

Artificial intelligence for Breast Cancer Detection

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ABSTRACT

Millions of women globally are affected by breast cancer, which is a serious global health issue. Accurate diagnosis and early detection are essential for enhancing patient outcomes. The development of artificial intelligence (AI) has completely changed the way that breast cancer is diagnosed and treated. For the goal of diagnosing breast cancer, several AI techniques have been used, which include CAD systems, models based on deep learning and machine learning algorithms. To develop models that can precisely categorize and identify malignant lesions, separate harmless from tumors that are malignant and predict patient outcomes, these techniques make use of enormous databases of annotated pictures.AI algorithms can help with risk assessment by spotting high risk people who can benefit from specialized screenings or preventive measures. Despite these encouraging advancements, issues including quality of data, consistency, and ethical issues still exist. This research paper primarily focuses on the significance of AI in detecting breast cancer, the techniques used by AI and the fundamental ideas around it.

Keywords-Breast Cancer, Diagnosis, Artificial Intelligence, Malignant Lesions.

1.Introduction

Artificial intelligence, or AI, has become a potent tool in many industries, including healthcare. AI has demonstrated significant promise in enhancing prognosis, treatment planning, and diagnosis of breast cancer. One of the most frequent malignancies to impact women globally is breast cancer, and early identification is key to enhancing patient outcomes.

By examining many kinds of medical pictures, including MRIs, mammograms, and ultrasounds, AI can help with the detection of breast cancer. This enables radiologists to make more intelligent judgments and might result in earlier and more precise diagnoses. Oncologists can create individualized treatment programs with the aid of AI algorithms.

It's crucial to remember that, despite the optimistic improvements, AI is not intended to take the position of healthcare personnel. Instead, it is a useful tool to increase their level of knowledge and enhance patient care.

2. Objectives

Depending on the applications and research fields, multiple goals for AI in breast cancer detection may be set. However, a few typical objectives are as follows:

- i. Early Detection
- ii. Improved Accuracy
- iii. Efficient Screening
- iv. Reduced Healthcare costs
- v. Decision Support for Healthcare Professionals
- vi. Continual Learning and Improvement
- vii. Research Insights



AI in breast cancer aims to increase detection precision, enable individualized treatment plans, improve patient outcomes, and deepen our understanding of the illness for improved management and prevention.

3. Basic Concepts

Breast cancer detection and diagnosis are greatly aided by artificial intelligence (AI). Here are some fundamental AI principles utilized in breast cancer detection:

I. Machine Learning (ML): ML is a subset of AI that allows systems to learn from data without having to be programmed. In breast cancer detection, machine learning (ML) techniques are developed on massive datasets of mammograms, medical records, and other pertinent information to uncover trends and generate accurate predictions.

II.Image Recognition: Artificial intelligence (AI) algorithms for image recognition can analyses mammograms and identify probable abnormalities or worrisome regions that may signal a risk of breast cancer. These algorithms can learn from thousands of labelled mammogram pictures to enhance their precision in detecting early signs of cancer.

III. Deep Learning: Deep learning is a branch of machine learning that focuses on multi-layered artificial neural networks. Convolutional Neural Networks (CNNs) are widely employed in the identification of breast cancer. CNNs can analyze mammograms and automatically identify information like tumor shape and texture to help with diagnosis.

IV. Computer-Aided Detection (CAD): CAD systems use AI algorithms to assist physicians in reading mammograms. These technologies can discover and label areas of interest for subsequent investigation. CAD systems serve as a second set of eyes, increasing the reliability and efficiency of breast cancer detection.

V.Data Integration: AI algorithms can integrate and analyze numerous forms of patient data, such as mammograms, patient histories, genetic information, and clinical records. AI systems can give more accurate predictions and personalized risk evaluations by analyzing multiple data sources.

VI. Predictive Modelling: Using numerous risk variables and indicators, AI may create predictive models that assess the possibility of developing breast cancer. AI models that have been trained on massive datasets can detect subtle patterns and forecast the likelihood of getting breast cancer in particular patients.

VII. Decision Support Systems: AI can help healthcare workers with breast cancer diagnosis and treatment planning by providing decision support. AI systems can make recommendations on the best course of treatment for a specific patient by analyzing patient data and taking guidelines and best practices into account.

