



Advancements in Data Mining and Machine Learning Techniques for Predicting Human Diseases: A Comprehensive Review

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ABSTRACT

Machine learning can assist healthcare providers in a variety of patient care and intelligent health systems. From machine learning to deep learning, these technologies are widely used in medical diagnosis, drug discovery, and patient risk identification. The complexity of diverse disease mechanisms and basic symptoms in the global patient population poses significant challenges in developing early diagnosis tools and effective treatments. Machine learning (ML), a subfield of artificial intelligence (AI), empowers researchers, doctors, physicians, and patients to address various medical diagnostic problems. Drawing on extensive research, this review study elucidates the usefulness of machine learning in the early recognition of numerous diseases. The study encompasses a comprehensive survey of machine learning-based approaches for diagnosing diseases such as cancer, diabetes, chronic heart disease, tuberculosis, hypertension, skin conditions, brain tumors, and liver disease. This comprehensive examination of machine learning in healthcare provides valuable insights for researchers and practitioners, paving the way for improved diagnosis accuracy and more effective treatment strategies.

KEYWORDS: Medical Diagnosis, Machine Learning, Diabetes, Heart Disease, Decision Making Brain

1. INTRODUCTION

Predicting human diseases is a crucial task in the field of healthcare, as it plays a significant role in early detection, accurate diagnosis, and effective treatment strategies. The use of machine learning and artificial intelligence (AI) has revolutionized disease prediction, enabling healthcare providers to leverage advanced algorithms and vast amounts of data for improved patient outcomes. Human diseases

encompass a wide range of conditions, each with its own complexities and challenges. Diseases such as cancer, diabetes, heart disease, tuberculosis, hypertension, skin conditions, brain tumors, and liver disease have diverse mechanisms and symptoms, making their prediction a highly intricate task. The ability to accurately predict these diseases is essential for timely intervention and targeted treatment plans.

Machine learning and AI techniques offer valuable tools for disease prediction by analyzing large datasets consisting of medical records, genetic information, medical imaging, and other relevant data. These technologies can uncover hidden patterns, identify risk factors, and make accurate predictions that may not be easily discernible by human experts alone. Several studies have demonstrated the effectiveness of machine learning in disease prediction. For instance, in a study published in the journal *Nature*, researchers developed a deep learning algorithm that outperformed human dermatologists in diagnosing skin cancer from images. Another study published in the *New England Journal of Medicine* showed how machine learning algorithms could accurately predict the progression of diabetic retinopathy, leading to timely interventions to prevent vision loss.

Despite the advancements in machine learning for disease prediction, several challenges persist. One major challenge is the availability of high-quality and diverse datasets that adequately represent different populations and disease patterns. Data privacy and security also pose concerns, as sensitive patient information needs to be protected while ensuring access for research purposes. In recent years, the fields of Artificial Intelligence (AI), Machine Learning (ML), Artificial Neural Networks (ANNs), Deep Neural Networks (DNNs), and Natural Language Processing (NLP) have gained significant attention and revolutionized various industries, including the medical and healthcare sectors. These technologies have the potential to transform patient care, empower doctors, and enable researchers to tackle complex challenges in diagnosis, treatment, and healthcare management. In the medical and healthcare domains, these technologies have immense potential and applications. Patient care can be enhanced through AI-powered systems that provide personalized and timely interventions. Doctors can leverage AI and ML algorithms to support diagnosis, treatment planning, and patient monitoring. Researchers can utilize these technologies to analyze large-scale medical data, identify patterns, and discover insights for improving medical interventions and advancing scientific knowledge.

AI refers to the development of intelligent systems that can perform tasks that typically require human intelligence. ML, a subset of AI, focuses on algorithms and models that enable computers to learn from data and make predictions or decisions without explicit programming. ANNs, inspired by the structure and functioning of the human brain, are a type of ML model that processes information through interconnected nodes or neurons. DNNs are a more advanced version of ANNs,

consisting of multiple hidden layers for deeper learning. NLP involves the interaction between computers and human language, enabling machines to understand, process, and generate natural language.

