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## RESEARCH ARTICLE

# Spatio-Temporal Change Detection Analysis of Land-use and Land cover: A study of Mysore-Mandya corridor region

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### Abstract

The present study has been conducted to analysis the land use and land cover of Mysore-Mandya corridor region between the years 1991 and 2014. The study has been done using remote sensing data and GIS techniques. The necessary satellite images (1991, 2001 and 2014) have been obtained from the United States of Geological Survey website and the GIS analysis has been done in ERDAS. The downloaded satellite has been corrected atmospherically and radio-metrically using ATCOR module in ERDAS. After preparing the corrected base imagery, supervised image classification has been carried out using maximum likelihood algorithm. To prove the accuracy of the classification result, the error matrix and Kappa index have been applied from which the classification accuracy was 95.91 per cent. The spatial extension LULC analysis showed that, the built-up land use has been increasing continuously while other land uses have been decreasing. The land transit analysis depicts the high conversion of vegetation and waste land into built-up during the study period. Finally, decadal wise comparison shows that, during the second period of study (2001 – 2014) the intensity of conversion is higher than the first period (1991 – 2001) for the land use categories.

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## INTRODUCTION

The land covers on the earth surface have been altering from the origin of the planet while land use has been changing ever since human started to live on a particular place. The study of land use and cover can help to manage the present as well as to take necessary actions for the future. (Zhou, Wu, & Peng, 2012). The expansion of population and their needs are altering the limited land and soil resources for agriculture, urban and industrial uses. Thus, the information regarding the rate and kind of shifting land resources is mandatory for proper management, planning and regularization of the use of such resources (Gautam & Narayanan, 1983; Gong et al., 2013).

The studies of land use and land cover using satellite imagery is very popular among the scientific communities as it is cost effective and technologically sound (de Sherbinin et al., 2002; L. Yang, Xian, Klaver, & Deal, 2003; Xiaojun Yang & Liu, 2005). Extensive research have been directed towards urban change detection by employing the remotely sensed imagery (for example; (Gomasasca, Brivio, Pagnoni, & Galli, 1993; Green, Kempka, & Lackey, 1994; Lo & Shipman, 1990; Royer, Charbonneau, & Bonn, 1988; Todd, 1977; Toll, 1985; Toll, Royal, & Davis, 1981; Xlaojun Yang, 2002; Xiaojun Yang & Liu, 2005; Yeh & Li, 1997). Remotely sensed data have been used to characterize patterns of land cover changes at scales from sporadic meters to few degrees in latitude by

longitude depending on the sensors (Council, 2008). Satellite imagery offers a historical and recent perspective on landscape dynamics (Peterson, Egbert, Price, & Martinko, 2004).

### **LAND USE LAND COVER CHANGE DETECTION**

The process of change detection identifies the modifications in the two sets of images taken from different time period. The process considers several factors namely the nature of the problem comprised in change detection, image processing method, selection of suitable variables and algorithms (Lu, Li, & Moran, 2014).

Several change detection techniques are developed and used by researchers. Notably the post-classification image comparison (Fichera, Modica, & Pollino, 2012; Rawat & Kumar, 2015; Tian et al., 2005), object based change detection (Blaschke, 2005), cross-correlation (Wang, Chen, Gong, Shimazaki, & Tamura, 2009; Z. Yang & Mueller, 2007), improved change-vector analysis (Chen, Gong, He, Pu, & Shi, 2003), texture analysis (Erener & Düzgün, 2009), multivariate alteration detection (Nori, Elsidding, & Niemeier, 2008) and change-vector analysis (Setiawan & Yoshino, 2012) are some of the change detection approaches.

The post classification comparison technique is the most logical method for change detection, which involves the cataloguing of each selected images independently followed by comparison of corresponding pixel value to identify areas where change has occurred (Roostaei & Kamran, 2012; Weismiller, Kristof, Scholz, Anuta, & Momin, 1977). The accuracy assessment of the classification to the real world can be analyzed using the error matrix system, through which the percentage of the categorization precision can be obtained.

Therefore, the present study has been conducted to analysis the land use and land cover changes along the Mysore-Mandya corridor region, where rapid changes have been taken recently by the construction of Bangalore-Mysore Industrial Corridor (Goldman, 2011).

