DEVELOPING A DECISION-MAKING FRAMEWORK FOR OPTIMAL MARINE OIL SPILL WASTE MANAGEMENT AND TREATMENT

by

Seyedeh Mahboobeh Jafari

B.Sc., Shiraz University, 2014 M.Sc., Shiraz University, 2017

THESIS SUBMITTED IN PARTIAL FULFILLMENT OF THE REQUIREMENTS FOR THE DEGREE OF MASTER OF SCIENCE IN NATURAL RESOURCES AND ENVIRONMENTAL STUDIES

UNIVERSITY OF NORTHERN BRITISH COLUMBIA

November 2024

© Seyedeh Mahboobeh Jafari, 2024

Abstract

This research explores estimating off-shore oily waste, considering waste-waste compatibility due to the heterogeneous nature of oily waste. Firstly, hyperparameters for Artificial Neural Networks (ANN), Support Vector Regression (SVR), and Improved Random Forest (IRF) models are optimized to develop a comprehensive oily waste estimation model incorporating liquid, solid, and total waste types. The results show that IRF is the most accurate model, with the lowest error indices and a higher correlation coefficient compared to ANN and SVR. This study then takes a step further to propose a waste allocation framework, which is tested using information on the Bella Bella oil spill incident in British Columbia. Incorporating treatment and receiving facilities' details, such as their location and capacity, the framework distinguishes all possible waste pathways for handling the waste from source to landfill. Genetic Algorithm (GA) is introduced to optimize waste transfer processes and successfully minimize transportation costs. The results show that the model can find the most optimized path to reduce transportation costs. The model's high customization, adaptability, and capacity to consider multiple nodes make it suitable for complex waste transfer networks, demonstrating its practicality in emergency situations. Efficiently allocating resources and ensuring cost-effective waste transportation while considering facility capacities and waste compatibility, the study holds practical implications for waste management practitioners, environmental authorities, and response teams.

TABLE OF CONTENTS

	F CONTENTS	
	TABLES	
	FIGURES	
	RY/LEDGEMENT	
CHAPTER		
_	t Is a Marine Oil Spill?	
	or Oil Spill Incidents	
1.3 Oil S	Spill Effects on Different Sectors	5
1.3.1	Aquatic Habitat and Ecology	5
1.3.2	Wildlife	5
1.3.3	Economy	6
1.3.4	First Nations and Local Communities	6
1.3.5	Tourism	7
1.3.6	Human Health	7
1.4 Wha	t Affects Oil Spill?	8
1.5 Oil S	Spill Clean-up Strategy	9
1.5.1	Off-shore Clean-up Techniques	9
1.5.2	On-shore Clean-up Techniques	11
1.6 Revi	ew of Canadian Petroleum Industry and Oil Shipping Activities	12
1.6.1	Canadian Marine Oil Response System and Practices	12
1.6.2	Legislative and Regulatory Structure	13
1.6.3	National Oil Spill Preparedness and Response Regime	14
1.7 Obje	ectives and Significance of this Study	14
1.8 Orga	nnization of The Thesis	15
CHAPTER		
	pill Waste Management and Modeling	
2.2 Rese	arch Gaps in Oily Waste Estimation and Waste Management	
CHAPTER		
	Collection and Assumption	
	icial Intelligence (AI)-based Model Development	
3.2.1	Artificial Neural Network (ANN)	
3.2.2	Support Vector Regression (SVR)	
3.2.3	Improved Random Forest (IRF)	25

3.3 Waste Estimation Model	26
3.3.1 Model Evaluation	29
3.4 Waste Management/Transfer Framework	30
3.4.1Problem Description	30
3.4.2 Mathematical Modeling	32
3.4.3 Objective Function and Constraints	35
Appendix A	38
CHAPTER 4 RESULTS and DISCUSSION	
4.1 Hyperparameter Optimization	43
4.2 Waste Estimation Model	45
4.3 Waste Allocation Framework	46
CHAPTER 5 CONCLUSION AND RECOMMENDATIONS	54
5.1 Conclusion	54
5.2 Recommendations	57
REFERENCES	59

LIST OF TABLES

Table 1.1	Twenty-two major oil spill incidents the world has seen
Table 3.1	AI-based model's hyperparameters, their description and search space used in the Bayesian optimization
Table 3.2	Notations used in the designed waste management framework41
Table 3.3	Description of the decision variable used in the designed waste management framework
Table 4.1	Optimized hyperparameters of ANN waste estimation model
Table 4.2	Optimized hyperparameters of SVR waste estimation model53
Table 4.3	Optimized hyperparameters of RF waste estimation model54
Table 4.4	Evaluation of the AI-based waste estimation models55
Table 4.5	Name and location of currently operating facilities in British Columbia56
Table 4.6	The Detailed information on oily waste handling facilities in British Columbia
Table 4.7	The identified path from source to the landfill with volume based on the Bella Bella oil spill incident in British Columbia

LIST OF FIGURES

Figure 1.1	Comparison between the number of tanker spills and growth in crude oil and other tanker trade between 1970 and 2020 (Adapted from ITOPF (2021)) 11
Figure 1.2	Number of medium and large tanker spills from 1970 to 2021 (Adapted from ITOPF (2021))
Figure 1.3	Canadian Coast Guard Regional boundaries (Adapted from Government of Canada (2022)
Figure 3.1	Parameters affecting the volume of generated off-shore oily waste31
Figure 3.2	Schematic structure of the ANN model
Figure 3.3	A Random Forest schematic view34
Figure 3.4	Schematic view of possible transportation paths between each two nodes in the framework
Figure 4.1	Results of Bayesian optimization for AI-based optimization models: a) ANN, b) SVR, and c) RF
Figure 4.2	The location of waste handling facilities in British Columbia56
Figure 4.3	The common practice of oily waste management in British Columbia57
Figure 4.4	Genetic Algorithm minimization cost graph

GLOSSARY

Abbreviations		
AI	Artificial Intelligence	
ANN	Artificial Nueral Network	
CC	Corrolation Coefficient	
CCG	Canadian Coast Guard	
DFO	Fisheries and Oceans Canada	
EPA	Environmental Protection Agency	
IMO	International Maritime Organization	
IRF	Improved Random Forest	
ITOPF	International Tanker Owners Pollution Federation Limited	
MPRI	Multi-Partner Oil Spill Research Initiative	
NOAA	National Oceanic and Atmospheric Administration	
NSERC	Natural Sciences and Engineering Research Council of Canada	
NP-hard	Non-deterministic Polynomial-time hard	
PPE	Personal Protective Equipment	
RMSE	Root Mean Square Error	
RMAE	Relative Mean Absolute Error	
SRM	Structural Risk Minimization	
SVR	Support Vector Regression	
WCMRC	Western Canada Marine Response Corporation	

ACKNOWLEDGEMENT

I stand at the end of a very long and challenging journey, one that truly tested my determination and resilience. This thesis is a culmination of academic effort and a testament to the power of hope, resilience, and unwavering support.

To my beloved family and friends, though miles apart, your love and encouragement were always close at heart. Your belief in me, even during the darkest times, fueled my determination to overcome challenges and power through.

I extend my deepest gratitude to my supervisors, Dr. Jianbing Li and Dr. Youmin Tang and my esteemed committee member, Dr. Chris Opio. Your guidance, patience and support steered me towards overcoming challenges. I am profoundly grateful for your mentorship and support.

The research was financially supported by the Multi-Partner Oil Spill Research Initiative (MPRI) of Fisheries and Oceans Canada, as well as the Natural Sciences and Engineering Research Council of Canada (NSERC). I would like to express my sincere gratitude to the following individuals whose invaluable assistance greatly contributed to the success of this research: Dale Bull, Senior Environmental Emergency Response Officer at the BC Ministry of Environment and Climate Change Strategy; Jeff Brady, Deputy Superintendent of Environmental Response at the Canadian Coast Guard; Scott Wright, Director of Response Readiness at Western Canada Marine Response Corporation (WCMRC); and David Ellwood, Regional Commercial Manager at Terrapure Environmental. Their provision of information and data played a crucial role in facilitating the progress and outcomes of my research.

To all who find themselves navigating their storms, may my thesis stand as a humble symbol of hope and an invitation to persist and dare to dream, even when the odds are saying otherwise.

Thank you to each and everyone of you for being a part of this journey.

Mahboobeh

CHAPTER 1 INTRODUCTION

1.1 What Is a Marine Oil Spill?

Over the past two decades, there have been ongoing debates surrounding the transition to more sustainable energy sources. Despite this, many industries continue to rely on fossil fuels. The rapid growth of industrialization and economic expansion has led to a rise in the transportation of fossil fuels, consequently increasing the risk of oil spill incidents. A marine oil spill incident occurs when petroleum hydrocarbons are accidentally released at sea due to human error or equipment failures (Li et al. (2014); Beyer et al. (2016)). Tankers, offshore platforms, drilling rigs, and subsea piping lines are identified as the primary sources of such spill incidents (Li et al. (2014)).

The International Tanker Owners Pollution Federation Limited (ITOPF) is a London-based organization that compiles statistics on global oil demand and incidents of oil and chemical spills. They also provide technical support services for oil spill response plans. This information is made available to tanker owners, their insurers responsible for covering oil pollution incidents, and international organizations like the International Maritime Organization (IMO), as adapted from ITOPF (2021). In essence, ITOPF's core services encompass spill response, analysis of claims and damages, training, contingency planning, and advisory and informational support.

Figure 1.1 depicts a long-term analysis of global trends in oil demand versus tanker spill incidents. As shown, despite an increase in tanker movements from 1970 to 2020, the number of oil spill incidents has consistently decreased over these years. This positive trend is attributed to advancements in the shipping industry, coupled with stricter regulations and a sustained commitment to enhancing maritime safety and environmental protection through investment and exploration of innovative solutions.

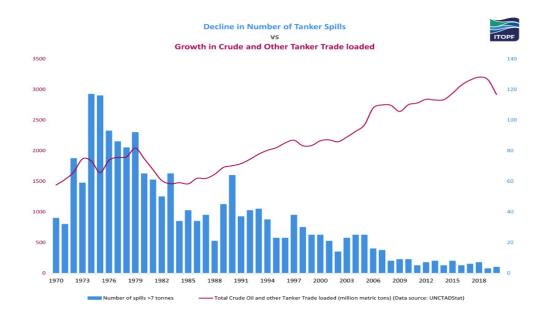


Figure 1.1 Comparison between the number of tanker spills and growth in crude oil and other tanker trade between 1970 and 2020 (Adapted from ITOPF (2021)).

Investigating the causes of oil spill incidents offers valuable insights for managerial tasks and risk analysis. An analysis conducted by ITOPF spans tanker spill incidents from around 1970 to 2021. Spills exceeding 700 tonnes, those ranging from 7 to 700 tonnes, and those below 7 tonnes are categorized as large, medium, and minor, respectively. Figure 1.2 illustrates the trend and volume of spill incidents in the medium and large categories during this period.

As shown, the total volume of oil released due to tanker spill incidents in 2021 is approximately 10,000 tonnes, with the majority classified as medium spill incidents. While annual fluctuations are present, the overall trend shows a decline in the number of oil spill incidents. This reduction is attributed to positive shifts within the shipping industry, reinforced regulations, and governmental support for research initiatives.

The following sections will provide more details regarding marine oil spill incidents, causes, environmental impacts and related regulations.

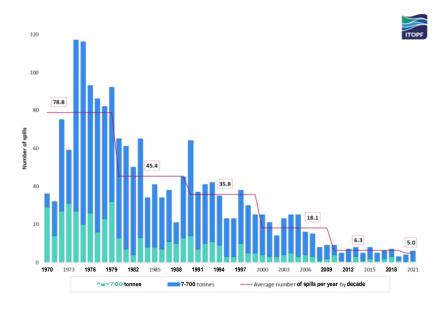


Figure 1.2 Number of medium and large tanker spills from 1970 to 2021 (Adapted from ITOPF (2021).

1.2 Major Oil Spill Incidents

Even though the number of oil spill incidents has decreased significantly, primarily due to stringent regulations, it's essential to recognize that a single oil spill incident can still lead to catastrophic events. A summary of the twenty most significant oil spill incidents worldwide is presented in the following table. This table includes renowned incidents such as Exxon Valdez and Prestige, although they are listed further down the rank column in Table 1.1. The subsequent section will provide detailed information about the consequences of oil spill incidents.

Table 1.1 Twenty-two major oil spill incidents the world has seen.

Rank	Ship name	Year	Location	Spill size (tonne)
1	Atlantic Empress	1979	Off Tobago, west India	287,000
2	ABT Summer	1991	700 nautical miles off Angola	260,000
3	Castillo de Bellver	1983	Off Saldanha Bay, South Africa	252,000
4	Amoco Cadiz	1978	Off Brittany, France	223,000
5	Haven	1991	Genoa, Italy	144,000
6	Odyssey	1988	700 nautical miles off Nova Scotia, Canada	132,000
7	Torrey Canyon	1967	Scilly Isles, UK	119,000
8	Sea Star	1972	Gulf of Oman	115,000
9	Sanchi	2018	Off Shanghai, China	113,000
10	Irenes Serenade	1980	Navarino Bay, Greece	100,000
11	Urquiola	1976	La Coruna, Spain	100,000
12	Hawaiian Patriot	1979	300 nautical miles off Honolulu	95,000
13	Independenta	1979	Bosphorus, Turkey	94,000
14	Jakob Maersk	1975	Oporto, Portugal	88,000
15	Brear	1993	Shetland Islands, UK	85,000
16	Aegean Sea	1992	La Coruna, Spain	74,000
17	Sea Empress	1996	Milfor Haven, UK	72,000
18	Khark 5	1989	120 nautical miles off the Atlantic coast of Morocco	70,000
19	Nova	1985	Off Khark Islan, Gulf of Iran	70,000
20	Katina P	1992	Off Maputo, Mozambique	67,000
21	Prestige	2002	Off Galicia, Spain	63,000
22	Exxon Valdez	1989	Prince William Sound, Alaska, USA	37,000

1.3 Oil Spill Effects on Different Sectors

1.3.1 Aquatic Habitat and Ecology

Aquatic habitats are the first areas heavily impacted by oil spill incidents. The marine environment is a complex system of plants, animals, and their surroundings. When the environment is harmed, it often affects species in the food chain, leading to further consequences for other species (IMO (2001)).

In open water, fish and whales can swim away from an oil spill by going deeper or in different directions. However, animals like turtles, seals, and dolphins, which usually live near the shore, are at a higher risk. They are affected by oil on beaches or by eating prey in oil. In shallow water, oil can also harm plants like seagrasses, kelp, and coral reefs, which many species rely on for food, homes, and breeding spots (IMO (2001)).

Oil spill incidents can also happen in swamps and marshes, where there's less water movement, making the situation worse than in flowing water. Lakes and ponds are also at risk because more types of animals can be exposed to oil. This is especially true for migrating birds, as they can spread the contamination over a larger area. While rivers are usually less affected than lakes, if a river is a drinking water source, it directly threatens human health. This concerns many communities, including indigenous communities, that rely on water bodies for their needs.

1.3.2 Wildlife

The species most impacted by oil spill incidents are birds, mammals, and plants inhabiting marine environments and adjacent shorelines. These organisms face various

threats, including direct physical contact with oil, toxic contamination, depletion of food sources or habitats, and reproductive challenges (IMO (2001)).

Different species exhibit varying degrees of tolerance to oil ingestion. For instance, oil spill incidents and their vapours can be fatal to seabirds. Despite their potential resilience, once exposed to oil, these birds can accumulate it within their bodies, leading to infiltration of their nervous system, liver, and lungs. Consequently, oil enters the food chain, posing a risk of contamination to predators.

1.3.3 Economy

Marine-based industries, including port operations, fisheries, tourism, and aquaculture enterprises, often experience significant adverse effects from oil spill incidents. These impacts manifest through direct losses of products due to mortality or habitat destruction, as well as restricted access resulting from harvesting bans and area closures. Furthermore, economic losses stem from decreased market demand, driven by concerns over the safety of products tainted by oil contamination (Li et al. (2014); Beyer et al. (2016)). These losses reverberate throughout the fisheries supply chain, affecting docks, processors, and supply businesses (Beyer et al. (2016)).

1.3.4 First Nations and Local Communities

First Nations represent a particularly vulnerable segment of society, highly susceptible to oil spill incidents' consequences. These communities often rely on fisheries and land resources from water bodies, making them particularly vulnerable to the impacts of oil contamination in terms of polluted soils from oil pipelines and tainted seafood (Li et al. (2014)).

In addition to First Nations, other indigenous communities and rural locals with strong ties to the natural environment for subsistence and cultural practices are also affected by oil spill incidents. While compensatory frameworks exist for these communities, delays in compensation disbursement may adversely affect trust in governmental authorities, potentially leading to significant social unrest and societal disruption.

1.3.5 Tourism

Tourism stands out as a particularly vulnerable sector in crises and disasters. This vulnerability arises from its interrelations with other industries, such as transportation and accommodation, hotels, airlines, and car rentals. Additionally, tourism is heavily influenced by external factors such as political stability, currency exchange rates, and weather conditions (Beyer et al. (2016)).

Oil spill incidents near shorelines and areas populated by humans pose aesthetic concerns, necessitating safety measures due to toxic volatile vapours. Cleaning such spills is more costly and may extend over longer periods, resulting in significant losses and market decline for businesses and properties situated near beaches and waterfronts, such as restaurants, hotels, and recreational facilities. Over time, tourism in the affected area may experience crises or collapse, as these events divert tourist traffic to alternative destinations.

1.3.6 Human Health

Human beings can be impacted by an oil spill incident in three significant ways, including disruption of ecological processes, resulting in direct harm. This includes ingesting seafood contaminated with oil toxins, which can lead to various health issues. For instance, consuming seafood bio-accumulated with oil toxins or breathing in oil vapours can cause

direct harm. Economic stressors affect individuals working in fields such as fishers and the tourism industry. According to Aguilera et al. (2010) and Major and Wang (2012), inhalation of vapours or consuming contaminated seafood can lead to harmful health effects ranging from dizziness and nausea to certain types of cancers and issues with the central nervous system. Although the long-term effects of hydrocarbon toxicity on humans are less studied, they have been associated with severe DNA degradation, cancers, congenital disabilities, reproductive defects, irreversible neurological and endocrine damage, and impaired cellular immunity (Binet et al. (2003); Aguilera et al. (2010)).

1.4 What Affects Oil Spill?

Once an oil slick forms on the water's surface following an oil spill incident, it undergoes various weathering processes, including photolysis, evaporation, dilution, formation of oil-water emulsions, and biodegradation (Dave and Ghaly (2011)). Specifically, the formation of oil-water emulsion leads to significant changes in water interfacial tension, density, and viscosity. Selecting the most effective method for spill clean-up largely depends on the characteristics of the oil and environmental factors. Therefore, understanding factors such as the quantity and type of spill, weather and ocean conditions, age of the spilled oil, and ocean behaviour is crucial.

The most common marine oil spills include bunker crude oil, refined petroleum products and by-products, and waste oils (Dave and Ghaly (2011)).