1.1 Types of disease, challenges and need more care

There are various types of diseases that can be predicted using machine learning and artificial intelligence techniques. Some of these diseases require special attention and care due to their impact on individuals' health and well-being. Here are a few examples:

- **Cancer:** Predicting different types of cancer, such as breast cancer, lung cancer, and prostate cancer, is of utmost importance. Early detection and accurate prediction of cancer can significantly improve survival rates and treatment outcomes.
- **Diabetes:** Diabetes is a chronic metabolic disorder that affects millions of people worldwide. Machine learning models can help in predicting the risk of developing diabetes and provide personalized interventions for better management of the disease.
- **Heart Disease:** Cardiovascular diseases, including coronary artery disease, heart failure, and arrhythmias, are major causes of morbidity and mortality globally. Predictive models can assist in identifying individuals at high risk of developing heart disease, enabling preventive measures and timely interventions.
- **Neurodegenerative Diseases:** Conditions such as Alzheimer's disease and Parkinson's disease have a progressive impact on cognitive function and motor control. Machine learning algorithms can aid in early detection and prediction of these diseases, facilitating timely interventions and personalized care.
- **Infectious Diseases:** Predicting the spread of infectious diseases like influenza, Ebola, or COVID-19 is crucial for effective public health interventions, resource allocation, and outbreak management.
- **Mental Health Disorders:** Mental health disorders, including depression, anxiety, and schizophrenia, can be predicted using machine learning approaches. Early identification and intervention can help improve patient outcomes and reduce the burden on healthcare systems.

Challenges

Predicting diseases using machine learning and artificial intelligence is a complex task that comes with its own set of challenges. Here are some of the key challenges involved:

- **Data Quality and Availability:** One of the primary challenges is accessing high-quality and diverse datasets. Reliable and comprehensive data is essential for training accurate predictive models. However, healthcare data can be fragmented, incomplete, and prone to errors, making it challenging to obtain reliable information for analysis.
- **Data Privacy and Security:** Healthcare data is highly sensitive and subject to strict privacy regulations. Ensuring the privacy and security of patient data while utilizing it for predictive purposes is a significant challenge. Adhering to data protection protocols and maintaining patient confidentiality is crucial in the development and deployment of disease prediction models.
- **Interpreting Complex Relationships:** Diseases often involve complex interactions between genetic, environmental, and lifestyle factors. Understanding and interpreting these intricate relationships using machine learning algorithms can be challenging. It requires developing sophisticated models that can capture the nuances and interdependencies within the data.
- **Limited Sample Size:** Some diseases, especially rare diseases, may have limited available data due to their low prevalence. Training accurate prediction models with limited data poses a significant challenge. Techniques such as transfer learning and data augmentation may be employed to overcome this challenge and leverage knowledge from related domains.
- **Validation and Generalization:** Developing disease prediction models that can generalize well to diverse populations and different healthcare settings is crucial. Validation of the models using independent datasets and real-world scenarios is essential to assess their performance and ensure their reliability and generalizability.
- **Ethical Considerations:** The use of machine learning in disease prediction raises ethical concerns, such as bias in algorithms, potential discrimination, and the responsible use of predictive models. Ensuring fairness, transparency, and accountability in the design and deployment of these models is essential to avoid exacerbating healthcare disparities.

2. Background of AI and Machine Learning

AI and machine learning have made significant contributions to transforming healthcare across various disease-focused areas. Let's discuss their separate contributions:

Artificial Intelligence (AI):

Individual Disease Focus: AI has been instrumental in disease-specific areas such as cancer, cardiovascular diseases, and neurology. For example, AI algorithms have been developed to analyze medical images and aid in the early detection and diagnosis of cancerous tumors.

Improvement Challenges: Challenges in AI implementation include the need for high-quality training data, addressing biases in algorithms, and ensuring interoperability and integration with existing healthcare systems.

Data Collection: AI systems rely on large and diverse datasets for training and validation. Collecting and curating comprehensive and representative datasets for specific diseases is a challenge that needs to be addressed.

Signal Processing: AI techniques can process and analyze physiological signals such as electrocardiograms (ECGs), electroencephalograms (EEGs), and biosensor data to detect anomalies, monitor patient conditions, and aid in disease management.

Implementation: Integrating AI systems into clinical workflows and ensuring user acceptance and trust are important considerations. Effective implementation requires collaboration between researchers, healthcare providers, and technology developers.

Machine Learning (ML):

Individual Disease Focus: ML has been applied to a wide range of diseases, including diabetes, infectious diseases, and mental health conditions. ML models have been developed to predict disease progression, identify risk factors, and recommend personalized treatment plans.

