

5-2017

Location-based Modeling and Statistics Internship with Location, Inc.

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Location-based Modeling and Statistics Internship with Location, Inc.

Location-based Modeling and Statistics Internship with Location, Inc.

Yixuan Wang

Degree will be conferred May 2017

A GISDE final project paper

submitted to the faculty of Clark University, Worcester, Massachusetts,

in partial fulfillment of the requirements for the degree of

Masters of Science in Geographic Information Sciences for Development and Environment

in the Department of International Development, Community, and Environment

Accepted on the recommendation of

Dr. Yelena Ogneva-Himmelberger, Project Advisor

Abstract

Location-based Modeling and Statistics Internship with Location, Inc.

Yixuan Wang

In the summer 2016, I had the honor to work as a GIS & statistics intern at Location, Inc. It is a Software as a Service (SaaS) private company that builds and delivers location-based big data, analytics, tools, and services. The company's mission is to create the best location-based technologies to help businesses make well-informed decisions (*Location, Inc.*). I was chosen by the company's CEO, Dr. Andrew Schiller, who is also a Clark Alumnus. During the internship period, I worked in the data science team supervised by Jonathan Glick, and my main responsibilities were handling geographical and statistical related tasks to support project development. In general, I participated in two projects and worked on many different tasks such as tabular data cleaning, descriptive statistics analysis, geo-datasets development, spatial analysis, etc. The samples of work I describe here are processing TransUnion datasets and network analysis in ArcMap. I have learned a few new skills such as processing big data or using basic SPSS syntax, and gathered wider experience from the work. Overall, it was a wonderful experience for me, and no doubt it would shape my future career. I would definitely recommend this internship opportunity to other GISDE students.

Dr. Yelena Ogneva-Himmelberger, Project Advisor

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Dedication

This paper is dedicated to my beloved parents, who are always supporting me to pursue what I love. I also dedicate this paper to my dear girlfriend, Lei, who has been striving together with me since college.

Acknowledgements

I want to specially thank Dr. Andrew Schiller for providing me such a wonderful internship position. I also want to thank Jonathan Glick for tutoring me during the internship.

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Chapter 1. Introduction

Geographic Information Science for Development and Environment (GISDE) master program at Clark University is a powerful and unique GIS program, and it shapes both my study and career paths. We develop expertise in geospatial techniques, remote sensing, and advanced spatial analysis, focusing on various fields, such as land change, biological conservation, earth system modeling, public health, urban development, etc. This program offers not only a number of high value courses and seminars but also several tracks to fulfill our different needs, including internship track, research track, as well as portfolios track. As most of my colleagues, I chose the internship track, which provides opportunities to apply our knowledge and skills learnt from classes directly to the real world, and more importantly, to prepare for careers after graduation.

In the summer 2016, I had the honor to join the local geographic analytics company, Location, Inc. I was chosen by the company's CEO, Dr. Andrew Schiller, which is also a Clark Alumnus. I worked as a GIS & statistics intern in the data science team supervised by Jonathan Glick. During the internship period, I participated in two projects and worked on many different tasks such as tabular data cleaning, descriptive statistics analysis, geo-datasets development, spatial analysis, etc. I applied my knowledge and skills to contribute to the company and meanwhile I have learned a few new skills and gathered experience from the work. Overall, it was a wonderful experience for me, and no doubt it would shape my future career.

This internship paper is a general report for my 3-month location-based modeling and statistics internship at Location, Inc., which includes the company's description, overall internship responsibilities, samples of work, as well as internship assessment.

Chapter 2. Description of the Organization

2.1 Background

Location, Inc. is a Software as a Service (SaaS) private company that builds and delivers location-based big data, analytics, tools, and services. Its headquarters is located in the downtown of Worcester, the heart of Massachusetts. The company was established by Dr. Andrew Schiller, a geographer and demographer, when he was pursuing his Ph.D. degree in geography at Clark University in 2000. He was inspired by his thought that people could spend much less time finding new locations and neighborhoods in which to live if they could search out the ideal imagined neighborhood based on specific criteria they input. Two years later, his idea had come to the real world in the form of the first and the most renowned product he created, NeighborhoodScout®, a web-based neighborhood search engine. Since then, the company has been keeping rapidly developing and innovating new services and products. Today, it serves over 55 million individuals and businesses with its own exclusive data and products (*Location, Inc.*).

