# Chemical Separation Technology

**https://chemical.journalspub.info/index.php?journal=IJCST**

Review International Contract of the Contract o

Employing ML and DL to Optimize an Electrochemical Biosensor

Yashas A Mithra $^{\text{\tiny{14}}}$ , Yashaswini D.M.<sup>2</sup>, Salma Itagi<sup>3</sup>, M.S. Ganesha Prasad<sup>4</sup>

## *Abstract*

*A major field of study in science, that have made the understanding of the world and madehomo sapiens the dominant species on the planet is our understanding of medicine. Underthis vast umbrella is the detection and classification of microorganisms. Electrochemical biosensors have emerged as promising tools in this regard due to their high sensitivity, selectivity, and potential for miniaturization. By integrating a Machine Learning & Deep Learning model for the detection of the bacterium in the sample and also to see how different physical and chemical environmental characteristics affect the working of the sensor. The working of the project is quite elementary, as a sensor is simulated on COMSOL Multiphysics, in a particular environment, with varying factors of Sensor factors, namely: Electrode Shape, Material & height, material of the substrate and the dielectric used. Thisis done to test varying factors and to find the best combination of materials that produce the most consistent and accurate values. Following which, the data generated from the simulation, which will be a set of electric field values can be processed and mutated to the required format and will be fed into the machine learning algorithm to be trained. The algorithm best suited is the Random ForestMethod, where a decision tree is formed to detect the bacteria present in the sample. Further, a Deep Learning model can also be trained to add a layer of complexity, which shows how each of the underlying chemical & physical properties of different bacterium play a role in generating the final peak electric field values. This works as the chemical &physical properties is unique to every bacterium. If implemented and developed, this project can act as a lifesaver in disease detection as "Prevention is better than cure". This can correctly can help communities, especially low- income and underprivileged ones. The applications include, but are not limited to MedicalDiagnostics, Environmental Monitoring, Food Safety, Agriculture, Biotechnology & Pharmaceuticals and Security and Defense.*

**Keywords:** Electrochemical Biosensor, Machine Learning, Deep Learning, Random Forest, Bacteria Detection, Convolutional neural network*.*

\*Author for Correspondence Yashas A Mithra E-mail: yashasamithra.20ec@saividya.ac.n

<sup>1</sup>Student, Department of Computer Science & Engineering, Sai Vidya Institute of Technology, Bengaluru, Karnataka, India <sup>2-3</sup>Professor, Department of Computer Science & Engineering, Sai Vidya Institute of Technology, Bengaluru, Karnataka, India <sup>4</sup>Principal, Sai Vidya Institute of Technology, Bengaluru, Karnataka, India

Received Date: May 08, 2024 Accepted Date: June 23, 2024 Published Date: July 30, 2024

Citation: Neha Sahu, Rizwan Arif. Employing ML and DL to Optimize an Electrochemical Biosensor. International Journal of Chemical Separation Technology. 2024; 10(1): 33–38p.

## **INTRODUCTION**

Our primary goal is to design an Electrochemical Biosensor capable of swiftly and accurately detectingpathogens without reliance on labels or probes. By avoiding the complexities of sample preparation, our biosensor streamlines the detection process, promising enhanced speed and efficiency in an environment. Since the fabrication of the sensor is beyond current reach, we will have to make-do with a simulation. [15]

Another goal, and the one we are predominantly focusing on is to detect the analyte for the bacteria it carries by training a ML/DL model. This is further backed by the various microscopic factors, both physical and chemical in nature, like wettability,

charge, roughness, topography, hydrophobia, size and concentration around the electrodes of the sensor that affect its working and to study how each of these parameters affect the electric field generated in the simulation. [611]

The implications of our project are significantly in the medical domain. With an emphasis on medical applications, this biosensor holds promise for transformative impact. This innovation stands out to expedite diagnostics, elevate healthcare standards, and enhance the efforts in combating infectious diseases. [10]

#### **OBJECTIVE**

Based on the research gaps that have been identified it is possible to formulate the following research objectives:

#### **Objective 1**

*Exploring the Impact of Different Electrode Shapes and Materials:* The aim is to investigate how various electrode shapes affect the performance of biosensors focusing on their influence, on sensitivity, selectivity and stability. The objective is to improve the efficiency and accuracy of biosensors by examining properties of electrode materials such as conductivity, stability and compatibility with target analytes.