### **METHODOLOGY**

The selected change detection method for the present study employs the independent thematic classification images taken at two different time spans, which can generate the results with 'from-to' change information that is easy to interpret. The correction of the atmosphere and sensor errors in the diverse satellite imageries would improve the accuracy of results. The methodology flowchart is displayed in figure 1.

To unearth the existing LULC in the study area, the satellite images of 1991, 2001 and 2014 have been corrected atmospherically, radio-metrically and geometrically. The atmospheric and radiometric corrections have been done employing the ATCOR module in ERDAS and geometric correction has been done through the ERDAS itself. The Figure 2 shows the details of the Landsat based map used in the present study which are downloaded from the USGS and they are likely to be as follows:

- For the first time period, 14<sup>th</sup> January 1991 Landsat-5 satellite data has been used.
- For the second time period, 3<sup>rd</sup> March 2001 Landsat-7 satellite data has been utilized.
- For the final time period, 11<sup>th</sup> February 2014 data Landsat-8 satellite has been employed.

To obtain higher accuracy of LULC, the base data should be on the similar scale or same spatial resolution which is of 30 meters for the present study. After the preparation of the base data, the supervised classification using Maximum Likelihood classification is implemented to classify the imageries. The supervised classification is a popular approach which is highly handled in India as well as in other countries (Maktav & Erbek, 2005), for Istanbul, Turkey; (Nori et al., 2008) for El Rawashda forest, Sudan; (Uma & Mahalingam, 2011) for Kanchipuram District, Coastal Stretch, Tamil Nadu, India). The training sample locations were identified for the year 2014 as 120 sample sites (Area of Interest) for Built-Up, 132 for Vegetation, 102 for Waste land and 57 for Water bodies have been collected and from this signature, a file has been created in ERDAS for the purpose of further analysis.

### **ACCURACY ASSESSMENT**

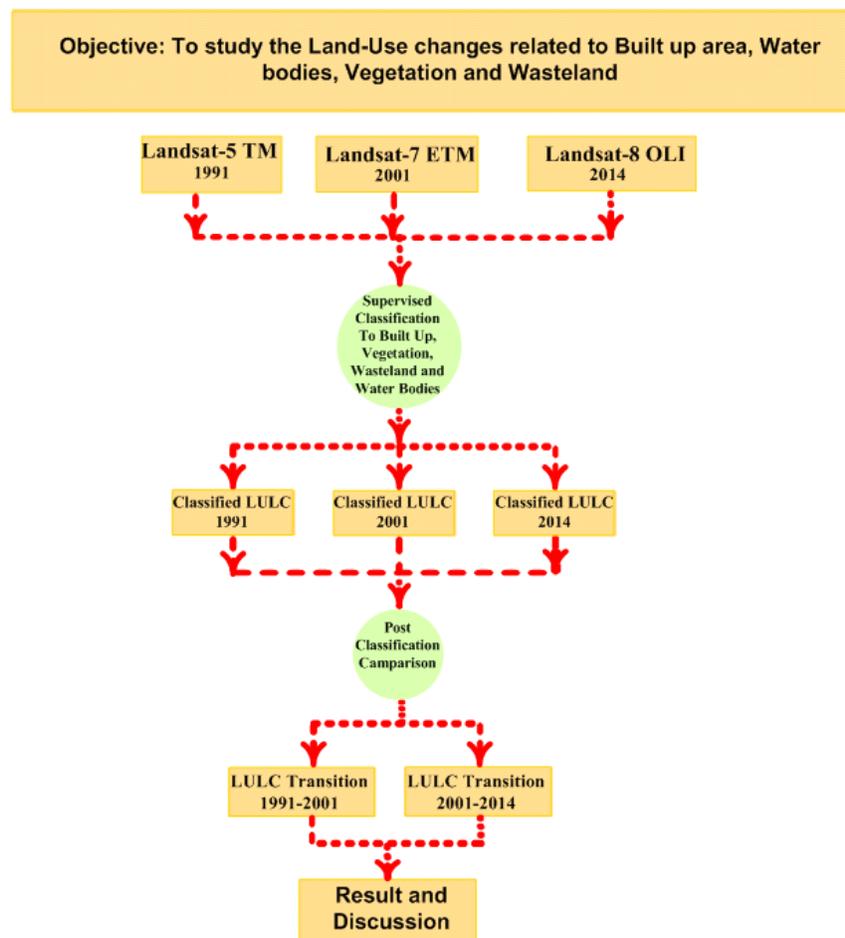
The result of classification cannot be accepted without proving the accuracy of the classification, therefore the present study has used the Confusion Matrix and Cohen's Kappa, which is highly accepted and followed by several researchers for the accuracy assessment of LULC studies (Hollister, Gonzalez, Paul, August, & Copeland, 2004; Kaul & Sopan, 2012).

