
**DETERMINING THE AREA OF A WATER BODY USING HIGH-RESOLUTION
SATELLITE IMAGES**

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Abstract

Water bodies play a crucial role in rural areas of India, serving multiple purposes. The country is home to numerous lakes and reservoirs, which are undergoing transformations due to the process of urbanization. Remote sensing and Geographic Information Systems (GIS) techniques are widely employed for the extraction of water bodies and the identification of changes in water bodies. The present study evaluates the process of extracting water bodies from satellite imagery of an urban area in India. Water resources are employed in various sectors such as ecological management, transportation infrastructure, urban planning, disaster management, industrial production, and agricultural activities. Conducting a comprehensive assessment of water bodies and accurately delineating their components is an essential preliminary measure in any planning endeavor, particularly in regions such as India, where the land surface is predominantly characterized by the presence of water bodies. The intermittent capture of satellite imagery does not consistently reflect the distinguishing characteristics of water in relation to non-water features, such as the shadows cast by large structures. The perception of the water body is often erroneously interpreted as the shadow of the high-rise structure due to the tranquil surface of the water, which facilitates the reflection of light and the subsequent generation of reverberation waves. Water transportation is considered to be the most cost-effective mode of transportation. Developing nations such as India stand to gain significant benefits if water transportation is promoted. In the realm of water transportation, it is imperative to establish a connection between various land masses by means of a suitable navigational infrastructure. Moving forward, it is imperative to establish a distinct demarcation between still bodies of water and the areas shaded by various structures. In the context of water transportation, it is necessary to establish a connection between various land masses, including infrastructure, by means of a suitable navigational system. Moving forward, it is imperative to establish a distinct demarcation between still bodies of water and the obscured areas created by architectural elements.

Keyword*Feature extraction, remote sensing, and water body, satellite images.***1. INTRODUCTION**

In recent years, remote sensing has emerged as a prominent form of distant sensing. The field of object detection utilizing remote sensing imagery has emerged as a prominent area of study,

garnering significant interest from both the scientific and industrial sectors. The effective management of water resources, specifically in relation to flood disaster mitigation, necessitates the accurate extraction of water body information from remotely sensed imagery. In recent times, a number of novel methodologies have emerged for the extraction of water from remote sensing images obtained from diverse sources. Object-oriented programming (OOP) is a contemporary technological paradigm that has surfaced within the past few years. The identification of water holds significant importance in various precise calculations and is essential for human survival. In recent decades, a plethora of image processing algorithms have been developed with the aim of detecting and extracting surface water areas from satellite data. There exist both single-band and multi-band methods. Remote sensing techniques have frequently been utilized in LandSat imaging to effectively detect and extract surface water areas. Additionally, a threshold value, which can be either positive or negative, has been employed in this process. The proliferation of machine learning methodologies for information representation, alongside conventional examination strategies, has been driven by the availability of extensive electronic databases and high-speed computing processors for data storage. The process of analyzing extensive pre-existing databases in order to extract novel information is commonly referred to as data mining. The pursuit of concealed connections and worldwide patterns within extensive datasets is underway [1]. "Data mining is the use of statistics to uncover patterns and trends in very big databases through exploratory data analysis and predictive models." (From the user documentation for Insightful Miner 3.0) Its definition shifts depending on who is using it "Data mining is the process of sifting through huge volumes of data kept in repositories, employing pattern recognition technology as well as statistical and mathematical tools to identify important new connections, patterns, and trends." (Source: Gartner Group) (3). Data mining is a technique employed to extract valuable insights from extensive datasets stored within a database [2]. Data mining plays a crucial role in the identification of patterns, the grouping of data for the purpose of data association, and the revelation of previously undisclosed information within extensive databases. Data mining technologies, such as statistical models, mathematical algorithms, and machine learning approaches, have the capability to identify predictive information that surpasses the boundaries of human expertise. Data mining, alternatively referred to as Knowledge Discovery in Databases (KDD), is a methodology employed to extract pertinent information from extensive databases.[3].

