

M.Sc. Engg. Thesis

# **Detecting Geo-Financial Solvency Using Social and Web Media Networks**

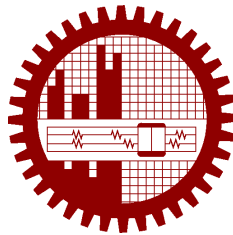
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Submitted to

**Department of Computer Science and Engineering**  
**Bangladesh University of Engineering and Technology**  
Dhaka, Bangladesh

in partial fulfillment of the requirements for the degree of  
Master of Science in Computer Science and Engineering

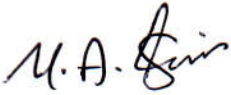
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*I dedicate this thesis to my late parents.....*

## Candidate's Declaration

I, do, hereby, certify that the work presented in this thesis, titled, "**Detecting Geo-Financial Solvency Using Social and Web Media Networks**", is the outcome of the investigation and research carried out by me under the supervision of Dr. Mahmuda Naznin, Professor, Department of CSE, BUET.

I also declare that neither this thesis nor any part thereof has been submitted anywhere else for the award of any degree, diploma or other qualifications.



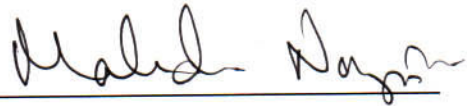
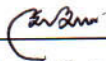
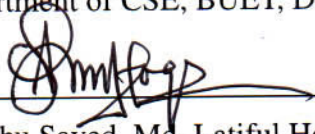


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The thesis titled “**DETECTING GEO-FINANCIAL SOLVENCY USING SOCIAL AND WEB MEDIA NETWORKS**”, submitted by Md. Aflatul Sagir, Student ID 0413052027, Session April 2013, to the Department of Computer Science and Engineering, Bangladesh University of Engineering and Technology, has been accepted as satisfactory in partial fulfilment of the requirements for the degree of Master of Science in Computer Science and Engineering and approved as to its style and contents on September 28, 2019.

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Dhaka  
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2019

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## **Abstract**

Financial solvency analysis of a geographic area is a very challenging task because it costs lots of money and a big volume of data handling is involved. Every country in the world checks its financial status, which is known as the economic census. There are some techniques to measure financial status like using of some questionnaires where data is collected from field level manually, or using financial statements. These techniques are old and these use data collection manually which increase the cost, money and effort. The use of social and web media network has been increased a lot. As a part of socio-activities people share their personal information like occupation, home town, events, plan with his/her friends on social media. On the other hand, web media or online access of information provide information for economic establishments (e.g. public institutions, private companies, etc). There are many research work like cultural boundary detection, community detection, food consumption statistics, sentiment analysis, and personality predicting, have been done using social media. It is very important to have economic census for a country. In this study, an automated economic zone finding framework has been designed using social-sensing and web mining. Here, data has been collected using social and web media. By using the proposed framework, the financial status of eight major divisional cities of Bangladesh has been successfully detected. And the last of all, the city's ranks have been measured according to their financial solvency. This technique is automated, so it requires less time and it costs less money than other traditional surveyed methods. According to best of our knowledge, this is the first technique in our country that uses the web and social media to verify the financial status of different cities and this technique performs better than existing surveyed approaches.

## Abbreviations

<b>AHE</b>	<i>Activities of households as employers</i>
<b>AER</b>	<i>Arts, entertainment and recreation</i>
<b>AEOB</b>	<i>Activities of extraterritorial organizations and bodies</i>
<b>AFF</b>	<i>Agriculture, forestry and fishing</i>
<b>AFSA</b>	<i>Accommodation and food service activities</i>
<b>ASSA</b>	<i>Administrative and support service activities</i>
<b>BBS</b>	<i>Bangladesh Bureau of Statistics</i>
<b>Bus.</b>	<i>Business</i>
<b>BP</b>	<i>Buying or Purchasing</i>
<b>Const.</b>	<i>Construction</i>
<b>DFA</b>	<i>Daily Financial Activities</i>
<b>ED</b>	<i>Eating and Drinking</i>
<b>Edu.</b>	<i>Education</i>
<b>EE</b>	<i>Economic Establishment</i>
<b>EGSA</b>	<i>Electricity, gas, steam, and air conditioning supply</i>
<b>Emp.</b>	<i>Employment</i>
<b>Ent.</b>	<i>Entertainment</i>
<b>FIA</b>	<i>Financial and insurance activities</i>
<b>GDP</b>	<i>Gross domestic product</i>
<b>HHSWA</b>	<i>Human health and social work activities</i>
<b>IC</b>	<i>Information and Communication</i>
<b>LR</b>	<i>Linear Regression</i>
<b>MAQ</b>	<i>Mining and quarrying</i>
<b>MFG</b>	<i>Manufacturing</i>
<b>NIPA</b>	<i>National income and commodity accounting</i>
<b>NR</b>	<i>Non Residents</i>
<b>Occ.</b>	<i>Occupation</i>
<b>OSA</b>	<i>Other service activities</i>
<b>OO</b>	<i>Other Occupation</i>

<b>PAD</b>	<i>Public administration and defense; compulsory social security</i>
<b>PPI</b>	<i>Producer price index</i>
<b>PSTA</b>	<i>Professional, scientific and technical activities</i>
<b>REA</b>	<i>Real estate activities</i>
<b>SMO</b>	<i>Sequential minimal optimization</i>
<b>SSM</b>	<i>Solvency Sensing Method</i>
<b>SVM</b>	<i>Support Vector Machine</i>
<b>TAS</b>	<i>Transportation and Storage</i>
<b>TNU</b>	<i>Total Number of Users</i>
<b>TNP</b>	<i>Total Number of Posts</i>
<b>TNNR</b>	<i>Total Number of Non-Residents</i>
<b>VP</b>	<i>Voted Perceptron</i>
<b>WRT</b>	<i>Wholesale and retail trade; repair of motor vehicles and motorcycles</i>
<b>WSWRA</b>	<i>Water supply; sewerage, waste management and remediation activities</i>



# Chapter 1

## Introduction

*Geo-Financial Solvency* measurement verifies the financial status of the people residing in a particular geographic area. Due to its importance, every country in the world performs checking of its financial status of the country's different locations in a certain period. It is an important activity with the aim of improving the quality of life based on economic statistics. Businesses use *Economic Census* data to help to make decisions about where to expand the business, where to set up a new business, how much to produce etc. To compare the performance to other businesses in the industry or community *Economic Census* plays a vital role. Local communities use *Economic Census's* results to attract new businesses, assess the economic health of their localities, to understand the characteristics of their business base, and to compare their community to other geographical areas. Individuals can use Economic Census results to identify emerging job markets and growing industries.

### 1.1 Motivation

Different types of tools and techniques are used to measure the financial health status of any geographic location. One of the most common techniques is to use of some questionnaire [1, 2], about the type of economic activities of the business units, type of the manufactured products or services, type of ownership, equity, sales and accounting practices. To accomplish such a process, information is collected from the field level (end users) manually.

In case of an organization, the financial solvency is verified from the balance sheet received from the financial statements of that organization [3, 4]. In that case, the financial solvency of the institution is determined using some accounting theories [5, 6], based on the total assets and total liabilities of that organization. Those techniques cost huge money and require huge time.

Some international organizations [7–10] use their own formulas and tools to measure the

financial conditions of different countries. Very recently Support Vector Machine (SVM) technique has been used to analyze data got from the third party [11]. These types of techniques reduce processing time after data is collected. However, the total time for data collection is not reduced.

Financial solvency measurement is a challenging task because:

- It costs lots of money.
- A big volume of data handling is involved.
- A large volume of man-power is required.

This can be handled if the following steps are done.

- Data collection from social and web media can reduced cost of money.
- Use of an automated tools can save processing time and a few people to operate the software.

To minimize the financial cost of conducting a census, an automated framework has been proposed in this thesis, where data has been collected from social and web media. To reduce data processing time, machine learning has been used in this framework.



Figure 1.1: The process of automatic geo-financial solvency detection using SSM.

The figure 1.1 shows the process of geo-financial solvency detection using proposed method what we call *Social Sensing Machine (SSM)*. In the first step, the data is downloaded from social and web media. In the second step, the downloaded data is entered into SSM for analysis. The last part of this figure compares the score of the targeted areas and ranks them according to the score value.

## 1.2 Challenges

In this section, we talk about some challenges in developing automated solvency framework. These are stated as follows:

1. *Data Extraction:* The social media data is too noisy. Therefore, the financial data extraction from numerous non-related data in social media can be tricky.
2. *Structured Data:* Making unstructured data into structured format is one of most difficult task in data processing. The success of this method depends on the structured data.
3. *Indicator Selection:* Choosing the right indicators for the dimensions and the proper weight distribution among the indicators is one of the most difficult tasks. In this case, we followed the international methods.
4. *Computation Method:* Finding a suitable score calculation method that will reduce the computation time in the total working process.

## 1.3 Applications and Advantages

There are many applications and advantages of this technique. Some applications of this method are pointed below:

1. **Economic Census Measurement:** The proposed framework can be used for area wise economic census measurement in national or international level with a very low cost and less time.
2. **Financial Status Measurement:** The proposed technique can be used to measure person's financial status by analysing daily financial activities through social media.
3. **Public Opinion Verification:** To verify the public opinion to start a new business in a city or in a country, this method can be helpful.
4. **Measurement of Product Demand:** Using SSM, the public demand of certain types of products can be measured.
5. **Dominant Factor Finding:** This method can help to find out the most dominant factors that makes an area most or less solvent. (e.g., manufacturing, non-residents etc.)

There are many advantages in the proposed method compare to traditional surveyed methods, some of them are describes in Table 1.1.

<b>Criteria</b>	<b>Traditional Method</b>	<b>SSM</b>
Cost	Huge money investment needed	Less money investment needed
Processing time	Need more data processing time	Need less data processing time
People	More people needed to engaged	As it fully automated software based, need a few people to operate.

Table 1.1: Advantages of the proposed method compared to conventional survey methods

## 1.4 Scope of The Work

This research has been designed for eight divisional cities of Bangladesh (1) Barishal (2) Chattogram, (3) Dhaka, (4) Khulna, (5) Mymensingh, (6) Rajshahi, (7) Rangpur, and (8) Sylhet.

The economic establishment data has been collected by using three websites and a search engine such as (1) <https://www.bangladeshyellowpages.com>, (2) <https://bangladesh.gov.bd>, (3) <https://en.wikipedia.org>, (4) <https://www.google.com>. This research has covered all economic units- permanent and temporary establishments, household based economic activities which are available at web media. It should be noted that some establishments engaged illegal activities like bar, smuggling, gambling, beggary, prostitution, etc. are kept out of this research.

This research uses social media data set which has collected from Facebook. In case of selecting user posts, we ignore the posts which are not written in English and the posts which are irrelevant to our research topic.

## 1.5 Contributions

The major contributions in this thesis are stated as follows.

1. *Area Ranking*: To the best of our knowledge, we have ranked eight divisional cities of Bangladesh according to their financial status using social and web media data.
2. *Detecting Dominant Factors*: We have been able to figure out the most influential factors those affect a region's financial stability. For example, the "Non-Resident" indicator enables the Sylhet city to gain better rank in the occupation and DFA described later.

3. *Financial Data Extraction:* Financial activities have been extracted from numerous non-related plain texts of a post.
4. **New Dimension Addition:** To verify financial solvency using social and web media data, new financial impact creating features have been created.
5. **Use of Modern Tools:** For data processing and mining, *R*, machine learning tools *WEKA* have been used for geo-financial analysis.
6. *Dictionary Construction:* Four dictionaries which contain 10000 words for solvency analysis have been built.
7. **Auto Framework Build-up:** A solvency sensing framework has been built up which can automatically detect geo-financial solvency.

## 1.6 Organization of the Thesis

The rest of the thesis is organized as follows. Chapter 2 discusses the existing work related to geo-financial survey methods and challenges. It highlights the scope of the works based on the limitation found in the state of the art methods. This chapter discusses benchmark solutions to measure financial stability by some standard organization or other countries. Chapter 3 explains the detail framework of our proposed method called *Solvency Sensing Method (SSM)*. The assumptions of this research is also discussed here. The mathematical models used for the score and rank calculation are explained at the end of this chapter. Results and analysis are discussed in Chapter 4. The tools and techniques, data-set are given in this chapter. The comparative result analysis and limitations of this research are also discussed here. The summary of the results are shown in the end of this chapter. Finally, in Chapter 5, we conclude the thesis by summarizing the proposed work and highlighting some future scopes.

# Chapter 2

## Related Work

There are many solutions which have been proposed to measure financial stability of an organization or a country. We divide them into three categories based on - (1) Financial Theorem (2) Census, and (3) Organization's Own Formula. Here, we also discuss some work done using social media network.

### 2.1 Financial Theorem Based Methodology

The financial theorem based method uses different financial theorems for analysis organizational financial stability. It uses different types of dimension and indicators. The most common types of dimension: (1) Cash solvency (2) Budgetary Solvency (3) Long run Solvency (4) Service-level solvency.

(1) *Cash solvency* [3] describes a government's (or organization's) ability to generate sufficient financial resources to pay it's current liabilities. The working capital to expenses ratio is a indicator of cash solvency that measures the liquidity. The corresponding equation is as follows:

$$(i) \text{ The Working Capital to Expenses Ratio} = \frac{\text{Current Assets} - \text{Current Liabilities}}{\text{Expenses}}$$

(2) *Budgetary solvency* [4] demonstrates a government's financial ability to maintain current or desired service levels within a period by sufficiently funding operating expenses. There are three indicators to measure Budgetary solvency: (1). the continuing services ratio, (2). the fund balance ratio, and (3). the operating surplus ratio. *Continuing services ratio* is an indicator that examines unrestricted net assets as a percentage of expenses for all funds government-wide.

$$(ii) \text{ The Continuing Services Ratio} = \frac{\text{Unrestricted Net Assets}}{\text{Total Expenses}}$$

The *fund balance ratio* is an indicator that compares unrestricted general fund balance to general fund expenditures, reflecting the operating savings that a government has accumulated relative to the government's operating expenditures for that fiscal year.

$$(iii) \text{ Fund Balance Ratio} = \frac{\text{Unrestricted General Fund Fund Balance}}{\text{General Fund Expenditures}}$$

The *operating surplus (deficit) ratio* shows the general fund operating surplus or deficit as a percentage of total operating expenses. This indicator reflects the difference between revenues and expenditures in completed fiscal years. Generally, a larger ratio reflects excess operating revenues over expenditures. Therefore, a ratio greater than zero and an increasing trend are considered favorable.

$$(iv) \text{ Operating deficit ratio} = \frac{\text{General Fund Surplus or Deficit}}{\text{Net Operating Expenditures}}$$

(3) *Long-run solvency* [5] assesses the availability of future resources to pay for existing long-term obligations. The *net worth ratio* is an indicator that measures restricted and unrestricted net assets as a percentage of total asset. It is a measure of the net worth of a government and signifies the government's ability to pay off existing long-term liabilities. A larger ratio indicates more accessible resources for the government and therefore, an increasing trend and higher ratio are considered favorable.

$$(v) \text{ Net Worth Ratio} = \frac{\text{Restricted and Unrestricted Net Assets}}{\text{Total Assets}}$$

The *debt service expenditure ratio* is the percentage of debt service expenditures out of total governmental fund expenditures. The indicator can be used to evaluate service flexibility by determining the amount of expenditure committed to the annual debt service. With a higher debt service expenditure ratio, a larger portion of expenditures is being allocated to paying for debt issued by the government rather than regular government services. A lower ratio and decreasing trend are considered favorable.

$$(vi) \text{ Debt Service Expenditure Ratio} = \frac{\text{Debt Service Expenditure}}{\text{Total Expenditures}}$$

(4) *Service-level solvency* [6] which reflects a government's ability to maintain services at the quality and level required to ensure the safety and welfare of citizens and to meet their expectations and desires. Expenses per capita divide the total expenses of the primary government, which include governmental activities and business-type activities, by population. Higher expenses per capita reveal a more expensive government and lower solvency to sustain that expense level. Therefore, a lower ratio than the average of all of the cities and decreasing trend are considered favorable.

$$(vii) \text{ Expenses per Capita} = \frac{\text{Total Primary Government Expenses}}{\text{Population}}$$

*Liabilities per capita* divides the total liabilities of the primary government by population and represents the government's relative indebtedness with regard to future taxpayers. A lower ratio and decreasing trend are considered favourable. Since higher than average or increasing liabilities can be a cause for concern, a lower ratio than the average of all of the cities and decreasing trend are considered favorable.

Taxes and fees per capita divides all taxes and charges for services for primary government activities by population. Higher taxes and fees per capita reflect a higher tax burden for residents and a lesser ability of the government to raise taxes or fees further to sustain current service levels. As such, a lower than average ratio and decreasing trend are considered favorable.

$$(ix) \text{ Taxes and Fees per Capita} = \frac{\text{Total Primary Government Taxes and Charges for Services}}{\text{Population}}$$

## 2.2 Economic Census Based Methodology

There are many approaches are used to measure economic census. We describe some of them. Bangladesh Bureau of Statistics (BBS) conducted economic census in 2013 [1] to measure financial condition of Bangladesh by calculating the number of economic households, establishments and employee's work on those organizations. They generate some questionnaire about economic establishments, households and employees. The data was collected manually by filling up the form. These census costs more than 500 crore and total processing and outcome time was more than 1 year.

Central Statistics Office (CSO) [2], conducted economic census of India 2013, they includes various agricultural and non agricultural activities, they divides all economic activities into 23 categories, outcome of that census was (1) total number of establishments counted is about 58.5 million, (2) 59.9% of the establishments belong to rural areas, (3) 20.5%



of the establishments operate from outside household without fixed structure, (4) total number of persons employed is about 128 million with rural share of 51.9%, (5) percentage of female workers is 25.6%, (6) growth rate in number of establishments over fifth EC (2005) is 41.7%, (7) growth rate in total employment over fifth EC (2005) is 34.4%. It was also very costly and time consuming.

## 2.3 Organization's Own Formula Based Methodology

The international organizations like World Bank, UNDP, World Economic Forum uses their own formulas and tools to measure the financial condition of different countries in the world [7–10].

The Civic Federation [3, 5], compared the relative fiscal performance of the city of Chicago to 12 other major U.S cities using nine financial indicators compiled from audited financial statements from FY2007 through FY2011. Primarily, rank was counted for each indicator based on five year financial change of the cities. The final rank is counted as the average of the indicators' ranks. The city which has the lowest total average indicator rank marked as Rank 1 and so on. In their report, Columbus (Ohio) scored Rank 1, Chicago ranked 11 and Detroit ranked 13.

The World Economic Forum [8], analysis the economic development of 140 countries in the world. They divided the development into three stages. (1) Factor driven (35 countries), (2) Efficiency driven (31 economies) and (3) Innovation driven (38 economies). *Stages* were calculated by three sub indexes: Basic requirements, Efficiency enhancement and Sophistication factors. There is no uniform and universal rule for measuring economic census. Most of the cases, data is collected manually which is costly and time consuming. It should be noted that, the results depend on the method what is adopted.

## 2.4 Social Media

Silva et al. [12] describes a technique of identifying people's cultural boundaries and similarities at different scales based on analysis of Foursquare check-ins. Firstly, they mapped food and drink related check-ins that is extracted from Foursquare into user's cultural preferences. Secondly, they identified specific personal preferences, such as the taste of a certain type of food or drink, e.g., pizza, coffee, as well as temporal habits, such as the time and day of the week when an individual visiting a restaurant. Thirdly, they analyzed this data to measure the cultural distance between two countries, cities or regions. Then they applied a simple clustering technique (k-means), using this cultural

distance measure, to draw cultural boundaries across countries, cities or regions.

