

# **Monitoring and analysis of Land use/cover Change Detection and Urban Expansion from 1988 to 2014 in the South Seattle, Washington, USA**

Eric M. Baer<sup>1</sup>, Mushtak T. Jabbar<sup>2</sup>

Professor, Department of Geology, Highline College, MS 29-3, 2400 S. 240th St., Seattle, Washington, 98198, USA<sup>1</sup>

Professor, Department of Geology, Highline College, MS 29-3, 2400 S. 240th St., Seattle, Washington, 98198, USA<sup>2</sup>

**ABSTRACT:** In this study we integrate multiple pre-processing, land use/cover classification, and change analysis detection techniques to develop a methodology and assess urban expansion change between 1988 and 2014 in the southern part of King County in Washington, USA using a 1:150 000 mapping scale. The results reveal an overall 2.38 % growth in the urban area. Output land use/cover maps show that urban expansion has been associated with loss of agricultural land. Urban expansion was mainly attributed to the high growth in population, including a large number of immigrants from neighboring countries and other socio-economic changes. These results can provide city planners and decision makers with information about the past and current spatial dynamics of land use/ cover change and strictly urban expansion towards successful management for better planning and environmental monitoring.

**KEYWORDS:** Geo-information techniques; Classification; Index; Seattle; USA.

## **I. INTRODUCTION**

Urban expansion has been the defining global phenomenon during the past century. Particularly in the developing countries, urban population increase has been dramatically fast. In the time period between 1950 and 2005, annual urban population growths in Africa, Asia and Latin America have been 3.9%, 2.9% and 4% while those in Europe and North America are 1.2% and 1.9% respectively [1]. Between 2000 and 2030, the urban population in developing countries is expected to double, and entire built-up areas are projected to triple if current trends continue. In this process, various land-use problems emerged like arable land loss, land pollution and erosion, etc.

The study location was selected based on the background experience where urban expansion was prevalent, since it had approximately 32% of the total population of Washington, USA. During the last 25 years urban expansion was very noticeably driven by high population growth rate of 2.1% [2].

Previous studies have relied on remotely sensed information to classify land use/cover types either use raw data number values or calibrated reflectance, if a more precise material identification is needed. Geological and biological applications of TM data have also used simple band ratio techniques and spectroscopy to accentuate spectral features indicative of specific surficial materials, biomass, and vegetation health [3-7]. Land use/cover classification of urban areas has been problematic due to the heterogeneity and small spatial size of the surficial materials, which leads to significant subpixel mixing [8-9]. This problem becomes exacerbated when discrimination of multiple classes is necessary. Significant improvements in the accuracy of land cover classification in urban areas have been achieved using a variety of sophisticated approaches including: (1) the use of neural networks [10-12]; (2) fuzzy classification [13-15]; and (3) image texture analysis [16-18].

# International Journal of Innovative Research in Science, Engineering and Technology

(An ISO 3297: 2007 Certified Organization)

Vol. 4, Issue 4, April 2015

The objective of this study is to detect and evaluate the landuse/cover change due to urbanization between 1988 and 2014, and to analyze the main factors governing urbanization and land use/cover change to inform urban planners of the changing conditions in the study area in order to develop urban planning strategies that can protect the local environment. To perform the objectives of this study, two time series analyses using remotely sensed data in combination with Geographical Information System (GIS) were employed. Specifically, land use/cover classification schemes for seven cities in the southern part of King County were identified using Landsat Thematic Mapper (TM) and Enhanced Thematic Mapper (ETM+); (1) to retrieve the two phases of urban boundary using supervised classification and NDBI respectively followed by comparison of the result derived from these two different methods, and (2) to map urban land use/cover, thus to identify the factors that caused land use/cover changes during the past 26 years; and (3) to propose strategies for development, while protecting limited farmland resources and ecological environment.

## II. MATERIALS AND METHODS

### Materials

**Study Area:** The study region selected was the southern part of King County in Washington state, USA (Fig. 1), representing a total area of 249.434 km<sup>2</sup> between longitude -122° 42' to -121° 99' W and latitude 47° 26' to 47° 51' N. To study the urban expansion monitoring, the cities of Burien, Des Moines, Federal Way, Kent, Normandy Park, SeaTac, and Tukwila were selected as a study area. The cities are situated in the southern parts of Seattle. Average population growth was estimated at 2.1% since 1990 [19-20]. In recent years, along with the increase of population, economic development, and improvement of investment environment, Seattle has witnessed a rapid urbanization, which has deeply affected the ecological environment in the urban and rural regions.