1.5 Oil Spill Clean-up Strategy

When the oil spills on top of the water's surface, it will slowly drift toward the shore due to tidal activities. Spill clean-up operations must be employed to prevent a catastrophic situation after an incident. Over the last 50 years, oil spill clean-up technologies have developed extensively. The clean-up processes at the spill location (at sea) and the shore are called off-shore and shoreline clean-ups, respectively.

1.5.1 Off-shore Clean-up Techniques

Off-shore response techniques are divided into mechanical/physical, chemical, biological and in-situ burning. (Dave and Ghaly (2011)).

a) Physical techniques primarily involve spatially controlling the oil slick using physical barriers, thereby keeping the oil's physical and chemical characteristics unchanged. Commonly used physical barriers globally include booms, skimmers, and adsorbent materials (Fingas (2016)).

Booms are floating barriers designed to restrict the movement of oil, ultimately facilitating higher oil recovery through skimmers or other response methods. They come in three main types: fence booms, curtain booms, and fire-resistant booms. Fire-resistant booms are typically employed in conjunction with insitu burning. Fence and curtain booms serve a similar function, with approximately 60% submerged underwater and only 40% floating on the surface. Booms are typically around 15 meters long and can be interconnected as needed. While fence booms are lightweight, easy to handle, and reliable in calm waters, their stability in rough conditions with high waves and strong winds is limited.

Skimmers represent the second physical response technique commonly used alongside booms to enhance oil recovery efficiency. Depending on the skimmer type, they exhibit high stability in rough ocean conditions and can recover up to 90% of the oil. However, they are less effective when dealing with oil mixed with dispersants. Regardless of their type, all skimmers are made of oleophilic materials, which attract oil slicks that adhere to the surface. This oil can then be scraped or squeezed from the surface and collected in a small storage tank.

Utilizing adsorbent materials offers another physical approach to handling oil spill incidents without altering the oil's characteristics. Hydrophobic sorbents are employed for clean-up after skimming as a final step in response operations to capture any remaining oil on the water.

- b) Chemical methods complement physical techniques to expedite clean-up, particularly in spill locations near shorelines or sensitive marine environments. Dispersants and solidifiers are commonly employed chemicals in oil spill clean-up, altering the characteristics of the oil. Dispersants typically consist of surfactants, which break down the oil slick into smaller droplets, facilitating faster biodegradation as they are dispersed into the deep-water column. On the other hand, solidifiers transform the oil phase from liquid to a rubber-like substance, enabling more accessible collection on the water's surface. In rough seas, solidifiers can effectively utilize wave energy to enhance dissolution in the water, resulting in a higher rate of solidification (Dave and Ghaly (2011)).
- c) In-situ burning represents a straightforward and rapid clean-up method. However, its usage is limited due to concerns over human health and environmental impacts associated with burning by-products, including residues and the emission of thick plumes of black smoke. This method is particularly effective in snowy

conditions, such as clean-ups after pipeline leaks or when oil spills occur on top of ice. One crucial requirement is that the oil slick be sufficiently thick to sustain burning and prevent it from cooling.

d) Biodegradation serves as the final option for marine oil spill clean-up. While environmentally friendly and safe, its application is restricted by the slow degradation process, leading to prolonged exposure and less suitable for environments with low microbial activity.

1.5.2 On-shore Clean-up Techniques

On-shore oil spill clean-ups are more straightforward than off-shore and do not require special equipment. The clean-up process can be summarized into three stages based on (ITOPF (2021)) as follows:

- A) Stage 1- Emergency phase: This phase revolves around a collection of floating oil near the shoreline and transfer to temporary storage, usually on the shore.
- B) Stage 2 project phase: Stage 2 is a complementary phase to Stage 1, often combined. This phase involves collecting oily contaminated materials left on the shoreline. In smaller projects, any remaining oil might be left to degrade naturally.
- C) Stage 3- Polishing phase: this stage includes the final clean-ups and removal of oil stains if required.

Removing bulked oil and treating oil-contaminated beach materials (Stages 1 and 2) typically involves using skimmers to collect the oil, which is then transferred using vacuum trucks to subsequent handling facilities. The oil removal process can be done using

mechanical equipment or manual methods. Pressure washing is commonly employed to ensure thorough cleaning for hard-to-access areas along the shore.

1.6 Review of Canadian Petroleum Industry and Oil Shipping Activities

Canada is among the world's top oil producers, with approximately 98% of its oil exported to the United States (Mohammadiun et al. (2021)). Alberta has the largest oil sand reserves and crude oil production, followed by Saskatchewan, which produces approximately 487,000 daily barrels. Canada relies on reliable transportation methods such as railways, pipelines, trucks, and oil tankers to transport the produced oil to its destinations, including the US and other parts of the world. The selection of transportation methods depends mainly on factors such as volume and destination.

Regarding marine oil shipping, around 87% of Canadian oil is transported through the Atlantic coast, the Great Lakes, the Gulf of St. Lawrence and St. Lawrence Seaway, and associated ports. In comparison, the remaining 13% is shipped through Pacific coastal ports. Seven major ports accommodate significant oil tanker traffic, including the Port of Vancouver, Port of Montreal, Port de Quebec, Newfoundland Off-shore, Port of Saint John, Port of Hawkesbury, and Nova Scotia.

Given Canada's strategic geographical location, multiple international transit routes, and the continuous growth of industrial activities, there has been an increase in oil tanker transit along the Canadian coastline, consequently heightening the risk of oil spill incidents.

1.6.1 Canadian Marine Oil Response System and Practices

The Canadian Coast Guard (CCG) operates as a strategic agency under Fisheries and Oceans Canada (DFO) and is entrusted with the task of providing timely and effective

responses to incidents involving ship-source or unknown-source pollutants in Canadian waters (Government of Canada (2022)). To facilitate efficient program delivery, Canada has been divided into three regions known as the Western, Atlantic, Central, and Arctic Coast Guards. These regional divisions were established in October 2012 to ensure swift administration of response efforts. Figure 1.3 illustrates the boundaries of the Canadian Coast Guard regions as determined in October 2012.

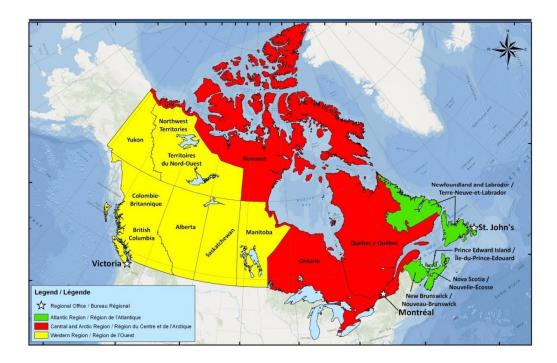


Figure 1.3 Canadian Coast Guard Regional boundaries (Adapted from Government of Canada (2022))

1.6.2 Legislative and Regulatory Structure

Under federal legislation and various international agreements, the federal government is responsible for cleaning up any pollutants spilled in Canadian waters.

1.6.3 National Oil Spill Preparedness and Response Regime

Since 1995, Transport Canada has taken the lead of federal regulatory agencies responsible for the regime based on the partnership between industry and government. This regime provides guidelines and regulatory structures for preparedness and response to marine oil spill incidents. In this regime, potential polluters pay for readiness. The three pillars of oil spill clean-up are Prevention, Preparedness and Response.

1.7 Objectives and Significance of this Study

Understanding the quantity of generated oily waste is essential for proactive preparedness for oil spill response, enabling the determination of necessary resources such as materials and labour required for urgent response. This helps decision-makers allocate resources effectively to address oil spills promptly, as the longer the response is delayed, the further the oil spreads across the water body. This spread necessitates increased use of resources such as products and labour, significantly increasing the volume of oily waste generated. Therefore, prompt action is vital to minimize waste accumulation and effectively address the environmental impact of oil spill incidents.

Therefore, the objectives of this study are twofold: a) to develop an AI-based model to effectively and accurately estimate the volume of generated oily waste, and b) to develop a waste management framework for handling the generated waste based on factors such as waste type and the availability of treatment and receiving facilities, as well as landfills, to minimize the cost of waste handling.

1.8 Organization of The Thesis

The thesis structure is as follows: Chapter 2 consists of a comprehensive literature review. The methods, materials, results, and discussion were described in Chapters 3 and 4. Chapter 5 explains the conclusion of the entire study and the recommendations for future studies.

CHAPTER 2 LITERATURE REVIEW

2.1 Oil Spill Waste Management and Modeling

In recent years, the focus on adequate pre-planning, risk assessment, and advancements in response techniques within the realm of oil spill management has intensified. This heightened attention stems from recognizing oil spills' profound environmental and economic consequences.

Researchers such as Marta-Almeida et al. (2013) and Azevedo et al. (2014) have made vital contributions to this field, whose models have emerged as valuable tools for addressing these challenges. The models proposed by Marta-Almeida et al. (2013) and Azevedo et al. (2014) have benefitted various stages of oil spill management. Specifically, these models have played instrumental roles in pre-planning activities, facilitating more informed decision-making processes. Additionally, they have proven invaluable in conducting comprehensive risk assessments, enabling stakeholders to anticipate and mitigate potential environmental and socioeconomic impacts (Bejarano and Mearns (2015)). Moreover, these models have contributed significantly to advancing response techniques during oil spill incidents. By providing insights into the behaviour and trajectory of spilled oil, as well as the effectiveness of different response strategies, they have empowered responders to devise more effective and efficient cleanup operations (Lehr et al. (2000); Ghanbari et al. (2021)).

Previous studies have underscored these contributions (Peterson et al. (2003); Dave and Ghaly (2011); Barron (2012)), which have highlighted the critical role of modelling approaches in enhancing preparedness and response capabilities. Through their empirical analyses and case studies, these researchers have demonstrated the tangible benefits of integrating modelling frameworks into existing oil spill management protocols (IMO (2001)).

Managing oily waste represents a crucial aspect of pre-planning and response strategies in oil spill waste management. Oily waste necessitates specialized handling and disposal procedures due to its hazardous nature and potential environmental impact. Despite the considerable attention given to oily waste management, a significant gap exists concerning estimating oily waste volumes generated during spill incidents.

While preliminary guidelines for oily waste management and factors influencing waste generation rates have been documented since the early 2000s (IMO (2001); IPIECA and IOGP (2014)), more emphasis should be placed on the accurate estimation of oily waste volumes. Metcalf (2014) attempted to address this deficiency by developing a general waste management plan to investigate the impact of cleanup strategies on waste generation. However, further advancements were reported as necessary to improve the accuracy and reliability of oily waste estimation methodologies (Beegle-Krause (2005); Bergstra (2012)).

Recent studies have turned to innovative approaches, such as self-learning or semisupervised learning techniques, to bridge this gap and enhance the model's accuracy. These methodologies leverage labelled and unlabeled data to improve model performance (Tkalich (2006)) iteratively. Combining information from unlabeled data with labelled data expands the labelled training dataset over successive iterations until the entire dataset is labelled. Such techniques are the foundation for machine learning algorithms, including Random Forest (RF).

The applicability of self-learning approaches has been demonstrated across various scientific disciplines, including GIS and remote sensing (Fatehi and Asadi (2017); Zhao et al. (2017); Lottes and Stachniss (2017)); medical diagnostics (Kourou et al. (2015)), and groundwater investigation (Sameen et al. (2019)). These studies have highlighted the

potential of self-learning methods, particularly in data-scarce environments, as they require fewer labelled training data than traditional supervised learning models.

Another aspect of marine oil spill response waste management involves efficiently handling the waste. Oily waste is categorized as hazardous waste, characterized by toxicity, corrosiveness, ignitability, and chemical reactivity, primarily stemming from industrial and manufacturing processes. In many countries, the manufacturing industry contributes over 75% of hazardous waste, emphasizing the criticality of its management due to environmental and human health risks. Efficiently managing hazardous waste involves collecting, transporting, treating, recycling, and disposing of it safely and cost-effectively (Xu et al. (2014); Ghanbari et al. (2021)).

The integration of location and vehicle routing decisions in hazardous waste transportation and disposal was explored in the early 1980s. Various models have emerged to optimize objectives like travel time, transportation, and disposal risk.

Alumur and Kara (2007) addressed optimizing hazardous waste management systems with treatment and disposal facilities, introducing realistic constraints on the properties of waste types. Alumur and Kara (2007) incorporated waste-technology compatibility constraints, while Peterson et al. (2003) considered waste-waste compatibility and disposal centers, considering government regulations and air pollution standards as constraints. Few studies have directly addressed the vehicle routing problem, while Peterson et al. (2003) tackled waste management with multiple incompatible waste types. Integrating inventory control into location and routing decisions has yet to be explored.

Addressing uncertainty is crucial, as it affects planning decisions. Zhao et al. (2017) explored waste generation amount uncertainties and transportation costs. Metaheuristic

approaches like Genetic Algorithms (GA) have gained traction due to the Non-deterministic Polynomial-time (NP)-hard nature of location-routing problems.

Despite advancements, several gaps still need to be addressed. Simultaneous optimization of location, routing, and inventory plans under uncertain conditions, multiperiod planning, and application of multi-objective solution approaches for medium to large-scale stochastic problems warrant further exploration. This comprehensive literature review outlines these aspects, providing a foundation for marine oily waste management framework research.

2.2 Research Gaps in Oily Waste Estimation and Waste Management

Understanding the magnitude of oily waste production during oil spill incidents is crucial for proactive response planning, facilitating the allocating necessary resources such as materials and labour required for prompt intervention. Rapid response is imperative to mitigate oil spread across water bodies, minimizing resource consumption and waste generation. Therefore, timely action is crucisl to reduce waste accumulation and effectively mitigate the environmental repercussions of oil spill incidents.

Hence, this study aims to achieve two primary objectives: firstly, to develop an AI-based model for the estimation of oily waste volume, and secondly, to formulate a comprehensive waste management framework considering factors such as waste type and the availability of treatment and disposal facilities with the ultimate goal of optimizing and/or minimizing waste management costs.

Based on the existing literature, this research introduces a novel, Improved Random Forest (IRF) model for marine oily waste estimation. By linking self-learning methodologies with the Random Forest algorithm, IRF aims to enhance predictive accuracy regarding oily

waste generation in marine contexts. To ensure the efficacy of the IRF model, this study employs Bayesian optimization for fine-tuning hyperparameters and conducts thorough performance evaluations through cross-validation. Furthermore, two conventional machine learning algorithms, namely Artificial Neural Network (ANN) and Support Vector Regression (SVR), are examined for comparative analysis to discern the optimal method for marine oily waste estimation.

Moreover, a comprehensive framework for managing oily waste has been developed to streamline the allocation of various waste volumes to suitable facilities, considering their capacities and geographical locations to minimize transportation costs. This model provides responders with a structured pathway for transferring waste volumes from generation points to treatment facilities, receiving stations, and ultimately landfills, considering compatibility between different types of waste and the capacities of the respective facilities. This model provides responders with optimized routes for waste transfer, ensuring efficient management even during facility disruptions or the availability of additional facilities.

CHAPTER 3 METHODOLOGY

This research aims to conduct comparative assessments of various well-known AI-based models, namely Artificial Neural Networks (ANN), Support Vector Regression (SVR), and Improved Random Forest (IRF), for estimating off-shore oily waste following oil spill clean-ups. ANN is often considered a primary choice among researchers due to its versatility and success in addressing many problems. Meanwhile, SVR has gained increased attention in recent years for its effectiveness in handling estimation problems comprehensively. The following paragraphs will provide data collection and background information on the abovementioned models. This section will then continue explaining how to develop and evaluate estimation models.

3.1 Data Collection and Assumption

Understanding the factors influencing the volume and composition of waste generated during oil spill incidents is crucial for estimating waste. However, accessing information about oil spill incidents presents significant challenges due to limited valid sources. This study compiles a comprehensive database detailing the physical and chemical characteristics of previous oil spill incidents globally to address this issue. Extensive efforts were made to gather information through literature searches, surveys, and consultations with relevant agencies such as the Western Canada Marine Response Corporation (WCMRC), DFO, and BC Ministry of Environment and Climate Change Strategy as provided in Appendix A. This database considers various aspects of oil spill incidents, including quantity of spilled oil, location, oil type, viscosity, ocean conditions, wind speed, water temperature, response techniques, duration of off-shore response operations, and volume of recovered oil

and waste (Figure 3.1). These factors significantly influence response strategies, personal protective equipment requirements (PPE), and the volume of generated waste.

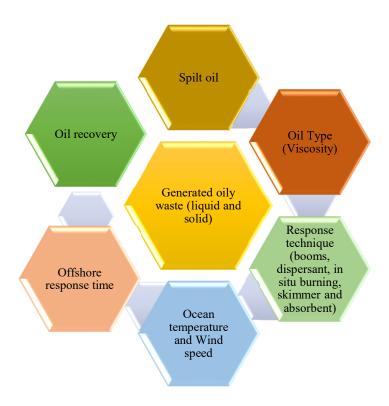


Figure 3.1 Parameters affecting the volume of generated off-shore oily waste

In cases where climatic data, such as ocean temperature, were unavailable in oil spill response reports, assumptions were made based on past 20-year average water temperatures for the incident location reported on different websites such as National Oceanic and Atmospheric Administration (NOAA), US Environmental Protection Agency (EPA), and ITOPF. Sixty oil spill incidents were ultimately selected as input data for the research model. The selection criteria prioritized oil spill incidents with 90% of the required information available.

While AI-based models typically benefit from larger datasets, the decision to use 60 oil spill incidents for testing purposes was deemed sufficient. This approach helps prevent

model overgeneralization and ensures efficient pattern recognition by excluding erroneous or irrelevant data. Notably, the model's adaptability allows for optimization based on new and existing data, promising more accurate waste volume estimations as additional information becomes available.

3.2 Artificial Intelligence (AI)-based Model Development

3.2.1 Artificial Neural Network (ANN)

Artificial neural networks are complex mathematical systems that mimic the human brain's neural system. In brief, ANNs comprise three different layers: input, output, and hidden. Each layer consists of neurons or nodes as processing elements of the network. Neurons on the input layer will distribute the input information to the neurons of the next layer, hidden layer(s), where the information will be processed and then transferred to the output layer, where the results are produced (Zhang and Friedrich (2003)). Figure 3.2 shows a schematic structure of a simple ANN model. In training such a network, weighted connections among layers only occur in a forward direction from the input to the output layers. These interconnections are then adjusted by distributing errors through the layers to produce the most accurate outputs, which is called backpropagation. Hidden layers are advantageous over traditional logistic regression analysis techniques by modelling the interactions and relations among all the input parameters. An essential consideration in designing ANNs is choosing the number of neurons at each layer and the number of hidden layers, commonly known as hyperparameters (Zhang and Friedrich (2003); Wang et al. (2009)). This study has applied a new approach to tuning these parameters, which will be discussed in section 3.2.

