1 ABSTRACT
Since 1980th, rapid population growth and urbanization have become issues in big cities in developing countries like Greater Cairo. As a consequence of explosive growth, the living conditions of Cairo Metropolis deteriorate. Development trends of the last twenty years have increased general wealth and modernization, at a time of a high rate of population growth, thus creating an increased demand for land combined with environmental degradation.

Planning a sustainable development of urban areas requires understanding of growth dynamics of urban systems. This talk will be concerned with monitoring and analysis of dynamic environment to capture and refine the urban patterns in Greater Cairo Metropolis on the basis of pixel-based and object-based classifications. Satellite images (TM, ETM+, & Spot) of different dates and resolutions, and ground truth data collected from available maps, field observation, and personal experience were used to execute the image segmentation analysis to reveal urban patterns and expansions.

By using Erdas Imagine software, land use/land cover image classifications were constructed, which showed regimes and trends in urban growth change.

The presented maps based on object segmentation have more accurate results than the produced maps based on pixel classification module.

Two main types of urban patterns could be detected (passing from center to periphery). The first one is informal and the second one is formal building. The informal type mainly comprises slums and urban encroachment on arable land. The formal one mostly consists of new cities and legal houses. Moreover, a rate of urbanization growth during the last three decades would be described as geometrical progression.

2 INTRODUCTION
Nowadays, three out of six human beings live in cities, two of whom in developing cities. These figures illustrate the scale of the challenge represented by urban development in today’s world and the growing pressure it currently puts on all levels of the city.

Due to population growth, the urban and metropolitan sprawl in developing cities will continue to increase the inadequacy of these cities’ capacities-institutional, financial and technical capacities, but also human and even conceptual capacities- in the face of “urban issue” that is harder and harder to grasp. Therefore, research on how to describe rapid increasing of megacities by urbanization and especially land conversion of peri-urban environments has been identified as a pivotal area of current and future land change research.

Urban planners and administrative bodies require reliable information to assess the consequences of urbanization, to ensure a sustainable functioning of megacities and to minimize negative impacts of rapid urbanization. Urban expansion in the developing world often takes place in unplanned manner and administration is unable to keep track of growth related processes (Griffiths et al., 2010).

Remote sensing based information is one of the most important resources to support urban planning and administration in megacities (Maktav et al., 2005). The remotely sensed information is hence of particular relevance in the world's developing countries because it provides fundamental information on growth related processes and their effect on the urban environment that are not available from other sources (Miller & Small, 2003).

However, satellite imagery provides a synoptic view of the urban mosaic; remote sensing data can be an important complement to limited in situ measurements in urban areas. Since Pixel based approach is based on conventional statistical techniques, such as supervised and unsupervised classification. In supervised classification the image analyzed “supervised” the pixel categorization process by specifying, to the computer algorithm, numerical description of the various land cover types present in a scene. This approach has shown good accuracy for images acquired by course resolution sensors, while in the case of very high resolution (VHR) images it was considered that the spectral mutability increases within a particular class,
making the extraction of thematic information more difficult (Foody 2002). Various new techniques take into account, besides the spectral data, also the texture features of the image as additional layer in classification process (Puissant et al., 2005). An improved accuracy, especially for urban land-use/cover classifications, has been proposed by the object-based classifications. Hence, object-based analysis facilitates work with meaningful image objects and their mutual relationships (Matinfar et al., 2007).

This paper describes and discusses different remote sensing (RS) techniques to monitor and analyse of dynamic expansion and urbanization in Greater Cairo. Since pixel-based analysis in complexly structured land use and land cover (LU/LC) areas is limited because the semantic information necessary to interpret an image is usually not represented in single pixel, both pixel-and-object-based approaches were applied and compared. This involves a certain trade-off between the richness of detail of very high resolution (VHR) remote sensing imagery such as Spot Image and the generalizing nature of moderate to high resolution sensors such as the 30 m spatial resolution of Landsat TM and ETM+ used in this study.

### 3 STUDY AREA

The selected area of study is the metropolitan area of Greater Cairo and its surroundings, which known as the capital of Egypt and one of the fastest growing megacities worldwide, (Fig. 1). The area covers about 600 km², encompassing major parts of the governorates of Cairo, Giza, Six October, and Helwan. The Nile forms the administrative division between these governorates, with Cairo and Helwan on the east bank of the river and, Giza and Six October on the west bank. The area includes a variety of land uses associated with a complex mix of land cover. Such as a central business district (CBD) urban/suburban residential areas and some rural areas (e.g. cultivated areas and soil). This area has encountered rapid urban development and population growth in the last 20 years (Fig. 2).

![Fig. 1: the study area-Greater Cairo-Egypt](image)

### 4 MATERIALS AND METHODS

This study is based on three Landsat TM & ETM images dates (path 176, row 39): 1984, 1990, and 2006 (the image of 1984 consider as the reference year for the other images). Moreover, a multi-spectral spot mosaic acquired on 2006 has been chosen. These data were processed using ground control points and have a geometric error within 30, 15, and 5 m respectively.