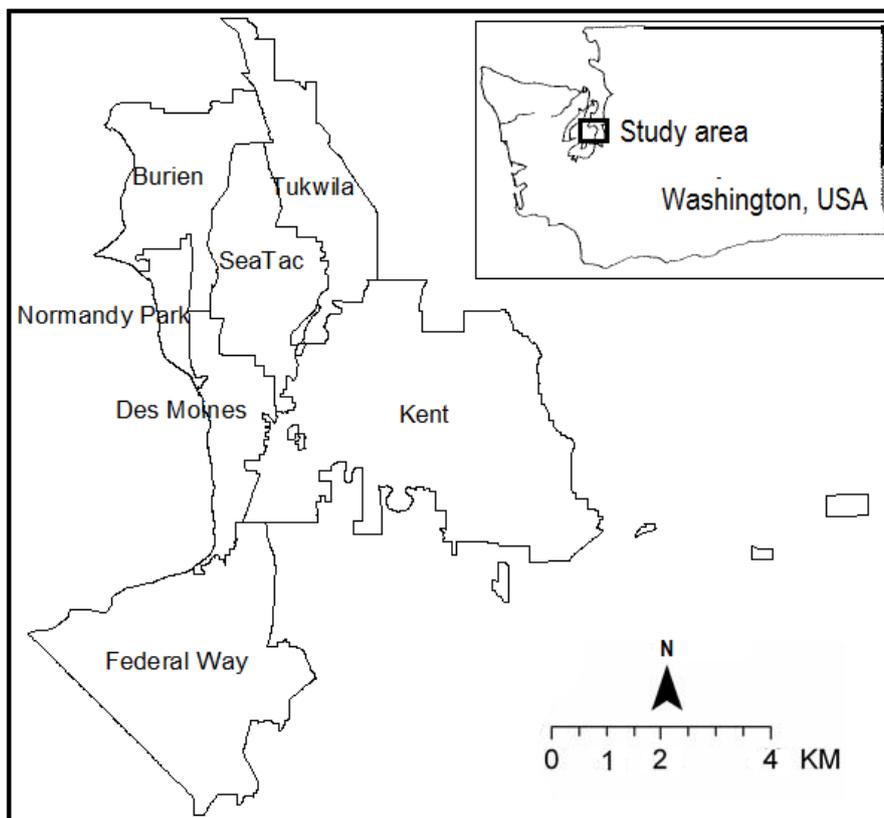


Fig. 1: General location of study area showing cities boundary.

# International Journal of Innovative Research in Science, Engineering and Technology

(An ISO 3297: 2007 Certified Organization)

Vol. 4, Issue 4, April 2015

**Spatial Database Using GIS and Remote Sensing:** In this study used a Landsat 5 TM image of July 1988 and a Landsat 7 ETM+ image of July 2014. The orbit data of the two images is 46/27, and their spatial resolution of 28.5 m. A city-level topographic map, geological map, soil map, meteorological data and all thematic layers were generated in a GIS environment at a scale of 1:150 000. The software's packages used for this study were (ERDAS ver. 9.2), GIS (ArcGIS ver. 10.1), and SPSS (statistical analysis).

## Methods

**Imagery pre-processing:** The pre-processing for the dataset included image registration, radiometric calibration, and radiometric normalisation. Rectification and registration of TM and ETM+ imageries were based on control points collected from vector files of the large and small rivers at the study area using forty ground control points (GCP). The remotely sensed dataset were geometrically corrected in the datum WGS84 and projection UTM N10 using the first order (linear) of polynomial function and nearest neighbor rectification re-sampling, which was chosen in order to preserve the radiometry and spectral information in the imagery. Image to image registration was done in order to register the ETM+ image (dated 2014) with geo-coded TM image dated 1988 (master image). The RMS error of the image-to-map was 0.38 to 0.49 pixels, while was 0.19 to 0.25 pixels with image to image registration. The Landsat imageries were radiometric calibrated for sensor differences, converted into spectral radiance and normalized for illumination properties through differences in sun-elevation angle and sun-earth distance by recalculating the pixel values into at-satellite reflectance [7].

**Imagery post-processing:** The basic idea of this research was to retrieve the urban boundary of different image phases using Landsat TM/ETM+ data, and then to analyze the impacts of urban expansion on land use/cover change (LUCC) (Fig. 2). A detailed illustration of data processing follows:

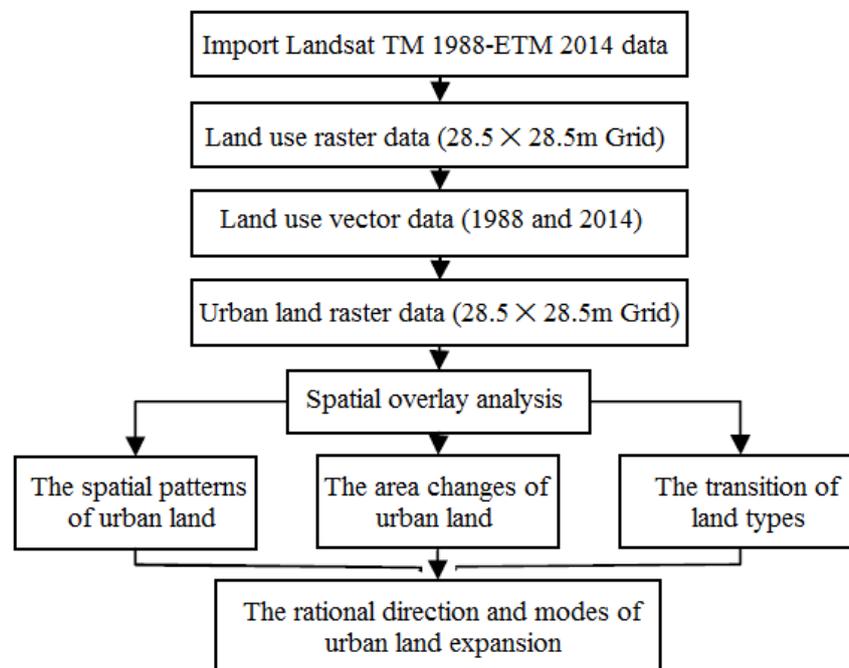


Fig. 2: The flow chart of analyzing of urban land changes

**Classification System:** A supervised classification system using maximum likelihood algorithm was applied for land use/cover mapping from the three images. To map changes that had occurred in the southern part of King County from

# International Journal of Innovative Research in Science, Engineering and Technology

(An ISO 3297: 2007 Certified Organization)

Vol. 4, Issue 4, April 2015

1988 to 2014, six non-thermal bands of both Landsat TM and ETM+ images were individually used as input for maximum likelihood classification system.

