

Impact of artificial intelligence on clinical radiography practice: futuristic prospects in a low resource setting

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ABSTRACT

Current trends in clinical radiography practice include the integration of artificial intelligence (AI) and related applications to improve patient care and enhance research. However, in low resource countries there are unique barriers to the process of AI integration. Using Ghana as a case study, this paper seeks to discuss the potential impact of AI on future radiographic practice in low-resource settings. The opportunities, challenges and the way forward to optimise the potential benefits of AI in future practice within these settings have been explored.

Some of the barriers to AI integration into radiographic practice relate to lack of regulatory and legal policy frameworks and limited resource availability including unreliable internet connectivity and low expert skillset.

These barriers notwithstanding, AI presents a great potential to the growth of medical imaging and subsequently improving quality of healthcare delivery in the near future. For example, AI-enabled radiographer reporting has a potential to improving quality of healthcare, especially in low-resource settings like Ghana with an acute shortage of radiologists. In addition, futuristic AI-enabled advancements such as synthetic cross-modality transfer where images from one modality are used as a baseline to generate a corresponding image of another modality without the need for additional scanning will be of particular benefit in low-resource settings.

The urgent need for inclusion of AI modules for the training of the radiographer of the future has been suggested. Recommendations for development of AI strategies by national societies and regulatory bodies will harmonise the implementation efforts. Finally, there is need for collaboration between clinical practitioners and academia to ensure that the future radiography workforce is well prepared for the AI-enabled clinical environment.

Keywords: Artificial intelligence, low-resource settings, Ghana, radiography, machine learning

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Introduction

The practice of radiography is highly reliant on technology and thus current innovations and advancements are contributing significantly to the quickly evolving profession. The transition from manual film processing, through to the phase of automatic processing and to the most recent use of digital daylight image processing is an example of such changes due to technological advancement.¹ These technological advancements revolutionised the practice of clinical radiography, particularly, with the introduction of cross-sectional imaging modalities like computed tomography (CT) and magnetic resonance imaging (MRI). Current trends in clinical radiography practice include the integration of artificial intelligence (AI) and related applications to improve patient care and enhance research in the field.^{2,3} In low resource settings with unique challenges including healthcare infrastructure inadequacies, a discussion relating to the future impact of these modern innovations on clinical radiography practice is necessary.

Using Ghana as a case study, this paper aims to discuss the futuristic impacts of AI on clinical radiography practice.

AI usage in clinical radiography practice in Ghana

The benefits of AI notwithstanding, very little activity is reported of AI implementation and usage from the clinical radiography setting in most low resource countries. In Ghana, the areas of radiography practice that have seen some form of AI implementation are in image processing/reconstruction (aimed at reducing image artifact), dose optimisation and diagnosis of tuberculosis (TB) and recently for COVID-19 detection. The epidemiological distribution and prevalence of TB in Ghana is nearly four times the World Health Organisation (WHO) average.⁴ The government therefore employed a commercially developed AI tool - Computer-Aided Detection for Tuberculosis (CAD4TB) (<https://www.delft.care/cad4tb/>) in an attempt to improve disease diagnosis especially within highly affected regions of the country. This system automatically detects TB on chest imaging systems that have CAD4TB installed using AI approaches.⁴ The computer-Aided Detection for COVID-19 (CAD4COVID) (<https://www.delft.care/cad4covid/>) AI software was also introduced in response to the COVID-19 pandemic to help triage suspected COVID-19 patients from the populations. Both CAD4TB and CAD4COVID operate through a similar technical pipeline by employing AI approaches in the generation of heatmap scores ranging between 0 and 100 to indicate the extent of abnormalities. These observed abnormalities are then quantified as a ratio or percentage of the total lung that is affected.⁵

In terms of dose optimisation and image processing, almost all latest CT and MRI scanners have AI systems that reduce poor iso-centering. In particular, integrated technical specifications (e.g., positioning sensors) that detect specific landmarks on the patient's body surface in relation to the area of interest for a scan series and adjust the positioning (i.e., automatically moves the table) such that the majority of the scanned anatomy is located at the isocentre.⁶ This helps to optimise patient dose. Other algorithms in a few of the