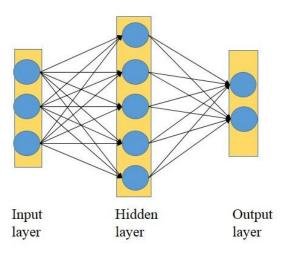


Figure 3.2 Schematic structure of the ANN model

3.2.2 Support Vector Regression (SVR)

Another studied model is SVR, initially designed for classification problems (Sain and Vapnik (1996)), but with the introduction of a new parameter, Vapnik's E intensive loss function, its application has been extended to solve non-linear regression estimation (Smola and Schölkopf (2004)), and time series forecasting as well (Thissen et al. (2003); Lin et al. (2006)). SVR implements Structural Risk Minimization (SRM) rather than empirical risk minimization implemented by most traditional ANNs, which ultimately results in providing an answer that is always unique and globally optimal (Lin et al. (2006)).

Similar to ANN, SVR uses an implicit feature space mapping from the dimension of the data to a possibly infinite feature space, which provides a non-linear representation of the modelled data, performed through a kernel function (Smola and Schölkopf (2004); Sadri et al. (2012); Goyal et al. (2014)). Defining kernel function and its parameters is not straightforward in the SVR network, as some settings might be prone to over-fitting or underfitting. Therefore, different kernel functions should be tested to choose the one associated

with the lowest error in the validation stage. This study uses a Bayesian optimization model to determine the most competent kernel function.

3.2.3 Improved Random Forest (IRF)

RF, the third studied model in this work, is a classification-based regression model introduced by Breiman (2001). It is a supervised learning algorithm that uses an ensemble learning method for regression. RF operates by creating and combining many decision trees (Figure 3.3). Decision trees handle high dimensional data well, overlook irrelevant descriptors, control multiple mechanisms of action, and are amenable to model interpretation (Svetnik et al. (2003)).

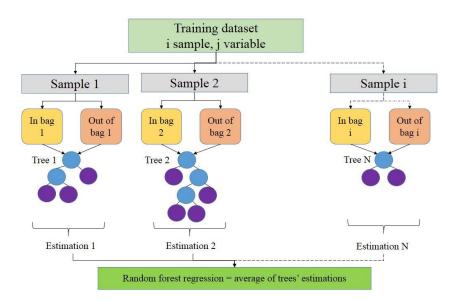


Figure 3.3 A Random Forest schematic view

The steps of training the RF regression model can be summarized in the following steps:

- About two-thirds of the data will be randomly selected as a training dataset called a bootstrapped dataset.
- 2. The remaining one-third will fall into an out-of-bag dataset, which will be used to measure the error.
- 3. The model's outcome is the average result of all the trees (Cutler et al. (2007)).

3.3 Waste Estimation Model

As mentioned, artificial intelligence and machine learning techniques can significantly expedite decision-making processes by analyzing subtle data patterns from previous oil spill incidents. No single machine learning algorithm consistently outperforms others, as their effectiveness depends on the nature of the problem and data formats. Hence, this thesis compares ANN, SVR, and RF for off-shore oily waste estimation. Generally, developing estimation models involves establishing a formula between input and output parameters that accurately represents the problem's nature and can be extended to estimate new variables within the input parameter range the model is trained with.

The effectiveness of AI-based models hinges on the proper combination of hyperparameters, which control the model's overall performance. Typically, hyperparameters are selected through trial and error, a time-consuming process with loose guidelines on their numerical range. Moreover, hyperparameters interact with each other, complicating the optimization process. To address this challenge, robust optimization methods are employed to find the optimal configuration of each AI-based model. Traditionally, grid and random search

methods have been used for this purpose. Grid search optimizes all possible model configurations, while random search involves iterative runs of evaluating models with randomly selected hyperparameters. However, these methods overlook historical information from previous evaluations.

This study uses Bayesian optimization to find the best combination of hyperparameters by minimizing an error index (objective function) within predefined search ranges (constraints). Unlike trial-and-error approaches, Bayesian optimization leverages past evaluations to create a probabilistic model mapping hyperparameters. The Bayesian optimization procedure involves several steps, including model initialization, acquisition function optimization, model updating, and hyperparameter selection. For detailed information on Bayesian optimization, readers are referred to Snoek et al. (2012).

- 1. Build a surrogate probability model of the objective function;
- 2. Find the hyperparameters that perform best on the surrogate;
- 3. Apply these hyperparameters to the actual objective function;
- 4. Update the surrogate model incorporating the new results; and
- 5. Repeat steps 2–4 until max iterations or time is reached.

Table 3.1 presents the name of hyperparameters along with their search range for optimization purposes for ANN, SVR and RF. To accurately compare the models, the objective function is set to minimize the Root Mean Square Error (RMSE) for all models (see Eq. (3.1)). Mean Square Error (MSE) and RMSE are widely used error indices for evaluating model performance. RMSE is chosen over MSE mainly because RMSE gives higher weight and punishes significant errors.

RMSE = =
$$\sqrt{\frac{\sum_{i=1}^{n} (O_i - S_i)^2}{n}}$$
 Eq. (3.1)

Where O and S denote the observed and estimated values, respectively, to achieve a more reliable estimation of oily waste, a 5-fold cross-validation method is employed alongside Bayesian optimization to tune hyperparameters finely. This process involves randomly dividing the data in the training subset into five equal-sized parts. One part is reserved for validating the model, while the remaining four are used for training. The cross-validation process is repeated five times for each run of Bayesian optimization. The estimated hyperparameters and their accuracy are then derived by averaging the results across all runs (Diamantidis et al. (2000)).

Table 3.1 AI-based models' hyperparameters, their description and search space used in the Bayesian optimization process

Model	Hyperparameter	Search space
ANN	Hidden Layer size	(1, 20)
AININ	Learning rate	(10 ⁻³ , 1)
SVR	Kernel Function	(Gaussian, RBF, Polynomial)
5110	Kernel scale	$(10^{-3}, 10^3)$
	Number of estimators (trees)	(100, 5000)
RF	Max_features	(1, 7)
KI	Max_depth (The maximum number of splits a	(1.20)
	tree should make before it makes a prediction)	(1, 20)

3.3.1 Model Evaluation

In this study, the three most common statistical error indices, including RMSE, Relative Mean Absolute Error (RMAE), and Spearman correlation coefficient, are considered to compare the performance of waste estimation models at the test stage as follow:

RMSE =
$$\sqrt{\frac{\sum_{i=1}^{n} (O_i - S_i)^2}{n}}$$
 Eq. (3.2)

RMAE =
$$\frac{\sum_{i=1}^{n} |O_i - S_i|}{\sum_{i=1}^{n} O_i} \times 100$$
 Eq. (3.3)

Correlation coefficient =
$$1 - \frac{6\sum_{i=1}^{n} d_i^2}{n(n^2 - 1)}$$
 Eq. (3.4)

Where n is the sample size, O and S denote the estimated waste from the AI-based model and the actual generated waste, respectively. This study uses Spearman correlation, a non-parametric correlation factor, to measure the association between the estimated waste from the AI-based model and the actual generated waste. Unlike other correlation measures, Spearman correlation does not assume a normal distribution of the variables. Instead, it evaluates based on the difference in their statistical ranks (as elaborated in Snoek et al. (2012)).

3.4 Waste Management/Transfer Framework

Off-shore oily waste management faces more constraints related to cost and resources compared to response plans, emphasizing the importance of careful pre-planning to maximize resource efficiency. This planning can be divided into two main components: waste estimation and designing an efficient waste management framework. This section will evaluate the most effective framework for waste transfer, considering factors such as route number and cost. Accurate waste estimation is crucial for responders to arrange temporary storage and allocate resources effectively before an incident occurs. Given its significant cost implications, waste management plays a pivotal role in contingency planning for oil spill response. Efficient and optimized allocation of generated waste relies on several critical factors, including the location of the oil spill incident and available facilities, the compatibility of waste types with treatment facilities, the capacities of these facilities, the treatment rate of treatment facilities, and the number of generated points (oil spill incidents).

3.4.1 Problem Description

Industrial hazardous waste management involves handling waste at its source, transporting it to treatment or receiving facilities, and ultimately disposing of it in landfills if necessary. Through consultations with hazardous waste management contractors in British Columbia (BC), such as Terrapure, it has been recognized that waste transportation constitutes the most costly and time-consuming aspect of waste management. This challenge is exacerbated when dealing with multiple types of waste, as considerations must be made for compatibility with facilities and transport vehicles, with some waste types posing risks of dangerous reactions if co-transported. Hence, understanding the available waste treatment network is crucial for optimizing waste transportation efficiency. The key focus of this

optimization lies in determining the most effective routing system to minimize transportation costs.

BC has been selected as a case study for this study due to its relevance to the research funded by DFO for waste management purposes on the West Coast. On a typical day without spill incidents, approximately three to four trucks transport around 60 tonnes of waste to landfills in BC and Alberta. Consequently, proactive planning is essential to transport waste to each facility based on its current waste load and capacity to prevent congestion and additional storage costs. The efficient utilization of temporary storage or receiving facilities' capacity constitutes the second component of this study. Each facility is designed to accommodate a finite amount of waste.

Lastly, the third component involves selecting the optimal facility based on its capacity and location among other potential facilities. Therefore, a network incorporating all these components can aid decision-makers in promptly managing generated waste, mainly since prolonged collection and transportation times result in increased contamination and storage costs. The formulation of this model aims to optimize transportation routes to waste-compatible facilities using waste-compatible vehicles to minimize total costs. Several assumptions are considered in designing the model, including the impact of uncollected waste at generation nodes and unprocessed waste at treatment facilities leading to additional storage costs, the determination of the number of vehicles at each generation node, and ensuring that the amount of waste transferred by cars does not exceed their capacity.

3.4.2 Mathematical Modeling

With the assumptions outlined, a non-linear model is developed to identify the optimal waste management strategy from the generation node to the final destination (landfill), considering facility capacities and waste types to minimize transportation costs. Before delving into the model's design, defining the terms and notations utilized for the modelling process is essential, as detailed in Table 3.2.

Based on the assumption above, a non-linear model is programmed to find the optimum way of handling waste from the generation node to the ultimate destination (landfill), considering the capacity of facilities and waste type to minimize the transportation cost. Before designing the model, the terms and notations used for modelling are stated in Table 3.2.

Table 3.2 Notations used in the designed waste management framework

Index	Parameter	Value/unit	Description
t			Number of treatment facilities
r			Number of receiving facilities
S			Number of generation nodes (source)
1			Number of landfills
	Vsol	40,000 m ³	Truck capacity (solid waste)
	Vliq	$60,000 \text{ m}^3$	Vacuum truck capacity (liquid waste)
	TR	10%	Treatment rate at solid treatment
	IK	1070	facilities
		5	Tuonanantatian aaat
	cc	CAD/km.tonne	Transportation cost
	TCt	750 tonne/day	Capacity of treatment facility number t

RCr	1200 tonne/day	Capacity of receiving facility number r
LCr	Inf	Capacity of landfill facility number l
DIStr	km	Distance between two nodes

As previously stated, the primary objective function of the model is to identify the most optimized path within a marine oily waste management network. Consequently, all feasible paths between each node for waste originating from a particular node until it reaches its designated destination are considered design variables (Figure 3.4). For instance, the volume of waste transferred from source 1 to treatment facility one is denoted by ZST(1,1), as per the notation outlined in Table 3.3. Subsequently, additional decision variables are defined based on the number of generation nodes (sources), treatment plants, receiving facilities, and landfills.

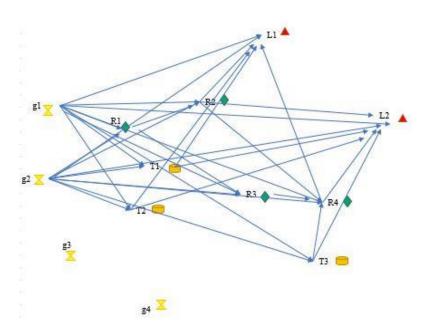


Figure 3.4 Schematic view of possible transportation paths between each two nodes in the framework

 Table 3.3 Description of the decision variables used in the designed waste management framework

Decision Variable	Description
ZST(1,1)	The volume of waste transferred from source 1 to treatment plant 1
ZST(1,2)	The volume of waste transferred from source 1 to treatment plant 2
ZST(1,t)	The volume of waste transferred from source 1 to treatment plant t
ZST(2,1)	The volume of waste transferred from source 2 to treatment plant 1
ZST(2,2)	The volume of waste transferred from source 2 to treatment plant 2
ZST(s,t)	The volume of waste transferred from source s to treatment plant t
ZSR(1,1)	The volume of waste transferred from source 1 to receiving facility 1
ZSR(s,r)	The volume of waste transferred from the source s to the receiving facility r
ZSL(1,1)	The volume of waste transferred from source 1 to Landfill 1
ZSL(s,l)	The volume of waste transferred from source s to Landfill l
ZTR(1,1)	The volume of waste transferred from treatment plant 1 to receiving facility 1
ZTR(t,r)	The volume of waste transferred from the treatment plant t to the receiving
	facility r
ZTL(1,1)	The volume of waste transferred from treatment plant 1 to landfill 1
	•••
ZTL(t,l)	The volume of waste transferred from the treatment plant t to the landfill l
ZRL(1,1)	The volume of waste transferred from receiving facility 1 to landfill 1
	•••
ZRL(r,l)	The volume of waste transferred from the receiving facility r to the landfill l

3.4.3 Objective Function and Constraints

As discussed in the previous section, the complexity of the problem involves numerous decision variables that cannot be readily solved. These issues fall under the Non-deterministic Polynomial-time hard (NP-hard) problems, which are computationally challenging and time-consuming to address using traditional algorithms. This challenge is exacerbated by the exponential growth rate of feasible solutions over time, a phenomenon known as combinatorial explosion (Hoang (2008)). Consequently, various meta-heuristic algorithms have been developed to tackle such optimization problems effectively.

One prominent meta-heuristic optimization approach is the Genetic Algorithm (GA), which has successfully addressed complex optimization challenges. Originating in 1975, GA is inspired by natural selection in biological evolution. In GA, a population of feasible solutions evolves through selection, recombination, and mutation, akin to genetic alterations aiming to enhance solution accuracy. The critical steps of a GA optimization problem are as follows:

- Initialization: A population of feasible solutions is randomly generated.
- Evaluation: Each solution is assessed against a fitness function, typically representing an error index quantifying its effectiveness in addressing the problem.
- Selection: A subset of solutions is chosen based on their fitness function values, expecting that subsequent generations will yield improved solutions.
- Recombination: Selected solutions undergo recombination operations such as crossover or mutation to generate a new population of solutions.

Replacement and iteration: The new population replaces the previous one,
 repeating steps 2 to 5 for a specified number of generations or until a
 satisfactory solution is attained based on the fitness function.

GA's main advantage lies in its ability to handle many feasible solutions while remaining time-efficient. However, GA's performance heavily relies on the randomness of the initially generated solutions, underscoring the importance of running the model multiple times to achieve a robust solution.

In designing an optimized transportation framework, the primary objective function aims to minimize transportation costs and determine the optimal volume and path for waste generated from a predefined source location to the ultimate destination (landfills). The objective function can be formulated as follows:

$$Min(f_z) = \sum_{s=1}^{s} \sum_{t=1}^{t} Z_{st} \times DIS_{st} \times cc + \sum_{s=1}^{s} \sum_{r=1}^{r} Z_{sr} \times DIS_{sr} \times cc + \sum_{s=1}^{s} \sum_{l=1}^{l} Z_{sl} \times DIS_{sl} \times cc + \sum_{t=1}^{t} \sum_{l=1}^{r} Z_{tr} \times DIS_{tr} \times cc + \sum_{t=1}^{t} \sum_{l=1}^{l} Z_{tl} \times DIS_{tl} \times cc + \sum_{r=1}^{r} \sum_{l=1}^{l} Z_{rl} \times DIS_{rl} \times cc$$
Eq. (3.5)

$$Z_{Total} = Z_{s1} + Z_{s2} + ... + Z_s$$
 Eq. (3.6)

Where the variables and decision parameters are described in Table 3.2 and Table 3.3. The optimization model's constraints are outlined s follows:

 Whenever a collection vehicle enters a node, it must exit it towards another destination.

- 2. Compatibility between wastes for transfer within a collecting vehicle is not a concern.
- Waste types must align with the technology at treatment facilities and the types accepted by receiving facilities and landfills.
- 4. The quantity of waste allocated to a route and/or facility must be equal to or greater than zero.
- 5. The total waste volume across all directions must match the overall volume that needs to be managed, excluding considerations for waste type.
- 6. The volume of waste being transferred to a facility should be at most of the facility's capacity.
- 7. some waste residues can be directed to a receiving facility or a landfill following treatment.