**Normalized Different Built-up Index (NDBI):** NDBI was used to retrieve urban land from the Landsat imagery. The NDBI which is sensitive to the built-up area [21-22], and was created by analyzing the spectral characteristics of different land use/cover types.

$$NDBI = [(band5) - (band4)] / [(band5) + (band4)]$$

Where, bands (band4 and band5) of the Landsat TM or ETM+ images.

**Rates of Land Use/Cover Change:** Many models and indicators exist that are used to analyse the magnitude, rate and trend of LUCC [23-24]. This study relies on statistics to determine actual conversion rates of urban expansion. The following equation was used to calculate conversion rates [25]:

$$X = \left[ 1 - \frac{C_2 - C_1}{C_1} \right]^{1/n} - 1$$

Where X is the change rate of the land use and land cover,  $C_1$  is the class area at time  $t_1$  (1988);  $C_2$  is the class area at time  $t_2$  (2014); n the difference in years between the two dates (i.e. 26 years in this study). The study made use of a transition matrix to describe land conversion over the entire study period. A transition probability was adopted to conduct a trend analysis of landscape patch dynamics [26].

### III. RESULTS AND DISCUSSION

#### Assessment of land use/cover changes

The GIS capabilities allowed the post-classification comparisons, and facilitated a qualitative assessment of the factors influencing urban expansion. There are seven land use/cover classes of interest in south part of King County: water, forested, forested urban, grass shrub crops, grass shrub urban, paved and bare land, and urban/built up areas. The spatial distributions of these classes were extracted from each of the land use/cover maps by using of GIS spatial analysis. The statistic land use/cover distribution for the two time series (1988 and 2014) as derived from the maps are presented in Fig 3.

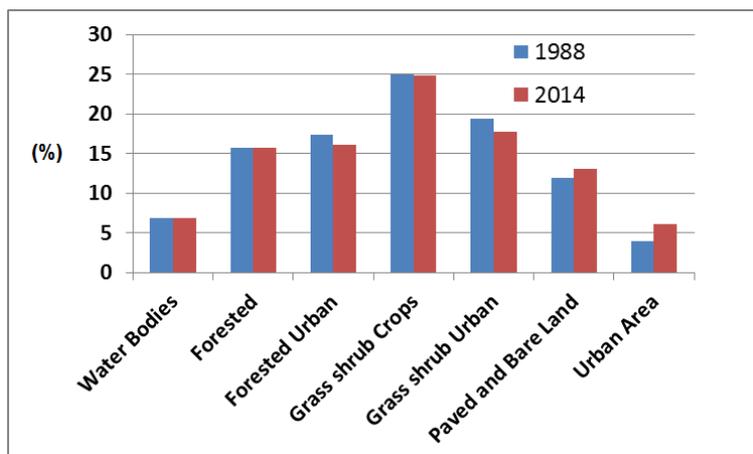


Fig. 3 :( LUCC) classes monitored from satellite image for the study area.

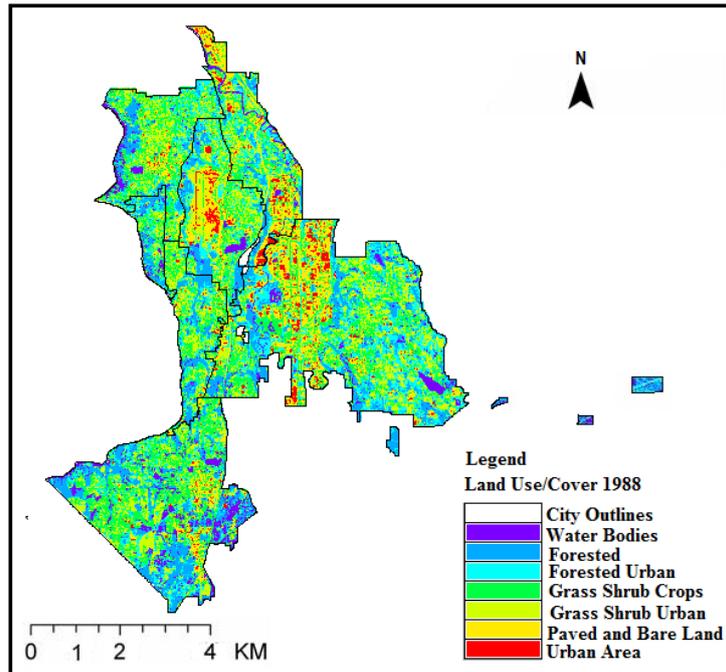
The data revealed that during the 26-year period several land use/cover changes occurred; in particular a 18.1% decrease in vegetation areas was highlighted, corresponding to an average loss of  $0.29\text{km}^2\text{year}^{-1}$ . On the contrary,