1 equipments in Ghana are the systems that allow the automatic selection of optimal protocol
2 for a specific modality based on a given condition. Moreover, some automatic exposure
3 control (AEC) systems used in CT operate on the principles of machine learning in the
4 selection of an optimal tube potential and current. Of note, some of the latest equipments,
5 mostly located in the major cities (and within private facilities) of the country have certain
6 operations based on the convolutional neural network (CNN)-based deep learning approach
7 to optimise image noise.⁷ In addition, some ultrasound machines have AI software, which
8 create multi-parametric reports and automates the analysis and quantification of cardiac
9 imaging parameters. Moreover, the radiotherapy treatment planning systems in Ghana also
10 use CNNs to produce auto-segmentation when showing organs that might be at risk while
11 the available linear accelerators have AI-incorporated algorithms that automate the
12 movement of some parts to ensure precision. With the important role of AI in the health
13 sector, and the increasing demand for quality healthcare delivery including clinical imaging
14 in low resourced countries, it is anticipated that many AI-integrated equipment modalities
15 will eventually find their way into these settings in the very near future.

21 **Challenges of AI usage in clinical radiography practice in Ghana**

22 The introduction of AI in Ghana and most low resource countries is faced with challenges
23 including the lack of necessary legal frameworks for implementation and monitoring, lack of
24 knowledge and commitment from the various authorities, lack of financial resources to
25 acquire appropriate AI tools and qualified personnel to utilise and manage these tools for
26 maximum benefit. Of note, Ghana and most other developing countries are still challenged
27 by limited resources.^{8,9} Thus, the very low imaging equipment to the population ratio,
28 especially when considering specialised modalities like CT and MRI^{8,9} makes it unlikely for
29 the required investments for AI-specific equipment implementation.

30 AI is heavily dependent on fast, reliable, and affordable internet. Ghana like most low
31 resource settings is faced with the challenge of unreliable internet bandwidth quality which
32 is mostly available only in major cities and relatively expensive.¹⁰ Governments have over
33 the years made attempts at improving the internet connectivity in the country, but enough
34 progress has not been made even though Ghana is considered to have one of the best
35 internet connectivity ratings in Africa¹⁰. In most low resource settings, it is known
36 anecdotally that governments have over the years not invested the required resources in
37 healthcare generally and radiography in particular. The radiography infrastructure is quite
38 expensive and because these countries have resource constraints, equipment's are almost
39 always very sparsely distributed when they are available and in most cases are not the most
40 modern.^{8,9} The pace of development of relevant technology is so fast that by the time low
41 resource countries procure a certain technology it might have become obsolete. Previous
42 reports indicated that radiographers feel that poor equipment maintenance culture might
43 not make AI a sustainable technology for Africa.¹¹ This finding agrees with that of other studies
44 on medical equipment infrastructure and management from across the West African sub-
45 region.^{9,12,13,}

46 In Ghana and most low resource countries, there are still many equipments that are not
47 digital and/or in very poor functioning state.^{9,12,13} Many African countries are still using

1 conventional radiographs with wet-image processing techniques. In such settings, the
2 integration of AI into clinical radiography practice is almost impossible. It is only recently
3 that Ghana as part of the TB control program with Delft acquired 52 digital x-ray
4 equipments for distribution in mostly district hospitals and other healthcare facilities.⁴ Most
5 hospitals in these countries do not have electronic patient records, hard copies of
6 radiographs are still printed as part of the clinical care pathway. It is therefore difficult for
7 such facilities to make use of AI technology without the necessary investments in AI-related
8 infrastructure.¹⁴ Interestingly, even in the few hospitals with fully digital equipment,
9 electronic health records are not linked with any other hospital.

