

Big Data Analytics for Municipal Waste Management

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ABSTRACT

Waste management is becoming a greater concern for cities and municipalities around the world because of the continual increase in population and waste. Big data analysis has the potential to not only help assess the current waste management strategies but also provide information that can be used to optimize the systems used in various institutions, local governments, and companies, among others.

KEYWORDS

Waste Management, Big Data, Local Government, HID305, HID319

1 INTRODUCTION

In the current fast-paced society, production of goods continues to increase, and new distribution chains continuously change. The generation of waste and the deprecated goods -from now on referred to as solid waste- has increased over the past ten years, rising from approximately 0.64 kg per person per day of solid waste to approximately 1.2 kg per person per day. Projections estimate that this number is expected to increase to about 1.42 kg by 2025 [15]. This continual change and increase make waste management a more complex and more intensive endeavor.

While the quantity of waste itself can and should be reduced by conscious use and discipline in recycling and reusing items, local governments and waste management companies can also make modifications to their systems to reduce waste in the actual collection and disposal.

Because of this, different local governments and organizations have recognized the need to develop more elaborate regulations to control the different features, segments, processes involved in waste disposal from the moment the material is discarded by the consumer to the moment the material reaches its ultimate destination, such as a recycling plant or landfill. This set of systematic regulations is called solid waste management. Solid waste management has changed over time. What used to be systems designed and implemented based on local needs and convenient disposal as moved into extensively researched and implemented management systems that consider complex multivariable and dynamic sources of data [1].

Currently, data used in waste management is collected from many sources and varies depending on the types of solid waste and the rate of disposal of a particular population. Because of the diversity of data available (including types of waste, weight collected, the location of collection and disposal), the quality of the data must be continually monitored and assessed. [6]. New management systems that seek to optimize waste management must collect large volumes of data from various data sources on a daily basis in order

to compile information essential for optimization. Multivariate data analysis methods provide exploratory data analysis, classification and parameter prediction using this data [4].

2 MUNICIPAL SOLID WASTE MANAGEMENT

There are various factors that contribute to the complexity of optimizing waste management systems and the types of data needed for analysis to do so. These include differences in composition of waste, environmental and logistic needs for individual communities, the changing forms of handling various materials such as recyclables, and the various ways that waste can be disposed of.

There are significant differences between the general composition of the waste generated in rural areas and the waste produced in urban areas. The waste produced in the later is profoundly influenced by the culture and the practices of our modern society. This distinction means that there will inevitably need to be differentiation in waste management practices and systems based on each communities needs, commerce, economy, and practices [6].

Municipal Solid Waste (MSW), commonly known as garbage or trash, consists of items that residential, commercial, institutional and industrial sites generate. This is much different from the type of waste produced, for example, in an agricultural community. Urban, or municipal, waste can be the result of everyday activities, such as leftover food, plastic bottles, packaging material, wooden furniture, electrical and electronic appliances, glass, medical waste, card boards, waste tires, office wastes, consumer goods, among others. Each type of waste needs to be handled in distinct manners.

The amounts of solid waste and its composition vary depending on the country, place, and activity performed at the site where the waste is generated [6]. For this reason, every process related to waste management (transportation, storing and final disposition, among others) must be engineered and tailored to fit the specific needs of community, organization, or local institution. The data collected for analysis will also be specific to the needs and realities of each entity, which is why waste management plans are not universal or easily applied from one community to another.

The forms of handling waste have also changed over the years, affecting the best ways to manage waste. One example of this is the increasing ability to separate and recycle various materials. Data on the recyclable material, dividing waste and recyclable material, how materials are sent or how they are disposed of has become significant. According to EPA statistics from 2014, Americans generated about 258 million tons of MSW of which more than 89 million tons is recycled and composted. This is an equivalent to a 34.6% recycling rate compared to 6.4% in 1960. With the increase in recycling capabilities and consciousness, adaptations must be made to the management systems [9].

Another variable that is how waste is disposed of. Americans have used the energy production process to combust approximately 33 million tons of waste, while as much as 136 million tons of waste ended up in landfills during the same year [9]. Local governments must determine which form of waste disposal is more efficient and cost-effective based on their unique context. They must take into consideration access to and distance from disposal facilities, types of waste being disposed of, and specific environmental implications of each process based on their geographic, geological, and environmental context.

In waste management, decision-makers are and will continue to be forced to make choices. When they develop or implement plans, they have a choice as to what information they will or will not consider for analysis. They have to choose what factors are influential in their jurisdiction. They make choices about routes, resources, and all the details between pick up to disposal. These choices can be classified as fortuitous, good, or optimal [1].

