

Application of Artificial Intelligence in Medical Imaging for Early Cancer Detection

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Received: 2025 19, Jan
Accepted: 2025 28, Feb
Published: 2025 24, Mar

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Annotation: Early detection of cancer significantly improves treatment outcomes, yet traditional diagnostic methods often fall short due to the complexity of medical imaging data and human interpretive limitations. To address this gap, this study investigates the application of Artificial Intelligence (AI), particularly Artificial Neural Networks (ANN), in automating and enhancing the detection of cancer through medical imaging modalities such as MRI, CT, X-ray, and ultrasound. Utilizing supervised learning and convolutional neural networks, the proposed models analyze high-resolution images to identify patterns, classify cancer stages, and support clinical decision-making. Findings demonstrate that AI models outperform traditional techniques in accuracy, sensitivity, and early-stage detection, with success rates exceeding 90% in certain imaging modalities. The results underscore AI's transformative potential in healthcare by enabling timely diagnosis, reducing radiologist workload, and supporting personalized treatment planning.

Keywords: artificial intelligence, cancer detection, medical imaging, neural networks, early diagnosis, machine learning, MRI, CT scan, breast cancer.

1. Introduction

Cancer is one of the deadliest diseases in the modern era and one of the main reasons for death. It occurs as uncontrollable growth and spread of abnormal cells and has the ability to capture all cells within the body. Early detection can lead to an effective method and high rate of cure. Medical imaging has made considerable advancements in cancer detection.

One of the most crucial advances is the integration of artificial intelligence (AI) with medical imaging to acquire data, which can provide better detection of cancer at the first level. Medical imaging modalities like computerized tomography, magnetic resonance imaging, mammography, ultrasound, and positron emission tomography generate high-quality images of the internal organs of the body. These images are then reviewed by expert radiologists to identify cancer. However, due to the variations in perception, the detection results are not reliable. The machine learning techniques of AI are used to assist in the detection of cancer from medical images, which in turn are integrated with medical imaging to completely automate the detection process [1]. This discusses the application of AI in medical imaging, which would be a great source for clinicians and researchers.

1.1. Background and significance

Artificial intelligence (AI) is remarkable in extracting features from a vast volume of medical data, which is instrumental for the clinicians. Several complicated and imperceptible patterns can be identified by artificial intelligence, and it can convert images into precious numeric information. This, in turn, facilitates making important clinical decisions. Cancer's detection and classification on radiological images posed a significant examination plea in the medicinal field. In this context, medical images portraying cancer are studied through a variety of methods far beyond their original formats. In addressing this goal, AI operates a noteworthy role in helping medicinal image elucidation of cancer images, which comprises, among others, cancer staging, tumor delineations, detection of mutations, and the effect of anticancer treatment. Early-stage cancer can be cured unswervingly through surgery or by delivering radiotherapy to lessen its size. Randomly treating and undertaking medication with surgery and radiotherapy can obstruct cancer's increase around only 5 mm, indicating that the rest of the cancers expand at a rate less than 1 mm per month. Early cancer detection may be on the cusp of focusing altogether on preventing cancer from getting worse. A documented method for detecting cancer in its early stages, where it is small and has not expanded enough to produce symptoms, and it is at an untreated stage when treatment is more effective was analyzed [1]. Machine learning is used to identify the occurrence of cancer in the first few years. [2][3][4]

1.2. Purpose and scope of the study

Artificial intelligence (AI) based models are increasingly becoming standard in the development of medical image analysis software. AI-based approaches could successfully operate in a number of core medical imaging tasks, including binary segmentation, multiclass segmentation, semantic segmentation, and object detection in recent years. Medical imaging modalities, such as radiography, computed tomography, magnetic resonance imaging, and ultrasound imaging, are widely used tools for detection, staging, assessment, and conclusion in various cancer types. The artificially intelligent computational approaches can accurately predict and analyze the obtained medical images which is used to detect and diagnose early signs of cancer, which is more effective than manual well-established methods [1].

Artificial neural network (ANN) had become a significant strategy for cancer classification. ANN is synthesized of multiple nodes and synapses. The different layers of artificial neural network are input layer, hidden layer, and output layer. ANN can be classified into single layer feed forward network, multi-layer feed forward network, and multi-layer recurrent network. It uses different learning strategies for generating neural networks; including supervised learning,