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13 Moreover, the lack of inclusion of AI theories, applications and techniques in radiography
14 curriculum coupled with the lack of experience with AI tools continue to widen the
15 knowledge gap about AI tools in Ghana and most low resource countries.^{11,15} In a recent
16 study¹⁶ of Ghanaian radiographers, as high as 82.8% of the respondents indicated they
17 lacked knowledge about AI. Most of the respondents have not seen any or learnt about AI
18 and so they could only imagine its applications to clinical practice. With any new technology,
19 some people would have various misgivings about it. Thus, the lack of knowledge and
20 misgivings have the tendency to breed misinformation and lack of understanding about the
21 AI tool. The religious beliefs of some people in Africa make them think that anything “super
22 technology” is evil. Some radiographers were of the view that AI tools are a necessary evil
23 and are prone to mistakes just like humans.¹⁶ Education would be needed to empower end
24 users of AI tools in low resource countries including Ghana.

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31 Importantly, AI comes with its own legal and data protection complexities. These issues will
32 require a robust legal framework that will ensure patients’ data is adequately protected to
33 avoid regulatory breaches as much as possible. The legal framework will also ensure that
34 data is only used for the purposes for which it was acquired, and only authorised persons
35 have corresponding access to certain levels of information. The major challenge in Ghana
36 and most low resource countries is that, although, there may be some form of legislation on
37 general data protection, nothing exist in relation to AI and policy implementation has always
38 been a challenge.¹⁷ This is known by most citizens and most are therefore very reluctant to
39 provide information.

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44 More so, cyber security is very important in the era of AI because of possible cyber-attacks
45 by hackers. It is important that patient data is fully secured to minimise or eliminate the
46 possibility of data breaches from hackers. To achieve enhanced cyber security, it is required
47 to have the necessary security infrastructure in place which is non-existent in most low
48 resource countries. Ghana has recently just passed the Cybersecurity ACT 2020 with the
49 mandate to set up a National Cyber Security Centre, to be responsible for the development
50 and implementation of a national policy and strategy on cybersecurity.¹⁸

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54 Botwe and colleagues¹⁶ in a study that assessed the perspectives of Ghanaian radiographers
55 on the integration of AI into medical imaging concluded that the participants showed a
56 general awareness of AI and believed it will improve and ease their work, however they
57 were concerned about possible job losses, AI related errors, security of data, lack of
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1 technical expertise, threats of cyber security and high equipment cost. These trends were
2 replicated in a follow-up Africa-wide study^{12,15}

3 **Potential future AI applications in radiography practice**

4 While it is unlikely that direct human interaction will be completely taken over by AI tools,
5 there are aspects of these interaction that will be affected. Activities like vetting of referrals,
6 verification of patient identity via electronic health records and assessing clinical history and
7 matching it with the modality requested are potential areas AI tools could be used.¹⁹ AI
8 tools do have the potential to access data from multiple sources much faster and more
9 accurately than a human and as such it is reasonable to expect that it is cost saving, efficient
10 and more preferred. However, there might still be the need for the radiographer to have an
11 oversight responsibility over the AI tools employed to ensure patient records have not been
12 corrupted¹. Additionally, the radiographer's role includes accurate positioning of patients in
13 all imaging examinations. Of note, previous reports^{19, 20}, on isometric positioning of patients
14 for CT and MRI indicate the possibility of automation of this procedure which is a traditional
15 radiographer role. Although, this is already happening, it is on a very limited scale, currently.
16 There is also the possibility of contrast volume being determined by AI tools based on
17 patient parameters²¹ and in some cases contrast media could be eliminated altogether by AI
18 systems by introducing synthetic contrast enhancement approaches.²²