## **Objective 2**

*Optimizing Substrate and Dielectric Materials:* This objective aims to determine how different substrate materials affect the support, electrical insulation and thermal stability of biosensors.The objective is to find materials that can improve these qualities. By investigating how dielectric materials affect insulation, leakage current avoidance, mechanical support, and thermal stability, this research aims to develop biosensor technology. The substance that can increase biosensor efficiency overall is the target.

## **Objective 3**

*Obtaining CSV of other bacteria*: One of the most important aspects is to obtain the values of various potentially life- threatening bacteria like Cholera, E Coli, Salmonella, H1N1, Dengue, SarsCOV2, Influenza, MERS, TB. These values need to be obtained by either conducting a deep literature survey to access values simulated by other researchers or by installing additional packages on the COMSOL Multiphysics software.

These research objectives address the areas of investigation identified in the research gaps and provideguidance for further research, in the field of electrochemical biosensors.

#### **Problem Statement**

The problem statement dictates 3 main topics of discussion and research as listed below:

- Investigating various dielectric, substrate, and electrode forms and materials for the detection of analytes (such as bacteria, fungi, viruses, and pathogens, among others).To train an ML/DL model that uses data sets from COMSOL Multiphysics to:
- Optimize the sensor observing how each parameter affects the working of the sensor and eventually, the value of the peak electric field generated.
- Detect the analyte in the sample provided.
- To provide a low-cost, portable, easy-to-use, easy to fabricate and rapid sensor under arduous conditions, ensuring optimal performance.

#### **Literature Survey**

Discusses the traditional methods for detecting pathogenic bacteria in labs are slow and require trained experts. To overcome these limitations, microelectrodes are employed for recognizing harmful microorganisms using dielectric impedance spectroscopy. The effectiveness of this method relies on the spatial dimensions and arrangement of the electrodes, which define key sensor parameters like sensitivity, signal-to-noise ratio (SNR), and dynamic range. The study involves the creation and evaluation of various sensor models, with finite element method (FEM) simulations conducted in both 2D and 3D. The researchers tested different electrode shapes, examined the impact of excitation amplitude, and introduced a mathematical concept to assess sensor quality. They also explored the influence of guard electrodes in blocking external factors affecting the electric field, ultimately leading to optimized sensor designs in terms of geometry. The research provides insights into suitable material choices for practical sensors and offers design and layout recommendations.

Presents an overview on utilizing cyclic voltammetry analysis, the performance of variousworking electrode designs with different geometries was compared to those with the same areabut varying perimeters. The findings indicate that electrodes with higher perimeters exhibitedsuperior current compared to those with smaller perimeters. Hence, the perimeter of the electrode geometry plays a significant role in determining the peak current of an electroanalytical cell. Therefore, in order to optimize the geometry, it is recommended to maintain a constant area while varying the scan rate. Among the different geometries tested, the Fibonacci geometry demonstrated the highest peak current value of 93mA. This particulargeometry holds promise for fabricating the proposed Electrochemical sensor and merits furtherinvestigation in determining ion concentrations.