Appendix A

																																	4									
Beng	05.03.1977	No cheng. Ta iwan	Task Yes of	213660 (Bardq		6 (By))	No select col	۰	z	Intensival Taker Overs Relation Vertexion	z						÷	8.8			refere	-	• Rhistoreth Shiemer	2000 (Red)					z	z	z	z	-2 feer incheden spe Lood codes tolded mannerally vir									OSyd Cas Hactor Surmerics of sight Cas of Hactor Surmerics of the Cas of Hactor Steal Hactor Hactor Surmerical Hactories of Necessary Delete.
The tegenor	(6.0).1979	Banky Bay Indani	Task Yes of	H720 (Bardo)	30000 (Bardo)		Meel Arabian Cruit Oil	+		Gelt Chasty Council								100			Decodop-fredgagben is Tagbent (Tagbent (Atenit or deepen specific or	-	- Oalf Ol Coopey Bay	approx. 250 (Meleo)			Mathyle, non-specified weight	llay	>	Z	Y (200 Barrdo)	z	Hity spread to a bareh of a shared or Explains on and a fire that better of old to be better of the bette				25,000 (raders celedi)		3((ma)			O Digit Can Harter Stemmer of suggesters (2) and the state of the suggesters of the suggesters of the suggesters of the suggesters of the suggester of the sugg
Bye Lifeach (Brg IV 20)	09.03-1973	Oper Oaksanes Bay, Boate Slip Cheese, Teas, U.S.A.	Task Ye out	1000 (Bardi)		2 (By))	Licel sing Coult Bankler C	μ.	z	-U.S. Creat Creating -Base Service Corpusy - Clean Characteristics	z)4 (flat)	Xo.35 (heats) Seedmant	>	Α.	Y	я	31 (Days)	28 (Byo)			10 by 110 Bergestylelben Sålenn er Clan Chand Infansy Sålenner MY Lady Ale (oppuly 170 gill ens)	5800 (Feet)			3 seriest places, new specified weight	Sevins Sebess		N	N	z	Marying toloning and for beath of using. Type Mark 114(DP), Mark-12-4(L) Filter Fraces and sorber policies from any sorter policies for contact effects of 11 sorbers Planck).				400 (oxbis) milit	519, 342 (gibbos)	80,00 (plens)	1000 (plikes per bear)		ON 94 Cas Harme Summers of supplied and supplied and heart and supplied of SA. Harmes Martin Hope our Anne supplied to Sales of SA. Mare supplied to Sales of SA.
Abesian Venture	23.04-1988	Southast of Cipe Rass, Newfordland Carada	Task Yeard	23329 (Berdo	253429 (Berdo		Chles ded Cassoline	Α.	z	US Cost Gard Casable Cost Gard						z	z	۰		0 (By)0	+5 Aplants +7 Methat works	NA	NA	NA	NA	N/A	NA	WA	z	N	×	z	*Nec constrain serve trius als expoent secret sken.		NA	NA				0 (Bereb)	•	4 Splitter II notes Scenario bigident US rad for melecul Als, NOAM Hass dens March expres sell he societal Desirio
Andrá	06-00-1903	Ones	Task Youd	37600 (Banko)	33600 (Banko	0,640	Light famine Crube Oil	>	z	Ohns Ontel Creed of Conservation from the convicement mid prevention of publics mid public of CONN mid public of CONN mid public of CONN mid public of CONN	>		(1000)	(Neat) Neth	>	h.	z	141	38 Deyo	0.00	-2 tig beas - Overflijft witch								z	Z	z	z	*Ye set I moved 200 rad has from shore suith. A set of the set of									N SQL Cas Harrie Steman des Albert U.S. and Brentzind Als NOAM Brentzind Albert Cast Albert Cast Poperate Albert Cast Talent Million
Abhad Dazken Gespay	62.03.1988	Monogoleki Bivor, West Blockets, Peersylveria, U.S.A.	Bully	23810 (Bard 9		2 (Bays)	Deed Peri	4	z	-US Constituent, Noberg Natural State Roce LAVEABIA, O State Constituent Postsonion -US Electroma Postsonion -US Electroma Postsonion -US Constituent of Electrom -US Constituent of Electrom -US Constituent	>				z		z		(жд)		District in District in District in District in District in District in District in	Makelanases		30 000 for 0			Maliple, non-specified seight			N	N	z	• Los, demos and dou la, baldend finders groud of the ol	* 150 prople						L965 (Beredo)		with the control of t
Arrow	04-02-19-70	Chelidratio Bay, Neva Sorita, Canala	Task Yes of	TT000 (Bardo)		10 00,00	Besker C OI		z	- Department of Milato, Transportation and Coast Gardel Canada	z			3437 (Bed1)	z		4	-0.1	99			Malgirius or LANTO	- Slid Lidors Okophile Bull Type						z	Z	T (10 towns)	z	ON disks (dippersed by new arkins and dispersed by the arkins and dispersed by the Branch Changelock behaves and Kinn Changelock behaves and Kinn Changelock behaves and Kinn Changelock behave the Changelock between the Changelock behave the C					22 COLORD (Mark)	3 %00 (Berrdo)			(2) by Car listen termine of aginn (1) of the listen to th
АдоМафи	15-12-876	29 mle s south of Namucket Island, Manuachesetts, U.S.A.	Task Yeard	153000 (Bardo	183660 (Bando))) (Dev)	•Ns. 6 FullON • Outer skels	z	×	VOS Cons Chard NOAM NOAM Nocial fiels Consegg plat Institute National Africe Influents Society Volvority of Receiving VOS Codegal Survey	z		(pg)	South-Restrant	z	>-	>	13		, T	4 + Yearth								z	Y.	z	z	CODO CERL cards deployed between spill med const. med									Of Spil Cas II to be de Sementio of spil Cas II to be de Sementio of spil spil se de self des des spil spil se de Spil se
ARCMadenge	21-12-898	od Angeles Harbeer, Washington, U.S.A.	Task Vessel	560 (Berdş	814000 (Band 9	1 (Dex)	Alaka Neeb Stye Orak Ol	z		Cemalism and U.S. Coast Grant Alexandrian Trans. Long Bank Culfornia, U.S.A.	>		610 fox)	Di (konto) En tand Wot	z	>	z	ū		216 (Radi)	- 2 T mis - 2 T mis - 3 Leg Berge - 3 Ober Berge - 1 Helstopier	٠		(T800 (Pet)			Mahijda, non specified weight confection 3,00004,0000 (squareyord)		z	z	z	z	• Belair on top for adhumomors to work or confinement by the 5 wash . •) Committees by the 5 wash . •) Committees the upon supplied common to memory to supply pressure water a memory to deep . • High pressure water a memory to death and the form 2. He so I was a confidence to be more 2. He so I was a confidence to the most a few or I was a media or that is written to the surface of the ANDO .	+2000 + volestor s			-2300(cebicywito) Indodeg Serbena -2300 (nes) of olid legs		50 (Bards (100 yild)	X126 (Barre b)	280 Bardo	O SpICes Herde chemoto of spices and spices of the spices Herde chemoto of the
Angen	19 17 1869	Materia Arbipdays, Perupi P	Task Yeard	17300 (Berdo	23800 forest	(By)	Motor May Ovid Ol		z	Perspecie Novy Intersibility Taket Over Politics Forest Commenty Taket Fore				11.9 (mpl)	z	>		17.9	0 (Bys)	46 (Days)				Malek					>	z	٨	٨	Highwoose he was waking				- X0,000 (seezes)	- BARD Bears	10,000 (cable meters)	00.0	0.0	ON SQL Core Haster Schemarkie of afficient US and treatment of the Core of the Core of the Broper and human Division. The Confidence of the Core of the Confidence of the Core of the Notice of the Core of the Percentage of the Core of the Marine on the Core of the Percentage of the Core of the Core of the Percentage of the Core of the Core of the Percentage of the Core of the Core of the Core of the Percentage of the Core of the Core of the Core of the Percentage of the Core of the Core of the Core of the Core of the Percentage of the Core o
Arbin Gdf Havai	P-61-1991	Person Gelf, Karsa k	Task Yeoreb Fasilites Pipelanes Matforms	socco flanto		72 (Dex)	Kness Crafe OI	>	z	* Use of Nato Chear Chearl * Sand Areha Areha American * Old Corporary * Mar aren'elypid and Environment Frotender Admir * Royd Corraminate for Albey (and * Yorke) (mi))				111		107		17	*ARMO	25 OMess					>	z	z	,	- Tweeken day life high felt oil enther life - Elikken Berna Uned						HORGO			of space of throse Scenario of space of
yea Hillings, Ayea Hill Barge	28.07.1960	Oakreke Bay, Teas, USA	- Task Bargs -4 Cargo Tasks	364% (Bardi)		1 (De)	No 5 Oil Neons Olf Guidyk Peol Nodo		z	- UNCO Mater Sultry Office Obsteron Rieds Formed Instructed Style - Materiological Style - Chan Gulf Nordaline - Chan Gulf Nordaline - Chan Office Sultry				11 (raph) SouthWest				27.5		11 (By0		*		presto	retere		Multiple, non-specified weight		>	Z	z	j.	though dock of 2 vaccoring as Loop property whileg at head re-									O Spd Con Hactor Summission of spd Con Hactor Summission of spaties U.S. and I starts from J. Spd. NOAN Hauston March Respects and As moreon of D 4 dec.
Amon Calls	16-61-1978	Britany, France	Task Yeard	BO19048 (Hawlio	S NOB (Bento	14 (Dby)	Arabian Light Orade Iranian Orade Humdor C	۸.	z	-17099	z			36 (Brith) Newth East	z		×	13.4		(Year)		enden		250080			630 (Metacrasa)	desphile chall	>	Z	٨	¥	*Receives askins and thick in source of Bones are delenment and the same and the thick is all conserved from the in- turbulence of European and In- received when the CO's sourced in the land as no food with frost and body on remarkly	• TOO precede suite ()			~ 20,000 (keens)	- 80,000 (beans)	Miller			O by Coefficients Seminated of a spike contrained by the contrained the contra
	074031980	Hazigatos Boach California, U.S.A.	Task Viscod	9450 (Bardo)	90148 (Bando)	0.000	Median North Stope Cristic Oil	z	4	-USCOA Arise Safey Office Group Los Anglise Long Bauth				35 (Bests)		z	z	0	10 (0.60)	55 (Dex)	+25 support been clow sessels.	15 +	-2 Open Water Of Containment Roosen y System Sharming Barters -7 Marco Class Y -3 Jugo of Shee	Make	2 leasins		Multiple, recespecific dise ight			z	z	z	- Estern Berns offs of w for despite evidents well high his small					MARK (Bands)				Jilyod Case (Baceloc Saemus a dydgoras U.S. and Heroniteda Al.). NOAM Heardina Maharisi loppes e anl Associente III visit
Amenda Hess Od Co. Strage Tarks	20.00-8989	PortAlonels Line vor Bey, St. Creb, U.S. Vigni falends	5 Steenge Tanks	16000 (Barolo)		0 (Dgy)	Havy Crafe Oil	z		- USCO Alante Steke Torne - Hoss Ol Vrigis Islands Corporation Power Bast				140 (skb)	z	٨	٨	29.3				opdy peus		2000 ff ed),24*			Makife, ne speafal wagt	of once	z	N	z	z	· Extern Berns soul				5000 (Barrds)			1000 (Barrds)		N Spi Case I Bard oz Szermeto A figika m U. Sasti bere aktivad Spika NOAM Handrach Marzile Beprete and Austream Division
AmonYature	G+E-1986	Swand Ring Gardin Gry, Gergin, USA	Task Ye out	1800 (Banko		1 (D6)	No. 6 Pad Oil	Y	z	Personal, Creates and by U.S.Coust Grand				11.9 (bmb)				163	19 (X-9K)	100 (Deys)	2 stombon is			St0+(Fed)	3 kentons				z	N	z	×										og Que cherons Serente de dyster (18 see beverdeed dyster (18 see beverdeed) glyk, DOAI Hawten Marien blyken in begense an Associeted Disk and the control of the control o
Abens	30-07-898	Okanicarhor Bar Chand. 11 rulos SI Cararros Loxisiana, USA	2T ask Yearsh	cato (Berdo		7 (Bys)	Yeacuda Mary Oute Plee Ords	z		ClassChilf Assistation				1520 (breb) Eastard North East	*			36.7) (A)	61 (Ray)	- 2 Bages - Tepbest			Miliple (ree-specifiel)					z	z	٨	z	- San Bartér							981989	63,000 %,000 (Bardo)	N Spil Case H to to Stemation of spilotate Library San International Conference of the Spilotate Conference of the Spilotate Conference of the Spilotate of Spilotate Spilotate Office of Spilotate Office o
A Baydrain	10-10-1977	Gora faly	Task Ye out	13.0 Barelo		2 (Dey1)	Navasi Crafe O.I.	٠	z	Guesa Poet Authority	>			26 (brob)		z	z	101	(Dex)	(Dec)	6 Toplesis	-	Oaku						z	z	Y (1190 hared)	z										O pp Continues Sommon of objects objects of objects objects of obj
Name of facilities	Date of incident (different-5039)	Loatin	MontTyle	Yelense of spik ol.	No. of ontpo	Time big for steeling response	OlPvdad	Confession (YN)	Red dest type Station (Y.N.)	Beyrane Apre y	Odes (Y.N.)	Ocean condition Why period	Wave In spire	Ward videotry	Skerry (Y/N)	Andless condition Bakey (CVX)	Choulty (TCN)	Ocean Temps (C)	Olikee	Repend time Cooker	No. 8 spense younds	No. Shimmor	Type of shamer	No. catainmen been	No. absorbest honer	Type brand of absorbert boom	No. Seebe nt pad	Type brand of Response Method serbest ped	Viewer Trads (Y/N)	Books bereig (TON)	Department (F/N)	Biomore dialon. (Y.O.)	Орис	Number of responders	Number	Opady	Quantity of generated solid to sate	Quantity of generated liquid was to	Oher	Yelsens of Recovered of Brees Marsons	Rices absorberas	Mérene