19 Moreover, radiographer reporting is well established²³⁻²⁵ with positive outcomes.^{26,27} It is
20 envisaged that AI-led radiograph triaging may soon render radiographer reporting
21 redundant; this is indeed captured in literature.²⁸ If AI-led image triaging becomes
22 widespread and clinically viable, what becomes of reporting radiographers? AI presents a
23 great potential in medical imaging and especially for low resource settings like Ghana,
24 where there are no reporting radiographers and very few radiologists within the healthcare
25 system. AI could be used in image interpretation across most of the country where there are
26 either no radiologist or very few of them. There is a possibility of cost reduction, increased
27 productivity, and increased accuracy.²⁹⁻³² There are early indications of the possibility of
28 synthetic cross modality transfer, such that an MRI image could be generated from a CT
29 scan image and vice versa,³³ thereby eliminating the need for a second scan altogether. This
30 will be a boost to quality healthcare delivery in low resource countries like Ghana where
31 very few facilities have both CT and MRI, and there would no longer be the need to
32 purchase separate CT and MRI units thereby saving scarce resources that could be used in
33 other areas within healthcare.

34 **Opportunities for radiographers in the AI era**

35 Topol³⁴ has predicted that every clinician, be they a consultant or a specialist or a
36 paramedic, will soon be required to use AI and deep learning applications in their
37 professional practice. This period is one of uncertainty for many professions in healthcare
38 and there has been some level of speculation that some professions like radiology could be
39 replaced by automation and AI tools.³⁵ AI also presents with it numerous opportunities and
40 as to whether one profession will become redundant or not will depend on how willing and
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1 ready that profession is ready to take up the opportunities presented by AI. Indeed, roles
2 will change, skillsets that were once very essential might no longer be needed and the
3 ability and willingness of any profession to learn new skillsets and take up new responsibility
4 will largely determine how relevant the profession will be going forward.³⁶ Whilst
5 radiographers and other healthcare professionals in Ghana and most low resource countries
6 will be tempted to believe that the introduction of AI in clinical radiography practice is in the
7 distant future, urgent plans are required to prepare the radiography workforce in order not
8 to be overtaken by events. AI tools are already creating good opportunities for some
9 radiographers in low resource settings to positively influence the patient management
10 pathway.^{11,15,16} The CAD4TB installed in Ghana in particular, is changing how radiography is
11 practised in facilities which have these software tools incorporated in their x-ray
12 equipments and related accessories. Far from the norm at various imaging facilities in
13 Ghana, radiographers now do not have to physically keep radiological images of suspected
14 TB patients and carry them to radiologists for comments which take between 3 days to a
15 week or more to complete. Some radiographers are now trained while others are being
16 trained on how to read pathological heatmaps of TB and COVID-19 using specialised
17 software and communicate results to the referring clinicians through virtual platforms in
18 real time. This is reducing the time spent by patients and improving care at the imaging
19 departments. In addition to that, radiographers are serving as first line managers of the AI
20 tools at the radiology departments to ensure effective operation. In these settings, they are
21 engaging in AI-assisted preliminary image interpretation (including reporting incidental
22 findings) as core parts of their role. Thus, the AI tools are extending the roles of
23 radiographers and creating new prospects in medical imaging practice in a low resource
24 setting.

25 Moreover, there is the need to periodically audit the reports being generated by the AI
26 systems and this is one area radiographers could be engaged in. Auditing of these
27 automated clinical AI tools will serve the purpose of establishing the sensitivity, specificity,
28 and accuracy while serving the additional purpose as a peer reviewer. Against this
29 background, urgent critical infrastructural developments and training/education will be
30 required for all radiographers especially those in developing countries to embrace these
31 futuristic changes.

32 Additionally, radiographers will still have to provide the human interface for patients as AI
33 tools have still not developed to the point where this activity is no longer required. The
34 current IR(ME) regulations³⁷ clearly places the ultimate responsibility of the acquisition and
35 processing of medical imaging in the hands of the radiographer regardless of the level of
36 automation. It is unlikely this will change anytime soon. It is therefore the responsibility of
37 the radiographer to provide radiation protection education to the patients and take steps to
38 ensure the patient is always protected.³⁷ Again, the radiographer led AI-supported reporting
39 is a cheaper alternative when compared to a single radiologist, especially for high volume
40 examinations or modalities like chest radiography, CT lung and mammography screening. It
41 is therefore imperative that radiographers develop and implement plans for radiographer-
42 led AI-supported reporting services in low resource setting like Ghana that is bedevilled with
43 huge imaging backlog and inadequate personnel.^{9,38}