The Cohen's Kappa is used to calculate the agreement between the two individual variables that measures the same land-use. The result of Kappa is less than or equal to 1 with the values near to 1 represents the perfect agreement while those near to 0 represents the worst agreement. The Table 1 accentuates on the error matrix table that is prepared to measure the accuracy of the classification.

Before selecting the samples, it is necessary to decide upon the number of samples for each class. Therefore, the study has followed rule of the thumb method to assign the number of samples. The total number of samples for each class has been determined based on the area covered, that is, for larger areas, a higher number of samples and lesser for small areas (for example: samples for vegetation are 158 with an area of 1543.81 km<sup>2</sup> and 31 samples for water bodies with an area of 72.53 km<sup>2</sup>).

As stated in the Table 1, an overall 269 points have been taken from the classified images randomly, from which, 32 samples belong to the waste land, 48 belong to the built-up section, 31 are attributed to the water bodies and 158 are related to the vegetation. For these selected samples, points of latitude and longitude have been identified and transferred to hand-held GPS for field verification. For ground verification and reference recording, the researcher has visited all the chosen sample areas. Finally, classified image results and the ground verification information have been used to prepare the error matrix and Kappa statistics (Figure 3).

The overall accuracy has been analyzed after the class wise accuracy assessment; the result represents 95.91 per cent of the overall accuracy, which is more than the minimum numeral of interpretation accuracy of 85 per cent (ANDERSON, HARDY, ROACH, & WITMER, 1976; Rao, GAUTAM, R.L.Karale, & Sahai, 1991). The 'kappa' calculation marks the value of 0.932 which is equal to 93 per cent.