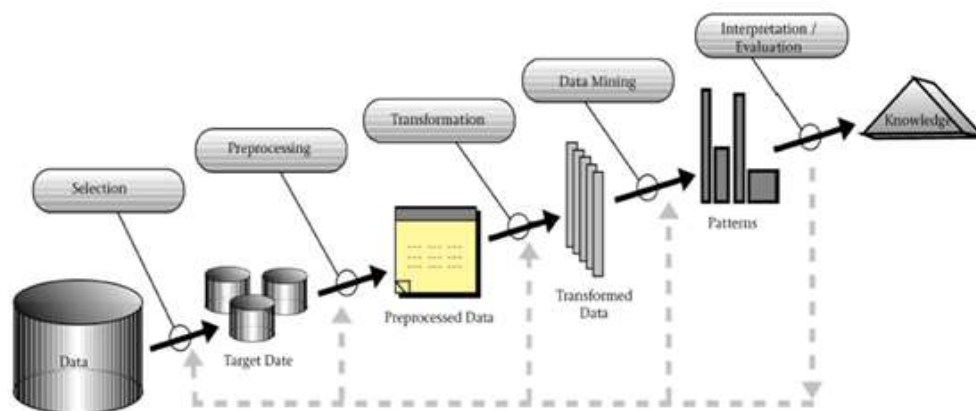


Figure 1: Stages of Data Mining Process [3]

1.1 Preface to the Field of Image Mining:

Image mining poses a formidable challenge for researchers in the field of data mining. Image mining refers to the process of extracting distinctive characteristics or attributes from a vast assemblage of images [4]. The images within the database undergo preprocessing procedures in order to enhance the overall quality of the images. Images are employed to assess prominent characteristics and user-targeted essential parameters through a variety of transformations, selections, and extractions. By leveraging these characteristics, the mining process can employ various exploration techniques to identify and uncover meaningful patterns. The aforementioned patterns are assessed and analyzed in order to ascertain knowledge. In contemporary times, the proliferation of digital satellite image capturing devices has led to the incorporation of geological and land topography information databases. These databases now encompass not only the structural details of land surfaces but also unstructured geophysical image information. Although geophysical satellite images provide a comprehensive depiction of Earth's resources, the process of extracting valuable information from these images may require additional time. The efficacy of management in geophysical engineering is contingent upon the quality and comprehensiveness of the geological database. Geological information holds significant importance within the domain of geology. The researchers must intensify their efforts in order to determine the precise feature selection of the geological satellite images.

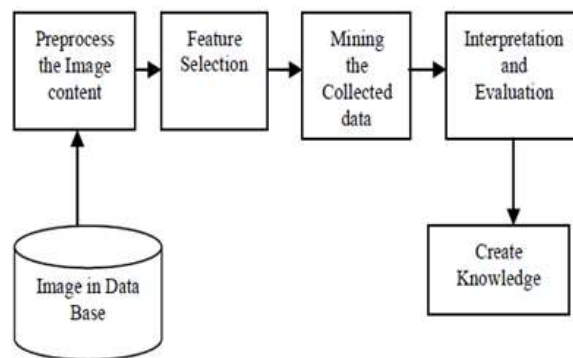


Figure 2: Image Mining Process [4]

1.2 An Overview on the Investigation of Water Supply Bodies:

Water bodies located on the Earth's outer layer are an essential component of the hydrological cycle. Surface water, groundwater, island water, streams, lakes, ephemeral waters, coastal waters, and springs are vital water resources (5). This study examines the retrieval of information pertaining to lakes, as water resources have experienced contamination and depletion in recent years. Water bodies play a significant role in causing various catastrophic events in different scenarios. Consequently, the extraction of waterbody data has become increasingly important. Understanding the fluctuations in regional water resources is imperative. Lakes are bodies of water that are located inland and characterized by their stationary nature. Lakes play a crucial role in the hydrological cycle and are therefore essential for monitoring water resources. The presence of inland water bodies is a crucial factor in determining the ability of humans to survive and progress socially [6]. The importance of people, food harvests, and biological systems is fundamental [7]. Accurate data regarding the spatial distribution of open surface water is crucial for assessing current and future water resources, environmental models, agricultural crop significance, river flow, wetland extents, watersheds, surface water surveys and management, flood planning, and weather monitoring [8].

1.3 Overview of Remote Sensing:

The rapid development of satellite imaging technology, particularly in terms of spatial and spectral resolution, has led to a remarkable expansion of extensive collections of high-resolution satellite images. The data obtained through the utilization of satellites, radars, and sensors contains crucial geographical information that is employed in remote sensing applications, including regional planning and disaster management. Spatial data classification and object recognition are typically considered to be the primary tasks associated with numerous applications. Satellite imagery assumes a prominent role as the primary resource for a multitude of remote-sensing applications. Remote sensing technology has made significant advancements in its ability to accurately identify and analyze the spatial distribution of land cover, water bodies, and vegetation. Each remote sensing image contains valuable information and various objects

within it. However, manually extracting useful patterns from these images is a complex and challenging task. The task of identifying and categorizing objects within images poses a significant challenge.