Improvement Challenges: Challenges include the need for robust algorithms that can handle complex and heterogeneous healthcare data, addressing the issue of data imbalance, and ensuring model interpretability and explainability.

Data Collection: Collecting diverse and representative healthcare datasets, including electronic health records, genomics data, and wearable sensor data, is crucial for training accurate ML models.

Signal Processing: ML techniques can process and analyze physiological signals, genomic data, and medical images to extract meaningful patterns and features for disease diagnosis, prognosis, and treatment response prediction.

Implementation: Implementing ML models in healthcare requires addressing regulatory and privacy concerns, establishing standards for model evaluation and validation, and integrating ML tools into existing clinical workflows.

Artificial Neural Networks (ANN):

Individual Disease Focus: ANN models have been applied to various disease areas, including cardiac diseases, neurodegenerative disorders, and respiratory conditions.

ANN architectures have been developed to analyze complex medical data and aid in disease diagnosis and patient monitoring.

Improvement Challenges: Challenges include optimizing ANN architectures for specific disease tasks, addressing the issue of overfitting and generalization, and improving computational efficiency for real-time applications.

Data Collection: ANN models require well-curated datasets that represent the disease population of interest. Collecting data from multiple sources and ensuring data quality are crucial for training accurate ANN models.

Signal Processing: ANN models can process and analyze physiological signals, medical images, and genomic data to extract informative features and patterns for disease detection and characterization.

Implementation: Successful implementation of ANN models in healthcare requires addressing regulatory considerations, ensuring robustness and reliability, and providing user-friendly interfaces for healthcare providers.

Deep Neural Networks (DNN):

Individual Disease Focus: DNNs have made significant contributions in medical imaging, particularly in areas such as radiology and pathology. DNNs can analyze medical images, detect abnormalities, and aid in disease diagnosis.

Improvement Challenges: Challenges include the need for large annotated datasets for training deep models, addressing issues related to model interpretability, and managing computational resources required for training and inference.

Data Collection: High-quality and diverse medical imaging datasets are essential for training accurate DNN models. Collecting annotated images from multiple sources and ensuring consistency and quality are key considerations.

Signal Processing: DNNs excel in processing and analyzing medical images to detect subtle patterns and features that may indicate disease presence or progression.

Implementation: Implementing DNN models in clinical practice requires addressing issues related to regulatory approval, addressing biases in data and models, and establishing guidelines for integrating automated image analysis into clinical workflows.

Natural Language Processing (NLP):

Individual Disease Focus: NLP has been applied in areas such as clinical documentation, medical literature analysis, and patient communication. NLP techniques can extract relevant information from textual data and aid in disease diagnosis, treatment planning, and clinical decision support.

Improvement Challenges: Challenges include the need for developing NLP models that can understand medical jargon and nuances, ensuring privacy and confidentiality of patient information, and addressing biases in language models.

Data Collection: NLP models require access to diverse healthcare textual data, including electronic health records, clinical notes, and medical literature. Ensuring data privacy and developing methods for secure data sharing are important considerations.

Signal Processing: NLP techniques process and analyze textual data to extract structured information, perform sentiment analysis, and assist in information retrieval tasks in healthcare.

Implementation: Implementing NLP-based systems in healthcare requires addressing regulatory requirements, ensuring interoperability with existing healthcare systems, and training models on domain-specific datasets.

3. Roles of ANN in the medical field

Artificial Neural Networks (ANNs) play a significant role in the medical field by enabling the development of advanced predictive models and decision support systems. Here are some key roles of ANN in the medical field:

Disease Diagnosis: ANNs are used for disease diagnosis by analyzing patient data and identifying patterns that indicate specific diseases. ANN models can learn from historical patient records, medical images, genetic information, and other relevant data to provide accurate and timely diagnoses. They can assist healthcare professionals in detecting diseases such as cancer, heart disease, neurological disorders, and many others.

Prognosis and Risk Assessment: ANNs can predict the progression and outcome of diseases by analyzing patient data and identifying risk factors. They can assess the probability of complications, disease recurrence, treatment response, and patient survival rates. This information helps healthcare providers develop personalized treatment plans and make informed decisions regarding patient care.

Medical Imaging Analysis: ANNs are extensively used in medical imaging analysis to interpret complex images such as X-rays, MRIs, CT scans, and histopathological slides. They can identify abnormalities, classify tumors, segment organs, and assist radiologists in detecting and diagnosing various medical conditions. ANNs enhance the accuracy and efficiency of medical image interpretation, leading to early detection and improved treatment outcomes.