**Figure 1:** Methodology flow chart

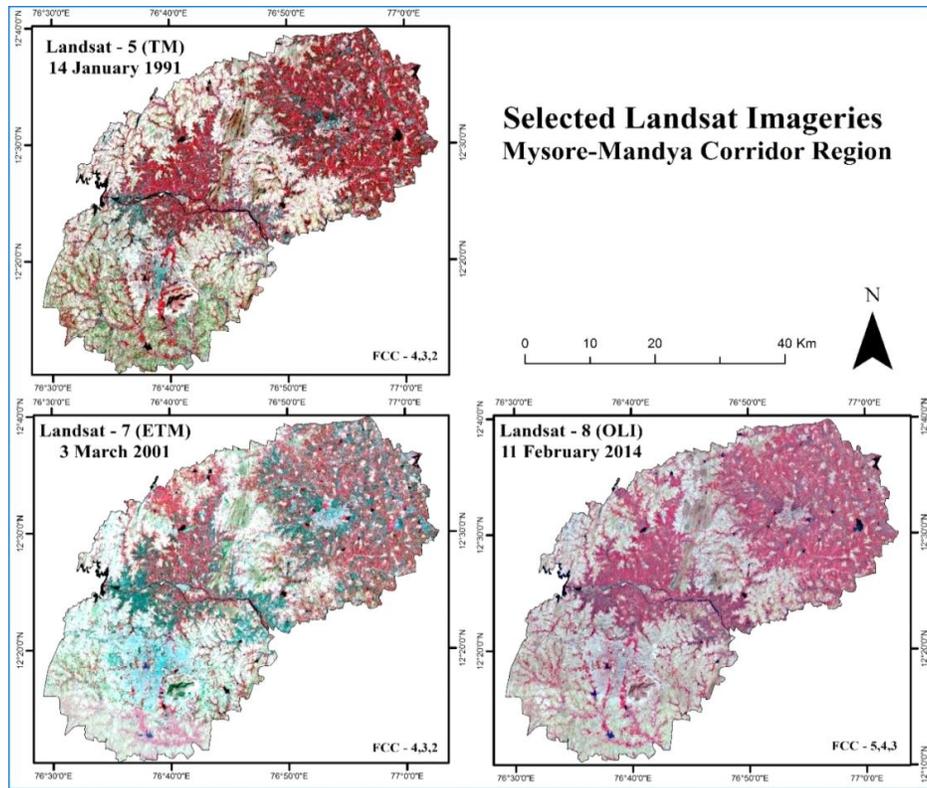


Figure: 4.1

Figure 2: Selected Landsat Imageries

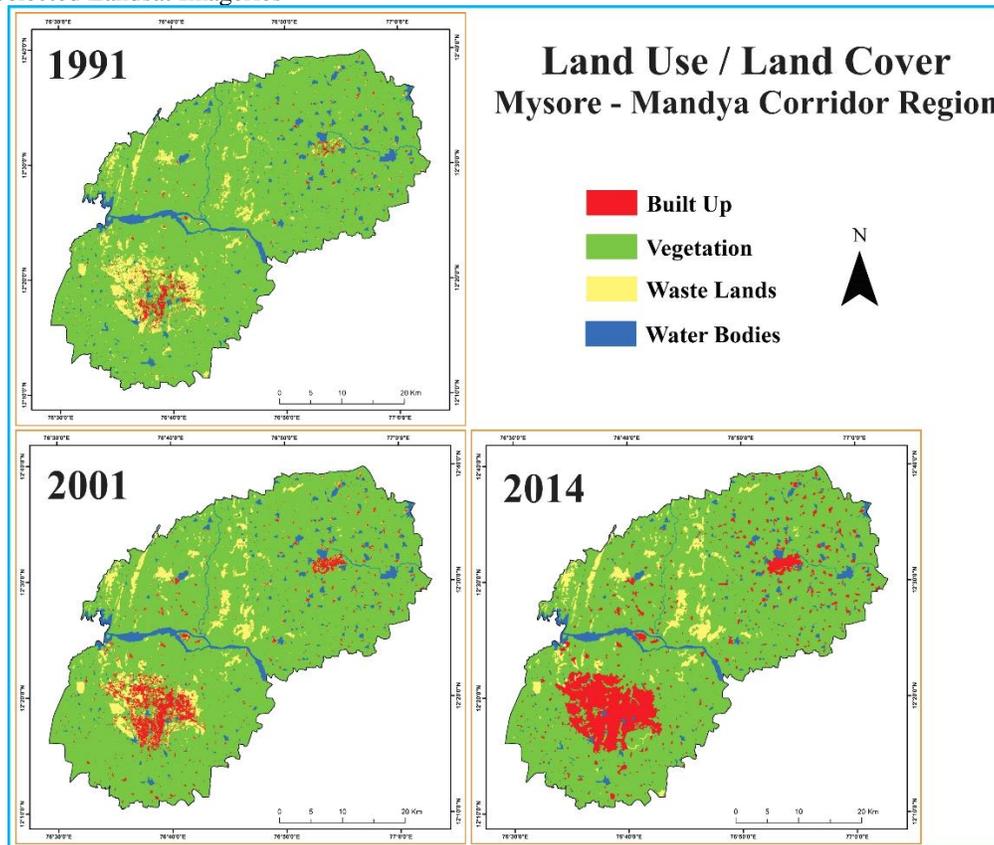


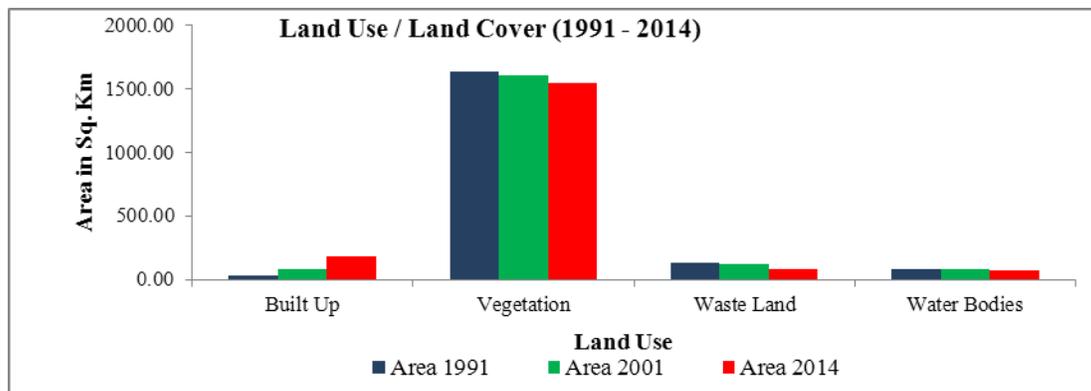
Figure 3: Land Use / Land Cover

S. No	Classified LULC	Wasteland	Built Up	Water Bodies	Vegetation	Total	User Accuracy
1	Wasteland	29	2	0	1	32	90.63
2	Built Up	2	45	0	1	48	93.75
3	Water Bodies	1	0	30	0	31	96.77
4	Vegetation	3	0	1	154	158	97.47
<b>Total</b>		35	47	31	156	269	
<b>Producer Accuracy</b>		82.86	95.74	96.77	98.72		
<b>Overall Accuracy = 95.91</b>							
<b>Kappa Result = 0.932</b>							

## DISCUSSION OF LULC RESULTS

After the confirmation of LULC classification result, the area covered by each class in the selected years has been scrutinized using ArcGIS and presented in Table 2 and Figure 4. The year wise classified result shows that during 1991, 2001 and 2014 the built up area has been increasing as 34.01 km<sup>2</sup>, 83.26 km<sup>2</sup> and 180.25 km<sup>2</sup> respectively, while other land use classes' vegetation, wasteland and water bodies have shown decrease for the respective three decades.