Yang et al. [13] developed a model called CESNA (Communities from Edge Structure and Node Attributes) for detecting overlapping communities in networks with node attributes. In which *Adjacency Matrix* and values of nodes in a graph network were generated from *Node Community Memberships*. Then by adopting a *block coordinate ascent approach* communities were identified.

Qi et al. [14] presented a technique of community detection which has been formulated by integrating contents of the edges of graph network along with graph structure through *matrix factorization* to derive a high quality latent vector representation for the edges based on an analysis of the link matrix. That representation of the latent vectors is used to extract communities by applying k- means clustering. known as EIMF (Edge Induced Matrix Factorization).

Hoang et al. [15] extracts sentiment from micro event on twitter. They at first found micro events about transportation system of Singapore, then analyzed the data, and found out the sentiment of transport and road communication system development.

Social media data can play an important role to find the financial status of the people of particular geographic locations,.

# Chapter 3

## Solvency Sensing Method

The main goal of this research is to provide geo-financial ranking. As a sample, this ranking has been done for divisional cities of Bangladesh according to the financial status by analyzing social media and web media networks. To achieve this goal, we use three dimensions with twenty eight indicators. Then, the score value of dimensions and indicators has been measured on a scale of 1-100. The city with the highest score earned to get the first position in the ranking. This research has been conducted on the data of eight divisional cities of Bangladesh- Barishal, Chittagong, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur, and Sylhet.

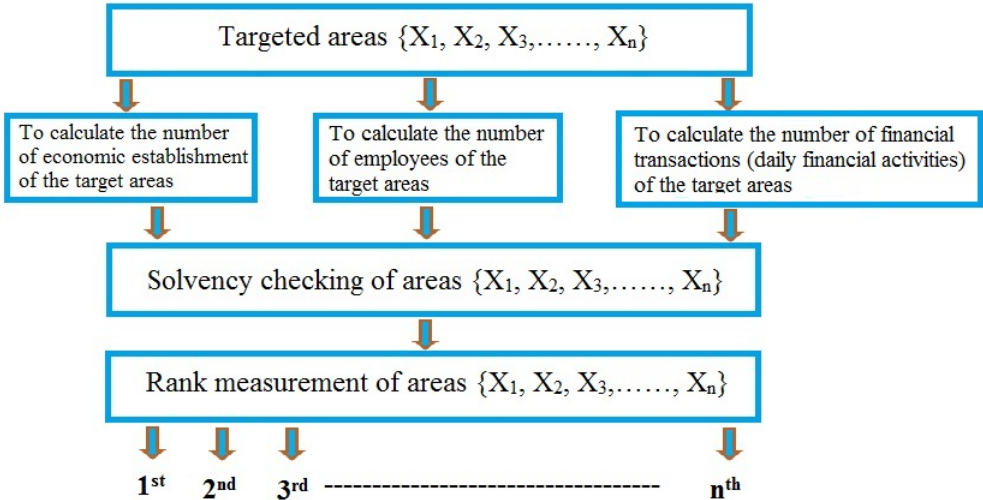


Figure 3.1: Financial Solvency measuring at researcher’s view.

The figure 3.1 depicts the number of economic establishments from web media, information on employment, and different financial events from social media. The economic establishments are considered here garments, textile, bank, transport, airlines and any kind of institutions w have money transactions. The number of employees has been

computed from the Facebook user's profile. The events related to financing detected from the user's posts. Scores have been calculated based on the number of establishments, employees, and events in a city. Lastly, the rank has been computed based on the score value.

### 3.1 Assumptions

There are some of our research described as follows.

1. *The area which has more economic establishment is more solvent.* We consider *Economic Establishment* as an asset. We think that the area which has more economic establishments, has more assets.
2. *The area which has more employed persons is more solvent area.* More employed person indicates the geographic area's more economic activities which results into solvency.
3. *The area which has more daily transaction indicates as the area with better financial status.* For example, if the citizens of an area have better purchasing capabilities than other cities, it is considered as more solvent area.

### 3.2 Detail Framework

Figure 3.2 describes the detailed framework of our proposed method Social Sensing Machine or SSM. There are five components in this framework. The first part involves in data collection from the social and web media, the second part involves in data cleaning, the third part involves in preparing the data suitable for score calculation, the fourth part involves in score computation for each dimension and indicator for each city, and the last part computes ranking of the areas according to the achieved score value.

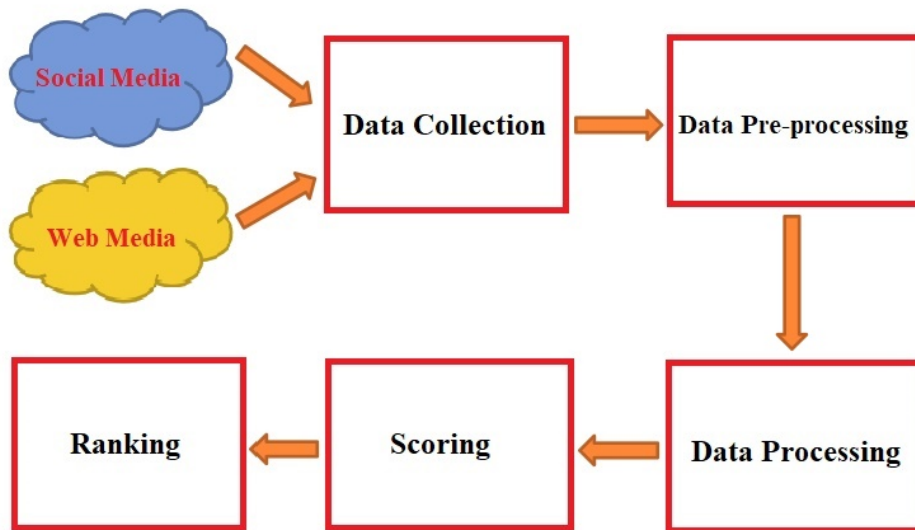


Figure 3.2: Solvency detection framework.

Therefore, our research has been conducted using three economic dimensions and their 28 indicators. The dimensions are as follows:

1. Economical Establishments (*EE*)
2. Occupation (*Occ.*)
3. Daily Financial Activities (*DFA*)

Figure 3.3 describes a sample scenario of area ranking. In the first part, two types of data (Economic Establishment related data and the social media data or user posts) are used as input. In the second part, the data is pre-processed for score calculation. The score is computed over some dimensions based on the indicator data. In the last part, ranked cities have been reported based on the score.

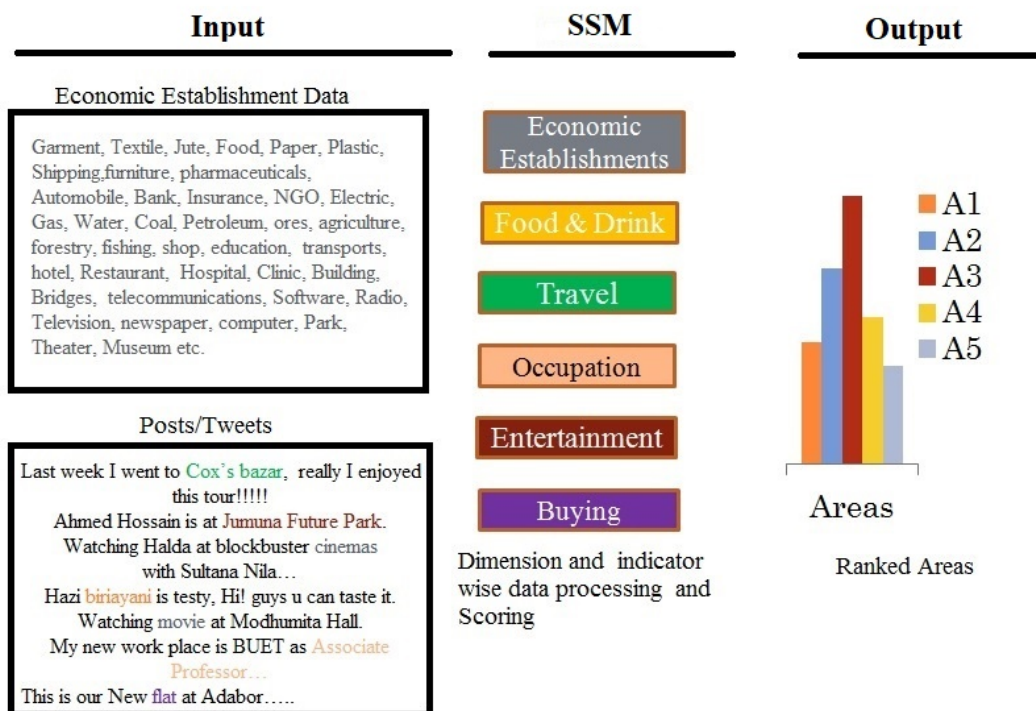


Figure 3.3: A sample scenario of geo-financial ranking using SSM.

### 3.3 Economic Establishment (EE)

An economic unit operates in a permanent or temporary place. Examples: hotel, restaurant, shops, ships, textile mills, banks, insurance etc. According to International Standard (ISIC), all economic activities are classified into 21 broad categories [16]. Those categories have been considered as indicators in this research. The objective is to count the number of economic establishment number of eight cities. The indicators of this dimension are describes in Table 3.1.

<b><i>Dimension</i></b>	<b><i>Indicators</i></b>
<b>Economic Establishment</b>	<ol style="list-style-type: none"> <li>1. Agriculture, forestry and fishing (AFF)</li> <li>2. Mining and quarrying (MAQ)</li> <li>3. Manufacturing (MFG)</li> <li>4. Electricity, gas, steam &amp; air conditioning supply (EGSA)</li> <li>5. Water supply; sewerage, waste management and remediation activities (WSWRA)</li> <li>6. Construction (Const.)</li> <li>7. Wholesale and retail trade (WRT)</li> <li>8. Transportation and storage (TAS)</li> <li>9. Accommodation and food service activities (AFSA)</li> <li>10. Information &amp; Communication (IC)</li> <li>11. Financial and insurance activities (FIA)</li> <li>12. Real estate activities (REA)</li> <li>13. Professional, scientific and technical activities (PSTA)</li> <li>14. Administrative and support service activities (ASSA)</li> <li>15. Public administration and defense; compulsory social security (PAD)</li> <li>16. Education (Edu.)</li> <li>17. Human health and social work activities (HHSWA)</li> <li>18. Arts, entertainment and recreation (AER)</li> <li>19. Activities of households as employers; undifferentiated goods- and services-producing activities of households for own use (AHE)</li> <li>20. Activities of extraterritorial organizations and bodies (AEOB)</li> <li>21. Education (Edu.)</li> </ol>

Table 3.1: Defining indicators of *EE* dimension.

A sample scenario of city ranking at *EE* dimension has been given in Figure 3.4.

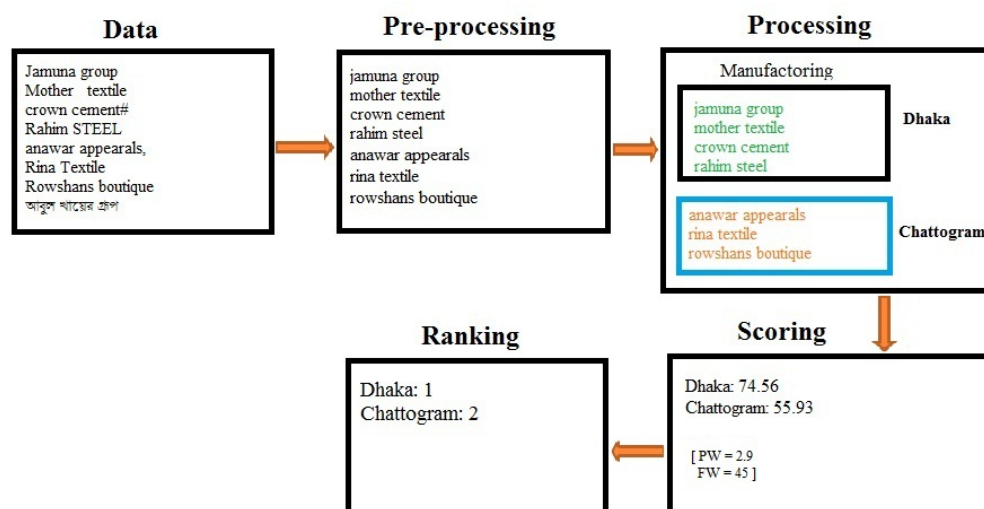


Figure 3.4: A sample scenario of geo-financial ranking at *EE* dimension.

### 3.3.1 Agriculture, forestry and fishing

This indicator includes the activities comprise the natural resources of animal and plants, raising and breeding of animals, comprising the activities of growing of crops, animal and animal product, harvesting of timber and other plants. We collect the firm or company name that performs the activities like:

1. Animal and crop production, and hunting related service activities
2. Forestry
3. Fishing and aquaculture

In the economic census in 2013, the agriculture sector was not included. In this research, we include this important sector. The organizations types have been given in Appendix B.1.

### 3.3.2 Mining and Quarrying

This indicator includes the extraction of minerals occurring naturally as gases (natural gas), liquids (petroleum) or solids (coal and ores). Extraction can be achieved by different methods such as seabed mining, underground or surface mining etc. This indicator also includes supplementary activities for the preparation of crude materials for marketing, for example, grinding, crushing, drying, cleaning, sorting, accumulation of solid fuels, and natural liquid gas.

We collect the firms names that engaged in:



1. Mining of coal and lignite
2. Extraction of crude petroleum and natural gas
3. Mining of metal ores
4. Quarrying of stone, sand and clay
5. Mining support service activities

The organizations types has given in Appendix [B.2](#).

### **3.3.3 Manufacturing**

This indicator includes the physical or chemical conversion of materials, substances or ingredients into new products, although it cannot be used as a single universal standard for defining production. Transformed materials, substances or ingredients are raw materials that are used as products in agriculture, forestry, fishing, excavation or mining and other manufacturing activities. Substantial changes, renovations or product remodeling are generally considered as manufacturing. However, there are units that convert materials or substances into new products at home or in the worker's home, and those are selling products to general public. For example, as custom tailors, bakery etc. are also included in this section. We select the companies which are engaged in:

1. Manufacture of food products and beverages
2. Manufacture of tobacco products
3. Manufacture of textiles and wearing apparel
4. Manufacture of leather and wooden products
5. Manufacture of paper and paper products
6. Printing and reproduction of recorded media
7. Manufacture of coke and refined petroleum products
8. Manufacture of chemicals and chemical products
9. Manufacture of medicinal, chemical, pharmaceuticals, and botanical products
10. Manufacture of rubber and plastics products
11. Manufacture of other non-metallic mineral products

12. Manufacture of basic metals and fabricated metal products
13. Manufacture of computer, electronic and optical products
14. Manufacture of electrical equipment
15. Manufacture of machinery and equipment
16. Manufacture of motor vehicles, trailers and semi-trailers
17. Manufacture of other transport equipment
18. Manufacture of furniture
19. Manufacture of jewellery, bijouterie and related articles
20. Manufacture of musical instruments
21. Manufacture of sports goods, games and toys
22. Manufacture of medical and dental instruments
23. Repair and installation of machinery and equipment

According to BBS, the contribution of Manufacturing sector to GDP 18.99% in fiscal year 2017-18. The organizations types details has given in Appendix [B.3](#).

### **3.3.4 Electricity, Gas, Steam, and Air Conditioning**

This indicator includes electric power, natural gas, steam, hot water supply activities and through the permanent infrastructure of the line, main and pipe. This section also included the gas, steam and hot water distribution activity in industries or residential buildings. This indicator, therefore, includes activities for electric and gas utilities, which generate, control and distribute electric power or gas. It also included steam and air-conditioning system supplies. We select the companies that are engaged in providing service:

1. Electric power generation, transmission, and distribution
2. Manufacture of gas and distribution of gaseous fuels
3. Steam and air conditioning supply

### **3.3.5 Water supply, sewerage, waste management and remediation activities**

This indicator includes activities related to the management of various types of waste such as industrial, household and other waste. Water collection, treatment and supply activities, sewage treatment, waste collection, treatment and disposal activities; remediation activities and other waste management services have been included this section. We select the companies that are engaged in providing service:

1. Water collection, treatment, and supply
2. Sewerage
3. Waste collection, treatment and disposal activities
4. Remediation activities and other waste management services

### **3.3.6 Construction**

This indicator includes general construction such as full housing, office buildings, stores, and other government and utility buildings, farm buildings, and specialized construction activities for buildings and civil engineering work like motorways, railways, airfields, roads, bridges, tunnels, ports and other water projects, sewer systems, irrigation systems, industrial facilities, pipelines, and sports facilities, electric lines, etc. These include new works, repairs, additions and alterations, the construction of pre-built buildings or structures on the site and the construction of a temporary nature. This work can be carried out on your own account or on a fee or contract basis. Parts of the work and sometimes even the whole practical work can be subcontracted. Building and engineering repairing works are also included. We select the companies that are engaged in providing service:

1. Construction of buildings
2. Civil engineering
3. Specialized construction activities

### **3.3.7 Wholesale and Retail Trade, Repair of Vehicles and Motorcycles**

This indicator includes wholesale and retail sales of any type of products and the relevant rendering in the sale of these products. Wholesale is the resale of goods retailers, to industrial, institutional, professional users, commercial, to other wholesalers. Retail

sales are basically the sale of new and used products to the general public. The retail product has the sale in shops, department stores, stalls, private or family Mail-order homes, door-to-door sales to people, hawkers and paddlers, consumer cooperatives, auction house, etc.

We collect the wholesale and retail points which are engaged in providing the following service:

1. Wholesale trade
2. Retail trade
3. All kinds of points which engaged in wholesale or retail trade

The related organizations types has given in Appendix [B.4](#).

### **3.3.8 Transportation and Storage**

This indicator includes the economic units that engaged in passenger or freight transport, by rail, road, water, air and associated activities such as terminal and parking facilities, cargo handling, storage and renting the transport equipment with driver or operator. The postal and courier service are also included here.

We collect the economic unit's information which is engaged in providing the following service:

1. Land transport and transport via pipelines
2. Water transport
3. Air transport
4. Warehousing and support activities for transportation
5. Postal and courier activities

The organizations types has given in the appendix [B.5](#).

### **3.3.9 Accommodation and Food Service**

This indicator includes the economic units that engaged in short-stay accommodation for visitors and the other travelers, and food suitable for immediate use and includes the provision of drinks. This section includes short term accommodation activities, camping grounds, recreational vehicle parks, and trailer parks, other accommodation services,

food, and restaurants, beverage service activities and event catering, mobile food service activities, and other food service activities, food and beverage delivery activities and so on. This section excludes long-term accommodation and food-drinks that are sold as wholesale or retail trade. We collect the economic unit's information which is engaged in providing the following service:

1. Accommodation
2. Food and beverage service activities

### **3.3.10 Information Services and Communication**

This indicator includes the economic units that engaged in the production and distribution of information and cultural products, data communication, information technology, data processing, and other information service related activities. The activities such as telecommunications activities, software publishing, radio, and TV broadcasting, motion picture, and sound recording activities, programming activities and information technology are included in this section. We collect the economic unit's information which is engaged in providing the services as follows.

1. Publishing activities
2. Motion picture, video and television program production, sound recording and music publishing activities
3. Programming and broadcasting activities
4. Telecommunications
5. Computer programming, consultancy, and related activities
6. Information service activities

The organizations types has given in the appendix [B.6](#).

