# GIS BASED MARKET SIZE ANALYSIS HOUSING PROPERTIES DEVELOPMENT USING REMOTE SENSING DATA 

## ZOUHEIR SALLMAN

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A Thesis Submitted to the College of Graduate Studies, Universiti Tenaga Nasional in Fulfilment of the Requirements for the Degree of

Master of Civil Engineering

## DECLARATION

I hereby declare that the thesis is my original work except for quotations and citations which have been duly acknowledged. I also declare that it has not been previously, and is not concurrently submitted for any other degree at Universiti Tenaga Nasional or at any other institutions. This thesis may be made available within the university library and may be photocopies and loaned to other libraries for the purpose of consultation.

## ZOUHEIR SALLMAN

Date: 07/11/2019


#### Abstract

Spatial data is among other important information in the marketing strategy. Urban growth and its spread changed the geographical extend to understanding the demand and existence of stocks. A business needs to know growth and market size accurately in order to better plan in a way of avoiding randomness in decision making. Recent days businesses have utilized geo-locations but still, many people do not use spatial data effectively in order to increase the accuracy of the results and make the work more efficient and flexible with the use of Geomarketing. In this study, remote sensing data is explored to play as an active role in order to provide accurate data on the growth of the buildings associated with population increases. In this study remote sensing data is explored to play as an active role in order to provide accurate data on the growth of the buildings associated with population increases. The population growth require food and buildings construction need building materials. The building growth will determine by explore the difference between two satellite images which related to two different dates. The resulting data will provide accurate information that will enable decision-makers to plan business as a result of obtaining accurate market readings. As a result, the production, sales demand or establish service centers can be provided in better ways.


## ACKNOWLEDGMENT

I would like to express my sincerely gratitude to my supervisor Dr. Fathoni Usman, Senior Lecturer, Department of Civil Engineering for providing me the full support and his valuable suggestion to carry out my research successfully.

I am grateful to Dr. Wong, Senior Lecturer, Department of Civil Engineering for his support during my submitted my request for the university.

## DEDICATION

I would like to express my sincere appreciation to my wife, my great daughters, for their encouragement and supporting me throughout the study.

My utmost gratitude for my mother's spirit, my father, my father in law for his prayer for me, my mother in law, Brothers and sisters
My utmost gratitude for Dr. Osama Ammar who planted the idea in my head and his supporting during my working in General Organization of remote sensing in Syria.

## TABLE OF CONTENTS

Page
DECLARATION ..... iii
ABSTRACT ..... iv
ACKNOWLEDGMENT ..... v
TABLE OF CONTENTS ..... vi
LIST OF TABLES ..... ix
LIST OF FIGURES ..... xi
LIST OF ABBREVIATIONS ..... xiv
LIST OF PUBLICATIONS ..... XV
CHAPTER 1 INTRODUCTION ..... 1
1.1. Studying Area ..... 1
1.2. Problem Statement ..... 2
1.3. Objective ..... 3
1.4. Scope of The Study ..... 4
CHAPTER 2 LITERATURE REVIEW ..... 5
2.1. Building Extraction ..... 8
2.2. GIS ..... 10
2.3. Forecasting ..... 10
2.4. Best Site Location ..... 10
2.5 Research Gaps ..... 11
CHAPTER 3 MATERIALS AND METHODOLOGY ..... 12
3.1. Introduction ..... 12
3.2. Work Flow ..... 12
3.3. Data Collection ..... 14
3.3.1. Remote Sensing Data ..... 14
3.3.2. Surveys ..... 15
3.3.3. Open Street Map ..... 17
3.3.4. Nielsen Data ..... 18
3.4. Variables ..... 18
3.4.1. Buildings ..... 18
3.4.2. Units Per Residential Buildings Type ..... 19
3.4.3. Family Member ..... 19
3.4.4. Vacant Buildings Percentage ..... 19
3.4.5. Required Time For Products Installation ..... 19
3.5. Software ..... 21
3.5.1. ARC GIS ..... 21
3.5.2. Erdas Imagine ..... 21
3.6. Methodology ..... 22
3.6.1 Data transformation ..... 22
3.6.2. Satellite Images Analyses ..... 22
3.6.2.1. January 2016 Satellite Images ..... 22
3.6.2.2. October 2016 Satellite Images ..... 23
3.6.3. Micro-Geographic Segmentation ..... 24
3.6.4. Population ..... 25
3.7. Discussion ..... 27
3.8. Conclusion ..... 27
CHAPTER 4 RESULT AND DISCUSSION ..... 29
4.1 Introduction ..... 29
4.1.1. January 2016 Population ..... 30
4.1.2. October 2016 Population ..... 32
4.2. Cumulative Population ..... 41
4.3. Milk Market Size ..... 46
4.3.1 January 2016 Milk Market Size ..... 46
4.3.2 October 2016 Milk Market Size ..... 50
4.3.2.1 Daily Milk Market Size Forecasting ..... 50
4.3.2.2. Yearly Milk market Size Forecasting ..... 60
4.3.2.3 Cumulative Yearly Milk Market Size Forecasting ..... 62
4.4 Building Material ..... 65
4.4.1 The Increase in Market Size According To Periods ..... 69
4.4.2 Cumulative Market Size For Switches And Sockets ..... 76
4.5 Determine The best Site Location ..... 82
4.5.1 Land Value ..... 82
4.5.2 Population Factor ..... 83
4.5.3 Land Use ..... 84
4.5.4 Distance To The Road Network ..... 85
4.5.5 Topographic Factors ..... 86
4.5.6 Climate Factors ..... 87
4.5.7 Competitors (Distance From Current Markets) ..... 87
4.5.8 Land Area ..... 89
4.5.9 Select the Best Location For New Retail ..... 90
CHAPTER 5 CONCLUSION AND RECOMMENDATIONS FOR ..... 93 FUTURE WORK
5.1 Conclusion ..... 93
5.2 Recommendations ..... 94
REFERENCES ..... 95

## LIST OF TABLES

Page
Table 3.1 The Average Number Of Pieces For Each Building Type ..... 16
Table 3.2 The Duration of The Stage For Building Type ..... 16
Table 3.3 Milk Consumption in KSA And Riyadh City (Nielsen Data) ..... 18
Table 3.4 Total Buildings Scanned in January With its Building Type ..... 22
Table 3.5 Total buildings scanned in October with its building type ..... 24
Table 3.6 Average Number of Units Per Residential Type And ..... 27 Population Per Building Type
Table 4.1 Expected Population In the Study Area According to ..... 30 Building Type
Table $4.2 \quad$ Expected Population at January 2016 ..... 31
Table 4.3 Expected Population According to Buildings Scanned at ..... 33October 2016
Table 4.4 Total Buildings For October 2016 Construction With its ..... 35 Buildings Stage
Table 4.5 Expected Population According to The Construction Stage ..... 36 For October Buildings
Table 4.6 Expected Cumulative Population According to The ..... 42Construction Stage For Total Buildings
Table 4.7 Expected Daily And Yearly Milk Consumption According ..... 47 to January Buildings
Table 4.8 Daily Expected Milk Market Size After Three Months ..... 53 "Litre" For October Buildings
Table 4.9 Daily Expected Consumption According to October ..... 55 Buildings (By Litre)
Table 4.10 Daily Expected Cumulative Consumption According to ..... 56 October Buildings (by Litre)
Table 4.11 Yearly Milk Consumption According to October Buildings ..... 61
Stage
Table 4.12 Yearly Cumulative Consumption According to October ..... 63 Buildings
Table 4.13 Total Construction Requirement For The Switches ..... 66
Table 4.14 Show The Expected Switches Market Demand ..... 67
Table 4.15 The New Buildings Distribution Depending on Construction ..... 68 Type
Table 4.16 The Demand For New Products According to Building ..... 70 Stage And Building Type
Table 4.17 Cumulative Market Size According to Building Stage And ..... 76 Building Type
Table 4.18 The Retails in The Study Area ..... 88

## LIST OF FIGURES

Page
Figure 1.1 The Growth in Riyadh City Since 2005 ..... 1
Figure 1.2 The Studying Area Northern Riyadh City ..... 2
Figure $3.1 \quad$ Work Flow ..... 13
Figure 3.2 Shown Spot 1.5 m Satellite Image ..... 15
Figure 3.3 Road Network in The Study Area ..... 17
Figure $3.4 \quad$ Buildings Type Segmentation For January 2016 ..... 23
Buildings
Figure $3.5 \quad$ Buildings Type Segmentation For October 2016 ..... 24 Buildings
Figure 3.6 Micro-Geographic Area Segmentations ..... 25
Figure $4.1 \quad$ People Segmentation at January 2016 ..... 30
Figure $4.2 \quad$ Expected Population at January 2016 ..... 32
Figure $4.3 \quad$ Expected Population at October 2016 ..... 34
Figure 4.4 Expected Population According to The ..... 37 Construction Stage For October Buildings
Figure 4.5 The Expected Increase in Population After 3 ..... 37 Months Per Districts For October 2016 Buildings
Figure 4.6 The Expected Increase in Population After 6 ..... 38 months Per Districts For October 2016 Buildings
Figure 4.7 The Expected Increase in Population After One ..... 39
Year Per Districts For October 2016 Buildings
Figure 4.8 The Expected Increase in Population After Two ..... 40 Years Per Districts For October 2016 Buildings
Figure 4.9 Expected Cumulative Population According to The ..... 43 Construction Stage For Total Buildings
Figure $4.10 \quad$ The Expected Cumulative Population One Years ..... 44
Later at Districts Level

Figure 4.11 The Expected Cumulative Population Two Years Later

Figure 4.12 Expected Daily Milk Consumption According to 48 January Buildings
Figure 4.13 Expected Yearly Milk Consumption According to 49 January Buildings

Figure 4.14 Daily Expected Milk Market Size (Three Month 51 Later)

Figure 4.15 Daily Cumulative Expected Milk Market Size (Three Month Later)
Figure 4.16 Difference Between Market Size And Cumulative 54 Market Size After Three Months (For October Buildings)

Figure 4.17 Contribution of Cumulative Daily Expected Milk 57 Market Size Over The Year

Figure 4.18 Daily Milk Market Size, Cumulative (Six Months 58 Later)
Figure 4.19 Daily Expected Milk Market Size (Six Months 59 Later)
Figure $4.20 \quad$ Yearly Expected Milk Market Size (Three Month 62 Later)

Figure 4.21 Yearly Cumulative Expected Milk Market Size (Three Month Later)
Figure 4.22 Expected Market Size For Switches Market Size 69 For The New Buildings
Figure 4.23 The Increase in Market Size 71
Figure $4.24 \quad$ Increase in Switches And Sockets Three Months 72 later

Figure 4.25 Increase in Switches And Sockets Six Months Later 73
Figure 4.26 Increase in Switches And Sockets One Year Later 74
Figure 4.27 Increase in Switches And Sockets Two Years Later 75
Figure 4.28 The Increase in Market Size 77
Figure 4.29 Expected Cumulative Switches Market Size (Three 78 Months Later)
Figure $4.30 \quad$ Expected Cumulative Switches Market Size (Six 79 Month Later)
Figure 4.31 Cumulative Switches Market Size (One Year Later) 80
Figure 4.32 Cumulative Switches Market Size (Two Years 81 Later)
Figure 4.33 Land Price Classification 83
Figure 4.34 Population Density 84
Figure 4.35 Land Use 85
Figure 4.36 Main Roads 86
Figure 4.37 The Slop in The Study Area 87
Figure 4.38 Retails (Competitors) Location in The Study Area 89
Figure 4.39 Empty Land in The Study Area 90
Figure 4.40 Raster Calculation 91
Figure 4.41 Best Site Locations 92

## LIST OF ABBREVIATIONS

| FMCG | Fast-moving consumer goods |
| :--- | :--- |
| POP VI | Population in a villa |
| POPAPA | Population in Apartment |
| V | The average units per villa |
| A | The average units per Apartment |
| P | The average units per palace |
| AVE | The average population per units |
| E | Non-inhabited buildings |
| POP PA | Population in Palace |
| CUM (n) | Cumulative population after n periods |
| Ex POP (n) | expected population growth after (n) period |
| POP (current) | Current population |
| DM | Daily Milk needs |
| AVE M | person's daily Milk consumption |
| CUM DM (n) | Cumulative milk market size a specific periods |
| DM (n-t) | previous period Market size |
| t | previous period |

## LIST OF PUBLICATIONS

1- Geomarketing using Remote Sensing A Study on Marketing and Planning Development Strategy at Northern Riyadh, iGRAD2018, Scopus journals.

## CHAPTER 1

## INTRODUCTION

### 1.1 Study Area

Riyadh city the capital of the Saudi Kingdom located in the middle of the country at the intersection of the transportation network, this city was significantly developed by 25 \% between 2005 and 2016 Figure 1.1. It is extended towards the Northern direction more than to the other directions due to natural obstacles or industrial cities. The study area (i.e., $240 \mathrm{sq} . \mathrm{km}$ ) located in the northern part of the town in which recently an essential urban growth and human activities based. Figure 1.1 shows the boundaries of the city in 2005 with yellow color while the blue boundaries at 2018, it shows how the town stretched and expanded. In general, Riyadh is about 30\% of the Kingdom's population and is increasing horizontally because the people here prefer to live in separate buildings and have privacy.


Figure 1.1 The Growth in Riyadh City Since 2005

According to the 2013 census (Authority of statistics) on Riyadh City:

- The population was 5.7 million
- The population growth rate of $4 \%$ per year
- Number of housing 960000 units
- Number Schools was 3060
- Number of hospitals bed was 14137 beds
- The age group for Saudis under 14 years was $30 \%$
- The age group for Saudis between 15-60 years was $60 \%$
- The average number of the family was 5.97.
- Riyadh city area approximately 1400 sq. km


Figure 1.2. The Study Area at Northern Riyadh City

### 1.2 Problem Statement

Geographical Information Systems (GIS) play a major role in different application such as urban area planning and environment studies but need to improve its usage in business like marketing, sales demand and market research. Using GIS for the spatial representation of marketing problems helps the decision-makers arrive at conclusions better related to marketing planning issues. The improve of use the spatial information (Shawky, 2016) associated with high-resolution satellite images in planning and
decision-making will have a positive impact in the long term by being able to understand the temporary and long-term changes occurring in the global market especially in the midst of intense competition. The analysis of high-resolution satellite images for detection the urban growth associated with identify buildings type (villas, apartments, mall, etc.) and its construction level will provide a clear vision for the current market size and create more accurate expectations, this information will not be available through other methods to helps the decision-makers for better analyze situation to generate a forecast based on more accurate data, and contribute to the selection of the best location for the establishment of services.

