








3 ORIGINAL ARTICLE

4 The use of artificial intelligence in
5 diagnosing emergency cases: a cross-
6 sectional study of physicians' perspectives
7 on the effectiveness of AI systems in
8 improving diagnostic accuracy and
9 reducing time to diagnosis

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14 ABSTRACT

15 **Background:** Artificial intelligence (AI) has the potential to improve diagnosis accuracy and speed in emer-
16 gency rooms; yet, its acceptance and use by emergency physicians in Saudi Arabia remain limited.

17 **Methods:** A descriptive cross-sectional survey was issued to emergency physicians in Saudi Arabia. The ques-
18 tionnaire looked at demographics, AI exposure, perceived benefits and problems, and overall views toward AI.
19 The data were evaluated with descriptive statistics, chi-square tests, ANOVA, and logistic regression.

20 **Results:** A total of 352 physicians participated, the majority of whom were young residents. Only 11.6% said
21 they used AI on a regular basis. The majority of physicians (65.3%) had a negative attitude, whereas 6.8% were
22 optimistic. Consultants had more favorable perceptions than residents. Attitudes differed considerably by
23 region, age, gender, country, and professional level ($p < 0.05$). Key perceived benefits included increased diag-
24 nostic accuracy and speedier decision-making, while major hurdles included a lack of training, system inaccur-
25 acy, and difficulties interpreting AI outputs. Male and non-Saudi physicians had higher positive sentiments.

26 **Conclusion:** Despite acknowledging AI's potential, practical adoption in emergency contexts remains low due
27 to poor training and dependability issues. Structured AI training and system development are required to
28 ensure a safe and effective integration into emergency practice.

29 **Keywords:** Artificial intelligence, emergency department, diagnostic accuracy, decision-making, triage, clinical
30 practice, healthcare technology, AI in medicine.

31 Introduction

32 Artificial intelligence (AI) is the development of
33 computer systems that can execute human cognitive
34 capabilities, including problem solving and decision
35 making. AI-based algorithms are rapidly being used
36 to anticipate a variety of medical factors, including
37 diagnosis, therapy selection, and risk assessment. AI has
38 showed great potential in improving the capacities of
39 emergency departments (EDs), which treat patients with

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44 a wide range of diseases from non-urgent to serious. To
45 achieve prompt and correct medical interventions in these
46 circumstances, proper prioritizing and patient evaluation
47 must take place using dynamic systems [1,2].

48 AI's significance in emergency departments has grown
49 dramatically, as it is utilized to increase diagnostic
50 accuracy, aid in clinical decision-making, and maximize
51 operational efficiency. AI aids diagnostic processes by
52 evaluating medical imaging, such as X-rays, computed
53 tomography (CT) scans, and magnetic resonance imaging
54 (MRI), to detect conditions including trauma, stroke, and
55 pneumonia [3]. It also uses real-time patient data to aid
56 clinical decision-making by recommending treatments
57 based on patient severity [4]. Furthermore, AI contributes
58 significantly to predictive analytics by forecasting
59 critical illnesses such as sepsis and cardiac arrest based
60 on vital signs and medical history [5]. AI-enhanced triage
61 systems select patients based on urgency, ensuring that
62 the most serious cases receive immediate attention [6].
63 Furthermore, AI helps to optimize resource management
64 in the ED by anticipating patient volumes and improving
65 staff allocation, potentially reducing overcrowding [7].
66 Continuous patient monitoring facilitated by AI exposes
67 clinicians to early indicators of deterioration, ultimately
68 improving patient outcomes [3].

69 Several research have examined the use of AI in emergency
70 case diagnosis, with varying results. A comprehensive
71 review conducted in Germany emphasized the increasing
72 usage of AI applications in emergency care; nevertheless,
73 most studies are still in early phases of development and
74 lack sufficient generalizability, underlining the need for
75 more advanced approaches in future study [6]. Another
76 study, done at a tertiary research hospital's emergency
77 department, demonstrated that AI, namely ChatGPT, was
78 effective in differentiating patients with high urgency
79 [8]. Furthermore, a diagnostic accuracy research at Imam
80 Khomeini Hospital Complex, including 215 patients with
81 acute abdominal discomfort, found that AI could create
82 a model with a high level of diagnostic accuracy for
83 emergency case triage [1].

