



**A survey of patient acceptability of the use of artificial intelligence in the diagnosis of paediatric fractures: an observational study**

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3 A survey of patient acceptability of the use of artificial intelligence in the diagnosis of  
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6 paediatric fractures: an observational study  
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45 **Competing and conflicting Interests**  
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47  
48 None.

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54 **Mini-abstract**  
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57 We performed a prospective, cross-sectional survey over 4-weeks at a tertiary hospital,  
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59 investigating carer attitudes towards the use of artificial intelligence in paediatric fracture  
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3 diagnosis. Our results demonstrate that carers think positively about AI but are not ready to  
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6 accept automated systems over human decision-making.  
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For Review Only

**Structured abstract**

**Objective:** To assessed carer attitudes towards the use of artificial intelligence (AI) in management of fractures in paediatric patients.

**Summary Background data:** As fracture clinic services come under increasing pressure, innovative solutions are needed to combat rising demand. AI programmes can be used to diagnosis fractures, but patient perceptions towards its use are uncertain.

**Methods:** We conducted a cross-sectional survey of carers of paediatric patients presenting to fracture clinic at a tertiary-care centre, combining single-best-answer questions and likert-type questions. We investigated patient perception of clinical review in the Emergency Department (ED); disruption to school to attend fracture clinic and attitudes towards AI.

**Results:** 45% of paediatric fracture patients were seen within two hours, 29% were seen between 2-4 hours and 26% were seen after 4 hours. 75% were seen by both a nurse and a doctor, 16% were seen only by a nurse, and 9% only by a doctor. 61% of children had to take time off school for their appointment, 59% of parents had to take time off. 56% agreed that more research is needed to reduce waiting times. 76% preferred a nurse or doctor to review their child's radiograph. 64% were happy for an AI programme to diagnose their child's fracture, and 82% were happy with an AI programme being used as an adjunct to a clinician's diagnosis.

**Conclusion:** Carer perceptions towards the use of AI in this setting are positive. However, they are not yet ready to relinquish human decision making to automated systems.

**Word count: 247**

## INTRODUCTION

Musculoskeletal injuries in children account for nearly half of the 4 million presentations to Paediatric Emergency Departments (ED) across the UK per year<sup>1</sup>. Of these, fractures are an important cause of morbidity, with a reported incidence between 1500 to 3600 per 100,000 children per year<sup>2</sup>. Most fractures do not require admission to hospital but may be managed as outpatients via local fracture clinics. The British Orthopaedic Association Standards for Trauma (BOAST) guidelines describe the standard of care that patients with a significant musculoskeletal injury should receive in an outpatient setting. The first point of guidance describes the timeframe for review by an Orthopaedic specialists, explaining that “patients should be seen in a new fracture clinic within 72-hours of presentation with the injury”<sup>3</sup>. Timely assessment is essential to optimal management, with delays leading to increased pain and loss of opportunity, particularly in the paediatric population<sup>4</sup>.

Fracture clinic services throughout the UK have been under pressure in recent years and the mismatch between service demand and service availability continues to pose a challenge to orthopaedic specialists<sup>5</sup>. The Covid-19 pandemic greatly exacerbated this problem, as an acute reduction in the provision of services and a shift in population health-seeking behaviour has compounded the pressure on NHS services, and increased patient backlogs<sup>6</sup>. There is, therefore, an important and continued need for innovation within orthopaedics to help meet this demand, evolving outpatient orthopaedic services at pace with developing technologies.

Artificial Intelligence (AI) has been defined as the ability of a computer to accomplish human-like tasks<sup>7</sup>. In medicine, AI has been used as diagnostic aide since the 1960s, where early-era