Parieta No. 5	17-01-1980	ger Delta, Fernaltos, Nigeria	Matron	30000 (Banks)		12 (Deyx)	Ngries Crube		Z	Bol Adás Corporales								292	NA	5 + (Marks)	sile to	NA	NA	100	MA	NA	MA	MA	z	z	Y (ARG pilon)	Z	*8,500 galaxis of old Crew detrymels				WA	W		0 (Barelo)		CO Spel Case Harrie Sammeries of eligibitists. Samistranderies of eligibitists. Samistranderies Beyone Machanists although the Beyone failed and control that Dates 1948. Control Spel Case (1948) Data Nober (1948) Il Spel Not (1948)
Exe tiles	34.03 - 8989	High Reaf, Drice William Society Andra, U.S.A.	Tank Ye and	24000 (Berrido	5300000 (Calco)	35 (Berr)	Profiles Bay Croit	×	N	* Alyocka Pelitic Service Coopury * Fourth State and Local Agentics: * lice hidings earth Chan. Oyen locus	z		18 (84)	up to T0 (Jests)	Z	>	Å	*	(10g(1)	884 (1/40G)	1)400 Yeards • No Areaf	Naipk	Ok ophili c Die Skimmer Ok ophili c Die Skimmer Egende Brand P aklis fich Skimmer (Egende Pil)	No (mks)	Make	Reded Pads, Srare	Malph, non pecified weight	benthen	Α.	Y (15,000 galeen)	Y (8500 plens)	Υ.	*Fiddl at brins Bermel *Joe off specific Design from groying Survear Thaday (OI * Sar from Streeters *Name Charing Procees: * Sar *Sarbert Bermer speed *Sarbert Bermer speed *On the Control of the	+1100 percend	-	Barge 13.000 (Galleris)	2500 (tens)	10 (00 (teems)		710 (Barelo)		A Spirit or the control of the contr
Doors Bayway Berthery	00011000	Arthur Kill, New York.	Nytoe	13500 (Bareh)		0 (Deys)	No 2 Harling OI	z	٨	Committee of the c				302 89/k0				12	77 (Dg9)	0, (A)	40 vacuus buds 10 buts 4 le laques	a	* Sof prepeled Shamer * Marco Shamer * Bit Shamer	0,000,000	400 (00) (6 40)		Mahiple, non specified weight	best best	ν.	z	z	Y	Translass were digited the trace craiming Translass were digited the trace craiming Translass were digited to the weapon tool shoulder It follows the ress.	- 1500xm percent					19,000 (Barelo)		HO,00 (calices)	of Self-can librates Stemmer of against US and "Self-can librates Stemmer of against US and "Self-can librates" - Report to the heavest Report and the Second Self-can librates of the heavest Report and the self-can librates of the heavest Report and the self-can librates of the heavest Report and the self-can librates of the self-can librates o
DIMI (I)	0440.1977	lation Bur, New York, New York, U.S.A.	Task Barge	10000 (Bards)	(Bards)	(6g)	No. 6 Fad Oil	٨.	z	U.S. Constituted wheates when I Tenna . Tenna . Tenna . Tenna . See Land Book orneral . Englanding Company . N	z			17.1 (Arek)	z	z	z	1	10 (Bays)		«Tugbons «ADAPTS	Matte	Lot blood Shims or Ma or Shims Cold Worker Shims	300 + (Sec	NA	NA	Makple, non specified weight	OlSane	Y	z	z	z	"Lightering Of search in pile spice of easted in be beyond to chanch a "Marring latter out (Showin, 19) pile follows, that it lags, then generates, high velocity void or the min on enclose impact)									O Spel Co tell meter Co Stematic of agicus of a special control of agicus of a special control of agicus of a
ESSO Pueto Sico	03.05-1988	So skylyt Sove Lexistera, U.S.A.	Task Yound	23.000 (Berrdo)			Other Blak Feelersk	z	٨	-0.1.8. Onest Gearchard Althorise Strike Town Marter Solliny Office New Orbans	>			3.9 (mp3) South	z	z	z	19.4			- Office bear						Muliple, nes specified seight		z	z	z	z	Dhig and station operation Ol d supated with their carration.		~	S (ples)			10 (flamb)			-U Spill Ce Barton Semantos el spillor U Se diferente local PAN N Che Randon Sacries Beyrese and housewast D view
Basiv	06.05 878	orloli. Soutest Cost of England	Task Yound	52500 (Berds)	117,280 (Bards)	0 (Pays)	Hony Fud Ol	Α.	z	Duck hard England Justiceles	*			11.1	z	z	Å	971	NA		* 5 Tup 22 Ye sods with spin ying capabilities	Nen Speaked	NA	Mahiple (non Speaffall)	NA	NA	NA	NA	N	z		Y	Now up of agreent bow so then of					100 toms		NO (cases)		A Spill Care His trees Summaries of suggest Care His trees to Sand Husenshood of Agricus of All North Husenshood and All North Husenshood and All North Husenshood and All North Husenshood All North Husenshood North Husenshood All North Husenshood North Husensho
Bech fras Ol File	23.04.877	Newsy North Son	Parlon	2028.1 (Bards)	NA.	(Bys)	Bodk Ová Ol	Α.	z	• Rad Mis V • Newsgan State Polition Control	z		4.6 (fat)	Species with	z	4	Å	9	8 (By)	NA	,	W	NA	NA	NA	NA	NA	NA	z	z	z	z	Capingot Rig Wave Author Wave Author Wave Author		۰	0			0	0	۰	Ol Spill Core Histories Stemmado of significant LOS and Interactional of significant LOS and Interactional topograms and Assemmed Devian horizon and Assemmed Devian - Michael Spill Company of the Core - Efficiency for Long Orderer Col pick, J.E. Dager Schmidt Efficiency Text. Nelsons.
E-24	22-13-8965	Shork Island Sound Fishers Mand, New York.	Tank You d	TI (Barels)	Stem (Barrels) No 6 Fool	2 (D ₃ y)	•No 6 Fod OI •Deel Fod	z	Υ.	Marine Pollution Council Council Council Child State May be greated Child State May be greated Child State Copies of the Post New York	z		5.6 (fart)	25 Ukaro Dae	z	*	Å.	10.4	301 (Days)	13 (Days)									Z	z	z	z	sking generals broken						BS (Oalbes)	Recognes		ON Squit Cost Havior Streamsis of squitter 15 and less extend the NOAA Hamilton Mark the Aspeter and Assessed Division
Contains	18.09.1902	Dilaware Store Manus Hode, Remobiated, U.S.A.	Tark Yound	2kato (Berds)	315000 (Bards)	17 (Mks)	Agotan Carde OH (Type 2)	z	٨	-U.S. Creat Grant P.Hs. dephased Abade Stale Trees - P. Madekal Fr. Department -U.S. Arroy and Novy				5.9 (right)				222						Malife	MA		NA	form	Z	z	z	z	A Alphamb She realizes on shoot for executify re enayed with shorter									Of Syl Case Historice Stemanics of digitization US and leaves about Syls, NOA, Historica March Syls, NOA, Historica March Syls, NOA, Horderd March NOAM, Indient More
Cardio	19-03-1961	Kill Yank of New York U.S.A.	Tank Yeard	1716 (Bards)	2073/0 (Bards)	(Bys)	No. 6 FactOR (Type 4)	z	٧	U.S. Cosst Clearl and Marke. Strike Town Dean Sideors Company									6 (Bys)		- Bargitto - ADAPTS	7	ABF Dyamic Indias Plans 1001 1003 Self Prepelled Skinners						N	z	z	Z	-Off building and driving Open deces- of, ightering with an AV-Dok wall to. And in Marter Transit Copysters And Arman and Transit Copysters. Eartha and prespingsoysters.							35.734 (Barnelo)		Notifice I form Stemmon of a galaxy 15 section as well of the NoVA Hearthean Market Septement II have been been been been been as the section of the section
Grites Biran	D-16-1978	Feb Ser, South Vales	Tank Yeard	23990 (Barelo)	25720 (Barelo)	(Byy)	lk sy Isalan Orde Ol	N	Y	Her Nije sy's Const Guad British Poresherr								13	2 (De)9	4 (Day)	Hem IP decken than The sock with therepart sproying Tuplen (Autoff Towns Tug Casaltena)	9	Marry Bo, Skimmer with amono- sockierpemp Skora a Miniskimmer Oversepark	Materia					N	z	Y (1600 Barelo	N	Of basing speakes adoutly 1846-Center 1846				335 (beau)		20,151, 200 (Bare b)	368 (Berrds)		8 Spi Case Hancie Stemmie De Spi Case US no benezioni Proceso de Novo Hancies Marche spener atti Assont all'Prises
GevreeMán Pas Bleck 41	0.62-1970	11 Miss Bat Mississips River Delta, Losiana, U.S.A.	Mafron	63000 (Barrds)		22 (Days)	Ondi Ol		z	* Cherra Ol Conpar *Rajber Repense Tens *U.S.Con (Gurd	z			11-22 (brosh)					48 (1961)		Octobers biology Octobers Brys Skinner Brys Skinner Bry Skinner Bry Ottober St. Skinner		6 Swelleh Skimmen *Chanc Skimmen	14 (MBcs)		- J M Been - Note Been - Kein Harden Been	Mahiple, non-specified weight	See (2100 bile) Ovice a Pub (60 1950)	z	z	Y G000 Bare to	z	Clearing was doub a reduct rate of a med shown: • All webs on platferers were on prod • Two damied the prints sproyed on platferers.	-2500	~	1000 formels)		6927 (Barnelo)		306 (Bare to		of golf-ce filtrate Sement of golf-ce filtrate Sement of golf-ce filtrate SAN III and sement filtrate ANN III and sement Debug filtrate and sement Debug filtrate Carbysha Annotate All Pobles for semint Nethern Pobles for semint Nethern Chang San Education Semint Chang San Education Semint Chang San Education Semint Annotate Agency A brief Seviet Changes and Service Semint Semint Semint Changes and Semint Semint Semint Semint Semint Semint Changes Semint
Chercellavai	66.61.69.10	Deer Park, Texas, U.S.A.	Tank Yeard	20000 Barelo	(Bardo)		Seal Maris Crade OI Oktablik Grader Footbands		Α.	Odišatie Tean Odišatie Tean	z			(94K)	Å	*	Å	197)(pag)		Mality of the board Mality of the board of the b	Malple	Large Exces Shimmer	Malple			riften popodo osa splipsy.	alsorbest massrabs	Å	z	z	z	rational is a charby on rational and the charby on	* A Chat Chard percent * 2 Coat Chat P dates incertigates								ON Spil Cast Hazri on Stemmate On Spil Cast Hazri on Stemmate Spile, NOAN His netous Marriel Response and Jones 1981 Division
ANNO	00-101508	Preu Des, Gâe	Tark Yound	40000 Barelo		2 (Deyx)	BAAP Caule OI	z	٨	- Witnesday Solvage - Chica Martine Authority	z				z	>	Å	19		73 (Deyo)		NA	NA	NA	NA	ки	NIN	NA	z	z	>	z	Corden 9527 (92 p less) and Shis gree LTE (33 p pless) of deperpents and	· Branca (realizer classics)				IIII plantic bugo) per da y	NA	NA	NA	O Liqui Cone Bed on Stammaio (Appl. 100) I Library and (Appl. 100) I Library and
Beenlings	0.11 - 979	Galvesten Bay, Texas, USA	Tank Yeard	24761 Barelo	466000 (Berels)	(Days)	• He ole of crude (high supidits contest)	Α.	N	Chen Ware Inc. Seek Interneted Inc. US Ney Separator of Selving.									92 (Dept)	27 (Deys)	Ower Ourse Ourse Ourse Ourse Outcome Outc	٠	Shimu ingloanius (DADCRS) (1) Ledde all drimmer Auto Class Visioner -2 (DANCRS)	106 E (Fed)	N/A	Gooff or Been (contained)	N/A	N/A	Y	>	z	Z	dead describer to the dead of the Wast of the dead of	One Coat Coat Otto Viles One Otto Viles	#1feep of DANDGES 1/ANN TS				1/2 100 (Sare b)	6100(Bare b)		Cl. Spd. Cor. Histories Scientwice of Spd. Cor. Histories Scientwice of Michael Conf. State of Spd. Cor. Michael Conf. Spd. Cor. Michael Conf. Spd. Cor. Michael Conf. Spd. Cor. Michael Conf. Theopen, Electric Poly. Michael Conf. Theopen, Cl. Cor. Poly. Michael Conf. Cor. Michael Conf. Cor. Michael Conf. Cor. Michael Conf.
Balays Pyline	30.03-1990	Maghary Store Knapp Star, Parasylvania, U.S.A.	Dylan	1790 (Barrds)		Days)	Nersome No. 2 Rel Oil	z	Α.	Regional Response Town License Adulte Soft of Office Pictobrough						>	¥					Mahite			sphere		Mahiple, non-specific divergite	llay	Y	z	z	z	-16 of Till or Dates						D1H1 Bardo			O Spil Can Hardon Stemmers of highest Day and the standard Spil NOAM Is are of a Marche Reporter of Associated Spiles
Braden Marina	09-01-19/8	Sao Sobasian Sao Pada Braid	Tank Yeard	TM00 Barelo			Karat Crafe Of	z	٨	U.S. Coat Clard U.S. Bardenman Procedure Apercy	z			North East				316									Makiple, non specified weight	llay	Y	z	Y (5.50 g los)	z	Old shared even road by freet and baday Hopeway spread on beachest to about ods									10 QCC in literary is memory - Ol QCC in literary incomes of the control of the c
Boudand #65	36.01-1977	Security Newschoots, U.S.A.	Task Begs	1932 (Barelo)	T6891 (Barelo)	(B)	No 2 Horse Hearing Oil	z	٨	Centre Station ing Coast Street Inc Made State or Mark Street	z	12 (mckm)		722 (Brob)		*	¥	13	24 De00		- No care Tracks - Statement of Vocare Units - Tray bods - Dages	3	Norry Marco Shimmer Lookhe all Clean Swoop Shimmer Endern Repe Shimmer	Multiple (non-specified)	Nahly (con-specified)		NA	Malipla (non-specified weight)	Y	Y (15,600 fems)	z	z	Lea (19 above) cortains (ibs of and proce that thought of the post of the	•63% percentle flighter ratio	1 places he feet	NGG(Bay)	576 (100) (10)	19990(Ban)	77.000 (See 6.)	2004 NO (Sers)	rediffers	ON Spd. Co. a literate Scientario of efficient 10. See determent of efficient 10. See determent of efficient 10. See determent of the property
arm of freibns	Secol beidet (diens-3353)	ceation	concliye	ridume of spik ol.	is d ongo	ine lag for statingropene	XIProduct	Cestiones (YN)	Sadden (Y-6V)	Stanly reads	oh Om	Dem coulies Wire pried	Wave biggle	Wad wholy	Seasy (Y.N.)	Rainy (Y-PO)	Cheels(TVN)	Sear Temp. (C)	Office	Onloce	to an primervant in	No Shiemor	Type of shimmer	No catalened bean	No abotest boss	Type beam of absorbest becen	No Soders pal	Type-brand of superne Method surbant pad	ManuTeds (FN)	Insula bening (Y-PO)	Disperses (Y.N.)	(YVN)	0900	shadeed Roponiko	Number	Ober et enge Operity	wantey of generated solid waste	wanty of generated head wante	Ober	dans of Pres skirners	Free absorberts	elevace

Sai Bach Will Blowat	31-16190	Stall Beats, California, U.S.A.	Patien	20 (Berds)		- (Week)	Grade Oil	>	z	-L.S. IV, apteo formation -US. This and Wide Service	¥			5545				17.8	0 (Days)	2 (Weeks)	Sharming Yes of North Sounder Pato Review Tag	-	Invend Plant Type Shemar Otophile Ibdi Shemar Otophile Reper Type Shemer	Nicee	Note		Maliple, non specified weight	even pe cified	z	z	z	٨	"A Agrication and process "A Agrication and process with a second and a second and a second and a second a second and a second a									-ON-pill Care I lavelo Stemanio of signific North Homosted Pill NOR Homosted Monthle Reports and Assement Delices (19)
Prosidure Rhura	24.06.1999	Mavare Bow, South of Marcin Hock, Perroybyana	Tank Vessel	73310 (Bareb)	45,000 (Barelo)	0 (x6q)	No 6 CI Beay inflatrial grafit)			- Main Agney Loud Response Ton - US Cost Comd	Y			N.4 (mp3) Egiz-sacio	z	Å.	×	0	+ (rég()	- (Yau)	Chem bid bendon Hopper Bugs Fieldig Yound or nicen to de not.	Mahle	Steldam Marmine Flooding Wor Shirman Dalescare Bay and River Corporative Shirman Didday	Mayte Mayte		• Goodyar J ff • 36° Sox Cartain	Miliple, no specifid veight	OlSeares		z	z	z	his his of the fact of the control o	300 personal from Nasional Count	per valer OI Octalement and Recovery System			S (ca)				1005 Con literate Stemat in 100 of spingles 100 to februaries 100 of the control of the November 100 of th
Ned Pudik	D-07-1988	free flatbox, Ospo Orbid. D	Tark Veset	15350 (Barrds)	(Bards)	l (feet)	Beaties (Nech Sea) Crade G1	z		- Copea Chai Ava ON Sell An oritin - Miller Environantal Sovice - Chann Marie Sovices - Chann Deliver Sovices - Under On Underhain Sovices - Under State Channel - Marie Sells Tom	Y			11.1-10.2 (ruph)	z	z	z	19.6	0 (54)	= B ₂				4400 (for 0			Mikiple, non speedind weight	ton speat fed	Å	z	z	z	Offerest demage task pestoped into me compy task. • Filter Percor into ments men offered with protecting the a nea		0		118 gabis yando		11,344 (Bends)			10 SQC or Bleeder Stemmen 10 of specifical Used International of Specifical Use of Stemmen of Specifical Specifical Specifical Specifical Beyons and Assessment Divisor 18
Nethoda	23-13-1988	inys Habour, Washington, U.S.A.	Task Barys	5500 (Bards)	((pred))	8 (Dept)	Beecher C	z	٨	CUS Nations Part Servic Washington Star Department of Biology CUS Coast Chang Coast Chang Coast Chang	×		640	(odes)	z	¥	z	6'6	0 (Dys)	1.89 (Days)		0		0	0		Maliple, ten specifed wight	of drama h • Ol Searce • Propents	z	*	×	z	**Obton manual champ **Oberts and no serges and doubt of free health; algues, and both of free health; algues, of the beath hand the beath hand the beath hand serges or of seal for her meetle with the Organie Manual Ann and the of the doubt dan free for the doubt day. The beath for committee of the period of				585 (cm)					Night Case II does Sommein Wegist and O seedsteen sivest with NDA I loant on March operase and Assessment Division
Nep Boy	0849-1930	of Ghoten Teas, USA	Tesk Youd	(Bare b)	819.80 (Bando)	Essiban I (Rost)	Angelon Pilance Orale Oil	þ.	z	• US Cost Cond is Obeston • USCS Safey Office	Y			10.85	¥	z	z	17-22	12 (Deys)	16 (Deyt)	* OPERIES * OPER	12+	U.S. Nacy Supervisor of Salvage Skimmen	(000 (64)		36			¥	٨	Y 01,300 galloso)	٨	Mach of the of exponent or beautiful to the of exponent of the of expose who is a valer and few splitation. A place of the or and th	• Numerous Volunteen					-tito(100 (Barrio) Educing -13 (GB (Barrio) Tax contare	0	0	N Spil Case Hatedoc Sheemad of the spilotest U.S. Sheemad of the spilotest sheet with the spilotest sheet sh
MCN-5	3101.898	Comes Cheed, Sheem Nee, Waltegen U.S.A.	Task Bergs	1004 (Bardo)	1900 Gardo Havy Cyde Gas Ol 24 (Barako Harrandak Fod Cif & Mario	(Week)	Iftery Cycle Car Ol Intersective Field Ol Naries Diesel Ol		z	-The Obyest Toping Bage Co of Stafe Workington									20 Deys)		• Tog Boat(s)	3	Largeins whed planeitype shimmer Obequing chall by pe shimmer Sonal Or oping steps of type Sonal Or oping steps of type Sonal Or oping of	Nahple					z	z	N	z	*Sid-uge Cyenifons						lifving experiments to oil was recovered	NA	NA	Ol Sell Core Hercies Surmanies of against U.S. rad learned and yells. Nov. All Bearders in Novel Hercies Surmanies of Sells Core of The Line Hercies of The Line of The Line of The Line of The Line of The Core o
Likebard Pipeline Company	05-03-1991	V. S.O. Krossey, Nieke Blanco	Pyline	(spang) 9,100		0 (Deyo)	papeo	z	٨	Marcela Count Guard Marcela Polarización Control Agrey Marcela Polarización Control Agrey Marcela Polarización Marcela Polarización Marcela Popartent of Intro quarcy Mangament Marcela Departent of Respersación Marcel Marcela Departent of Natural Meserres	N			18.2 Own to	z	A	¥					Makific		Make					Å	Å	N	z	Whites used sign greet o push A being iss on a removal basis on Too blocks were gas you with hot want committee out of and recovered white teams									A Night Con Harvier Stemarie degris and Sarai Bernarien gleg A. XXXII Earsten Maraia forgrass and Assessment White is
Mardian	15401-1979	93 less receb-out of Sydney (Nova Scota) mar Capella toe Idand, Nova Scota, Canada	Tank Vessel	4300 (Banels)	13500 (Berels)	0 (Dept)	Bedon C	>	z	Elovicenza al Prot céteo Service Ragio cela la revera al al Energie con Teen Consilia o Cost Garet Amorphe et El nivora est Service	z			21.7 (brid)				10-			• Barge (s)			Nakyle			Natiple, non specified weight	servers material	z	z	z	z	Bow so ken twee d 20 miles free Nova. Se cals and sank She not do known for the Nova for the world to Nova flaving how shown Sonta to server free in the Nova Sonta to server free in the Nov		Makele	55 (pike)		(cling per pl (00/90/)	2,300 (Bursts)			Old Spice Blacker Streets of Spice Spice Blacker Spice
Aphe	1600-1900	Sugarav Bloce Bay City, Nikhinger, U.S.A.	Task Nesed	3000 (Berds)		30 (Ma)	Undersited Groodine	>	z	Banger County For Departs at USCS present White meters Company Company	Y			8.6.10 saph)				р		6 (Medio)				Malph						z	z	z	• High and Som actors • The broke of front The broke of front Marketing resentations on a stress front						242 (Barelo)			Dipg. Car II acto. Sumario of spigor of Sari Brewaria pipor of Sari Brewaria pipor of Sari Brewaria Saria Saria Sa
Irea/s Oallings	1701-1977	Pacific Ocean, 50 miles North Lisienski Hassa inn Edunth, U.S.A.	Task Yeard	337600 (Bends)			Light Coule Oil	÷	z	Regional Response Trans States Advisory Citrap US Creat Guad	z			и					5 (Deys)		- 2 outox - Nator - Javis	NA	NA	NA	MA	NA	NA	NA	z	z	z	z	. Were askee degreed be of except.				MA	NA	20	SN.	W.	Ol St. Con librate Sommeto el el gale del Control Sommeto el el gale Donne Sommeto el el gale Donne Sommeto el Son Milla del Son Sommeto del Son Sommeto del Son Sommeto el Son
enspanding	15-11-1979	Bottachol, Turkoy	Task Yeard	(84T78) (Barrdo)	714.780 (Barelo)		SS Skr Cala Oil	z	*	- Trakach Ney - Doter of the Mersers Ru- Dotest Dotest - Betrakend Taker Over Politics Federates Lid.	×			17 (Amil)				333	- Gras)	NA		Make		Make					Å	z	z	z	No Beyeve de demep up odaed skowskie ree dod									Ol Syl Cas litted or Same ale of agelous US and interested Spike, NOA Alla a stoke to be to be Represented Assessment Division
	10.05.1978	Per Seiten Charrell ampallay, Florida, U.S.A.	NecTak Yearl	952 (Berrds)		- (By)	Busher C Ol, LightD and	÷	z	His browness (Constitution) Constitution of Constitution Noted Device of the Office of Constitution - Registed Response Constitution - NOCCOLOR Stake Town				(phs)				. W			Begs CO3200 was rife pilaten paralteat	Neilpe	U.S.Nay Stemar	1000 (640			Makife, non specified weight		٠	z	z	z	*Manual dements of the set of saids in the sas of frost and lookers, the large of the said same on, and the said care of the said same on, and care on and and same pack a. Look Prost set working * Polywalte a shouling covered joint in the said said covered plant in the said said growing plant in the said and growing descriptions.						(48 Bands)			O by Don Harder Serration of against User Berarder Angle of Don Harder Serration Angle of Don Harder Represented The Server Represented The Server
Buco	1104-1901	Generalisty.	Tank Yesod	142857 (Barneb)	100000 (Berelo)		brain Harry Ords OI	×	z	His deer Man tr'in Coron Italy Increased Trade to Coron Publishes Fred as in Hodden Coron Count				(9ko)				691				Malph		Malyk						z	z	z	Namalemos of clint forby Divigity are a surry do Roore safen of and order red roar falling.	same ones volume as			26 MD (subsympt)			35,700 (Barelo)		ON PAIC SE HEART OF NORMER EN CARREST OF A TOTAL PAIR OF A SERVICE OF A TOTAL PAIR OF A PROPERTY OF A TOTAL PAIR PROPERTY OF A TOTAL PAIR OF A PROPERTY OF A TOTAL PAIR OF A TOTAL PAIR OF A PAIR OF A TOTAL PAIR OF A TOTAL P
	G-10-1962	Caffor Arbei 250 for NW Quar. 140 Mer N Sand Arbeit	Platices	(Barrds)		S (Deyt)	No grap Of	z	>	Sinte Tak contrated by AdANCO Bahrain Perchant Company and code of front hells havin Cooperson Quar Committee Out Committee Cold Annual Management Management									n Deco	8 (Medio)		Metiple		Maliple		dage dod SMB hose. Light hosess Navy Borns				z		z	* Wace boen were abed in septy, discretional should ago one fore cell building pay hars were used for building pay hars were used for a preconstruct boroning has be of near	· 30 · promede	Malife Ol Darss	25 (pallor)				15000 (Barelo) Oily votor ceatains		O Spd. Cariffices on Stemanian Spd. Ann All Littleway Market. It shows to and someous Devices of the Indian Friend Works) The Indian Friend Works of the Indian Friend Works of the Indian Friend Works of the Indian Indian Friend Works of the Indian Friend Works of the Indian Friend Works of the Indian Friend Works of the Indian Indian Friend Works of the Indian F
Hatlerrad Drawy Took Deep	26.05.1976	Backenseck, New Jersey, U.S.A.	Belby	((Bards)		0 (A)yi)	No Fred CII	z	*	US Cost Could Histories has along res Deed greent Corne i sien				v 0.88				103	1 G G					300 fe d					٨	z	N	z	Of recovered no water by map on an and sometimes Config ON of up galan midel by there much grow it would by the recent may be a second by the second beautiful to the second				700 (640)					on - ON-Spill Case II skeete Stermanies A of Spiriter U.S. sell Skeete Seetel Spill St. O'Co. Nison and Seetel B. Repense and Seetel Spiriters
Grand Eagle	28.09-1985	Debrure River, Max on Hook, Peres Sciente, U.S.A.	Task Yessel	1057 (Bare b)	530659 (Barnels)	0 0000	Neise Ceole Oil	z		• US Cost Cond	Α.			17-21 faces) Neeb-Norbook	à.	z	z	222	Period Period	41 (Deyt)	*21 Vacentratio *Bage *Tag *4 valid Routs	9	DELBAY Skiener Necy Skieners		55,495 (fat)		Nahyle, nonspecified weight			z	×	z	Change in idea that the formed formed in another in another in the control of th	+442 pen erect			200 julie yardo			Seo (bank)		O SECO, Bare Secondo COSE Collegica contrato cologia Coll Bare Cologia Collegia col
Globali Sun	B-66-1975	Calf of Masico	Task New d	7000 (Barrds)		30 (Min)	Anthian Cruit Oil	z	*	Varies Roponies forment by Pariems of the Galf of Mexico	Α.		1-2 (feat)	10 (knek) Southwat	i.	z	z	90	0 (Dys)	0 (May)	Colfiy Boxs Coar Coard Cutters -4 ale colf Trapeat	NA	N.Y.	NA	NA	NA	NA	NA	z	z	z	z	Majolity of oil based off, no classing roots in d Walve a robe Vosel Towel for lightering.									1 - CO SQL Cue illustrice Scientisis II - CO SQL Cue illustrice sono constitution in constitut
Garant Odecaresis	0703.1968	Hosbers Mand, Bahaman	Tank Newol	3700 (Bareb)	119000 (Barch)	NA	Nazaratha Crade Oll Dies d Beel	z		Yes	z			11.1 Objes				3423			Boo Margaria Tarker Fear Tarker Pear Tarker	NA	N.	Malyle (ne Sprofted)	NA	NA	Multiple conspectful wight	Sea Const. Serpresent	z		Y (10,900 gallon)	z	Ol spryod with chargest when co want a described and the Travelle from verkey of keewing renders and charge keewing renders and charge republy channel of rend Lightwening and lishable power.				MA	N		0 (Bareb)		Chipie Con it notes its month; Chipie and US and beneated in Pall, NATA, Romono Marsh Report and Annah Report and Annah NATA, Resident Nove NATA, Resident Nove
New of red & m	Date of incident (ill) eme yyyy)	Looke	We need Type	Volume of spith of.	No of ago	Time lag for starting response	Offivelat	Octions (TN)	End dett 5 yes Sadden (107N)	Regions chaptery	O. No.	Ocen codilice Wavepried	Wavebuight	Wed wholey	Steep (CO)	Bain (YN)	Glossly (TYN)	Ocean Teemp (C)	Offdore	Repress clene Orabore	No. response vessels	No. Shirman	Type of shimmer	No. oceanire out to	No. absorbed beam	Type beautied and advantumation or advantumation or an advantumation or advant	No. Sodour pad	Type besided Report e Mahadi terbase pe d	Wom Tado (Y?)	h-sin bening (Y/N	Diponentiff?)	Biceauchice (073)	Ober	Newbor of no presit o	Newbor	offstere strage Capacity	Quesity of generated solid waste	Questity of generated Equil water	ote	Volume of Brees shimm on a second col	free sharbats	De lierce