While AI has showed potential in the identification of breast cancer, it is not a replacement for clinical experience. Artificial intelligence systems should be viewed as tools to help healthcare professionals make more accurate diagnoses and treatment decisions.

4. Methodology

Algorithms based on machine learning and image processing techniques are frequently used in AI breast cancer screening strategies. An overview of the process for AI-based breast cancer diagnosis is provided below:

A significant amount of breast scan information is gathered, generally through mammograms, ultrasound scans, or MRIs. Images of both healthy and malignant patients should be included in the dataset. The preprocessed images are cleaned up to get rid of noise, artifacts, and extraneous data. Techniques for picture scaling, normalization, and enhancement may be used in this stage. The preprocessed photos are then used to extract pertinent characteristics. These features can be created manually and include traits like texture, form, or density, or they can be learned features that are retrieved utilizing architectures for deep learning like convolutional neural network models (CNNs).



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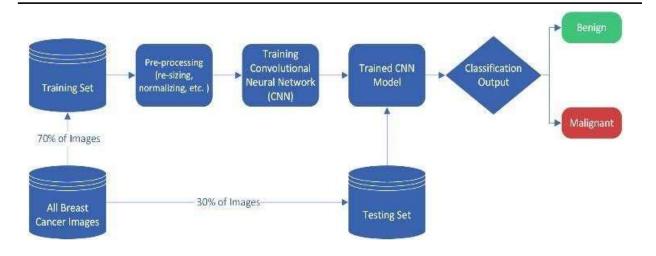


Fig 1. Processes involved in detection.

The characteristics extracted are utilized to train models for machine learning, often deep learning models like CNNs or supervised learning techniques like support vector machines (SVM) and random forests. The labeled dataset is given to the model during training, and internal parameters are changed to enhance the model's capacity for correctly identifying pictures. A different dataset that was not utilized during training is used to test the trained model to determine how well it performed. Accuracy, sensitivity, specificity, precision, and F1 score are typical assessment criteria. To enhance the performance of the model, optimization approaches like regularization or hyperparameter modification may be used. This process helps the model be improved for better results.

The model can be used for detecting breast cancer missions after being trained and tuned. The model can be provided with fresh, previously unseen breast photos, and it will produce predictions about the existence or lack of malignant tumors. To keep the deployed model updated with the most recent developments and to raise its accuracy, it can be updated from time to time with fresh data.

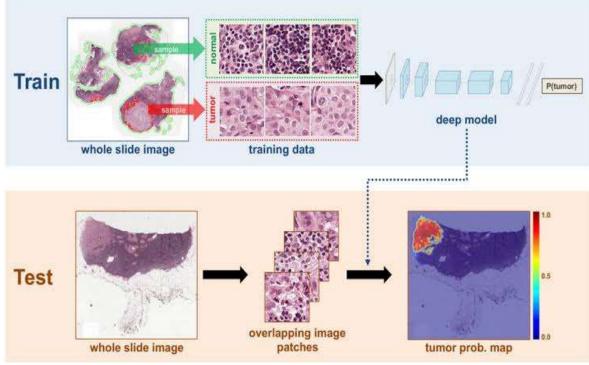


Fig 2. Understanding cancer using Machine Learning.



5. Comparison

Artificial intelligence (AI) can improve accuracy, efficiency, and scalability in breast cancer diagnosis. Here's a comparison of breast cancer detection with and without artificial intelligence: > Accuracy

With AI, medical imaging such as mammograms or MRIs may be analyzed with high precision and accuracy. AI models can acquire from enormous datasets of labelled images, allowing them to spot subtle patterns and anomalies that human observers may miss.

Without AI: Breast cancer detection is totally dependent on the competence of human radiologists. While doctors are highly skilled, medical image interpretation can be subjective, and there is the possibility of human error or oversight.

Scalability:

AI-based breast cancer detection: AI-based detection of breast cancer can be easily scaled to handle massive volumes of medical pictures. Once taught, AI algorithms can analyze several cases at the same time, allowing them to screen many patients rapidly and consistently. This scalability is particularly useful in areas where specialized healthcare practitioners are scarce.