## International Journal of Innovative Research in Science, Engineering and Technology

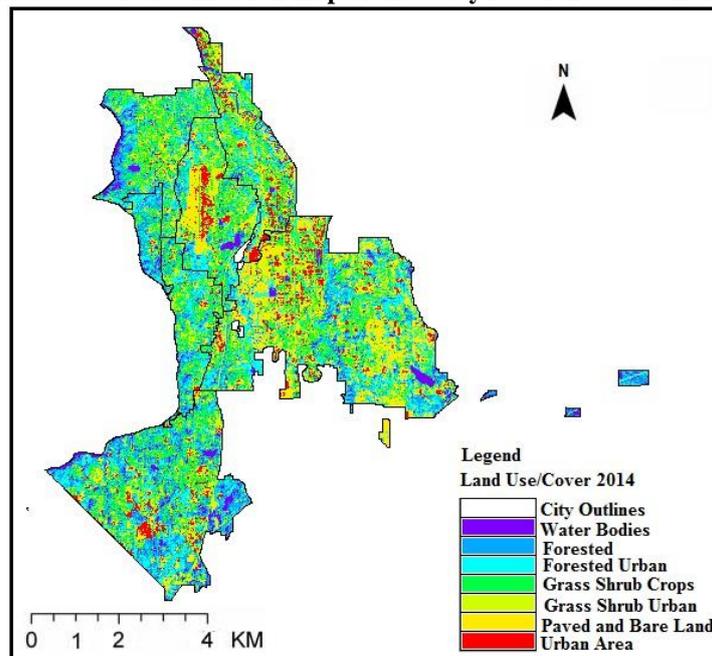
(An ISO 3297: 2007 Certified Organization)

Vol. 4, Issue 4, April 2015

paved and urban/built up areas increased by 13.1% and 6.1% respectively. The spatial distributions of the seven classes were extracted from each landuse/cover maps of 1988 and 2014 as shown in Fig.4A and 4B.



**Fig.4A: Land use/cover classification map of the study location based on analysis of TM.**



**Fig.4B: Land use/cover classification map of the study location based on analysis of ETM.**

## International Journal of Innovative Research in Science, Engineering and Technology

(An ISO 3297: 2007 Certified Organization)

Vol. 4, Issue 4, April 2015

Using Table 1 and Fig. 3 together, significant changes in the LUCC can be recognized and type of LUCC conversion taking place can be identified. From 1988 to 2014 urban/built up areas saw an increase. Primarily due to a loss of vegetated land. This may suggest logging and development. During the same time span, paved and bare land also increased in size. The majority of this change came from the development of vegetation land into an urban class. Some of the marshes converted to unused land, while some of it was converted to urban. Finally vegetation areas saw a decrease in size. The largest portion was converted to urban/built up areas and a significant portion was developed in to paved and bare land.

**Table 1: Transition Matrix (Km<sup>2</sup>) of land use/cover maps (1988 and 2014)**

|             |                | 1988   |        |        |        |        |        |        |             |
|-------------|----------------|--------|--------|--------|--------|--------|--------|--------|-------------|
|             | Transition     | W.B.   | F      | F.U.   | G.C.   | G.U.   | B.L.   | U.A.   | 2014 Totals |
| <b>2014</b> | Water Bodies   | 2.264  | 3.015  | 1.825  | 2.423  | 2.095  | 1.942  | 2.247  | 15.811      |
|             | Forested       | 5.123  | 5.039  | 7.483  | 6.527  | 8.126  | 4.816  | 2.311  | 39.425      |
|             | Forested Urban | 1.481  | 6.796  | 6.124  | 11.186 | 9.184  | 4.818  | 1.087  | 40.676      |
|             | Grass Crop     | 2.126  | 11.036 | 13.069 | 18.299 | 11.469 | 4.913  | 1.143  | 62.055      |
|             | Grass Urban    | 2.168  | 6.942  | 6.316  | 10.328 | 9.842  | 7.129  | 1.116  | 43.841      |
|             | Bare Land      | 2.618  | 4.316  | 7.129  | 9.923  | 5.294  | 2.516  | 0.879  | 32.675      |
|             | Urban Area     | 0.792  | 2.004  | 1.546  | 3.585  | 2.352  | 3.437  | 1.235  | 14.951      |
|             | 1988 Totals    | 16.572 | 39.148 | 43.492 | 62.271 | 48.362 | 29.571 | 10.018 |             |

As the accuracy assessment required intensive visual analysis, the sub-change categories of each land cover type were aggregated into a single change class [27-29]. Therefore, the 10 different classes were aggregated into seven major subcategories, i.e., Water, Forested, Forested Urban, Grass shrub Crops, Grass shrub Urban, Paved and Bare Land, and Urban/built up areas. The overall accuracy, kappa, producer and user accuracy values are presented in Table 2.

**Table 2: Information of accuracy assessment data used in this study.**

| Class                                | Produce accuracy (%) | User accuracy (%) |
|--------------------------------------|----------------------|-------------------|
| Water Bodies                         | 76.89                | 95.25             |
| Forested                             | 81.76                | 93.56             |
| Forested Urban                       | 75.88                | 86.38             |
| Grass shrub Crops                    | 77.45                | 89.21             |
| Grass shrub Urban                    | 76.96                | 85.58             |
| Paved and Bare Land                  | 78.35                | 81.59             |
| Urban Area                           | 75.91                | 83.97             |
| Overall accuracy = (215/256) = 83.98 |                      |                   |
| Kappa coefficient = 0.78             |                      |                   |

Fig. 5 shows change rate of the land use and land cover processes between 1988 and 2014 with all lands grouped in two clusters. (1) Decreased of vegetation land for more than 1.2 % of the total land area and water surface loss for more than 0.2 per cent of the total land area, (2) More than 2.1 % of urban/built up areas and 1.2 % of the paved land was increased. The changes and spatial patterns of environment degradation processes during 1990–2003 will be analysed hereafter.