A Schoolyk, Bigs hland (DOIR) 203 (TS)	04-12-2020	High Island, Bletch Columbs, Greats	New			16-17 barn	Field COI	¥	z	ED: Nationary of Bovious and used Cleaner Change Steame (SE) Energy stry, Management BC Chandles Creat Guard Electroperson and Cleaner Change Cannical ELECTR Cannical ELECTR Cannical ELECTR Workers Cannical Marine Response	0.00000								21 Day)	111 (Days)	* HYATL BAGBI ON WAY * GTV ATL BAGBI ON WAY * CCGS Meeting BHB)	1	- Titon 20 Skimmer - Trien (0 Skimmer	950 fed)	over IS, IDO (Sat)	GRS Bores Greent Perpose Bores Abdust on Bores		Seed Bags					- Sid sym commission betti plang - High Spood Severy System						(40 (Nesers) Are ping of HFO and reades dissel	48311305040		-Government of Hash Colon by Hydroceta, Sorten Y and High Island - Aless Drife - Aless Drife
Queen of the North	21-03-2006	Oli Island Bhitish Chambia Cearda	Ne saf			(Pe()				· WORRC														Maliple	Nobjek (non-profiled)																	-worec
Watercol Jean	04-06-2006	Squarish Bitish Oddes, Onale	Carrier	50,000 (Liken)		StracDay	Black Oil	N	٨	WOARC Constant Cost Goal														Malple	Nakiph (non-specifici)																	**************************************
InterDrive	24.07.2507	ant Highea yo: Berand blat, IIC, Grada	Pipeline (Steam dishins	244,000 (Liken)		Same Day	Syahele Bitmen	N	٨	*WOMRC *Consider Coard * bland Trig and Bargs Co.										06 (Dept)					Muliple (son-specified)															210000 (Surv)		• WOMEC
N.GZdiski	1908-07-08	100 sandest rates seeth of Place B. Bapert, B.C. Canada	Amy Shp			67 (year)	Besler C Rel	Å.	z	* WOARC * Caption Count * Caption Notice * Caption Notice * WOARC Table on Notice * WOARC Table on Notice * Notice would be a cheery and short fees as secured care short fees as secured care	z				z	-	*			90 (cárd)													- Sabage open fore									- WCMBC
M.Y. Mathewa	01-04-20 15	English Bay, Yancover, British Colombia, Centils	Bilt Carter	2,700 (Likero)	81000 (LDAT)	D (bean)	Boslor C Fool	z	>	Consider Cost Gord WANDEC WANDEC WANDEC Per May Second Per May Second On A Wander Dayment of Their or also care Tray port Consider	CONTROL BOOK AND ALL SECONDS								N (pak)	18 (bes)	+3 WOARC Skiers by vone la	Maliple		Muliple	Makiple (nen speallast)				λ.	×	z	z		*Yancener Webster Cops, 4000 promote						1,400(Bero)		MICAN TO ANY AND ANY
Phosper Bay	0.0.2016	Esperal Habor, Baid Oberita Orada	sket Bergs		Dozon - Xt.000 (Lkers)	Sens Day	Diesd Fiel	z	٨	Partners of National Defence WOMEC Canadas Goat Gurd Engines hand Steple of Part Nations Keysteen Environment	z			high winds					t (New)	>4 (recelbs)		Muliple	NA.	Multiple (non-specificit)	Maliple (son speafed)	NA	Malpla, non spacific weight	Serbert Palls	٨	Z	z	z	*Shouline Massal Charap					11,000 (Ress)		1600 (ker)		-Che fibe at qui bron desep ventroses. Tre et clédei
Bella Didla (Nathan E Savvar)	D-10-2016	Bitgs Reef off of Abbase Shed birmon of Northern BC Sestieth Charnel, Comdo	Tagilion	100,000 (Litero)	129.000 (Liken)	18 Board	Diesel Fort	z	٨	Constitution Constitution Constitution Constitution ILLC, Resolven TX, U.S.A. ILLC, Resolven TX, U.S.A. Filed sections and Constitution Constitution Constitution Constitution	Z	, (mtm)	(1) (mstars)	10 (Bech) NE	z		*	=	(Krig)		WCMRC Clowbert Spirit WCMRC Clowbert Spirit WCMRC von dEngle Bay COG Mee P. Toly Start Hollwalve and	7	(D.) Lenot LOBS2 Slavnary (D.S. throad)	хон	Moliph (ten specified)	NA	Seve 26 L dans	Serbent Padis	¥	z	z	z	Selvage and Demphing Operations	*200 cm w					119,000 (L) dood (lebox aels: Salvage operation	10(1)1 (Bero) Olyvatar		**WEXIBLE **Mainstriff, cer *Mainstriff, cer *Mainstriff, cer **Mainstriff, cer **Mainstriff, cer **Mainstriff, cer **Mainstriff, cer **Traped Safey Read Canala **Traped Safey Read Canala
Rahmano R or Peoline		Market, Niskiges, USA	Phyline	3,100m /3			Orale of sife beneric allocat		z													Muliple							, A	×	z	z					Rein, me harsenfour 1/40 (Neuro)	*Debris boundon 0.700 to 13.		2,500 (ser.3)		Newson of Natigered
Havaian Pariet	23403.1977	SSIXes West of Harmin Islands	Tasker	50,000 (besset)	99,000 (beers)	NA	light Indonesian Crube OII	z	٨	NA	z				z		*		(of g	0 (D ₃ yi)	- Merchanten of Follopine Batan (Veset)	0	NA	0	۰	NA	0	NA	z	×	z	z	• Wind and Wine asten • Evipeation	۰	0	NA	0	0	0	0	0	-Crode -TDOP -TDOP -TDO Agent Of Adelests -TDO Agent Obtained -TDO Agent Obtained
TYOQUEY	1611-198	Cleads	Tacker (Vessel	92 099 (Barelo)	132,157 (berse)	l Bost)	Noth Sea Croit Of	z	٨	· Candan Coat Gast	z	> 11	25 (fed)	(ylo)	z	-	٨	10	0	(Bys)	- Bassian Worker Skip, Bysst, - Balk, Carric Nasitran Wisdom	NA	N/A	N/A	N.A	N/A	NA	NA	z	z	z	z	*Narah Dispunion	N.A	N/A	N/A	0	0	0	0	0	The World West Of Disars. - CARE.: 260 - Offer - Off
IXTOC	03-06-1979	Bay of Campe the, Mexico	PW.	-3,32,000 flamb)	N/A	- 8 (days) - USCO 2 (Mandes)	Crafe Oil	¥	z	Parado o Maria anno 195 Mil Xu Ol Mayo ha. NAXA NAXA NAXA NA hard harmadond of Basaco Davaz Const Avistin	T. Company			12 (bross) NW Sqs 20, 1979	z	z	z	25.27	٠	(mostb))	TOTAL STORY	Natyle	Skims of 6) Skims of 6) Open Ware Skims or Shaft On 2008 Skims or Outlet 10 to 2008 Skims or	several forcement (for t)	Makple (nen Speatif all)	- Good Year Bores	Maliple, sen spac Ficweight	N/A		Α.	Å	Α.	- Onling of 2 roles with - Onling or construction of toolow option - Core (927) Energyate - Longer (927) Energyate - Core	Mark th reponse included 30 personal on site				DODGO abbie ya alfa	tensored at well-sale 23,000 (steric tensor)	161,000,086)	N/A	On Suppose of Spikito the Old Company of Spikito of Spi
Dapon at Horizon	25-04-2030	Macordo Propest, (NarceChe contec' Lesbines, Calfeé Moiso	OIR	4,900,000 (Bar ds)	TOOGOO Qulkao deed	Nghot	Mense o Grab GI	Α.	z	- Birkish Persisten - NOA - USCO - ERA - E	, v	<1 (float)	0.6 (84)	6 (Nace) 2 Hi dage on NSW	z	z	z	212	is de distribution de distribu	H55 (D4)0)	- USCGuatas - DSCGuatas - Printeedilecentryly wash	h 20 IX 26 Shirming Ne mada	- News Wer Shares or	4200 (00 B)	0.0000000000000000000000000000000000000	- Tige Borns - Long dark damberboess	• Our 10 (side)	* Sand Borns * GIMops	٨	Y (-05665086)	Α.	Α.	ACC was a second to see the control of the control	47(00 Pva codis(200) - inducing 13 WCMRC saff - 1,100 Excision National Conditrorps			89,000 (henses) * (p. to 20 M	16,260 g cenes)* up to 2004	[TMcm-8Jmn] floreby Persping at set lead	- 1925 99 (harrelo)	- 1,000 (barels)	A CONTRACTOR OF THE CONTRACTOR
Espeiral Hatour		Perper Bay Engineth Barbary, 3 IE, Oznda	Bage	34 000 (Serv)			• Dissel Oil			Worker On thirting Worker On the Marine Represent Coperation On Mannational Informational Socker On paramet of	2000								2 (060)			Malple		Malph			Makple, non spadit weight		z	N	z	z	- Packing of bandwise sake a seas apring reasoff with booses. - Materil Hooking of Waker								27.000 (Bers)	-W-MCC CheWare Candip to Parager Day Yancoore (Pd.Cheng Pari Spillreden, Parks Cotanie Pari Spillreden, Parks Cotanie
World Prodige	23-05-1999	Nevport, Rhotle Island	Task Yeard	873	19230 (Barrido)	II. (Mis)	No 2 Heating Oil	Y	z	US CoatGard - Aleric Schot Torn - Federal On	N N			5-10 (lece)	٨		z	157	1000	4000	T resourchips • Several USCG vessels	Matiple		Matiple	3,000(84)		Mikiph, non specified weight		٨	N	×	N	- You of firstly sourceach ovek by byour of continuous bove - Officialitie - Sigh energy vedesce fished by finitely for my sedesce fished by finitely for my should seem	Noticed Combine National Compinion Instrument from the Nate Add Correlated Institute								O Spi Can Harder Stemman An Agent St. and International Spirit St. Mark Harder Sherich Beyen et al. Non seems II view
CNTB 285	25-13-1998	Median Sharib, Aladia	Begs	47020 (Barek)	47/20 (Bardo)	12 (Deyt)	DesdFul	Α.	z	Chine d Markes Trays and Bangus free. Scart learned Sign	z		2025 (84)	5040 (kmb)	z		٨	5.2	23 (Dey1)	0,000	US Coast Gaald Outer Stripe The Sid to go Chief Remodd y Open to d'Yodaide	Nices		None	Nees		Nese		z	z	z	z	Stating of vond No cleans a ratio is sucknown	-T dives				2691 (Berds)				O SQUEST BANKS Stematics of supports and supports and housement between the supports and housement Delines by
Tanyo Maru	22-07-3991	Nob Bay Wadegee, U.S.A.	Non-Task Yeard	T43 (Bardo)	herrodite Parlos - 1105 (Bardo) Destroi		- Dond OI - Like OI	Å	z	Consider Cost Correl US Cost Cost Correl Canadan (US) of a Narios Politim Cost Cost (Narios Politim Cost (Narios Politim) Narios Cost (Narios Politim) Narios Cost (Narios Politim)				5.0	z	÷	z	101	36 (Na)	38 (Deys)	*2 high *Bernatdy Opported Yochicle	Maliple		Muliple		• Rollaces	Mikiple, son specific Overight		z	Z	z	z	Od trucking via over head flights. Marinal colorier comoved using shootly the host sederates, are farmatheing a rather through on the by the colorier colorier to the property of the true of the salent world profit the true of the salent world profit the true of the salent world profit the true of the salent world.		-	14 (00 (Oddss)	35,000 (plants top)	9000 (plees)	200 (ples)	Th (Bar do)		N Spill or Bridge Sentration of Agel Care Bridge Sentration of Agel Care Bridge Sentration of Sentra
Stel of Corplex	23.04-1988	Carquino Strais, Maximo, Caffornia, U.S.A.	Padity	STO (Bards)		0 (04)	Sur Joseph Valley Heavy Coole OI	Α.	z	41.8 Coast Casall Class Boy Free Nation Boy Free Nation Armsphare free Fin Venneral Protection Agency Fin Venneral Protection Agency LLS Navy Supervice of Salangs	The state of the s			5.54.3 8940					(King)	107 (Dey 1)		Muliple	Max o Gao IS bire ror US Newy Max Gao V Skimmer Supplied by Gao Bay Jos. Week / Pemp Skimmer	Muliple	Muliphi		Mikiple, non specific dronight	penpan	, A	z	z	z	Chaining of finish vator rembles. "Wood Chaining working a chaining social and theorems to the annual and water than the annual business of the annual business							780 Barelo		A STATE OF THE PROPERTY OF T
Name of Ireidens	Date of insided (different 5305)	Location	We used Type	Yelsens of spit of.	No of cargo	Time lay for starting response	Oil Probust	Confinence (YVX)	Introductiyyo Suddon (Y.PV)	Respection Agency	O.No.	Overn conflice Nave prind	Wase beight	Wind videol by	States (Y'00)	Notifier consist on Ruley (TIN)	Georgy (F/70)	Owen Teerp (C)	Of dee	ne press tra e Ondre e	No respressive sels	No. Shimmer	Type of chierae	No. containment boxes	No. als obset been	Type bened of absolve at boxes	No. Seebest pad	Type bend of Represe Method without pall	Noves Tracks (Y-N)	Stook a bereing (Y17)	Okpanses (3170)	Bivern datin (WN)	Open	Natibe of superden	Number	officers skrape.	Quartity of generated sold waste	Quartity of generated kipids to see	Obser	Volume of recovered od Fro es skinner s	Fio rs a hochesta	Defendance