2.2 Mission and Expertise

The company's cutting-edge algorithms and models provide risk assessment and management, site selection, market analysis, and real estate investments for all kinds of enterprises within the United States. The company's mission is to create the best location-based technologies it can that help businesses make well-informed decisions, which could transform into greater revenue and reduced costs (*Location, Inc.*). "Our mission is simple. What we do is not." Just like the words in its website describing themselves, the company owns quite a few

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exclusive or patented location-based technologies which have been refined over a decade of research.

Location, Inc. currently has two major web-based products: SecurityGauge® and NeighborhoodScout®. SecurityGauge® is a crime risk assessment tool that is able to evaluate the risk of crime for any address in the U.S. immediately with extremely high accuracy based on reported crime incident data from over 18,000 law enforcement agencies in the U.S. (*Location, Inc.*). NeighborhoodScout® provides instant, and comprehensive neighborhood reports for any address in the U.S., which contains about 300 statistics including demographic statistics, lifestyle, house value, public schools, and crime statistics. It is recognized as the largest database of neighborhood statistics available today. The company also provides a number of standalone location-based data products including crime risk data, fire risk™ data, fraud risk/ecommerce loss prevention, real estate, house values & appreciation rates, school data and ratings, and lifestyle and demographics data. In addition, it offers geo-fence boundary products such as neighborhood boundaries, zip code boundaries or city & metro boundaries.

With advanced technologies and abundant experience of location-based analytics, the company offers comprehensive industry solutions serving different sectors from government to academia and from small business to big corporation (*Location, Inc.*). It covers real estate, retail, fraud detection, insurance, relocation, direct marketing, and site selection. Although all the products and services currently are for the U.S. only, the company intends to cover their business in Canada in the near future, and it has already initiated developing and analyzing crime data in Canada.

2.3 Organizational Structure and Culture

When Dr. Andrew Schiller first started Location, Inc. in 2000, it had only two people. Now, it has around 15 employees and the majority of them are in full-time position. Gender composition is quite balanced, namely male and female staff are nearly half-and-half. The management team consists of the CEO and the chair of the board, Dr. Andrew Schiller, a vice president, sales, services and operations managers. The company has three departments: sales, services, and development, which includes web engineering team and data science team (see Figure 1). Both the web engineering team and the data science team are the core of productivity in the company. The primary responsibilities of the web engineering team are website maintenances and development as well as data management. Since many products and services are delivered on the internet, the web engineering team is the backbone of the company. The data science team is led by an outstanding statistician, Jonathan Glick, who has been working for the company for many years. He and the data science team are the muscle and blood of the company, which focuses on analyzing data, constructing algorithms and models for tools and products.

The culture of in the company tends to be informal. Since the company and the departments are relatively small, staff communication either within a department or among different departments is easy. Normally, every workday at noon, the staff members gather together along with the CEO in the conference room to have lunch. Sometimes, the whole company staff members go to a restaurant nearby to have a bigger lunch usually on someone's birthday or a special day. Every week, the company is likely to have a happy hour in a bar nearby. Dress code in the company is not very restricted, so business casual or even casual are

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generally encouraged. Moreover, the company is a dog friendly place so some staff members bring their lovely dogs into the office.

Although the size of Location, Inc. is not large, implementations are relatively simpler and faster than at large companies or government departments. Since it was established, the company has been rising gradually. It has increasing number of products and services, serving more and more businesses and individuals, and the reputation has reached a higher level. All these successes could not have been achieved without the leadership and innovation of Dr. Andrew Schiller. The products and the services he and the whole company created have very little competition because they are unique and powerful. With the successful products like the NeighborhoodScout® and the SecurityGauge®, the company has increasing revenues and draws more investments to develop more new products.