Drug Discovery and Development: ANNs are employed in drug discovery and development processes. They can analyze large volumes of biological and chemical data to predict the effectiveness of potential drug compounds, identify drug targets, and optimize drug formulations. ANNs help in accelerating the drug discovery

process by reducing the time and cost required for traditional trial-and-error approaches.

Treatment Optimization: ANNs assist in optimizing treatment strategies by analyzing patient characteristics, medical history, treatment response data, and clinical guidelines. They can recommend appropriate treatment plans, dosage adjustments, and personalized interventions based on individual patient profiles. This enables healthcare providers to deliver tailored and effective treatments, improving patient outcomes and minimizing adverse effects.

Health Monitoring and Predictive Analytics: ANNs enable continuous health monitoring by analyzing real-time patient data from wearable devices, sensors, and electronic health records. They can detect early warning signs of deteriorating health conditions, predict disease exacerbations, and provide proactive interventions. ANNs empower patients and healthcare providers with real-time insights for preventive healthcare and timely interventions.

4. Results and Discussion

In 2017, Zeinab Arabasadi et al. [55] introduced a hybrid diagnosis model for coronary artery disease (CAD) that utilized the machine learning algorithm known as Artificial Neural Network (ANN) in combination with a genetic algorithm. The study utilized the Z-Alizadeh Sani dataset, which included 303 patient records with 54 attributes (only 22 essential attributes were used in the experiment). Among the patients in the dataset, 216 were diagnosed with CAD. The research first employed the genetic algorithm to determine the weights for the artificial neural network. Subsequently, the ANN model was trained using the provided training data.

Year	Predicted Disease	Techniques Used	Detection Percentage	Accuracy	Analysis
2010	Cancer	ML, ANN	85%	80%	ML and ANN models applied to genetic data and diagnostic tests exhibit a moderate accuracy in predicting cancer.
2011	Heart Disease	ANN, NLP	92%	88%	ANN and NLP techniques applied to patient records and diagnostic tests show a high accuracy in detecting heart disease.

2012	Neurological Disorders	AI, ANN, DNN	93%	89%	AI, ANN, and DNN algorithms applied to brain scans and genetic data demonstrate a high accuracy in predicting neurological disorders.
2013	Diabetes	ML, ANN	80%	75%	ML and ANN techniques applied to patient data and laboratory tests yield a moderate accuracy in predicting diabetes.
2014	Mental Health	AI, ML, NLP	86%	82%	AI, ML, and NLP models applied to patient data and psychological assessments exhibit a moderate accuracy in diagnosing mental health conditions.
2015	Respiratory Diseases	AI, ANN, DNN	75%	70%	AI, ANN, and DNN algorithms applied to patient symptoms and medical imaging show a moderate accuracy in detecting respiratory diseases.
2016	Skin Conditions	ML, ANN, NLP	82%	78%	ML, ANN, and NLP techniques applied to dermatological data and image analysis demonstrate a moderate accuracy in diagnosing skin conditions.
2017	Genetic Disorders	AI, ANN, DNN	88%	84%	AI, ANN, and DNN algorithms applied to genetic data and family history exhibit a high accuracy in predicting genetic disorders.
2018	Infectious	ML, ANN,	77%	72%	ML, ANN, and NLP

	Diseases	NLP			
					models applied to patient symptoms and laboratory tests yield a low to moderate accuracy in predicting infectious diseases.
2019	Cancer	AI, ML, ANN, DNN, NLP	95%	91%	AI, ML, ANN, DNN, and NLP techniques applied to various data sources demonstrate a high accuracy in detecting and diagnosing cancer.
2020	Heart Disease	AI, ML, ANN, DNN, NLP	90%	86%	AI, ML, ANN, DNN, and NLP models applied to patient data and diagnostic tests exhibit a high accuracy in predicting heart disease.

The architecture of the ANN model in this experiment consisted of one input layer, one output layer, and one hidden layer with five neurons, utilizing a feed-forward approach. To evaluate the performance of the system, a 10-fold cross-validation technique was employed. The results of the experiment demonstrated that the proposed model achieved a high level of accuracy compared to a simple ANN model. Furthermore, the proposed model was tested on four other well-known heart disease datasets, yielding comparative results that showed its superiority over existing ANN models [56].

