



ISSN: 2321-2152

**IJMECE**

*International Journal of modern  
electronics and communication engineering*

E-Mail

[editor.ijmece@gmail.com](mailto:editor.ijmece@gmail.com)

[editor@ijmece.com](mailto:editor@ijmece.com)

[www.ijmece.com](http://www.ijmece.com)

# Investigations Into The Use Of AI In The Diagnostic Imaging Of Disease

<sup>1</sup> S. Gouthami, <sup>2</sup> Patan Sumiya,

<sup>1</sup>Assistant Professor, Megha Institute of Engineering & Technology for Women, Ghatkesar.

<sup>2</sup> MCA Student, Megha Institute of Engineering & Technology for Women, Ghatkesar.

## Abstract-

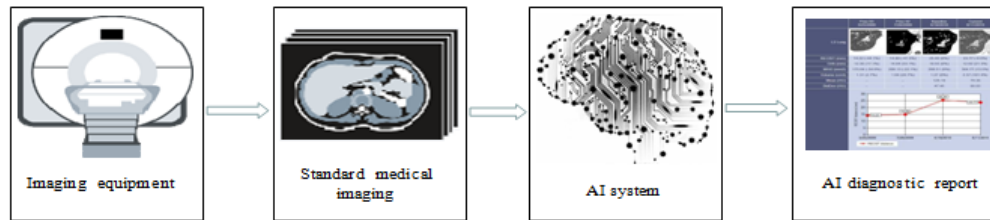
First, the article explains the current state of AI research in medical imaging diagnosis and uses examples to show how important computer-aided diagnosis is; Second, we take a close look at the technological, industrial, and application-related elements of the present development constraints of computer-aided diagnostic technology; Lastly, the study offers some recommendations for improving the use of artificial intelligence technology in medical imaging diagnostics, taking into account the present-day real-life circumstances, based on the earlier research.

