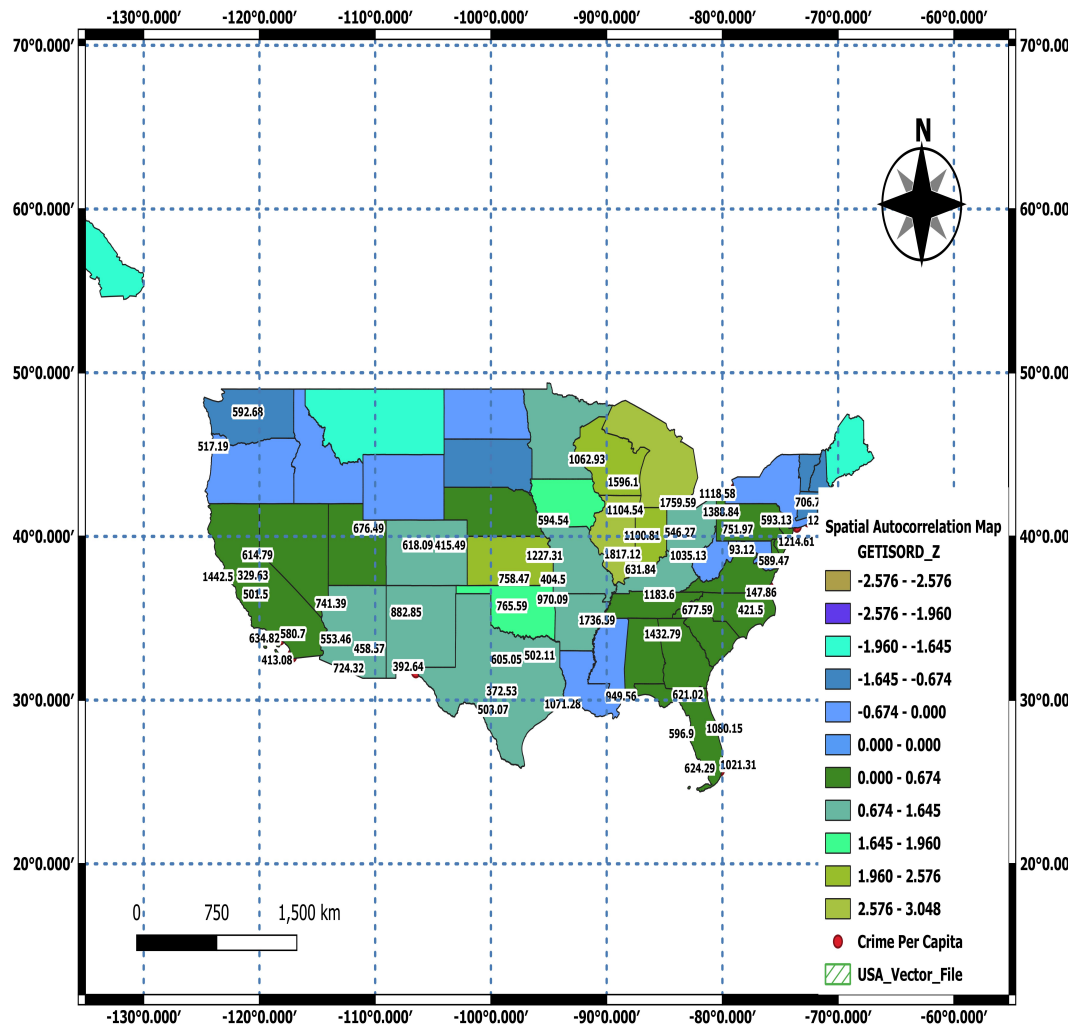


Figure 7: Getis-Ord G_i^* technique of spatial autocorrelation implemented to identify crime hotspots and coldspots.

861 **6. Conclusion and Future Scope**

862 The report consisted a thorough literature survey on spatial interpolation
 863 techniques as well as spatial autocorrelation methods used in the field of geospa-
 864 tial crime analytics. The report also entailed a proposed framework which epit-

865 omized the implementation of the merger of spatial interpolation and auto-
866 correlation for a better geospatial crime analysis paradigm. The interpolation
867 techniques entailed in this report includes Kringing interpolation, TIN interpo-
868 lation, IDW interpolation and kernel desity interpolation. Further, the two spa-
869 tial autocorrelation techniques i.e., also known as LISA indicators were enclosed
870 within this report that comprised of Moran's I and Getis-Ord G_i^* techniques.
871 In order to practically realize the efficacy of the proposed framework to perform
872 geospatial crime analytics, a dataset comprising of diverse crime types enclosed
873 within the attribute "violent crimes" for the regions in U.S.A. has been consid-
874 ered. Further, the dataset has been geocoded to automatically allocate latitude
875 and longitude values to each region. The dataset after preparation has been
876 overlaid on the vector shape file of U.S.A. so as to perform the weighted-overlay
877 analysis. The weighted-overlay analysis maps have been labelled with different
878 key attributes. Then, heatmap analysis has been performed and crime hotspots
879 has been identified based on crime statistics. Further, the choropleth analysis
880 has been performed in order to epitomize the join operation between the vec-
881 tor shape file and the corresponding dataset. Also, the interpolation technique
882 chosen from the proposed framework was kernel density interpolation. Kernel
883 density interpolation by specifying kernel function to be "quartic" has been im-
884 plemented to visualize the interpolated map. Finally, spatial autocorrelation
885 was implemented using the immensely used technique in the domain of crime
886 analytics i.e., Getis-Ord G_i^* . The considered interpolation technique is well-
887 known since it can classify both hotspots as well as coldspots (crime hotspots
888 and coldspots in our case). Thus, a detailed geospatial crime analytics has been
889 performed in this dissertation to entail all the available GIS-based features which
890 can be used to efficiently perform geospatial crime analytics. The emerging rage
891 of GIS technology has led towards immense applicability areas where GIS can
892 play a key role. Sectors like crime mapping, disaster management, healthcare
893 and many such have immense applicability of proficient techniques. Hotspot
894 identification is a prime task in GIS and it can be done with the assistance of
895 spatial interpolation by generating heatmaps based on density as well as spatial

896 autocorrelation techniques. The ever-growing demand of GIS technology can
897 have future spaces where the task of interpolation and autocorrelation might be
898 merged with techniques of artificial intelligence, machine learning, cloud com-
899 puting paradigm, serverless computing framework and many such.

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