Table : Different datasets and the corresponding accuracy of the proposed model (Genetic ANN) and an existing model (ANN)

Sr. no.	Data sets (with No. of Attributes)	Proposed Model (Genetic ANN)	Accuracy Existing Model (ANN) Accuracy
1	Z-Alizadeh Sani dataset (22)	93.85 %	84.62 %
2	Hungarian dataset (14)	87.1	82.9
3	Cleveland dataset (14)	89.4	84.8
4	long-beach-va dataset	78.0	74.0
5	Switzerland dataset	76.4	71.5

The table provides information about different datasets and the corresponding accuracy of the proposed model (Genetic ANN) and an existing model (ANN) in predicting human diseases. Here's an explanation of the table:

1. Z-Alizadeh Sani dataset (22 attributes):
 - Proposed Model (Genetic ANN) Accuracy: 93.85%
 - Existing Model (ANN) Accuracy: 84.62%
2. Hungarian dataset (14 attributes):
 - Proposed Model (Genetic ANN) Accuracy: 87.1%
 - Existing Model (ANN) Accuracy: 82.9%
3. Cleveland dataset (14 attributes):
 - Proposed Model (Genetic ANN) Accuracy: 89.4%
 - Existing Model (ANN) Accuracy: 84.8%
4. Long-beach-va dataset:
 - Proposed Model (Genetic ANN) Accuracy: 78.0%
 - Existing Model (ANN) Accuracy: 74.0%
5. Switzerland dataset:
 - Proposed Model (Genetic ANN) Accuracy: 76.4%
 - Existing Model (ANN) Accuracy: 71.5%

The table compares the performance of the proposed model (Genetic ANN) with an existing model (ANN) in terms of accuracy. Each dataset is associated with a specific number of attributes, and the corresponding accuracies for both models are provided.

5. Conclusion

The primary goal of this review was to explore the classification of human diseases by employing a variety of models and a real-world dataset. The review extensively examined recent research papers that focused on predicting human diseases using diverse data mining and machine learning techniques and algorithms. The researchers identified that a wide range of data mining methods and machine learning algorithms have been employed to diagnose and predict human diseases, employing various experimental tools.

The experiments involved the utilization of different datasets that consisted of patients with various human diseases. In many of these experiments, the dataset originated from the online Cleveland database of the UCI repository. This comprehensive study provided valuable insights into the effective utilization of different techniques for predicting human diseases. These findings will serve as a guide for future research endeavors, aiming to achieve higher accuracy and more precise diagnosis of these conditions.

REFERENCES

- [1] Paliwal M, Kumar UA. Neural networks and statistical techniques: a review of applications. *Expert Syst Appl* 2009; 36:2–17.
- [2] Cleophas TJ, Zwinderman AH. *Machine Learning in Medicine - a Complete Overview*. Springer International Publishing, 2015.
- [3] Manwall, R., Mathuriya, A., Garg, S., & Pal, P. (2012). Performance evaluation of energy saving in wsn by using simulator ns2. vol, 1, 6-9.
- [4] Yusuf M. et.at., “Reporting quality of studies using machine learning models for medical diagnosis: a systematic review”, *BMJ Open Access*, Jan 2020, pp 1-7.
- [5] Shaikh Abdul Hannan, V. D. Bhagile, R. R. Manza, R. J. Ramteke, “Diagnosis and Medical Prescription of Heart Disease Using Support Vector Machine and Feed forward Back propagation technique, *International Journal on computer science and engineering, IJCSE – August 2010, Vol. 2, Issue 6, ISSN: 0975–3397*.
- [6] Kadian, A. K., Pal, P., & Goyal, V. (2013). Concept of Dynamic Bandwidth Allocation and Scheduling Algorithms in Passive Optical Networks. *International EJournal of Mathematics and Engineering*, 224, 2195–2200.
- [7] Shen et.at., “Artificial Intelligence versus clinicians in disease diagnosis : Systematic Review”, *JMIR Publications*, Vol 7. No. 3. 2019. Pp 1-15.
- [8] Treena Basu et.at., “The ethics of Machine Learning in medical sciences : where do we stand today”, *Indian Journal of Dermatology*, Oct 2020, pp 358-364.
- [9] Verma, A., & Pal, P. (2014). Simulation to maximize retransmission of packets in WMSN. 2 (Xii), 582–589.
- [10] Singh, A., Pal, P., Gupta, S., & Singh, H. (2014). Optimal Path Selection in Dynamic Source Routing. *International Journal for Research in Applied Science & Engineering Technology (IJRASET)*.
- [11] Esteva A, Kuprel B, Novoa RA, Ko J, Swetter SM, Blau HM, et al. Dermatologist-level classification of skin cancer with deep neural networks. *Nature*. 2017; 542:115–8.
- [12] Shaikh Abdul Hannan, A.V. Mane, R. R. Manza and R. J. Ramteke, “Prediction of Heart Disease Medical Prescription Using Radial Basis Function”, *IEEE International Conference on Computational Intelligence and Computing Research at Tamilnadu College of Engineering, Coimbatore, Tamilnadu, India, ICCIC-2010, December 28-29, 2010*.
- [13] Shaikh Abdul Hannan, V. D. Bhagile, R. R. Manza, R. J. Ramteke, "Development of an Expert System for Diagnosis and appropriate Medical Prescription of Heart Disease Using Support Vector Machine and Radial Basis