## Keywords-

Computer-assisted diagnostics; medical imaging; artificial intelligence

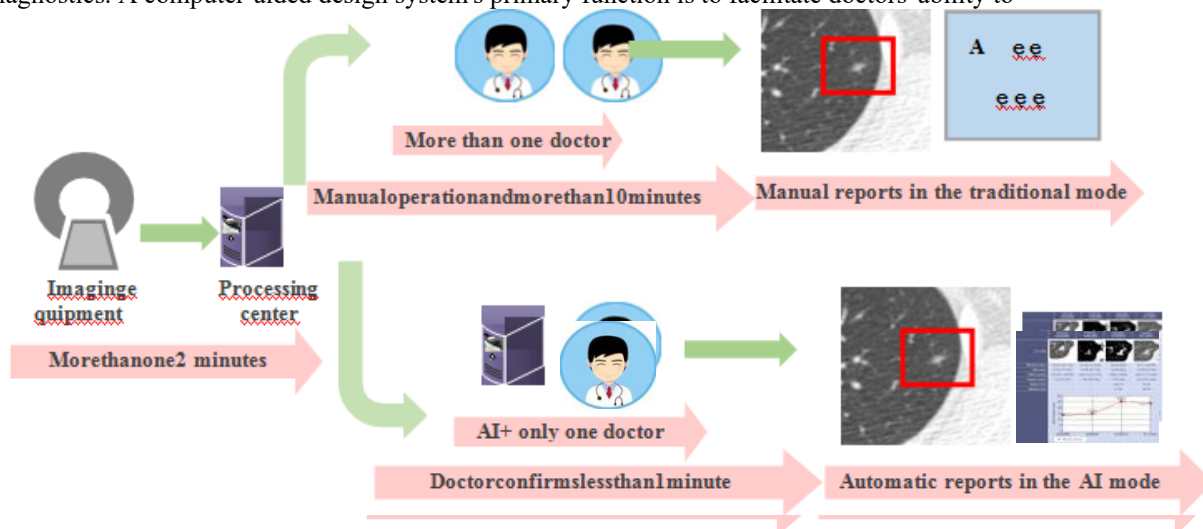
## I. INTRODUCTION

Cancer accounts for 22.32 percent of all fatalities globally, putting it in first place among the leading causes of death worldwide, along with cardiovascular disease, diabetes, and chronic respiratory illnesses (WHO, 2011). From 17.2 million in 2016 to 19.3 million in 2020 and is projected to reach 20.2 million in 2022, the number of diagnosed patients with cancer has been steadily rising, reflecting the increasing trend in cancer incidence worldwide in recent years. Even more concerning is the fact that cancer incidence in my nation has been trending downwards in age groups [2]. Disparities in economic and social development among regions are a major factor, particularly in rural areas where healthcare is less developed and more expensive. There is a high percentage of long-term death because many malignancies have been identified and treated with, but the optimal period for treatment has passed. Consequently, it is crucial to regularly examine for linked disorders in order to recognize the affected area in time and the appropriate therapy may be administered to boost the likelihood of survival. Integrating computer technology with other domains also supports societal growth in the modern era, which is happening at the same time as information technology and technical innovation are slowly but surely changing people's way of life. Various disorders have been screened for with the use of medical imaging technologies [3] in this context. Medical imaging technology has been used in over 70% of clinical diagnoses, according to statistics[4]. It primarily pertains to the technological means and procedure of directly acquiring pictures of inside tissues using non-penetrating methods for specific parts of the body. In clinical practice, it primarily serves three purposes: auxiliary medical clinical examination, injection processing method determination, and interventional therapy guidance. The former two use features like 3D visualization, while the latter two rely on features like identification and marking. The third function is to aid in evaluations and cuts during injection processing. Modern medical imaging detection techniques have become more diverse, including a wide range of imaging modalities such as computed tomography (CT), magnetic resonance imaging (MRI), computerized radiography (CR), ultrasound, endoscopy, angiography, PET-CT, etc. As a general rule, alterations in human physiology are a common byproduct of disease pathogenesis, and these alterations will manifest in various imaging findings across diagnostic tests. By analyzing this data effectively, doctors may find possible causes, track the progress of associated illnesses, and decide what to do next in terms of diagnosis and therapy.



**Figure1.** The over all process of AI –based computer –aided diagnosis.

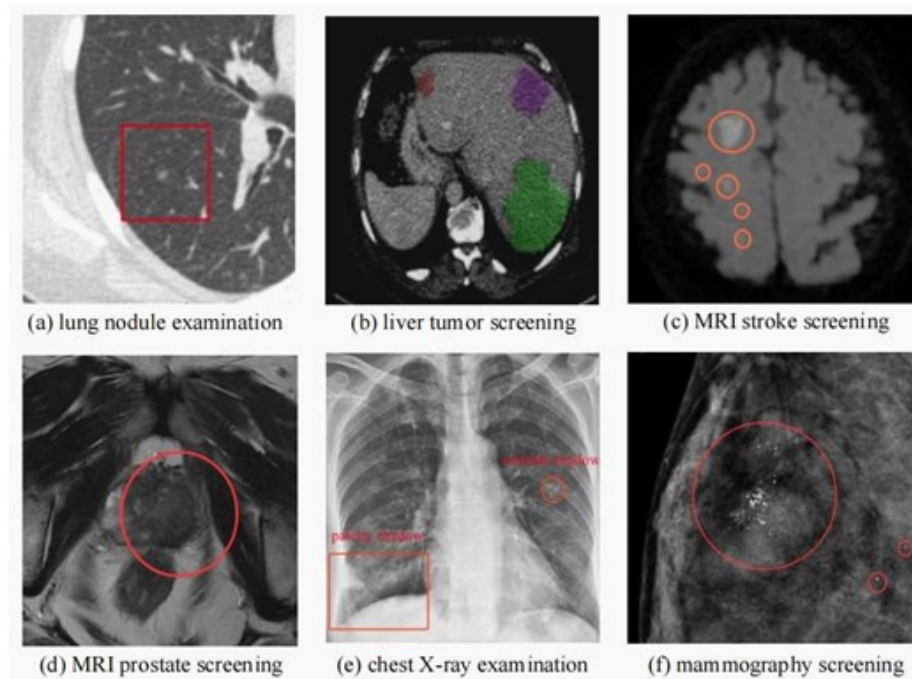
The disparity between the supply and demand for medical image analysis, however, persists and even worsens as the number of hospitalized patients rises. The efficiency of manual reading is also likely to be below average since primary-level doctors lack expertise with diagnosis and therapy. Because of this, the development of computer-aided diagnosis (CAD) [6] has the potential to speed up the process of illness diagnosis and treatment, improve access to imaging data, and facilitate specific quantitative and qualitative analyses. Improving the efficiency and accuracy of physicians' diagnoses, it has proven enormous beneficial results in intelligent medicine. The general procedure of AI-based CAD is shown in Fig. 1. A large number of researchers both domestically and internationally place a premium on the CAD system design process and work tirelessly to find ways to use this technology in clinical diagnostics. A computer-aided design system's primary function is to facilitate doctors' ability to