S.No	Classes	Area (km <sup>2</sup> )			Changes (km <sup>2</sup> )		
		1991	2001	2014	1991 - 2001	2001-2014	1991-2014
1	Built Up	34.01	83.26	180.25	49.25	96.99	146.24
2	Vegetation	1634.35	1597.54	1543.8	-36.81	-53.74	-90.55
3	Waste Land	127.03	121.23	83.47	-5.80	-37.76	-43.56
4	Water Bodies	84.68	78.04	72.55	-6.64	-5.49	-12.13



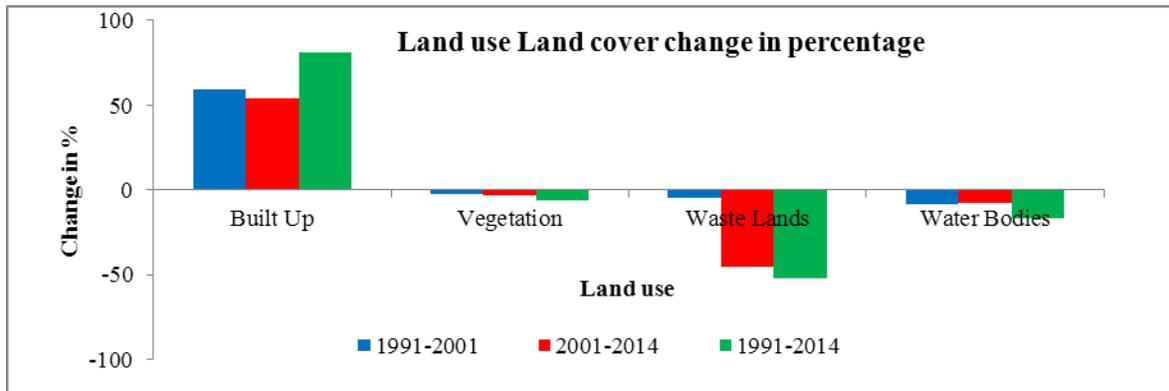
**Figure 4: Land Use Land Cover (1991-2014)**

Moreover, post classification comparison method has been used to calculate the LULC changes between the selected years and the change values are given in Table 2. The temporal changes of the classified LULC results prove that except built up areas all others having negative change. The increase in built up is systematically compensated by the decrease in the area of vegetation, waste land and water bodies. This indicates the positive growth in built up land use and decreasing rate of vegetation is compensated by the escalating trend in the built up area. Similarly, wasteland and water bodies shows decreasing trend following that of vegetation.