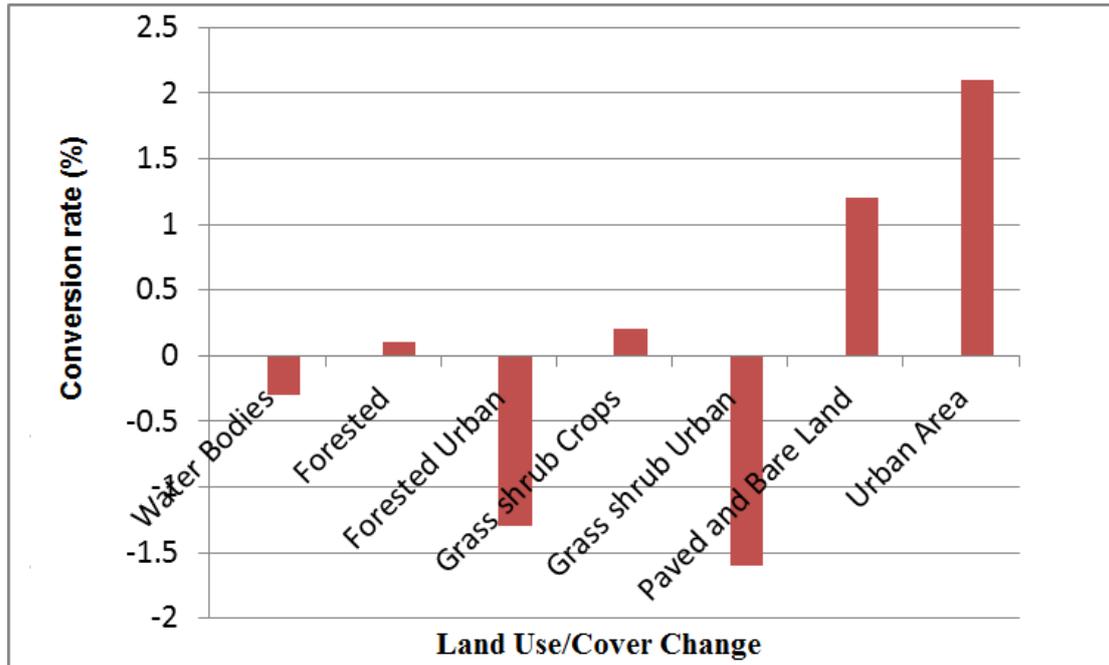
According to statistics of land use/cover thematic map, the total study area is 249.434 km<sup>2</sup>. There into, area of forested urban is about -2.816 km<sup>2</sup> of total area, forested is about 0.277 km<sup>2</sup>, grass shrub crops is about -0.216 km<sup>2</sup>, grass shrub urban is about -4.521 km<sup>2</sup>, paved and bare land is about 3.104 km<sup>2</sup>, and urban/built up area is about 4.933 km<sup>2</sup> (Fig. 3) . In the Table1, we calculated areas affected by various levels of land use/cover change in the southern part of King

## International Journal of Innovative Research in Science, Engineering and Technology

(An ISO 3297: 2007 Certified Organization)

Vol. 4, Issue 4, April 2015

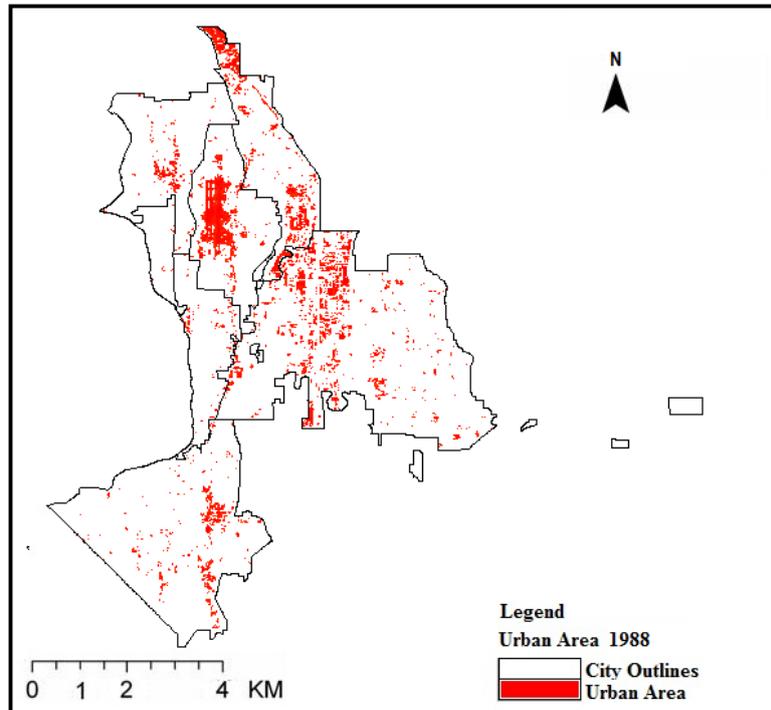
County in Washington. It is supposed that the whole area is subject to urbanization, as we mentioned previously, mainly by anthropogenic activities.