1.4 Brief Overview of Image Processing:

The process of feature extraction and classification is carried out utilizing image processing techniques, while patterns are identified through data mining methodologies [9]. This approach aims to acquire pertinent knowledge for the purpose of recognizing objects in satellite imagery. Every image contains significant information and objects, making the manual extraction of useful patterns from them a laborious task. Image segmentation is the prevailing technique employed for the generation of such objects [10]. Image segmentation is the process of partitioning neighboring pixels in an image into distinct regions based on their similarity, following a predefined set of criteria that considers shared characteristics. The segmentation process yields various attributes for each extracted object, including spectral parameters, texture, morphological characteristics, and contextual features. These attributes are subsequently employed for image analysis. Moreover, the objects present in an image exhibiting distinct patterns are assigned to specific classes through the process of comparing various identified objects. The process of object classification involves the utilization of object-oriented classification techniques, which take into account the uniformity of object features. Data mining techniques are utilized to perform automated object recognition and image classification. In recent years, the utilization of remote sensing satellites with diverse spatial, spectral, and temporal resolutions has become essential for the detection and extraction of inland water bodies and their dynamic changes. These satellites generate a substantial volume of data, which serve as crucial sources for this purpose. In recent times, a number of image processing algorithms have been put forth with the aim of extracting water features from satellite data. Data pertaining to aquatic bodies is collected. The utilization of remote sensing and data mining techniques has become increasingly important in various applications such as water surveying, wetland protection, disaster prevention, and water resource monitoring. Various techniques, such as the threshold method, exponential method, decision tree, and other spectral methods, are employed to gather data on water bodies. In the process of gathering data on water bodies, researchers employ both single-band and multi-band techniques to measure the reflectance of the surface. This measurement is valuable for extracting crucial information, such as determining ratios and spectral relationships, as exemplified by the Normalized Difference Water Index [7].

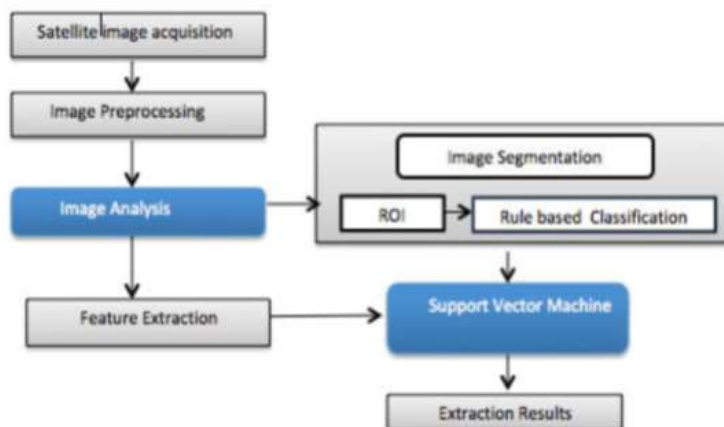


Figure 3: Flow Chart to illustrate proposed Image analysis and Data mining approach to classify water bodies

2. Methodology:

The integration of various algorithms for satellite image processing and data mining has been proposed as a means to enhance effectiveness and efficiency. However, the challenge of effectively and efficiently integrating these algorithms remains unresolved. The majority of current image processing algorithms demonstrate a 90% accuracy rate in extracting water bodies and detecting changes in lakeshore contours. To achieve a higher level of accuracy, it is necessary to employ additional algorithms in image processing and utilize data mining queries. There is a limited selection of literature pertaining to the integration of water body extraction and data mining techniques. The literature lacks a comparative analysis of various image processing algorithms integrated with database queries in the context of data mining, specifically focusing on their respective performance.