In general, using Geomarketing in Arab countries needs to improve to better understand the market and its needs (European Commission, 2017). Therefore, this research presents the use of satellite images in Geomarketing in:
a) Improving market size reading
b) Supporting thinking at micro-geographic area
c) Supporting decision-making
d) Improving marketing strategy

This study was carried out in the north of Riyadh City as shown in Figure 1.2. In the Kingdom of Saudi Arabia, in an area that has a fast-growing and forms the vital area of the city, with the aim to improve plans and determining market size using spatial data. Many programs and projects may falter in the absence of accurate data that reflects the current reality that enables the governmental authorities and companies an Evaluate the needs of the community, determine market size and forecast for the future. Therefore, government agencies and companies are looking forward to having the data to help them to address the challenges and create appropriate plans to meet the community requirements. However, current data and methodologies can't solve the demands and needs of each geographical area. The proposed method will help decision-makers in building and developing plans more effectively and include the following.

### 1.3 Objective

In this research, the building growth and its effect on building materials market size will be studied. Growth in buildings will be used to study the population at districts level to expect growth and the cumulative number of community in a specific time; this
expectation also will help to calculate milk market size at the micro-geographic area, the population estimation also will use to set up a new methodology to determine the best site location in the study area.

The objective of this study is as follow:

1. To determine the effect of adding geospatial information to marketing and business.
2. To determine market size and develop forecasting methodology by using remote sensing
3. To improve the best site location methodology by Geomarketing.

### 1.4 Scope of Study

The overarching purpose of this research to examine using the High-resolution satellite images in Geomarketing and improve marketing strategy with a combination of other factors to have better results by distinguishing the buildings type (existing and new) and support decision-maker in their decisions. We will study the impact of buildings growth and the units per buildings to the market size for the studying products. This growth also affects the population increase which will study the community who live or can live in this area by having information about the family member. The population density will determine according to this methodology.

## CHAPTER 2

## LITERATURE REVIEW

The utilization of remote sensing and GIS give various advantage for the urban industry due to the power of GIS application in storing, querying and analyzing data on the ground which including the information required by the urban industry. This is the fact of GIS implementation can carry out computer and spatial data analysis methods developed in varying related disciplines which including surveying, geography, land management, and Urban planning, remote sensing (Mzainora, 2016). With a GIS-based details system, many data can be displayed correctly on maps to support the users who interested in urban area planning, business or such as retailers, buyers, sellers, decision-makers about searching for the best site location or be estimating market values.

Geomarketing is a field that uses geoformation in the process of marketing activities and business planning. A lot of fields can use geo-targeting such is marketing such as 4 ps marketing or geo-targeting. Geomarketing analysis information uses huge data such as location residential areas, marketing, and business, analyzes demographic information like gender, age, genre, their economic situations, and lifestyle. This data supports the users to improve their planning in order to achieve marketing aims (Suhaibah, 2016). One of the important applications of Geomarketing is market segmentation. It divides data into several ranges based on its geographic factors.

Geomarketing and geolocation theories may effectively and efficiently support the business establishment in best locations, allowing any businessman to implement his business and achieve profits. The benefits of the Geomarketing as a revolutionary methodology (Casalino, 2018):

- Establish a link or relation among geographic business elements and their relevant spatial data.
- Ability to recognize and analyze the composition of business phenomena through the use of Geomarketing
A lot of endeavors are constantly made to better recognize the consumer and customer distributions, In order to develop a more accurate way of marketing within these areas. However, there are many barriers to marketing processes such as data restrictions, limited access, administrative boundaries, lack of market information and market size (Nicholas,
2014). (Amanda, 2014) studied how Geomarketing can be used to discover and acquire marketing skills and are effective in solving real problems. (Veland, 2018) built on data collected by researchers through questionnaires and government data, and he used the classification of companies how sites and spatial data in cities can affect decision-makers and the results are presented on appropriate maps. (Somnath, 2018) discussed the usage of Web-Based Geographical Information System (GIS), for business; especially through the Internet, with a plan to develop electronic-promotion and strategic marketing by implementing Web GIS tools for Geo business.

The subject of electronic-business is present study permeates the utilization of GIS to spread the greatest level of data to the clients. Furthermore, it represents the growing combination of GIS in target marketing, business management, and strategic decisionmaking processes. (Hosseini, 2016) used a model to select the best site for suitable places and took into account the following aspects: maximizing the well-being of the DP, minimizing the negative impacts on neighborhood life, minimizing the public expenditures on TH , minimizing the negative environmental impacts, and maximizing the well-being of the people involved in the TH process.

During the site selection process, this model help decision-makers in observing and comparing the indicator values of all alternatives. At times, decision-makers select other option that has weaknesses caused by limitations; based on the mentioned feature of this model, decision-makers can select the weak parts of a determined site and then control these weaknesses with suitable work. Spread around sites located in different microGeographic areas have the best social indicator value.

Zhang (2016) used a new method of using satellite imagery to estimate location adjustment factors where they do not exist. This method for estimating location modulation index was evaluated against a determined cost index database and results show that satellite imagery can be used to effectively select the best site adjustment elements. This work participates in the information by bringing an accurate way to assessing location modification index which can enhance cost estimation for select new projects.
A big-data approach to extract buildings from a DEM created from aerial laser scanning imagery. This process includes two steps. The first one is a Map Reduce process where neighboring points in a digital surface model are mapped into cubes. The second one uses anon-Map Reduce algorithm first to remove trees and other obstructions and then to
extract adjacent cubes. Present to this process, all of the adjacent elements belong to the same objectives which this belong to one or more adjacent buildings (Aljumaily, 2015). Housing market forecasts can provide insight into the essential sustainability of housing and construction. The home sales index (HSI) reflects one of the most important indexes for forecasting housing market trends in the real estate function, they have tried to develop relevant forecasting models for the HSI. There are many demands for effective HSI forecasting by identifying the various social factors influencing the HSI. He suggested a new way for forecasting model development with the provision of fundamental data and the pros and cons of each model to which the multiple regression analysis (MRA) and the artificial neural network (ANN) were applied. He compared it with ARIMA. Forecast HSI forecast data using ARIMA are more accurate than those of MRA and ANN (Han, 2018).

Li , (2017) suggested a new method for road feature extraction from aerial imagery data that consisted of multi-steps:

- Removal of elevated objects.
- Removal of shadows and vegetation.
- Extract road features from the fused data;
- Final extraction of road surfaces and centerlines.

This new method is suggested for data fusion of satellite images and to extract road features by utilizing color components, such as saturation. Housing growth provides economic expansion with Geolocation. In addition to their own experiences, employees should consider several factors during the selection of residential or commercial investments (Aktas, 2017). These factors may be evaluated using multicriteria decisionmaking analyses. The weights of many evaluation indexes were determined using the fuzzy analytic hierarchy process (FAHP) and fuzzy entropy (FENTROPY) methods (Bostancı, 2017).

Darani (2018), studied the multicriteria decision-making (MCDM) method by coupling the analyses of the hierarchy process (AHP) and the technique for order of preference by similarity to ideal solution (TOPSIS) methods under a fuzzy environment to select the best site for new parking. Boostani (2018), told that it is important to determine the factors that affect the best location of accommodation or new activity sites. He used his expert and the Fuzzy Delphi method, three main criteria including infrastructural factors, access,
and sustainability. To specify the weight of the identified factors, the fuzzy DEMATELbased analytic network process (FDANP) method is used.

Recently Wu (2019), studied a shopper preference-based competitive location model (SPCLM) to find the best solutions to the location problem for the nearest retail. He used multi-criteria for this model construction: size, diversity of the tenant inside the shopping center, retail agglomeration near the shopping center, distance to metro stations, and distance between consumers and shopping center.

Geomarketing is a cornerstone of market segmentation. The segmentation gathers the data into groups based on their geographic factors. He suggested a method to improve the process to gather the data using a gather algorithm. In this study, the geomarketing is active in urban areas needs a three-dimensional (3D) way and this a constraint in GIS. To avoid this issue, He suggested a combination of market segmentation based on geographic factors and gathers algorithm for 3D geomarketing data management. He minimized the overlap areas during the segmentation (Suhaibah, 2016). Garud (2017), studied the food supply chain, foodshed, food miles for identified diets for PMR, this a basic scenario for FMCG consumption style and produce is mapped. Fukun (2019), present a new method called a DBA detection uses hierarchical structural constraints in remote sensing images. This way was managed in two main steps. (1) During keypoint generation (2) To match the screened key points.

### 2.1 Building Extraction

Jin used object-oriented segmentation and classification methodology by using texture, characteristics, feature, shape feature to extract the building in remote sensing images (Jin, 2012). Jian used object-oriented classification to have individual object-classes, and fuzzy determined rules system to be created for earthquake collapsed building extraction (Jian, 2013). Huang and lot of researcher used many indexes to extracted the building as result of the different scale of different type building and merged the results as a whole, Suggested novel morphological building index for automatic building extraction by using high-resolution images with optimization and improvements (Huang and Zhang, 2011), Jian yang used another method to extract the building information by combination the contoured transform and PCNN segmentation algorithm to obtain the multiscale and
multidirectional characteristic of the building. Other efforts have been made to extract buildings in automated ways. Applied the Perceptual grouping way to obtain the building. Lin and Nevatia detected the edges of the image to find the parallel lines from the edges. After that, they searched for the parallel lines to find a rectangle which meets the geometric and projective constraints as the building object, (Lin and Nevatia, 1998). A study proposed a building extraction method by combining edge preserving, smoothing bilateral filter with the line segment detector. Wang used a filtration for smooth the original satellite image, detected the line segments and grouped the lines to construct a rectangular building, so these methods are suitable for extracting simple square buildings, (Wang, 2015).

The proposed framework consists of two main stages: generating foreground and background area, minimizing the energy function, (Yihua and Yujie, 2016). The classification of high-resolution images was studied by a lot of researchers More recently, finer texture and more certain boundaries of the building can be obtained from HSR imagery and applied to build extraction, (Ok, 2013). So it is still difficult to discern building types from high-resolution images by computers programs because it is hard to find appropriate segmenting scales to completely capture even an individual building from complex patterns of combining pixels,( Zhang 2015), Elevation data and building contours have also applied in the classification of the building type. Airborne light detection and ranging (LiDAR) is particularly useful to collect the elevation data for the expression of building structural characteristics, (Kraus and Pfeifer, 1998).

In EMRS-SBP scheme, (Extended multiresolution segmentation) EMRS serves to guide the design of descriptor and SBP to generate a more natural classification is used for the classification of urban building type, (Junfei and Jinhua, 2017). Change detection for remote sensing images has been applied to many fields, such as urban area and so on. It has drawn great attention in recent years since it is an efficient way to find the differences between images with different dates, typical methods often firstly consider spectral features and then use threshold segmentation to detect the change. However, the thresholds are usually determined by experience, which brings about a series of problems, (Wen, Huang, 2016), Huo, J. Cheng introduced two different multiscale fusion strategies conducted the experiments on QuickBird high-resolution remote sensing images and completed the object-oriented change detection. But the robustness of this algorithm is not good enough, (Huo, Cheng, 2008). Qingle Guo and others studied a new method which is a multiscale segmentation and decision-level fusion, Multiscale segmentation
refers to segmenting the same image in different scales, and the segmentation results can reveal the multiscale land-cover features and spatial structure information of images, (Qingle Guo, Junping Zhang, 2017).

### 2.2 GIS

GIS used to support the decision-maker in many fields like working to decrease food waste, (San Martin1 2017), decision making is based on numerous data concerning the problem. It has been estimated that $80 \%$ of data used by managers and decision-makers are geographical (spatial) in nature, (Worral 1991), GIS using in the most critical and farreaching decisions faced by operations managers are deciding where to locate new industrial facilities. This is a strategic decision involving irreversible allocation of the firm's capital, (Bhatnagar \& Sohal, 2005), GIS is tools to solve the spatial decision problem typically includes a large set of feasible alternatives, (Rikalovic and Aleksandar, 2014). Amparo BAVIERA-PUIG told Geomarketing viewpoint, and the model shows that sociodemographic characteristics of the supermarket's trade area affect firms' location strategies, (Amparo Baviera-Puig, Juan Buitrago, 2016.

### 2.3 Forecasting

Arvydas and Simon, studied statistical estimates to confirm the claim that goodness-offit does not imply good forecasting performance and that increased model complexity does not necessarily yield greater forecasting accuracy. This, therefore, calls for the adoption of a Laconic or "Keep It Sensibly Simple "modeling approach. In other words, the recommendation is for analysts to make forecasts user-friendly so that they are easy to use and understand. Researchers should put simplicity at the core of them.

### 2.4 Best Site Location

The geographic information system can combine spatial data, especially competitor sites, with solutions designed to help open a new facility to enhance the competitiveness of the new entity. Competitive sites consist of several criteria and are primarily related to consumers and service providers. Here, customer behavior must be considered so that we
can meet their demands so that they can have all their requirements in one place (Rafael, 2012). The opening of a new branch for any company or service is one of the critical works that requires careful study and depends on scientific methodologies in which the review of the situation in various aspects to mitigate any risks that may result from the expansion of activity. The population density that will be used to determine the new location has been studied based on the data obtained from the Statistics Authority (Norat, 2013).

Xuefeng, (2010) used ArcGIS Geoprocessing to develop spatial analysis modeling and decision making by using implementation technology, modeling and multiple options using ARC GIS, using the classification and reclassification method, and calculating the required spaces to set up the new branch within the conditions and result. "Kazem" presented a study to select the best location for parking as a result of the traffic jam of Tehran city and performed the study of land use and peak areas of traffic and restrictions that prevent the establishment of this position such as cultural restrictions and historical monuments, all those features have been added all to the maps to determine the outputs that correspond with Determine the best location for parking (Kazem, 2015).

### 2.5. Research gaps:

Through the above, it becomes clear to us that there is a lack of using remote sensing data in the calculation of market size especially at the micro-Geographic area and the development of the using Geomarketing that would enhance the use of spatial and attribute data to determine the best locations for establishing service sites or centers, depending on the development of the Geomarketing concept, and that what has been presented from other researchers has not touched on the issue of using remote sensing data in developing market studies and sales planning with a reflection on the strategic decisions taken Companies in terms of an accurate reading of the market and building more accurate future forecasts.

Below we will be explaining how we can develop these concepts and uses to serve the industry and work to bridge an important gap in this area.