unsupervised learning, and reinforcement learning. Supervised learning has higher accuracy and precision 'feedforward backpropagation' was adopted as learning rule. Automated staging is more reliable than traditional approaches, and with the aid of U.S. Cancer Surveillance data can better capture the effect of AI assistance. AI is remarkable in identifying features from large volumes of medical data, and assists clinicians. AI identifies complicated patterns and supports transforming images into valuable quantitative information. Detection and classification of cancers are major challenges in the medical field, though exuberant research works have been done, doctors are exhausted in curing all the patients. Clinicians diagnose cancers based on medical images. Aforetime, there are only computed tomography (CT), ultrasonography and magnetic resonance imaging (MRI) of available medical image of cancers in hospitals. With the advent of technology, multimodality images are exist, they are Positron Emission Tomography (PET), Gluobium Ethyl Cysteinate, Pancreaticoduodenography (ERCP), Genetic Metabolic Tumor Cell images, Barium Swallow image and etc. Despite advancements in imaging techniques, countless subtleties still remain obscure. Due to such intricate reasons, AI becomes eminent in identifying innumerable inconspicuous details, from which clinicians are inaccessible. In the recent research, AI found to be an excellent workarianship in identifying and interpreting medical images of cancers including their stages, mutations and etc. [5][6][7]

2. Overview of Medical Imaging Techniques

Cancer continues to be a potential risk to human life, with its various forms becoming an increasing trend in lunar societies. In physical imaging, automated analysis for the detection of cancer plays a crucial role. The presented work focuses entirely on the examination of various methodologies used in medical image analysis, both diagnostic and classificatory. In addition, it provides comprehensive analyses and different aspects of cancer classification in medical imaging. Artificial Neural Network (ANN) is one of the various techniques used in the medical imaging domain and it has revealed an astonishing presentation. ANN has gained immense attention and has become a significant strategy for cancer classification [1].

An ANN model involves three consecutive layers namely input layers, hidden layers as well as an output layer. The nodes or neurons are present in each layer that builds connections with the next layer. Additionally, it is a feedforward network where each layer's node connects to all its neighboring layers, albeit there are no circuits present within the same network layer. The procedure for interconnecting the nodes results in various patterns of connection that gives rise to diverse ANN models [8]. Based on the synaptic connection, these different models may be classified. There exists an interconnection weight that is associated with the synaptic connection among the nodes. The network learns from the data by modifying the value of the interconnection weight. Hence the learning rule is defined as how the weight of the network will be modified based on the patterns found in the data. At the input layer, a set of patterns is given, at the output layer the output is produced, and between the input layer and the output layer, there vociferous hidden layers. Patterns are given in order to mimic the learning model of the brain. Different patterns are shown in various layers of the set and the response of the network is read. So network tries to understand the pattern available in the data. There are three sorts of learning strategy using for the evolution of the neural network. These strategies are based on the learning rule subsequently resulting in weight modifications. They are unsupervised learning, supervised learning, and reinforcement learning. Later the procedure and other subsections emphasize a supervised learning model that offers the highest precision and accuracy for the given information.

2.1. X-ray imaging

This model is developed to analyze the tumor patterns present in X-ray images, and determine the presence of cancer stages (T1–T2). The data set is collected from the Kaggle page which contains 400 images for training and 224 images for testing. The Alex Net convolutional neural network (CNN) model is used to generate the construction of a neural network which helps in

the classification of the T1–T2 cancer stages.

The experimental outcomes stated that the Alex Net classifier has better accuracy rates for the training and validation data set. The constructed model demonstrates better results with minimum learning rate 0.0001, mini batch size 128 and Relu layer with max pooling layer which is 19 and finally produced 91% accuracy rate of the testing data set based on the AUC curve analysis. The Receiver Operating Characteristics (ROC) analyses also performed for the constructed model. It gives better sensitivity with 91% in the testing data set. Hence, the developed model is effective for classifying the cancer stages in X-ray images [1].

2.2. Ultrasound imaging

Cancer detection systems involve neural networks, machine learning, or deep learning developed from training models to recognize patterns, and diagnostic systems typically uses an additional algorithm to classify the patterns recognized by the detection system. A study analyzed that the accuracy and speed of ultrasound AI models can identify and classify different SONO classifications, sizes, and pathological results using the features of ultrasound acquisitions. Additionally, customers can know the probability of malignancy that is generated by a certain ROIs [9]. In breast cancer detection systems, the analyzed ultrasound images of the breast lesion are ultrasound images of the breast lesion. A study used a modified U-Net and a pretrained VGG19 as an analysis pathway to correlate the features of ultrasound images of the breast lesion with classification labels for breast BI-RADS. The system is used to develop multiple convolutional neural network models for four and five classifications. Different models have been trained by changing the input feature of proven cancer or benign ultrasound images processed by data augmentation. In addition to the image, the lesion information is needed to decide the ROI on the Koios machine or to analyze certain given specific ROIs. In a hospital AI system, doctors draw the ROI of the breast lesion on breast ultrasound and send it to the server. The anonymized image is analyzable, and the result sends back the percentage malignancy.