Provides an overview of the use of ECBs in pathogen detection, including their recognitionelements and transduction strategies. In addition to their application in point-of-care single detection scenarios, ECBs also show promise for online multiplex monitoring. While ECBs are primarily utilized for assessing bacteria, there has been an increasing focus on virusdetection, particularly COVID-19, in the past three years [2, 3]. Furthermore, ECBs can be miniaturized and made portable, offering convenience in various settings. A significantadvantage of ECBs is their label-free detection capability, eliminating the need for culturing orpolymerase chain reaction-based methods. This review specifically examines the modificationof working electrodes with nanocomposites, such as carbonand metal-based materials, which enhance ECB specificity and electrochemical signals. It draws attention to the enormous potential that ECBs have to stop the spread of extremely dangerous illnesses and enhance healthcare around the world. discusses the challenges of using traditional tools to detect specific pathogens, likening it to finding a needle in a haystack. However, the emergence of advanced biosensors offers a promising solution. These cutting-edge biosensors have the ability to detect infections in various scenarios without the need for sample preparation. They can now be employed to identify a wide range of illnesses, including those found in food, bodily fluids, and even on object surfaces. While optical techniques have higher sensitivity compared to electrochemicalmethods, they are often expensive and difficult for most users to utilize. However, whereas electrochemical approaches are easier to use, their accuracy in identifying diseases is restricted. However, with the ongoing commercialization and expansion of biosensor technology, it may soon be possible to have small biosensors placed near a patient's bed, in a doctor's office, or even at home. This would revolutionize disease detection and monitoring, transforming the current approach.

The study addresses the recent developments in bioanalytical systems for bacteria detectiongo beyond traditional biosensors. Solid surfaces can be chemically or physically modified to attract target bacteria without traditional capture probes. In these interactions, surface characteristics like as hydrophobicity, charge, and topography are vital. Surface-immobilized glycans and bacteria-imprinted polymers have shown promise in capturing pathogenic bacteria.However, the exact mechanisms behind these systems are not fully understood, and their rational design is still evolving. While traditional capture probes are prevalent, this counter opinion advocates for incorporating bio-solid interactions into sensor designs, aiming to bridgethe gap between these approaches for more effective bacterial detection and monitoring.

Tells us that the systematic optimization and nonlinear relationships have been compromisedfor electrode fabrication and data analysis. Machine learning and experimental designs are chemometric tools that have proven useful in method development and data analysis. The most recent uses of experimental designs and machine learning in electroanalytical chemistry are compiled in this minireview. Various experimental designs, including complete factorial, central composite, and Box-Behnken, are examined as methodical ways to optimize the production of electrodes while taking individual variable effects and their interactions into account. In order to improve calibration and analyte classification, machine learning algorithms—such as neural networks, support vector machines, and logistic and linear regressions—are introduced to extract complex relationships between chemical structures and their electrochemical properties and analyze complex electrochemical data. Given that these chemometric methodologies will hasten the development and improve the performance of electrochemical devices for point-of-care diagnostics and commercialization, the future of machine learning and experimental designs in electrochemical sensors is discussed.

Reviews recent progress in machine learning application to biosensors, discussing potentialbenefits such as improved sensitivity, selectivity, and accuracy. Additionally, it covers a range of machine learning methods used with biosensors, such as feature extraction, data preprocessing, classification, and data analysis models. Machine learning has the potential to improve biosensors by enabling realtime biological process monitoring, analyzing massive and complex data sets, and identifying minute changes in biomolecular interactions. Challenges associated with integrating machine learning and biosensors are addressed, including data availability, sensor performance, and computational requirements. The development of portable, affordable biosensors and the application of machine learning algorithms for effective data analysis are only two of the potential and obstacles for integrating biosensors with machine learning that are highlighted in the report. Artificial intelligence and deep learning algorithms for biosensors, as well as the possibility of developing a fully autonomous biosensing system, are future trends and new technologies in the field.