#### 4.1 Image processing

A series of processing operations was performed on these images. The images were georeferenced using UTM map projection for zone 36 and datum of WGS84. The images were resampled to 30 m for 1, 2, 3,4,5,7
bands, 15 m for panchromatic and 60 m for thermal bands per pixel, using the nearest neighbor technique. In order to produce test area, false color composite from ETM+ bands of 7, 5 and 3 were used, while all of the six bands (ETM+ 1,2,3,4,5 and 7 bands) were used for classification by method pixel-based. Additionally, a subset of spot images mosaic of the study area was compiled, which were used for soft method of object-based classification.

![Figure 2: Population growth in Greater Cairo (modified after CAPMAS)](image)

4.2 Pixel-base classification

Supervised classification was performed using TM and ETM+ bands. In supervised classification, the basic steps followed are (1) select training samples which are representative and typical for that information class; (2) perform classification after specifying the training samples set and classification algorithms. Training samples are selected according to the ground truth. These homogenous areas are identified in the image to form the training samples for all of the information classes. The selected algorithm for performing the supervised classification is the minimum distance classification. In this algorithm first the mean spectral value in each band for each class is determined. These values comprise the mean vector for each class. This distance threshold could vary for each class depending on the expected degree of compactness of that class. Compactness might be estimated from the standard deviation for each feature of the pixels making up the training sample for a given class.

4.3 Object-base classification

Erdas Imagine 2010 was used for object-oriented analysis and classification. Segmentation is the main process in the classification module in Erdas software and its aim is to create meaningful objects. This means that an image object should ideally represent the pattern of each object in question. This pattern combined with further derivative color and texture properties can be used to initially classify the image by classifying the generated image objects.

Thereby the classes are organized within a class hierarchy. With respect to the multi-scale behavior of the objects to detect a number of small objects can be aggregated to form larger objects constructing a semantic hierarchy (Matinfar et al.; 2007). In performing the segmentation of Spot image, four spectral bands (5 m resolution 1, 2, 3, & 4) took in the segmentation process with a full weight.

5 RESULTS

Six land use/land cover (LU/LC) classes are recognized based change detection in the first level (pixel-based classification) to be distributed in the studied area: Urban areas (U), Cultivated Land (CL), Cultivated to Urban (CU), Desert (D), Bare Soil (BS), and Water Bodies (WB) (Fig. 3).

The following are shortcomings that limit the accuracy of classification based multi-resolution/multi-temporal change detection by using the pixel-based classification techniques:

- Limited spectral separation of classes.
- The statistical independence assumption pixel-based classifications involve the DN values individually without considering the neighbourhood pixels (Castelli et al., 1999).
Fig. 3: LC/LU multitemporal classified maps represented the change detection in study area

<table>
<thead>
<tr>
<th>LU/LC</th>
<th>1984 %</th>
<th>1990 %</th>
<th>2006 %</th>
</tr>
</thead>
<tbody>
<tr>
<td>U</td>
<td>12</td>
<td>25</td>
<td>45</td>
</tr>
<tr>
<td>CL</td>
<td>24</td>
<td>17</td>
<td>8</td>
</tr>
<tr>
<td>CU</td>
<td>23</td>
<td>22</td>
<td>18</td>
</tr>
<tr>
<td>D</td>
<td>35</td>
<td>32</td>
<td>24</td>
</tr>
<tr>
<td>BS</td>
<td>4</td>
<td>3</td>
<td>4</td>
</tr>
<tr>
<td>WB</td>
<td>2</td>
<td>1</td>
<td>1</td>
</tr>
</tbody>
</table>

Table 1: Percentage of LU/LC classes in the study area
• Land cover maps derived from classification of images usually contain some sort of errors due to several factors that range from classification techniques to methods of satellite data capture. Hence, evaluation of classification results is an important process in the classification procedure.

• The intrinsic limitation of classifiers- comparison of image data in a change detection procedure requires more research that would involve new approaches of multi-scale analysis.

Two levels for automatic identification of the urban classes on the multi-spectral spot mosaic were applied by image segmentation module in Erdas Imagine software for the **object-oriented classification**:

• Formal-informal urbanization level (Fig. 4)
• Informal urbanization segmentations level (Fig. 5)

![Fig. 4: The distribution of formal (green) and informal urbanization (red) in the study area](image)

![Fig. 5: The classes of informal urbanization in the study area according to the textural density and shape](image)
Objects are described by their specific spectral information, shape, and texture. The brightness relationship of adjacent image objects was worked out explicitly. For that purpose, the relationships between objects were established. Based upon the objects and the class descriptions, a fuzzy soft-classification was implemented. Whatever, shape characteristics such as size and density of an object, and neighbourhood characteristics for topological and brightness relationships analyses, were raised.

The first level classification of pixel-based method is integrated by the object-based segmentation, which represented the second level of classification. Objects in the second level classification were concerned to describe the formal and informal urbanization. Further classification process can be carried out to generate the third level of segmentation. This level is distinguished by four child classes (LUC1, LUC2, LUC3, LUC4). These child classes are described as low, moderate, high, very high dense urbanization respectively.

6 CONCLUSION

It can be concluded that the pixel-based classification, which is applied on Landsat TM and ETM+ time-series images, is using to monitor the dynamic increasing of megacities. While the object-oriented classification deals more sufficiently with the urban environment including formal and informal buildings. The high level of object oriented segmentation, which applied on the very high resolution images, is recommended for sub-object accuracy classification. Moreover, a rate of urbanization growth in Greater Cairo during the last three decades could be described as geometrical progression.

7 REFERENCES


