



A Study on Water Security for Bengaluru-A Predictive Analytics using Sentinel data

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ABSTRACT

Water is the lifeline of living beings. It is one of the most essential fundamental element to sustain the life. Bengaluru-“The City of Gardens”, “The city of Thousand Lakes” or “Necklace of Lakes” has now become like a once upon a time story because of heavy exploitation of the resources and uprooting of the trees in the name of Urbanization. Bengaluru was said to have around 1000 lakes earlier and now because of real estate revolution many lakes are dried up and sky scraping buildings are seen everywhere. It is said that now around 100 odd lakes only are there and no lake is Potable. Almost all the lakes in Bengaluru fall in D or E category. We need to safeguard our lakes, recharge, maintain our wells, and enhance the utilization of our groundwater aquifers effectively [1 May 7 2024, Avinash Murthy]. It is important to study and understand the declination of the areas of existing potable water lakes. Sentinel data, which is supplied by the Copernicus Program of the European Space Agency, is essential for tracking changes in urban areas, aquatic bodies, and land usage. Data from Sentinel-1 (SAR) and Sentinel-2 (optical) are very pertinent. In order to analyse satellite data and develop prediction models, geospatial approaches are essential. This involves the use of remote sensing tools and Geographic Information Systems (GIS). Using the sentinel images [2] of 4 lakes of Bengaluru North, the diameter of the water body is obtained from previous records and the primary data of 2025 is gathered in our work. Further using the Image processing technique, we would attempt to do Predictive Analysis using Artificial Intelligence to build an algorithm that will

give suggestions for water security in Bengaluru in the coming days. This research work is dedicated to sensitize the importance of Water bodies to the mankind in general and Bengaluru City in particular, which helps in conservation and recharging of the water bodies and resilience to the underlying aquifers.

KEYWORDS: Water Security, Aquifers, Sentinel data, Image Processing, Artificial Intelligence, Algorithm, Predictive Analytics, Remote Sensing, Urbanization.

Introduction:

Society benefits from urbanization because it makes it easier for people to live more ordered lives. However, if not properly planned and managed, it can also be detrimental [1]. (Rimba et al. 2021). Mohan et al. (2011). The population of Bengaluru has increased historically in just 25 years due to the city being technological hub. The garden city is now popularly called as Silicon city. Majority of influx is from rural side of the state and from other state seeking for Jobs. Urbanization happening quickly [2] has drastically changed the region's natural resources, impacting both the ecosystem and society (Sun et al. 2013, 2016; Bharath et al. 2021). (Bhat et al. 2017; Mundia and Aniya 2006; Berling-Wolff and Wu 2004). The rapid growth of cities and industries has raised interest in learning more about the environmental effects of these developments. Data on water levels must be gathered in order to manage water bodies effectively. It is a key component that reflects the changes in lake's water status. The European Space Agency (ESA)'s Sentinel-2 mission, which is a component of the Copernicus Program, offers multispectral, high spatial resolution pictures of the Earth's





surface. Sentinel-2 finds its applications in ; Assessing the quality of water, keeping an eye on factors including turbidity, collared dissolved organic matter, Algal bloom detection is the process of locating and keeping an eye on dangerous algal blooms. Monitoring shifts in the size of lakes, rivers, and reservoirs is known as "water body mapping." Mapping flooded areas both during and after floods is known as flood monitoring. In our study we have used sentinel-2 images for its high resolution and accuracy in providing the accurate coordinates of the lake for the study of our topic.

The idea behind applying deep learning neural networks to water level analysis is comparable to the conventional approach of taking pictures of water gauges [17–19] or water level lines [20–26] in order to measure water levels. Nonetheless, the two methods differ greatly from one another. The location of the water gauge and the water level line must be manually set up in order to perform classic image analysis of water levels. When a result, when additional conditions are taken into account, the program structure grows larger and more intricate.