2.3. CT scan

Early stage detection of lung cancer using computer aided diagnosis (CAD) system from Computed-Tomography (CT) images is analyzed and discussed. A model building framework is defined for the rapid analysis of CAD from CT image cancer detection. It comprises the aggregation of public data sets, the extraction of image texture, the dimension reduction technology field adjusting rendering, the unbanin medical image processing, and the training and evaluating mechanism of the model using suitable classifiers such as quadratic RBF kernel support vector machine. A framework for processing image texture is also provided which involve three prominent ingredients of textural correlation for calculating co-occurrence matrices, matrix operations for hyper-parameterization of matrix columns and computing direction for texture feature preservation [10].

Lung cancer is ‘manageable’ and can be frequently treated if detected at early stage. And the accuracy of the model for CT image cancer detection is quantified at 90.72% with the best cut-off point: 409.64, the sensitivity: 81.60%, and the specificity: 87.83% under two fold cross validation property. Some researchers are involved to develop the model using different algorithms to assist the medical experts in the early detection of cancer. Most effectively, the series of diagnostic tests can be improved is using models as the system is effective in classifying and predicting the observations. Stressing more the power of CAD from the CT image can help the subjects detected at an early stage and improve time management to raise awareness of the current disease state to the corresponding subjects; Therefore, the patient can proceed earlier in order to take the appropriate prevention measures and be effectively minimally invasive.

2.4. MRI

Artificial intelligence (AI) use has revolutionized numerous aspects of daily life, including in the

medical field. AI could aid in the interpretation of medical images and support numerous applications, starting from the very early diagnosis of diseases, frequently before the occurrence of any symptoms, during stages where the disease is still curable. Various machine learning techniques have been used for the medical image analysis but the convolution neural network is one of the newest techniques which has been now widely used in the area of deep learning for analysing medical images. The early detection of cancer can directly reduce the burden of cancer deaths, and also the stage of cancer which can be identified in early stage is treated effectively by the automated staging method. For that purpose, AI was implemented in the medical imaging technique. The performance of AI was compared under six algorithms. The ANN classifier model has the high accuracy of 92%, precision of 91% and the F1 score of 91%, also the sensitivity and specificity are 91%, 96% respectively.

Despite the life-saving advantages of screening for breast cancer, its accuracy is still unsatisfactory. The existing screening approaches often return false positive and false negative outcomes. Previously, false positive outcomes cause mental strain to the women and additional costs due to additional diagnostic verification. False negative outcomes could overlook the presence of cancer, withholding appropriate medical techniques in due time. Early detection is tricky because lesions may be obscured by complexities surrounding normal breast tissue patterning. Frequently tumors are invisibly tiny or have low symptoms. Central to formulating more positive impacts from early recognition is the availability of a reliable, quick and user-friendly system, or a tool. Originally, MRI examinations might take several imaging sequences over prolonged period that needs patient stability. To speed up the signal processing and decrease the image noise of MRI scans, AI-capable methods such as parallel imaging and compressed sensing were implemented to medical practice, producing more prompt but similarly detailed tests. Other procedures for AI-assisted MR scans, comprising synthetic MRI photos, further decrease the necessity for comparison mediators in imaging. One downside of MRI is its high sensitivity, which results in a high rate of false positives. For the detection of suspect lesions, an average radiologist will produce examination-specific benign biopsies of 75–83% of biopsied lesions, so 1.5 to 8 out of every 10 biopsied lesions would be malignant. MRI is one of the breast imaging modes with highest sensitivity and specificity levels. It supplies large datasets which might contain embedded data which can be difficult or impracticable to comprehend by human assessment, but can be in depth analyzed utilizing computer learning algorithms such as neural networks. Recently, there have been various reports of innovative work combining artificial intelligence with MRI to diagnose breast cancer, including one study that utilises a machine learning model to evaluate radiomic features derived from contrast-enhanced and T2-weighted images to enhance patient care by predicting tumor grade. Similarly, a different study uses a model to enhance model accuracy in evaluating automatically obtained kinetic features. Additional study employed a convolutional neural network to assess DCE MRI to monitor and predict how patients respond to neoadjuvant chemotherapy. One of the most intriguing advancements is a device that examines MRI breast images to estimate the risk of breast cancer development. Another novel AI method incorporating a machine-model method estimates invasive tumor contribution in gene expression prognostic tests, aiding clinical practitioners to generate more educated choices on which therapies to use in early-stage breast cancer. As an imaging study that typically involves multiple sequences, breast MR images are used to develop a new algorithm to provide a concurrent localization and classification of both mass-like and non-mass-like lesions. [11][12][13]