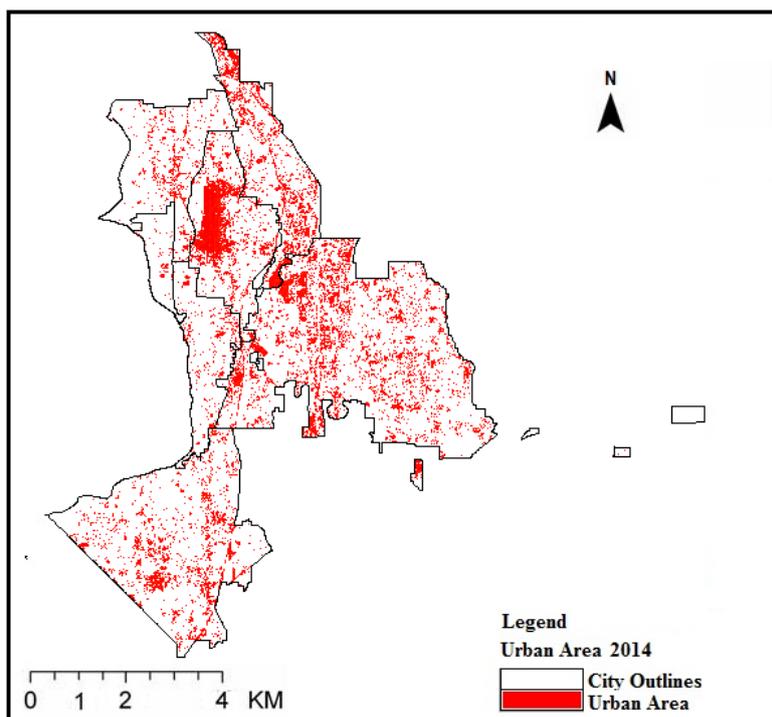
**Fig.5. Change rate of the land use and land cover types expressed in percentages**

### *Assessment of Urban Expansion*

In this research, the differences in urban area of seven cities between 1988 and 2014 were analyzed (Fig.6A and Fig.6B). The measurements make it possible to notice a significant increase of urban/built up area with respect to the time interval. The urban land cover increased from 3.6% to 6.1% (Table 3). The data reveals a significant increase in urban area with time; however, the magnitude of increase varied by location. The southern part of King County in Washington is considered a great economic and administrative center, the increase in its urban area was 14.9 Km<sup>2</sup> from 1988 to 2014. Although the cities of Burien, Des Moines, Federal Way, Kent, Normandy Park, SeaTac, and Tukwila all experienced an increase in urbanization, with increases of 1.2%, 2.7%, 10.7%, 1.7%, 7.9%, 1.9%, and 1.5%, respectively. The statistical data of the census showed a general increase in the population of the whole study area. Population in the region increased from 166,389 in 1988 to 328,819 in 2014 with an accompanying increase in population density from 0.671 to 1.318 thousand Per/Km<sup>2</sup>(Table 4). From Table 4 we know that the changes of total population level of study location had presented a tendency to increase constantly in the past 26 years. In this period, the population has increased 162,430 with an annual urbanization growth rate of 2.5% per year (Table 3). The results of the statistical analysis showed that urban area has a significant correlation with increase population positive change (0.92). In general, the population is the major factor that influenced the urban expansion.



**Fig. 6A:** The spatial change of the urban expansion of the study location in 1988.



**Fig. 6B:** The spatial change of the urban expansion of the study location in 2014.

## International Journal of Innovative Research in Science, Engineering and Technology

(An ISO 3297: 2007 Certified Organization)

Vol. 4, Issue 4, April 2015

**Table 3: Urban expansion statistics results for the study area during the period from 1988 to 2014.**

| Study site    | Area (Km <sup>2</sup> ) | Area(Km <sup>2</sup> ) |      |        |       | Increase (Km <sup>2</sup> ) | Percent (%)        | Increase (%)      |
|---------------|-------------------------|------------------------|------|--------|-------|-----------------------------|--------------------|-------------------|
|               |                         | 1988                   | (%)  | 2014   | (%)   |                             |                    |                   |
| Burien        | 26.785                  | 0.369                  | 1.38 | 0.487  | 1.82  | 0.118* <sup>a</sup>         | 31.9* <sup>b</sup> | 1.2* <sup>c</sup> |
| Des Moines    | 16.812                  | 0.262                  | 1.56 | 0.447  | 2.66  | 0.185                       | 70.6               | 2.7               |
| Federal Way   | 57.931                  | 0.787                  | 1.36 | 2.983  | 5.15  | 2.196                       | 279.0              | 10.7              |
| Kent          | 89.658                  | 4.330                  | 4.83 | 6.267  | 6.99  | 1.937                       | 44.7               | 1.7               |
| Normandy Park | 6.844                   | 0.028                  | 0.41 | 0.084  | 1.24  | 0.056                       | 200.0              | 7.9               |
| SeaTac        | 26.567                  | 1.418                  | 5.34 | 2.138  | 8.05  | 0.72                        | 50.7               | 1.9               |
| Tukwila       | 24.837                  | 1.862                  | 7.05 | 2.592  | 10.44 | 0.73                        | 39.2               | 1.5               |
| Total         | 249.434                 | 9.056                  | 3.63 | 14.998 | 6.01  | 5.942                       | 65.6               |                   |