### **3.3.11 Financial and Insurance Services**

This indicator covers financial services activities, including insurance, banking, reinsurance, pension funding and activities to support financial services. This indicator also covers financial services activities like central banking, monetary inter-mediation, other financial intermediaries, activities of funds, holding companies, trusts and similar financial entities, financial leasing, auxiliary financial service activities, credit granting,

financial market administration, fund management activities and activities of insurance agents and brokers. According to BBS, the contribution of this indicator to GDP was 3.93% in fiscal year 2017-18.

We collect the economic unit's information which is engaged in providing the following service:

1. Financial service activities
2. Insurance, reinsurance and pension funding activities
3. Auxiliary financial service activities

The organizations types are given in Appendix [B.7](#).

### **3.3.12 Real-Estate Services**

This indicator covers the economic units that include selling or buying real estate, renting real estate, providing real estate valuation or other real estate services such as acting as a real estate escrow agent. The activities of this section may be conducted on its own or lease property and may be performed on a fee or contract basis. These include structural buildings, such as those associated with owning or leasing structures.

We collect the economic unit's information which is engaged in providing the following service:

1. The real estate activities with own or leased property
2. The real estate activities on a fee or contract basis

### **3.3.13 Professional, Scientific and Technical Activities**

This indicator covers the economic units that include specialized professionals, scientific and technical activities. These types of activities require a high degree of training, and make specialized knowledge and skills available to user. This indicator includes legal and accounting activities, book keeping and auditing activities; architectural and engineering activities, technical testing and analysis, tax consultancy activities, management consultancy activities, scientific research and development, research and experimental development on natural sciences and engineering, specialized design activities, advertising and market research, public opinion polling, veterinary activities, photographic activities and other professional, scientific and technical activities. We collect the information which is engaged in providing the following services.

1. Legal and accounting activities
2. Management consultancy activities
3. Architectural and engineering activities
4. Technical testing and analysis
5. Scientific research and development
6. Advertising and market research
7. Other professional, scientific and technical activities

The organizations types have been given in Appendix [B.8](#).

### **3.3.14 Administrative and Support Services**

This indicator covers the economic units that includes a variety of activities that support general business operations such as rental and leasing activities, leasing of intellectual property and similar products, travel agency, tour operator, reservation service and related activities, activities of employment placement agencies, activities of temporary employment agencies, other human resources provisions, maintenance service activities, services to buildings and landscape activities, general cleaning of buildings, combined facilities support activities, cleaning activities, office administrative, office support and other business support activities, security and investigation activities, security systems service activities, private security activities, investigation activities, other building and industrial cleaning activities, landscape care etc. We collect the economic units of information related to the following services:

1. Rental and leasing activities
2. Employment activities
3. Tour operator, travel agency, reservation service agencies and related service activities
4. Security and investigation activities
5. Services to buildings and landscape activities
6. Office administrative and office support activities

The organizations types has given in the appendix [B.9](#).

### **3.3.15 Public Administration, Defence, Compulsory Social Security**

This indicator covers the units that include the activities of a governmental nature, normally carried out by the public administration. This indicator includes administration of the state and the economic and social policy of the community, legislative activities, taxation, general public administration activities, defense activities, regulation of the activities of providing education, health care, cultural services and other social services, laws and judicial activities regulation of and contribution to more efficient operation of businesses, administrative programmes, provision of services to the community as a whole, foreign affairs, public order and safety activities, compulsory social security activities and compulsory social service activities etc. We collect information which is engaged in providing the following service:

1. The administration of the state and the economic and social policy of the community
2. General public administration activities
3. Regulation of the activities of providing health care, education, cultural services
4. Other social services
5. Regulation of more efficient operation of businesses
6. Foreign affairs
7. Defense activities
8. Public order and safety activities
9. Compulsory social security service activities

### **3.3.16 Education**

This indicator covers the establishment that engaged in private or public education at any level or for any profession. The education may be oral or written as well as by radio and television or other any kind of communication. It includes the education by the different establishments in education system at its different levels as well as literacy programs, special education for physically or mentally handicapped pupils, adult education, religious education, computer education, technical education, training centres, professional coaching centres, military schools and academies, prison schools etc. We collect the establishment information which is engaged in providing the following education services.



1. Pre-primary and primary education
2. Secondary education
3. General secondary education
4. Technical and vocational secondary education
5. Higher education
6. Sports and recreation education
7. Cultural education
8. Educational support activities
9. Other education

### **3.3.17 Human Health and Social Work**

This indicator covers the establishment that involves in health and social work activities. This includes hospital, clinic, or any kind of health care providing organization, and the organizations those have provided social service activities.

We collect the establishment information which is engaged in providing the following service:

1. Human health activities
2. Residential care activities
3. Social work activities without accommodation

### **3.3.18 Arts AND Entropy**

The indicator covers the establishment that includes a wide range of activities to cater to the diverse cultural, recreational and recreational interests of the general public, including live performances, management of museum sites, activities of amusement parks and theme parks, sports and recreation activities, botanical and zoological parks and nature reserves activities.

We collect the establishment information which is engaged in providing the following service:

1. Creative, arts and entertainment activities
2. Libraries, archives, museums and other cultural activities
3. Sports activities and amusement and recreation activities

### **3.3.19 Activities of households as employers; undifferentiated goods- and services-producing activities of households for own use**

This indicator covers the activities of household such as house maids, waiters, cooks, valets, laundresses, butlers, chauffeurs, gatekeepers, stables-lads, caretakers, tutors, secretaries, governors, babysitters etc. It also includes the undifferentiated subsistence goods-producing and services producing activities of households. This section includes the activities as follows:

1. Household activities as employers of domestic workers
2. Different types of goods and service production activities of private families for their own use

### **3.3.20 Activities of extraterritorial organizations and bodies**

This indicator includes activities of international organizations such as the United Nations and the specialized agencies of UN, regional bodies etc., the IMF, the World Bank, the World Customs Organization, the Organization for Economic Co-operation and Development, the OPEC, the European Communities, the European Free Trade Association and other international organizations. This indicator also includes the activities of diplomatic and consular missions.

We collect the establishment information which is engaged as the following organization:

1. Activities of extraterritorial organizations and bodies

### **3.3.21 Other service activities**

This indicator includes the activities that were not covered elsewhere in the classification. It includes the activities of membership organizations, repairing of computers and personal and household goods, funeral and related activities, washing and cleaning of textile and fur products, hairdressing and other beauty treatment, and last of all, a variety of personal service activities (e.g., student, housewife) not covered elsewhere in the classification. We collect the information of the organization or persons which is involved with the following activities:

1. Activities of membership organizations
2. Repairing of computers and personal and household goods

### 3. Other personal service activities

The organizations types has given in the appendix [B.10](#).

## 3.4 Occupation Finding

Occupation means an activity or work with which one occupies himself; especially productive activities, services, commerce, or craft for which one is regularly paid; Job. It is a person's regular work or profession. Occupation can be any work like engineering, government job, a family business, actors, business man, doctors, clerk, banker, teacher, singer, writer etc. Occupation provide us fulfill our basic needs-food, cloth and shelter.

We select that user who shares his/her occupation in social media. Our goal is count the number of employees of each city. There are four indicators of Occupation Dimension as given in the table [3.2](#).

<i>Dimension</i>	<i>Indicators</i>
<b>Occupation</b>	1. Employment 2. Business 3. Non-Residents 4. Other Occupation

Table 3.2: Defining indicators of *Occ.* dimension.

A sample scenario of city ranking at *occupation* dimension has given in the figure [3.5](#).

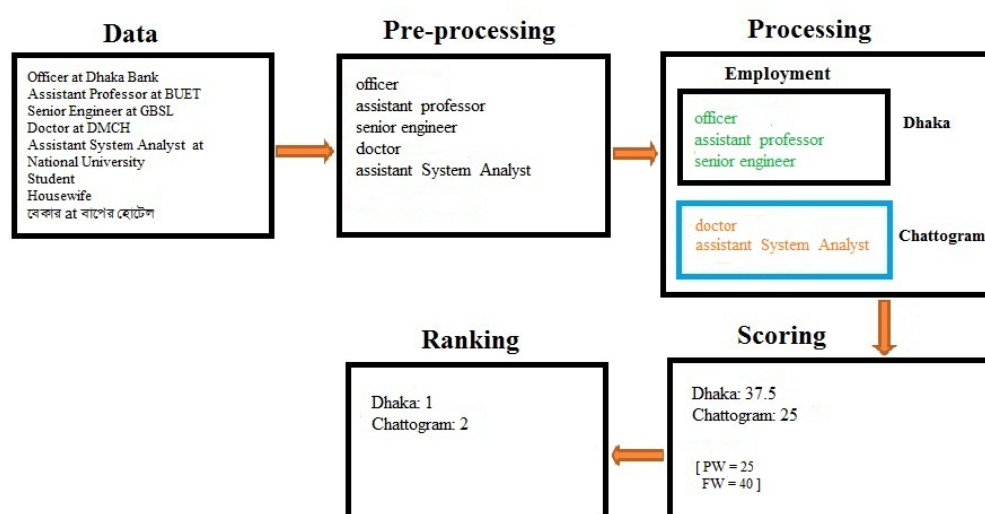


Figure 3.5: A sample scenario of geo-financial ranking at Occupation dimension.

### 3.4.1 Employment

Employment is a relationship between two parties namely, the employer and the employee. It is usually based on a contract where the work is paid for in the profitable or non profitable organization or in the company. Employees work for payments which can be in the form of hourly, weekly, monthly or an annual pay, depending on the kind of work an employee is doing or in which sector he is working. Employees in some fields or sectors may receive gratuity, bonus payments or stock options depending on the nature of employer. In some types of jobs, employees can receive payment as well as some benefits such as health insurance, housing, disability insurance or use of gym etc. Employment is usually governed by employment law, regulation or legal contract.

The five main employment types are:

1. Permanent or fixed-term employees
2. Casual employees
3. Apprentices or trainees – employees
4. Employment agencies staff – known as labor hire
5. Contractors and sub-contractors – known as hired staff

An employment is a job such as teacher, doctor, engineer, banker etc. This indicator includes the professional designation that an employee holds. In this research, we count the number of employees for each city through social media with the aim of detecting city's financial solvency.

### 3.4.2 Business

Business is an activity of producing, buying and selling products to customers. There are three major types of business:

1. **The Service Business:** This type of business provides intangible products such as professional skills, expertise, advice, and other similar products to their customers. The examples of service businesses: salons, repair shops, accounting firms and law firms, schools, banks etc.
2. *Merchandising Business:* In merchandising business, products are buys at wholesale price and make profit by selling those products at higher price than purchase price. So, this types of business is known as “buy and sell” business. In this business, the product is selling without any change of its actual form. Examples of these businesses are: grocery stores, distributors, convenience stores, and other re-sellers.

3. *Manufacturing Business:* In a manufacturing business, a new product is produced from raw materials with using manufacturing process. In this business, the product is selling with the change actual form. This type of business use raw materials to produced a new product. Last of all, the manufactured products is selling to customers.

*Hybrid Business:* A Hybrid business is a business where the companies may be classified in more than one type of business. A restaurant, for example, combines ingredients in making a fine meal (manufacturing), sells cold water (merchandising), and fills customer orders (service).

**Business ownership:** There are seven basic forms of business ownership's:

1. **Sole proprietorship business:** A business owned by a single person.
2. **Partnership business:** In a partnership business, the business is owned by two or more people who share responsibilities, profits and loss of business.
3. **Limited Partnership business:** General partner(s) run the business, while limited partners invest.
4. **Corporation:** A fully- independent business with shareholder.
5. **Limited Liability Company (LLC):** A mixture of a partnership and a corporation business which is designed to make easier to start small business.
6. **Non Profit Organization:** A business that uses its profits for charitable purposes.
7. **Cooperative (Co-op) business:** A business owned and operated for the benefit of the members of the organization that use it services.

### 3.4.3 Non Residents

A Non-Resident Bangladeshi (NRB) is a Bangladeshi citizen who holds a Bangladeshi passport and has temporarily emigrated to another country for work, residence or any other purpose and living abroad six months or more. The term "NRB" covers both the permanent and temporary of Bangladeshi origin– who are working or living abroad for a long time (six month or more). The permanent migrants living in different countries are known as the "Bangladeshi diaspora".

More than 10 million Bangladeshis are living abroad of which about 2.4 million Bangladeshis are living abroad permanently and the rest of living temporarily. They lives in 162 countries like Saudi Arabia, United Arab Emirates, Qatar, Kuwait, Lebanon,

Oman, Bahrain, Japan, Australia, Yemen, UK, USA, Italy, Greece, South Africa, France, Netherlands, Belgium, Canada, Spain, Germany, Switzerland and other countries.

The NRB sends remittance on a regular basis. Remittance is second highest source of foreign currency earnings of Bangladesh after exports of ready-made garments (RMG). The annual remittance of the country has increased significantly as the migrant workers from Bangladesh gradually increase over the years. So, remittance has become a major contributor to the Bangladeshi economy. In the year 2018, the expatriates have sent 15.54 billion of foreign currency. The contribution of foreign remittance sent by the NRB makes up about 12 percent of Bangladesh's GDP.

### 3.4.4 Other Occupation

This section includes other categories that are not included in previous indicators. Example: student, housewife, house maid etc.

## 3.5 Daily Financial Activities (DFA)

How much a person costs or earned daily? An event related to financial that occurred in our daily life and we share it with our friends in social media. For example: buying a new flat, traveling in Nepal, having dinner in a five star hotel. we add this dimension to find the financial transaction status of cities. We think this is more dominant factor to solvency analysis. Our target is count the number of event related to financial as city wise. There are three indicators of DFA Dimension. The table 3.3 shows the indicators of DFA dimension.

<i><b>Dimension</b></i>	<i><b>Indicators</b></i>
<b>Daily Financial Activities</b>	<ol style="list-style-type: none"> <li>1. Buying or Purchasing</li> <li>2. Eating and Drinking</li> <li>3. Entertainment</li> </ol>

Table 3.3: Defining indicators of *DFA* dimension.

A sample scenario of Ranking at *DFA* dimension has given in the figure 3.6.

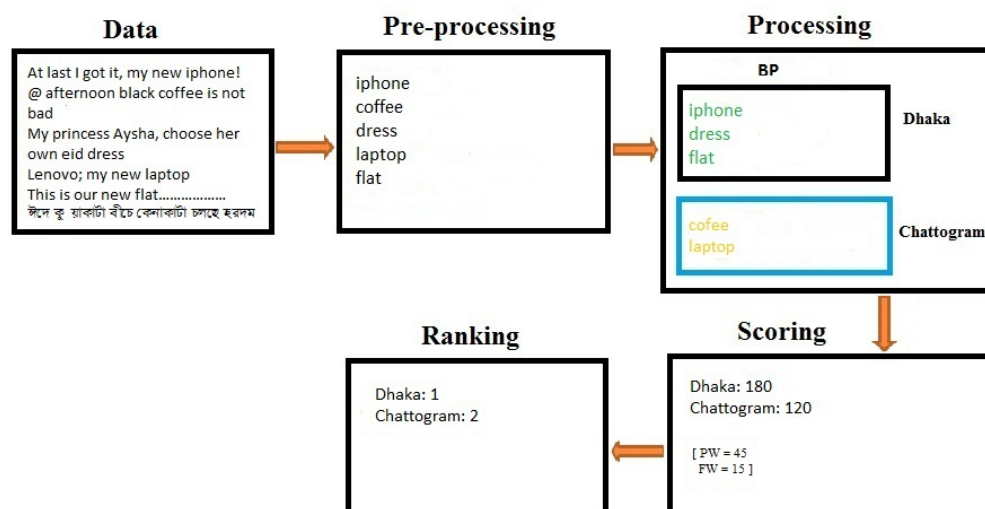


Figure 3.6: A sample scenario of geo-financial ranking at DFA dimension.

### 3.5.1 Buying or purchasing

Buying is most commonly used to refer to everyday goods and commodities, on the other hand, Purchasing is considered to be a more formal term than buy. Purchasing is often used to refer to contracts and large products, on the other hand, buying refers to smaller products.

For example, *'I purchased a new flat'*. *'The government purchased a huge defense contract from the U.S. military'*. However, *'I bought a new phone'*.

Still, it must be noted that in daily usage, the both terms (buy and purchase) are synonymous. For example, *'I bought a new book'*. *'I purchased a new book'*.

### 3.5.2 Eating and Drinking

This indicator includes all kinds of food and drinks that we eat daily or occasionally and share them with our friends on social media (e.g., Facebook). For example, *'With my best friend, Sajib, at Modhumitta Coffee shop'*. *'Haji Biriani is so tasty'*.

### 3.5.3 Entertainment

The arts, entertainment, and recreation are comprised of a variety of organizations that provide benefits or services to their patrons to meet different cultural, entertainment and recreational interests. This index includes companies that are engaged in the production, promotion or participation in live performances, events or exhibitions intended for public viewing; the establishments that preserve and display historical, cultural or educational

objects and sites; and establishments that provide services or operate facilities that enable peoples to engage in recreational activities or pursue amusement, hobbies and leisure interests.

## 3.6 Data Collection

There are two types of data. (1) Economic Establishment data which has been collected from web media (Bangladesh Yellow Pages), and (2) Social media data which has been collected from Facebook.

### 3.6.1 Economic Establishment Data

Economic Establishment data has been collected from many web sites which collects company information of Bangladesh. We collect data through web scraping. We take bangladeshyellowpages.com data. We use *R* scripts use for downloading data. The web page contains many kinds information, all are not required for us. We take those data which is needed for our research. Therefore, to select those type of data we installed the “Selector Gadget” to Browser to select relevant data table. The data collection procedure is given as below:

1. Install Selector Gadget on google browser from <https://selectorgadget.com/>
2. Select the relevant data table of bangladeshyellowpages.com using Selector Gadget
3. Using *R* script for downloading selected data
4. Run the *R* script on *R* studio
5. Then the unstructured data downloaded as text format and we saved it in a csv file

The figure 3.7 shows the data collection process using web media.

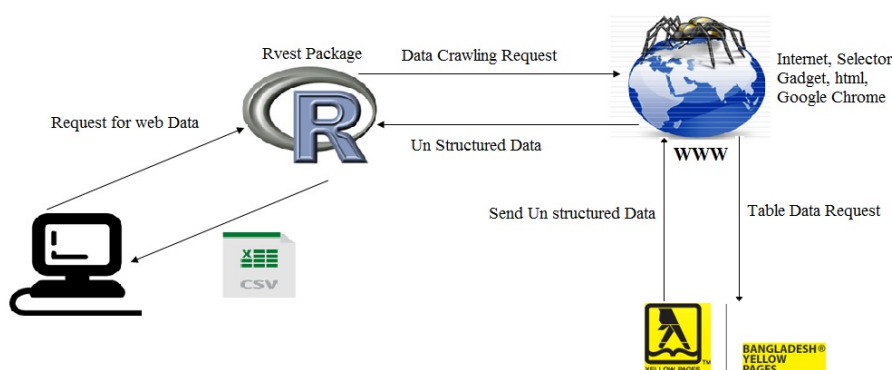


Figure 3.7: Economic establishment data collection.



### 3.6.2 Social media data collection

For Occupation and DFA data collection, we analyze social media posts. Firstly, we considered twitter, facebook, yahoo etc. as social media. However, mostly Bangladeshi people use Facebook. Therefore, we target facebook users for data collection. We collected data from user who permitted us to access their personal data. The figure 3.8 shows the data collection process using social media.

#### Personal Information

Users provide a wealth of personal information in his/her Facebook profile. We collected the raw data included features like the user's name, hometown, current city, profession, and workplace. From current city we detect them as the residents or non-residents. This include values of 0 for the users who did not supply any information.