**Figure2.** Comparison between traditional and AI- based methods for disease diagnosis.

Artificial intelligence (AI) medical imaging systems based on deep learning [7] are able to circumvent the aforementioned drawbacks, thanks to the fast advancements in both software and hardware. Classification, segmentation, and object recognition have all been greatly improved by deep learning technology, which has recently made significant strides in computer vision. In particular, convolutional neural networks (CNNs) can precisely localize lesions and collect their details by integrating deep learning technology with medical imaging expertise, offering a quantitative diagnostic foundation for patients' future treatment plans. Through ongoing picture training, it is able to automatically extract visual characteristics, allowing for the detection and illness diagnosis of diverse medical images. Artificial intelligence (AI) can now do tasks like accurately identifying lesion areas, automatically making a clear division, and labeling lesions. The system's suitability for clinical diagnosis of illnesses is shown by the time-saving and efficient detection, which lets patients uncover their own conditions in time. Consequently, diagnostic analysis of medical images in conjunction with computer technology plays a crucial role in enhancing the precision and efficacy of lesion diagnosis. Conventional medical image analysis algorithms are still heavily employed in CAD systems today, despite the fact that they have a number of procedural and technological limitations. As an example, conventional segmentation algorithms rely on basic data and don't fully use semantic information. As a consequence, they don't effectively extract picture aspects like texture and edges, which leads to subpar diagnostic outcomes. Compared to more conventional approaches, AI-based solutions for illness detection are both quicker and more efficient (Fig. 2).medical picture categorization, segmentation, target recognition, and

retrieval; target area delineation; and 3D reconstruction. Its primary use is in the medical fields of screening, diagnosis, and therapy. Therefore, AI offers numerous advantages to medical image analysis, including the following: first, rapid image processing and analysis with timely auxiliary judgment results; second, high diagnostic sensitivity with a correspondingly low missed diagnosis rate; third, accurate data analysis to close the knowledge gap between doctors, leading to better primary disease screening. As shown in Fig. 3, the diagnosed lesion locations are represented by the color-labeled sections in the deep learning-based CAD findings for a variety of medical pictures. In particular, the screening of lung nodules, liver tumors, MRI strokes, MRI prostates, chest X-rays, and mammograms are shown by Fig.3 (a)–(f), in that order.



**Figure3.** Results of CA Dusing AI technology.

## II. CHALLENGES FACED BY AI MEDICAL IMAGING

Thanks to ongoing software and hardware optimization, deep learning technology has made remarkable strides in medical imaging diagnosis. Market investment is also continuously improving, and there is a rising number of organizations and businesses entering this industry. The following three areas, however, remain largely unbridged between the theoretical development and practical implementation of AI-based medical imaging diagnostic technologies: Getting your hands on annotated datasets to train your models is technically challenging. A large training dataset is mostly responsible for the predictive performance of medical picture diagnostic models based on deep learning, according to studies. More samples trained on a model usually means more accurate predictions. The following issues, however, make it challenging to gather high-quality datasets: When it comes to medical imaging data, there is a lack of standards and digital imaging data is hard to come by. As a second point, data set labeling is hard. The medical field is very professional, which means that not only is there a high bar for participation in labeling, but the quality of the labels themselves is uncertain. In addition, many issues in clinical medicine still lack a clear description. B. Particular criteria for judging the efficacy of AI-assisted diagnostics are lacking in the business world. To prevent misunderstanding or mixed usage of indicators, the assessment indicators for various illnesses should be distinct, and they should correspond to their own unified criteria for screening or



diagnosis. two, the judicial and regulatory framework has not yet been established in relation to registration, access, and oversight. The Chinese market is just as popular as the Western market, but the enforcement of rules and regulations is much slower. One major obstacle preventing linked businesses from achieving product optimization is the fact that, for instance, access standards and assessment methodologies for AI software products are still in the conceptualization phase. C. AI still has a ways to go before it can effectively address real-world issues with medical imaging. To get better picture features for future illness detection, deep learning has to learn a lot from a big quantity of sample data, because it has its own traits of self-learning and self-adaptation [9]. Second, bad results are the consequence of the complexity of medical image information processing, which makes it hard to construct a trustworthy model and train pictures with minor data variations and different structures and forms, including organs and dynamic images.