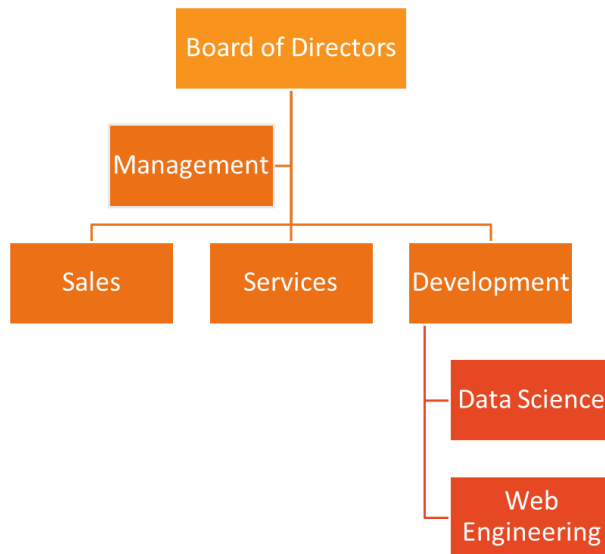


Figure 1. The structure of Location, Inc.

2.4 GIS and Mapping

Since the business of Location, Inc. is all about location-based analytics, GIS is used massively within the company. ArcGIS software is generally used to develop useful data in the data science team. For most of the services and the products, GIS is highly involved in developing data, especially for some basic spatial analysis tools. In the NeighborhoodScout®, geocoding using Google Maps Geocoding API is applied in early stage of data development for data that only contain address information. Most data require precise x and y coordinates to be used later. Spatial join is another GIS tool that is normally used. For example, some data that is in a very small scale are needed to be joined into the census tract scale in order to be further analyzed. Geovisualization is another big part. Web mapping is a part of the final report in the NeighborhoodScout®. The crime product SecurityGauge® requires surface interpolation to show crime map reports based on crime data. Network analysis is recently utilized for a developing product, for instance, calculating driving time to jobs from the centroids of the block group.

Chapter 3. Description of Internship Responsibilities

3.1 Overall Responsibility

As a GIS & statistics intern, I worked in the data science team supervised by the lead statistician, Jonathan Glick, who was always working remotely in Seattle, WA. I communicated with him through skype every workday. In general, he directly assigned tasks to me and usually provided helpful statistical instructions. I also came to him with advice or questions when I encountered some issues on the tasks. We used JIRA Software to track our individual tasks and project flows. Once I finished a task, I reported my work to him directly via Dropbox. He would then evaluate the task and let me know if it needs to be improved. In addition, I was obligated to report processes to the CEO Dr. Schiller often. My main responsibilities were handling geographical and statistical related tasks to support project development. More specifically, they included cleaning, merging, and aggregating tabular data sets, running descriptive statistics in SPSS, cleaning and developing geodatabases, performing spatial analysis in ArcMap, as well as geoprocessing datasets using spatial join tool, clip tool, merge tool, etc. With regards to data, all original data were provided by either Jonathan or Andrew which were initially downloaded from websites or servers or were sent from other companies. Once I had the data, I would start with preprocessing to make the data useful in the following analysis. During the summer, I was partially involved in two large projects, the first one was creating new variables for a new model of housing value prediction project, and the second one was processing FEMA fire datasets and RealtyTrac datasets for the second version of structure fire assessment model project. The first project was in the full development stage, and the second one was only in the pre-development stage. In the following section, I will give more details about the work I did for the first project.

3.2 Samples of Work

During the internship period, I completed a number of tasks, some of them were relatively simple and several of them were challenging. Here I will describe three tasks that were relatively time consuming and challenging.

3.2.1 Processing TransUnion Datasets

The first task of my internship was to process TransUnion datasets. Dr. Schiller handed over a “black box” to me on my first day, which was a book-size external hard drive that contained TransUnion datasets including over 500 credit variables for the entire US. The assignment was examining and familiarizing myself with the datasets and then aggregating it from current zip+4 level (see Figure 2) into both block level and block group level. The original datasets were extremely large and separated into 101 sub files. Each file was approximately 2 gigabytes with no file-type extensions. The entire tabular dataset had about 550 columns and over 39 million rows. The file type was fixed width text format, which is not very friendly to data processing. I noticed that the first step was to convert the fixed width text format into the comma delimited csv format which is a very common and friendly tabular data format. With the help of the web engineering team, I learned how to operate the Terminal in the Linux Operation System installed in the Virtual box software. The csv kit, which is a command tool in the Terminal, is able to convert fixed width text format into the comma delimited csv format. A layout reference schema file which recorded how many spaces for each column was required in order to complete format change. Fortunately, this layout sheet was provided along with the original data. A single line Shell Script to execute csv converting function in the Terminal was