#### Activities and Preferences

Providing lists of personal activities or favorite things have been always been a part of Facebook. We collected the user's posts which are related to financial activities like travel, buying something, about food, about entertainment. We counted the number of activities as city wise.

*R* scripts have been used for processing the posts or data. The following steps are taken for social media data collection:

1. To create an apps through <https://developers.facebook.com/tools>
2. To invite the facebook users to give permission to use on their data by registration on that app or by using graph API explorer.
3. To get the user access token
4. We write a *R* script for download the user data
5. Run the *R* script on *R* studio
6. Then the unstructured data downloaded as csv format

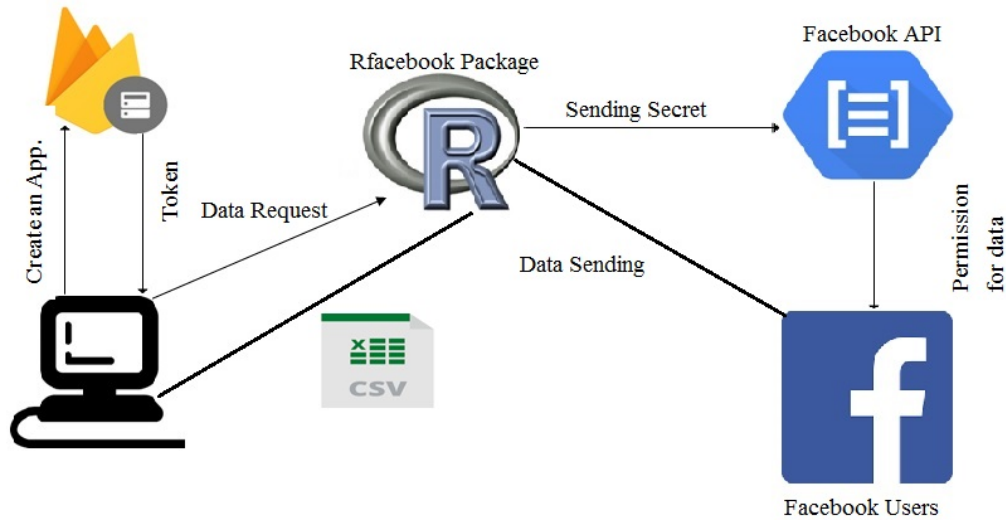


Figure 3.8: Social media data collection process

## 3.7 Data Pre-processing

*Data pre-processing* technique involves the conversion of raw data into an understandable format. Real-world data is usually incomplete, inconsistent, and/or is a lack of specific behaviors or tendencies and can have many flaws. The *Data pre-processing* technique is a proven method to resolve such flaws. There are many techniques [17–19] which have been studied for pre-processing the data. Data pre-processing steps involves: (1) citizen selection in social media, (2) Data cleaning.

### 3.7.1 Citizen selection from social media

To find the citizen of our targeted areas we performed the following steps:

1. Check the user home town with area dictionary.
2. Remove the user who is not the citizen of our targeted areas.

The figure 3.9 shows a scenario of citizen selection procedure in social media.

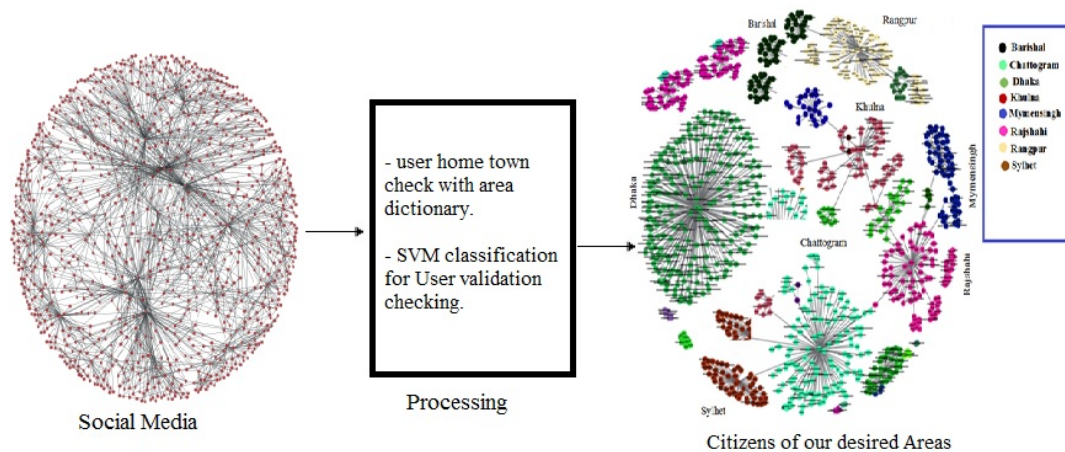


Figure 3.9: A scenario of citizen selection in social media.

### 3.7.2 Pre-processing of Economic Establishment Data

Pre-processing of establishment data includes the following tasks:

1. To make all text into lowercase using tolower() function.
2. To remove all blank lines using regular expression.
3. To remove all URLs, hash tags, targets using the function removePunctuation().
4. To removes tabs and extra spaces using stripWhiteSpace()
5. To remove unnecessary words using removeWords()
6. To remove Non-English lines

The figure 3.10 shows the steps of data pre-processing to get structured data.

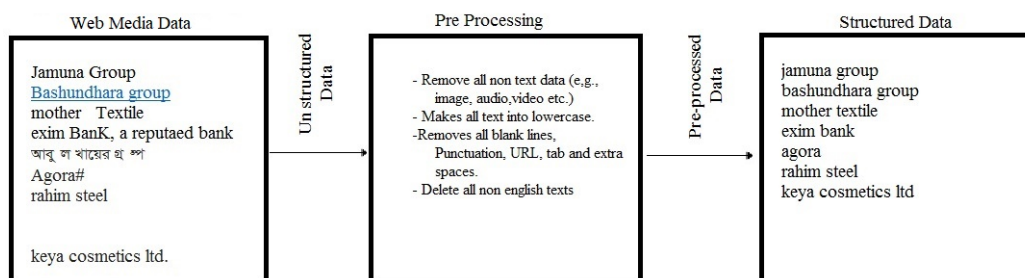


Figure 3.10: The pre processing steps of web media data at *EE* dimension

### 3.7.3 Pre-processing of Social Media Data

Pre-processing of social media data includes the following tasks:

1. To make all text into lowercase using `tolower()` function.
2. To remove all blank lines using regular expression.
3. To remove all URLs, hash tags, targets using the function `removePunctuation()`.
4. To remove tabs and extra spaces using `stripWhiteSpace()`
5. To remove unnecessary words using `removeWords()`
6. To remove Non-English lines
7. to correct the spellings; sequence of repeated characters is to be handled.
8. to remove all punctuations, symbols, numbers
9. to remove Stop Words: There are some words that are frequently used in English sentences but provide little information. These are known as stop words, and we may want to remove them from our analysis. For example, "I", "she'll", "the", etc. There are 175 common English stop words are found in `tm` package which is given as below.

#### STOP WORDS:

you, your, yours, yourself, yourselves, we, our, ours, ourselves, i, i'm, me, my, myself, he, him, too, very, will, his, himself, she, her, hers, herself, it, its, itself, they, them, their, theirs, themselves, have, has, had, having, do, does, did, doing, what, which, who, whom, this, that, these, those, am, is, are, was, were, be, been, being, i've, you've, we've, they've, i'd, would, should, could, ought, you're, he's, she's, it's, we're, they're, you'd, he'd, she'd, we'd, they'd, doesn't, don't, didn't, i'll, you'll, he'll, she'll, we'll, they'll, isn't, aren't, wasn't, weren't, hasn't, haven't, hadn't, won't, that's, who's, what's, wouldn't, shan't, shouldn't, can't, cannot, couldn't, mustn't, let's, here's, there's, when's, where's, why's, of, at, by, for, with, how's, a, an, the, and, but, if, or, because, as, until, while, about, to, from, up, down, in, against, before, after, above, below, when, where, why, how, no, nor, not, out, on, off, over, under, again, all, between, into, through, during, any, both, each, few, more, most, other, than, further, then, once, here, there, some, such, only, own, same, so.

## 3.8 Data Processing

Data processing has been done for manipulating the data indicator wise and to make data fit for score calculation. There are many techniques [20–23], have been studied for data processing. In this research the data processing is done with the following tasks:

1. User validation checking
2. Dictionary construction
3. Financial word counting from user posts
4. Number of non-residents counting
5. Number of employee counting

### 3.8.1 User validation checking

To detect valid user of social media we perform the following steps:

1. We use SVM classifier for User validation checking
2. We take those users who provide at least three types of information in his/her facebook profile : home town, location/Current City and work/profession.
3. We remove those users who post under twenty [24], posts to discourage insufficient information for profile computing.
4. SVM provides the best classification accuracy and therefore, we take SVM for social media user validation checking. (Classification is done with WEKA software.)

The figure 3.11 shows the flow Chart for valid user checking in social media.

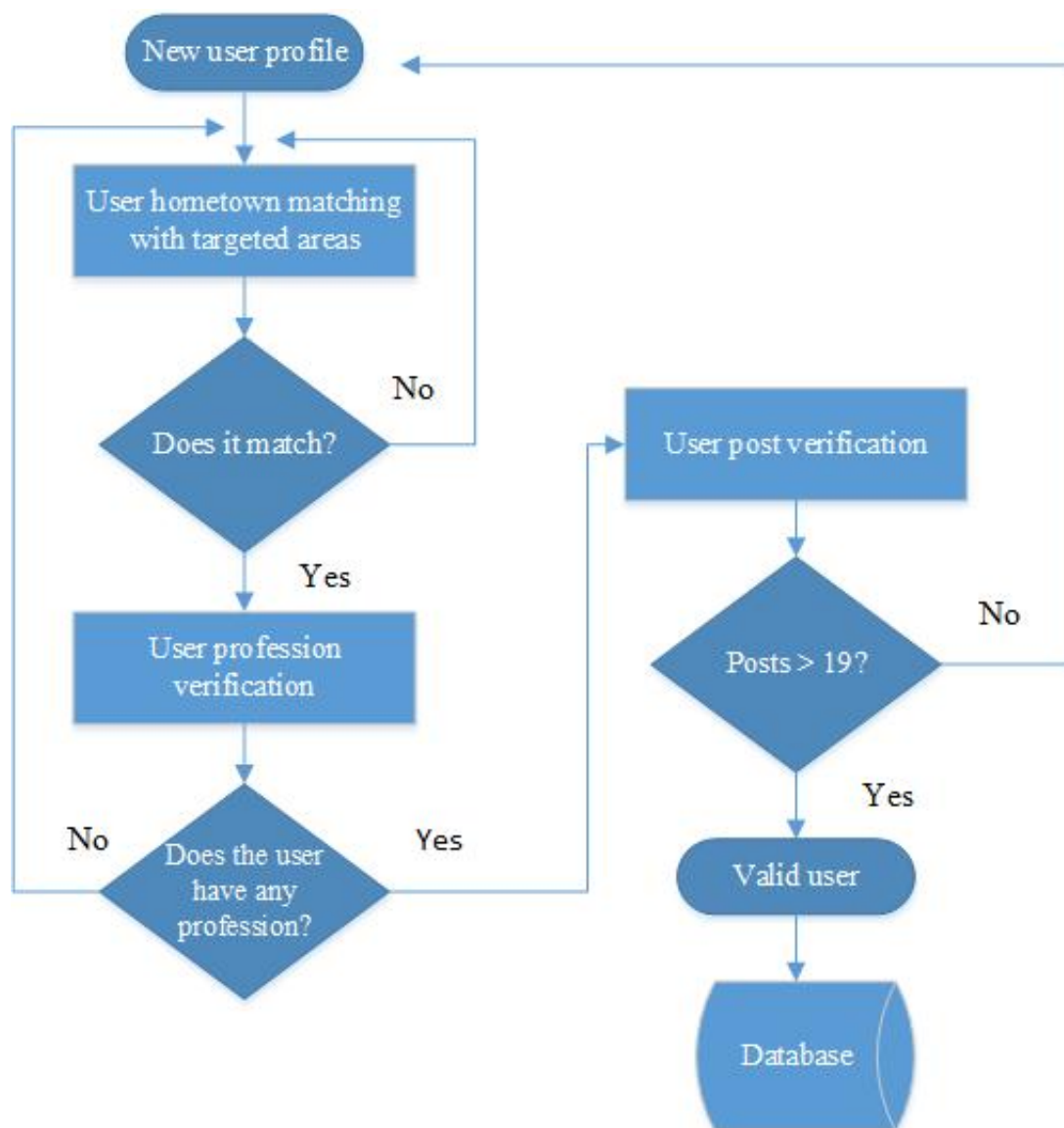


Figure 3.11: The flow chart of citizen selection procedure in social media.

The table 3.4 shows SVM classification to select valid social media user.

<b>Id</b>	<b>Hometown</b>	<b>Location</b>	<b>Work</b>	<b>Posts</b>	<b>Valid</b>
10214826143312460	1	1	1	20	Yes
558523984664452	0	1	1	31	Yes
104725000622012	0	0	0	21	No
664032562182458	1	1	1	8	No

Table 3.4: A sample of data classification procedure to detect valid citizen in social media

The user provides information=1 and the user provides no information= 0

### Using different Classifiers

We use different classifier like linear regression, voted perception and SMO to check user validity. Among three the SVM performs better, so, we use it for social media user selection.

### 3.8.2 Financial word counting from user's posts

We constructs four dictionaries as follows:

1. **Area dictionary:** It contains the name of our 8 divisional cities for citizen selection.
2. **Word dictionary:** It contains financial related words for counting financial word from user posts.
3. **Country dictionary:** It contains all countries and famous cities name in the world for finding non-residents. But it excludes the target areas name (eg., in this research, Bangladesh and the eight divisional cities name).
4. **Occupation dictionary:** It contains the designation name of different profession.

The figure 3.12 describes the steps of financial word counting from user's posts.

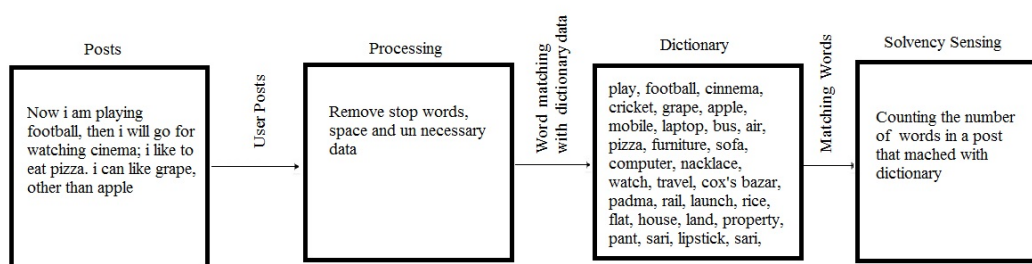


Figure 3.12: The steps of finding financial related word from user's post in social media.

### 3.8.3 Non-residents computation

The objective of this section is counting the number of non-residents by matching the user location with country dictionary. The figure 3.13 depicts the steps of non-residents detection from user's profile.

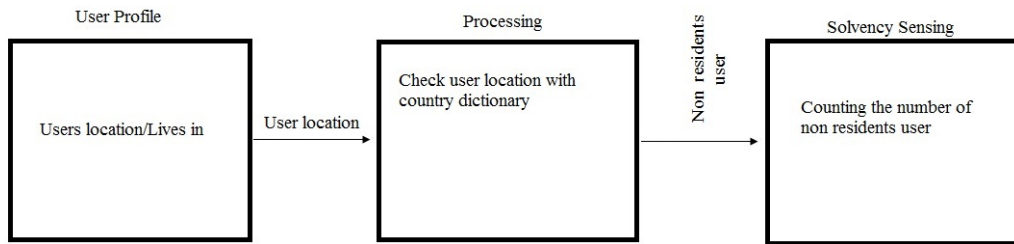


Figure 3.13: The steps of finding Non residents from social media users.

### 3.8.4 Employee finding

The objective of this section is counting the number of employees in an area by ignoring non-employed users. When we get a user's profession we match it to occupation dictionary. If it matching we just count the number of employees, businessman as city wise. In the case of user who does not provides the information of designation, we just count there the number of organization name. The figure 3.14 express the steps of employee finding.

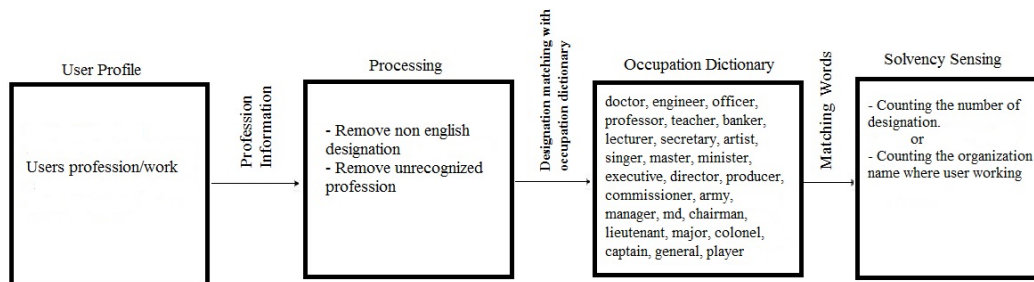


Figure 3.14: Finding employees from user profession

## 3.9 Score Calculation

The purpose of this section is to calculate dimensions and indicators scores for each city. There are two steps for calculating score which are noted as follows:

1. **Weight Distribution:** which describes the distribution of primary and final weight among dimensions and indicators.
2. **Scoring on *EE*, *Occ.* and *DFA* dimension:** which describes the score calculation of the areas based of proposed method.



### 3.9.1 Weight Distribution

The objective of this section is to assign weight for each dimension and indicators. To achieve this goal, we assign two types of weight (1) Primary Weight and (2) Final Weight.

- The Primary Weight, PW, is assigned to 28 indicators to measure the indicator level score.
- The Final Weight, FW, is assigned to 3 dimensions to measure the dimension level score.

#### Weight assigning at Dimension level

To assign weight among dimensions, we have studied many national and international standard methods [1, 8, 16]. We called the dimension level weight is the Final Weight (FW). The final weight of 3 dimensions (*EE*, *Occ.*, *DFA*) has distributed on a scale of 1-100. The table 3.5 describes the weight distribution statistics among three dimensions.

<b>Dimension</b>	<b>FW</b>
<b>Economic Establishment</b>	45
<b>Occupation</b>	40
<b>Daily Financial Activities</b>	15
<b>Total</b>	100

Table 3.5: Distribution of weight (FW) at dimension level

#### Weight distribution to the indicators of EE dimension

The national contribution of 21 indicators in Bangladeshi economy finding by economic census-2013 [1], is considered as primary weight in economic establishment indicators. The table 3.6 describes the primary weight distribution statistics among 21 indicators of EE dimension.

S/N	Indicator Name	PW
1.	AFF	2.9
2.	MAQ	0.22
3.	MFG	11.76
4.	EGSA	0.15
5.	WSWRA	0.75
6.	Const.	2.39
7.	WRT	45.91
8.	TAS	13.65
9.	AFSA	4.77
10.	IC	0.55
11.	FIA	0.47
12.	REA	0.15
13.	PSTA	0.65
14.	ASSA	0.39
15.	PAD	0.10
16.	Edu.	3.93
17.	HHSWA	2.03
18.	AER	0.39
19.	AHE	0.34
20.	AEOB	0.03
21.	OSA	8.48
	TOTAL	100

Table 3.6: Primary weight distribution on indicators of *EE* Dimension

### Weight distribution to the indicators of *Occ.* dimension

The table 3.7 describes the primary weight distribution statistics among 4 indicators of *Occ.* dimension. The primary weight of the indicators of *Occ.* dimension has calculated on a scale of 1-100.