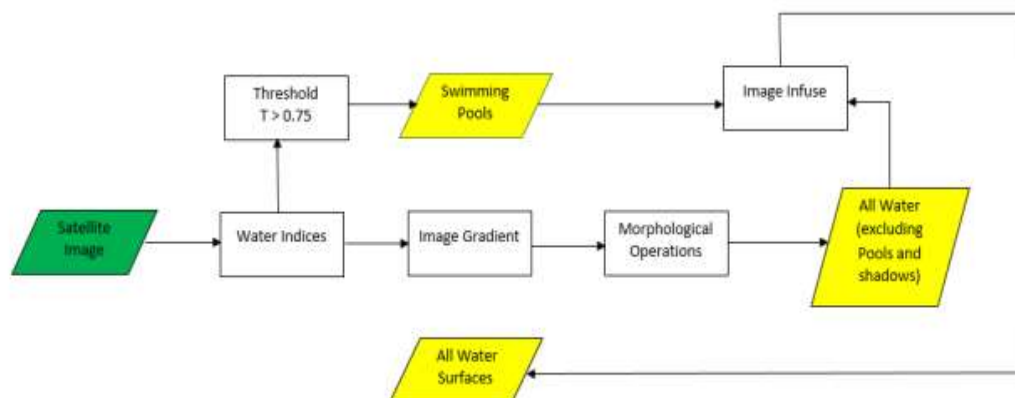


Figure 4 : Water Bodies Feature Extraction Methodology

Figure 4: Methodology for Extracting Features of Water Bodies Based on the aforementioned flowchart, it is possible to derive water capabilities from the Satellite Image. The application of

specialized Water Indices, the calculation of the Gradient of the Image, and the subsequent utilization of Morphological Operations. The subsequent sections of this paper will explicate each of the aforementioned steps and elucidate the significance of each step in the process of extracting the capabilities of water bodies.

2.1. MACHINE LEARNING AND OBJECT DETECTION:

Object Detection is a technique that uses computer imagination and image processing to identify the many objects seen in an image. If we delve any further into this area, we may also discuss weather, emotion detection, and facial recognition. It's used to identify things like traffic signals, roads, pedestrians, zebra crossings, animals, and more inside a shot. Everything is distinct and accurate in its own way. Picture processing can also employ an object's geometric properties to zero in on it: a square will be perpendicular, a face will have two eyes, noses, lips, etc., and so on.

2.2. TECHNOLOGY USED PYTHON:

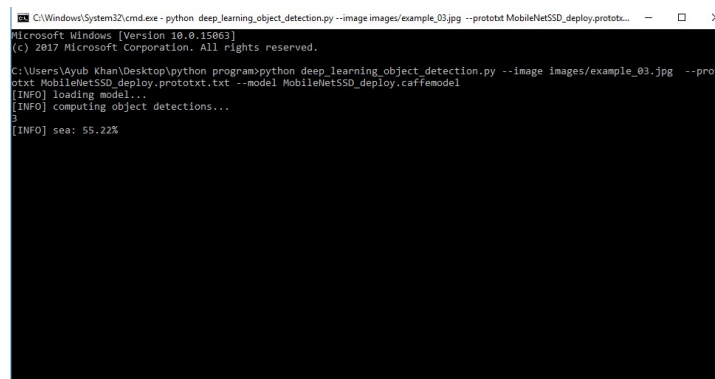
Python is an interpreted, interactive, and object-oriented high-level programming language. Python is well developed such that it is easy to read. In addition to using English key phrases frequently, it also has fewer syntactical formations than extraordinary languages that use punctuation. Python code is "interpreted" when it is entered at runtime with the help of the interpreter. You should not gather your code before running it. Similar to PHP and Perl. You can sit down in front of a Python installation and type your code directly into the interpreter, making it highly interactive. Python supports the object-oriented programming style, which hides operations in objects for safer analysis.

3. IMPLEMENTATIONS

1. `deep_learning_object_detection.py`
2. `# USAGE`
3. `# python deep_learning_object_detection.py --image images/example_01.jpg \`
4. `##--prototxt MobileNetSSD_deploy.prototxt.txt --model MobileNetSSD_deploy.Caffemodel`
5. `# import the necessary packages`
6. `import numpy as np`
7. `import argparse`
8. `import cv2`
9. `# construct the argument parse and parse the arguments`
10. `ap = argparse.ArgumentParser()`
11. `ap.add_argument("-i", "--image", required=True,`
12. `help="path to input image")`
13. `ap.add_argument("-p", "--prototxt", required=True,`
14. `help="path to Caffe 'deploy' prototxt file"`
15. `if not check_integrity(os.path.join(root, "EuroSAT.zip")):`
16. `download_and_extract_archive(URL, root, md5=MD5)`

17. # Apparently torchvision doesn't have any loader for this so I made one
18. # Advantage compared to without loader: get "for free" transforms, DataLoader
19. # (workers), etc
20. def __init__(self, paths: [str], loader=default_loader, transform=None):
21. self.paths = paths
22. self.loader = loader
23. self.transform = transform
24. def __len__(self):
25. return len(self.paths)
26. def __getitem__(self, idx):
27. image = self.loader(self.paths[idx])
28. if self.transform is not None:
29. image = self.transform(image)
- 30.
31. # WARNING -1 indicates no target, it's useful to keep the same interface as torch vision
32. return image, -1