By creating a system for determining the location of the water gauge and the water level line during training, deep learning networks for water level analysis overcome this difficulty. As a result, the bar for starting a program is lowered. Finding the water level line is the first step in performing semantic segmentation for water level analysis. The location of the water level is then represented by the dividing line between the two parts of the image, which are the water body area and the non-water body area. [20]

Predictive analysis for water security of lakes in Bengaluru involves using historical and real-time data, statistical modeling, and machine learning techniques to forecast future conditions and identify potential risks to the lakes' water quantity and quality. The methodology involves Data acquisition and preprocessing, the historical data can be obtained from **Government Agencies such as, Karnataka Lake Board Authority**, Bangalore Water Supply and Sewerage Board (BWSSB). Sentinel-2 data from remote sensing satellites, in-situ sensors and monitoring stations. Further the predictive modeling techniques and machine learning models and AI tools are used to develop an algorithm, validation of the algorithm is done by providing different data sets. Once the algorithm meets the required standards, the trained and validated models with the latest available data to forecast future water quantity and quality indicators over different time horizons is used (e.g., short-term: days/weeks, medium-term: months, long-term: seasons/years). The summary reports are generated for

predictions; the predictive tools are integrated into a user friendly platform for regular forecasting.

Literature Survey

Water level measurement technologies can be categorized as contact and non-contact sensors. Contact sensors such as floating sensors and pressure sensors are low-cost and easy to use. However, they are prone to inaccurately measured data and instrument damage due to bed sediment interference [3]. Radar and ultrasonic instruments must be positioned directly above the river surface, typically on bridges. However, they are vulnerable to bridge vibration, and ultrasonic instruments are also susceptible to wind speed interference during typhoon events. Alternatively, imaging instruments offer a different solution, and there are two ways to measure water level using this technology. The first method involves utilizing satellite images to calculate the coordinates of water body boundaries [11–13]. While this approach is not always feasible, it is useful in defining the boundary between the river and land areas [14] and examining long-term changes in the river area [15, 16]. The conventional approach of taking pictures of water gauges [17–19] or water level lines [20–26] is comparable to using deep learning neural networks for water level analysis. The location of the water gauge and the water level line must be manually set up in order to do a significant traditional image analysis of water levels. Consequently, when more conditions are taken into account, the program structure grows larger and more intricate. By creating a system for determining the location of the water gauge and the water level line during training, deep learning networks for water level analysis overcome this difficulty. As a result, the bar for starting a program is lowered [3].

Bengaluru has seen a population increase from 394,794 persons in 1901 to 9,621,551 in 2011 as per the Census of India. With increase in population, urbanization also increases. Urbanization, which is an output of population increase in cities, which overloads the cities natural resources includes land and water through activities such as building of more residential areas, industrialization and commercialization. This urbanization impacts not only the drainage network, which is often broken due to encroachment, but also the quality of run off and thus the quality of water stored in the lakes. Atmosphere of urban areas adds to the pollution load in the lakes too (Lazaro, 1979). Thus, quality of lake water has to be periodically monitored and communicated to authorities and public as they are the main stakeholders in preserving their environment and the benefits also go to them [6]. Sentinel-2, in effective turbidity management for dynamic urban coastal





environments such as Los Angeles. Remote sensing allows for improved spatial-temporal monitoring that could capture seasonal and storm-related variations in turbidity, which might be difficult for traditional monitoring methods to catch. Incorporating machine learning models, such as random forests, with variables like precipitation, tides, and human activity may further enhance turbidity prediction and support proactive management strategies to protect coastal ecosystems and public health. Satellite data can also help storm water management by identifying turbidity hotspots from a broader spatial aspect to guide interventions like sediment control and public health advisories [9].

Predictive modelling techniques, particularly machine learning algorithms, are increasingly being used in conjunction with Sentinel-2 data to forecast various water-related parameters. Studies have demonstrated the potential of using machine learning models like Artificial Neural Networks (ANN), Support Vector Machines (SVM), Random Forest, XG Boost, and Partial Least Squares Regression (PLSR) to predict water quality parameters such as turbidity, chlorophyll-a, and other indicators based on Sentinel-2 spectral data. These models often show good performance, with reported R^2 values and Root Mean Square Errors (RMSE) indicating a strong relationship between Sentinel-2 data and water quality measurements. The choice of the specific machine learning model often depends on the water quality parameter being predicted and the specific characteristics of the water body under study. Beyond water quality, Sentinel-2 data is also being used for forecasting surface water extent and hydrological events. While the optical nature of Sentinel-2 limits its ability to directly monitor floods in real-time due to cloud cover, it proves valuable for post-flood assessment and for integration with hydrological models and climate data to improve forecasting of water availability and drought conditions. Studies have also focused on predicting water scarcity and water stress conditions using Sentinel-2 data. This often involves using water indices derived from Sentinel-2, such as NDWI, or monitoring changes in surface water extent and vegetation health, particularly in agricultural contexts for assessing crop water stress [13].