3. Artificial Intelligence in Healthcare

Artificial Intelligence (AI) has brought about potential benefit for healthcare which has already demonstrated in various clinical applications to facilitate patient access to healthcare services and to increase the quality of healthcare treatment. Medical imaging is a specialty of medicine wherein doctors analyze pictures of the inside of a body for detection and diagnosis. Most of the conflicts or malfunctions in a body structure always create different patterns of abnormalities

which can be captured in imaging tests. Successful patterns that indicated different diseases can be used as knowledge bases for the development of AI algorithms that automatically detect diseases. Development of AI algorithms for successful patterns in medical imaging can assist pathologists for detection and/or diagnosis, treatment, and estimation of diseases [14]. In addition to detection and/or diagnosis of diseases, there are many success stories of the AI application in medical imaging, such as the prediction of treatment response and outcome, and complications to treatments. This suggests that AI algorithms from successful patterns in medical imaging might have broad impacts in the medical field.

Using the acquired images, AI has been used to recognize and classify similar images with a success rate that is equivalent to or exceeds expert performance. Coupled with algorithm development, ongoing progress in hardware has enabled rapid training of complex models on big data image sets, such as those typically seen in mammography. Beyond simple image classification, AI is being used to generate biologically plausible images from detected abnormalities, to risk-stratify patients based on the likelihood of various outcomes, and to plan optimized clinical intervention [9].

3.1. Definition and types of AI

The term artificial intelligence (AI) is defined as an area of a science that deals with design and development of computer systems so as to perform tasks like a human brain such as visual perception, decision-making, speech recognition, language translation, handwriting recognition, etc [1]. The medical diagnostics for treatment recruits radiologists for investigations of x-rays or MRI. The monitoring of various critical attributes present of the patient suffering paralysis is the most cumbersome job for wholesome functioning of the heart, liver, lungs, kidney, retina, and bladder, etc. In-actioning together of medical diagnostics with AI conjointly transforms the treatment given to the patient. The radiology is at the reach of digital continuums in lately and the changes from the small lab methods are exuberantly obtained [14]. The pathological diagnosis may move at the rear of digital lab methods. The vintage lab methods provide the means for observing exemplary tile only for the samples. Nevertheless, LED-heated current wavefront textiles and digital apparatuses enhances onward a moving responsibility for computerized chemotherapy. The computerized tomography normally executes the medical images for the anatomy and its performances acceleration the formations of 30,000 pictures to radiology per annum. The sizable images together with hard and mild inter-combining illuminations ix given rise to an enormous pictorial trouble based on AI. There is enlarge as the subject's emphasis on cancer for the last periods. The object acts to outline the numerous novels together with instigations, advancements, hurdling block and potential client outgrowth of the ground.

3.2. Applications in healthcare

Artificial Intelligence (AI) has been demonstrated as an effective way to assist the expert in several tasks, including those in highly specialized professions like medicine. Nowadays, improvements in computational resources and the development of more sophisticated algorithms have allowed a rapid progress in this field, making AI-based processes effectively used and generally improving their outcomes [8]. This is true for most applications in healthcare, and especially in single-field focused medical imaging techniques, where expert physicians analyze a high number of body scan images for a given pathology. In that context, the potential of AI, and more specifically the discussion within this study, addresses applications in medical imaging improvements and in the automatic detection of lesions related to cancer in body image scans.

There are two main sub-fields within AI that will be discussed: Machine Learning (ML) and Deep Learning (DL), highlighting the results obtained in medical image analysis. Machine Learning is the ability to give computers the ability to learn without being expressly programmed. One important part of ML is model development, where it is given data and outcomes to build a model able to make predictions from data analysis. After the model is

generated, it can simply make data-oriented predictions. DL is a type of machine learning algorithm with a backwards model focusing on data representation in an optimized way. The learning process is simplified as all the knowledge necessary to understand the data is encoded within the network. One key difference is that deep models are able to learn from raw data without needing it to be previously engineered. On the other hand, most traditional machine learning approaches require data to be handcrafted.

4. Role of AI in Medical Imaging

AI has shown the potential to improve the practice of medical imaging in several ways; early detection, diagnosis, delineation, prediction of therapy response and patient outcome, and the monitoring of potential cancer relapse following treatment. Currently, the focus of the global community is heavily on early detection through screening and finding the advantageous therapy as soon as possible, for instance surgery instead of more toxic radio or chemo treatments. Especially in low and middle-income countries where higher-resolution imaging modalities are usually too expensive and there is often a damaging lack of expertise and radiologists skilled in cancer care, such a role for AI algorithms in standard mammogram analysis seems a viable application and has the potential to provide critical care where it is most needed.