According to [7], ML is being used to generate data-driven gains in accuracy, repeatability, and sensitivity. This involves employing multi-output regression models to determine numerous targets from a single measurement. A significant trend in the literature has been the development of supervised machine learning models that are trained on massive data sets generated by electrical and electrochemical biosensors. These models enable precise analyses despite problems like electrode fouling, low signal-to-noise ratios, chemical interferences, and matrix effects. In order to alleviate the need for adopting experimental techniques to address these interfering issues, this trend article provides examples from the literature that show how beneficial ML algorithms can be. This will ultimately help translate testing technologies into real-world, useful, and everyday applications. A wide range of fields, including computer vision, speech processing, finance, locomotion, personality profile, gaming, organic synthesis, bioinformatics, drug discovery, material design, and sensors/biosensors (bio/sensors) have all benefited from machine learning. The inherent benefits of bio/sensors, such as their speed, affordability, ease of use, non-structural nature, and capacity for on-site applications in the food, biomedical, and environmental domains, have led to their widespread development.

Explains how machine learning (ML) in biosensors can be used for diagnosis, analysis, and detection [8]. While machine learning, or ML, has advanced significantly in recent years, deep learning—which is particularly useful for image analysis, facial recognition, and speech recognition—has proven to be quite difficult for the biosensor community to use. The writers talk about the benefits and limitations of well-known machine learning algorithms, including deep learning techniques like convolutional

Based on the analysis of sensory data, neural networks (RNN) and recurrent neural networks (RNN). They also discuss diverse ML-assisted electrochemical biosensors, wearable electronics, SERS, fluorescence biosensors, and colorimetric biosensors. The authors also introduce biosensor networks and multi- biosensor data fusion. According to the authors, machine learning (ML) can offer innovative approaches to help biosensors overcome obstacles and develop into intelligent biosensors that can automatically forecast species or concentrations based on a decision system for diagnosis, analysis, and detection.

Presents an overview of the COVID-19 pandemic and the extensive discussions surrounding testing. The significance of early and widespread testing for SARS-CoV-2 has been well- established as it reduces mortality rates and aids in contact tracing efforts. However, the accuracy and accessibility of diagnostic tests have a major impact on how successful testing is. With concerns mounting over the complexity of RT-PCR and ELISA-based tests, as well as inventory shortages, it is crucial to explore alternative platforms that are user-friendly while maintaining sensitivity and simplicity. Using electrochemical-based sensors, which have several benefits, is a viable way to overcome these issues. These sensorsemploy various methods of signal transduction, enabling the detection of a wide range of analytes with exceptional accuracy, depending on the specific biorecognition element employed. The versatility of electrochemical based devices makes them a potential game- changer in the field of rapid diagnostics for COVID-19.

Talks about the recent advancement of machine learning (ML) algorithms and theories hasoffered new opportunities and insight to address these challenges adequately. Real-world applications that are already benefiting from such ML algorithms include healthcare analytics, business, economics, analytics, and computer science. The power of machine learning (ML) algorithms resides in their capacity to automatically learn and extract patterns and features from a given collection of data—a task that often calls for a domain expert. These algorithms can be excellent candidates to replace traditional sensing approaches with expanding datasets and poorly understood system models.

## **Proposed System & System Diagram**

The system under development will first explore existing studies on how different electrode shapes and materials impact biosensor performance, focusing on sensitivity, selectivity, and stability. Further research to be done on properties of electrode materials, substrate material and investigate the dielectric materials impact on the sensor and the resultant fluctuation in the electric field. Since the fabrication of the said sensor is beyond current capabilities, a simulation can be conducted on the COMSOL Multiphysics application. FIG 1



**Fig 1:** Basic Block Diagram.

A diversified dataset must be ready for the model's training after the simulation. After preprocessing and engineering the data, the model will be trained with fine parameter tuning, meaningthe study of each of the aforementioned properties affect the final electric field graph. The model will predict the type of bacteria present in the sample.

Early research and analysis suggest that a Random Forest Classifier using an Ensemble Learning algorithm, with the aid of a Convolutional neural network is best fit for the job.