3.1. SCREENSHOTS:



```
C:\Windows\System32\cmd.exe - python deep_learning_object_detection.py --image images/example_03.jpg --prototxt MobileNetSSD_deploy.prototxt...
Microsoft Windows [Version 10.0.15063]
(c) 2017 Microsoft Corporation. All rights reserved.

C:\Users\Ayub Khan\Desktop\python program>python deep_learning_object_detection.py --image images/example_03.jpg --prototxt MobileNetSSD_deploy.prototxt --model MobileNetSSD_deploy_caffeemodel
[INFO] loading model...
[INFO] computing object detections...
[INFO] sea: 55.22%
```

Figure 5 : Water Bodies Program Running

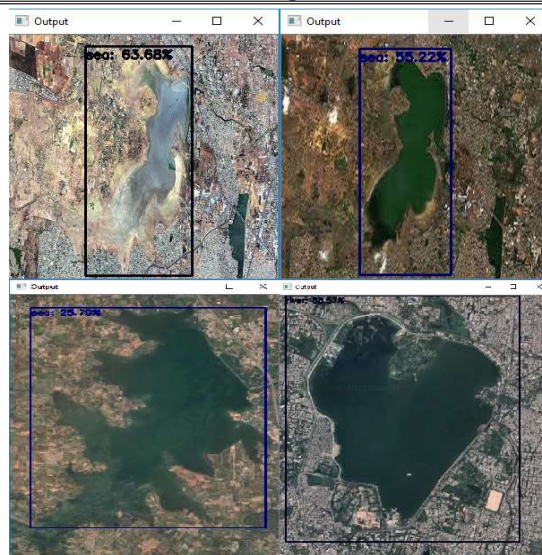


Figure 6 : Water Bodies Program Output

4. CONCLUSIONS

The introductory section of this paper elucidates the significance of water body statistics, outlines the objectives of water characteristic extraction, and highlights the inherent complexities involved in water frame segmentation. Furthermore, this paper elucidates the various types of satellites and methodologies employed by researchers in acquiring satellite imagery for the purpose of water feature extraction. Limited discussion has taken place regarding the outcomes. A subsequent attempt was undertaken to reach a conclusion regarding the present challenges and future prospects of water resource extraction methods. The utilization of satellite data in water body mining is contingent upon the land use and land cover characteristics of the studied landscape. The present investigation employed the landscape structure of the urban setting in the rapidly expanding city of Hyderabad. It is necessary to evaluate the efficacy of the proposed methodology in diverse topographical configurations, such as bodies of water within heavily wooded terrains or rural regions. The scenario entails the presence of water bodies interspersed within agricultural lands featuring diverse crops, or alternatively, water bodies situated in close proximity to expansive rivers and streams. The significance of landscape lies in the variation of land use classes surrounding water bodies across different landscapes. Water bodies are delineated by differentiating between water and non-water features, which act as boundary classes for the water bodies. However, identifying these classes can be challenging due to the potential overlap of spectral characteristics among different features. Considering the aforementioned points, forthcoming investigations in the domain of perceptron models hold substantial potential for constructing and advancing a distinct approach capable of effectively operating on diverse satellite data sets. This method has the ability to extract water bodies with enhanced precision, encompassing smaller water bodies as well. The primary objective of this study was to develop a methodology that improves the accuracy of water extraction by increasing spectral distinctiveness between water and non-water surfaces, particularly in areas

characterized by shadows and urban infrastructure. The methodology employed in this study was implemented on various datasets acquired from satellite sources. This report presents various methods for delineating water bodies from satellite imagery, considering different techniques that have proven to be robust in terms of both subjective and quantitative (visual and numerical) outcomes. The proposed methods can be considered beneficial for both the examination of Picture Analysis, specifically in relation to highlight extraction, as well as for research involving data extraction in Remote Sensing images. These methods utilize the automatically or semi-automatically extracted features from digital images to assign them semantic attributes. The significance of this study within the field of Image Analysis lies in its direct relevance to the application of Digital Image Processing techniques for the purpose of water body extraction. The present procedure can be classified as a methodical extraction due to its utilization of specific attributes of the focal point of interest as guidelines for defining the sequence of extraction techniques, which in this instance pertain to the water bodies such as rivers, lakes, and oceans in satellite imagery. Furthermore, there is a necessity for enhancing the algorithm utilized in the extraction of water bodies. This enhancement should involve automating the system to handle various types of sensor images. Additionally, it should be integrated with other tools to ensure accurate data is provided regarding flood conditions and the availability of underground water. The aforementioned perspectives constitute a fundamental concern in developing nations. Collecting data manually can often be a tedious task.

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