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3 devices provided statistical analyses of numerical data derived from radiological images to  
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5 support human clinicians in their diagnoses<sup>8</sup>. Advancements in both technological innovation  
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7 and computer processing power has driven the development of increasingly complex and  
8  
9 capable machines, with AI research now moving beyond simply mimicking intelligence and  
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11 into the exploration of areas such as experiential learning<sup>7,8</sup>. Today's AI has the potential to  
12  
13 improve the diagnosis and management of myriad medical conditions and is already seeing  
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15 effective use in specialities such as Oncology and Dermatology<sup>9,10</sup>. In the orthopaedic setting  
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17 AI has seen a variety of applications, from clinical prognostication to outcome calculation.  
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19 Notably, research has explored the use of AI in fracture identification with promising results.  
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21 AI has been shown to perform at a level equal to human diagnosticians when diagnosing  
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23 common fractures, and a specific AI outperformed both general physicians and orthopaedic  
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25 surgeons in the setting of proximal humeral fracture diagnosis. AI has also been shown to  
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27 equal human performance in recognising plain radiographic fractures of the ankle, wrist, and  
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29 hand with at least 83% accuracy. Yet evidence of the efficacy of AI in accurately diagnosing  
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31 subtle and occult fractures is lacking<sup>11</sup>.  
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42 The relative novelty of AI in healthcare means there are many barriers to its successful  
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44 implementation that are independent of the efficacy of the machine itself. Integration at an  
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46 organisational level requires transparent collaboration between organisations and AI vendors.  
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48 Yet a paucity of vendors may render healthcare organisations vulnerable to acquiring  
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50 inappropriate products, particularly where companies have a limited understanding of how  
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52 to apply their AI to the particular needs of a healthcare organisation. There is also a wide  
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54 range of computer literacy amongst clinicians and, although it is advantageous to develop  
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56 user-friendly programmes, this is not always possible. A highly effective AI may, therefore, be  
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3 untenable if the clinicians it is directed at are unable to integrate it into their daily practice<sup>12,13</sup>.  
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5 Critically, AI must also be acceptable to patients. Yet little is known in this regard, particularly  
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7 with respect to the paediatric population. The literature highlights the dehumanisation of the  
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9 clinician-patient relationship, low trustworthiness of AI, and a perceived lack of regulation as  
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11 key patient concerns and, although patients may be comfortable with the use of AI as an  
12  
13 adjunct in certain settings, they still exhibit a preference for a clinician<sup>14</sup>. It is, therefore,  
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15 essential to further elucidate patient opinions if AI is to be meaningfully employed in the  
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17 future.  
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25 This study aimed to assess parent attitudes towards use of artificial intelligence in the  
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27 management of orthopaedic injuries in paediatric patients.  
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## 32 **METHODS**

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34 This study was a noninterventional, cross-sectional survey of parents or guardians of  
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36 paediatric patients presenting to an outpatient orthopaedic fracture clinic at a tertiary care  
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38 centre in London June to August 2022. Parents or guardians of patients referred to the  
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40 fracture clinic were invited to participate when checking into their appointment and prior to  
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42 them being seen by a clinician. Participation was voluntary and the study period was four  
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44 weeks. The study was conducted as a service evaluation under audit guidelines and was  
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46 registered with the trust audit department: registration number AUDI003065.  
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54 The survey was an 11-item questionnaire ([Figure 1](#)). Data was collected on the child's initial  
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56 presentation to ED (length of time to be seen and whether they were seen by a doctor or  
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58 nurse), disruption to school or work in order to attend the outpatient appointment and  
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3 perceptions towards use of AI in managing orthopaedic injuries. Questionnaires were  
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5 anonymous and no biometric or identifiable information was collected. Questions were either  
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7 single best answer, or Likert-type with a scale ranging from strongly disagree to strongly agree.  
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12 Completed questionnaires were returned to a locked 'post box' held behind the reception  
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14 desk. Responses were manually loaded onto a secure electronic database held on a trust  
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16 computer. Data was analysed using Microsoft Excel (Microsoft Corporation, version 16). Data  
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18 was collected under the audit framework and thus ethical approval for this study was not  
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20 required.  
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## 27 **RESULTS**

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29 184 responses were obtained. 123 surveys were completed in full, with 61 surveys partially  
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31 completed.  
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### 37 *Section 1 – Regarding the child's presentation to ED*