MSCNpdi	1801-2007	is the of Trap selCots of Armor, Fraces, Watern Channel	Cargo Yes sid	*50 (kenso) 630 (kenso)	-1 FO -Mailer Doed OI																									z	z	z	• Number Geration						*352 (seers) IFO *B2 (crees) Mains Deed Of Persping Operation			Condidence Was a Mangamani condidence Was a Mangamani for the Award Mass of for the Mass Award was a Mangamani for the Mass Award Links May a Mark Award Mass and Again for the Mass and Again of White Perchandadis Conty Control
JTBH 6, domon)	13-07-2006	South of Lebrero, Mediumeron Son		10,000 B,000 (ten)			Beenedate FoldOl 190	Å	z	*SASSAR *Traps *Traps *Department of chartic Cost *Department of control Agency Mail tree Sorte and Markine Affine #EMPIC				SE to NE																			· Staff Washing - Staff Washing - Staff Washing - Staff Washing				3600 to 13)	\$46 (m²))	115,649 (trens) ceology me dunical recovery	52,5 II (1000.9)		CVID'S Regional information by the programme of Todaha in more than the former of Staff Vibina more than the former of Staff Vibina more staff of the Staff Vibina and the staff Vibina in the programme of the more staff vibina in the control of the the staff Vibina in the more staff vibina in the more staff vibina in the manual to the more staff vibina in the manual to manual staff vibina in manual staff vibina in manu
Profig	1)-11-2000	Cape Phinkers Gallot, Sprin		(100)		1,003)																															199,100 foreso)	\$4 000 (Neere s)				Comparison and American of White Contrario Design (2014). 2011 International of Sept. ACM, Under East, Victorial Sept. And American of Sept. American Sept. And American of Sept. American Addresses of Sept. American of Sept. Addresses of Sept. American of Sept. Addresses of Sept. American of Sept. American of Sept. American Sept. American of Sept. American of Se
BildicCerrier	3-0-201	Babic Sos Bleve on Gens an and Danish Waxes	Ol Taker	2,700 (tenso)	34 000 (connect)	3 (00%)	Beny Fed OI	z	*	Dasish, Stratish and Green an Androdee Dasis Crid Defere Gerps EC bask freee	z			AL .	z	٨	Å.				B Gerran, Darah and Seedah Yoneb	NA	NA	Maligite, Non-speeding	Militite, Non speedfold				>	z	z	×		* 220 respendars			10'50 (seeses)	905 Ecensis	boom and onth moving machinery 250 feemes)	NA	NA	Corpation influences of the Corpation influences of 2014 decreased Od Spil Geferess T. Wallworth
Edu	D-D-8009	Bay of Bluxy, France		20,000 (ton)						French Malesia Madess French Madess French Androcks Spain Androcks French French French															Nahilipto, Nonspoolikel	Makiple, Non-spession		Mop Nas	*	z	z	z		-> 5000 pan cerdir & valencers			200,000 (cereo)	(100 (bene)		L30 (tens) making		Comprison and Automater Waster Control
Nissos America	2840-1997	Marse also Clared, Verende I	Tarker (Yound	3,600 (ceres)	75,000 (beans)	6 (bean)	Venezatio ende ed	٨	z	· Pordes d'Vescada · Mareco · Lagoro			2 (186.0)	8-10 (nts) NII, IINII							- Shimming 18 and												Tracked concribing may be extend to receve	- 200 presente				40 (00 (heres)				City the to study becomes of White Courself behaviors of White Courself behaviors of Study
Nahhedla	3501-1020	West side, filest best, Japan	Off Weer (Dodde Hill)	6240 (remed)	19000 (neaso)	2 (84.90)	No leaf subserveriti	Å	z	* April Coard Coard * Martins to Danter Provision Castes * Appender Costs of or * Martins Saley Agree * App to Nate Danter Danter Or Gard * Appender Costs of or * App to Nate Danter Danter Or * App to Appende Costs of the Costs of	z		0 - 6 (mater)	20 m i NW	×	Y	Å	156	0KRQ145	B4(Dgy)	Over the vest in • Care Bago • Severi bardeo Palag Bots • ASDWA of the Palas Of Securge	adedorated and second		Maliple (new specified)	NA	N/A	N.A	VN		z	4		ended tog crush iron from on the date is a self the lets. 2 Rannowy System is a first and a self the lets in the antiquest must first and men's wholeg to be remove to a bed to crush on the self and to the lets and to crushes a lets in the lets and to crushes a lets a vealed as the lets and to crushes a lets a vealed and to the lets and the	\$20,000 Per and	+Ol Dem	NA		5(000 (nems)	* 6,600 - 1,500 (como) Lightering * 6,600 (s) crabon raccharia1 recovery	8,700,613		Citypian and Asternat of Spik Asternation of Spik Obstructured of Spik Conformer 1: Wash were Conformer 1: Wash were Market 1: His 1997. The National Market 1: His 1997. The National Asternation of Spik Asternation of Spik Asternation of Spik Asternation of Spik Asternation of Spik Market of Percentage Asternation of Market Market of Percentage Asternation of Market Market of Percentage Asternation of Market Market of Percentage Asternation of Market Historia of Market of Market of Market Historia of Market of Market of Market Historia of Market of Market of Market of Market Historia of Market of Ma
An age one	30011988	of the cost of Frisies, Worken Britary, France	Ollmker	21,000 (mmms))2,500 (oneso)	48 facer 9	eage of benefitie	N	A	age) - aaa Kanggasaj - sapagay yaag -	z		10-12 (recky)	Force 12 Winds	N	N	N	0	(séng) L	(1/0/Q) (E	Fitted New Youch Skimmer Vene by the French Newy Part of Bons	opigny	Stress 20 Airms on Netstack tof an liter oursise	(149 (14) (10)	NA	NA	NA	YON	z	z	z	z	- Manual and	Over 80 Welees *Velation *Hedges			(),000 (onese)	0				Comprison and Assessed of the Confession of Spill Confession of Spill Confession Confession
Teeko	07-03-1980	Broto Gut of Britary lance	Ol Taker	99, 938 (Bereko)	150,800 (Barelo)	48 (barn)	No.6FedOI	A	z	*Preeds Authorities *Received Novy Aury *Polency Plan *TOOP ender the forensioned ON Publics Compensation (DNO) Publics Compensation (DNO) *Compensation (DNO) *Compensati	Z		-26 (Fot)	Rese 11 NW	z	N	z	0	\$12 (Days)	T6(Days)	Abell el argandre Abell el argandre Abell el Bardr Frida	Makple	- Egys de p skiemer	> 16300 (materio)	NA	NA	Yo, No specific Waght	- Grazal Minral Sorbats (Ellopari 33)	٨	и	¥	Å	Das to neters spring weather conditions containers a not condition containers a not for excessive as the design of the condition of the c	X00 persons le mobile of co.200 charsep sites	NA	NA	25000 (treers)	0	-54 975 (Barnels) Lightening -20,000 tons)	NX	NA	Corporation of Automated of Money and Money
Hibrithia	T012.1307	Port of Enchesis en the west casa of State Korea, seeth of Sand	OlTaker	(0000) (000)	2/0,000(series)	4(0s)0	Light Anabous Crosk Oil	N	A	Kerean Nay Kerean Nay Kerean Nay Kerean Name Kerean Name Kerean Name Varion Name Name Name Name Name Name Name Name Name Name Name Name Name	z		(mptor)	M-16 (mis) NW	N	N	N			0.6ep) q2~	More than 100 vessels 15000 private fiking boards	ndersk	NA	riften peprok an tridgen	Maligit, No Specified Wages	NA	Maligite, New Specified Weight	YN	٨	z	Å	A	Liphwing Operators Alarea I Orabors Rocovey	* Nameron Volenty * Nameron Volenty *21,000 tops, public off ash, position and delicated of the second of the se			37,630 (tomos)	2,440 (terms)	21 feer ecb. Diving Operations			*Comprison and bearents of Their Comprison of Objects of Conference II. Wadversh *Conference II. Wadversh *Conference II. Wadversh *Conference II. Wadversh *Conference III. Wadversh *Conference III. Wadvership *Conference III. Wadvership *
Abos I	2611-2006	Dds varellives (folloddyba, USA	Ol Taker	1000 (heates)	U,000 000 (gallow)	1(De/)	Vocazala orde ol	z	Y	Using Stand Creat Goard NODA	۸.				Y	N	z				Need Subnerged Of Bacovey System			30 (les)	NA	NA	N/A	NA	z	×	Y	Α.	- Scherring of OLS gill gracese - Deer Paping Operation				6,100 (areas)	200 (comes)				Comparison and Assessment of Seph. Asset Comparison and Assessment Of Seph. Considerance of Theorem. Asset Comparison of Totals Only Seph. Asset Comparison of Totals Only Seph. Comparis
Sea for press	B-G-1996	Miked Haveel' souge, Usked Kingshen	OITmke	71,000 (bead)			OvdeOil			- Aries Repone Crists - Marie Philates Crared Uni - ON Syll Respress Limited	z													2000 (me kers)					٨	z			-Tenden vorreling puriled to see to cache to dard voice reteree. -Lightering Cyventiese	- X0 Percela sa			(1),000 (100000)	25,000 (tomot)	64 000 (caso) Lightwing	8,500 (com) 4000 (com oil)	NA	Comparisonal Autometric Comparisonal Autometric Page 18, 18 februarian Collegel Conference C. Wedeverle Conference Confere
Buce	5-01-198	Question lay Started Mo.	Ol Taker	*84700 (venes) *800 (venes)	85,000 (lones)	II (born)	Cutting Custs Of Basisar C	٨	z	Marke Potation Control Usi Status Interd Control	z		49	50 - 100 (righ) (farrians foce vints					ş (x6q)			NA	N.A	NA	NA	N/A	N.A.	NA	z	z			- Rock Barrier instead of the use of				2,000 (mesos)	0	afevices	N/A	N/A	Compains and Assessed of Second Distriction of Second 1914 International Sec
Sta Phinoe	23-07-1995	portof Yosa, Stefn Kersa	Of Tasker	5,000 (Team)	85,000 (Tomes)	1 (De/)	Avides Croit Of	N		Poun Martin Pole (Keen) Makes Martin Fole (Keen) Perden Martin Fol or Agnoy Perden Amerikan of Apm Mark Repend Limited	z	91.8	(Red Typhren fon ewints	40 (rs inc)	Z	¥	A		(King) (8	\$ (media)	*Oer 5007 onch	Meliple	- Tra of Shimmer	13,766 (sacars)	NA	NA	239 678 640	ndens		z		z	Lightwing of would On these methat of itsoway, with prestle high pressure pumps	elboure per table 100,001 -			\$100 (benes)	2000 (Nemes)	soye sob Sankand (so)(369'18	(40) (44)		Comparison and American of Data Green and Changes (Chighel, Age 20) Harman and Chigh (Age Age 20) Harman and Chigh (Age Age Children and Ching Children and Ching (Age of Age Age Age Age Age Age Children and Ching (Age Age Age Age Age Age (Age Age Age Age (Age Age (Age Age (Age Age (A
an col Indian	ate of incident (Scienters) ()	parken	osetTye	alarms of spill cel.	actory	ine lig for stating respense	d Probat	Continue (YN)	eideat hype Seadden (Y.N.)	Seattly retained	0.00	oan cardilen Nave paried	Wave height	Wadvelocky	Steay (YYN) Suby coelline	Rainy (YN)	Check (TN)	oun Temp. (C)	Offidore	Ocubero	o respensor sed s	No. Steman	Type of skimmer	No actains at been	No. sheetest born	Type beand of absorbant boom	No. Sahest pad	Typebrank of opense Melsod methan pad	Vacur Teds (Y/N)	finein bondag (T'N)	Dipenset(T/3)	Bicondition (YA)	Other	unka d'ropodan	Number	Shore stongs Capacity	candy of greened sold water	cutally of green hed liquid water	Other	chem of covereded frees skinners	Free absorbants	dienc

CHAPTER 4 RESULTS and DISCUSSION

4.1 Hyperparameter Optimization

In this section, the optimized values for the hyperparameters of Artificial Neural Networks, Support Vector Regression, and Random Forest models are presented. As previously stated, this study aims to develop an estimation model for three types of oily waste: liquid, solid, and total solid and liquid waste. Bayesian optimization has been employed to fine-tune the hyperparameters of the estimation models listed in Table 3.1 to minimize the models' RMSE index. The scripts for all estimation models, including the hyperparameter optimization, were implemented in MATLAB. Figure 4.1 illustrates the results of Bayesian optimization for ANN, SVR, and RF models. Additionally, Tables 4.1, 4.2, and 4.3 provide detailed reports on the outcome of the optimization process and the best-determined values for the hyperparameters of ANN, SVR, and RF waste estimation models, respectively.

Table 4.1 Optimized hyperparameters of ANN waste estimation model

Type of waste	Best Hidden layer size	Best Learning rate
Solid Waste	8	0.04087
Liquid Waste	8	0.03352
Total Waste	9	0.03156
Total Waste	,	0.03130

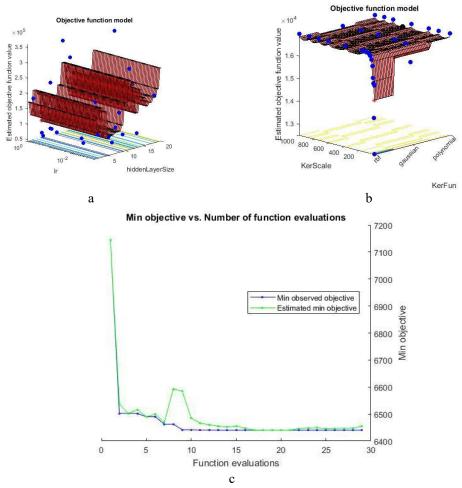


Figure 4.1 Results of Bayesian optimization for AI-based optimization models: a) ANN, b) SVR, and c) RF

Table 4.2 Optimized hyperparameters of SVR waste estimation model

Kernel scale	Best Kernel Function	Type of waste
1.127	RBF	Solid Waste
1.127	RBF	Liquid Waste
1.127	RBF	Total Waste
	KBr	Total Waste

Table 4.3 Optimized hyperparameters of RF waste estimation model

Type of waste	Number of trees	Best max feature no.	Best split no.
Solid Waste	3270	6	5
Liquid Waste	112	3	13
Total Waste	259	1	6

4.2 Waste Estimation Model

After optimizing the hyperparameters, the models were tested with the tuned settings, and the results of the estimation models are summarized in Table 4.4. It is observed that the error values decrease progressively from ANN to SVR and finally to RF. Notably, RF exhibits the lowest reported error indices (RMSE and RMAE) across all types of waste. Furthermore, RF demonstrates a higher correlation value (0.77) between the estimated and observed wastes compared to ANN and SVR, which average at 0.4 and 0.59, respectively. This superiority of RF can be attributed to its increased randomness and minimal risk of overfitting due to the large number of decision trees involved. Additionally, RF provides a more robust output estimation by ensuring low correlation among individual trees, achieved through diversification of the forest by limiting the number of input parameters for each tree. The RF-based model developed for oily waste estimation holds significant applicability to numerous oil spill incidents, as the database can be continuously updated, and the model will automatically adjust accordingly. This model is of immense value to oil spill response practitioners and waste disposal contractors by providing crucial insights for waste management planning.

Table 4.4 Evaluation of the AI-based Waste Estimation Models

Model	Type of Waste	RMSE (tonne)	RMAE (%)	CC
	Solid	9.98	4.86	0.39
ANN	Liquid	9.98	4.88	0.39
	Total	9.98	4.80	0.41
	Solid	3.69	5.32	0.61
SVR	Liquid	3.69	6.09	0.55
	Total	3.69	5.32	0.61
	Solid	1.20	0.61	0.77
RF	Liquid	1.22	0.66	0.77
	Total	1.25	0.61	0.77

4.3 Waste Allocation Framework

The Bella Bella oil spill incident in British Columbia, Canada, serves as a case study to assess the proposed framework outlined in the methodology section. To this end, comprehensive details regarding treatment and receiving facilities, including their capacities and distances from one another, have been compiled and presented in Table 4.5. This data was acquired through numerous meetings with representatives and operation managers of Terrapure, the primary waste contractor on the West Coast. The geographical locations of each facility are indicated on the map depicted in Figure 4.2.

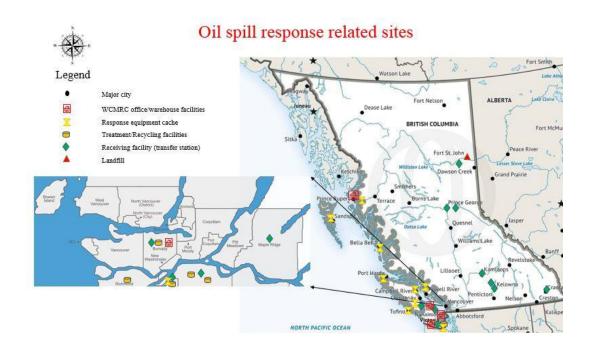


Figure 4.2 The location of waste handling facilities in British Columbia

Table 4.5 Name and location of currently operating facilities in British Columbia

Treatment/Recycle facilities	Receiving facilities (Transfer stations)	
Safety Clean: Delta	Aveitas: Maple Ridge	
Secure Energy: Richmond	GFL: Victoria, Nanaimo, Port Alberni, Surrey,	
Sumas Environment: Burnaby	Cranbrook, Kelowna, Prince George	
Stericycle: Surrey	Safety Clean: Delta	
	Secure Energy: Richmond, Fort St. John	
	Sumas Env: Burnaby, Kamloops	
	Stericycle: Surrey	

As depicted in Figure 4.2, most facilities are situated in the lower mainland area or on Vancouver Island, underscoring the importance of pre-established frameworks for waste

transport in the event of emergency incidents occurring in the central or northern regions of the province. Through research and discussions with the primary waste contractor for the province, it has been determined that the treatment path for generated oily waste varies depending on whether it contains a high concentration of BTEX or bunker oil, as illustrated in Figure 4.3. Liquid high BTEX oily waste is primarily transported to oil refineries in Alberta via pipelines. At the same time, its solid form can undergo processing within the province, being stabilized at a soil processing facility in Princeton before being transported to designated landfills. In cases where the generated solid waste stems from a bunker oil spill incident, it is typically transported directly to landfills without segregation. Conversely, the liquid waste undergoes treatment at specialized facilities, where processes such as dewatering, emulsion breaking, and physical separation are employed to separate oil from water. In many instances, the separated oil can subsequently be sold.

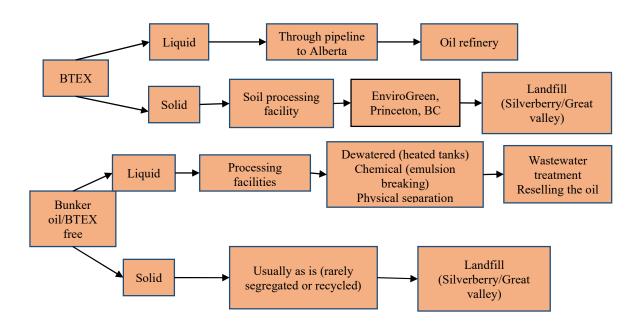


Figure 4.3 The common practice of oily waste management in British Columbia

Considering the heterogeneous nature of oily wastes, it is imperative to consider waste-waste compatibility. Furthermore, the transportation of each waste type necessitates specific types of trucks and treatment facilities. To address these requirements, comprehensive details regarding all operational facilities have been compiled and are presented in Table 4.6.