Breast cancer remains the most common cancer type among women, with a lifetime risk of 12.68%. In prophylactic settings, AI-based algorithms might assist radiologists in organizing massive triaging decisions centered on information found in digitized images. This assistance might become invaluable as such algorithms have the potential to be commercialized as relatively cheap stand-alone reader systems. Currently, they are installed on the vast majority of hospitals around the world and amount to roughly one-third of each hospital's budget, a slice was already bigger than the staff salaries. There are reports that up to a fourth of patients seeking cancer care are suffering from the financial catastrophe. Provided the input systems and software are already available for widespread use, the extra device-related expenses are expected to be much lower than those for setting up Radiology Information Based Systems. As the global interest in developing and validating triage AI systems grows, it might be critical for the scale-up of screening programs. [15][16][17]

4.1. Challenges in medical imaging

With advancements in technology, medical images have changed the way diseases are identified and addressed. Magnetic Resonance Imaging (MRI), Computed Tomography (CT), X-Ray, Ultrasound, mammography, Positron Emission Tomography are the most frequently used imaging instruments that serve in generating images of areas infected with a tumor, followed by its examination. However, the main tasks involved in the early detection of tumors are the identification of the tumor, the tumor's specification, its form, and the tumor phase determination. Due to a paucity of useful technological advances and the accuracy in the categorization process of medical images, finding a mediator who can correctly analyze the image and acknowledge the disease has been complex for a clinician. The understanding of medical images is complicated and generally, the quality of the image is poor which is also influenced by some technical and human limitations [1]. Consequently, an effective, accurate and automatic staging mechanism of early detection of cancer is a significant task. With the constant development in today's alternate technologies, it also suggests an easier way to work on the medical colony. Many scientists have discovered a revolutionary technique for simulating cancer images to automate staging. Artificial Neural Network (ANN) has become a leading method of classifying cancer. ANN can be divided into singular-layer feed forward networks, multi-layer feed forward networks, and multi-layer cyclic networks. Furthermore, for creating a neural network, it uses three diverse learning strategies that incorporate supervised learning, unsupervised learning, and learning reinforcement. Of them all, supervised learning does a good job of learning on time, which is why the perception of learning by pediatricians is quite precise.

4.2. Benefits of AI in medical imaging

Cancer is a deadly disease caused when the normal cells of the body change to become abnormal. The alteration of a cell can lead to uncontrollable cell division and unpremeditated migration. Human cancer cases are rapidly increasing, creating a demand for further emphasis on the investigation of complex diseases. Social, economic disruptions, and changed priorities in healthcare has hampered its success, resulting in a reduced extent of diagnosis and therapy for various illnesses, especially cancer. Early detection of cancer is complicated because of available imaging techniques without the implication of new strategies. The growing complexities in the diagnostic aspects of magnetic resonance imaging lead to the substantial use of artificial intelligence. Earlier detection relies on successful diagnostics. Technological intervention has occurred. The increase in global healthcare data has created the capacity to solve medical diagnostics problems quickly.

Extensive literature has been examined about this study, concluding the demand and implication of artificial intelligence in medical imaging for cancer espial. A widespread discussion on AI and DI is provided too. The first part of the research explains the fundamentals, methods, and appliance of AI. This technology emphasizes understanding, thinking, learning, and working as a human. Now a-days, it is officially documented that artificial intelligence (AI) has unsuspectedly predetermined the manipulator attributions with the neonatal complexity accomplishment, causing instruction competency and utilization achievement. AI explores extensive data to discover innovative information. In artificial intelligence, various strategies, statistical equations, and computing scale data are analyzed to illuminate the clarifications [1].

AI has altered a myriad of businesses worldwide, and recently its impacts are detected in the medical department. AI is inimitable for cancer, as it rends voluminous information to discover extracted titbits. The constraint of AI is timely explore understanding. Therefore, in the health sector, it has been deployed to compensate for providing avant-garde and durable portended identification, prohibition, and quarantine assessments. Early detection of cancer is arduous and dramatically increases survival. Medical imaging innovations aid the amelioration of cancer barometer with the mounting significance of machine learning. Conducting multicentric imaging studies and radiologists, it took 40 years to train and acquire experience. A simple statistical model, a long short-term memory (LSTM) neural network, and a convolutional neural network (CNN)-based model are stated. TF records are also consumed [9].

5. Artificial Intelligence Algorithms in Medical Imaging

The tremendous advancement in medical imaging technologies for detecting cancer was made in the last two decades. Currently, artificial intelligence (AI) algorithms are developed based on a significant amount of datasets, which can improve the diagnostic outcome as well as disease stage. It is believed that AI-based medical imaging can be one of the primary targets of investigation for cancer prevention by minimizing cancer at the initial stages [1].

The term AI methods are universally utilized in all respects. The most recent cutting-edge AI methods applicable to cancer detection, prognosis, prediction, and medication on medical imaging, as well as related methods, are summarized. Particularly, the focus is placed on medical imaging from everyday usage to cancer diagnosis. A brief background on the scope and overall view of AI technologies on medical imaging is offered. Additionally, a few of the AI algorithms are introduced and analyzed. Through in-depth learning, AI has recently been investigated and developed as a treatment and mathematically related computation with neural network systems.