Fortuitous decision-making has no scientific base; the person who is in charge of making decisions must always try to solve the problem with little or no research or data. On the other hand, good decision-making is primarily based on experience, comparison of elements and trial and error. Optimal decision making, however, requires understanding and analysis of techniques and technologies provided by other fields [1].

This is where Big Data comes in.

3 BIG DATA AND WASTE MANAGEMENT

Big data has the capacity to facilitate analysis of information so that better decisions can be made.

In Big Data, the expert in charge takes many data sets coming from diverse and dynamic data sources and applies technologies to analyze these data sets. Authors of papers and books like "The Fourth Paradigm" state that big data exploration works to find patterns in data by analyzing the trends and outliers found in the data sets mentioned above, to generate knowledge [14].

Characteristics (such as the large volume of data generated in waste management, how dynamically the data is generated, and the variety of formats in which the data comes) make the task of producing knowledge an ideal task for Big Data. Scientists can interpret the findings of Big Data in a way that allows individuals and institutions involved in waste management to make optimal decisions [20].

4 SOLUTIONS FOR EFFECTIVE WASTE MANAGEMENT

4.1 Examples of Big data and waste management

Various institutions around the world have explored and implemented Big Data analytic to optimize their waste management systems based on their unique needs.

In Manchester, England, the Greater Manchester Waste Disposal Authority, England's most significant waste management institution, has started to use Big Data to better orchestrate the waste management services they provide. To get the most out of their Big Data approach, the Greater Manchester Waste Disposal Authority is

working in collaboration with the research done by the University of Manchester. Together, they are trying to create environmentally sustainable solutions for Manchester, and are attempting to develop optimal solutions to the 1.1 million tons of waste that Manchester produces every year. While their work has not been implemented yet, they have recognized the need for both research and partnership with research institutions to compile and analyze data [19].

Another example of a local government implementing Big Data Solutions is the city of Songdo, South Korea. In said city, every citizen needs to use a chip card while disposing of their garbage. Data collected from these chip cards is being used to analyzing on the quantity of disposed waste and their locations. Each trash bin is incorporated with sensors to provide the height of the garbage accumulated in the reciprocal, temperature, and air pollution levels. These multiple parameters help municipal authorities forecast ideal times to collect the trash and optimize the routes to save time and expenses [19].

Researchers in Ethiopia are combining socioeconomic data alongside geographic data in order to get a clearer understanding of the patterns of how household waste is being collected and distributed. This study helped local authorities to better manage waste practices in urban areas [19].

A group of researchers from the University of Stockholm is using Big Data to identify how to optimize waste management routes in their city. By using a wide variety of data sets collected from various sources, roughly around half a million entries including trash bin locations, weights, and truck routes, researchers have developed waste generation maps of Stockholm. This research has helped reveal various inefficiencies in the current waste management system and will be integral in helping them improve their local waste management [19].

4.1.1 Vehicle Routing Problem. Vehicle route optimization is one of the primary concerns in waste management. It is termed as Vehicle Routing Problem (VRP) [7]. While routes may be designed based on geography, merely looking at a map and designing what seems to be functional movement through the city, this is underestimating the extent of the factors involved. There are multiple factors that either directly or indirectly affect waste collection that can also be analyzed and considered while designing routes. Common known factors that influence vehicle routing are the type of vehicle, vehicle capacity, number of collection stops, volume per capita and the route length.

Two of the most fundamental VRPs are the Travelling Salesman Problem (TSP) and the Chinese Postman Problem (CPP) [2]. However, when too many constraints and attributes are considered, both of the TSP and CPP become more difficult problems to solve. Many researchers have studied and published articles on waste management vehicle routing problem VRPs, and yet it is a persistent problem. Mathematical models need to be developed to provide city administrators with tools to make effective long-and short-term decisions relating to their municipal disposal system [3].

In addition to the typical known factors causing VRP problems, indirect attributes must also be taken into consideration. Time of day, traffic, weather patterns, energy prices, demand fluctuations, vehicle health, dump site inventory also have potential to affect route efficiency.

A research team at OSI came up with a better solution for solving VRP problems using Big Data technologies. Mixed Integer Programming (MIP) formulation was designed to interact with millions of attributes in live environments providing real-time decisions to optimize the VRP. Big Data technologies are being used to enable prediction of travel times and forecast and address demand on a tactical time line. This approach has helped improve VRP forecasting between 5% to 10% [12]. Any improvement, even less than 5% created on VRP is a significant improvement [13].

Scientists and engineers have designed a system to collect data that communicates between waste reciprocals, waste collection vehicles and a central system using sensors that measure how full containers are and can send real-time data [10].