The way forward

Given the fast pace at which technological innovations are occurring in healthcare and particularly in medical imaging, it is understandable that some radiographers are concerned about their level of knowledge to keep active in clinical practice.^{11,15,16} Unfortunately, one common trait of radiographers in Ghana is the penchant of underestimating their power, to influence decision making in the workplace. The unique position of radiographers, ensures they understand the way technology and innovation impacts both the patient and the flow of work volumes.³⁹⁻⁴¹ Radiographers in Ghana turn to act unconcerned and when decisions are taken that affect them then they shout the loudest. Radiographers must depart from this attitude and ensure that their voice is heard throughout the process from procurement, vendor demonstration, implementation and conducting required research to inform practice and policy.⁴² The Ghana Society of Radiographers (GSR) and similar professional bodies must as a matter of urgency develop an AI strategy to guide future implementation and practice monitoring activities. Societies like the European Federation of Radiography Societies (EFRS), Canadian Association of Medical Radiation Technologists (CAMRT) and the Society and College of Radiographers (SCoR) are taking steps to ensuring that their membership is abreast with AI.^{43,44} They have set up a subcommittees and special interest working groups to provide education, develop policies for their members on the usage of AI applications in radiography practice.⁴³ Awareness creation and capacity building must be of prime importance to all professional bodies and societies and particularly those in developing countries like Ghana considering the relatively poor awareness of these latest developments.

It is important that radiographers take steps to learn the intricacies of AI in healthcare. More importantly, the GSR and similar professional bodies across low resource countries should outline their strategy towards AI. Obviously, membership will be looking up to their respective professional bodies to give them leadership, position papers and direction.^{43,45} It is also crucial that radiographers demand that any tool that will be suggested for implementation will ensure a healthy balance between efficiency and well-being of both patients and staff.⁴⁵

Critically, there is the urgent need for collaboration between radiographers in clinical practice and academia. There is also the need to ensure that the educational curriculum is reviewed to include advanced information technology (IT) and AI skills to reflect the changing times¹. Higher education will do the profession a lot of good if the next generation of radiographers are adequately trained in AI and IT skills. Higher education institutions in collaboration with employers and professional societies should organise continuous professional development (CPD) programs for radiographers already in clinical practice to update their skills¹. This will ensure they stay relevant within the ever-changing profession.¹

It is important that strategic partnerships are created between clinical practice, academia, and industry to ensure AI tools developed are relevant to the profession and keep academia in the loop on what is being developed so that in educating the next generation of radiographers, AI is factored into their curriculum. This synergy will ensure the profession is well placed with regards to usage and implementation of AI applications.

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Lastly, if radiographers are not to be left behind with AI, then conscious efforts must be made to invest in AI related research. Other countries are investing heavily in AI and low resource countries cannot afford to be left out. Investments in AI related research rose from \$80 million in 2016 to a projected value of over \$500 million dollars in 2019.^{45,46} Whilst it is unreasonable to expect Ghana, a low resource country, to invest such huge sums of money into AI related research, it is important that Ghana begin to invest in research. Moreover, there would also be need for more research to be conducted in low-resourced settings to empirically develop data to further understand the AI usage, challenges, potential future AI applications and opportunities for radiographers.

Conclusion

Artificial intelligence has come to stay, radiographers equipped with clinical, analytical and research skills should be harnessed to ensure the safe and ethical use of AI. As professionals we must accept AI, embrace it, learn it and own it. The profession of radiography must be positioned strategically to play key roles in the introduction and management of AI to ensure that practitioners can direct how it is implemented. AI will take up some roles played by the radiographer and improve upon efficiency, speed, and accuracy but there will always be the need for the indispensable element of human support and interaction. Radiographers should ensure they make inputs because if they don't, this AI ship will still sail, maybe just not in our preferred direction.

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