Objectives:

- Data Acquisition from Google Search Engine of Sentinel-2
- Data Preprocessing using appropriate software [13] (Sandeep Gupta et al.) Data preprocessing involves three steps; Data cleaning, Data Transformation and Data Reduction using various software's viz. Python, R,

Jupyter Notebook

- Data Mapping

Data Mapping is an integral part of Data Management, Data Mapping if not done appropriately can cause the data corruption or data loss by the time it reaches the destination. It is the important tool to get data out of Data cleaning, Data transformation and Data reduction

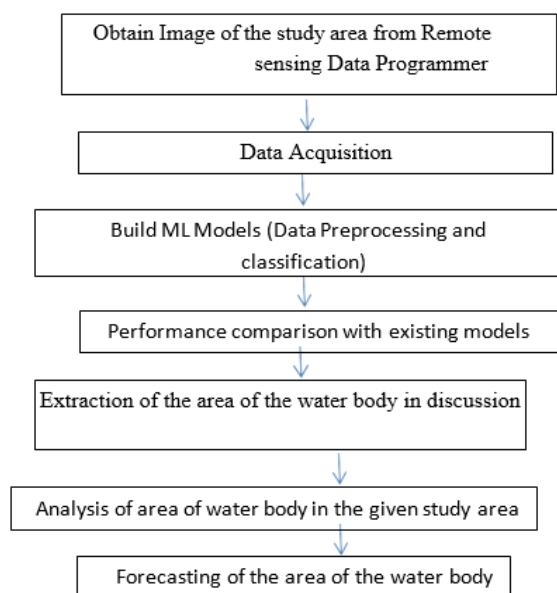
- Data Validation

Proposing the best band combination of sentinel image and efficient deep neural network architecture. Results were evaluated with various sentinel band combinations namely, Near Infrared, Red, Green and Blue. Deep neural network and Deep Convolutional Neural Network architectures were customized and hyper parameters were fine tuned to improve the classification results.

Methodology

The study is supervised using various data sets to analyze the cultivated crops using the temporal data in the area. To effectively manage this region's growth, it is imperative to comprehend the dynamic phenomena of urban sprawl. Jupyter Notebook and MATLAB 18 are the two software tools used in this investigation.

1.1 Proposed flowchart



Discussion

To effectively design methods to measure the water body, it is imperative to obtain a full understanding of the fluctuations in the area at regular intervals and seasons. The





complexity of the procedure makes classifying and mapping the land cover pattern extremely difficult. The same operation may now be carried out on satellite photos thanks to the growing variety of image processing techniques and algorithms. After downloading the 2001 and 2021 satellite photos from the USGS, MATLAB 18 software is used to classify the images using a supervised image classification method.

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This paper is guided by my coauthor and guide Dr. Veena C S who also has contributed key points and pointers in writing this paper. Around more than 30 papers are referred and understood before writing the paper.

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Conclusion:

After the completion of this work the changes in the area of The water body is tabulated and changes in the water body in The future will be understood and forecasted.

References:

- <https://citizenmatters.in/bengaluru-water-security-indices-data-challenges/>
- *Advances in machine learning and IoT for water quality monitoring: A comprehensive review*
- *Ismail Essamlali, Hasna Nhaila, Mohamed El Khaili*
- *Evaluation of deep learning computer vision for water level measurements in rivers by Wen-Cheng Liu * , Wei-Che Huang, Department of Civil and Disaster Prevention Engineering, National United University, Miaoli, 360302, Taiwan*
- *WATER FEATURE DETECTION USING SENTINEL-2A DATA IN LAKE YLIKI*
- *Terpsichori Mitsi, Nicholas Stoupas*