## CHPATER 3

## MATERIALS AND METHODOLOGY

### 3.1 Introduction

In this section, we will study the data used, the sources of these data, and how it will be processed and used in a manner appropriate to this study, what the variables which used and how can it affect the study, also explain the software that used. What are the satellite data and spatial data, their dates, how can estimate the population by using remote sensing, and how we can expect market size for a given product?

### 3.2 Workflow

This work started after selecting the studying area which the methodology will be applying to study it, we have to choose the period that we will investigate it then the suitable satellite images which be optical which will be used to extract existing buildings from the older image then determine the growth by extract the new buildings from the latest image, here we should have information about the family member per building unit or FMCG daily consumption (Milk) and buildings needs from the products for each building type to determine the market size and improve the forecasting. The following flowchart explains the process.


Figure 3.1 Workflow

### 3.3 Data Collection

### 3.3.1. Remote Sensing Data

The primary data used to conduct this study is the remote sensing data for determining to the urban growth that will require the use of two satellite images in two different dates for the studying area, the specification for it was:

- Satellite: Optical Spot 6
$\square$ Resolution: 1.5 m , color
The first image: its date was January 2016, it is used to create a feature class containing all the existing buildings at that date and classified it according to the building type and construction level.

The second image: as in Figure 3.1, its date was October 2016, it is used to create a feature class containing the new buildings which constructed after the date of the first image in order to know the exact number of new buildings that were created in the study area and which classified according to the building type and construction level, The construction level are:
a) Basements
b) Under construction
c) Construction completed

Here we have the all the building in the studying area, existing and new which classified depend on a field observation (explained in the methodology)

Extraction method: extract the buildings from satellite images were manually Firstly, digitize the first image and extracted the existing buildings, the growth was determined by digitize the second image to identify the difference between the two images, the difference was the growth (October 2016).


Figure 3.2 Shown Spot 1.5 m Satellite Image

### 3.3.2. Surveys

Since one of the objectives of this study aims to increase the accuracy of the identify market size by remote sensing, one of the building materials products was studied to estimate the expected demand of these products (switches and sockets) during a specific
period and determine the cumulative demand. To achieve this goal, a field survey of a sample of electricians and contractors ( 50 persons) was conducted in the study area to determine the average need for each building type. (Villa, apartment, palace, mall, mosque). For these products during the finishing phase.

The survey included the following questions:

1. The average number of pieces (switches and sockets) needed by each type of buildings, Table 3.1.
2. The expected time for these products to installing to those buildings type, table 3.2.

Table 3.1. The Average Number of Pieces For Each Building Type

| Buildings type | Switches Average (pieces) |
| :---: | :---: |
| villa | 193 |
| Apartment | 771 |
| Esteraha | 48 |
| Mosque | 104 |
| tower | 4188 |
| Gas station | 93 |
| Industry | 300 |
| School | 425 |
| Mall | 227 |
| showroom | 351 |
| Palace | 2000 |
| warehouse | 50 |
| Governmental | 1613 |

Table 3.2. The Duration of The Stage for Building Type

|  | Duration of the stage( month) |  |  |
| :---: | :---: | :---: | :---: |
| Building type | Basement | Under <br> construction | construction <br> completed |
| Apartment | 6 | 12 | 24 |
| Esteraha | 3 | 6 | 12 |
| Gas Station | 3 | 6 | 12 |
| Governmental | 6 | 12 | 24 |
| Mall | 6 | 12 | 24 |
| Mosque | 6 | 12 | 24 |
| Palace | 6 | 12 | 24 |
| School | 6 | 12 | 24 |
| Showroom | 3 | 6 | 12 |
| Tower | 6 | 12 | 24 |
| villa | 3 | 6 | 12 |
| warehouse | 3 | 6 | 12 |

### 3.3.3. Open Street Map

The road network in the study area was extracted using the open Street map where the road network in the area was extracted, checked, completed and classified into primary and secondary roads to help classify the buildings. The apartments are often built on the main roads, while the villas are inside districts alongside the secondary roads, the roads also help to demarcate the boundaries of the micro-geographic area, and the roads also will be taken into consideration for choosing the best location, Figure 3.2.


Figure 3.3 Road Network in The Study Area

### 3.3.4. Nielsen Data

Nielsen data used to estimate Milk market size as FMCG products by using Remote sensing data at the micro-geographic area.
a) The average per capita consumption of milk per day in Riyadh was 0.11 liters while in Saudi Arabia it is 0.09 liters per day in 2016 (Nielsen company)
b) Market size includes white/plain and flavored milk. Excludes powder milk \& fermented milk drinks (Laban or yogurt)
c) This study only for residential sectors (Villas, Apartments, Palaces)

The following table shows the total consumption in Saudi Arabia and Riyadh City

Table 3.3. Milk Consumption in KSA and Riyadh City (Nielsen data)

| KSA -MILK <br> CONSUMPTION <br> (Million Litre/ year) | 2012 | 2013 | 2014 | 2015 | 2016 | 2017 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Fresh Milk | 506.8 | 458.1 | 483.5 | 518.7 | 550.3 | 579.7 |
| Growth (\%) | $9.40 \%$ | $-9.60 \%$ | $5.50 \%$ | $7.30 \%$ | $6.10 \%$ | $5.40 \%$ |
| Total KSA | 807.4 | 888.6 | 945.1 | 997.7 | 1034 | 1077 |
| Growth (\%) | $9.30 \%$ | $10.10 \%$ | $6.40 \%$ | $5.60 \%$ | $3.70 \%$ | $4.10 \%$ |
| RIYADH ONLY | 283 | 311 | 331 | 349 | 362 | 377 |

### 3.4. Variables

### 3.4.1. Buildings

Buildings Growth means Growth in market demand for a lot of products like buildings materials. Also, the population will live in the new buildings which food, water, and Milk will necessary for them, this variable will have it by Satellite images analysis.

### 3.4.2 Number of Residential Units per Residential Buildings Type

This variable changes according to the city and districts. It means the number of units that each residential building type contain, this factor info obtained from the authority of statistics.

### 3.4.3 Family Member

It's a critical variable to estimate the population in the study area which helps to calculate the expected FMCG demand like Milk market size. The family member obtained from the authority of statistics.

### 3.4.4 Vacant Buildings Percentage

According to the authority of statics, there were $12 \%$ of vacant residential units in Riyadh in 2016 which should consider it when calculating the population in the studying area. This factor affects the expected community so when this factor increase that is mean the population will decrease.

### 3.4.5 Required Time for Products Installation

Required time for product installation variable is necessary to estimate the required time that each building type needs to install the products or required time that the residential buildings ready for living. This variable has been compiled from a survey of the experience of contractors and electricians as in Table 3.2. When this variable was increased the required time for products installation will increase and also the require time for living.

The geographical information system (GIS) used in this study in order to identify the existing and new buildings which determined depends on the experience and the interpretation of the images by manual digitizing. A point placed in front of each building and its classification according to the building type and the construction level, and this done for the images dated at January 2016 and October 2016. The types of buildings determined according to the following:
a) The majority of the buildings that are next to the main roads are apartments, which are the features of the building systems in Saudi Arabia, especially in the main cities.
b) Villas located inside the districts next to secondary roads.
c) The Esteraha is a building consisting of a few rooms and one floor, it has a small garden with a pool and usually rented during the holidays.
d) Towers: It is a high building with more than ten floors like hotels that typically located next to some of the main roads such as King Abdul Aziz Road or the Northern Ring.
e) Mosques: Its shape is distinctive, especially the towards of Mihrab to the Qibla
f) Gas station.
g) Malls: which occupy large areas.
h) Showrooms: usually characterized by the large building and its positions near to the main roads.
i) Schools and government buildings: characterized by the occupied area and by the method of construction, consisting of four blocks in general and especially for schools.
j) Palaces and large villas: characterized by the large occupied area located inside the districts green areas within it.
All types of buildings identified by the satellite image and classified according to the construction level, table (3.2) and the construction level that includes:
a) Basement
b) Under construction
c) Construction is complete

For example, to explain this methodology to determine the market size, we can apply it to calculate the market needs for two kinds of products (FMCG like Milk and building materials like switches and sockets.

In the following will process and study:
a) Estimating the population in the study area at the Micro-Geographic area levels to using it to estimate the demand for a wide range of services and products.
b) Estimate the Milk market size and cumulative milk market size at the MicroGeographic area levels.
c) Estimate the material of products (switches and sockets) market size and cumulative market size the Micro-Geographic area levels.
d) Carry out analysis to select the best site location for new services depend on the population density which estimated by using satellite images.

### 3.5 Softwares

### 3.5.1 ARC GIS 10.1

ArcGIS (Geographical information system) is a computer-based system that collects, maintains, stores, analyses, outputs and distributes data and spatial information? These systems collect, input, process, analyse, display and produce spatial and descriptive information for specific objectives, and assist planning and decision-making concerning agriculture, urban planning, housing expansion, as well as reading the infrastructure of any city through the creation of called Layers. This system introduces geographic information (maps, aerial photographs, and satellite images), descriptors (names, tables), processing them (revising the error), storing them, retrieving them, analysing them, and analysing them on a computer screen or on paper in the form of maps., Reports, and graphs.

GIS helps to answer many of the specific questions of specificity (e.g., what is the agricultural pattern, what types of suitable crops to cultivate in the farm unit), measurements, location (What is the relationship between population distribution and Milk or water market size) and hydrological scenarios (what happens if the water used for irrigation).

### 3.5.2 ERDAS Imagine

ERDAS Imagine (Earth Resource Development Assessment System) is a digital image processing software used for analysis and study the satellite imagery, By using it for extraction of Digital Number values of the pixels of the image, and Import-export raster's and vectors satellite image, working with various bands of satellite data, to perform detailed analysis of multiple objects and information using the pattern recognition technique, Land use land cover analysis. Everything you do on ERDAS Imagine has a unique concept of Visual Interpretation of satellite imagery.

### 3.6 Methodology

3.6.1 Data transformation: Determine the volume of urban growth in the study area in general by means of remote sensing and linking it with the average number of family members per housing unit and dividing it according to the Micro-Geographic areas will give the ability to identify the population in each district, thus calculate their food needs, also determine the buildings growth in will necessarily lead To know the expected materials buildings market size which is a result of the use of satellite imagery data.

### 3.6.2 Satellite Images Analyses

### 3.6.2.1 January 2016 Satellite Images

The existing construction opportunities in the study area have been extracted to determine the number of existing buildings and classify it according to the building type.

The buildings under construction were not considered as part of the existing buildings on that date and were therefore considered part of the subsequent buildings.

A description of the existing building included in its attribute table which included the building type and construction level (here the construction stage is considered as completed), Table 3.4 shows the total existing building that detected in January 2016 by Satellite images analysis.

Table 3.4. Total Buildings Scanned in January with its Building Type

| Building type | Number of existing buildings | percentage $\%$ |
| :---: | :---: | :---: |
| villa | 44603 | 88.77 |
| Apartment | 3345 | 6.66 |
| Esteraha | 914 | 1.82 |
| Showroom | 526 | 1.05 |
| Palace | 277 | 0.55 |
| Mosque | 237 | 0.47 |
| School | 93 | 0.19 |
| warehouse | 88 | 0.18 |
| Tower | 85 | 0.17 |
| Gas Station | 43 | 0.09 |
| Governmental | 20 | 0.04 |
| Compound | 6 | 0.01 |
| Industry | 3 | 0.01 |
| Mall | 3 | 0.01 |
| Grand Total | 50243 | $100 \%$ |



Figure 3.4 Buildings Type Segmentation For January 2016 Buildings.

Table 3.4 and figure 3.4 show that the villas reflect $89 \%$ of the existing buildings while the Apartments reflect only $8 \%$.

### 3.6.2.2 October 2016 Satellite Images

The analysis of the following satellite image dating to October 2016 will determine the growth and trends in the study area. The new buildings classified according to the buildings type and construction level, Table 3.5 .


Figure 3.5 Buildings Type Segmentation for October 2016 Buildings

Table 3.5. Total Buildings Scanned in October with its Building Type

| Building type | Number of new <br> buildings | percentage \% |
| :---: | :---: | :---: |
| villa | 3626 | 84.96 |
| Apartment | 429 | 10.05 |
| Showroom | 94 | 2.20 |
| Esteraha | 57 | 1.34 |
| Palace | 18 | 0.42 |
| Mosque | 15 | 0.35 |
| School | 9 | 0.21 |
| Gas Station | 8 | 0.19 |
| Tower | 5 | 0.12 |
| Governmental | 3 | 0.07 |
| warehouse | 2 | 0.05 |
| Compound | 1 | 0.02 |
| Mall | 1 | 0.02 |
| Industry | 0 | 0.00 |
| Grand Total | 4268 | $100 \%$ |

Also, the majority of the new buildings consist of villas by $85 \%$, apartments $10 \%$ while the other buildings type include the percentage of the other building types.

### 3.6.3 Micro-Geographic Segmentation

The study area was divided into Small districts (micro-Geographic area) to identifying the growth at this level to understand the requirements for those areas by determining the business volume and improve the planning. The main roads were selected as the borders for those micro-Geographic areas and these areas were labeled with unique numbers starting from 1 to 44 , the areas for those districts vary from 2.3 Sq. to 7.7 Sq. , only one district its area 23.8 sq. Which contain a large university, districts 25 .


Figure 3.6 Micro-Geographic Area Segmentations

### 3.6.4 Population

Determine the building type and construction level will be valuable for urban area planning and improve business, in other meaning, by having information about the total Buildings with its geospatial information which detected by remote sensing, also we have the average family member per residential unit, this will give us a new tool by identifying the population at micro-geographic areas level, so the buildings will give us the reading and indicator for the population. By using the geospatial data, we will be able to have more accurate information about the community in the studying area and market needs at the districts level. Population information will support decision-makers in many sectors (i.e. governmental, business, etc.) by determining where the highest population density
areas are to study their requirements in the best way to have the right decision depending on the new data.

The total Buildings in January 2016 was 50243 buildings, the residential buildings were 48225 ( 95.98 \% from total), While total new buildings in October 2016 was 4268 buildings, the residential buildings in October were 4037 Buildings ( $95.43 \%$ from total). According to the 2013 census (Authority of statistics) In KSA:
a) The average units per villa are 3.5 unit. (V)
b) The average units per Apartment are 15.5 units. (A)
c) The average units per palace are 1 units. (P)
d) The average population per unit is 5.97 person per unit. (AVE)
e) Percentage of non-inhabited buildings in Riyadh is $12 \%$ ( $88 \%$ inhabited).(E).