S/N	Indicator Name	PW
1.	Emp.	25
2.	Bus.	30
3.	NR	35
4.	OO	10
	TOTAL	100

Table 3.7: Primary weight distribution on indicators of *Occ.* Dimension

### Weight distribution to the indicators of *DFA* dimension

The table 3.8 describes the primary weight distribution statistics among 3 indicators of *DFA* dimension. The primary weight of the indicators of *DFA* dimension has calculated on a scale of 1-100.

S/N	Indicator Name	PW
1.	BP	45
2.	ED	30
3.	Ent.	25
	TOTAL	100

Table 3.8: Primary weight distribution on indicators of *DFA* Dimension

### 3.9.2 Score Computation for EE, Occ. and DFA

Let,  $X = x_1, x_2, x_3, \dots, x_n$  is the name of areas. In this research we have eight areas,  $X = Barishal, Chattogram, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur, Sylhet$ . Let,  $D = d_1, d_2, d_3, \dots, d_n$  is the number of dimensions. In this work we have three dimensions,  $D = EE, Occ., DFA$ .

Let,  $I = i_1, i_2, i_3, \dots, i_n$  is the number of indicators. In this work we have total twenty eight indicators,  $I = AFF, MAQ, Emp., \dots, Ent..$

We have six steps for score calculation at dimension and indicators levels. We repeat these steps for each city at each indicator and dimension level. At first, we count the numbers of Economic Establishments, Occupation and DFA data for each city using Equation 3.1.

$$S_i = \sum_{i=1}^n L_i \quad (3.1)$$

Where  $S$ = total number of counted data at each indicator level,  $n$  = total number of indicators in a dimension (For EE dimension,  $n=21$ ; for Occupation  $n=4$ , and for DFA,  $n=3$ ), and  $L$ =number of establishments, posts, employees for each indicator of a city.

For example, in this research, we counted 120 manufacturing companies in Dhaka city at *EE* dimension. We get 6 employees in social media in *Occ.* dimension and we counted 192 buying and purchasing activities at *DFA* dimension.

Let, the counted number of EE, Occ. and DFA data are stored in the data table A. The A has M rows and N. where M is the row of the table for EE,  $M=21$ , for Occ.,  $M=4$  and for DFA,  $M=3$  where  $N$ = the number areas, here  $N$  is constant and its value is 8 as we have 8 cities. Now, we will make average the data as row wise. Point Calculation (row level averaging) is as follows:

$$Avg_i = \left(\frac{S_i}{T_i}\right) * 100 \quad (3.2)$$

Where  $T$ = Total counted number of data as row wise.  $T_i = \sum_{i=1}^n S_i$

The equation 3.2 shows the row level averaging of counted data.

Now, we will multiply the row level average value with our distributed weight. Firstly, multiply Avg value with primary weight at indicator level and secondly, multiply Avg

value with the final weight at dimension level.

$$AW_i = Avg_i * weight \quad (3.3)$$

The equation 3.3 shows the multiplication of row level average value with distributed weight. Now, we will find the Total point value by performing column wise summation of data table A. The equation 3.4 shows the column wise summation at indicator and dimension level.

$$Point_x = \sum_{i=1}^l AW_i \quad (3.4)$$

Where  $l$  = the number of indicators mentioned in [3, 4, and 21] for three dimension levels. To compute the score, we divide the total point value by total weight (100) at indicator and dimension wise. The equation 3.5 shows the score calculation.

$$Score_x = \frac{Point_x}{T_w} \quad (3.5)$$

Where  $T_w$  is total weight per dimension level

Now, we will compute the Final score for each city by adding the score of three dimensions.

$$FScore_{x1} = (Scored1_{x1} + Scored2_{x1} + Scored3_{x1}) \quad (3.6)$$

The equation 3.6 shows the Final score calculation.

### 3.10 Computation of Rank

*Rank* is calculated for each dimension level based on weighted score value. We use R built-in-function to compute the rank at each dimension level. The equation 3.7 shows the rank computation at each dimension level.

$$Rank_X = (Score_{x1}, Score_{x2}, Score_{x3}, \dots, Score_{xn}) \quad (3.7)$$

Final rank is computed over the final score of each city's in three dimensions occupation, EE, DFA. We uses R built-in-function to compute the final rank of a city. Equation 3.8 computes the final ranking of divisional cities. .

$$FRank = rank(FScore_{x1}, FScore_{x2}, FScore_{x3}, \dots, FScore_{xn}) \quad (3.8)$$

# Chapter 4

## Results and Analysis

In this chapter, the experimental results and analysis are reported. At first, we provide the description of testing environment.

### 4.1 Tools and Techniques

We use following tools for processing our data.

1. R software: RStudio-1.1.383.
2. Weka 3.8
3. Classification method: SVM

The computation has been processed in a machine with Intel i3-6100 CPU, 3.70 GHz, RAM: 4.00 GB.

### 4.2 Results

In this section, we report our findings. We report the weighted score found for Occupation, Economic Establishment and Daily Financial Activities.

#### 4.2.1 Economic Establishment

Here are the indicators of Economic Establishment of eight cities reported in Table [4.1](#).

Indicator	Barishal	Chattogram	Dhaka	Khulna	Mymensingh	Rajshahi	Rangpur	Sylhet	Total
AFF	15	64	90	31	13	12	18	21	264
MAQ	3	9	13	4	2	2	1	7	41
MFG	11	65	120	18	12	15	15	18	274
EGSA	9	63	180	18	7	19	12	37	345
WSWRA	5	32	131	18	17	18	12	14	247
Const.	15	67	141	39	14	18	16	24	334
WRT	46	152	281	38	36	39	52	32	676
TAS	95	160	300	91	90	75	60	85	956
AFSA	40	95	160	70	60	60	55	98	638
IC	12	45	81	48	10	14	11	20	241
FIA	30	48	85	36	27	31	26	51	334
REA	16	50	110	19	12	14	9	47	277
PSTA	8	31	67	21	8	8	13	12	168
ASSA	7	18	28	9	6	8	6	11	93
PAD	12	22	57	19	13	14	13	15	165
Edu.	14	19	55	20	12	18	9	11	158
HHSWA	20	53	110	24	19	21	14	18	279
AER	9	14	47	8	7	10	11	16	122
AHE	20	24	78	27	16	18	10	21	214
AEOB	4	12	32	6	5	5	3	6	73
OSA	25	37	119	28	26	24	16	19	294
<b>TOTAL</b>	<b>416</b>	<b>1080</b>	<b>2285</b>	<b>592</b>	<b>412</b>	<b>443</b>	<b>382</b>	<b>583</b>	<b>6193</b>

Table 4.1: Statistics on the data obtained in the indicators of *EE* Dimension

### 4.2.2 Demographic Data from Social Media

We collect data of ninety users of Facebook, from them we extract full data of 41 users and partial data of 49 users. From user profile we extract name, id, profession, hometown, current city and user posts.

*Full Dataset:* We call it *Full Dataset* when we extract name, id, profession, hometown, current city and posts.

*Partial Dataset:* We collect 49 user data from non-resident Bangladeshis. In this case, we collect name, id and current city. We call it *Partial Dataset*.

City	TNU	TNP	TNNR
Barishal	7	240	1
Chattogram	14	360	5
Dhaka	13	420	2
Khulna	9	251	1
Mymensingh	11	191	1
Rajshahi	8	260	2
Rangpur	6	185	2
Sylhet	22	521	13
<b>Total</b>	<b>90</b>	<b>2428</b>	<b>27</b>

Table 4.2: Statistics of demographic data of social media users.

Figure 4.2 depicts the social media data which we are collect from *Facebook*. Here, TNU denotes Total Number of Users, TNP denotes Total Number of Posts, and TNNR denoptes Total Number of Non-residents.

### Occupation

Table 4.3 shows the statistics for different cities for *Occupation* dimension.

City	Emp.	Bus.	NR	OO	Total
Barishal	3	1	1	2	7
Chittagong	2	4	5	3	14
Dhaka	6	4	2	1	13
Khulna	4	2	1	2	9
Mymensingh	3	3	1	4	11
Rajshahi	2	3	2	1	8
Rangpur	1	1	2	2	6
Sylhet	2	6	13	1	22
<b>Total</b>	<b>23</b>	<b>24</b>	<b>27</b>	<b>16</b>	<b>90</b>

Table 4.3: Data obtained for the indicators of Occupation.

### Daily Financial Activities

Table 4.4 shows the statistics of cities for DFA.

City	BP	ED	Ent.	Total
Barishal	44	111	85	240
Chattogram	109	98	153	360
Dhaka	192	153	75	420
Khulna	75	87	89	251
Mymensingh	67	69	55	191
Rajshahi	58	89	113	260
Rangpur	53	47	85	185
Sylhet	120	231	170	521
<b>Total</b>	<b>718</b>	<b>885</b>	<b>825</b>	<b>2428</b>

Table 4.4: Statistics on the data obtained in the indicators of *DFA* Dimension

## 4.3 Classifier Performance

We tested social media user data on three classifiers (1). Linear Regression (LR) (2). Voted Perceptron (VP) and (3). SVM The classification results have been reported as follows.

*Evaluation Metrics:*

We utilize four metrics for evaluating classifier performance to find the validity social media users in our research.

1. *Accuracy:* The accuracy of classification is defined as the number of correctly classified patterns to the total number of patterns. It can also be expressed as the ratio of the sum of true positives and true negatives to the total number of trials.

$$Accuracy = \frac{(TP+TN)}{(TP+FP+FN+TN)} = \frac{(Citizens\ Correctly\ Identified+Non\ Citizens\ are\ Correctly\ Identified)}{Total\ Number\ of\ Trials}$$

2. *Precision:* Precision tells us about the success probability of making correct positive class classification. It is computed as the number of True Positives divided by the total number of positive calls.

$$Precision = \frac{TP}{(TP+FP)} = \frac{Citizens\ Correctly\ Identified}{(Citizens\ Correctly\ Identified+Individuals\ Incorrectly\ Labeled\ as\ Citizens)}$$

3. *Recall:* Recall explains how sensitive the model is towards identifying the positive class. This is computed as the total number of True Positives divided by the total sum of True Positives and False Negatives.

$$Recall = \frac{TP}{(TP+FN)} = \frac{Citizens\ Correctly\ Identified}{(Citizens\ Correctly\ Identified+Citizens\ Incorrectly\ Labeled\ as\ Non\ Citizens)}$$

4. *F1 Score:* F1 score compute the score by considering both precision and recall of the test and It is the harmonic mean of the precision and recall. It reaches its best value



1 (perfect precision and recall) and worst at 0. The F1 score cannot be greater than accuracy.

$$F1\ Score = 2 * \left( \frac{Recall * Precision}{Recall + Precision} \right)$$

<b>Metrics</b>				
<b>Classifier</b>	<b>Accuracy</b>	<b>Precision</b>	<b>Recall</b>	<b>F1 Score</b>
<b>LR</b>	0.5	0.7	0.47	0.56
<b>VP</b>	0.54	0.01	1	0.02
<b>SMO</b>	0.58	0.78	0.53	0.63

Table 4.5: Classifiers to detect citizens from social media.

From Table 4.5, we have found that SMO performs better than other three classifiers. Therefore, we have selected SMO for data classification. The corresponding performance chart has given in the figure 4.1.

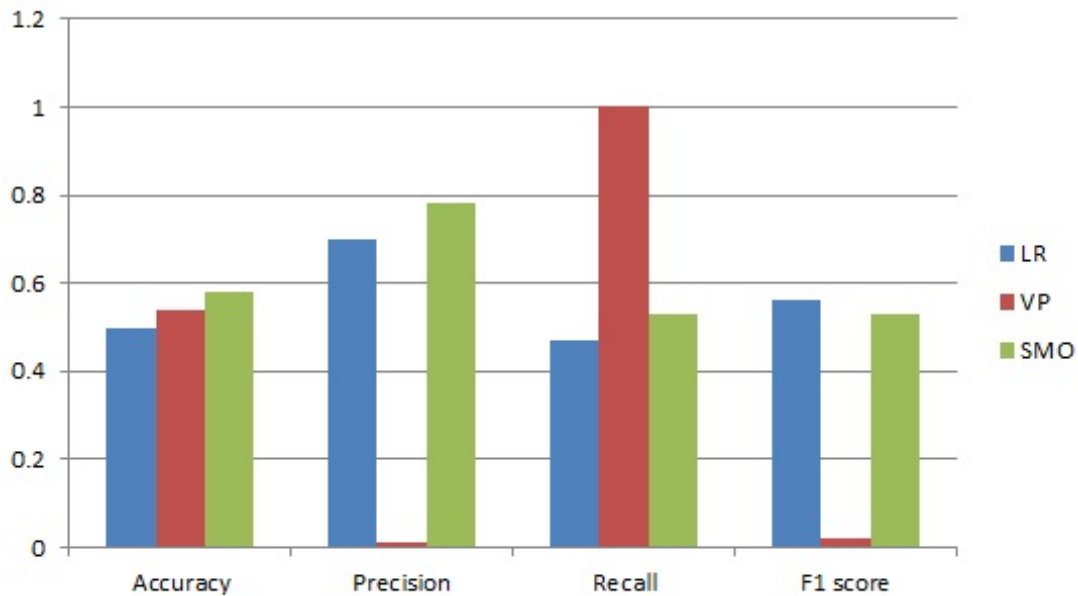


Figure 4.1: Classifier performance chart

## 4.4 More Results

From the analysis we have found Barishal, Chittagong, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur and Sylhet score 6.98, 19.88, 38.92, 7.73, 6.37, 6.66, 6.79 and 6.67 respectively. Here, Dhaka have got the highest score and Mymensingh scored lowest. The table 4.6 describes the score of cities at *EE* dimension.

City	Score
Barishal	6.98
Chattogram	19.88
Dhaka	38.92
Khulna	7.73
Mymensingh	6.37
Rajshahi	6.66
Rangpur	6.79
Sylhet	6.67

Table 4.6: Score for cities in *EE* dimension.

#### 4.4.1 Occupation Dimension

From the analysis we have found Barishal, Chattogram, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur and Sylhet's scores are 7.06, 15.53, 14.74, 9.39, 10.81, 9.14, 6.18 and 27.15 respectively. Here, Sylhet has got the highest score, and Rangpur has scored the lowest. The table 4.7 expresses the score of cities at *Occ.* dimension.

City	Score
Barishal	7.06
Chattogram	15.53
Dhaka	14.74
Khulna	9.39
Mymensingh	10.81
Rajshahi	9.14
Rangpur	6.18
Sylhet	27.15

Table 4.7: The achieved score value of cities at *Occ.* dimension.

#### 4.4.2 Scoring in *DFA* dimension

In *DFA* dimension, we have found Barishal, Chattogram, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur and Sylhet's scores 9.1, 14.79, 19.49, 10.35, 8.2, 10.08, 7.49 and 20.5 respectively. Here, Sylhet has got the highest score, and Rangpur has scored the lowest. The table 4.8 shows the score of cities at *DFA* dimension.

City	Score
Barishal	9.1
Chattogram	14.79
Dhaka	19.49
Khulna	10.35
Mymensingh	8.2
Rajshahi	10.08
Rangpur	7.49
Sylhet	20.5

Table 4.8: The achieved score value of cities at *DFA* dimension

### 4.4.3 Overall Score Computation

The overall score is the summation of three dimensions's scores. In this research We found Barishal, Chattogram, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur and Sylhet scores 7.33, 17.38, 26.33, 8.79, 8.42, 8.17, 6.65 and 16.94 respectively. Here, Dhaka has got the highest score and Rangpur has scored the lowest. The table 4.9 express the overall score of all cities in three dimension.

City	Score
Barishal	7.33
Chattogram	17.38
Dhaka	26.33
Khulna	8.79
Mymensingh	8.42
Rajshahi	8.17
Rangpur	6.65
Sylhet	16.94

Table 4.9: Final scores for all cities as measured by *SSM*.

## 4.5 Rank Computation

This section shows the ranks of the cites based on three dimensions, and the last of all we report the overall ranking.

### 4.5.1 Rank in *EE*

In *EE* dimension, we have found Barishal, Chittagang, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur and Sylhet ranks 4, 2, 1, 3, 8, 7, 5 and 6 respectively. Here, Dhaka got

the 1st rank and Barisal has got the 8th position. The graph 4.2 expresses the rank of the cities at *EE* dimension.

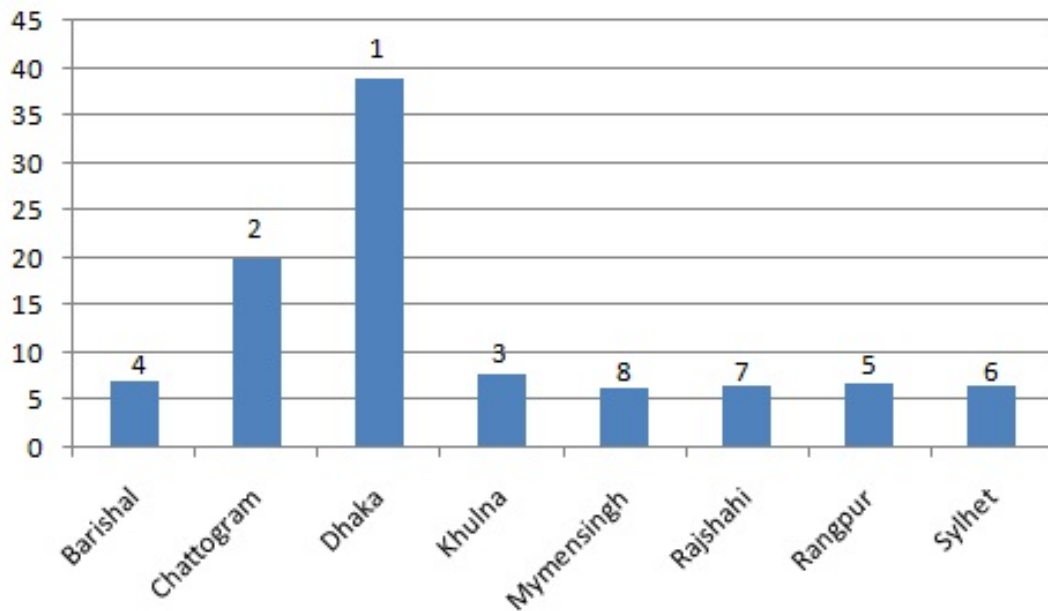
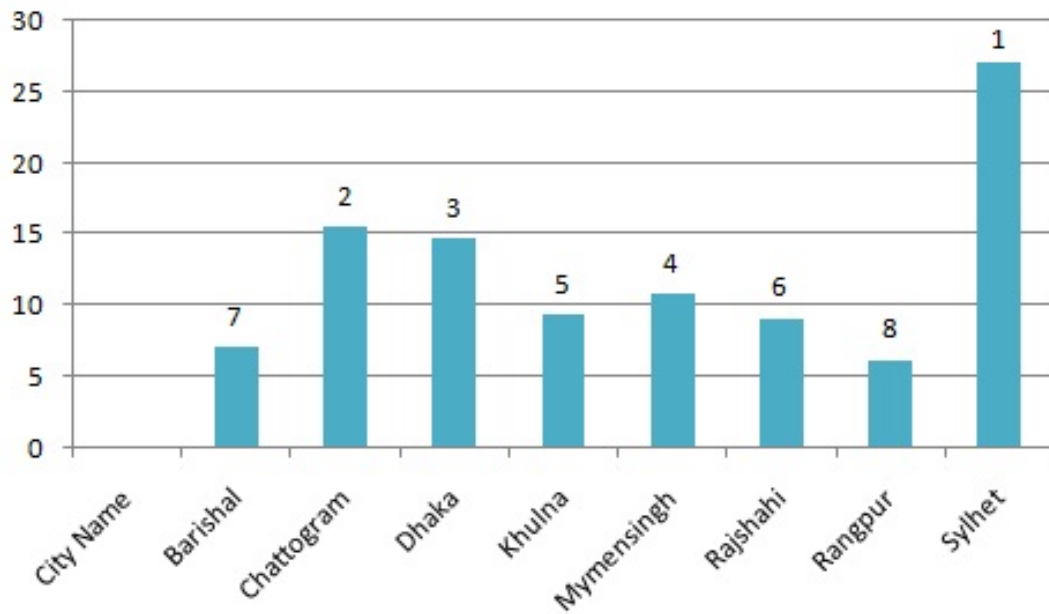