### III. SEVERAL SUGGESTIONS FOR AI MEDICAL IMAGING

An immediate need for smart medical care is the rapid application of deep learning-based computer-aided diagnostic technologies to clinical diagnosis [10], in reference to the extensive study mentioned before. The article then uses this information to provide situationally appropriate recommendations for the use of AI in medical imaging diagnosis: Topic A: Technical Advice It is crucial to optimize the deep learning algorithm model and create a high-quality data set. This involves doing things like: (1) conducting more research on a combined rule-and-deep learning model; (2) creating training data sets with a variety of sources and standards; and (3) making use of transfer learning to decrease annotations in model training in order to improve annotation efficiency. B. Intelligence from the AI sector The industrial sector might benefit from more research into and development of registration standards for AI software. First, relevant institutions should form a new professional team of artificial intelligence medical application researchers whose main task is to study the model evaluation framework and indicators; second, the content should center on the model's pre-, mid-, and late-stage process specifications; third, the model's performance and safety evaluations; and finally, the model's access mechanism to medical institutions. Expert alliances in imaging medicine and other similar fields should research AI for medical imaging, develop a professional consensus on the topic, and then advise businesses and consumers on medical matters. C. Tips for Users of AI Apps In order to enhance the interoperability of data output in the CAD system, it is necessary to standardize and standardize the training of digital annotation in accordance with the application standards of medical pictures in the clinical diagnostic and treatment process. When training deep learning models for medical imaging diagnosis, it is important to use the patient's symptoms, medical history, and other test findings in addition to the image signals itself. This will allow the models to make more thorough assessments. This will allow for the general medical AI level to be raised, which in turn will facilitate the shift from auxiliary examination to predictive diagnosis and auxiliary judgment.

### IV. CONCLUSION

From a technological, industrial, and practical standpoint, the article offers solutions to the problems with the present state of artificial intelligence technology in medical picture analysis that are based on deep learning. Although AI-based medical imaging computer-aided diagnosis mode can significantly lessen the workload of front-line clinical imaging doctors, there is still a long way to go before the entire disease diagnosis process is more effective and reliable. So, how can we guarantee the accuracy of lesion area detection? The primary focus of future research will be on improving the accuracy and reliability of the auxiliary diagnosis system's detection quality and providing accurate illness diagnosis information, with the ultimate goal of enhancing the whole diagnostic process.

### REFERENCES

- [1] Sung H, Ferlay J, Siegel R L, et al. Global cancer statistics 2020: GLOBOCAN estimates of incidence and mortality worldwide for 36 cancers in 185 countries[J]. CA: a cancer journal for clinicians, 2021, 71(3): 209-249.
- [2] Clinton SK, Giovannucci EL, Hursting SD. The world cancer research fund/American institute for cancer research third expert report on diet, nutrition, physical activity, and cancer: impact and future directions[J]. The Journal of nutrition, 2020, 150(4): 663-671.
- [3] Chandy A. A review on iot based medical imaging technology for healthcare applications[J]. Journal of Innovative Image Processing (JIIP), 2019, 1(01): 51-60.

- [4] Smith N B, Webb A. Introduction to medical imaging: physics, engineering and clinical applications[M]. Cambridge university press, 2010.
- [5] Kasban H, El-Bendary M A M, Salama D H. A comparative study of medical imaging techniques[J]. International Journal of Information Science and Intelligent System, 2015, 4(2): 37-58.
- [6] Doi K. Computer-aided diagnosis in medical imaging: historical review, current status and future potential[J]. Computerized medical imaging and graphics, 2007,31(4-5): 198-211.
- [7] ZhouSK, GreenspanH, DavatzikosC, et al. A review of deep learning in medical imaging: Imaging traits, technology trends, case studies with progress highlights, and future promises[J]. Proceedings of the IEEE, 2021, 109(5): 820-838.
- [8] Han C, Rundo L, Murao K, et al. Bridging the gap between AI and healthcare sides: towards developing clinically relevant AI-powered diagnosis systems[C]//IFIP International Conference on Artificial Intelligence Applications and Innovations. Springer, Cham, 2020: 320-333.
- [9] Janiesch C, Zschech P, Heinrich K. Machine learning and deep learning[J]. Electronic Markets, 2021, 31(3): 685-695.
- [10] Fujita H. AI-based computer-aided diagnosis (AI-CAD): the latest review to read first[J]. Radiological physics and technology, 2020, 13(1): 6-19.
- [11] Cho J, Lee K, Shin E, et al. How much data is needed to train a medical image deep learning system to achieve necessary high accuracy?[J]. arXiv preprint arXiv:1511.06348, 2015.