#### **CONCLUSION**

The integration of ML & DL into the optimization of electrochemical biosensors represents a groundbreaking advancement in the realm of bacterial detection. The ML of Ensemble Learning algorithm, combined with the DL of a Convolutional neural network will prove to. Be very efficient due to its Space and Time Complexities. The combined power of ML and DL not only enhances the speed and accuracy of electrochemical biosensors but also opens avenues for innovation. These technologies enable the development of low-cost, portable biosensors that are easy to use and fabricate. As we continue to explore and refine this convergence, the future holds the promise of accessible and impactful biosensor solutions, particularly in resource-limited settings where traditional methods may fall short.

#### **REFERENCES**

- 1. Pfeffer, C.; Liang, Y.; Grothe, H.; Brederlow, R. "Towards Easy-to-Use Bacteria Sensing: Modeling and Simulation of a New Environmental Impedimetric Biosensor in Fluids". Sensors 2021, 21, 1487. https://doi.org/10.3390/s21041487
- 2. S.J. Bharathi, S. H. Thilagar and V.Jayasurya, "Design and Modeling of ElectrochemicalSensor for Determining ION Concentration", 2019 IEEE 1st International Conference on Energy, Systems and Information Processing (ICESIP), Chennai, India, 2019, pp. 1-5, doi: 10.1109/ICESIP46348.2019.8938357.
- 3. Jasmina Vidic, Marisa Manzano, "Electrochemical biosensors for rapid pathogendetection, Current<br>Opinion in Electrochemistry", Volume 29.2021,100750, ISSN 2451-9103, Opinion in Electrochemistry", Volume 29,2021,100750, ISSN 2451- 9103, https://doi.org/10.1016/j.coelec.2021.100750.
- 4. Banakar, M.; Hamidi, M.; Khurshid, Z.; Zafar, M.S.; Sapkota, J.; Azizian, R.; Rokaya, D. "Electrochemical Biosensors for Pathogen Detection: An Updated Review". Biosensors 2022, 12, 927. https://doi.org/10.3390/bios12110927
- 5. Ofer Prinz Setter, Xin Jiang, Ester Segal, "Rising to the surface: capturing and detecting bacteria by rationally-designed surfaces", Current Opinion in Biotechnology, Volume 83, 2023, 102969, ISSN 0958-1669, https://doi.org/10.1016/j.copbio.2023.102969.
- 6. "Machine learning and chemometrics for electrochemical sensors: moving forward to the future of analytical chemistry" by Pumidech Puthongkham, Supacha Wirojsaengthonga and AkkapolSuea-Ngam on 7 Sept 2021. DOI: https://doi.org/10.1039/D1AN01148K
- 7. "Machine Learning for Biosensors" by Gayathri Anapanani in 2023, for Statler College of Engineering and Mineral Resources at West Virginia University. DOI: https://doi.org/10.33915/etd.12118
- 8. "Machine learning toward high-performance electrochemical sensors" by Gabriela F. Giordano, Larissa F. Ferreira, Ítalo R. S. Bezerra, Júlia A. Barbosa, Juliana N. Y. Costa, Gabriel J. C. Pimentel & Renato S. Lima. DOI: 10.1007/s00216-023-04514-z
- 9. "Advancing Biosensors with Machine Learning" by Feiyun Cui, Yun Yue, Yi Zhang, Ziming Zhang, and H. Susan Zhou on November 13, 2020 in 2020 American Chemical Society. https://doi.org/10.1021/acssensors.0c01424
- 10. Sahar Sadat Mahshid, Sarah Elizabeth Flynn, Sara Mahshid, "The potential application of electrochemical biosensors in the COVID-19 pandemic: A perspective on the rapid dianostics of SARS-CoV-2, Biosensors and Bioelectronics", Volume 176, 2021, 112905, ISSN 0956-5663, https://doi.org/10.1016/j.bios.2020.112905
- 11. Nam Ha, Kai Xu, Guanghui Ren, Arnan Mitchell, and Jian Zhen Ou, "Machine Learning-Enabled Smart Sensor Systems" in Advanced Intelligent Systems, 14 July 2020 https://doi.org/10.1002/aisy.202000063