- *WATER QUALITY STUDIES IN URBAN LAKES Mahantesh Kodli1 , Student, Dept of Geoinformatics, Karnataka State Rural Development and Panchayat Raj University Gadag. Ravi Jadi2 , Faculty, Dept of Geoinformatics, Karnataka State Rural Development and Panchayat Raj University Gadag. Dr. S. Rama Subramoniam2 , Scientist/Engineer 'SF', RRSC-South, Dept of Space, ISRO Bengaluru - 560 037. Mr.Ramchandra Hebbar3 , Deputy General Manager, Scientist/Engineer 'G', RRSC-South, Dept of Space, ISRO Bengaluru - 560 037.*
- *Impact of urbanisation on lakes—a study of Bengaluru lakes through water quality index (WQI) and overall index of pollution (OIP) Khushbu K. Birawat · Hymavathi T · Mathuvanthi C.Nachiyar · Mayaja N.A · Srinivasa C.V Received: 15 January 2021 / Accepted: 17 May 2021*
- *Satellite remote sensing to improve source water quality monitoring: A water utility's perspective*
- *John Lioumbas, Aikaterini Christodoulou, Matina Katsiapi, Nikolett Xanthopoulou, Panagiota Stournara, Thomas Spahos, Georgia Seretou, Alexandros Mentis, Nopi Theodoridou*
- *Sentinel-1 SAR Images and Deep Learning for Water Body Mapping*
- *Fernando Pech-May,Raúl Aquino-Santos* and Jorge Delgadillo-PartidaDepartment of Computer Science, TecNM: Instituto Tecnológico Superior de los Ríos, Balancán 86930, Mexico, Universidad Tecnológica de Manzanillo, Las Humedades s/n Col. Salagua, Manzanillo 28869, Mexico.*
- *Monitoring Coastal Water Turbidity Using Sentinel2—A Case Study in Los Angeles by Yuwei Kong*
- *Advances in image acquisition and processing technologies transforming animal ecological studies*
- *Sajid Nazir ", Muhammad Kaleem*
- *Bengaluru uses big data analytics to check unaccounted water supply <https://cio.economictimes.indiatimes.com/news/case-studies/bengaluru-uses-big-data-analytics-to-check-unaccounted-water-supply/30732789>*
- *Digital Signal Processing By Dr. Sandeep Gupta*
- *A machine learning-based strategy for estimating non-optically active water quality parameters using Sentinel-2 imagery Hongwei Guo , Jinhui Jeanne Huang , Bowen Chen , Xiaolong Guo & Vijay P. Singh*
- *Geospatial Techniques and their Role in Natural Resources Management B.U. Choudhury and Patiram ICAR Research Complex for NEH Region, Umiam - 793103, Meghalaya*
- *Citizen science can improve conservation science, natural resource management, and environmental protection Author links open overlay panelDuncan C. McKinley , Abe J. Miller-Rushing , Heidi L. Ballard, Rick Bonney , Hutch Brown, Susan C. Cook-Patton, Daniel M. Evans, Rebecca A. French , Julia K. Parrish , Tina B. Phillips, Sean F. Ryan, Lea A. Shanley , Jennifer L. Shirk, Kristine F. Stepenuck, Jake F. Weltzin , Andrea Wiggins, Owen D. Boyle , Russell D. Briggs , Stuart F. Chapin III , David A. Hewitt ...Michael A. Soukup Image Processing Techniques for Analysis of Satellite Images for Historical Maps Classification—An Overview by Anju Asokan J. Anitha Monica Ciobanu Andrei Gabor Antoanela Naaji And D. Jude Hemanth.*
- *Comparative assessment of remote sensing-based water dynamic in a dam lake using a combination of Sentinel-2 data and digital elevation model Muhittin Karaman · Enre Özelkan*
- *M.A. Hossain, T.Y. Gan, A.B.M. Baki, Assessing morphological changes of the Ganges River using satellite images, Quat. Int. 304 (2013) 142–155. [16] X. Liu, Q. Zhang, Y. Li, Z. Tan, A.D. Werner, Satellite image-based investigation of the seasonal variations in the*