**SPATIAL EXTENSION:** Table 3, depicts the results of the spatial extension of each classified class in the study area. It is clear that even though vegetation has covered more than 80 per cent of the corridor region in all the three observed periods, its area has incessantly shrunk (Figure 5) followed by the waste land and water bodies. The built up has shown a positive augmentation in its area which was 1.81 per cent in 1991 and has gone up to 4.43 per cent to 9.59 per cent for the year 2001 and 2014.

S.No	Classes	Area (%)			Changes (%)		
		1991	2001	2014	1991 - 2001	2001-2014	1991-2014
1	Built Up	1.81	4.43	9.59	59.15	53.80	81.13
2	Vegetation	86.93	84.97	82.11	-2.30	-3.48	-5.86
3	Waste Land	6.76	6.45	4.44	-4.78	-45.23	-52.18
4	Water Bodies	4.50	4.15	3.86	-8.50	-7.56	-16.71

Source: Result of LULC Analysis



**Figure 5:** Land use land cover change

#### CHANGE IN PERCENTAGE OF LULC

The variation in the percentage of land use and land cover (Table 3) has been calculated to identify the changes that have been taken place in the study area for the period of 1991-2001, 2001-2014 and for 1991-2014. The built up has shown a positive change which was 59.15 per cent in 1991-2001 and 53.80 per cent for 2001-2014, followed by an overall rate of 1991-2014 where the growth recorded reached up to 81.13 per cent, whereas a marginal negative change is noticed in vegetation that is -2.3 per cent for the period of 1991-2001 noted a further decrease of -3.48 per cent in its area in 2001-2014. Ultimately, for the general period, the figure has reduced to -5.86 per cent.

#### LULC TRANSITION FROM 1991 TO 2001

The analysis of LULC transition helps to find out how the transformation of a particular land use or land cover has been shifted in to another category over the period of time. The table 4 shows the LULC transition that has taken place between the classified classes for the selected years (Figure 6) of the corridor region. The analysis result for the year 1991 to 2001 represents that totally 55.62 km<sup>2</sup> of vegetation area has been converted into other categories from which 25.87 km<sup>2</sup> turned into built-ups, highest area of 29.17 km<sup>2</sup> into waste land and the least of 0.58 km<sup>2</sup> into water bodies. The conversion of the area of the waste land shows that 40.84 km<sup>2</sup> areas have been transfigured into other categories, as highest area of 29.36 km<sup>2</sup> into built-up, 11.46 km<sup>2</sup> into vegetation and 0.02 km<sup>2</sup> into water bodies. Further the transition trend in the water bodies reveals that 7.01 km<sup>2</sup> of area were converted for other uses, as 0.56 km<sup>2</sup> into built-up, 5.25 km<sup>2</sup> into vegetation and 1.20 km<sup>2</sup> into waste land.

The general assessment of the LULC transition for 1991 to 2001 clearly states that larger quantity of vegetation and waste land have been converted into built-up lands, a hefty amount of vegetation land has been converted into waste land and notable amount of waste land has been converted into vegetation. After the analysis of each class areal transition to another class, the percentage of change occurred from the total area has been calculated and analyzed is as follows.

Table 4 indicates the transition between the vegetation and built-up areas. The tabled data highlights that 25.87 km<sup>2</sup> of the vegetation land in 1991 has been systematically converted into built-up within 2001. During the year 1991, the total area of vegetation was 1634.35 km<sup>2</sup>, that is,  $(25.87/1634.35)*100$ , which derives the answer of 1.58. This figure means that 1.58 per cent of the vegetation territory from the total area has been transformed to built-up from 1991 to 2001. Highest change of 1.78% of vegetative area has been transferred into waste land. The percentage wise calculation of each class transition clearly explains that 23.12 per cent of the waste land has been rehabilitated into built up. Following the trends, water bodies have been converted to a highest rate of 6.20 per cent