needed to convert all original files into comma delimited csv files. Due to the size of the data, this process took about 2 hours to finish. After transformation, the next step was to merge all sub files. Another single line of command would do the merging. In order to aggregate from zip+4 level into either block level or block group level, I first needed such georeferencing data to be joined. Dr. Schiller also provided me with that zip+4 georeferencing data which contained the block ID field and the block group ID field from GeoLytics database. This georeferencing data was relatively small, only about 4 gigabytes. The ideal technique was to join this smaller georeferencing data into the big TransUnion table with the same unique key field: zip+4 ID. The size of the data was too large so that I could not find an efficient way to perform merging. After some research, experiment, and failure, I finally found a relatively fast and simple, but may not be the smartest way to merge them. I imported both of the datasets into SPSS and first sorted them based on the zip+4 ID. It is always a good idea to sort first when dealing with big data process because it could save a number of time in later data process such as merge or aggregate. Next, I only extracted the sorted zip+4 ID field from the TransUnion datasets and joined the georeferencing data into it. In this case, I kept all records of zip+4 ID from the TransUnion data and only joined matched georeferencing data with common zip+4 ID. Then, I hard-joined this semi-joined georeferencing data into the large TransUnion data with all the variables since they have the same order of the zip+4. Finally, I aggregated this joined data by choosing median values into block level as well as block group level by using the aggregate tool in SPSS. After aggregating, I performed the data process evaluation and found that about 70 percent of the original data were successfully aggregated at the end. The reason why some of the original data were not able to be aggregated was that there were some differences in the zip+4 field between the datasets.

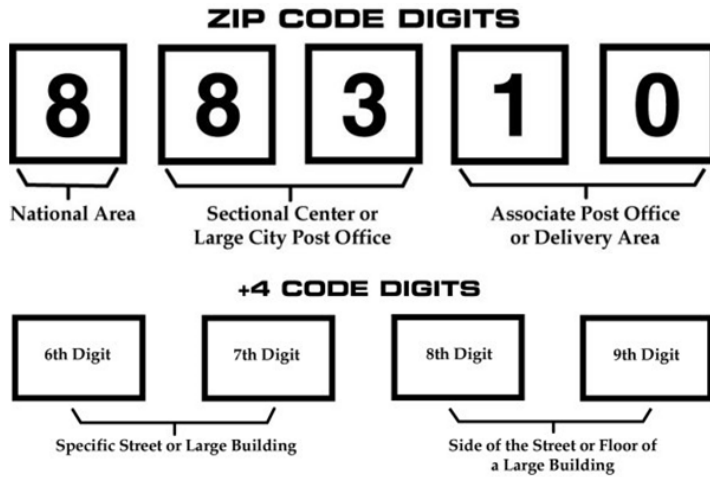


Figure 2. An illustration of zip+4 digits.

3.2.2 Network Analysis in ArcMap

Network analysis in ArcGIS helps solve spatial problems based on network attributes such as roads. This entire network analysis task was to help building two variables for a new model of housing value prediction project. One was calculating how many jobs in census blocks that can be reached by different driving times from the centroid of each block groups in the US, and the other was calculating how many rail transit stations that can be accessed by various walking times also from the centroid of each block groups in the US. To finish either of the tasks, I needed to first create a nation-wide network dataset in ArcMap. Jonathan provided me with a North American street layer to get started.