Authority of statistics: depend on 2013 census:
a- Population in a villa: In general, the ground floor in villas is occupied by a Saudi family which have a driver and housekeeper at $50 \%$.

$$
\begin{aligned}
& \text { POP VI }=(\mathrm{V} * \text { AVE })+[(\text { driver +housekeeper }) * 50 \%] * \mathrm{E} \\
& =(3.5 * 5.97)+[(\text { driver }+ \text { housekeeper }) * 50 \%] * 88 \%=19.26 \text { person }
\end{aligned}
$$

b- Population in Apartment:38 \% of Apartments population is Saudi ( $25 \%$ of this Saudi segments has a housekeeper)

$$
\begin{aligned}
& \text { POPAPA }=\{(\mathrm{A} * \mathrm{AVE})+[(\mathrm{A} * \mathrm{AVE}) * 38 \%] * 25 \%\} * \mathrm{E} \\
& =\{(15.5 * 5.97)+[(15.5 * 5.98) * 38 \%] * 25 \%\} * 88 \%=(92.69+8.790825) * 88 \% \\
& =89.16 \text { person }
\end{aligned}
$$

c- Population in Palace

$$
\begin{aligned}
& \text { POP PA }=(\mathrm{P} * \text { AVE })+(\text { Driver }+ \text { housekeeper }+ \text { cooker }+ \text { gardener }) * \text { E } 3.3 \\
& =(1 * 5.97)+(\text { Driver }+ \text { housekeeper }+ \text { cooker }+ \text { gardener }) * 88 \%=(5.98+ \\
& 4) * 88 \%=8.77 \text { person. }
\end{aligned}
$$

Table 3.6 Average Number of Units per Residential Type and Population per Building Type

| Building type | Average No. units | Average population per <br> building type |
| :---: | :---: | :---: |
| Apartment | 15.5 | 89.16 |
| Villa | 3.5 | 19.26 |
| Palace | 1 | 8.77 |

After digitizing the existing and new buildings, this gives a unique identity by GIS tools, this identity contains Micro-Geographic code and coordinate system. The other information entered manually during the digitizing like building type and stage.

### 3.7 Discussion

What we have now, while digitizing the building will have the required information:
a) Districts code: to know how much building in each district.
b) Coordinate system: to create geospatial data
c) Building type: to Determine the number and type of buildings in each district, if we know the average population in each building type then will determine the population in each district and the study area, where we can expect the milk market size as FMCG study product, or the required products for each building type then expect the potential market size in district and area.
d) Building stage: This information was necessary to determine the required time for each building type to install the products or ready for living. In this study, I hypothesized that they needed time for products installing is the same time that the building will be prepared to live (the time difference will be ignored). This information is useful for the planning team and logistics operation.

### 3.8 Conclusion

The study area in this section has determined and its general specification and the data used in this research. Also. The methodology that used to extract and classify the construction, the relationship between the data collected with the buildings were
explained to determine Building needs to products or the expected population for each building then at districts levels.

## CHAPTER 4

## RESULT AND DISCUSSION

### 4.1 Introduction

In this chapter will study how we can use this methodology to determine the population at districts level and the expected growth, calculate the cumulative number of community in a specific time; also the population estimation will use to develop a methodology to determine the best site location in the study area. The expected need for an FMCG product (i.e. milk) in the study area will be estimated at the Micro-Geographic level and the expected growth according to time with the cumulative needs for the same periods. Also, determine the growth will help us to calculate the building needs from materials products and applying this methodology to estimate (i.e switches and sockets) market size.

## Population analysis:

If we take into account the Authority of statistics, table 3.5 and 2013 census, we can expect the population in the residential buildings (Villa, Apartment, and palace) then by using the building stage the population growth and cumulative growth will calculate accordingly. In the following, the expected population which their buildings were detected by satellite image in January 2016 will be studying and analyzing, then the predicted population which can live in the buildings which scanned by satellite image in October 2016 (nine-month difference).

In addition to buildings type, the buildings that extracted by October satellite image will classify to:
a) Completed building: it is ready for a living or installs material products.
b) Under construction buildings: it will be prepared for living or installing after a period which differs according to building type.
c) Basement Buildings: it will be ready for living or installing after a period which changes according to building type. But will take time more than the buildings in the previous stage.

### 4.1.1 January 2016 Population

The residential buildings extracted by January 2016 satellite images shown in the following Table 4.1. It explains the number of residential buildings with the average population in those building according to building type, and this estimation depends on the table 3.6.

Table 4.1 Expected Population in the Study Area According to Building Type

| Building type | NO. of <br> building | Average population per <br> building type (person) | Total population <br> (person) |
| :---: | :---: | :---: | :---: |
| Villa | 44603 | 19.2676 | 859392 |
| Apartment | 3345 | 89.16672 | 298262 |
| Palace | 277 | 8.7809 | 2432 |
|  |  | Sum | 1160086 |

The majority of people live in Villas while who lives in Palace approaching zero.
Figure 4.1, now we search where those people live; in which district; where the highest population live. GIS will help to give each building the district code which located in, here will know how much buildings in each district and how much people in each district.


Figure 4.1 People Segmentation at January 2016

This methodology will help the decision-makers to have a good view of the expected demand in the area and change their plan accordingly.

Table 4.2. Expected Population at January 2016

|  | Building type |  |  | expected population per Building type |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| District | Apartment | Palace | Villa | Apartment | Palace | Villa | Grand Total |
| 1 | 176 | 20 | 1928 | 15693 | 176 | 37148 | 53017 |
| 2 | 205 | 10 | 2692 | 18279 | 88 | 51868 | 70235 |
| 3 | 183 | 11 | 2365 | 16318 | 97 | 45568 | 61982 |
| 4 | 148 | 49 | 1618 | 13197 | 430 | 31175 | 44802 |
| 5 | 99 | 5 | 729 | 8828 | 44 | 14046 | 22917 |
| 6 | 20 | 36 | 871 | 1783 | 316 | 16782 | 18882 |
| 7 | 9 | 16 | 783 | 803 | 141 | 15087 | 16030 |
| 8 | 31 | 17 | 608 | 2764 | 149 | 11715 | 14628 |
| 9 | 159 | 1 | 430 | 14178 | 9 | 8285 | 22471 |
| 10 | 576 | 4 | 2030 | 51360 | 35 | 39113 | 90508 |
| 11 | 130 | 15 | 1775 | 11592 | 132 | 34200 | 45923 |
| 12 | 122 | 14 | 2363 | 10878 | 123 | 45529 | 56531 |
| 13 | 37 | 6 | 1732 | 3299 | 53 | 33371 | 36723 |
| 14 | 34 | 0 | 266 | 3032 | 0 | 5125 | 8157 |
| 15 | 79 | 5 | 3192 | 7044 | 44 | 61502 | 68590 |
| 16 | 194 | 6 | 3243 | 17298 | 53 | 62485 | 79836 |
| 17 | 239 | 21 | 1701 | 21311 | 184 | 32774 | 54269 |
| 18 | 8 | 3 | 400 | 713 | 26 | 7707 | 8447 |
| 19 | 9 | 5 | 418 | 803 | 44 | 8054 | 8900 |
| 20 | 135 | 1 | 773 | 12038 | 9 | 14894 | 26940 |
| 21 | 163 |  | 1421 | 14534 | 61 | 27379 | 41975 |
| 22 | 110 | 6 | 3234 | 9808 | 53 | 62311 | 72172 |
| 23 | 87 | 4 | 2439 | 7758 | 35 | 46994 | 54786 |
| 24 | 20 | 0 | 337 | 1783 | 0 | 6493 | 8277 |
| 25 | 100 | 0 | 909 | 8917 | 0 | 17514 | 26431 |
| 26 | 1 | 0 | 85 | 89 | 0 | 1638 | 1727 |
| 27 | 17 | 0 | 365 | 1516 | 0 | 7033 | 8549 |
| 28 | 4 | 0 | 353 | 357 | 0 | 6801 | 7158 |
| 29 | 2 | 0 | 184 | 178 | 0 | 3545 | 3724 |
| 30 | 16 | 4 | 384 | 1427 | 35 | 7399 | 8861 |
| 31 | 41 | 1 | 612 | 3656 | 9 | 11792 | 15456 |
| 32 | 0 | 0 | 37 | 0 | 0 | 713 | 713 |
| 33 | 0 | 0 | 41 | 0 | 0 | 790 | 790 |
| 35 | 2 | 0 | 14 | 178 | 0 | 270 | 448 |
| 36 | 4 | 0 | 211 | 357 | 0 | 4065 | 4422 |
| 37 | 48 |  | 1604 | 4280 | 18 | 30905 | 35203 |
| 38 | 12 | , | 579 | 1070 | 18 | 11156 | 12244 |
| 39 | 2 | 1 | 329 | 178 | 9 | 6339 | 6526 |
| 40 | 58 | 0 | 449 | 5172 | 0 | 8651 | 13823 |
| 41 | 19 |  | 357 | 1694 | 18 | 6879 | 8590 |
| 42 | 17 | 0 | 150 | 1516 | 0 | 2890 | 4406 |
| 43 | 27 | 2 | 406 | 2408 | 18 | 7823 | 10248 |
| 44 | 2 | 1 | 186 | 178 | 9 | 3584 | 3771 |
| Grand Total | 3345 | 277 | 44603 | 298263 | 2432 | 859392 | 1160087 |



Figure 4.2 Expected Population at January 2016.

Map analysis: This figure shows the expected population according to January buildings, districts with dark Red are the highest population density while the dark blue colour reflects the lowest.

### 4.1.2 October 2016 Population

The urban growth in the study area will associate with an increase in the population who will live in the new buildings according to the completion of those constructions which was built after January 2016 and detected in the October 2016 Satellite image. The whole
buildings were 4073 while the residential was 4268 buildings. Below table explain the expected population that can live in those buildings as listed in Table 4.3.

Table 4.3 Expected Population According to Buildings Scanned at October 2016

|  | Building type |  |  | expected population per Building type (person) |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| District | Apartment | Palace | villa | Apartment | Palace | villa | Grand Total |
| 1 | 3 |  | 27 | 268 |  | 520 | 788 |
| 2 | 7 |  | 64 | 624 |  | 1233 | 1857 |
| 3 | 6 |  | 60 | 535 |  | 1156 | 1691 |
| 4 | 3 |  | 11 | 268 |  | 212 | 479 |
| 5 |  |  | 20 |  |  | 385 | 385 |
| 6 | 3 |  | 18 | 268 |  | 347 | 614 |
| 7 |  | 2 | 47 |  | 18 | 906 | 923 |
| 8 | 1 | 4 | 54 | 89 | 35 | 1040 | 1165 |
| 9 | 5 |  | 9 | 446 |  | 173 | 619 |
| 10 | 14 |  | 16 | 1248 |  | 308 | 1557 |
| 11 | 3 |  | 23 | 268 |  | 443 | 711 |
| 12 | 2 |  | 45 | 178 |  | 867 | 1045 |
| 13 | 5 | 9 | 101 | 446 | 79 | 1946 | 2471 |
| 14 | 6 |  | 58 | 535 |  | 1118 | 1653 |
| 15 | 31 |  | 130 | 2764 |  | 2505 | 5269 |
| 16 | 5 |  | 38 | 446 |  | 732 | 1178 |
| 17 | 7 |  | 42 | 624 |  | 809 | 1433 |
| 18 | 3 | 1 | 77 | 268 | 9 | 1484 | 1760 |
| 19 | 6 |  | 37 | 535 |  | 713 | 1248 |
| 20 | 28 |  | 145 | 2497 |  | 2794 | 5290 |
| 21 | 15 |  | 69 | 1338 |  | 1329 | 2667 |
| 22 | 8 |  | 44 | 713 |  | 848 | 1561 |
| 23 | 23 |  | 165 | 2051 |  | 3179 | 5230 |
| 24 | 16 |  | 141 | 1427 |  | 2717 | 4143 |
| 25 | 34 |  | 211 | 3032 |  | 4065 | 7097 |
| 26 | 17 |  | 97 | 1516 |  | 1869 | 3385 |
| 27 | 24 |  | 259 | 2140 |  | 4990 | 7130 |
| 28 |  |  | 29 |  |  | 559 | 559 |
| 29 | 1 |  | 3 | 89 |  | 58 | 147 |
| 30 | 13 |  | 41 | 1159 |  | 790 | 1949 |
| 31 | 14 |  | 117 | 1248 |  | 2254 | 3503 |
| 32 |  |  | 23 |  |  | 443 | 443 |
| 33 |  |  | 57 |  |  | 1098 | 1098 |
| 35 | 1 |  | 17 | 89 |  | 328 | 417 |
| 36 | 1 |  | 14 | 89 |  | 270 | 359 |
| 37 | 6 |  | 151 | 535 |  | 2909 | 3444 |
| 38 | 20 |  | 459 | 1783 |  | 8844 | 10627 |
| 39 |  |  | 133 |  |  | 2563 | 2563 |
| 40 | 39 |  | 225 | 3478 |  | 4335 | 7813 |
| 41 | 19 |  | 84 | 1694 |  | 1618 | 3313 |
| 42 | 8 |  | 34 | 713 |  | 655 | 1368 |
| 43 | 30 | 2 | 120 | 2675 | 18 | 2312 | 5005 |
| 44 | 2 |  | 111 | 178 |  | 2139 | 2317 |
| Grand Total | 429 | 18 | 3626 | 38253 | 158 | 69864 | 108275 |

Figure 4.3 shows the expected population according to October 2016 buildings, districts with dark red are the highest population density while the dark blue color reflects the lowest. But this map does not reflect the reality because there are different building construction stages that should consider it during the analysis. Its meaning the duration to complete the construction should be taking in our account, and this will let us determine the expected population growth in the following periods.


Figure 4.3 Expected Population at October 2016.

The details of the residential buildings according to its construction stage which detected in October image is shown in table 4.4, If we take into account the table 3.2 which explain the building duration) and table 4.4 below, the results are the constructions which will be ready to living depending on the required time to complete, which is helping to identify the buildings that will be ready for living in that time and its population capacity, table 4.5.

Table 4.4 Total Buildings for October 2016 Construction with its Buildings Stage

|  | Construction level |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| Building type | Basement | Under <br> construction | construction completed | Grand Total |
| Apartment | 81 | 127 | 221 | 429 |
| Palace | 6 | 9 | 3 | 18 |
| villa | 764 | 640 | 2222 | 3626 |
| Grand Total | 851 | 776 | 2446 | 4073 |

Table 4.4 shows us the expected population after three months, six months, one and two years for October buildings depending on the districts. After three months (starting from October 2016) the expected of the new population is 62545 person, after six months the expected population will increase by 12331 people and so on, the total expected increase in population 108275 Depending on the October buildings.