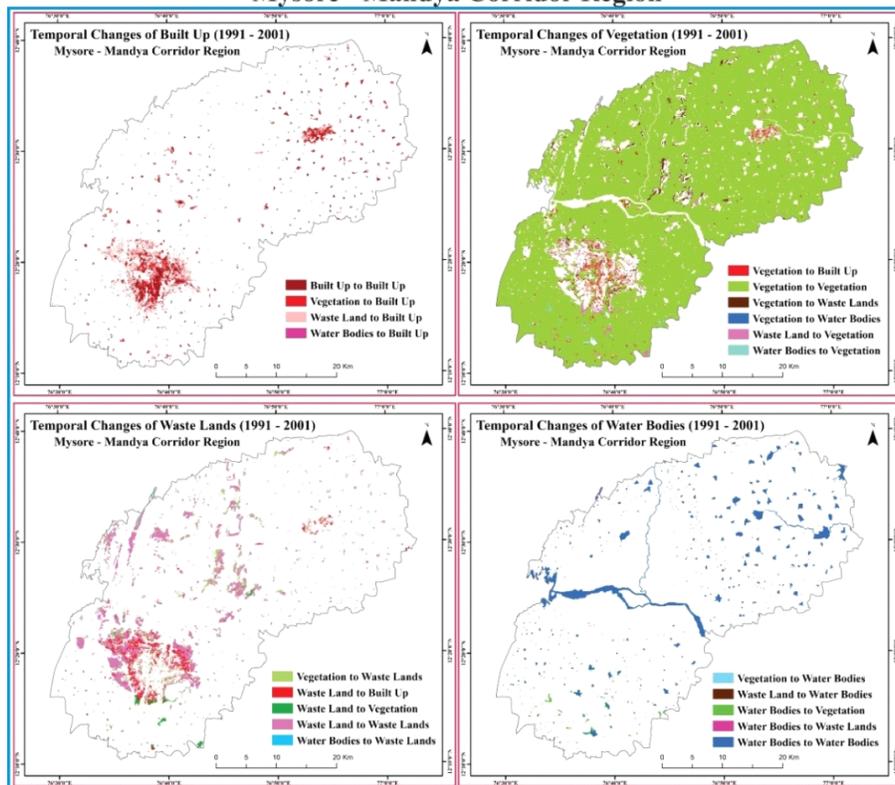
into vegetation. The overall assessment of the percentage of the recurring transition describes that waste land were highly converted trailed by water bodies and vegetation.

**Table 4: LULC Transition - Mysore-Mandya Corridor Region**

S. No.	Transition	1991-2001 (km <sup>2</sup> )	2001-2014 (Km <sup>2</sup> )	1991-2001 (%)	2001-2014 (%)
1	Built Up to Built-Up	34.01	83.26	Nil	Nil
2	Vegetation to Built-Up	25.87	42.90	1.58	2.69
3	Vegetation to Vegetation	1578.56	1533.32	96.59	95.98
4	Vegetation to Waste Lands	29.17	15.52	1.78	0.97
5	Vegetation to Water Bodies	0.58	3.34	0.04	0.21
6	Waste Lands to Built-Up	29.36	55.78	23.12	46.01
7	Waste Lands to Vegetation	11.46	0.12	9.02	0.10
8	Waste Lands to Waste Lands	86.17	64.97	67.83	53.59
9	Waste Lands to Water Bodies	0.02	0.21	0.02	0.18
10	Water Bodies to Built-Up	0.56	0.13	0.67	0.17
11	Water Bodies to Vegetation	5.25	8.57	6.20	10.98
12	Water Bodies to Waste Lands	1.20	0.00	1.42	0.00
13	Water Bodies to Water Bodies	77.42	68.97	91.42	88.37