**Table 4: The changes of total population growth in different cities.**

| Study site    | Area (Km <sup>2</sup> ) | % area | 1988                 |      |                                | 2014       |      |                                | Percentage Growth  |
|---------------|-------------------------|--------|----------------------|------|--------------------------------|------------|------|--------------------------------|--------------------|
|               |                         |        | population           | (%)  | Density (Per/Km <sup>2</sup> ) | population | (%)  | Density (Per/Km <sup>2</sup> ) |                    |
| Burien        | 26.785                  | 10.7   | 30,000* <sup>1</sup> | 18.1 | 1.120                          | 34,129     | 10.4 | 1.274                          | 13.7* <sup>3</sup> |
| Des Moines    | 16.812                  | 6.8    | 14,120               | 8.4  | 0.841                          | 30,030     | 9.2  | 1.786                          | 112.6              |
| Federal Way   | 57.931                  | 23.2   | 71,678* <sup>2</sup> | 43.1 | 1.281                          | 90,225     | 27.4 | 1.557                          | 25.8               |
| Kent          | 89.658                  | 35.9   | 32,350               | 19.4 | 0.361                          | 121,400    | 36.9 | 1.354                          | 275.2              |
| Normandy Park | 6.844                   | 2.8    | 6,130                | 3.6  | 0.896                          | 6,375      | 1.9  | 0.931                          | 3.9                |
| SeaTac        | 26.567                  | 10.7   | 7,351* <sup>2</sup>  | 4.5  | 0.277                          | 27,450     | 8.4  | 1.033                          | 273.4              |
| Tukwila       | 24.837                  | 9.9    | 4,760                | 2.9  | 0.192                          | 19,210     | 5.8  | 0.773                          | 303.5              |
| Total         | 249.434                 | 100    | 166,389              | 100  | 0.671                          | 328,819    | 100  | 1.318                          | 97.6               |

\*<sup>1</sup> 1998, \*<sup>2</sup> 1996, \*<sup>3</sup> Amount of change / population 1988 \* 100

#### IV. CONCLUSION

Evaluation of land use/cover patterns change is necessary not only for urban planning purposes, but also to improve the management of land resources. This research has demonstrated the importance of using satellite remote sensing and digital image processing together with GIS technique in producing accurate land use/cover maps and change statistics for the southern part of King County in Washington, which is valuable to monitor urban expansion effectively over a time. The change detection results of the study location show that urban/built-up areas covered 9.056 km<sup>2</sup> (3.6%) in 1988 and 14.998 km<sup>2</sup> (6.1%) in 2014. This represents a net increase of 5.942 km<sup>2</sup>, which is mainly attributed to the increase population due to the high growth rate, large number of immigrants and other socio-economic changes. In particular, during the period study a high percentage of agricultural lands were converted to urban areas and paved land. However, urban expansion was increased during the same period, which is considered as a key indicator that urban planning strategy should be given more attention. The preprocessing stage, such as the geometric corrections of multi-temporal satellite images as well as using the same classification scheme, is very important to enhance the accuracy of the results. This study concluded the advantage of remote sensing and GIS techniques that are indispensable for dealing with the dynamics of land use/cover change in the south part of Seattle over the last 26 years. Therefore, it is highly recommended that urban planners and decision makers can use remote sensing and GIS techniques for effective monitoring of urbanization trends. It will improve their predictions toward the amount of urbanization changes and the location of future built-up areas, and enhance the existing urban strategies for better sustainable land management.

#### V. ACKNOWLEDGMENTS

This research was supported by the Highline College. The authors are grateful to the anonymous reviewers for their critical review and comments on drafts of this manuscript.

# International Journal of Innovative Research in Science, Engineering and Technology

(An ISO 3297: 2007 Certified Organization)