Table 4.5 Expected Population According to The Construction Stage for October
Buildings.

|  | Expected population according to the construction stage for October buildings |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| (person) |  |  |  |  |  |



Figure 4.4 Expected Population According to the Construction Stage for October Buildings.

Figure 4.4 analysis: the majority of population growth will be three months later starting from October 2016, while the lowest growth will be two years later beginning in October 2016, so this is related to the buildings increasing which also the building types have an impact. The results in Table 4.5 explains in figures 4.5 to 4.8 as the population segmentation, as a visual map to be more readable.


Figure 4.5 The Expected Increase in Population Three Months Later per Districts


Figure 4.6 The Expected Increase in Population Six Months Later per Districts


Figure 4.7 The Expected Increase in Population One Year Later per Districts


Figure 4.8 the Expected Increase in Population Two Years Later per Districts

Maps analysis: Figures 4.5 to 4.8 shows the expected population increases according to October 2016 buildings for the periods three months to two years. Districts with dark red are the highest population growth while the dark blue colour reflects the lowest.

### 4.2 Cumulative Population

Furthermore, if we want to know the cumulative population for the same periods (three months, six months, one year and two years), namely the expected population growth after three months was 1874 but the cumulative population is the population living in the studying area is the current population (January 2016) plus the expected population growth after three months 1160087 plus 62545 this equal to 1222632 people. By applying this equation:
CUM (n) = POP (Current) + Ex POP (n)
$\mathrm{n}=$ after three-months, $\operatorname{Ex} \operatorname{POP}(\mathrm{n})=$ expected population growth after $(\mathrm{n})$ period POP (current) = January 2016
CUM (3) $=1160087+62545=1222632$ person
While the six-month population cumulative is three months cumulative plus six months expected growth $1222632+12331=1234963$ people.

It is essential for forecasting and planning; the decision-makers will know with reliability and accuracy what the region needs to, what the market size when the market needs the services, the government will be able to improve its planning to meet the user's requirements.

Table 4.6 Expected Cumulative Population According to the Construction Stage for
Total Buildings

|  |  | October Buildings influences |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| District | January population | Three Month cumulative | Six Month cumulative | One Year cumulative | Two Year cumulative |
| 1 | 53017 | 53619 | 53638 | 53805 | 53805 |
| 2 | 70235 | 71587 | 71683 | 72003 | 72093 |
| 3 | 61982 | 63314 | 63391 | 63584 | 63673 |
| 4 | 44802 | 44918 | 45014 | 45281 | 45281 |
| 5 | 22917 | 23226 | 23226 | 23303 | 23303 |
| 6 | 18882 | 19349 | 19349 | 19407 | 19496 |
| 7 | 16030 | 16415 | 16473 | 16935 | 16953 |
| 8 | 14628 | 15227 | 15362 | 15775 | 15793 |
| 9 | 22471 | 22854 | 22874 | 23001 | 23091 |
| 10 | 90508 | 91166 | 91224 | 91798 | 92065 |
| 11 | 45923 | 46128 | 46340 | 46634 | 46634 |
| 12 | 56531 | 57371 | 57429 | 57487 | 57576 |
| 13 | 36723 | 38028 | 38356 | 39105 | 39194 |
| 14 | 8157 | 9106 | 9164 | 9720 | 9809 |
| 15 | 68590 | 72337 | 72645 | 73592 | 73859 |
| 16 | 79836 | 80701 | 80797 | 80925 | 81014 |
| 17 | 54269 | 55313 | 55428 | 55614 | 55703 |
| 18 | 8447 | 9441 | 9653 | 10109 | 10207 |
| 19 | 8900 | 9476 | 9515 | 10148 | 10148 |
| 20 | 26940 | 29807 | 30192 | 31785 | 32231 |
| 21 | 41975 | 43548 | 43991 | 44553 | 44642 |
| 22 | 72172 | 73336 | 73471 | 73734 | 73734 |
| 23 | 54786 | 57966 | 58313 | 59481 | 60016 |
| 24 | 8277 | 10577 | 11136 | 11796 | 12420 |
| 25 | 26431 | 30534 | 31246 | 33082 | 33528 |
| 26 | 1727 | 3623 | 3950 | 4577 | 5112 |
| 27 | 8459 | 12241 | 13166 | 15055 | 15679 |
| 28 | 7158 | 7351 | 7659 | 7717 | 7717 |
| 29 | 3724 | 3743 | 3781 | 3871 | 3871 |
| 30 | 8861 | 9501 | 9636 | 10542 | 10810 |
| 31 | 15456 | 17152 | 17287 | 18781 | 18959 |
| 32 | 713 | 732 | 771 | 1156 | 1156 |
| 33 | 790 | 1118 | 1349 | 1888 | 1888 |
| 35 | 448 | 544 | 660 | 865 | 865 |
| 36 | 4422 | 4608 | 4723 | 4781 | 4781 |
| 37 | 35203 | 36930 | 37681 | 38380 | 38647 |
| 38 | 12056 | 19376 | 21130 | 22692 | 22871 |
| 39 | 6526 | 8106 | 8530 | 9089 | 9089 |
| 40 | 13734 | 18887 | 19619 | 21279 | 21636 |
| 41 | 8590 | 10465 | 10773 | 11725 | 11903 |
| 42 | 4406 | 5179 | 5276 | 5596 | 5774 |
| 43 | 10159 | 12533 | 13362 | 14441 | 15252 |
| 44 | 3771 | 4746 | 5247 | 5999 | 6088 |
| Total population | 1159633 | 1222178 | 1234509 | 1261087 | 1268363 |



Figure 4.9 Expected Cumulative Population According to the Construction Stage for Total Buildings

Figure 4.9 analysis: the charts show us that the highest population cumulative was two years later after October 2016. While the lowest population cumulative was three months later after October 2016.


Figure 4.10 The Expected Cumulative Population one Year Later


Figure 4.11 The Expected Cumulative Population Two Years Later

Maps analysis: The figures 4.10 and 4.11 show the expected cumulative population which calculated for all buildings that detected in January and October 2016. Districts with dark Red are the highest population growth while the dark blue colour reflects the lowest.

### 4.3 Milk Market Size

The information that obtained from Satellite images and Geomarketing which provide information about the existing buildings which detected at January 2016 and new buildings which built after January 2016 and detected at October 2016, these information will support the business and industry with the new tools which gives the required data about the new business volume with associated geospatial data for better understanding about the market situations, market size and the potential market, potential manpower or any required information that can support the decision- makers. The big challenge for the Science and the researchers are using it in industry or business, we studied the population forecasting according to the periods and micro-geographic areas, and this will be valuable for studying FMCG products like Milk. Which is an essential strategic product to find out how much people need it and what time it takes to need it, this kind of FMCG cannot keep for a long time it would, therefore, be useful to know how much people will need (Quantity), where and when? This is a challenge to improve planning, control production to meet the need of people for fresh milk daily and prevent any waste in the resources.

### 4.3.1 January 2016 Milk Market size

We will discuss in the following the milk market size depending on the population, which was estimated based on the existing buildings in the previous sections. The estimation population in January 2016 was 1160087 person in the studying area, table 4.2, by taking it in our account and consider the Nielsen survey, table 3.3. The following table 4.7 explains the milk consumption daily, and yearly in the study area for the population which lives in January 2016 buildings, we need this total volume to understand the expected market size Regardless of the subsequent increase in the population which we will study it for October buildings.

Table 4.7 Expected Daily and Yearly Milk Consumption According to January

> Buildings

| District | Yearly Consumption(L) | Daily Consumption(L) |
| :---: | :---: | :---: |
| 1 | 2128628 | 5832 |
| 2 | 2819950 | 7726 |
| 3 | 2488576 | 6818 |
| 4 | 1798797 | 4928 |
| 5 | 920137 | 2521 |
| 6 | 758093 | 2077 |
| 7 | 643585 | 1763 |
| 8 | 587320 | 1609 |
| 9 | 902225 | 2472 |
| 10 | 3633912 | 9956 |
| 11 | 1843824 | 5052 |
| 12 | 2269704 | 6218 |
| 13 | 1474442 | 4040 |
| 14 | 327498 | 897 |
| 15 | 2753899 | 7545 |
| 16 | 3205410 | 8782 |
| 17 | 2178918 | 5970 |
| 18 | 339136 | 929 |
| 19 | 357346 | 979 |
| 20 | 1081647 | 2963 |
| 21 | 1685292 | 4617 |
| 22 | 2897724 | 7939 |
| 23 | 2199670 | 6026 |
| 24 | 332302 | 910 |
| 25 | 1061201 | 2907 |
| 26 | 69336 | 190 |
| 27 | 343223 | 940 |
| 28 | 287399 | 787 |
| 29 | 149501 | 410 |
| 30 | 355751 | 975 |
| 31 | 620574 | 1700 |
| 32 | 28623 | 78 |
| 33 | 31717 | 87 |
| 35 | 17990 | 49 |
| 36 | 177549 | 486 |
| 37 | 1413392 | 3872 |
| 38 | 491577 | 1347 |
| 39 | 262025 | 718 |
| 40 | 554986 | 1521 |
| 41 | 344899 | 945 |
| 42 | 176900 | 485 |
| 43 | 411446 | 1127 |
| 44 | 151401 | 415 |
| Total | 46577525 | 127610 |



Figure 4.12 Expected Daily Milk Consumption According to January Buildings.


Figure 4.13 Expected Yearly Milk Consumption According to January Buildings.

Maps analysis: Figure 4.12 and Figure 4.13 show the highest districts potential that required the planning team, and decision maker's to focus in. The dark blue reflect the highest expected consumption while the medium yellow is the lowest potential market size.

### 4.3.2 October 2016 Milk Market Size

### 4.3.2.1 Daily Milk Market Size Forecasting

This section discusses the methodology of forecasting the demand for fresh milk according to the population growth studied, based on the increase in urban growth. As we reviewed, we have population growth depending on the new buildings and its stages, which will give us the chance to calculate the population's needs for different products like fresh milk continuously. Will consider the expected needs after three months, six months, one year, two years and the cumulative needs for the same periods, will study it daily and yearly.

## A. Daily Milk market size after three months

At the beginning of 2016 (at January) the expected population was 1160087 person, (table 4.2), so we calculated its expected Milk needs ( 127609.68 L per day), the planned capacity of the study area can increase by 62545 persons after three months, here we can know the expected human milk needs after this period by applying this equation:
$\mathrm{DM}=\operatorname{Ex} \operatorname{POP}(\mathrm{n}) * \operatorname{AVE} M$ 4.1.

DM = Daily Milk needs
$\operatorname{Ex} \operatorname{POP}(\mathrm{n})=$ expected population growth after $(\mathrm{n})$ period, $\mathrm{n}=$ three month
AVE M = person's daily milk consumption
Daily Milk needs= population for this period * person's daily consumption ( 0.11 L ) $=62545^{*} 0.11=\underline{6879.95} \mathrm{~L} /$ day .
This means we will need after three months (starting from April 2016) 6879.95 L Milk every day for those new people in studying area, Figure (4.14), here the concerned companies must plan to meet these requirements.


Figure 4.14 Daily Expected Milk Market Size (Three Month Later)

Then we can expect the cumulative market size for the same period (three months), we can calculate it with this equation:
$\operatorname{CUM} \operatorname{DM}(\mathrm{n})=\mathrm{DM}(\mathrm{n})+\mathrm{DM}(\mathrm{n}-\mathrm{t})$
CUM DM (n) = Cumulative milk market size a specific period
DM $(\mathrm{n}-\mathrm{t})=$ previous period Market size
$\mathrm{N}=$ October
$\mathrm{t}=$ previous period (here January 2016)
Cumulative milk market size= Daily milk needs for this period + previous period Market size $($ here January $)=6879.95+127559.65=127559.65$ L per day, Figure 4.15


Figure 4.15 Daily Cumulative Expected Milk Market Size (Three Month Later)

The consumption at the districts level shown in the following table 4.8. It showed the difference between expected cumulative consumption and expected to increase for the same period as a result of the increase in population in each district.

Table 4.8 Daily Expected Milk Market Size Three Months later

| District | Three Month later (litre) | Three months cumulative (Litre) |
| :---: | :---: | :---: |
| 1 | 66 | 5898 |
| 2 | 149 | 7875 |
| 3 | 147 | 6965 |
| 4 | 13 | 4941 |
| 5 | 34 | 2555 |
| 6 | 51 | 2128 |
| 7 | 42 | 1806 |
| 8 | 66 | 1675 |
| 9 | 42 | 2514 |
| 10 | 72 | 10028 |
| 11 | 23 | 5074 |
| 12 | 92 | 6311 |
| 13 | 144 | 4183 |
| 14 | 104 | 1002 |
| 15 | 412 | 7957 |
| 16 | 95 | 8877 |
| 17 | 115 | 6084 |
| 18 | 109 | 1039 |
| 19 | 63 | 1042 |
| 20 | 315 | 3279 |
| 21 | 173 | 4790 |
| 22 | 128 | 8067 |
| 23 | 350 | 6376 |
| 24 | 253 | 1163 |
| 25 | 451 | 3359 |
| 26 | 209 | 399 |
| 27 | 416 | 1347 |
| 28 | 21 | 809 |
| 29 | 2 | 412 |
| 30 | 70 | 1045 |
| 31 | 187 | 1887 |
| 32 | 2 | 81 |
| 33 | 36 | 123 |
| 35 | 11 | 60 |
| 36 | 20 | 507 |
| 37 | 190 | 4062 |
| 38 | 805 | 2131 |
| 39 | 174 | 892 |
| 40 | 567 | 2078 |
| 41 | 206 | 1151 |
| 42 | 85 | 570 |
| 43 | 261 | 1379 |
| 44 | 107 | 522 |
| Grand Total | 6880 | 134440 |



Figure 4.16 Difference between Market Size and Cumulative Market Size after Three

## Months

Figure 4.16 explains how we can distinguish between the cumulative market size and the increasing market size for the same period. The cumulative size will tell us how much the market size will be after the period that we want to prepare to forecast to it, while the increase in market size tells us only what the expected value that added to the market, the cumulative demand is higher than the other value.
B. Daily Milk market size for the others periods

With the same methodology, Milk market size can expect for different periods. For Milk market size, the table (4.9) shows the market size for varying periods, while the table (4.10) shows us the expected cumulative milk market size for the same periods.