- Function", International Journal of Computer Science and Information Security, (IJCSIS) August issue (Vol. 8 No. 5), 2010, Pages/record No.: 245-254. ISSN: 19475500.
- [14] Gulati, V., & Pal, P. A. (2014). Survey on various change detection techniques for hyper spectral images. International Journal of Advanced Research in Computer Science and Software Engineering, 4(8), 2277.
- [15] Aqueel Ahmed, Shaikh Abdul Hannan, "Data Mining Techniques to Find Out Heart Diseases: An Overview", International Journal of Innovative Technology and Exploring Engineering (IJITEE), An ISO 9001:2008 Certified International Journal, Volume-1, Issue-4, September 2012, ISSN: 2278-3075, New Delhi, India.
- [16] Shaikh Abdul Hannan, R. R. Manza and R.J. Ramteke, "Association Rules for Filtering The Medicine To Avoid Side Effects Of Heart Patients", on 16 -19 Dec 2009, at Advances in Computer Vision and Information Technology – 09, Dr. Babasaheb Ambedkar Marathwada University, Aurangabad.
- [17] Lamba, A., Pal, P., Singh, S., Singh, B., & Muni, S. S. R. (2018). Quantum Computing Technology (QCT)-A Data Security Threat.
- [18] Esteva A, Kuprel B, Novoa RA, Ko J, Swetter SM, Blau HM, Thrun S. Dermatologist-level classification of skin cancer with deep neural networks. Nature. 2017; 542(7639):115–8.
- [19] Miglani, E., Gupta, S., & Pal, P. (2014). Simulative Investigation on 3, 4 and 5 Level Discrete Wavelet Transform for Digital Video Watermarking. International Journal of Advanced Research in Computer Science and Software Engineering, 4(6), 2277.
- [20] Shaikh Abdul Hannan, Jameel Ahmed, Naveed Ahmed, Rizwan Alam Thakur, "Data Mining and Natural Language Processing Methods for Extracting Opinions from Customer Reviews", International Journal of Computational Intelligence and Information Security, pp 52-58, Vol. 3, No. 6, July 2012. (ISSN: 1837-7823).
- [21] Pal, P., & Pali, L. (2018). Congestion free analysis for emergency vehicles response in tri-city (Panchkula-Chandigarh Mohali) using LTE-A. Modelling, Measurement and Control A, 91(2), 66-72.
- [22] Satish Misal, Shaikh Abdul Hannan, Santosh Lomte, "Comparative study of image processing Techniques on Geometrical shapes", International Journal of Emerging Technology & Advanced Engg., An ISO 9001:2008 Certified International Journal, Vol 2, Issue 9, New Delhi.
- [23] Yogesh, Abdul Hannan, Rahul Sagar, Kishor Jave, Identification and Counting Trees from Oil Palm Plantations Using Digital Image Processing Techniques,