- hydrological connectivity of a large floodplain (Poyang Lake, China), *J. Hydrol.* 585 (2020) 124810.
- Does the existing liquid level measurement system cater the requirement of future generation by Jayalaxmi R. Hanni, Santhosh Krishnan Venkata science direct
 - J. Pan, Y. Yin, J. Xiong, W. Luo, G. Gui, H. Sari, Deep learning-based unmanned surveillance systems for observing water levels, *IEEE Access* 6 (2018) 73561–73571.
 - G. Bai, J. Hou, Y. Zhang, B. Li, H. Han, T. Wang, R. Hinkelmann, D. Zhang, L. Guo, An intelligent water level monitoring method based on SSD algorithm, *Measurement* 185 (2021) 110047.
 - C. Chen, R. Fu, X. Ai, C. Huang, L. Cong, X. Li, J. Jiang, Q. Pei, An integrated method for river water level recognition from surveillance images using convolution neural networks, *Rem. Sens.* 14 (23) (2022) 6023.
 - River Water Level and Water Surface Slope Measurement From Spaceborne Radar and LiDAR Altimetry: Evaluation and Implications for Hydrological Studies in the Ganga River Pankaj R Dhote +7
 - GPRS based river water level monitoring and measuring system By Ya-Jun wang+1
 - Using Multi-Temporal Sentinel-1 and Sentinel-2 Data for Water Bodies Mapping By Luiji Russo
 - SIS2-Water: A Global Dataset for Semantic Segmentation of Water Bodies From Sentinel- 1 and Sentinel-2 Satellite Images by Marc Wieland +5
 - C. Huang, Y. Chen, S. Zhang and J. Wu, "Detecting extracting and monitoring surface water from space using optical sensors: A review", *Rev. Geophys.*, vol. 56, no. 2, pp. 333-360, 2018.(Referefrom above)
 - F. Isikdogan, A. C. Bovik and P. Passalacqua, "Surface water mapping by deep learning", *IEEE J. Sel. Topics Appl. Earth Observ. Remote Sens.*, vol. 10, no. 11, pp. 4909-4918, Nov. 2017.(Ref from above)
 - T. James, C. Schillaci and A. Lipani, "Convolutional neural networks for water segmentation using Sentinel-2 red green blue (RGB) composites and derived spectral indices", *Int. J. Remote Sens.*, vol. 42, no. 14, pp. 5338-5365, 2021.(Ref from above)
 - Building a Smart City: A Conceptual Approach to Real-Time Urban Flood Control System by anil himire+1
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 - Leveraging ML with XGBoost, CatBoost and LGBost Classifiers to Optimize Water Quality Assessment and Prediction
 - By Preet Singh +2
 - Google Earth Engine Cloud Computing Platform for Remote Sensing Big Data Applications: A Comprehensive Review [Meisam+4]
 - Predictive Analytics: An Optimization Perspective
 - Chowdhury M, Hasan ME, Abdullah-Al-Mamun MM (2020) Land use/land cover change assessment of Halda watershed using remote sensing and GIS. *Egyptian Journal of Remote Sensing and Space Science* 23:63–75. <https://doi.org/10.1016/j.ejrs.2018.11.003>
 - Sangeet. 2020. "Advanced Applications of PLM Solutions in Data Center Infrastructure Planning and Delivery." *International Journal of Applied Mathematics & Statistical Sciences (IJAMSS)* 9(4):125–154.
 - Prasad, Rohan Viswanatha, Priyank Mohan, Phanindra Kumar, Niharika Singh, Punit Goel, and Om Goel. "Microservices Transition Best Practices for Breaking Down Monolithic Architectures." *International Journal of Applied Mathematics & Statistical Sciences (IJAMSS)* 9(4):57–78.
 - Prasad, Rohan Viswanatha, Ashish Kumar, Murali Mohana Krishna Dandu, Prof. (Dr.) Punit Goel, Prof. (Dr.) Arpit Jain, and Er. Aman Shrivastav. "Performance Benefits of Data Warehouses and BI Tools in Modern Enterprises." *International Journal of Research and Analytical Reviews (IJRAR)* 7(1):464. Retrieved (<http://www.ijrar.org>).
 - Dharuman, N. P., Dave, S. A., Musunuri, A. S., Goel, P., Singh, S. P., and Agarwal, R. "The Future of Multi Level Precedence and Pre-emption in SIP-Based Networks." *International Journal of General Engineering and Technology (IJGET)* 10(2): 155–176. ISSN (P): 2278–9928; ISSN (E): 2278–9936.