Table 4.9 Daily Expected Consumption According to October Buildings at Districts
Level (by Litre)

| District | Januar y | Three Month later | Six Month later | One Year later | Two Year later | Grand Total |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 5832 | 66 | 2 | 18 | 0 | 5919 |
| 2 | 7726 | 149 | 11 | 35 | 10 | 7930 |
| 3 | 6818 | 147 | 8 | 21 | 10 | 7004 |
| 4 | 4928 | 13 | 11 | 29 | 0 | 4981 |
| 5 | 2521 | 34 | 0 | 8 | 0 | 2563 |
| 6 | 2077 | 51 | 0 | 6 | 10 | 2145 |
| 7 | 1763 | 42 | 6 | 51 | 2 | 1865 |
| 8 | 1609 | 66 | 15 | 45 | 2 | 1737 |
| 9 | 2472 | 42 | 2 | 14 | 10 | 2540 |
| 10 | 9956 | 72 | 6 | 63 | 29 | 10127 |
| 11 | 5052 | 23 | 23 | 32 | 0 | 5130 |
| 12 | 6218 | 92 | 6 | 6 | 10 | 6333 |
| 13 | 4040 | 144 | 36 | 82 | 10 | 4311 |
| 14 | 897 | 104 | 6 | 61 | 10 | 1079 |
| 15 | 7545 | 412 | 34 | 104 | 29 | 8125 |
| 16 | 8782 | 95 | 11 | 14 | 10 | 8912 |
| 17 | 5970 | 115 | 13 | 20 | 10 | 6127 |
| 18 | 929 | 109 | 23 | 50 | 11 | 1123 |
| 19 | 979 | 63 | 4 | 70 | 0 | 1116 |
| 20 | 2963 | 315 | 42 | 175 | 49 | 3545 |
| 21 | 4617 | 173 | 49 | 62 | 10 | 4911 |
| 22 | 7939 | 128 | 15 | 29 | 0 | 8111 |
| 23 | 6026 | 350 | 38 | 129 | 59 | 6602 |
| 24 | 910 | 253 | 61 | 73 | 69 | 1366 |
| 25 | 2907 | 451 | 78 | 202 | 49 | 3688 |
| 26 | 190 | 209 | 36 | 69 | 59 | 562 |
| 27 | 931 | 416 | 102 | 208 | 69 | 1725 |
| 28 | 787 | 21 | 34 | 6 | 0 | 849 |
| 29 | 410 | 2 | 4 | 10 | 0 | 426 |
| 30 | 975 | 70 | 15 | 100 | 29 | 1189 |
| 31 | 1700 | 187 | 15 | 164 | 20 | 2085 |
| 32 | 78 | 2 | 4 | 42 | 0 | 127 |
| 33 | 87 | 36 | 25 | 59 | 0 | 208 |
| 35 | 49 | 11 | 13 | 23 | 0 | 95 |
| 36 | 486 | 20 | 13 | 6 | 0 | 526 |
| 37 | 3872 | 190 | 83 | 77 | 29 | 4251 |
| 38 | 1326 | 805 | 193 | 172 | 20 | 2516 |
| 39 | 718 | 174 | 47 | 61 | 0 | 1000 |
| 40 | 1511 | 567 | 81 | 183 | 39 | 2380 |
| 41 | 945 | 206 | 34 | 105 | 20 | 1309 |
| 42 | 485 | 85 | 11 | 35 | 20 | 635 |
| 43 | 1117 | 261 | 91 | 119 | 89 | 1678 |
| 44 | 415 | 107 | 55 | 83 | 10 | 670 |
| Grand <br> Total | 127560 | 6880 | 1356 | 2924 | 800 | 139520 |

Table 4.10 Daily Expected Cumulative Consumption According to October Buildings (by Litre)

| District | January | Three months cumulative | Six months cumulative | One year cumulative | Two year cumulative |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 5832 | 5898 | 5900 | 5919 | 5919 |
| 2 | 7726 | 7875 | 7885 | 7920 | 7930 |
| 3 | 6818 | 6965 | 6973 | 6994 | 7004 |
| 4 | 4928 | 4941 | 4952 | 4981 | 4981 |
| 5 | 2521 | 2555 | 2555 | 2563 | 2563 |
| 6 | 2077 | 2128 | 2128 | 2135 | 2145 |
| 7 | 1763 | 1806 | 1812 | 1863 | 1865 |
| 8 | 1609 | 1675 | 1690 | 1735 | 1737 |
| 9 | 2472 | 2514 | 2516 | 2530 | 2540 |
| 10 | 9956 | 10028 | 10035 | 10098 | 10127 |
| 11 | 5052 | 5074 | 5097 | 5130 | 5130 |
| 12 | 6218 | 6311 | 6317 | 6324 | 6333 |
| 13 | 4040 | 4183 | 4219 | 4302 | 4311 |
| 14 | 897 | 1002 | 1008 | 1069 | 1079 |
| 15 | 7545 | 7957 | 7991 | 8095 | 8125 |
| 16 | 8782 | 8877 | 8888 | 8902 | 8912 |
| 17 | 5970 | 6084 | 6097 | 6118 | 6127 |
| 18 | 929 | 1039 | 1062 | 1112 | 1123 |
| 19 | 979 | 1042 | 1047 | 1116 | 1116 |
| 20 | 2963 | 3279 | 3321 | 3496 | 3545 |
| 21 | 4617 | 4790 | 4839 | 4901 | 4911 |
| 22 | 7939 | 8067 | 8082 | 8111 | 8111 |
| 23 | 6026 | 6376 | 6414 | 6543 | 6602 |
| 24 | 910 | 1163 | 1225 | 1298 | 1366 |
| 25 | 2907 | 3359 | 3437 | 3639 | 3688 |
| 26 | 190 | 399 | 435 | 503 | 562 |
| 27 | 931 | 1347 | 1448 | 1656 | 1725 |
| 28 | 787 | 809 | 843 | 849 | 849 |
| 29 | 410 | 412 | 416 | 426 | 426 |
| 30 | 975 | 1045 | 1060 | 1160 | 1189 |
| 31 | 1700 | 1887 | 1902 | 2066 | 2085 |
| 32 | 78 | 81 | 85 | 127 | 127 |
| 33 | 87 | 123 | 148 | 208 | 208 |
| 35 | 49 | 60 | 73 | 95 | 95 |
| 36 | 486 | 507 | 520 | 526 | 526 |
| 37 | 3872 | 4062 | 4145 | 4222 | 4251 |
| 38 | 1326 | 2131 | 2324 | 2496 | 2516 |
| 39 | 718 | 892 | 938 | 1000 | 1000 |
| 40 | 1511 | 2078 | 2158 | 2341 | 2380 |
| 41 | 945 | 1151 | 1185 | 1290 | 1309 |
| 42 | 485 | 570 | 580 | 616 | 635 |
| 43 | 1117 | 1379 | 1470 | 1589 | 1678 |
| 44 | 415 | 522 | 577 | 660 | 670 |
| Grand Total | 127560 | 134440 | 135796 | 138720 | 139520 |



Figure 4.17 Contribution of Cumulative Daily Expected Milk Market Size over the Year

Figure 4.17 shows the contribution of population growth rates to the population's milk need according to time during the study period, two years after October 2016.


Figures 4. 18 Daily Expected Milk Market Size (Six Months Later).


Figures 4. 19 Daily Cumulative Expected Milk Market Size (Six Months Later).

Maps analysis: Figure 4.18 and 4.19 show the highest districts potential that required the planning team and decision maker's to focus on. The dark red reflect the highest expected consumption while the dark blue is the lowest potential market size, (district with no colour mean there are no buildings at that time), and by using this maps we can
determine where the decision makers have to take the right action on time to prevent any bottleneck in the future.

### 4.3.2.2 Yearly Milk Market Size Forecasting.

Knowledge of the expected annual market size will enhance the planning process and gives the ability to meet the requirements in time for having the necessary resources. With same mechanism which we used in the previous section (Daily Milk market size forecasting) to calculate the results for the yearly expected demand, Table 4.11 shows us the annual expected milk market size for the October population growth at districts level, its useful to have good knowledge about the people Milk needs for the required periods according to populations growth.

Table 4.11 Yearly Milk Consumption According to October Buildings Stage

| District | January milk Demand | Three Month later | Six Month later | One Year later | Two Year later | Grand Total |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 2128628 | 24179 | 774 | 6674 | 0 | 2160256 |
| 2 | 2819950 | 54259 | 3868 | 12863 | 3580 | 2894520 |
| 3 | 2488576 | 53486 | 3094 | 7736 | 3580 | 2556472 |
| 4 | 1798797 | 4642 | 3868 | 10740 | 0 | 1818047 |
| 5 | 920137 | 12378 | 0 | 3094 | 0 | 935609 |
| 6 | 758093 | 18764 | 0 | 2321 | 3580 | 782758 |
| 7 | 643585 | 15472 | 2321 | 18566 | 705 | 680650 |
| 8 | 587320 | 24046 | 5415 | 16598 | 705 | 634084 |
| 9 | 902225 | 15382 | 774 | 5127 | 3580 | 927088 |
| 10 | 3633912 | 26410 | 2321 | 23027 | 10740 | 3696410 |
| 11 | 1843824 | 8222 | 8510 | 11802 | 0 | 1872356 |
| 12 | 2269704 | 33750 | 2321 | 2321 | 3580 | 2311676 |
| 13 | 1474442 | 52381 | 13151 | 30094 | 3580 | 1573648 |
| 14 | 327498 | 38104 | 2321 | 22344 | 3580 | 393846 |
| 15 | 2753899 | 150417 | 12378 | 38014 | 10740 | 2965447 |
| 16 | 3205410 | 34722 | 3868 | 5127 | 3580 | 3252706 |
| 17 | 2178918 | 41882 | 4642 | 7448 | 3580 | 2236469 |
| 18 | 339136 | 39939 | 8510 | 18278 | 3933 | 409795 |
| 19 | 357346 | 23118 | 1547 | 25438 | 0 | 407449 |
| 20 | 1081647 | 115103 | 15472 | 63938 | 17900 | 1294059 |
| 21 | 1685292 | 63164 | 17793 | 22542 | 3580 | 1792371 |
| 22 | 2897724 | 46721 | 5415 | 10542 | 0 | 2960402 |
| 23 | 2199670 | 127660 | 13925 | 46919 | 21480 | 2409654 |
| 24 | 332302 | 92363 | 22434 | 26500 | 25060 | 498660 |
| 25 | 1061201 | 164720 | 28623 | 73707 | 17900 | 1346151 |
| 26 | 69336 | 76118 | 13151 | 25150 | 21480 | 205235 |
| 27 | 339643 | 151840 | 37133 | 75830 | 25060 | 629505 |
| 28 | 287399 | 7736 | 12378 | 2321 | 0 | 309833 |
| 29 | 149501 | 774 | 1547 | 3580 | 0 | 155402 |
| 30 | 355751 | 25726 | 5415 | 36376 | 10740 | 434009 |
| 31 | 620574 | 68094 | 5415 | 59962 | 7160 | 761205 |
| 32 | 28623 | 774 | 1547 | 15472 | 0 | 46416 |
| 33 | 31717 | 13151 | 9283 | 21661 | 0 | 75812 |
| 35 | 17990 | 3868 | 4642 | 8222 | 0 | 34722 |
| 36 | 177549 | 7448 | 4642 | 2321 | 0 | 191959 |
| 37 | 1413392 | 69336 | 30170 | 28047 | 10740 | 1551685 |
| 38 | 484064 | 293893 | 70397 | 62743 | 7160 | 918257 |
| 39 | 262025 | 63435 | 17019 | 22434 | 0 | 364913 |
| 40 | 551406 | 206890 | 29397 | 66654 | 14320 | 868667 |
| 41 | 344899 | 75254 | 12378 | 38211 | 7160 | 477902 |
| 42 | 176900 | 31051 | 3868 | 12863 | 7160 | 231842 |
| 43 | 407865 | 95342 | 33265 | 43339 | 32573 | 612383 |
| 44 | 151401 | 39165 | 20113 | 30170 | 3580 | 244430 |
| Grand Total | 46559272 | 2511174 | 495100 | 1067117 | 292099 | 50924762 |



Figure 4.20 Yearly Expected Milk Market Size (Three Month Later)

### 4.3.2.3 Cumulative Yearly Milk Market Size Forecasting

We need to calculate the cumulative annual milk volume to form a broader idea of the actual need for the population based on the consumption data that previously estimated in the daily tables. Table 4.12 shows us the yearly cumulative expected milk market size for October growth at districts. This table provides us with the reading for the requirements to put it in our plan.