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39 There were 141 complete responses to section 1. Total waiting time to be seen by a clinician  
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41 was represented in brackets of 1-hour, from less than 1-hour to more than 5-hours. 24%  
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43 (34/141) were seen within 1-hour, 21% (30/141) within 2-hours, 12% (17/141) within 3-hours,  
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45 and 13% (19/141) within 4-hours. 13% (19/141) reported waiting longer than 5-hours. 75%  
46  
47 (106/141) of respondents reported being seen by both a nurse and a doctor. 16% (22/141),  
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49 were seen only by a nurse and 9% (12/141) were seen only by a doctor. One respondent  
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51 reported not being seen by either.  
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6 *Section 2 – disruption to work or school to attend an outpatient clinic*  
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8 There were 165 complete responses to section 2. 61% of respondents agreed (51/165) or  
9 strongly agreed (49/165) that their child had to take time off for the appointment, 14%  
10 (23/165) were neutral, and 25% either disagreed (19/165) or strongly agreed (23/165). 59%  
11 of respondent agreed (53/165) or strongly agreed (45/165) that they personally had had to  
12 take time off for the appointment, 11% (18/165) were neutral, and 30% either disagreed  
13 (18/165) or strongly disagreed (31/165). 56% of respondents agreed (52/165) or strongly  
14 agreed (40/165) that more research is needed to reduce waiting times, whilst 35% (58) were  
15 neutral, and 9% either disagreed (7/175) or strongly disagreed (8/165). There were 177  
16 responses to the mode of transport used to attend fracture clinic appointment. 90%  
17 (159/177) of respondents attended their appointment at fracture clinic by either car, train, or  
18 bus. 69% (120/177) reported attending via private vehicle.  
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40 *Section 3 – Attitudes towards AI*  
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There were 165 complete responses to section 3. 76% (125/165) of respondents said they  
would prefer a nurse or doctor to review their child's radiograph. 64% (105/165) said they  
would be happy if an AI programme was used to diagnose their child's fracture, and 82%  
(135/165) reported being happy with an AI programme being used to help in the diagnosis of  
fractures. 8% (13/165) of respondents reported no preference for how their child's fracture  
was diagnosed but preferred AI not to be involved, 4% (7/165) reported no preference but  
would be happy for AI to assist in the diagnosis, and 12% (20/165) described no preference  
but would be happy for AI to make a diagnosis of fracture. 10% (16/165) of respondents would  
prefer a healthcare professional to make the diagnosis and preferred AI not to be involved,

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3 15% (24/165) preferred a healthcare professional to make the diagnosis (and not AI) but  
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5 would be open to having an AI programming assisting, and 52% (85/165) reported preferring  
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7 a healthcare professional to make the diagnosis and being open to AI both assisting in, or  
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9 making, the diagnosis of fracture.  
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## 15 **DISCUSSION**