3.2.2.1 Calculating Number of Jobs That Can Be Reached by Different Driving Time Zones

First, I extracted US streets only into a new file geodatabase by using the select by attribute tool in ArcMap. Then, I needed to create a new field that calculated time needed to drive over a certain segment of a street before I created the network datasets. By examining the attribute table of the US street layer, I found out there were a speed category field and of course a distance field that I could use to figure out the time field. By referring the metadata, I applied the mean speed of each speed class to get the time. In order to simulate the real world situation, Jonathan also suggested using a travel time index for all urban areas to take into account traffic congestion. The travel time index was acquired from the University of Texas. It was an index showing the level of congestion in urban areas. If the traffic did not have any congestion in an urban area, the travel time index is 1. The more severe the traffic congestion, the higher the travel time index. Therefore, the time field in second unit was calculated by dividing the length of a road segment in meters by the mean travel speed in meters per second and then multiplying by the travel time index. Now, a new network dataset could be built based on this modified street layer by using the time field “second” as impedance in the new network dataset wizard. It took about 8 hours to configure the entire national street network dataset. After the network dataset was complete, I could start to analyze new service areas in the network analyst toolbar. Next step was to create a new network service area in the network toolbar and import all the centroids of the block groups as facilities. The driving time zones we wanted to generate were 5 minutes, 10 minutes, 15 minutes, 30 minutes, 45 minutes, and 1 hour (see Figure 3). After the analysis was finished on each driving time service area, I exported it into the shapefile. Once exported into the shapefile, I applied spatial join tool to spatially join jobs point layer into each driving time service area. The variable “number of jobs” in block level data were provided by Jonathan (see Figure 4). The results in the attribute table contained block group ID, driving time zone, number

of jobs, and number of high pay jobs (see Figure 5). For each output, I also performed quality checks to make sure the results are reliable and correct. Once all analysis was completed, I uploaded all results in shapefiles into Dropbox.

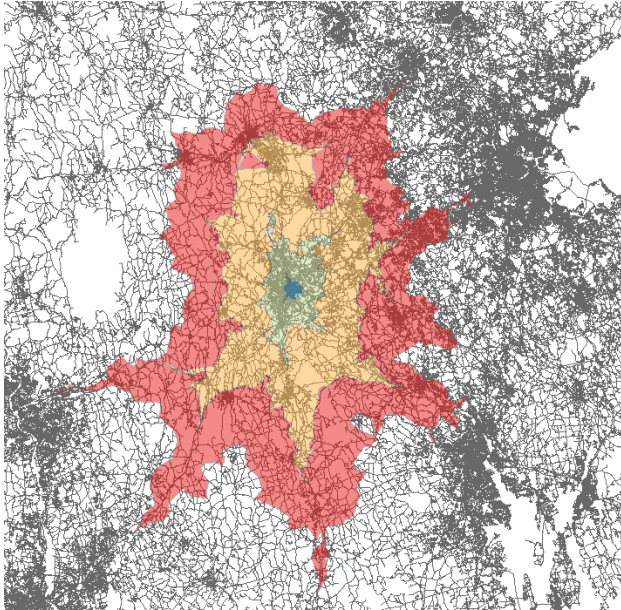


Figure 3. A sample shows 5minutes, 10minutes, 15minutes, and 30 minutes driving time service areas from a block group centroid in Worcester, MA. Vehicles are able to travel more distance within the same time to Connecticut than to Boston due to the travel time index is larger in Boston urban area.

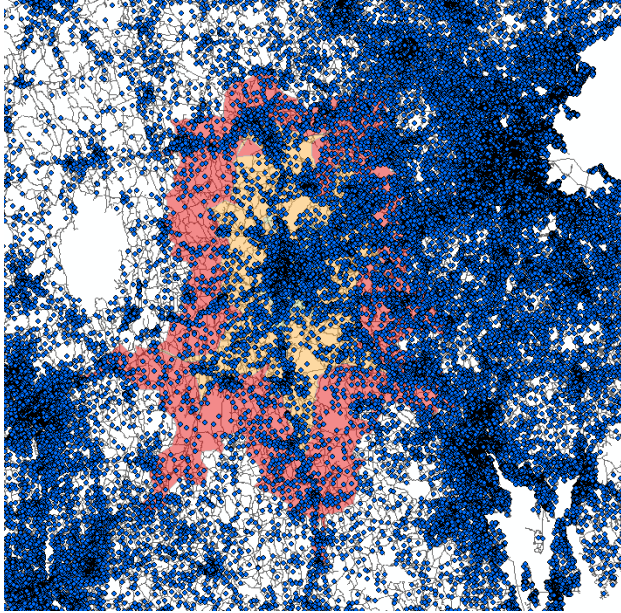


Figure 4. A sample shows jobs points based on block centroids in Worcester-Boston area.