Table 4.12 Yearly Cumulative Consumption According to October Buildings (Litre)

| District | $\begin{gathered} \text { January } \\ \text { milk } \\ \text { Demand } \end{gathered}$ | Three Month cumulative | Six Month cumulative | One Year cumulative | Two Year cumulative |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 2128628 | 2152808 | 2153581 | 2160256 | 2160256 |
| 2 | 2819950 | 2874209 | 2878077 | 2890940 | 2894520 |
| 3 | 2488576 | 2542062 | 2545156 | 2552892 | 2556472 |
| 4 | 1798797 | 1803439 | 1807307 | 1818047 | 1818047 |
| 5 | 920137 | 932515 | 932515 | 935609 | 935609 |
| 6 | 758093 | 776857 | 776857 | 779178 | 782758 |
| 7 | 643585 | 659057 | 661378 | 679944 | 680650 |
| 8 | 587320 | 611366 | 616781 | 633379 | 634084 |
| 9 | 902225 | 917607 | 918380 | 923508 | 927088 |
| 10 | 3633912 | 3660321 | 3662642 | 3685670 | 3696410 |
| 11 | 1843824 | 1852045 | 1860555 | 1872356 | 1872356 |
| 12 | 2269704 | 2303454 | 2305775 | 2308096 | 2311676 |
| 13 | 1474442 | 1526823 | 1539974 | 1570068 | 1573648 |
| 14 | 327498 | 365601 | 367922 | 390266 | 393846 |
| 15 | 2753899 | 2904316 | 2916694 | 2954707 | 2965447 |
| 16 | 3205410 | 3240131 | 3243999 | 3249126 | 3252706 |
| 17 | 2178918 | 2220799 | 2225441 | 2232889 | 2236469 |
| 18 | 339136 | 379075 | 387584 | 405863 | 409795 |
| 19 | 357346 | 380463 | 382010 | 407449 | 407449 |
| 20 | 1081647 | 1196749 | 1212221 | 1276159 | 1294059 |
| 21 | 1685292 | 1748457 | 1766249 | 1788791 | 1792371 |
| 22 | 2897724 | 2944445 | 2949860 | 2960402 | 2960402 |
| 23 | 2199670 | 2327331 | 2341255 | 2388174 | 2409654 |
| 24 | 332302 | 424665 | 447099 | 473599 | 498660 |
| 25 | 1061201 | 1225922 | 1254545 | 1328251 | 1346151 |
| 26 | 69336 | 145453 | 158604 | 183755 | 205235 |
| 27 | 339643 | 491482 | 528615 | 604444 | 629505 |
| 28 | 287399 | 295135 | 307512 | 309833 | 309833 |
| 29 | 149501 | 150275 | 151822 | 155402 | 155402 |
| 30 | 355751 | 381477 | 386893 | 423269 | 434009 |
| 31 | 620574 | 688668 | 694083 | 754045 | 761205 |
| 32 | 28623 | 29397 | 30944 | 46416 | 46416 |
| 33 | 31717 | 44868 | 54152 | 75812 | 75812 |
| 35 | 17990 | 21858 | 26500 | 34722 | 34722 |
| 36 | 177549 | 184997 | 189638 | 191959 | 191959 |
| 37 | 1413392 | 1482728 | 1512898 | 1540945 | 1551685 |
| 38 | 484064 | 777957 | 848354 | 911097 | 918257 |
| 39 | 262025 | 325460 | 342479 | 364913 | 364913 |
| 40 | 551406 | 758296 | 787693 | 854347 | 868667 |
| 41 | 344899 | 420153 | 432530 | 470742 | 477902 |
| 42 | 176900 | 207951 | 211819 | 224682 | 231842 |
| 43 | 407865 | 503207 | 536472 | 579810 | 612383 |
| 44 | 151401 | 190567 | 210680 | 240850 | 244430 |
| Grand Total | 46559272 | 49070445 | 49565546 | 50632663 | 50924762 |



Figure 4.21 Yearly Cumulative Expected Milk Market Size (Three Month Later)

## Analysis:

In this section, population data that were estimated based on the type and quantity of existing residential buildings and the size and type of growth in the residential building sector were used to determine the approximate milk market size in various periods, depending on the speed of establishing suitable conditions for the occupancy of the new buildings. This prior knowledge will enable a broad and comprehensive understanding of strategic requirements Of the population to meet their needs of the required products.

### 4.4 Building Materials

The High-resolution satellite image has been used to extract the new buildings built after January 2016. The number of these opportunities was 4268 and classified according to the building type and its construction level as in Table (3.2). Based on the new division of the study area, in each micro-geographic area using a geographic information system (GIS) to determine growth in micro-geographic.

By identifying the number of new buildings and the type of each building in each microgeographic, and using Table 3.2 which explain the average number of (switches and sockets) needed by each building type, then we can calculate the market size in the studying area, it is possible to know the needs of each micro-geographic area for the products and their distributions using geographic information systems. Figure 4.22 explains the market size at the micro-geographic area level which using GIS to convert the data to a visual map to know the potential market size for each district.

Building materials are a critical factor in human activity and urban development; Logistic operations are the cornerstone to provide the sites with these materials and keeping sufficient stock to prevent any bottlenecks in the stock or to keep due to manufacturing excess quantities. To keep our resources, we have to forecast the market size for these materials by expecting when the activities to conduct on dates that are aware of the duration that we may need these materials. At this time, remote sensing provides new tools and readable data after analysing it; this information will give us about the expected time for product installation and how much we need. In the following, we will use satellite data for products forecasting and planning to study the switches and socket market size.

A survey conducted for this products to know the number of switches which each building needs during the finishing stage, groups that were targeted by this survey is the electricians and contractors at the sites, the question was:
a) How many products each building type needs?
b) Time to install the product (finishing stage)!

Results tabulated in Table (3.1 \& 3.2)
Determine the current building construction stage is very important to estimate the periods that building needs to install this product, the construction building stages are:
a) Basements
b) Under construction
c) Construction completes

Switches study targeted the building which began construction after January 2016 or in another meaning after the first images have taken, if we have the number of buildings for each building type then we can expect the quantities of the products that we need table
4.13.

Table 4.13 Total Construction Requirement for the Switches

|  |  | Time to installation (Month) |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Buildings <br> type | Switches <br> Average | Basement | Under <br> construction | construction <br> completed | Number of <br> Buildings | Products <br> Quantities |
| villa | 194 | 12 | 6 | 3 | 3627 | 702410 |
| Apartment | 771 | 24 | 12 | 6 | 429 | 330919 |
| Esteraha | 49 | 24 | 12 | 6 | 57 | 2772 |
| Mosque | 105 | 24 | 12 | 6 | 15 | 1570 |
| tower | 4188 | 24 | 12 | 6 | 5 | 20942 |
| Gas station | 93 | 12 | 6 | 3 | 8 | 746 |
| School | 426 | 24 | 12 | 6 | 9 | 3830 |
| Mall | 227 | 24 | 12 | 6 | 1 | 227 |
| showroom | 351 | 12 | 6 | 3 | 94 | 33030 |
| Palace | 426 | 24 | 12 | 6 | 18 | 7660 |
| warehouse | 30 | 9 | 6 | 3 | 2 | 60 |
| Governmental | 1677 | 24 | 12 | 6 | 3 | 5031 |

At the moment, we have become familiar with the expected building requirements of the products which can install in these buildings, but this information will be more valuable in case determine the time to install it and demand at the district level. By using Geospatial information, we studied that the new constructions distributed over the districts, and we have an idea about how many new constructions for each district, Table 4.14 show the expected switches and sockets market size at districts level. Identify the number and type of new buildings in each district will allow determining the expected need for the products required to these buildings based on previous surveys which carried out.

Table 4.14 Show the Expected Switches Market Demand

| District | Market size | District | Market size |
| :---: | :---: | :---: | :---: |
| 1 | 7543 | 23 | 52649 |
| 2 | 18571 | 24 | 41603 |
| 3 | 17702 | 25 | 71700 |
| 4 | 4796 | 26 | 34347 |
| 5 | 5902 | 27 | 71055 |
| 6 | 6503 | 28 | 5968 |
| 7 | 10379 | 29 | 2552 |
| 8 | 14016 | 30 | 18566 |
| 9 | 5600 | 31 | 34235 |
| 10 | 16506 | 32 | 4806 |
| 11 | 8279 | 33 | 11439 |
| 12 | 10609 | 35 | 4415 |
| 13 | 29885 | 36 | 5034 |
| 14 | 15861 | 37 | 34678 |
| 15 | 51532 | 38 | 104527 |
| 16 | 14027 | 39 | 25757 |
| 17 | 13533 | 40 | 75627 |
| 18 | 17652 | 41 | 52217 |
| 19 | 12571 | 42 | 13043 |
| 20 | 50431 | 43 | 48719 |
| 21 | 25789 | 44 | 23181 |
| 22 | 15395 |  |  |
|  |  | Total | 1109197 (piece) |

The knowledge of the distribution of new buildings depending on the building type according to the districts is so essential that it helps to plan and organize the logistics activities better, Table 4.15

Table 4.15 The New Buildings Distribution Depending on the Construction Type

|  | the new buildings distribution at districts level depending on the construction type |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| District | Apartment | Compound | Esteraha | Gaz Station | Governmental | Mall | Mosque | Palace | School | Showroom | Tower | villa | wherehouse | Grand Total |
| 1 | 3 |  |  |  |  |  |  |  |  |  |  | 27 |  | 30 |
| 2 | 7 |  |  |  |  |  |  |  | 1 | 1 |  | 64 |  | 73 |
| 3 | 6 |  | 1 |  |  |  |  |  |  | 4 |  | 60 |  | 71 |
| 4 | 3 |  |  |  |  |  |  |  |  | 1 |  | 11 |  | 15 |
| 5 |  |  |  |  | 1 |  |  |  |  | 1 |  | 20 |  | 22 |
| 6 | 3 |  |  |  |  |  |  |  |  | 2 |  | 18 |  | 23 |
| 7 |  |  |  |  |  |  |  | 2 | 1 |  |  | 47 |  | 50 |
| 8 | 1 |  |  |  |  |  |  | 4 |  | 3 |  | 54 | 1 | 63 |
| 9 | 5 |  |  |  |  |  |  |  |  |  |  | 9 |  | 14 |
| 10 | 14 |  |  |  |  |  |  |  | 2 | 5 |  | 16 |  | 37 |
| 11 | 3 |  |  |  |  |  | 1 |  |  | 4 |  | 23 |  | 31 |
| 12 | 2 |  |  |  |  |  |  |  |  | 1 |  | 45 |  | 48 |
| 13 | 5 |  |  |  |  |  | 1 | 9 | 1 | 6 |  | 101 |  | 123 |
| 14 | 6 |  |  |  |  |  |  |  |  |  |  | 58 |  | 64 |
| 15 | 31 |  |  | 2 |  |  |  |  | 2 | 4 |  | 130 |  | 169 |
| 16 | 5 |  |  |  |  |  |  |  |  | 8 |  | 38 |  | 51 |
| 17 | 7 |  |  |  |  |  |  |  |  |  |  | 42 |  | 49 |
| 18 | 3 |  |  |  |  |  |  | 1 |  |  |  | 77 |  | 81 |
| 19 | 6 |  |  |  |  |  |  |  | 1 | 1 |  | 37 |  | 45 |
| 20 | 28 |  | 1 |  |  |  |  |  |  | 2 |  | 145 |  | 176 |
| 21 | 15 |  | 1 |  |  |  | 1 |  |  | 2 |  | 69 |  | 88 |
| 22 | 8 |  |  |  |  |  |  |  |  | 2 |  | 44 |  | 54 |
| 23 | 23 |  | 1 | 1 |  |  |  |  |  | 8 |  | 165 |  | 198 |
| 24 | 16 |  |  | 1 |  |  | 1 |  |  | 5 |  | 141 |  | 164 |
| 25 | 34 |  | 2 |  | 2 |  | 1 |  |  | 3 |  | 211 |  | 253 |
| 26 | 17 |  | 7 |  |  |  |  |  |  | 6 |  | 97 |  | 127 |
| 27 | 24 |  | 1 |  |  | 1 |  |  |  | 6 |  | 259 |  | 291 |
| 28 |  |  |  |  |  |  |  |  |  | 1 |  | 29 |  | 30 |
| 29 | 1 |  | 3 |  |  |  |  |  |  | 3 |  | 3 |  | 10 |
| 30 | 13 |  | 1 | 1 |  |  | 1 |  |  | 1 |  | 41 |  | 58 |
| 31 | 14 | 1 | 12 |  |  |  |  |  |  |  |  | 117 |  | 144 |
| 32 |  |  |  |  |  |  |  |  |  | 1 |  | 23 |  | 24 |
| 33 |  |  | 1 |  |  |  |  |  |  | 1 |  | 57 |  | 59 |
| 35 | 1 |  |  |  |  |  |  |  |  | 1 |  | 17 |  | 19 |
| 36 | 1 |  | 3 |  |  |  |  |  |  | 4 |  | 14 |  | 22 |
| 37 | 6 |  | 1 | 1 |  |  | 3 |  |  | 1 |  | 151 |  | 163 |
| 38 | 20 |  |  |  |  |  | 2 |  |  |  |  | 459 |  | 481 |
| 39 |  |  |  |  |  |  |  |  |  |  |  | 133 |  | 133 |
| 40 | 39 |  | 3 |  |  |  | 4 |  |  | 4 |  | 225 |  | 275 |
| 41 | 19 |  |  |  |  |  |  |  |  | 1 | 5 | 84 |  | 109 |
| 42 | 8 |  | 4 | 1 |  |  |  |  |  |  |  | 34 |  | 47 |
| 43 | 30 |  | 14 |  |  |  |  | 2 | 1 | 1 |  | 120 | 1 | 169 |
| 44 | 2 |  | 1 | 1 |  |  |  |  |  |  |  | 111 |  | 115 |
| Grand Total | 429 | 1 | 57 | 8 | 3 | 1 | 15 | 18 | 9 | 94 | 5 | 3626 | 2 | 4268 |



Figure 4.22 Expected Market Size for Switches for The New Buildings

### 4.4.1 The Increase in Market Size According to Periods

Estimating the market size in the studying area and Taking into account the time to products can be installed is very useful if we can do it, remote sensing gives the ability to have it, if consider tables 4.13 and 4.15 , the results are the products increasing depending on the building stage and time to install it, table 4.16. This is meaning: three months starting from October 2016 (approximately at January 2017) we will need 445604 new switches while after two years, 67173 new switches will need to install it and so on, at this time we don't care for other buildings that Which can be built in the studying area but we can be taking into our account later.