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17 Our results demonstrate a positive attitude towards the use of AI in diagnosing fractures in  
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19 the paediatric setting. Only 18% of respondents did not want AI to assist in the diagnosis of  
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21 their child's fracture and, whilst 76% preferred a healthcare professional to make the  
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23 diagnosis of fracture, 82% were happy for AI to augment this interaction. 16% had no  
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25 preference for whether their child was seen by a clinician and would be happy for the process  
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27 of fracture diagnosis to be automated. These results emulate previously presented data in  
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29 this area<sup>15</sup>, offering evidence in favour of automation of diagnoses in the paediatric setting,  
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31 which has far-reaching implications.  
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40 The pathway of fracture management can be lengthy. Patients presenting to ED are triaged,  
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42 undergo an initial assessment, and then have imaging requested by a healthcare professional.  
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44 Once this is reviewed and initial management suggested, most patients are then discharged  
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46 home with an outpatient fracture clinic appointment for specialist orthopaedic review<sup>15</sup>.  
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48 Although well established, this process can be inefficient and is prone to bottlenecking. For  
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50 instance, there is commonly a time-delay between initial assessment and subsequent  
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52 suggested diagnosis in the ED. Poor staffing levels, high patient volume and/or acuity, and  
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54 limited availability of services, in particular radiology reporting, have all been cited as possible  
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3 influencing factors on this<sup>16,17</sup>. The immediate-term consequences to patients include  
4 possible long waiting times and a delay in the acquisition of appropriate high-quality care.  
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10 The navigation through both ED and fracture clinic can be very time intensive, causing  
11 significant disruptions to the patient and their carer. Regular and ongoing reviews at fracture  
12 clinic can compound this problem and may result in multiple missed days of school and work  
13 for both individuals. This survey's result support this idea. 55% of participants waited more  
14 than 2 hours to be seen in the ED, and 23% waited more than 4-hours. Further, in the  
15 outpatient setting, 65% of participants agreed or strongly agreed that they had to take time  
16 off to attend the fracture clinic. The negative effects of missing school on childhood academic  
17 attainment are well established. There is a proportionally detrimental effect of absence on  
18 attainment, with this effect beginning after just a few days' absence.<sup>18</sup> As well as the  
19 important implications for patients highlighted above, this also incurs significant loss of  
20 departmental resources and clinician time. Reducing instances of absence and their duration  
21 is, therefore, highly important. Innovations such as virtual fracture clinics, use of which has  
22 increased significantly since the covid-19 pandemic, have been effective at reducing the rate  
23 of referrals from ED to fracture clinics and, therefore, school and work days missed<sup>19</sup>.  
24 Interestingly, studies investigating the efficacy of virtual fracture clinics show that the rate of  
25 discharge, rather than onward face-to-face assessment, is between 33% and 60%<sup>20-22</sup>, which  
26 implies that there are a significant number of unnecessary referrals made to fracture clinic.  
27 Indeed, in one centre they found that 37% of paediatric fracture clinic referrals had no  
28 confirmed fracture prior to referral, and 29% of all suspected fractures were subsequently  
29 found not to have one<sup>23</sup>. Using AI to improve diagnostic accuracy may serve to further reduce  
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3 the rate of unnecessary referral, safeguard patient and carer time, and improving efficiency  
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5 of hospital systems.  
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10 Although research into the development and implementation of AI in radiological diagnostics  
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12 has existed for many years, the vast majority of current and historical AI programmes  
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14 represent investigational proofs of concept with minimal near-future clinical applications. A  
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16 recent review examined the availability of licensed AI programmes in this field, highlighting  
17  
18 only six. Of these, 50% used plain radiography as their modality (OsteoDetect, FractureDetect,  
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20 BoneView). Each demonstrates high sensitivity (88.0% – 95.0%) and specificity (88.0% – 90.2).  
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22 One (OsteoDetect) shows a performance comparable to a clinician, and all have been shown  
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24 to improve clinician performance when used as an augmentative measure<sup>24</sup>. This is supported  
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26 by a recent systematic review with meta-analysis, which found that, across all available  
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28 literature, including grey-literature, the “pooled diagnostic performance from the use of  
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30 artificial intelligence (AI) to detect fractures had a sensitivity of 92% and 91% and specificity  
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32 of 91% and 91%, on internal and external validation, respectively”<sup>25</sup>.  
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42 However, no currently licensed AI programme has been approved as the sole diagnostic agent  
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44 capable of replacing a clinician, nor has any been licensed for use in the paediatric setting.  
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46 Furthermore, current machine and deep-learning AI programmes are designed to review  
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48 specific body parts or regions, with no single programme yet capable of performing at  
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50 clinician-level in all musculoskeletal regions. As such, they currently have limited practical  
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52 application in isolation to one-another, except in regions with high individual fracture  
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54 prevalence. The acceptability of AI to patients is a key factor that cannot be ignored. Yet, as  
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56 highlighted earlier, research in this area is lacking, particularly in the paediatric population.  
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3 One study, similar in design to this research, investigated the hypothetical use of an AI  
4 programme versus a clinician in radiograph interpretation to explore patient perceptions of  
5 the use of AI as an adjunct to clinician diagnosis. It found significantly higher confidence of  
6 patients in the accuracy of a clinician's diagnosis when compared with AI (9.20 vs 7.06,  
7  $p < 0.001$ ) and, when asked to determine their preference in case of a disagreement between  
8 the two, 95.4% indicated a preference for a clinician. Additionally, this study reported a  
9 significantly higher patient confidence in AI-assisted interpretation versus AI-assisted  
10 management (7.06 vs 4.86,  $p < 0.001$ )<sup>26</sup>. Our study demonstrates a similar pattern, where the  
11 majority of carers reported a preference for a clinician (76%), but were open to AI being used  
12 as an adjunct to diagnosis (82%). Further research, particularly in the paediatric population,  
13 is needed to bolster these and other initial, promising results.  
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33 Utilising an AI programme for autonomous fracture diagnosis may be beneficial at both the  
34 individual and departmental level. Obtaining a rapid diagnosis would allow for faster decision  
35 making and appropriate management strategies to take place, which could improve both  
36 patient safety and treatment outcomes by reducing waiting times and time-to-treatment. The  
37 varied skill mix and diagnostic confidence and accuracy of ED clinicians means that certain  
38 non-fracture injuries may be inappropriately immobilised and referred on for specialist review.  
39 For individual clinicians, AI could improve the diagnostic accuracy and confidence of non-  
40 specialists, thereby reducing cognitive load. This is significant as it reduces the risk of missed  
41 diagnoses that can result from cognitive fatigue<sup>27</sup>. It may also give expert clinicians more time  
42 and mental capacity to review and diagnose more complex emergency pathologies. At the  
43 departmental level, implementing AI programmes in imaging diagnostics has the potential to  
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3 reconfigure patient streaming pathways, reducing bottlenecks to diagnosis and management,  
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5 and reducing overall capacity issues through the ED.  
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9 Another positive impact of AI in this context may be through the reduction of unnecessary  
10 travel to outpatient appointments. In this survey 69% of respondents used a private car to  
11 attend their appointment. The recent coronavirus pandemic has affected travel behaviours,  
12 with working-from-home become the norm now for many, and fewer people preferring public  
13 transport or sustainable commuting over private vehicle<sup>28</sup>. Research has demonstrated that  
14 the rate of climate change is accelerating, which poses a threat to both the national and global  
15 public health gains of the last century<sup>29</sup>. The Greener NHS Programme<sup>30</sup> seeks to reduce the  
16 environmental impact of healthcare and create a sustainable model for the future. Virtual  
17 fracture clinics, originally implemented to reduce the burden on outpatient services, already  
18 dovetail well with this initiative. They have been shown to be highly effective, improving  
19 patient outcomes and satisfaction, and reducing face-to-face attendances by up to half<sup>31</sup>. AI  
20 would support this new green initiative, as a reduction in unnecessary appointments through  
21 improved diagnostic accuracy would reduce unnecessary vehicular travel.  
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41 There are several limitations that may negatively influence this study's results. The study's  
42 completion rate was only 67%. This may be due to the binary nature of several of the  
43 questions, as people may not have felt the answers available were representative of their  
44 opinions. Similarly, the questionnaire was divided into three sections that covered different  
45 time periods during their child's management journey, meaning that some respondents may  
46 have been unable to remember and recall information accurately. The data collection period  
47 was also relatively short, and there was no biometric data collected which limits this study's  
48 generalisability. The wording of the questionnaire is may also be limiting, as respondents were  
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3 not directly asked if they would prefer an AI programme over a human to diagnose a fracture  
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5 in their child. Expanding the questionnaire to obtain a more detailed understanding of  
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7 respondent preferences would serve to significantly strengthen these initial results.  
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## 10 11 12 13 **CONCLUSION**