FID	Shape *	Join_Count	TARGET_FID	ObjectID	FacilityID	Name	FromBreak	ToBreak	Name_ID	jobs	jobs_hipay	GEOID
0	Polygon	345	0	6826	14873	Location 14873 : 0 - 300	0	300	Location 14873	12773	3279	1500000US060730220001
1	Polygon	61	1	6827	27003	Location 27003 : 0 - 300	0	300	Location 27003	5824	2277	1500000US060730174012
2	Polygon	124	2	6828	23146	Location 23146 : 0 - 300	0	300	Location 23146	5452	2059	1500000US060590422011
3	Polygon	38	3	6829	25143	Location 25143 : 0 - 300	0	300	Location 25143	3696	971	1500000US06059046001
4	Polygon	185	4	6830	22067	Location 22067 : 0 - 300	0	300	Location 22067	6440	1556	1500000US060373004005
5	Polygon	23	5	6831	19912	Location 19912 : 0 - 300	0	300	Location 19912	4571	2826	1500000US060371064061
6	Polygon	7	6	6832	29091	Location 29091 : 0 - 300	0	300	Location 29091	495	123	1500000US06050419102
7	Polygon	131	7	6833	29563	Location 29563 : 0 - 300	0	300	Location 29563	17529	5981	1500000US060710072001
8	Polygon	113	8	6834	20722	Location 20722 : 0 - 300	0	300	Location 20722	5710	1277	1500000US060374016013
9	Polygon	144	9	6835	28842	Location 28842 : 0 - 300	0	300	Location 28842	48809	6960	1500000US060710013093
10	Polygon	189	10	6836	14696	Location 14696 : 0 - 300	0	300	Location 14696	7564	1640	1500000US060710020131
11	Polygon	31	11	6837	12924	Location 12924 : 0 - 300	0	300	Location 12924	1271	255	1500000US060710118003
12	Polygon	141	12	6838	23761	Location 23761 : 0 - 300	0	300	Location 23761	3879	1012	1500000US060710120022
13	Polygon	2	13	6839	26439	Location 26439 : 0 - 300	0	300	Location 26439	17	2	1500000US060470020002
14	Polygon	0	14	6840	18189	Location 18189 : 0 - 300	0	300	Location 18189	0	0	1500000US060190083023
15	Polygon	107	15	6841	29253	Location 29253 : 0 - 300	0	300	Location 29253	8102	2572	1500000US061130112051
16	Polygon	78	16	6842	18272	Location 18272 : 0 - 300	0	300	Location 18272	1301	334	1500000US060210102002
17	Polygon	39	17	6843	118724	Location 118724 : 0 - 300	0	300	Location 118724	1784	683	1500000US320199601032
18	Polygon	7	18	6844	118723	Location 118723 : 0 - 300	0	300	Location 118723	2204	708	1500000US320199601031
19	Polygon	46	19	6845	30615	Location 30615 : 0 - 300	0	300	Location 30615	737	154	1500000US060890122003
20	Polygon	11	20	6846	162115	Location 162115 : 0 - 300	0	300	Location 162115	272	92	1500000US410290029005
21	Polygon	15	21	6847	163217	Location 163217 : 0 - 300	0	300	Location 163217	958	360	1500000US410390003001
22	Polygon	51	22	6848	163349	Location 163349 : 0 - 300	0	300	Location 163349	3651	760	1500000US410470103032
23	Polygon	13	23	6849	164429	Location 164429 : 0 - 300	0	300	Location 164429	1497	445	1500000US410510105001
24	Polygon	14	24	6850	208195	Location 208195 : 0 - 300	0	300	Location 208195	2602	1497	1500000US530110404112
25	Polygon	3	25	6851	207538	Location 207538 : 0 - 300	0	300	Location 207538	6	0	1500000US530150020021

Figure 5. A sample result shows number of jobs covered by 5 minutes driving time from centroid of block group.