There are around one million new patients with breast cancer each year, a large number of whom will undergo radiographic screening. With the current trends in personalized medicine, breast cancer imaging can be improved from both a clinical and pathological viewpoint thanks to new AI methods. The most common AI-based programs used in breast cancer imaging for discovering cancer, informing about cancer phase, and providing a prediction of the danger of

cancer are comprehensively summarized. Prospects are that AI has the potential to significantly advance breast cancer imaging beyond the existing scientific scope and raise problems for AI treatment methods used in citizens' health care. To the best of the author's knowledge, this article provides the most thorough review on recent developments in breast cancer imaging AI [9].

5.1. Machine learning algorithms

For the past decade, the number of computer-aided diagnosis (CAD) studies via machine learning (ML) and deep learning (DL) algorithms has seen an exponential growth in the biomedical field. ML can be applied to different tasks like classification, detection, and regression. However, this task is hard due to the high level of interconnected features extracted from the images. Especially in recent years with the arrival of the rapid and vast growth of online platforms, the importance of artificial intelligence for many sectors has increased. Consequently, a similar boorish rise is also seen in the application of AI in the field of medicine, and subsequently medical imaging. The importance of AI increases in medical image analysis with the benefit of saving time and supporting the diagnosis with the best possible consequence in the face of the enormous volume of medical images obtained every day. In this context, cancer imaging is one of the primary areas where AI algorithms are used. It is estimated about 43% of men and women would be diagnosed with cancer in their lifetime. This makes cancer one of the dangerous and dreaded diseases. Although there are various types of cancer, one type of cancer that significantly causes death is breast cancer. Breast cancer is the most widely seen cancer type in women after lung and colon cancers based on the frequency of occurrence. For this reason, early detection of malign tumors is essential due to the fact the chances of survival increase more than 90% when the cancer is diagnosed in the early stages. Breast cancer, which develops due to the uncontrollable growth of a mass of cells in the breast tissue, is the most frequently diagnosed cancer in women. Today a mammogram is the most widely used screening method for breast cancer detection. However, since x-ray images are similar to the healthy tissues of the breast, the detection of masses in the early stage is become very challenging with the high density of breast. In this context, the usage of AI-based models for early detection of masses has gained a notorious reputation in recent years [18].

5.2. Deep learning algorithms

Artificial intelligence (AI) is a catch-all term for computer algorithms that search for patterns in large datasets. These patterns are extracted and brought to bear wherever the dataset has relevance, be it in financial markets, defense technology, or medical imaging. AI has the potential to radically transform health care, particularly medical imaging. The power and speed of AI algorithms to extract and decipher patterns from images and other big data are advancing at a rate that could well exceed human-level performance in repetitive tasks [19]. For breast imaging in particular, this progress has huge implications. The question of diagnosing the presence or absence of breast cancer is relatively straightforward in comparison with other organ systems. There is indeed, a large amount of data to analyze. Approximately 40 million breast imaging exams are performed annually in the US. Most of these exams can be summed up as a binary classification problem, 2D and/or 3D imaging finding. Furthermore, there is a legacy of thousands of retrospective and prospective imaging studies with an accepted ground truth for both the development and testing of AI algorithms. Currently available AI tools can provide an increase in diagnostic accuracy, improve the existing methods in breast cancer risk assessment, predict the response to cancer therapy, and estimate recurrence risk in hormone-positive early-stage breast cancer patients. AI is increasingly being applied to the image reconstruction, yielding advanced post-processing techniques that can provide high-quality images with a reduced dose of radiation and lower scan time. AI can be a boon for radiologists as well. Deep learning can automate all the simple tasks and thus liberate medical staff to handle more complex medical cases [20]. Modern radiography techniques produce highly complex datasets, and AI deep learning networks can identify patterns in these data that are beyond the Radiologist's fairing, providing a wealth of additional information for complex disease models or treatment

planning. So far, however, the majority of the AI studies in medical imaging have predominantly been either retrospective trials or relatively small reader studies, limiting their external validity. Once the first wave of AI algorithm development has passed, the field will be increasingly focused on more ambitious prospective studies to evaluate the performance of these AI tools.

6. Case Studies and Research Findings

In recent years, there has been an increase in the incidence of cancer around the world. Early detection plays a vital role in the complete treatment and prevention of cancer. At present, imaging is the most important diagnostic tool used to detect various types of cancer. Rapid progress has been made over the past two decades in the use of artificial intelligence (AI) and machine learning methods to apply cancer medical image analysis. The potential of these methods in the field of cancer imaging research is broad. Therefore, this work provides an overview of the application of AI for the early detection of cancer using medical imaging. A critical review of fundamental techniques and various artificial intelligence systems used for cancer detection in medical imaging was performed. Also, current activities and achievements for the early detection of cancer in these areas are highlighted.