The data collected can be used to optimize routes and space in the waste collection vehicles. Sensors can identify which containers do not need to be collected, allowing them to shorten routes when containers do not have sufficient volume to be collected. In addition, sensors can measure the remaining capacity of collection vehicles, allowing them to extend routes when they still have cargo space, ensuring that they return to the disposal center only when they are carrying a full load. After the implementation of this system, the city was able minimize inefficient travel and fewer collection vehicles were needed. It was estimated that in three years, the expense of purchasing and implementing this system would be recovered [18].

In addition to the real-time data collection that helps optimize collection and routes, remote self-diagnostics are also helping optimize vehicle use. By being able to monitor the vehicles use and health, maintenance and repair are more efficiently managed. Through this system, managers are made aware of parts that need to be ordered in advance, limiting the time that the vehicles may be out of service. Hand held devices have been developed for service verification, further helping minimize external providers or assessments, optimizing asset management and costs [11].

4.1.2 Problem of Landfill Disposal. A municipal solid waste landfill (MSWLF) is an individually isolated area of land or a trench where the household waste is collected and stored. These landfills are designed to store municipal solid waste as well as other wastes like construction and industrial waste. These landfills can be open-pits or below ground refuse chambers. In recent years, it's becoming more expensive to operate and maintain these landfills as well as to protect the environment liquid pollutants that drain from the waste and which can cause water and surface contamination. These liquid pollutants are commonly called as leachate. Leachate forms when water originating from rainfall or groundwater dissolves gradually through a porous surface and dissolves the chemicals from the refuse especially if the protective layer of the landfill allows liquid or gases to pass through it.

Leachate contamination problems have become more problematic in older landfill sites where they lacked appropriate barriers above or below the landfills. This contamination has been found to cause pollution which may be a cause of diseases affecting the citizens residing in close proximity to these landfills. In modern days, Governments across the world are motivated to prevent environmental contamination due to waste disposal and to reduce the size and expansion of landfill deposits. Steps are taken to reduce the

number of landfill areas and to extend landfill capacity at current sites.

Landfill disposal, in itself, has its own set of complex factors including measurement and control of gases, management of waste water, and precipitation patterns affecting water volume and run off in the geological basin.

Identifying these factors creates an opportunity to develop sensors that could collect meaningful data that would be an asset to optimizing landfill disposal. Sensors could measure gas emission, water volume in disposed waste, and precipitation. Having this information would allow data scientists to develop systems that would identify when disposal would need to be varied, how water treatment plants could efficiently manage runoff and basin water and how to communicate with water treatment plants to use more or less of their mechanical resources based on real-time needs.

Manual calculation would be time-consuming subject to human error and physical observation opens space for inefficiency and delay. In order to compute these calculations promptly so that the information could contribute to optimization, it would be ideal to develop a computer program to compute the mathematical operations. These calculations could translated into real time information that would help manage the systems in place. This would contribute significantly to optimization of water treatment, for example, in landfill disposal [1]

Waste management is already in the process of transforming from using older methodologies to modern Big Data technologies. Big Data has a lot more potential to eradicate many of the problems faced by government in waste disposal. Most of the government organizations are taking a leap to minimize disposal of waste and to achieve "zero waste" goals [17].

A program in government of District of Columbia called "Zero Waste DC" has initiated developments in order to provide resources that will help its residents move towards zero waste. Its primary motive is to divert 80% of MSW to recycling or source of energy where the remaining 20% non recoverable waste will be sent to landfills. By collecting vast amounts of data from all sources, they will analyze the data collected to implement cost-effective strategies for converting waste to re-usable resources, improving environmental conditions, taking measures on human health, reducing greenhouse gas emissions and conserving of natural resources. By obtaining data from the sensors measuring contamination of air with harmful gases, water purification levels from the nearby lakes, physical properties as well as chemical properties of the soil, Big Data will be able to help analyze this data and provide solutions to prevent further pollution and help design better systems in the future [17].

One of the reasons for Big Data to have such high volumes of information in these examples, is that the sensors that collect the data gather it on a daily basis and from different sources like digital meters, sensors and social media.

In some cases it is possible to read information from the sensors placed in recycle bins, the data they provide allows the company in charge to identify the type of waste is being disposed in the trash bins and which resident does it. If somebody places hazardous materials in the bins, authorities can detect them before being collected by the waste management vehicles. Besides preventing unwanted incidents, this kind of information can be used to develop tailored

waste management training for the public. Big Data Analysis can also be employed in other areas of education, to encourage citizens to use food scraps as fertilizer and even develop strategies like "farm to farm" practices where the public institution can promote urban organic farming [16].

On the same line, the use of Big Data technologies employed in the waste disposal process, provides visibility to the amounts and characteristics of the organic waste, particularly when it comes to food waste. Identifying inefficiencies in food management, can be used to plan and improve the food production and transportation chains. These and other applications of Big Data, enable food related companies to make effective decisions in their purchase and procurement departments, as well as their management of their own organic waste [5].