Table 4.6 The detailed information on oily waste handling facilities in British Columbia

ID	Name	Type	City
S1	Bella Bella	Oil Spill Incident	Bella Bella
T1	GFL		Victoria
T2	GFL		Surrey
Т3	Safety Clean	T	Delta
T4	Secure Energy	Treatment Facility	Richmond
T5	Sumas		Burnaby
	Environmental		
Т6	Stericycle		Surrey
R1	Aveitas		Maple
			Ridge
R2	Safety Clean		Delta
R3	Secure Energy		Richmond
R4	Secure Energy	Secure Energy	
R5	Sumas	Receiving Facility	Burnaby
	Environmental		
R6	Sumas		T Z 4
	Environmental		Kamloops
R7	Stericycle		Surrey

R8	GFL		Victoria
R9	GFL		Nanaimo
R10	GFL		Port
	GrL		Alberney
R11	GFL		Surrey
R12	GFL	GFL Cra	
R13	GFL		Kelowna
R14	GFL		Prince
			George
L1	Silverberry	Landfill	Fort St. John
L2	Great Valley		

As outlined in the methodology section, the optimization of decision variables surpasses the cognitive capacity of human processing, classifying these issues as NP-hard problems that are nearly insurmountable without computational aid. GA emerges as a robust solution for optimizing decision variables involved in waste transfer processes, effectively integrating the capacities of diverse handling facilities into the model without necessitating intricate programming. This accelerates processing time and enhances the model's practicality, which is a critical consideration in emergency scenarios. The real-world application of this methodology is exemplified through the analysis of the Bella Bella oil spill incident in British Columbia, where estimated solid waste volume and associated cleanup costs serve as pivotal parameters. However, detailed information regarding the incident volume is private. Therefore, based on the most recent study published on the estimated solid waste in the Bella Bella oil spill incident, the volume of generated solid waste is estimated to be 13.8×10^3 tonnes. According to the published articles, the cleanup costs of the Bella Bella oil spill incident amount to around \$2.7 million. The cost breakdown is reported as follows:

• Cleanup costs: \$2.7 million

• Fines for non-compliance: \$500,000

• Environmental testing during spill: \$50,000

• Labor expenses: \$500,000

• Office equipment, boats, etc.: \$100,000

• Hidden costs: \$100,000

• Transportation and disposal expenses: \$1.45 million

• Disposal fee (approx.): \$100/tonne of solid waste

• Total transportation cost: Approximately \$1 million

After running the model, the transportation cost was minimized, as shown in Table 4.7. The optimization process took approximately 30 minutes, resulting in a total minimized cost of 662,600 CAD. Figure 4.4 illustrates the trial-and-error process of the algorithm to achieve the lowest transportation cost for waste disposal from the source to the landfill. It is evident from the analysis that utilizing the model would yield savings of 337,400 CAD.

The model's high level of customization is evident in its ability to adjust capacities for different facilities, preventing traffic congestion in smaller plants. Furthermore, the model accommodates diverse treatment rates and relevant costs across facilities, contributing to a more cost-effective waste allocation strategy. The model's adaptability is a key highlight, enabling rapid adjustments in response to emergencies, such as the malfunction of a treatment plant, ensuring an optimized flow of waste transportation.

Table 4.7 The identified path from source to landfill with volume based on the Bella Bella oil spill incident in British Columbia

ID	Path	Volume (tonne)	Description
26	Zs111	7.28	Volume of waste from source 1 to landfill 1
58	Zs1r1	6.4	Volume of the waste from source 1 to receiving facility 1
59	Zr111	6.4	Volume of waste from receiving facility 1 to landfill 1

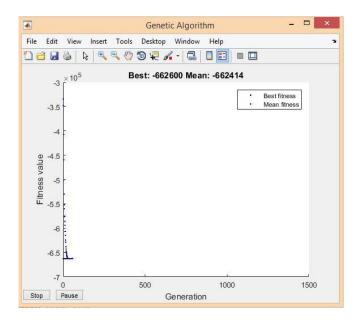


Figure 4.4 Genetic Algorithm minimization cost graph

The model's capacity to consider multiple generation nodes reflects its applicability to real-world scenarios where waste transfer involves intricate networks. The reported daily transfer of 60 tonnes of non-marine oily waste in British Columbia underscores the

significance of such a comprehensive approach. Moving beyond immediate applications, the study offers valuable insights for decision-makers regarding future facility expansions and strategic placements. It serves as an evaluative tool, providing feedback on the current performance of facilities and proposing a holistic optimization of the waste management system. The model's versatility extends to different geographical areas, making it a valuable tool for calculating distances based on the coordinates of new facilities or potential oil spill sources.

CHAPTER 5 CONCLUSION AND RECOMMENDATIONS

5.1 Conclusion

In conclusion, this research represents a significant step forward in the field of Oily waste management, offering an in-depth exploration of estimation models and allocation frameworks. The study systematically examined three prominent models – Artificial Neural Networks (ANN), Support Vector Regression (SVR), and Random Forest (RF) – and found that RF stands out as the most adept in accurately predicting volumes of oily waste. Incorporating Bayesian optimization in hyperparameter tuning enhanced the precision and adaptability of the estimation model, rendering it a potent tool for practical applications in diverse scenarios.

Utilizing the Genetic Algorithm (GA), the waste allocation framework emerged as a robust solution to the intricate challenges of transporting varied types of oily waste to treatment and disposal facilities. Its flexibility enables decision-makers to tailor the framework to specific needs, accounting for facility capacities, waste compatibility, and real-time conditions. This research showcased its practical efficacy by applying the framework to the Bella Bella oil spill incident in British Columbia, highlighting its pivotal role in facilitating informed decision-making during emergencies.

Beyond its theoretical contributions, this research holds practical implications for waste management practitioners, environmental authorities, and response teams. The rigorous evaluation and validation of the proposed models and frameworks establish a solid foundation for enhancing waste management practices in offshore environments. As the offshore industry evolves, integrating artificial intelligence and optimization techniques becomes imperative for developing sustainable and efficient waste management strategies.

The models and frameworks introduced in this study address current challenges and set the stage for future innovations in the field, emphasizing the continual need for advancements in off-shore waste management.

The success of the waste allocation framework lies in its adaptability, providing decision-makers with the ability to customize the model based on facility capacities, waste compatibility, and real-world constraints. Furthermore, applying the Genetic Algorithm enables quick adjustments to the network in response to emergencies, ensuring that the optimized flow of waste transportation can be identified promptly. The model's capacity to handle multiple generation nodes and provide insights into facility performance and potential expansions further strengthens its utility in practical waste management scenarios.

In summary, this research underscores the pivotal role of advanced modelling and optimization techniques in addressing the complexities of off-shore oily waste management. The presented models and frameworks advance our understanding of waste estimation and allocation and offer tangible tools for improving decision-making in the face of environmental emergencies. As the industry progresses, integrating these innovations will be essential in shaping sustainable practices and minimizing the ecological impact of off-shore activities.

The summary of advantages and disadvantages of the incorporated models in this study are presented as follows:

Strengths:

- Precision in Waste Estimation: Bayesian optimization enhances the precision of waste estimation models, ensuring accurate predictions.
- Optimized Waste Allocation: Genetic Algorithm-based waste allocation framework optimizes transportation routes and minimizes the costs.

- Model Adaptability: The waste allocation framework is highly customizable, allowing adjustments based on real-time conditions and facility capacities.
- Practical Efficacy: Application of the framework in the Bella Bella oil spill incident demonstrates its practical efficacy in real-world emergencies.
- Flexibility for Decision-Makers: Decision-makers can tailor the waste allocation framework to specific needs, considering waste compatibility and facility capacities.
- Quick Network Modifications: Genetic Algorithm enables adjustments to the waste transportation network in response to emergencies or facility malfunctions.
- Handling Multiple Generation Nodes: The model can consider multiple generation nodes, addressing the complexity of waste transfers to various facilities.
- **Insights for Facility Expansion:** Provides insights into the performance of existing facilities, aiding in strategic decisions for future expansions.
- Feedback Mechanism: The model acts as a feedback mechanism for the performance of present facilities, allowing for continuous improvement.
- Applicability in Different Areas: The model's ability to calculate distances by entering geographical coordinates makes it applicable in diverse geographical locations.

Weaknesses:

- Limited Public Data: Lack of publicly available data for actual incidents, such as the
 Bella Bella oil spill incident, limits the accuracy of model validation.
- Complexity of NP-Hard Problems: Waste allocation problems are NP-hard, making manual handling impractical due to the extensive number of decision variables.
- **Dependency on Initial Solutions (GA):** The performance of the Genetic Algorithm is highly dependent on the randomness of the initial generated solutions.

- Incomplete Information for Bella Bella Oil Spill Incident: Limited information on the volume of the Bella Bella oil spill incident waste hinders precise modelling and validation.
- Challenges in Modeling Waste-Waste Compatibility: Incorporating waste-waste compatibility in the models is challenging due to the heterogeneous nature of oily wastes.
- Specificity of Truck and Facility Requirements: Shipping each type of waste requires specific trucks and treatment facilities, adding complexity to the waste allocation problem.
- Continuous Model Improvement: The model's performance depends on frequent updates and adjustments based on evolving waste management practices.
- **Resource-Intensive Bayesian Optimization:** Bayesian optimization, while effective, can be resource-intensive, requiring significant computational power.
- Inherent Uncertainties in Oil Spill Incidents: The unpredictable nature of oil spill incidents introduces uncertainties that may affect the accuracy of waste estimation.
- Need for Real-Time Data Integration: Real-time data integration is crucial for the models, and its absence may impact the adaptability and accuracy of waste allocation.

5.2 Recommendations

Given the strengths and weaknesses outlined in the previous section, broadening the study by integrating various approaches or components will contribute to a more thorough evaluation of off-shore oily waste estimation. The suggested viewpoints are summarized below:

- Integration of Real-Time Data: Developing methods for real-time data integration to enhance the adaptability and accuracy of waste management models.
- Enhanced Validation Processes: Implementing comprehensive validation processes, including creating realistic scenarios, to overcome data limitations for incidentspecific models.
- Exploration of Advanced Optimization Algorithms: Investigating the application
 of advanced optimization algorithms beyond Genetic Algorithms for waste allocation
 to improve efficiency.
- Incorporation of Waste-Waste Compatibility Models: Developing models that
 explicitly consider waste-waste compatibility, addressing the challenges posed by the
 heterogeneous nature of oily wastes.
- Dynamic Facility Capacity Updates: Implementing mechanisms for dynamic updates of facility capacities, allowing for real-time adjustments in waste allocation frameworks.
- **Human-Computer Interaction:** Conducting studies on human-computer interaction to enhance the user-friendliness of waste allocation frameworks for decision-makers.
- Integration of Machine Learning for Incident Volume Estimation: Investigating
 the use of machine learning techniques for more accurate estimation of incident
 volumes, considering the complexities of real-world incidents.
- Environmental Impact Assessment: This includes an environmental impact assessment component in future studies to evaluate the sustainability of waste management practices proposed by optimization models.

REFERENCES

- Aguilera F, Méndez J, Pásaro E, Laffon B (2010) Review on the effects of exposure to spilled oils on human health. Journal of Applied Toxicology, 30(4), 291-301. doi: 10.1002/jat.1521
- Alumur S, Kara B (2007) A new model for the hazardous waste location-routing problem.

 Computers & Operations Research, 34(5), 1406-1423.
 - Azevedo A, Oliveira A, Fortunato AB, et al. (2014) A cross-scale numerical modelling system for management support of oil spill accidents. Mar Pollut Bull 80:132–147. doi: 10.1016/j.marpolbul.2014.01.028
 - Barron MG (2012) Ecological Impacts of the Deepwater Horizon Oil Spill: Implications for Immunotoxicity. Toxicol Pathol 40:315–320. doi: 10.1177/0192623311428474
- Beegle-Krause CJ (2005) General NOAA oil modeling environment (GNOME): A new spill trajectory model. In: 2005 International Oil Spill Conference, IOSC 2005. Allen Press, pp 3277–3283
- Bejarano AC, Mearns AJ (2015) Improving environmental assessments by integrating Species Sensitivity Distributions into environmental modelling: Examples with two hypothetical oil spills. Mar Pollut Bull 93:172–182. doi: 10.1016/j.marpolbul.2015.01.022
- Bergstra J (2012) Random search for hyper-parameter optimization. jmlr.org 13:281–305
- Binet S, Leszkowicz A, Brandt H, et al. (2003) Bitumen fumes: Review of work on the potential risk to workers and the present knowledge on its origin. The Science of The Total Environment 300(1-3):37-49. DOI:10.1016/S0048-9697(02)00279-6

- Beyer J, Trannum HC, Bakke T, et al. (2016) Environmental effects of the Deepwater Horizon oil spill: A review. Mar. Pollut. Bull. 110:28–51
- Breiman L (2001) Random forests. Mach Learn 45:5-32. doi: 10.1023/A:1010933404324
- Chen Z, An C, Chen X, et al. (2021) Inexact inventory-theory-based optimization of oily waste management system in shoreline spill response. Sci Total Environ 777:146078. doi: 10.1016/j.scitotenv.2021.146078
- Cutler DR, Edwards TC, Beard KH, et al. (2007) Random forests for classification in ecology. Ecology 88:2783–2792. doi: 10.1890/07-0539.1
- Dave D, Ghaly AE (2011) Remediation technologies for marine oil spills: A critical review and comparative analysis. Am. J. Environ. Sci. 7:424–440
- Diamantidis NA, Karlis D, Giakoumakis EA (2000) Unsupervised stratification of cross-validation for accuracy estimation. Artif Intell 116:1–16. doi: 10.1016/S0004-3702(99)00094-6
- Fatehi M, Asadi HH (2017) Data integration modeling applied to drill hole planning through semi-supervised learning: A case study from the Dalli Cu-Au porphyry deposit in the central Iran. J African Earth Sci 128:147–160. doi: 10.1016/J.JAFREARSCI.2016.09.007
- Fingas M (2016) Oil Spill Science and Technology: Second Edition. Gulf Proffesional Publishing
- Ghanbari F, Kamalan H, Sarraf A (2021) An evolutionary machine learning approach for municipal solid waste generation estimation utilizing socioeconomic components.

 Arab J Geosci 14:1–16. doi: 10.1007/s12517-020-06348-w

- Government of Canada (2022) Marine Spills Contingency Plan https://www.ccg-gcc.gc.ca/publications/environmental-environmentale/marine-pollution-deversements-en-mer/plan-overview-eng.html. Accessed 10 Oct 2022
- Goyal MK, Bharti B, Quilty J, et al. (2014) Modeling of daily pan evaporation in subtropical climates using ANN, LS-SVR, Fuzzy Logic, and ANFIS. Expert Syst Appl 41:5267–5276. doi: 10.1016/j.eswa.2014.02.047
- Guo W (2017) Development of a statistical oil spill model for risk assessment. Environ Pollut 230:945–953. doi: 10.1016/j.envpol.2017.07.051
- Guo WJ, Wang YX (2009) A numerical oil spill model based on a hybrid method. Mar Pollut Bull 58:726–734. doi: 10.1016/j.marpolbul.2008.12.015
- Hoang DT (2008) Metaheuristics for Np-Hard Combinatorial Optimization Problems.

 System
- IMO (2001) Guidelines for the Development of Shipboard Marine Pollution Emergency Plans. International Maritime Organization.
- IPIECA, IOGP (2014) Oily Waste Management for Offshore Petroleum Operations.

 International Association of Oil & Gas Producers.
- ITOPF (2021) Statistics ITOPF. In: Int. Tanker Owners Pollut. Fed. https://www.itopf.org/knowledge-resources/data-statistics/statistics/. Accessed 2 Oct 2022
- Kourou K, Exarchos TP, Exarchos KP, et al. (2015) Machine learning applications in cancer prognosis and prediction. Comput. Struct. Biotechnol. J. 13:8–17

- Lehr W, Wesley D, Simecek-Beatty D, et al. (2000) Algorithm and interface modifications of the NOAA oil spill behavior model. In: Environment Canada Arctic and Marine Oil Spill Program Technical Seminar (AMOP) Proceedings. pp 525–539
- Li P, Chen B, Li Z, Jing L (2014) An agent-based simulation-optimization coupling approach for device allocation and operation control in response to offshore oil spills. In: International Conference on Marine and Freshwater Environments, iMFE 2014
- Lin HY, Hsu PY, Yeh YT (2006) Application of the AHP in data warehouse system selection decisions for SMEs in Taiwan. Int J Manag Enterp Dev 3:599–617. doi: 10.1504/IJMED.2006.010356
- Lottes P, Stachniss C (2017) Semi-supervised online visual crop and weed classification in precision farming exploiting plant arrangement. In: IEEE International Conference on Intelligent Robots and Systems. Institute of Electrical and Electronics Engineers Inc., pp 5155–5161
- Major DN, Wang H (2012) How public health impact is addressed: a retrospective view on three different oil spills. Toxicol Environ Chem. 2012;94:442–467. doi: 10.1080/02772248.2012.654633.
- Marta-Almeida M, Ruiz-Villarreal M, Pereira J, et al. (2013) Efficient tools for marine operational forecast and oil spill tracking. Mar Pollut Bull 71:139–151. doi: 10.1016/j.marpolbul.2013.03.022
- Metcalf S (2014) Developing Location-based Oil Spill Waste Management Plans. Int Oil Spill Conf Proc 2014:1633–1646. doi: 10.7901/2169-3358-2014.1.1633

- Mohammadiun S, Hu G, Alavi Gharahbagh A, et al. (2021) Optimization of integrated fuzzy decision tree and regression models for selection of oil spill response method in the Arctic. Knowledge-Based Syst 213:106676. doi: 10.1016/j.knosys.2020.106676
- Peterson CH, Rice SD, Short JW, et al. (2003) Long-Term Ecosystem Response to the Exxon Valdez Oil Spill. Science (80) 302:2082–2086
- Sadri S, Research DB-WR (2012) Nonparametric methods for drought severity estimation at ungauged sites. Wiley Online Libr 48:12505. doi: 10.1029/2011WR011323
- Sain SR, Vapnik VN (1996) The Nature of Statistical Learning Theory. Technometrics 38:409. doi: 10.2307/1271324
- Sameen MI, Pradhan B, Lee S (2019) Self-Learning Random Forests Model for Mapping Groundwater Yield in Data-Scarce Areas. Nat Resour Res 28:757–775. doi: 10.1007/S11053-018-9416-1/FIGURES/10
- Smola AJ, Schölkopf B (2004) A tutorial on support vector regression. Stat Comput 14:199–222. doi: 10.1023/B:STCO.0000035301.49549.88
- Snoek J, Larochelle H, Adams RP (2012) Practical Bayesian optimization of machine learning algorithms. In: Advances in Neural Information Processing Systems. pp 2951–2959
- Thissen U, Van Brakel R, De Weijer AP, et al. (2003) Using support vector machines for time series prediction. Chemom Intell Lab Syst 69:35–49. doi: 10.1016/S0169-7439(03)00111-4
- Tkalich P (2006) A CFD solution of oil spill problems. In: Environmental Modelling and Software. Pergamon, pp 271–282

- Svetnik V, Liaw A, Tong C, et al. (2003) Random Forest: A Classification and Regression Tool for Compound Classification and QSAR Modeling. J Chem Inf Comput Sci 43:1947–1958. doi: 10.1021/CI034160G
- Wang WC, Chau KW, Cheng CT, Qiu L (2009) A comparison of performance of several artificial intelligence methods for forecasting monthly discharge time series. J Hydrol 374:294–306. doi: 10.1016/J.JHYDROL.2009.06.019
- Xu L, Li J, Brenning A (2014) A comparative study of different classification techniques for marine oil spill identification using RADARSAT-1 imagery. Remote Sens Environ 141:14–23. doi: 10.1016/j.rse.2013.10.012
- Zhang Z, Friedrich K (2003) Artificial neural networks applied to polymer composites: A review. Compos Sci Technol 63:2029–2044. doi: 10.1016/S0266-3538(03)00106-4
- Zhao M, Chow TWS, Tang P, et al. (2017) Route selection for cabling considering cost minimization and earthquake survivability via a semi-supervised probabilistic model. IEEE Trans Ind Informatics 13:502–511. doi: 10.1109/TII.2016.2593664