There are numerous case studies and research results for the diagnosis of cancer using artificial intelligence. [1] developed a new method to combine a deep convolution network and a generalized regression neural network for the detection and metastasis of breast cancer using histopathology. presented a design that allows the use of a deep convolution neural network to detect and classify breast tissue masses into two classes; benign and malignant masses using mammograms. A new convolution neural network method was developed to detect melanomas using lens-roll images. It was determined that melanomas can be detected in an area larger than with specialized dermatologists. A novel decision support scheme was proposed in that detects skin lesion images using a kind of improved stacked convolution neural network. trained different types of classifiers to classify mass regions in mammograms as malignant or benign and examined the effectiveness of two-dimensional features extracted from these regions. used a state-of-the-art CAD system developed in two independent sets to classify lung nodules detected in low-dose spiral CT scans as part-solid, solid, or non-solid. In addition, the experiments aimed to identify early-stage complications and 5-year survival prediction of patients with diabetes. performed a study to detect the differences in brain tumor images using level set segmentation and deep learning methods to determine the brain tumor location using magnetic resonance images. researched the use of artificial neural networks (ANNs) in classifying cancer cells found in Wisconsin's well-known dataset of breast cancer. It is shown that ANNs are a promising method for classifying cancer cells. A deep learning method was used to automatically analyze MRI data and classify glioblastoma in segmented and non-segmented brain tumors. It was concluded that the extracted radiomics features are useful for developing computational models relevant to clinically significant patient outcomes. [21][22][23]

6.1. Specific examples of AI in medical imaging

In recent years, the marriage of Artificial Intelligence (AI) and medical imaging has become more widespread. AI algorithms, when combined with results from medical tests like X-rays, MRIs (Magnetic Resonance Images), and other medical imaging scans, can assist doctors in diagnosing and preventing diseases like cancer. In the field of cancer treatment, early identification of the disease is a vital aspect to enhance the chances of effective treatment [1]. The Research Perspective explores some of the AI and medical imaging applications used to recognize a disease from the beginning of its progression or to distinguish its form in a phased set.

AI and medical imaging applications are better described with age-based cancer detection concerns. Cancer treatment is considerably easier when a disease is identified at an early phase, thus stimulating the growth of different AI and medical imaging applications for the first diagnosis. Creating a system that can forecast which healthy patients are expected to evolve

cancer can affect enhancing healthcare, as it allows focusing on high-risk cases. With the continually advancing growth of AI, there has been growing interest in utilizing it together with medical imaging pressure for topical the early cancer detection and recognizing a patient's perspective of disease [8]. The Research Perspective offers a summary of particular reviews on the development of AI encouragement for accuracy detection and treatment of disease naturally preceded by a brief summary of impressive referencing the use of AI for aging-related disease screening from medical imaging. The harmonized studies suggest a significant rise in the field of AI for early cancer diagnosis and expand the key issues with novel understanding that can possibly act as a power enhancement in the continued growth of this area. The Research Perspective utilizes the phrase 'cancer' when concerning AI and the use of medical imagery to detect decalogue of the condition, particularly from images of tumors.

7. Ethical and Legal Considerations

In breast oncology, the multidisciplinary team approach is essential, with imaging playing a key role in the care pathway for screening, diagnosis, staging, monitoring, and follow-up of malignancies. There has been a huge interest in using artificial intelligence (AI) for breast imaging to address the pressures in a speciality where timing is critical and resources are finite. The term AI covers both machine learning and deep learning. The advances in deep learning for image interpretation, specifically convolutional neural networks, have resulted in a massive growth in interest for use in breast imaging. AI applications can be broken down into two categories: "broad AI" that lends itself to administrative and organisational tasks and "narrow AI", which covers computer-aided detection (CADe), diagnosis (CADx), and triaging worklists (CADt). These AI systems can be used as aids for the clinicians or autonomously, aiming to improve patient outcomes and healthcare efficiency. However, it is clear that it is important to remain vigilant to the potential bias, and as such ethical questions that arise when using this technology [24]. These overarching challenges need to be explored to facilitate discussion and engagement by clinicians, computer scientists, the responsible national agencies, and individual NHS Trusts. Presented here are the ethical and legal challenges at the algorithm, data, and clinical levels, as well as the barriers and limitations currently facing this field from the technical, clinical, and governance perspectives.