United States' Environmental Protection Agency (EPA) is using Big Data in their waste management and recycle research. Big Data Analysis is used to identify quantities of solid waste being recycled and informing the public about the benefits they may receive when they recycle. A Municipal Solid Waste Characterization Report issued by the EPA in 2007, contains the data collected on generation and disposal of waste in the United States within the 30 years prior to the release of said report. This data is being analyzed to measure waste reduction and plan recycling programs across the country [8].

4.2 Statistics and Waste Management

The data collected from the different activities related to Municipal Waste Management, can be run through mathematical models developed to predict behavior and understand causes to complex issues. In other words, these data and applying statistics can be used by Big Data professionals to draw conclusions and foresee possible outcomes as related to effective policy making and actual waste production, collection, and disposal patterns.

In the field of statistics and probability, there are many data analysis methods that are used to study waste management and production, but the two most popular are: PCA (Principal Component Analysis) and PLS1 (or Partial Least Squares Regression). The Principal Component Analysis reveals relevant parameters within a large parameter set. This allows the researcher to find the most significant and essential properties of a sample when studying a particular question. On the other hand, scientists use the Partial Least Squares Regression to identify in two matrices, the major internal and external correlations [4].

There are many applications tools and packages that are well known in statistics that can be applied to waste management, some of them are SPSS, Canoco, The Unscrambler and R [4]. There is also a mathematical modeling language developed particularly for formulating and solving optimization problems using linear programming, called Lingo. Lingo optimization software uses the branch and bound methods to solve problems similar to the ones found when studying waste management [1].

These resources are valuable to the continual analysis of data being collected, continually helping in waste management optimization.

4.3 GIS Analytics

When it comes to Geographical Information Systems (From now on GIS), there are multiple software and hardware options in the market. From paid software like ArcGIS to Open and free software like QGIS, some solutions can help interpret large data sets, apply statistics and algorithms of different kinds and display them in a way that refers to a geographical space.

Two optimal routing algorithms have been used to calculate routes for waste collection; Solomon's insertion algorithm and a clustering algorithm. Data was used to help create more efficient routes by minimizing driving distance while taking into consideration factors like lunch breaks (time windows) that affected the distribution of human resources and time/vehicle management. By adding vehicle depots, by rerouting, and by sharing various routes, they predicted that they would be able to shorten vehicle expenditure by as much as 10,000 km.

An additional routing algorithm was used to present information on the environmental significance for optimizing waste collection routes. The study assessed the use of roll-on/roll-off containers while also considering schedules and lunch breaks. Through the utilization of an adaptive large neighborhood search algorithm, alongside a clustering method, a residential waste collection was analyzed, taking into consideration actual pick up points and employee lunch breaks. Using this information, time windows and starting conditions were adapted alongside the routes to reduce the total distance by as much as forty-five percent [18].

5 CONCLUSION

Waste management has been a growing concern and will continue to be an important area for optimization as both consumer waste and population increase. As institutions, governments, and individuals look to assess and optimize resources, minimize cost and create less of an environmental impact of waste management, Big Data has the potential to continue to help provide the information needed for future advancement.

There are various tools being used to optimize the different waste management practices, and there is space to develop additional tools to continue to the information available to decision makers.

One of the main reasons to use Big data in Municipal Waste Management is to provide local governments with tools that would facilitate the implementation of systems to more efficiently manage how much, where and the growth rate of the material that the community disposes. Optimizing the waste management systems can help minimize the environmental impact of the community's actions, reduce pollution, increase rates of recycled materials, optimize routes to reduce time and expenses, among others. By using Big data, local governments can also track the number and quantity or weight of disposals at different locations; this information can be mapped to reveal the locations of the most significant waste generators. This information will help entities develop specific strategies to reduce waste and help implement permanent solutions for better environments.

Waste Management is not only government issue. Citizens should take the initiative and educate others on how to recycle and reduce waste. With the help of the collected data, governments

can notify the different entities (individuals, communities, companies and other organizations) a equip them through education and awareness, and also share valuable information about the importance of waste management through different media, such as mobile phones, email, among others.

Local governments have just started to adopt Big Data technologies for solving problems involved in MSW, but there is plenty of room for growth and further use and development. By using Big Data Analytic, large amounts of data sets pertaining to specific communities waste, routes, and disposal can be used to identify trends and patterns that could highlight opportunities for improvement. Big Data can play a significant role in managing cities more efficiently, benefiting not only those managing the systems, but the communities in which they are implemented as well.

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6 BIBTEX ISSUES

7 ISSUES

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