3.2.2.2 Calculating Number of Rail Transit Stations That Can Be Reached by Different Walking Time Zones

After the driving times zones had been completed, I was tasked with calculating the number of rail transit stations that can be reached by different walking time zones, including 5 minutes, 10 minutes, 15 minutes, 20 minutes, as well as 30 minutes. This process was very similar to the previous one. First, I needed to calculate a new time field in the built network dataset. I needed to calculate how much time persons need to travel through each segment of roads. Based on the research found online, the average human walking speed is 3 miles per hour, which is about 1.34 meters per second. Therefore, the time field is calculated by dividing the length of a road segment in meter by the average human walking speed in meters per second. After finishing the configuration, the network dataset needed to be rebuilt in order to update it. It took a much shorter time to rebuild the network dataset than to create a new one. Once the updated network dataset was complete, I created a new service area and import all the centroids of the block groups just like the previous one. I computed 5 minutes, 10 minutes, 15 minutes, 20 minutes, and 30 minutes walking zones. The national rail transit station shapefile provided by Jonathan, which contains all Amtrak stations, local light railway stations, and subway stations was used to be spatially joined with each walking zone. At the end, I preformed quality checks as well to check the reliability of the results. The results were uploaded into Dropbox after the results were determined as reasonable.

3.3 Connection to the Mission

My responsibilities and work I accomplished were all connected to the company's mission. The data science team is always dedicating to improve and design algorithms, analysis,

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and models. As a member in the data science team, developing new datasets or making new variables for a new model were definitely counted as the dedication. The work I have done and my daily responsibilities were all efforts to create better location-based technologies that help businesses to make well-informed decisions.

Chapter 4. Assessment of the Internship

4.1 Learning during the Internship

In general, I did have a great learning experience in the internship period at Location, Inc. I learned how to utilize Linux Shell Scripts to process tabular data in batch and got familiar with the csv kit operated in the Terminal. They were very handy for dealing with big data. Also, I learned basic SPSS syntax which was very useful in importing and analyzing data. In addition, because of some very large shapefiles, I figured out how to download and use the ArcGIS 64-bit Background Geoprocessing Tool. This tool was able to let the computer use all available memory when dealing with big data in ArcMap. This was a big saver when I encountered memory limit errors running the spatial join tool. Moreover, our company trained us to use JIRA software to manage and track tasks or projects, which was a very useful platform. In addition, I learned how to work with other teams, and how to work for a supervisor that is always working remotely.

4.2 Useful Skills Acquired at Clark University

Most classes I took in my first year at Clark University were environment and raster analysis related such as Landscape Ecology or Wildlife Conservation Seminar. However, the internship I participated in was more focused on demographic and vector analysis. So I have to say that most of the hard skills learned at Clark University were not applied during this internship. Yet, I did get some useful soft skills at Clark University that were able to help in the internship. For example, I learned how to participate in a large projects and cooperate with team members in the Wildlife Conservation Seminar instructed by Dr. John Rogan and Dr. Florencia

Sangermano, which I found very important in the internship. Furthermore, my experience at Clark has taught me to think and solve problems differently and smartly. This ability allowed me working through the difficult tasks smoothly during the internship.

4.3 Relating to Studies and Goals

In general, my study focuses on GIS applications and my goal is applying geospatial techniques to let more people realize the power of GIS. Therefore, this internship was quite appropriate for my study interests and goals. The internship allowed me to touch and process big data and even enjoy dealing with them. It also lets me realize that these location-based data are quite powerful. Moreover, I realized the importance of scripting when dealing with either tabular data processing or spatial analysis. Therefore, I took Dr. Jie Tian's Python class as well as Computer Programming for GIS class in the fall 2016 semester. Since this internship, I have come to enjoy Vector GIS and demographic data. In the future studies, I would like to work on both environmental GIS and demographic GIS.

4.4 Recommendation to Others

Location, Inc. is a local company and its atmosphere is harmonious. It would always be a pleasure for me to work here. It is a great place for a GISDE or geography student to apply concepts to solve real world problems, and a unique place to practice skills on big data processing. I would definitely recommend this internship opportunity to other GISDE students.

Chapter 5. Conclusion

I appreciate that Dr. Andrew Schiller, the CEO of Location, Inc. offered me this wonderful opportunity to work as a GIS & statistics intern at Location, Inc. this summer. I have learned many new and useful skills in terms of tabular data processing as well as spatial analysis. This internship provided me a full angle of view to look at the challenge of big data processing and the power of location-based data. I was glad that I was able to apply my 4 years of GIS experience and skills to real world situations. This internship was a different way of learning which afforded me abundant new experience that would be very valuable for future studies as well as careers.

Reference

Location, Inc. Location, Inc. www.locationinc.com. Accessed 20 Sep. 2016.