Figure 4.2: City rank in *EE* dimension

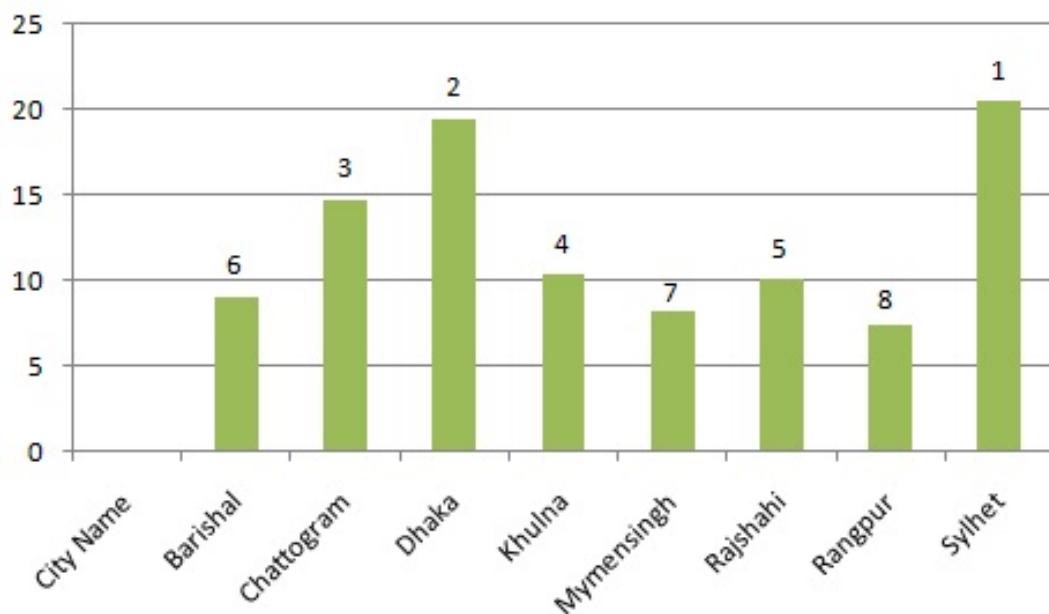
### 4.5.2 Rank in *Occupation*

In *Occupation*, we have found Barishal, Chittagong, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur and Sylhet ranks 7, 2, 3, 5, 4, 6, 8 and 1 respectively. Here, Sylhet has got the 1st rank and Rangpur has got the 8th rank. The graph 4.3 describes the rank of cities at *Occupation*.

Figure 4.3: City rank in *Occupation*.

### 4.5.3 Ranking in *DFA*

In DFA, we have found Barishal, Chittagong, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur and Sylhet's ranks as 6, 3, 2, 4, 7, 5, 8 and 1 respectively. Here, Sylhet got the 1st rank and Rangpur got the 8th. The graph 4.4 shows the rank of cities for *DFA*.

Figure 4.4: City rank in *DFA*.

#### 4.5.4 Overall Rank

Overall Rank The overall rank has been computed over the final score of each city's in three dimensions. In summary results, we have found Barishal, Chittagong, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur and Sylhet rank as 7, 2, 1, 4, 5, 6, 8 and 3 respectively. Here, Sylhet has got the 1st rank and Rangpur has got the 8th rank. The graph 4.5 expresses the overall ranking of eight divisional cities of Bangladesh.

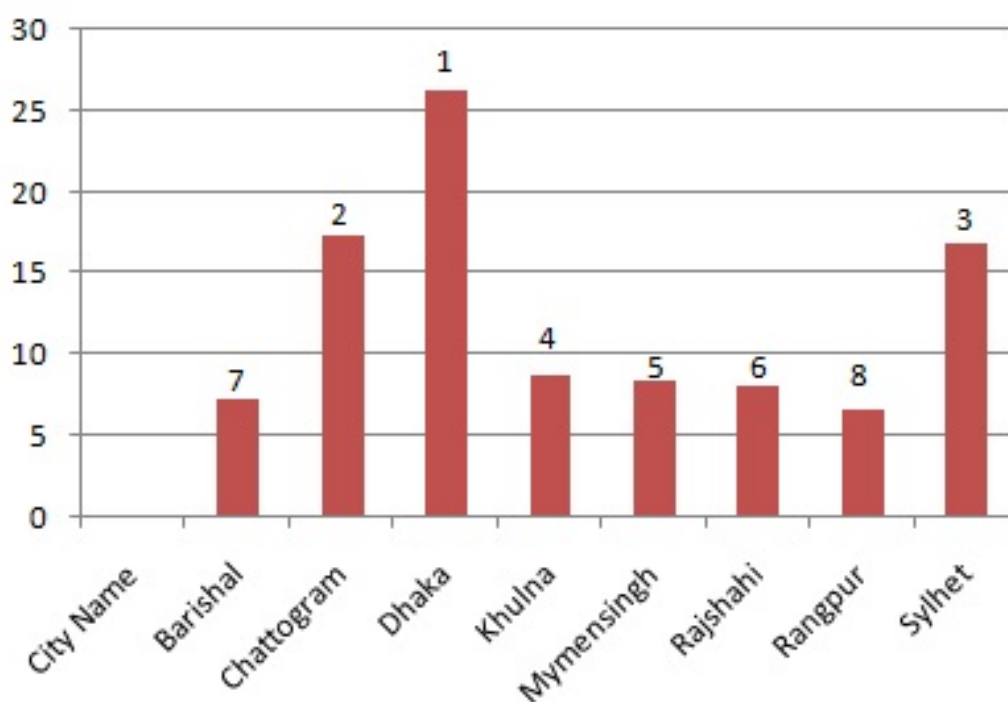


Figure 4.5: Overall ranking.

## 4.6 Result Comparison

Here, we compare our results with BBS's [Bangladesh Bureau of Statistics] data. Figure 4.6 shows the score comparison between BBS's and our computed data in *Economic Establishments* dimension. Here, we find some deviations for the following reasons:

- BBS results are derived from the conducted district wise surveys, on the other hand our SSM computes rank from the city wise data.
- SSM uses small amount of data while BBS uses big amount of data.
- In SSM, we use three dimensions but BBS uses only two dimensions.

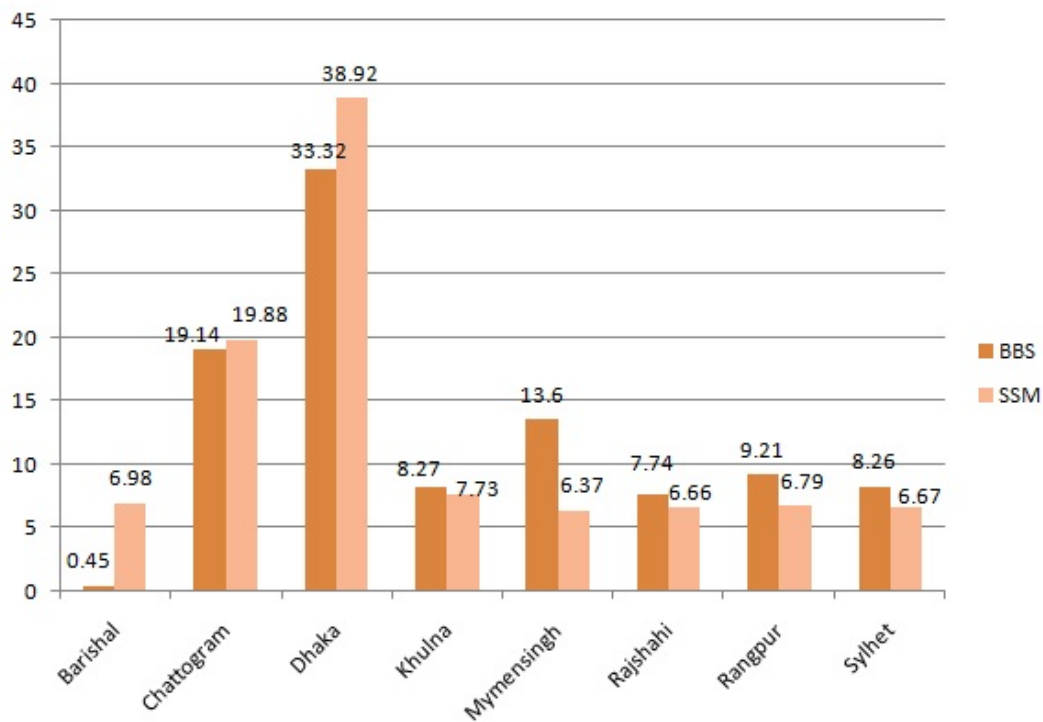


Figure 4.6: Comparison between scores computed by SSM and BBS for *EE* dimension.

Figure 4.7 shows the ranking obtained from BBS and our SSM for *EE* dimension. Here, we have found that four cities ranking match with BBS ranking (Dhaka, Chittagong, Rajshahi and Sylhet).

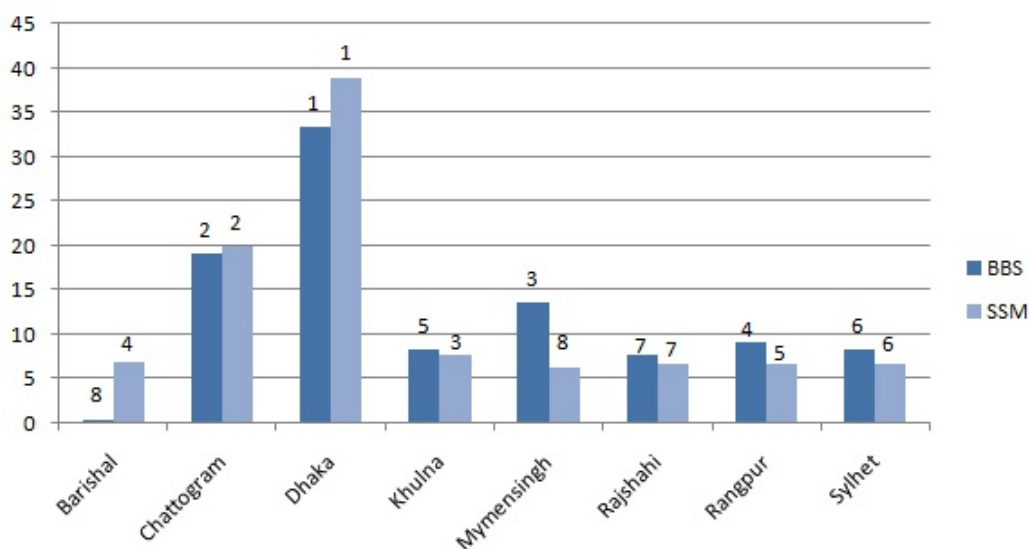


Figure 4.7: Comparison between ranking computed by SSM and BBS for *EE* dimension.

Figure 4.8 describes the comparison of scores computed by BBS and SSM for *Occupation* dimension. Here also, we find some variations in score because:

- SSM considers non-residents while in BBS collected data with only residents.
- BBS collected data occupation district wise. On the other hand, in SSM we count only city base data.

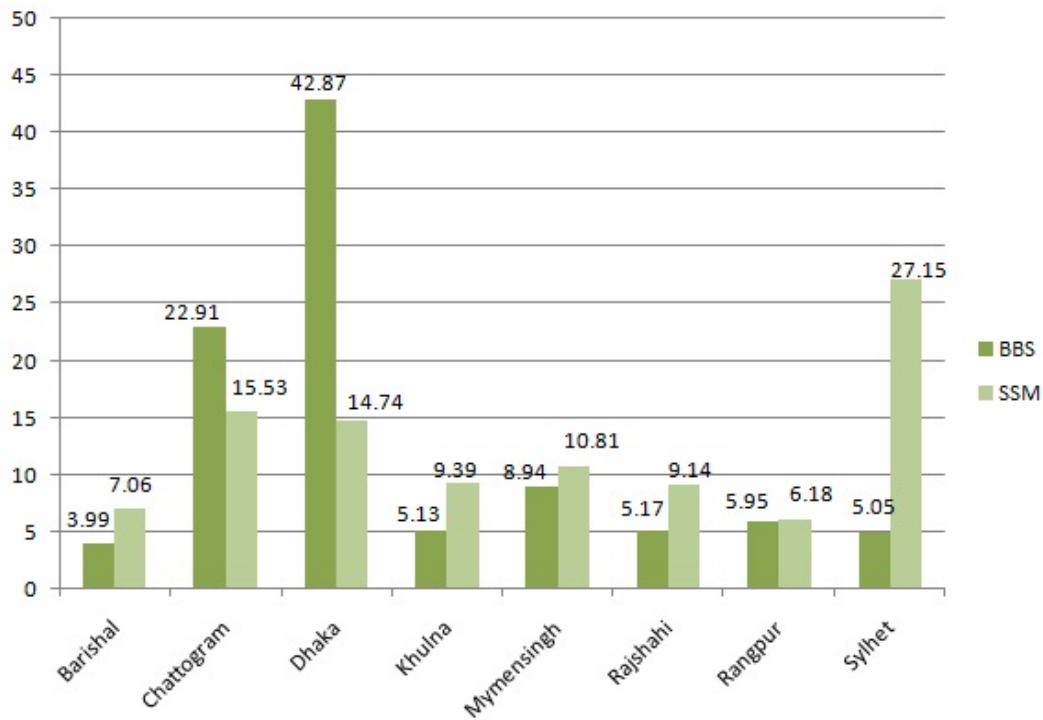


Figure 4.8: Comparison of scores between SSM and BBS for *Occupation*.

The figure 4.9 shows the comparison in ranking between BBS and SSM for occupation. Here we have found dissimilarities in cities's ranking and BBS's ranking (Barishal, Dhaka, Khulna, Mymensingh, Rajshahi, Rangpur and Sylhet). Only Chittagong matches.



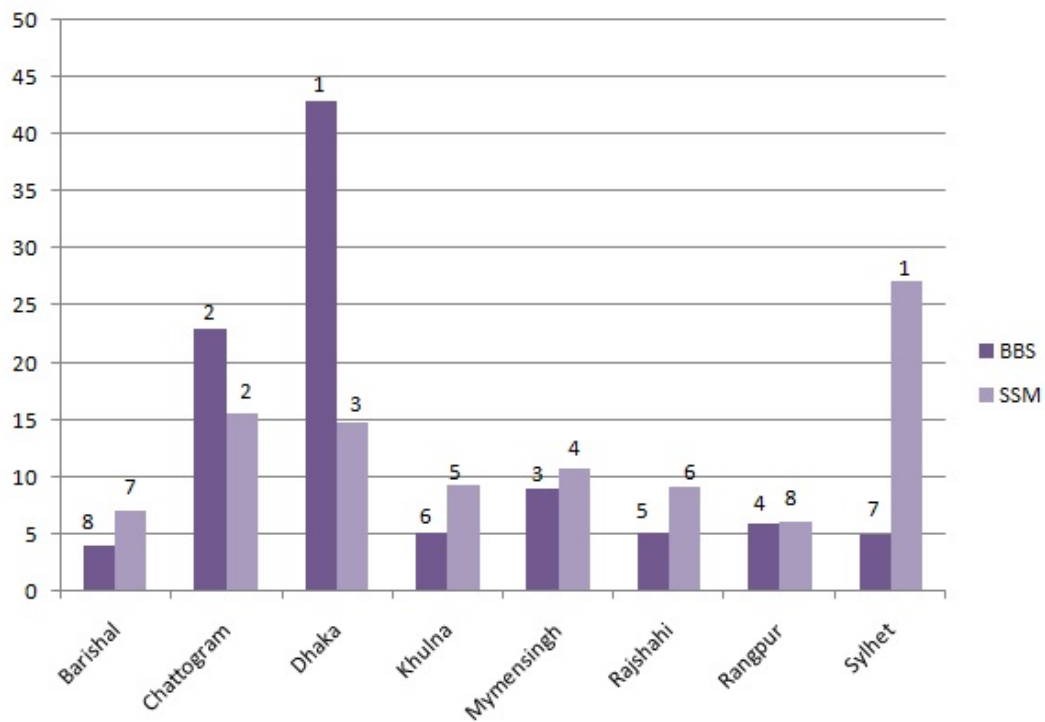


Figure 4.9: Comparison of ranking computed by SSM and BBS for *Occupation*.

Now, Figure 4.10 shows the comparison in overall score between BBS and SSM. Here, we find some differences in the score for the reasons described as follows.

- BBS results conducted district wise, on the other hand SSM uses city wise data.
- SSM uses small amount of data, while BBS uses big amount of data.
- SSM uses three dimensions, but BBS uses only two dimensions.
- SSM considers non-residents, while BBS considers only residents.

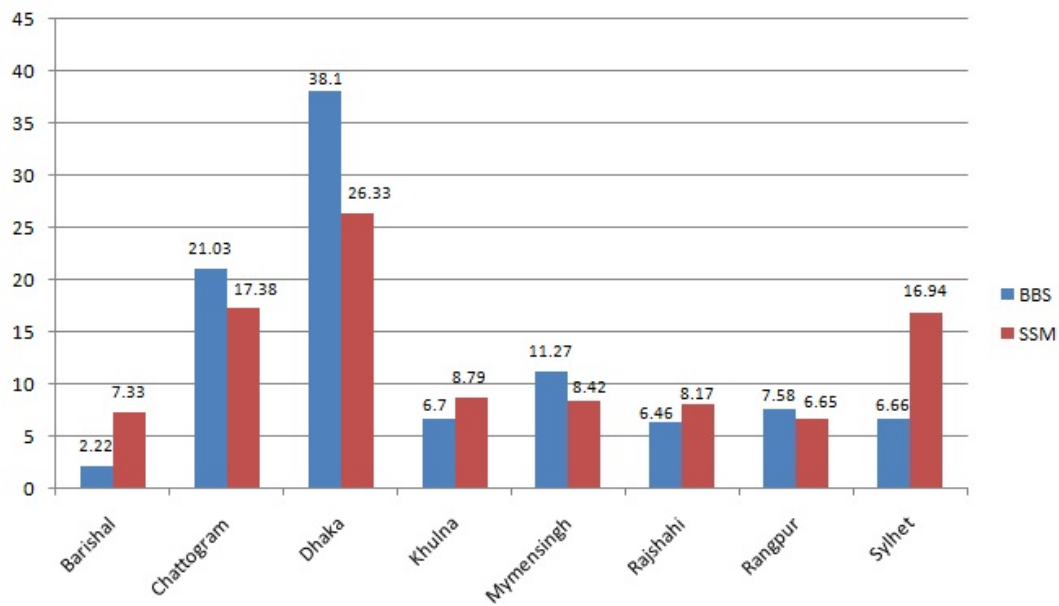


Figure 4.10: Comparison of overall score between SSM and BBS.