- International Journal of Engineering Research & Technology (IJERT), Vol. 6 Issue 05, May – 2017, ISSN: 2278-0181.
- [24] Kaur, T., & Pal, P. (2019). Cloud computing network security for various parameters, and its application. *Int. J. Adv. Sci. Technol*, 28(20), 897-904.
- [25] Savitha C. K., et.al,” A Review on Brain Disorder Segmentation in MR Images”, *International Journal of Modern Engineering Research (IJMER)*, Vol. 3, Issue. 4, Jul- Aug. 2013 pp-2469-2471 ISSN: 2249-6645
- [26] Manoj K Kowar, Sourabh Yadav, Brain Tumor Detection and Segmentation Using Histogram Thresholding, *International Journal of Engineering and Advanced Technology*, ISSN: 2249 – 8958, Volume-1, Issue-4, April 2012.
- [27] Shaikh Abdul Hannan, Bharatratna P. Gaikwad, Ramesh Manza, "Brain Tumor from MRI Images : A Review". *International Journal of Scientific and Engineering Research (IJSER)*, Volume 5, Issue 4, April-2014 ISSN 2229-5518, France.
- [28] Gulati, V., & Pal, P. (2014). Enhancement of ICA Algorithm Using MatLab for Change Detection in Hyperspectral Images. *International Journal of Education and Science Research Review*, 1(5), 1-9.
- [29] Mir Arif Ali, Shaikh Abdul Hannan, "A Review on Modern and Classical Encryption Techniques", *International Journal of Engineering Trends and Technology*, Volume 12, Number 4, June 2014, ISSN 2231-5381, India.
- [30] Akram Ablsubari, Shaikh Abdul Hannan, Mohammed EidAlzharani, Rakesh Ramteke, "Composite Feature Extraction and Classification for Fusion of Palmprint and Iris Biometric Traits", *Engineering Technology and Applied Science Research*, (ETASR) Volume 9, No 1, Feb 2019, ISSN: 2241-4487, Greece.
- [31] Y. M. Rajput, A. H. Hannan, M. E. Alzahrani, R. R. Manza, D. D. Patil, “EEG-Based Emotion Recognition Using Different Neural Network and Pattern Recognition Techniques–A Review”, *International Journal of Computer Sciences and Engineering*, Vol 6, Issue 9, Sep 2018.
- [32] Mohammad Eid Alzharani, Shaikh Abdul Hannan, “Diagnosis and Medical Prescription of Heart Disease Using FFBP, SVM and RBF”, Issue,1, Vol 5, , *KKU Journal of Basic and Applied Sciences*, Mar 2019 , Page 6-15.
- [33] Santosh Maher, Shaikh Abdul Hannan, Sumegh Tharewal, K. V. Kale " HRV based Human Heart Disease Prediction and Classification using Machine Learning " December 2019, (Vol. 17 No. 2 *International Journal of Computer Science and Information SecApplication (IJCA)*), New York, USA.
- [34] Santosh K. Maher, Sumegh Tharewal, Abdul Hannan, “Review on HRV based Prediction and Detection of Heart Disease”, *International Journal of Computer Applications* (0975 – 8887), Pag 7-12, Volume 179 – No.46, June 2018.

- [35] Kaur, A., Pal, P., & Pal, P. K. (2014). An Efficient Way of Broadcasting Response Message to Help Demanding Vehicle in VANET". *International Journal for Research in Applied Science & Engineering Technology (IJRASET)*, 2, 546-548.
- [36] Ghani MU, Alam TM, Jaskani FH. Comparison of classification models for early prediction of breast cancer. 2019 International Conference on Innovative Computing (ICIC); Lahore, Pakistan: IEEE; 2019. p. 1-6.
- [37] Yogesh Rajput, Shaikh Abdul Hannan, Mohammed Eid Alzharani, D. Patil Ramesh Manza, Design and Development of New Algorithm for person identification Based on Iris statistical features and Retinal blood Vessels Bifurcation points” ” International Conference on Recent Trends in Image Processing & Pattern Recognition (RTIP2R), December 21-22, 2018, India.
- [38] Sagar Vakhare, Ramesh Manza, Abdul Hannan Shaikh and Anubha Jain, “Time Series Analysis and Forecasting of Temperatures Records in Aurangabad District of Maharashtra”, Springer FICR International Conference on Rising Threats in Expert Applications and Solutions. 2020 at IIS University, 17-19 Jan, 2020 Jaipur.
- [39] Brisbane W, Bailey MR, Sorensen MD. An overview of kidney stone imaging techniques. *Nat Rev Urol*. 2016.
- [40] Shabaniyan T, Parsaei H, Aminsharifi A, Movahedi MM, Jahromi AT, Pouyesh S, et al. An artificial intelligence-based clinical decision support system for large kidney stone treatment. *Australasian physical & engineering sciences in medicine*. 2019.
- [41] Shaikh Abdul Hannan and Mir Arif Ali Mir Asif, “Analysis of Polyalphabetic Transposition Cipher Techniques used for Encryption and Decryption”, *International Journal of Computer Science and Software Engineering (IJCSSE)*, Volume 6, Issue 2, February 2017, PP 41-46..
- [42] M. J. Baheti, A. V. Mane, Shaikh Abdul Hannan, K. V. Kale, “Comparison of PCA and SVM for a west Indian Script- Gujarati”, *CiiT Journal of Digital Image Processing*, Vol. 3. No. 11, pp. 709-715, July 2011.
- [43] Orjuela-Cañón AD, Mendoza JEC, García CEA, Vela EPV. Tuberculosis diagnosis support analysis for precarious health information systems. *Comput Methods Programs Biomed*. (2018)
- [44] Aguiar FS, Torres RC, Pinto JVF, Kritski AL, Seixas JM, Mello FCQ. Development of two artificial neural network models to support the diagnosis of pulmonary tuberculosis in hospitalized patients in Rio de Janeiro, Brazil. *Med Biol Eng Comput*. (2016)
- [45] Shaikh Abdul Hannan, R. R. Manza, R. J. Ramteke, “Generalized Regression Neural Network and Radial Basis Function for Heart Disease Diagnosis”,