Source: Analytical Result

**Land Use and Land Cover Transition (1991-2001)  
Mysore - Mandya Corridor Region**



**Figure 6: Land Use Land Cover Transition (1991 – 2001)**

**LULC TRANSITION FROM 2001 TO 2014**

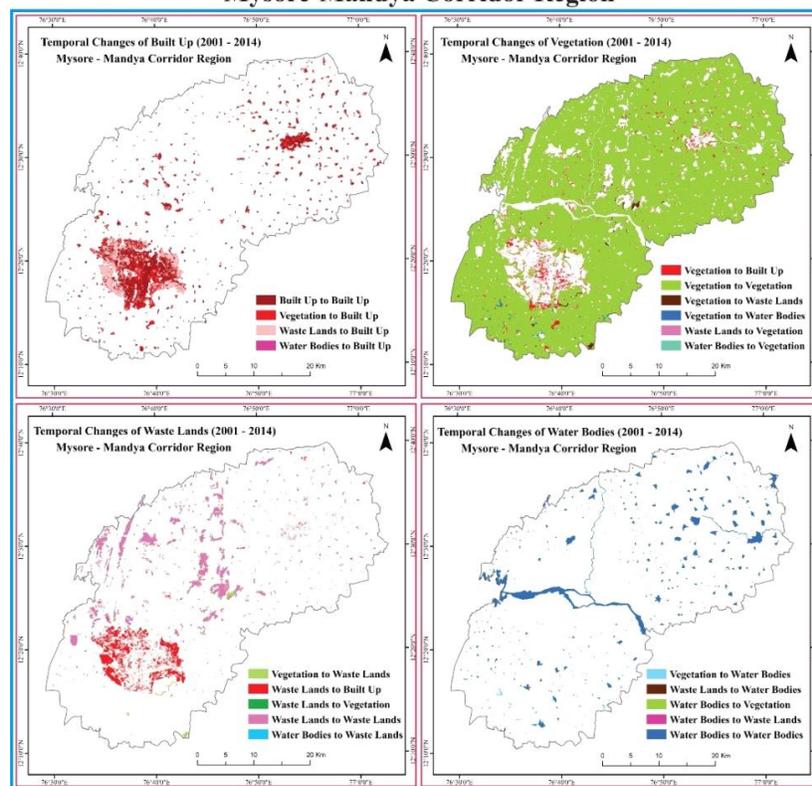
The transition calculation for the years 2001 to 2014 denotes (Table 4) that totally 61.76 km<sup>2</sup> areas of vegetation have been transformed into other (Figure 7) categories, which are higher than the previous decade. 42.90

km<sup>2</sup> areas have been converted into built-up terrains. The transition in the waste land shows that 55.78 km<sup>2</sup> have been converted into built-up. Finally, the transition of water bodies into other classes depicts that 8.57 km<sup>2</sup> were converted into vegetation.

The overall assessment of transition within 2001 to 2014 depicts that the subsequent conversion of vegetation into built-up, waste lands into built-up and water bodies into vegetation nearly doubled contrasted to the previous decade. From this analysis, it can be concluded that the major reason for the LULC changes in the study area was the undeniable influence of the anthropogenic activities, which were more elevated than that of the previous decade.

The percentage transition of each class illustrates that 2.69 per cent from the total vegetative area has been converted into built up. The transition of waste land displays 46.01 per cent has been changed into built up. Lastly, the transition of water bodies represents that 0.17 per cent has been converted into built up and 10.98 per cent into vegetation. The percentage evaluation of the transition between 2001 and 2014 depicts that the waste land have transitioned highly into built up, which is almost 50 per cent of its total area and the water bodies into vegetation which consists of 11 per cent.

### Land Use and Land Cover Transition (2001-2014) Mysore-Mandya Corridor Region



**Figure 7:** Land Use Land Cover Transition (2001 - 2014)

## CONCLUSION

The study of changes in land use and land cover in Mysore-Mandya corridor depicts that, spatial extension wise the built-up land use has been increasing continuously while other land uses have been decreasing constantly, notably the increasing built-up land use between the years 1991 and 2001 was 49.25 km<sup>2</sup>; then between the years 2001 and 2014 was 96.99 km<sup>2</sup> this states that, the changes had occurred highly in the second period than the first period. The result of land transition represents that, during the period 1991 to 2001 vast area of vegetation and waste land have been converted into built-up, but it is less when compared with the second period between 2001 and 2014 when rapid amount of transition had occurred. This result from the anthropogenic activities of increasing built-up land use on vegetation and water bodies are higher during the second period. Based on the results of present study it can be concluded that, the changes of LULC in the study area is happening like other places, but the conversion of vegetation and waste land into the built-land is high which may cause problem to the environment if the same trend continues. From the study it can be suggested that, there is a need to analyses the existing land conversion policies

and rules and implement new necessary policies and rules to control the high rate of conversion of vegetation in built-up land which may protect the future environment.

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