Figure 4.11 shows the comparison of overall ranking between BBS method and SSM. Only Dhaka and Chittagong match in overall ranks.

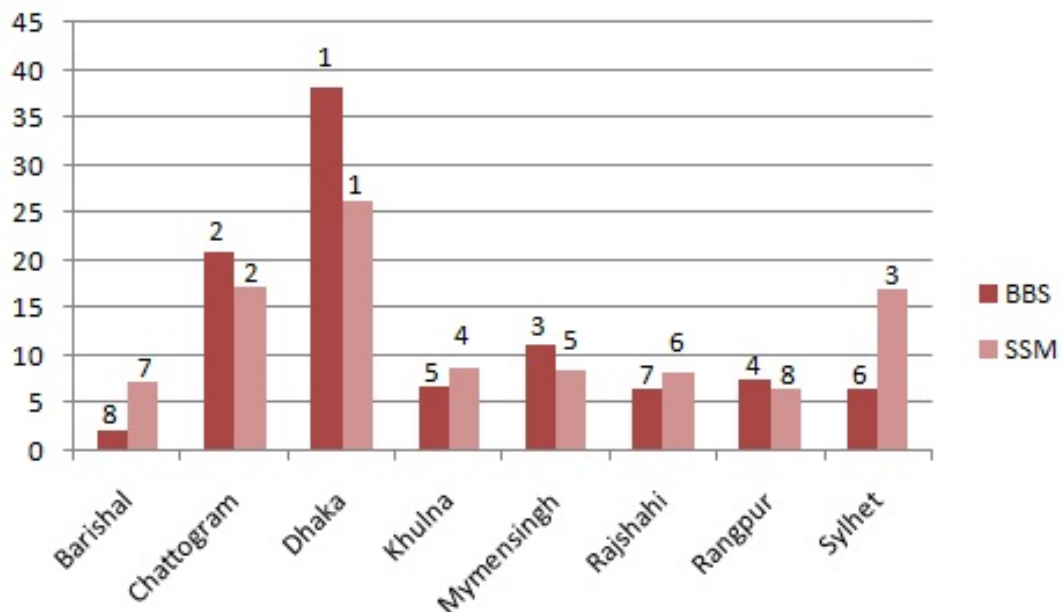


Figure 4.11: Comparison between overall ranking between SSM and BBS.

## 4.7 Limitations in Data Pre-Processing

The models we have built are not free of error. The probable causes for errors along with a complete analysis are given below:

1. **Language Problem:** Some users upload posts using English and Bengali both. Our model is designed for only English. For example, “*ami akta natun gari kinsi*”. Our framework is not able to extract financial data from these types of posts.
2. **One Count Problem:** Sometimes, our model counts only one activity while a user upload two or more financial activities in a post. Such as an example, ‘*Last week, I enjoyed a long trip, from Sunday to Wednesday i was in Cox’s Bazar and from Thursday to Saturday I was in Sajek*’. From this example, our model counts here only one activity “trip” but in here two activities occurred one is travel in cox’s bazaar and the another is travel in Sajek.
3. **VIP Keeps Out:** Some valuable people could not be automatically included in our research. Many business-VIP who have good business but there information are not available in social or web media. Our framework is based on only available data about the financial activities of general people.

## 4.8 Summary Results

Table 4.10 shows the summary results of our proposed framework.

	Overall		EE		Occ.		DFA	
	Rank	Score	Rank	Score	Rank	Score	Rank	Score
<b>Barishal</b>	7	7.33	4	6.98	7	7.06	6	9.1
<b>Chittagong</b>	2	17.38	2	19.88	2	15.53	3	14.79
<b>Dhaka</b>	1	26.33	1	38.92	3	14.74	2	19.49
<b>Khulna</b>	4	8.79	3	7.73	5	9.39	4	10.35
<b>Mymensingh</b>	5	8.42	8	6.37	4	10.81	7	8.2
<b>Rajshahi</b>	6	8.17	7	6.66	6	9.14	5	10.08
<b>Rangpur</b>	8	6.65	5	6.79	8	6.18	8	7.49
<b>Sylhet</b>	3	16.94	6	6.67	1	27.15	1	20.5

Table 4.10: Summary results.

# Chapter 5

## Conclusions

Financial solvency measurement publishes a directory of national business statistics and overall national economy status. In most countries, the authorities carry out an economic census every five or six years. During a census, they gather lot of valuable information about financial institutions, employees, sales, receipts, assets etc. The data getting from the economic census is very important for industry, community and business because:

- Statistics from the economic census provide policy-makers with evidence-based information used to make appropriate programmatic decisions.
- Census data is used as the basis for a variety of economic activities such as gross domestic product (GDP), national income and commodity accounting (NIPA), and producer price index (PPI).
- To determine organizational structure and product trends the trade and business associations use economic census data for finding key business facts.
- To make decisions about capital investments, operating sites and product development, the individual businesses use the economic census data.

It is a difficult work process due to manual demographic data collection. The manual data collection process increases the costs lot of money and engaged huge people.

Therefore, we have proposed an automated framework called Solvency Sensing Method (SSM), in which data has been collected from social and web media. We measure solvency using Economic Establishment (EE), Occupation, and web based Daily Financial Activity (DFA). Our method saves money, time and effort.

In this research, we have found Dhaka division scored 1st rank, Chittagong scores 2nd, Sylhet 3rd, Khulna 4th, Mymensingh 5th, Rajshahi 6th, Barishal 7th and Rangpur 8th.

We would like to continue our research with a large number of social media users. If we get more data the results will be more accurate. We would like to include more financial indicators. Here, we ignored natural disasters, environmental and time impact. In future we would like to include those for better accuracy.

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# Appendix A

## Algorithms

### A.1 Ranking

---

**Algorithm 1** Rank Calculation

---

**Input:**  $DATA[][]$ ,  $ROW$ ,  $COL$ ,  $W_i[ROW][COL]$

**Output:**  $RANK$  (Rank values in the array)

0: *Set*  $I \leftarrow 0$ ,  $J \leftarrow 0$ ,  $RANK[]$ ,  $SCORE[]$

0:  $SCORE \leftarrow SCORE - CALCULATION(DATA[][]$ ,  $W_i[ROW][COL]$ ,  $ROW$ ,  $COL$ )

    {Rank calculation}

0: **for**  $I \leftarrow 1$  to  $COL$  **do**

0:      $RANK[I] \leftarrow rank(SCORE[I])$

0: **end for**

    {Print Score}

0: **for**  $I \leftarrow 1$  to  $COL$  **do**

0:      $Print(SCORE[I])$

0: **end for**

    {Print Rank}

0: **for**  $I \leftarrow 1$  to  $COL$  **do**

0:      $Print(RANK[I])$

0: **end for**

    =0

---

## A.2 Scoring

---

### Algorithm 2 Score Calculation

---

```

0: function SCORE-CALCULATION(DATA[[[]], Wi[[[]], ROW, COL)
0:   Set I ← 0, J ← 0, Sum ← 0, addrow[], avg[[[]], aw[[[]], score[]
   {Adding Row}
0:   for I ← 1 to ROW do
0:     for J ← 1 to COL do
0:       Sum ← Sum + DATA[I][J]
0:     end for
0:     addrow[I] ← Sum
0:     Sum ← 0
0:   end for
   {Row level averaging}
0:   for I ← 1 to ROW do
0:     for J ← 1 to COL do
0:       avg[I][J] ← ((DATA[I][J]/addrow[I]) * 100)
0:     end for
0:   end for
   {Weighted average without percentage}
0:   for I ← 1 to ROW do
0:     for J ← 1 to COL do
0:       aw[I][J] ← (avg[I][J] * Wi[I])
0:     end for
0:   end for
   {column wise summation and score calculation}
0:   for I ← 1 to COL do
0:     for J ← 1 to ROW do
0:       Sum ← (Sum + aw[I][J])
0:     end for
0:     score[J] ← Sum
0:     Sum ← 0
0:   end for
0:   return score
0: end function
=0

```

---

# Appendix B

## Categorical Data

### B.1 Agriculture, Forestry and Fishing

Table B.1: Agriculture, Forestry and Fishing

<b>Indicator-1</b>	<b>Description</b>
AFF	<b>Crop and animal production, hunting and related service activities</b>
	Growing of non-perennial crops
	Growing of cereals (except rice), leguminous crops and oil seeds
	Growing of rice
	Growing of vegetables and melons, roots and tubers
	Growing of sugar cane
	Growing of tobacco
	Growing of fibre crops
	Growing of other non-perennial crops
	Growing of perennial crops
	Growing of grapes
	Growing of tropical and subtropical fruits
	Growing of citrus fruits
	Growing of pome fruits and stone fruits
	Growing of other tree and bush fruits and nuts

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	Growing of oleaginous fruits
	Growing of beverage crops
	Growing of spices, aromatic, drug and pharmaceutical crops
	Growing of other perennial crops
	Plant propagation
	Animal production
	Raising of cattle and buffaloes
	Raising of horses and other equines
	Raising of camels and camelids
	Raising of sheep and goats
	Raising of swine/pigs
	Raising of poultry
	Raising of other animals
	Mixed farming
	Support activities to agriculture and post-harvest crop activities
	Support activities for crop production
	Support activities for animal production
	Post-harvest crop activities
	Seed processing for propagation
	Hunting, trapping and related service activities
	<b>Forestry and logging</b>
	Silviculture and other forestry activities
	Logging
	Gathering of non-wood forest products
	Support services to forestry
	<b>Fishing and aquaculture</b>
	Fishing
	Marine fishing
	Freshwater fishing

*continued on next page*

*continued from previous page*

	Aquaculture
	Marine aquaculture
	Freshwater aquaculture

## B.2 Mining and Quarrying

Table B.2: Mining and Quarrying.

Indicator-2	Description
MAQ	<b>Mining of coal and lignite</b>
	Mining of hard coal
	Mining of lignite
	<b>Extraction of crude petroleum and natural gas</b>
	Extraction of crude petroleum
	Extraction of natural gas
	<b>Mining of metal ores</b>
	Mining of iron ores
	Mining of non-ferrous metal ores
	Mining of uranium and thorium ores
	Mining of other non-ferrous metal ores
	<b>Other mining and quarrying</b>
	Quarrying of stone, sand and clay
	Mining and quarrying n.e.c.
	Mining of chemical and fertilizer minerals
	Extraction of peat
	Extraction of salt
	Other mining and quarrying
	<b>Mining support service activities</b>
	Support activities for petroleum and natural gas extraction

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	Support activities for other mining and quarrying
--	---

## B.3 Manufacturing

Table B.3: Manufacturing

Indicator-3	Description
MFG	<b>Manufacture of food products</b>
	Processing and preserving of meat
	Processing and preserving of fish, crustaceans and molluscs
	Processing and preserving of fruit and vegetables
	Manufacture of vegetable and animal oils and fats
	Manufacture of dairy products
	Manufacture of grain mill products, starches and starch products
	Manufacture of grain mill products
	Manufacture of starches and starch products
	Manufacture of other food products
	Manufacture of bakery products
	Manufacture of sugar
	Manufacture of cocoa, chocolate and sugar confectionery
	Manufacture of macaroni, noodles, couscous and similar farinaceous products
	Manufacture of prepared meals and dishes
	Manufacture of other food products n.e.c.
	Manufacture of prepared animal feeds
	<b>Manufacture of beverages</b>
	Distilling, rectifying and blending of spirits
	Manufacture of wines
	Manufacture of malt liquors and malt

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	Manufacture of soft drinks; production of mineral waters and other bottled waters
	<b>Manufacture of tobacco products</b>
	Manufacture of tobacco products
	<b>Manufacture of textiles</b>
	Spinning, weaving and finishing of textiles
	Preparation and spinning of textile fibres
	Weaving of textiles
	Finishing of textiles
	Manufacture of other textiles
	Manufacture of knitted and crocheted fabrics
	Manufacture of made-up textile articles, except apparel
	Manufacture of carpets and rugs
	Manufacture of cordage, rope, twine and netting
	Manufacture of other textiles n.e.c.
	<b>Manufacture of wearing apparel</b>
	Manufacture of wearing apparel, except fur apparel
	Manufacture of articles of fur
	Manufacture of knitted and crocheted apparel
	<b>Manufacture of leather and related products</b>
	Tanning and dressing of leather; manufacture of luggage, handbags, saddlery and harness; dressing and dyeing of fur
	Tanning and dressing of leather; dressing and dyeing of fur
	Manufacture of luggage, handbags and the like, saddlery and harness
	Manufacture of footwear
	<b>Manufacture of wood and of products of wood and cork, except furniture; manufacture of articles of straw and plaiting materials</b>

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	Sawmilling and planing of wood
	Manufacture of products of wood, cork, straw and plaiting materials
	Manufacture of veneer sheets and wood-based panels
	Manufacture of builders' carpentry and joinery
	Manufacture of wooden containers
	Manufacture of other products of wood; manufacture of articles of cork, straw and plaiting materials
	<b>Manufacture of paper and paper products</b>
	Manufacture of pulp, paper and paperboard
	Manufacture of corrugated paper and paperboard and of containers of paper and paperboard
	Manufacture of other articles of paper and paperboard
	<b>Printing and reproduction of recorded media</b>
	Printing and service activities related to printing
	Printing
	Service activities related to printing
	Reproduction of recorded media
	<b>Manufacture of coke and refined petroleum products</b>
	Manufacture of coke oven products
	Manufacture of refined petroleum products
	<b>Manufacture of chemicals and chemical products</b>
	Manufacture of basic chemicals, fertilizers and nitrogen compounds, plastics and synthetic rubber in primary forms
	Manufacture of basic chemicals
	Manufacture of fertilizers and nitrogen compounds
	Manufacture of plastics and synthetic rubber in primary forms
	Manufacture of other chemical products

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	Manufacture of pesticides and other agrochemical products
	Manufacture of paints, varnishes and similar coatings, printing ink and mastics
	Manufacture of soap and detergents, cleaning and polishing preparations, perfumes and toilet preparations
	Manufacture of other chemical products n.e.c.
	Manufacture of man-made fibres
	<b>Manufacture of pharmaceuticals, medicinal chemical and botanical products</b>
	Manufacture of pharmaceuticals, medicinal chemical and botanical products
	<b>Manufacture of rubber and plastics products</b>
	Manufacture of rubber products
	Manufacture of rubber tyres and tubes; retreading and rebuilding of rubber tyres
	Manufacture of other rubber products
	Manufacture of plastics products
	<b>Manufacture of other non-metallic mineral products</b>
	Manufacture of glass and glass products
	Manufacture of non-metallic mineral products n.e.c.
	Manufacture of refractory products
	Manufacture of clay building materials
	Manufacture of other porcelain and ceramic products
	Manufacture of cement, lime and plaster
	Manufacture of articles of concrete, cement and plaster
	Cutting, shaping and finishing of stone
	Manufacture of other non-metallic mineral products n.e.c.
	<b>Manufacture of basic metals</b>
	Manufacture of basic iron and steel

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	Manufacture of basic precious and other non-ferrous metals
	Casting of metals
	Casting of iron and steel
	Casting of non-ferrous metals
	<b>Manufacture of fabricated metal products, except machinery and equipment</b>
	Manufacture of structural metal products, tanks, reservoirs and steam generators
	Manufacture of structural metal products
	Manufacture of tanks, reservoirs and containers of metal
	Manufacture of steam generators, except central heating hot water boilers
	Manufacture of weapons and ammunition
	Manufacture of other fabricated metal products; metal-working service activities
	Forging, pressing, stamping and roll-forming of metal; powder metallurgy
	Treatment and coating of metals; machining
	Manufacture of cutlery, hand tools and general hardware
	Manufacture of other fabricated metal products n.e.c.
	<b>Manufacture of computer, electronic and optical products</b>
	Manufacture of electronic components and boards
	Manufacture of computers and peripheral equipment
	Manufacture of communication equipment
	Manufacture of consumer electronics
	Manufacture of measuring, testing, navigating and control equipment; watches and clocks
	Manufacture of measuring, testing, navigating and control equipment
	Manufacture of watches and clocks

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	Manufacture of irradiation, electromedical and electrotherapeutic equipment
	Manufacture of optical instruments and photographic equipment
	Manufacture of magnetic and optical media
	<b>Manufacture of electrical equipment</b>
	Manufacture of electric motors, generators, transformers and electricity distribution and control apparatus
	Manufacture of batteries and accumulators
	Manufacture of wiring and wiring devices
	Manufacture of fibre optic cables
	Manufacture of other electronic and electric wires and cables
	Manufacture of wiring devices
	Manufacture of electric lighting equipment
	Manufacture of domestic appliances
	Manufacture of other electrical equipment
	<b>Manufacture of machinery and equipment n.e.c.</b>
	Manufacture of general-purpose machinery
	Manufacture of engines and turbines, except aircraft, vehicle and cycle engines
	Manufacture of fluid power equipment
	Manufacture of other pumps, compressors, taps and valves
	Manufacture of bearings, gears, gearing and driving elements
	Manufacture of ovens, furnaces and furnace burners
	Manufacture of lifting and handling equipment
	Manufacture of office machinery and equipment (except computers and peripheral equipment)
	Manufacture of power-driven hand tools
	Manufacture of other general-purpose machinery

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	Manufacture of special-purpose machinery
	Manufacture of agricultural and forestry machinery
	Manufacture of metal-forming machinery and machine tools
	Manufacture of machinery for metallurgy
	Manufacture of machinery for mining, quarrying and construction
	Manufacture of machinery for food, beverage and tobacco processing
	Manufacture of machinery for textile, apparel and leather production
	Manufacture of other special-purpose machinery
	<b>Manufacture of motor vehicles, trailers and semi-trailers</b>
	Manufacture of motor vehicles
	Manufacture of bodies (coachwork) for motor vehicles; manufacture of trailers and semi-trailers
	Manufacture of parts and accessories for motor vehicles
	<b>Manufacture of other transport equipment</b>
	Building of ships and boats
	Building of ships and floating structures
	Building of pleasure and sporting boats
	Manufacture of railway locomotives and rolling stock
	Manufacture of air and spacecraft and related machinery
	Manufacture of military fighting vehicles
	Manufacture of transport equipment n.e.c.
	Manufacture of motorcycles
	Manufacture of bicycles and invalid carriages
	Manufacture of other transport equipment n.e.c.
	<b>Manufacture of furniture</b>
	Manufacture of furniture

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	<b>Other manufacturing</b>
	Manufacture of jewellery, bijouterie and related articles
	Manufacture of jewellery and related articles
	Manufacture of imitation jewellery and related articles
	Manufacture of musical instruments
	Manufacture of sports goods
	Manufacture of games and toys
	Manufacture of medical and dental instruments and supplies
	Other manufacturing n.e.c.
	<b>Repair and installation of machinery and equipment</b>
	Repair of fabricated metal products, machinery and equipment
	Repair of fabricated metal products
	Repair of machinery
	Repair of electronic and optical equipment
	Repair of electrical equipment
	Repair of transport equipment, except motor vehicles
	Repair of other equipment
	Installation of industrial machinery and equipment

## B.4 Wholesale and Retail Trade

Table B.4: Wholesales and Retail Trades, Repair Industries.