Table 4.16 The Demand For New Products According to Building Stage and Building
Type

|  | Expected market size according to the period (piece) |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| District | Three Month | Six Month | One Year | Two Year | Grand <br> Total |
| 1 | 4261 | 1736 | 1546 | 0 | 7543 |
| 2 | 9102 | 4825 | 3447 | 1197 | 18571 |
| 3 | 9611 | 4632 | 2688 | 771 | 17702 |
| 4 | 1162 | 1320 | 2314 | 0 | 4796 |
| 5 | 3099 | 351 | 2452 | 0 | 5902 |
| 6 | 2905 | 1543 | 1284 | 771 | 6503 |
| 7 | 3873 | 581 | 4648 | 1277 | 10379 |
| 8 | 5387 | 2934 | 4844 | 851 | 14015 |
| 9 | 1162 | 2508 | 1159 | 771 | 5600 |
| 10 | 2130 | 5141 | 6495 | 2740 | 16506 |
| 11 | 1162 | 4307 | 2809 | 0 | 8279 |
| 12 | 7553 | 1352 | 932 | 771 | 10609 |
| 13 | 12989 | 5261 | 10864 | 771 | 29885 |
| 14 | 7746 | 2124 | 5219 | 771 | 15861 |
| 15 | 17394 | 21636 | 10188 | 2314 | 51532 |
| 16 | 6706 | 3282 | 3267 | 771 | 14027 |
| 17 | 6004 | 5019 | 1740 | 771 | 13533 |
| 18 | 9102 | 2902 | 4451 | 1197 | 17652 |
| 19 | 3099 | 3127 | 6345 | 0 | 12571 |
| 20 | 16619 | 15072 | 14882 | 3857 | 50431 |
| 21 | 7746 | 11445 | 5826 | 771 | 25789 |
| 22 | 5774 | 6755 | 2866 | 0 | 15395 |
| 23 | 25585 | 10360 | 12076 | 4628 | 52649 |
| 24 | 17645 | 11120 | 7438 | 5400 | 41603 |
| 25 | 25262 | 25273 | 17203 | 3962 | 71699 |
| 26 | 14987 | 8935 | 5797 | 4628 | 34347 |
| 27 | 30103 | 17761 | 17564 | 5627 | 71055 |
| 28 | 1937 | 3450 | 581 | 0 | 5968 |
| 29 | 545 | 387 | 1620 | 0 | 2552 |
| 30 | 4999 | 3096 | 8156 | 2314 | 18566 |
| 31 | 10460 | 7527 | 14705 | 1543 | 34235 |
| 32 | 194 | 387 | 4225 | 0 | 4806 |
| 33 | 3292 | 2324 | 5823 | 0 | 11439 |
| 35 | 1320 | 1162 | 1933 | 0 | 4415 |
| 36 | 1671 | 2636 | 727 | 0 | 5034 |
| 37 | 16603 | 8885 | 6876 | 2314 | 34678 |
| 38 | 63715 | 26318 | 12952 | 1543 | 104527 |
| 39 | 15880 | 4261 | 5616 | 0 | 25757 |
| 40 | 32585 | 25310 | 14542 | 3190 | 75627 |
| 41 | 9877 | 11164 | 29634 | 1543 | 52217 |
| 42 | 3385 | 4825 | 3290 | 1543 | 13043 |
| 43 | 11924 | 19747 | 9255 | 7793 | 48719 |
| 44 | 9050 | 5807 | 7553 | 771 | 23181 |
| Grand Total | 445604 | 308586 | 287833 | 67173 | 1109197 |



Figure 4.23 the Increase in Market Size

Analysis: This chart shows that the majority of the increase in the market size happen three months later starting in October 2016 while the lowest market size two years later.


Figure 4.24 Increase in Switches and Sockets Three Months Later


Figure 4.25 Increase in Switches and Sockets Six Months Later


Figure 4.26 Increase in Switches and Sockets one Year Later


Figure 4.27 Increase in Switches and Sockets Two Years Later
Maps analysis: Figure 4.24 to 4.27 show the highest expected market size over time, those districts required the planning team and decision maker's to focus on, the dark red reflect the most upper expected switches and socket market size while the dark blue is the lowest potential market size, (district with no colour mean there are no buildings at that time), by using this maps we can determine where they have to take the right action on time to prevent any bottleneck in the future.

### 4.4.2 Cumulative Market Size for Switches and Sockets

Knowing the amount of expected increase in market size will lead us to understand the cumulative increase that may occur in the market, which requires identifying the priorities and preparing the necessary plans for that.

It's essential to know that, so we will have a full idea of the expected total number that we will need according to previous dates, table 4.17, here will have information about The total quantities of new products we will need, example: after two years the cumulative products that the studying area will need is 1109196 new switches.

Table 4.17 Cumulative Market Size According to Building Stage and Building Type.

|  | Expected cumulative market size according to the period (piece) |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| District | Three Month | Six Months cumulative | One year cumulative | Two year cumulative |
| 1 | 4261 | 5997 | 7543 | 7543 |
| 2 | 9102 | 13927 | 17374 | 18571 |
| 3 | 9611 | 14243 | 16931 | 17702 |
| 4 | 1162 | 2482 | 4796 | 4796 |
| 5 | 3099 | 3450 | 5902 | 5902 |
| 6 | 2905 | 4448 | 5731 | 6503 |
| 7 | 3873 | 4454 | 9102 | 10379 |
| 8 | 5387 | 8321 | 13164 | 14016 |
| 9 | 1162 | 3670 | 4828 | 5600 |
| 10 | 2130 | 7271 | 13766 | 16506 |
| 11 | 1162 | 5469 | 8279 | 8279 |
| 12 | 7553 | 8905 | 9838 | 10609 |
| 13 | 12989 | 18250 | 29114 | 29885 |
| 14 | 7747 | 9870 | 15089 | 15861 |
| 15 | 17394 | 39030 | 49218 | 51532 |
| 16 | 6706 | 9989 | 13256 | 14027 |
| 17 | 6004 | 11022 | 12762 | 13533 |
| 18 | 9102 | 12004 | 16455 | 17652 |
| 19 | 3099 | 6226 | 12571 | 12571 |
| 20 | 16619 | 31691 | 46574 | 50431 |
| 21 | 7747 | 19192 | 25018 | 25789 |
| 22 | 5774 | 12529 | 15395 | 15395 |
| 23 | 25585 | 35944 | 48020 | 52649 |
| 24 | 17645 | 28765 | 36203 | 41603 |
| 25 | 25262 | 50535 | 67738 | 71700 |
| 26 | 14987 | 23922 | 29719 | 34347 |
| 27 | 30103 | 47864 | 65429 | 71055 |
| 28 | 1937 | 5387 | 5968 | 5968 |
| 29 | 545 | 932 | 2552 | 2552 |
| 30 | 4999 | 8096 | 16252 | 18566 |
| 31 | 10460 | 17987 | 32692 | 34235 |


| District | Three <br> Month | Six Months <br> cumulative | One year <br> cumulative | Two year <br> cumulative |
| :---: | :---: | :---: | :---: | :---: |
| 32 | 194 | 581 | 4806 | 4806 |
| 33 | 3292 | 5616 | 11439 | 11439 |
| 35 | 1320 | 2482 | 4415 | 4415 |
| 36 | 1671 | 4307 | 5034 | 5034 |
| 37 | 16603 | 25488 | 32364 | 34678 |
| 38 | 63715 | 90032 | 102985 | 104527 |
| 39 | 15880 | 20141 | 25757 | 25757 |
| 40 | 32585 | 57895 | 72437 | 75627 |
| 41 | 9877 | 21040 | 50675 | 52217 |
| 42 | 3386 | 8211 | 11501 | 13043 |
| 43 | 11924 | 31671 | 40926 | 48719 |
| 44 | 9050 | 14857 | 22410 | 23181 |
| Grand <br> Total | 445604 | 754191 | 1042023 | 1109197 |



Figure 4.28 The Increase in Market Size

Analysis: This chart shows that the highest cumulative market size happened two years later starting in October 2016. While the lowest cumulative market size three months later starting in October 2016.


Figure 4.29 Expected Cumulative Switches Market Size (Three Months Later)


Figure 4.30 Expected Cumulative Switches Market Size (Six Month Later)


Figure 4.31 Cumulative Switches Market Size (One Year Later)


Figure 4.32 Cumulative Switches Market Size (Two Years Later)

Maps analysis: Figure 4.29 to 4.32 show the highest expected cumulative market size overtime to tell us the where the districts with the highest potential which is dark red, while the dark blue is the lowest potential market size, by using this maps we can determine where they have to take the right action on time to prevent any bottleneck in the future.

### 4.5 Determine the Best Site Location

Many studies carried out to select the best sites for new activities that can provide suitable conditions and meet the specific requirements. The select new location should have criteria, which vary according to the business situation. So far, the population density in these studies has not been given what should have been given to it, but in this research using remote sensing to analyses the high-resolution satellite images to provide us the necessary information that can estimate the population in the studying area (as shown in this study).

The question of choosing the best location for the establishment of facilities and services is a matter that depends on the requirements and needs of the stakeholders and varies according to the purpose. Selection becomes more complicated depending on the number of conditions required and vice versa. The significant development that is taking place in Riyadh city especially the study area requires us to develop the methods of study and research commensurate with the massive increase in urban and human activities taking into account the need for new Malls. Several factors can influence selection processes, including the following:

### 4.5.1 Land Value

The land value determined in many ways, including:

- The average value of land sales and purchases in the area, obtained from the Ministry of Justice website.
- Distance to the main roads where the land value rises.
- Distance to the city center which the prices downtown are highest.

Figure 4.33 shows the land price (value) calcification according to the previous three points; the Red color is the highest price segmentation while the dark blue is the lowest segmentation. The highest segmentation located beside the main two roads (Northern road and King Abdul Al Aziz Road) and between King Abdul Al Aziz Road and King Fahed Road.


Figure 4.33 Land Price Classification

### 4.5.2 Population Factor

This factor is most important because people who will do the business, purchase and visited retail sites, therefore, a new way of finding out the population density has to be found. So, as we have seen, satellite imagery has helped us detect new and old buildings and calculate the population in areas better than other methods, which depend on tables or governmental data.

We studied the population in this research and calculated the estimation at districts level to know which places people live? Here the area with high population density will be attractive places for investment and establish new activities. By using ArcGIS the population data converted to population density, Figure 4.34 , will guide the process to target the areas with high population density and give it top priorities. In figure 4.34 the dark blue reflects the high density which has the number (1) in the legend. Also, the population who lives in the buildings which scanned in January 2016 will be taking into account.


Figure 4.34 Population Density

### 4.5.3 Land Use

Land use map created by using the Spot 6 satellite image (October 2016) to distinguish the lands to take it in our account and avoid it.
This classification includes, Figure 4.35:
a) Garden
b) Mall
c) Palaces
d) Unknown area
e) Industry


Figure 4.35 Land Use.

### 4.5.4 Distance To The Road Network

There is a wide road network in the study area, even within the small districts, which is a modern network and is continually being developed. Which will help the search and selection of the best location? So as not to put many phenomena on the maps and to avoid set all roads network have placed only the main road network, figure 4.36.


Figure 4.36 Main Roads.

### 4.5.5 Topographic Factors

The slope in the study area is very low, and the maximum value is $20 \%$. The areas which very slopping is outside the study area. This factor will ignore because it does not affect the selection of sites.


Figure 4.37 The Slop in The Study Area

### 4.5.6 Climate Factors

Mostly this factor has no significant effect because there is no difference between the temperatures in the study area. Where the temperature in the summer between 40 to 44 degrees.

### 4.5.7 Competitors (Distance from Current Markets)

When you create an activity that requires careful study of the market. So that the proposed activity should be away from competitors in a way that provides the right investment away from conflicts of interest with them, taking the current competitors and creating a map showing the distance to the competitors must be taken into account which is beneficial for investors.

The choice of the best site based on the required conditions will be more accurate by using remote sensing and Geomarketing, which provides the possibility of dealing with the places through equations given to the computer according to the required factors to offer a number of suggested sites that are compatible to the criteria for the establishment of sites required.

Afield survey for the Retail Stores in the study area was conducted to identify the competitors. The search included the branded stores and did not include the hundreds of small stores in the districts, this search was done in two ways:
a) Field visit
b) Stores websites

The information that recorded, only retail market name and its location
The following stores recorded in the study area, table 4.18:
While Figure (4.38) show the map of retails location in the study area

Table 4.18 the Retails in the Study Area

| Retail brand | number of retails |
| :---: | :---: |
| Aljazeerah | 1 |
| Al-Sadhan | 1 |
| Banda | 19 |
| Carrefour | 2 |
| Danube | 4 |
| Otahaim | 7 |
| Tamimi | 8 |
| Total Retails | 42 |



Figure 4.38 Retails (Competitors) Location in the Study Area

### 4.5.8 Land Area

This stage is a complex process, and it depends on dropping the sites of buildings, roads, and land use, by using ArcGIS the results are the lands which has not any activities. Here only the land which has enough area to build the retail building will consider, in Riyadh city the average area for the Malls is 25000 Meter square for that only the land with an area more than 25000 Meter square will take it into account as in Figure 4.39.


Figure 4.39 Empty Land in the Study Area

### 4.5.9 Select the Best Location For New Retail

After processing and preparing the required layers, we will be able to start by selecting the suitable sites for that.

We will weight the following layers:
a) Land price weight is $15 \%$
b) Distance to current market $25 \%$
c) Population density weight $60 \%$

We will use raster calculation to calculate the final results as in Figure 4.40. In this map, light blue (number 1) reflect the high priority.


Figure 4.40 Raster Calculation

The results will be a raster, by converting it to layer and by taking the intersection with empty lands as in Figure 4.39, we take into account the flowing at this stage:
a) Distance to the main road is less than 100 meters (According to regulations, also site need a car parking)
b) Keep only the priority number one and two

The results are the best site location for new retail by considering the previous criteria as in Figure 4.41.


Figure 4.41 Best Site Locations

## Conclusion:

It found that the use of remote sensing and GIS provides us with the necessary assistance to understand and improve our planning, increase the benefits of using it in industry. The high-resolution satellite images have provided what can be said as a qualitative leap in the search and put the science in the service of the community and the business world.

## CHAPTER 5

## CONCLUSION AND RECOMMENDATIONS FOR FUTURE WORK

### 5.1. Conclusion

The use of remote sensing in marketing, demand sales planning and develop a methodology to select the best location to establish new service centers or commercial center by developing the existing methods is a critical question that explored in this study. Different types of data have been used, it was processed and used in a manner appropriate to this study, also variables checked and used to determine its effect on the study, which supported the purpose.

1) Two different satellite images were used to calculate the building's growth and analyze the population increasing in micro-geographic areas using geospatial analyses. The locations of the new buildings have been linked to the variables studied in order to give a clearer idea for how spatial information can be used to identify growth at micro-geographic so that decision-makers can quickly access these markets.
2) GIS was used to store, manage and analyze data to determine the population in the study area and used this information an indicator for determining milk consumption at the micro-geographic level which could be used by FMCG distributors to improve demand planning, also another product was studied to calculate market size as a building material product.
3) The process of studying the best location for the establishment of a commercial market or any new service took a new form by calculating the population density reached as a result of determined the population in each small area as a result of the analysis of satellite images and statistics authority information, this will enable to add a new dimension to the new mechanism to determine the new location.
Thus, using remote sensing and GIS data for Geomarketing able to give an insight picture of urban growth and its concomitant increase in population and determine the best location to create a new market.

### 5.2. Recommendation

This research is based on the manual digitizing for the boundaries and components of the study area, including new buildings or under-construction buildings. It may take much time, the output preparation depends on equations and processes within the GIS sequence and logical but requires preparation for each. We can improve this research by enhancing the automatic extraction of the new buildings and classify it to its building type and stage. This improvement will allow the full usage of this new methodology and improve the visibility studies for the targeting products that we need to prepare to forecast and determine the market size for it. Besides, it is possible to increase efficiency through the development of scripts within the GIS program, which includes tabs for study parameters. This development in the program will allow the user to obtain the results immediately and shorten all the operations that will be done by the user and also through the model builder under GIS to shorten some processes by incorporating them in the one environment.

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