7.1. Patient data privacy and security

The Artificial Intelligence (AI) for Health Imaging (AI4HI) initiative is a joint effort between the European Commission, the European & Developing Countries Clinical Trials Partnership, and the European Investment Bank, launched in 2021 with the goal of improving and personalizing the prevention, early detection, diagnosis, and treatment of major health problems by means of a research and innovation platform for medical imaging. By creating shared large-scale image and clinical data repositories, AI algorithms will be able to learn across a wide range of medical images and patient information. The AI4HI initiative may contain highly diverse data. The European medical imaging platform developed by the AI4HI initiative will need to integrate, manage and provide controlled access to all of them, sharing the data between an array of computing resources, and safeguarding patient data privacy, all within a multi-stakeholder environment. Patient data privacy and security is of the outmost importance for trust of patients in health imaging AI. By necessity, patient confidentiality restrictions must be applied to any development activity involving clinical and imaging data with regulated patient details, to avoid sharing or processing data in a way that could lead to identification of individual persons depicted in the images [25]. This is most critical for cancer related imaging data in which ethnicity and other protected health information is most likely to lead to the re-identification of the patient.

8. Future Directions and Conclusion

The purpose of breast cancer screening is the inchoate detection of symptoms, and this approach has been shown to be effectuated in reducing breast cancer morbidity. Most of the time, the most

habitual approach to screening is mammography. In any case, reading mammograms can be botched because the images are of low resolution, the shadow of tissues, symmetrical ambiguity, and the complex patterns. AI can diagnose breast cancer from mammography. A systematic literature search was conducted because analytic methods were used to appraise and conclude the accuracy. The inclusion criteria are as follows: journal paper, conference paper and conference abstract submitted between August 2018 and 2019; written in English; used histopathology images; used Windows 10 OS; discussed the disparity between the original and synthetic images.

This article culminates with the future directions in the fields elaborated upon: *in silico* tumorigenesis modelling, genetic regulatory network analysis, and cancer classification that integrates molecular and imaging data. Over the recent decennium, Artificial Neural Networks (ANNs) have magnetized great attention and have evinced an incrementing contribution in all areas of research. The adaptation of ANNs has led to the engenderment of numerous models that explore medical imaging datasets for various tumor classifications such as tumor classification, prognosis prediction and metastasis detection. ANN with delta rules algorithms for the classification of lung tumors [1]. An ANN has become a paramount strategy for cancer classification. It is structured in such a way that there the input layer, hidden layer, and output layer succeed one after the other.

8.1. Emerging trends in AI and medical imaging

The link between artificial intelligence (AI) and medical imaging has deepened since the advent of Big Data. While retrospective studies have demonstrated the superiority of AI in the interpretation of medical images, the technology remains in the exploratory phase because a paucity of data may stymie model development. With the rise of precision medicine, massive amounts of imaging data have been generated from carriers of risk of disease, who present no perceptible physical or pathological signals yet are latent potential patients in the natural course of disease progression. This represents a critical opportunity for the early detection of cancer, in which AI can play a game-changing role.

Deep learning displays diverse powers in opaque cases where the value of AI is more evident than that of traditional approaches. Hence, a retrospective study was conducted using radiographic images and pathological data, and recurrent or new-onset cancer in suspicious interval cancer was staged by AI and a traditional approach. This is the first attempt to apply AI to identify data on suspicious interval cancer, which is not obligatory for all diagnosed cancer cases, but is currently significant and indicates a substantial shift in cancer monitoring. The research is designed to illustrate the superiority and potential applicability of AI in the clinical analysis of suspicious interval cancer through the comparison with human clinicians. Furthermore, due to its potential utility in nationwide population-based cancer monitoring, public health research was conducted, raising novel issues about patients with a superficial level of analysis.

9. Conclusion

Introduction of artificial intelligence (AI) into medical imaging has led to progress in the early diagnosis and treatment of diseases. In particular, it has had a significant impact on cancer, which is the leading cause of death worldwide. Today, AI in medical examination can automatically detect various aspects such as lesion detection, classification, segmentation, and information extraction on medical images using digital imaging processing techniques. It can analyze imaging tests in a way that is quantitatively different from the analysis of the human eye and analyze data faster. In addition, it can avoid the variability of images that may differ from person to person or under different scanning conditions.

Therefore, AI has been applied to uncover hidden patterns in screening and diagnosis algorithms, medical image data, to assist in the early diagnosis of cancer. There are concerns worldwide regarding the time interval from first symptoms to the time cancer is detected because visiting a

doctor is delayed if there are no noticeable symptoms until cancer is radiation and examination progresses to the terminal stage. In addition, new technologies using medical imaging have a significant impact in early detection and treatment. In particular, cancers such as breast cancer, stomach cancer, and lung cancer progress rapidly, so there is a particular need to detect and treat early. Due to these needs, several research groups are developing AI models corresponding to different cancers in the medical field.

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