<b>Indicator-7</b>	<b>Description</b>
WRT	<b>Wholesale and retail trade and repair of motor vehicles and motorcycles</b>
	Sale of motor vehicles
	Maintenance and repair of motor vehicles

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	Sale of motor vehicle parts and accessories
	Sale, maintenance and repair of motorcycles and related parts and accessories
	<b>Wholesale trade, except of motor vehicles and motorcycles</b>
	Wholesale on a fee or contract basis
	Wholesale of agricultural raw materials and live animals
	Wholesale of food, beverages and tobacco
	Wholesale of household goods
	Wholesale of textiles, clothing and footwear
	Wholesale of other household goods
	Wholesale of machinery, equipment and supplies
	Wholesale of computers, computer peripheral equipment and software
	Wholesale of electronic and telecommunications equipment and parts
	Wholesale of agricultural machinery, equipment and supplies
	Wholesale of other machinery and equipment
	Other specialized wholesale
	Wholesale of solid, liquid and gaseous fuels and related products
	Wholesale of metals and metal ores
	Wholesale of construction materials, hardware, plumbing and heating equipment and supplies
	Wholesale of waste and scrap and other products n.e.c.
	Non-specialized wholesale trade
	<b>Retail trade, except of motor vehicles and motorcycles</b>
	Retail sale in non-specialized stores
	Retail sale in non-specialized stores with food, beverages or tobacco predominating
	Other retail sale in non-specialized stores
	Retail sale of food, beverages and tobacco in specialized stores
	Retail sale of food in specialized stores

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	Retail sale of beverages in specialized stores
	Retail sale of tobacco products in specialized stores
	Retail sale of automotive fuel in specialized stores
	Retail sale of information and communications equipment in specialized stores
	Retail sale of computers, peripheral units, software and telecommunications equipment in specialized stores
	Retail sale of audio and video equipment in specialized stores
	Retail sale of other household equipment in specialized stores
	Retail sale of textiles in specialized stores
	Retail sale of hardware, paints and glass in specialized stores
	Retail sale of carpets, rugs, wall and floor coverings in specialized stores
	Retail sale of electrical household appliances, furniture, lighting equipment and other household articles in specialized stores
	Retail sale of cultural and recreation goods in specialized stores
	Retail sale of books, newspapers and stationary in specialized stores
	Retail sale of music and video recordings in specialized stores
	Retail sale of sporting equipment in specialized stores
	Retail sale of games and toys in specialized stores
	Retail sale of other goods in specialized stores
	Retail sale of clothing, footwear and leather articles in specialized stores
	Retail sale of pharmaceutical and medical goods, cosmetic and toilet articles in specialized stores
	Other retail sale of new goods in specialized stores
	Retail sale of second-hand goods
	Retail sale via stalls and markets
	Retail sale via stalls and markets of food, beverages and tobacco products
	Retail sale via stalls and markets of textiles, clothing and footwear

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	Retail sale via stalls and markets of other goods
	Retail trade not in stores, stalls or markets
	Retail sale via mail order houses or via Internet
	Other retail sale not in stores, stalls or markets

## B.5 Transportation and Storage

Table B.5: Transportation and Storage.

Indicator-8	Description
TAS	<b>Land transport and transport via pipelines</b>
	Transport via railways
	Passenger rail transport, interurban
	Freight rail transport
	Other land transport
	Urban and suburban passenger land transport
	Other passenger land transport
	Freight transport by road
	Transport via pipeline
	<b>Water transport</b>
	Sea and coastal water transport
	Sea and coastal passenger water transport
	Sea and coastal freight water transport
	Inland water transport
	Inland passenger water transport
	Inland freight water transport
	<b>Air transport</b>
	Passenger air transport
	Freight air transport
	<b>Warehousing and support activities for transportation</b>

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	Warehousing and storage
	Support activities for transportation
	Service activities incidental to land transportation
	Service activities incidental to water transportation
	Service activities incidental to air transportation
	Cargo handling
	Other transportation support activities
	<b>Postal and courier activities</b>
	Postal activities
	Courier activities

## B.6 Information Technologies and Communication

Table B.6: Information Technologies and Communication.

Indicator-10	Description
IC	<b>Publishing activities</b>
	Publishing of books, periodicals and other publishing activities
	Book publishing
	Publishing of directories and mailing lists
	Publishing of newspapers, journals and periodicals
	Other publishing activities
	Software publishing
	<b>Motion picture, video and television programme production, sound recording and music publishing activities</b>
	Motion picture, video and television programme activities
	Motion picture, video and television programme production activities
	Motion picture, video and television programme post-production activities

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	Motion picture, video and television programme distribution activities
	Motion picture projection activities
	Sound recording and music publishing activities
	<b>Programming and broadcasting activities</b>
	Radio broadcasting
	Television programming and broadcasting activities
	<b>Telecommunications</b>
	Wired telecommunications activities
	Wireless telecommunications activities
	Satellite telecommunications activities
	Other telecommunications activities
	<b>Computer programming, consultancy and related activities</b>
	Computer programming activities
	Computer consultancy and computer facilities management activities
	Other information technology and computer service activities
	<b>Information service activities</b>
	Data processing, hosting and related activities; web portals
	Data processing, hosting and related activities
	Web portals
	Other information service activities
	News agency activities
	Other information service activities

## B.7 Financial and Insurance Activities

Table B.7: Financial and Insurance Activities.

Indicator-11	Description
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FIA	<b>Financial service activities, except insurance and pension funding</b>
	Monetary intermediation
	Central banking
	Other monetary intermediation
	Activities of holding companies
	Trusts, funds and similar financial entities
	Other financial service activities, except insurance and pension funding activities
	Financial leasing
	Other credit granting
	Other financial service activities, except insurance and pension funding activities, n.e.c.
	<b>Insurance, reinsurance and pension funding, except compulsory social security</b>
	Insurance
	Life insurance
	Non-life insurance
	Reinsurance
	Pension funding
	<b>Activities auxiliary to financial service and insurance activities</b>
	Activities auxiliary to financial service activities, except insurance and pension funding
	Administration of financial markets
	Security and commodity contracts brokerage
	Other activities auxiliary to financial service activities
	Activities auxiliary to insurance and pension funding
	Risk and damage evaluation
	Activities of insurance agents and brokers
	Other activities auxiliary to insurance and pension funding

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	Fund management activities
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## B.8 Professional, Scientific and Technical Activities

Table B.8: Professional, Scientific, and Technical Activity Data

Indicator-13	Description
PSTA	<b>Legal and accounting activities</b>
	Legal activities
	Accounting, bookkeeping and auditing activities; tax consultancy
	<b>Activities of head offices; management consultancy activities</b>
	Activities of head offices
	Management consultancy activities
	<b>Architectural and engineering activities; technical testing and analysis</b>
	Architectural and engineering activities and related technical consultancy
	Technical testing and analysis
	<b>Scientific research and development</b>
	Research and experimental development on natural sciences and engineering
	Research and experimental development on social sciences and humanities
	<b>Advertising and market research</b>
	Advertising
	Market research and public opinion polling
	<b>Other professional, scientific and technical activities</b>
	Specialized design activities
	Photographic activities

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	Other professional, scientific and technical activities n.e.c.
	<b>Veterinary activities</b>
	Veterinary activities

## B.9 Administrative and Support Services

Table B.9: Administrative and Support Services

<b>Indicator-14</b>	<b>Description</b>
ASSA	<b>Rental and leasing activities</b>
	Renting and leasing of motor vehicles
	Renting and leasing of personal and household goods
	Renting and leasing of recreational and sports goods
	Renting of video tapes and disks
	Renting and leasing of other personal and household goods
	Renting and leasing of other machinery, equipment and tangible goods
	Leasing of intellectual property and similar products, except copyrighted works
	<b>Employment activities</b>
	Activities of employment placement agencies
Temporary employment agency activities	
Other human resources provision	
<b>Travel agency, tour operator, reservation service and related activities</b>	<b>Travel agency, tour operator, reservation service and related activities</b>
	Travel agency and tour operator activities
	Travel agency activities
	Tour operator activities
Other reservation service and related activities	
<b>Security and investigation activities</b>	<b>Security and investigation activities</b>
	Private security activities

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	Security systems service activities
	Investigation activities
	<b>Services to buildings and landscape activities</b>
	Combined facilities support activities
	Cleaning activities
	General cleaning of buildings
	Other building and industrial cleaning activities
	Landscape care and maintenance service activities
	<b>Office administrative, office support and other business support activities</b>
	Office administrative and support activities
	Combined office administrative service activities
	Photocopying, document preparation and other specialized office support activities
	Activities of call centres
	Organization of conventions and trade shows
	Business support service activities n.e.c.
	Activities of collection agencies and credit bureaus
	Packaging activities
	Other business support service activities

## B.10 Other Activities

Table B.10: Other Service Activities.

<b>Indicator-21</b>	<b>Description</b>
OSA	<b>Activities of membership organizations</b>
	Activities of business, employers and professional membership organizations
	Activities of business and employers membership organizations
	Activities of professional membership organizations

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	Activities of trade unions
	Activities of other membership organizations
	Activities of religious organizations
	Activities of political organizations
	Activities of other membership organizations n.e.c.
	<b>Repair of computers and personal and household goods</b>
	Repair of computers and communication equipment
	Repair of computers and peripheral equipment
	Repair of communication equipment
	Repair of personal and household goods
	Repair of consumer electronics
	Repair of household appliances and home and garden equipment
	Repair of footwear and leather goods
	Repair of furniture and home furnishings
	Repair of other personal and household goods
	<b>Other personal service activities</b>
	Washing and (dry-) cleaning of textile and fur products
	Hairdressing and other beauty treatment
	Funeral and related activities
	Other personal service activities.

# Appendix C

## Data and Code

### C.1 Data for Economic Establishments

```
1 library('rvest')
2 url <- 'http://bangladeshyellowpages.com/search/
3 results.html?what=&where=sylhet'
4 webpage <- read_html(url)
5 node<- html_nodes(webpage , '.search-result')
6 data <- html_text(node)
7
8 fileConn<-file("cc.csv")
9 eedata <- writeLines(data, "cc.csv")
10 close(fileConn)
```

## C.2 Social Media Data for Occupation and DFA

```
1 install.packages("Rfacebook")
2 install.packages("RCurl")
3 install.packages("httpuv")
4 install.packages("httr")
5 install.packages("rjson")
6 install.packages("tm")
7 install.packages("wordcloud")
8 install.packages("textclean")
9
10
11 library("Rfacebook")
12 library("RCurl")
13 library("httpuv")
14 library("httr")
15 library("rjson")
16 library("tm")
17 library("wordcloud")
18 library("textclean")
19
20 access_token <-
21 EAAIS21AOiRABAAZADvqvOQysQDTHQShcCp0uQ4awpz1zQ5MSBj1rxub9E41i9RwWUMbp
22 OND6taK2aLqnHBrKtxZCkzUSD2kxYTBm0pOhpOuuGunwJZB18k2149dhc2WaQAoCnxZBt9Y
23 TZCHQyT1uCCzBIWuG5f9cgeXmTuUZBO1fEZBziBbrWhA3bLUB2SqG7oZD
24
25 Options(RCurlOptions =
26 list(verbose=FALSE,capath=system.file(CurlSSL,cacert.pem,package =
27 RCurl),ssl.verifypeer=FALSE))
28 me <- getUsers(me, token= access_token)
29 myFriends <- getFriends(access_token,simplify = FALSE)
30 fileConn<-file("cc.csv")
31 fb_data <- writeLines(myFriend, "cc.csv")
32 close(fileConn)
```

### C.3 User Selection from Social Media Data

```
1 file1<-file("user.csv")
2 file2<-file("area.txt")
3 x<- grep("hometown", readLines("area.txt"), value = TRUE)
4 if(x == 0){
5 file.remove("user.csv")
6 }
7 close(file1)
8 close(file2)
```

### C.4 Pre-processing of EE Data

```
1 fileConn<-file("dd.csv")
2 xx <- readlines("dd.csv")
3 xx<-tm_map(xx,removeNumbers, lazy=TRUE, 'mc.cores=1')
4 xx<-tm_map(xx,stripWhitespace, lazy=TRUE, 'mc.cores=1')
5 xx<-tm_map(xx,removePunctuation, lazy=TRUE, 'mc.cores=1')
6 k <- writeLines(xx, "ee.csv")
7 close(fileConn)
```

### C.5 Pre-processing of Social Media Data

```
1 File4<-file("ll.csv")
2 yy <- readlines("ll.csv")
3 yy<-tm_map(yy,removeNumbers, lazy=TRUE, 'mc.cores=1')
4 yy<-tm_map(yy,stripWhitespace, lazy=TRUE, 'mc.cores=1')
5 yy<-tm_map(yy,removePunctuation, lazy=TRUE, 'mc.cores=1')
6 yy<-tm_map(yy,strip_retweets, lazy=TRUE, 'mc.cores=1')
7 yy<-tm_map(yy,removeWords,stopwords(english), lazy=TRUE, 'mc.cores=1')
8 fb <- writeLines(yy, "ll.csv")
9 close(file4)
```

## C.6 Data Cleaning

```
1 install.packages("tm")
2 install.packages("stopwords")
3 library(tm)
4 library(stopwords)
5
6 fn="f3.txt"
7 connect <- file(fn,open="w")
8 tx1 <- readLines(connect)
9 tx2 <- tolower(tx1)
10 tx3 <- removeWords(tx2,stopwords())
11 low <- writeLines(tx3, con = connect, sep = "\n", useBytes = FALSE)
12 tx3
13 close(connect)
```

## C.7 EE Computation

```
1 nrow(read.csv("agriculture-forestry.csv", header= T/F))
2 nrow(read.csv("mining-quarrying.csv", header= T/F))
3 nrow(read.csv("manufacturing.csv", header= T/F))
4 nrow(read.csv("electricity-gas.csv", header= T/F))
5 nrow(read.csv("water-sewerage.csv", header= T/F))
6 nrow(read.csv("construction.csv", header= T/F))
7 nrow(read.csv("wholesale-retail.csv", header= T/F))
8 nrow(read.csv("transportation-storage.csv", header= T/F))
9 nrow(read.csv("accommodation-food.csv", header= T/F))
10 nrow(read.csv("information-communication.csv", header= T/F))
11 nrow(read.csv("financial-insurance.csv", header= T/F))
12 nrow(read.csv("real-estate.csv", header= T/F))
13 nrow(read.csv("professional-scientific.csv", header= T/F))
14 nrow(read.csv("administrative-support.csv", header= T/F))
15 nrow(read.csv("public-administration.csv", header= T/F))
16 nrow(read.csv("education.csv", header= T/F))
17 nrow(read.csv("health-social_work.csv", header= T/F))
18 nrow(read.csv("entertainment.csv", header= T/F))
19 nrow(read.csv("household-activities.csv", header= T/F))
20 nrow(read.csv("extraterritorial-organization.csv", header= T/F))
21 nrow(read.csv("other-activities.csv", header= T/F))
```

## C.8 Dictionary of Financial Words Using Social Media

```
1 file1<-file("user.csv")
2 file2<-file("word.txt")
3
4 x<- readLines("user.csv")
5 y<- readLines("word.txt")
6 z<- grep(x, y, value = TRUE)
7 if(z == 1){
8   count <- count +1 }
9 k <- writeLines(count, "cc.csv")
10 close(file1)
11 close(file2)
```

## C.9 Detecting Non-Residents

```
1 File4<-file("user.csv")
2 File5<-file("country.txt")
3 x<- grep("location", readLines("country.txt"), value = TRUE)
4 if(x == 1){
5   count <- count +1 }
6 k <- writeLines(count, "non-resident.csv")
7 close(file4)
8 close(file5)
```

## C.10 Employee Computation

```
1 File5<-file("user.csv")
2 File6<-file("employee.csv")
3 x<- grep("student", readLines("user.csv"), value = TRUE)
4 y<- grep("house_wife", readLines("user.csv"), value = TRUE)
5 if(x == 1) break
6 else if (y==1) break
7 else {count <- count +1 }
8 k <- writeLines(count, "employee.csv")
9 close(file5)
10 close(file6)
```

# Appendix D

## Sample Dataset

### D.1 Web Media Dataset

1 Name, Address, Contact

2 Hotel Herocity, "3/A, \_Sadarghat\_Road\_", "031\_612645, \_619630"

3 Hotel Ovisar (Pvt) Ltd., "Sea-Beach\_Road, \_Cox's\_Bazar\_", 0341-63061

4 Hotel Regent Park, "4/A, \_Jubilee\_Road, \_Chittagong\_", 031-2852072

5 Hotel Saint Martin Limited, "25, \_S.\_K.\_Mujib\_Road, \_Agrabad\_C/A\_", 725961-2

6 Hotel Abakash, Agrabad , "8821548, \_9899290"

7 Meridian Hotel & Restaurant, 1367 CDA Avenue , 031 654000-1

8 Motel Shaikat, Chittagong , 880-31-209845

9 Mistimela, "1205, \_SK\_Mujib\_Road\_", 031-714832

10 Pagasai Restaurant, "Sajib\_Plaza,\_(1st\_Floor), 12/A, \_Block\_-G, Halishahar\_Ho

11 Pizza Hut Bangladesh (Chittagong Outlet), "805/B, \_C.D.A\_Avenue, \_JB\_Complex

12 Restaurant (Ordinary Food), "Delight, \_Sadarghat\_", 616158

13 Swiss, "125, \_K.B.\_Fazil\_Kader\_Road.\_", 031 653096

14 Asian SR Hotel Chittagong Office, "291, \_Satation\_Road, \_Chittagong-4000, \_Ba

15 Duncan Products Ltd., "Ispahani\_Building\_(1st\_Floor)\_Sheikh\_Mujib\_Road, \_Ag

16 Fried Peppers Restaurant, 69 Agrabad C/A , "723407, \_2513030"

17 Genesis Hotel Raj Complex, "154, \_Kabi\_Nazrul\_Islam\_Road, \_Sadarghat\_", "031-

18 Hotel Agrabad, "P.O.\_Box\_#\_147, \_Post\_Code\_#\_4000, \_Agrabad\_", 031 713311-8

19 Hotel Al-Hera, "Hafiz\_Plaza\_151-152, \_Kapasgola\_Road\_", "031\_636947, \_612364"

20 Hotel Atlantic Ltd, "Zilla\_Porishad\_Super\_Market, \_Court\_Road\_", 031 621460

21 Hotel Broadway, 680 A. Jublee Road , 031 619795

22 Hotel Gateway, "82, \_Station\_Road\_", "031\_634341, \_0171861002"

23 Hotel Gateway, "82, \_Station\_Road\_", 031 634341

## D.2 Social Media Dataset

1 558523984664452,Masquer Rahman,0,Dhaka,0,CEAT AKKHAN LTD,  
2 Elysa's\_Eid\_Shopping\_Done\_and\_she\_wear\_her\_one\_of\_Eid\_dress\_in\_the\_market  
3 Good\_Morning\_from\_Bogra,,,,,  
4 Masquer\_Rahman\_checked\_in\_to\_Appollo\_Hospitals\_Dhaka\_with\_Torikul\_Udoy\_an  
5 Roaming\_on\_Eid-\_with\_Masquer\_Rahman\_and\_2\_others,,,,,  
6 Team\_Lunch\_with\_Masquer\_Rahman\_and\_Mahmoodul\_at\_Alfresco\_Tejgaon,,,,,  
7 Good\_Luck\_Team\_Bangladesh,,,,,  
8 On\_the\_way\_to\_kishorgonj,,,,,  
9 Masquer\_Rahman\_eating\_ifter\_with\_Raisa\_Tasnim\_Rahman\_and\_2\_others\_at\_Adda  
10 Alhamdulillah.\_Thnx\_Raisa\_Tasnim\_Rahman\_for\_all\_hard\_work\_for\_preparing\_del  
11 "Sharif\_Islam\_eating\_pizza\_with\_Masquer\_Rahman\_at\_Bella\_Italia,\_gulshan1"  
12 CEAT\_Annual\_Picnic,,,,,  
13 Darjeling\_Mall,,,,,  
14 Birthday\_Celebration\_in\_office\_&\_Dinner\_with\_Family\_without\_my\_princess\_a



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