14  
15 This study assessed participant attitudes towards the use of AI in the diagnosis of fractures in  
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17 the paediatric setting. The results show that perceptions towards the use of AI in this context  
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19 are positive, but that carers still prefer a clinician with respect to fracture diagnosis. Patient  
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21 education around AI and its potential benefits may improve its acceptability as a diagnostic  
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23 tool.  
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3 Figures and Table Legends  
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8 Figure 1 – illustrating the questionnaire completed by parents of children attending the  
9 fracture clinic.  
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For Review Only

Paediatric artificial intelligence fracture study

On a scale of 1 to 5 where 1 is not likely and 5 is very likely please answer the following questions:

When we went to the emergency department after my child's accident we had to wait \*\*\* hours to be seen  
Please circle:  
<1 h      1-2 h      2-3h      3-4h      4-5h      >5 hours

I was seen by a nurse  
Yes      No

I was seen by a doctor  
Yes      No

---

My child had to take a day off school for this appointment

Strongly Disagree	Disagree	Neutral	Agree	Strongly Agree
1	2	3	4	5

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I had to take time off work for this appointment

Strongly Disagree	Disagree	Neutral	Agree	Strongly Agree
1	2	3	4	5

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We came to the appointment today by:  
Walking      Bicycle      Bus      Train      car      other

---

I would prefer that a nurse or doctor looks at my child's X ray in the emergency department  
Yes      No

---

I would be happy if an 'artificial intelligence' computer programme was used to look at my child's X-ray and diagnose the fracture  
Yes      No

---

I would be happy if an 'artificial intelligence' computer programme was used to help doctors and nurses diagnose a fracture  
Yes      No

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I think more research is needed to reduce appointments and make it easier for my child to be seen and treated if they have a suspected fracture

Strongly Disagree	Disagree	Neutral	Agree	Strongly Agree
1	2	3	4	5

---

I would be happy to be contacted again to help in the design of a research study looking at the accuracy of artificial intelligence programmes in diagnosing fractures in children

Strongly Disagree	Disagree	Neutral	Agree	Strongly Agree
1	2	3	4	5

Please write your name and email address if you would be happy to take part in a future study  
Name:      Email address:

Figure 1 - illustrating the questionnaire completed by parents of children attending the fracture clinic

279x203mm (300 x 300 DPI)