In the absence of AI: The availability of skilled radiologists limits the capacity of breast cancer detection in the absence of AI. In locations where radiologists are in limited supply, the demand for diagnosis and screening might exceed the ability of healthcare facilities, resulting in delays in identification and treatment. Systems can swiftly screen images for potentially worrisome areas, helping radiologists to priorities cases and better manage their time.

> Efficiency:

AI-powered systems can interpret and analyze medical images far more quickly than human radiologists. This can drastically minimize the amount of time needed for the planning of treatment and diagnosis. AI systems can swiftly screen images for potentially worrisome areas, helping radiologists to priorities cases and better manage their time.

Without AI, physicians must manually study and analyze each medical image, which takes time and results in lengthier patient wait times. The absence of AI aid can limit radiology departments' throughput and efficiency.

> Helpful Resource:

In the absence of AI, radiologists must rely completely on their own knowledge and experience to interpret medical pictures. The lack of AI as a supplementary tool may increase the likelihood of human errors or oversight.

While AI has showed considerable promise in breast cancer screening, it should not be used to replace human radiologists. Instead, it can supplement their abilities, improve their performance, and lead to more accurate results.

AI can help radiologists by providing a second opinion and minimizing the probability of missed diagnosis or false positives/negatives. It can aid in the detection of minor abnormalities, early-stage malignancies, and calcifications that are difficult to detect visually. Artificial intelligence can also help with risk assessment and personalized treatment planning.

6. Limitations

While AI has shown considerable potential in the identification of breast cancer, its present application still has major drawbacks. Here are some significant restrictions:

To properly learn and generalize, AI algorithms require a huge amount of high-quality training data. It can be difficult to find large datasets with precise classifications for breast cancer because manual annotation by professionals is frequently required. Decreased accuracy and generalization of AI models can be caused by biased or insufficient training data.

In some cases, AI algorithm might yield false positives or false negatives, ignoring malignant lesions while detecting cancer where none exists. False positive results may end up in needless invasive treatments and patient concern, while false negative result can lead to missed diagnoses and postponed treatment.



Concerns about privacy of patients, informed approval and potential biases are brought up using AI in the detection of breast cancer. Given the potential repercussions of incorrect or prejudiced predictions, it is imperative to solve these issues to guarantee that AI is utilized responsibly and fairly.

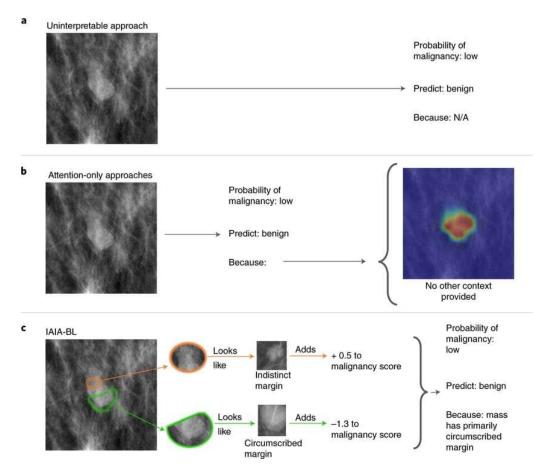


Fig 3. Most AI systems that identify pre-cancerous lesions in mammograms don't provide any mention of their decision-making method (top). If they do, its frequently a map of saliency (middle) that only instructs medical professionals as to where to look. In addition to telling doctors where it is searching, a new AI system (bottom) also explains which prior experiences it is using to form its judgements.

7. Conclusion

Artificial Intelligence has showed considerable promise in the identification of breast cancer, and it has potential to dramatically increase the precision and effectiveness of breast cancer diagnosis. The AI algorithms can discover patterns and traits related to breast cancer having a high level of accuracy since they can learn from large datasets of annotated photos. They can assist in breast cancer early identification which may result in earlier treatment and better patient outcomes. AI powered tools can save both time and resources by automating some procedures and supporting radiologists in their analysis, freeing up medical personal to concentrate on more challenging cases or patient care.

However, there are still difficulties and restrictions that need to be addressed despite the encouraging outcomes. By enhancing precision, effectiveness and early detection rates, AI has the potential to transform the detection of breast cancer. Even if there are obstacles to be addressed, continuous research and improvements in AI hold enormous promise for enhancing the diagnosis of breast cancer and ultimately saving lives.



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