Vol. 4, Issue 4, April 2015

## REFERENCES

- [1] UN, World Urbanization Prospects. United Nations Publications, New York, 2006.
- [2] Robinson L, Newell JP, Marzluff JM "Twenty-five years of sprawl in the Seattle region: growth management responses and implications for conservation. Landscape urban" <http://dx.doi.org/10.1016/j.landurbplan.2004.02.005>, 2005.
- [3] Huete, A. R. "A soil-adjusted vegetation index (SAVI)", Remote Sensing of Environment, 25, 295–309, 1988.
- [4] Lyon, J. G., Yuan, D., Lunetta, R. S., & Elvidge, C. D. "A change detection experiment using vegetation indices". Photogrammetric Engineering and Remote Sensing, 64 (2), 143–150, 1998.
- [5] Mattikalli, N. M. "Soil color modeling for the visible and near-infrared bands of Landsat sensors using laboratory spectral measurements" Remote Sensing of Environment, 59, 14–28, 1997.
- [6] Sultan, M., Arvidson, R. E., Sturchio, N. C., & Guinness, E. A. "Lithologic mapping in arid regions with Landsat Thematic Mapper data": Meatiq dome, Geological Society of American Bulletin, 99, 748–762, 1987. Aria Pezeshk and Richard L. Tutwiler, "Automatic Feature Extraction and Text Recognition from Scanned Topographic Maps", IEEE Transactions on geosciences and remote sensing, VOL. 49, NO. 12, 2011
- [7] Jabbar, M. T., Zhi-Hua Shi, Tian-Wei Wang, and Chong-Fa CAT, "Vegetation change prediction with geo-information techniques in the three gorges area of China", Pedosphere, 16, no. 4, pp. 457-467, 2006.
- [8] Foody, G. M. "Estimation of sub-pixel land cover composition in the presence of untrained classes" Computers & Geosciences, 26, 469–478, 2000. Mr. Rajesh H. Davdal, Mr. Noor Mohammed, "Text Detection, Removal and Region Filling Using Image Inpainting", International Journal of Futuristic Science Engineering and Technology, vol. 1 Issue 2, ISSN 2320 – 4486, 2013
- [9] Ridd, M. K. "Exploring a V-I-S (vegetation-impervious surface-soil) model for urban ecosystem analysis through remote sensing: comparative anatomy for cities" International Journal of Remote Sensing, 16, 2165–2185, 1995.
- [10] Berberoglu, S., Lloyd, C. D., Atkinson, P. M., & Curran, P. J. "The integration of spectral and textural information using neural networks for land cover mapping in the Mediterranean" Computers & Geosciences, 26, 385–396, (2000).
- [11] Kumar, A. S., Basu, S. K., & Majumdar, K. L. "Robust classification of multispectral data using multiple neural networks and fuzzy integral" IEEE Transactions on Geoscience and Remote Sensing, 35, 787–790, 1997.
- [12] Paola, J. D., & Schowengerdt, R. A. "A detailed comparison of back propagation neural networks and maximum likelihood classifiers for urban land use classification" IEEE Transactions on Geoscience and Remote Sensing, 33, 981–996, 1995.
- [13] Bastin, L. "Comparison of fuzzy c-means classification, linear mixture modeling and MLC probabilities as tools for unmixing coarse pixels" International Journal of Remote Sensing, 18, 3629–3648, 1997.
- [14] Fisher, P. F., & Pathirana, S. "The evaluation of fuzzy membership of land cover classes in the suburban zone" Remote Sensing of Environment, 34, 121–132, 1990.
- [15] Foody, G. M. "Estimation of sub-pixel land cover composition in the presence of untrained classes" Computers & Geosciences, 26, 469–478, 2000.
- [16] Gong, P., & Howarth, P. J. "The use of structural information for improving land-cover classification accuracies at the rural–urban fringe" Photogrammetric Engineering and Remote Sensing, 56 (1), 67–73, 1990.
- [17] Iron, J. R., & Petersen, G. W. "Texture transforms of remote sensing data" Remote Sensing of Environment, 11, 359–370, 1981.
- [18] Stuckens, J., Coppin, P. R., & Bauer, M. E. "Integrating contextual information with per-pixel classification for improved land cover classification" Remote Sensing of Environment, 71, 282–296, 2000.
- [19] Office of Financial Management State of Washington (2007) Population projections for Washington State. Available from <http://www.ofm.wa.gov/pop/gma/projections07.asp> (accessed 1 July 2008)
- [20] PSRC, "Decennial Change in Population and Land Area of Cities, Towns, and Counties in the Central Puget Sound: 1990 to 2000". US Census 2000. P.L. 94-171 Redistricting Data. Puget Sound Regional Council, 2001.
- [21] Zha Y.; Gao J. and Ni S, "Use of normalized difference built-up index in automatically mapping urban areas from TM imagery" International Journal of Remote Sensing, 24(3), 583-594, 2003.
- [22] Zhao, H. M., & Chen, X. L. "Use of normalized difference bareness index in quickly mapping bare areas from TM/ETM+" *Geoscience and Remote Sensing Symposium*, 3(25–29), 1666–1668, 2005.
- [23] Awasthi KD, Sitaula BK, Singh BR "Land-use change in two Nepalese watersheds: GIS and geomorphometric analysis" J Land DegradDev 13:495–513, 2002.
- [24] Jabbar, Mushtak T., and Jing-xuan Zhou., "Environmental degradation assessment in arid areas: a case study from Basra Province, southern Iraq", Environmental Earth Sciences, 70, no. 5, pp. 2203-2214, 2013.
- [25] Velazquez A, Duran E, Ramirez I et al., "Land use-cover change processes in highly biodiverse areas: the case of Oaxaca, Mexico" Glob Environ Chang 13:175–184, 2003.
- [26] Jia BQ, Zhang ZQ, Ci LJ et al., "Oasis land-use dynamics and its influence on the oasis environment in Xinjiang, China" J Arid Environ 56:11–26, 2004.
- [27] Zhou W, Troy A, Grove M "Object-based land cover classification and change analysis in the Baltimore Metropolitan area using multitemporal high resolution remote sensing data" Sensors 8:1613–1636, 2008.
- [28] Jabbar, Mushtak T., and Xiping Zhou., "Eco-environmental change detection by using remote sensing and GIS techniques: a case study Basrah province, south part of Iraq", Environmental Earth Sciences, 64, no. 5, pp. 1397-1407, 2011.  
Yanli, You, Mushtak T. Jabbar, and Jing-Xuan Zhou, "Study of environmental change detection using Remote Sensing and GIS application: A case study of northern Shaanxi Province, China", Polish Journal of Environmental Studie,s 21, no. 3, 2012.