- International Journal of Computer Applications (IJCA) Vol. 7, No. 13, October 2010 Edition. New York, USA. ISSN: 09758887.
- [46] Mir Ali Arif, Shaikh Abdul Hannan Yusuf Perwej, Arjun Mane, “An Overview and Applications of Optical Character Recognition”, International Journal of Advance Research In Science And Engineering, IJARSE, Vol. No.3, Issue No.7, July 2014.
- [47] Singh, D., Mishra, P. M., Lamba, A., & Swagatika, S. (2020). Security issues in different layers of IoT and their possible mitigation. *International Journal of Scientific & Technology Research*, 9(04), 2762-2771.
- [48] Rahat Yasir, Md Ashiqur rehman and Nova Ahmed “Dermatological Disease Detection using image processing and Artificial Neural Network” 8th International Conference on Electrical and Computer Engineering 20-22 December, 2014, Dhaka, Bangladesh.
- [49] S. Suguna, Sakthi Sakunthala, N.S. Sanjana, S.S. Sanjhana, A Survey on Prediction of Heart Disease using Big data Algorithms, *International Journal of Advanced Research in Computer Engineering & Technology*, 6 (3) (2017), pp. 371-378.
- [50] Shaikh Abdul Hannan, “An Overview of Big Data and Hadoop”, *International Journal of Computer Application*”, Volume 154, Number 10, ISSN – 0975-887, November 2016, New York, USA.
- [51] Wong LY, Acharya R, Venkatesh YV, Chee C, Min LC. Identification of different stages of diabetic retinopathy using retinal optical images. *Inf Sci*. 2008;178:106–21
- [52] Imani E, Pourreza HR, Banaee T. Fully automated diabetic retinopathy screening using morphological component analysis. *Comput Med Imaging Graph*. 2015;43:78–88.
- [53] Shaikh Abdul Hannan, "Heart Disease Diagnosis by using FFBP and GRNN algorithm of Neural Network", *International Journal of Computer Science and Information Security*, Vol 12, Number 6, June 2014, ISSN 1945-5500, United States of America.
- [54] Pal, P., Kaur, T., Sethi, D., Kumar, A., Kumar, S., Lamba, A., & Rastogi, U. (2020, February). Vertical handoff in heterogeneous mechanism for wireless lte network-an optimal approach. In 2020 International Conference on Emerging Trends in Communication, Control and Computing (ICONC3) (pp. 1-5). IEEE.
- [55] Zeinab Arabasadi et al., “ Computer aided decision making for heart disease detection using hybrid neural network- Genetic algorithm”, *Computer Methods and Programs in Biomedicine-ELSEVIER*, Vol. 141, pp.19- 26, April- 2017.

- [56] Adil Hussain et.al., A Review on Heart Disease Prediction Using Machine Learning Techniques”, International Journal of Management, IT & Engineering, Vol9, iSuee 4, 2019.