84 The goal of this study is to determine the effectiveness of
85 AI in diagnosing situations in the emergency room.

86 **Subjects and Methods**

87 This descriptive cross-sectional study evaluated
88 emergency department (ED) physicians' attitudes,
89 experiences, and perceived usefulness of AI in
90 emergency medical diagnosis across Saudi Arabia.
91 Data were collected between March and September
92 2025. The target population consisted of licensed
93 emergency physicians aged 24 and above, representing
94 several specialties practicing in emergency settings, as
95 well as a wide range of educational backgrounds, job
96 situations, and geographical locations. The inclusion of
97 ED physicians from various professional categories was
98 intended to provide a complete perspective on AI-related
99 diagnostic techniques.

100 To improve data validity and assure accurate depiction
101 of real-world ED procedures, part of the data collection
102 method entailed visiting emergency departments (EDs)
103 in person and approaching and inviting on-site physicians

to participate. This strategy enabled the research team
to directly contact active ED physicians during clinical
shifts, enhancing response quality and reducing selection
bias. Additional responses were gathered online to
increase sample size and geographic diversity.

Only physicians who provided complete responses and
informed consent were considered in the final analysis.
Participants who were under the age of 24, unlicensed,
or had no prior experience with diagnostic or AI-assisted
tools were eliminated. According to the most recent
Ministry of Health statistics for physicians and dentists
in Saudi Arabia (2023), a minimum sample size of 344
physicians, stratified by gender, nationality, specialty,
and professional category, was calculated using OpenEpi
with a 95% confidence interval and a 5% margin of error.
However, data collection was extended beyond this
criterion to improve statistical robustness, and a total of
352 responses were examined.

Data were obtained using a validated, self-administered
online questionnaire created with Google Forms. To
optimize accessibility and reach, the questionnaire was
circulated electronically via social media platforms such
as WhatsApp, X (previously Twitter), Telegram, and
email groups, as well as in-person within educational
institutions. The instrument underwent professional
evaluation and pilot testing to assure its clarity, relevance,
and logical flow.

The questionnaire has various sections that were aimed
to capture demographic and AI-related information.
Demographic data included age, gender, nationality,
specialty, job status, years of experience, and practice
region. Additional questions assessed physicians' current
usage of diagnostic tools, prior exposure to AI platforms,
and their overall health and workload characteristics.
The AI component investigated physicians' impressions
of AI's diagnostic accuracy, dependability, impact on
diagnosis time, utility in clinical decision-making,
comfort level with AI tools, and perceived barriers and
benefits of incorporating AI into emergency departments.

Data was initially exported to Microsoft Excel for
cleaning and coding. Statistical analyses were carried
out with IBM Statistical Package for the Social Sciences
(SPSS) version 28. The demographic variables and
AI-related replies were summarized using descriptive
statistics such as frequencies, percentages, averages, and
standard deviations. Pearson's chi-square test or Fisher's
exact test were used to investigate associations between
independent variables and physicians' attitudes toward
AI-assisted diagnosis, if appropriate. A p -value of <0.05
was judged statistically significant.

Ethical approval and consent to participate

This study was conducted in accordance with the
ethical principles of the Declaration of Helsinki. Ethical
approval was obtained from the Biomedical Research
Ethics Committee at Umm Al-Qura University, Saudi
Arabia (Approval No. HAPO-02-K-012-2025-03-2589).
Informed consent was obtained electronically from all
participants prior to their participation. Participation
was voluntary, and confidentiality and anonymity of the
collected data were strictly maintained.

164 **Data analysis**

165 Data analysis was conducted using IBM SPSS Statistics
 166 version 28 (IBM Corp., Armonk, NY, USA). Descriptive
 167 statistics were used to summarize the participants’
 168 sociodemographic and professional characteristics, as
 169 in emergency diagnosis well as their responses to items
 170 assessing perspectives on AI. Categorical variables
 171 were presented as frequencies and percentages, while
 172 continuous variables were summarized using means and
 173 standard deviations (mean ± SD).

174 To assess physicians’ perspectives on AI use in
 175 emergency settings, a scoring system was developed
 176 based on responses to five Likert-scale items, with each
 177 item scored from 1 to 5. The total perspective score
 178 ranged from 8 to 40. Based on the total score, attitudes
 179 were categorized into three groups:

180 Negative attitude: total score ≤ 23

181 Neutral or moderate attitude: score 24–31

182 Positive attitude: score ≥ 32

183 Comparisons of mean perspective scores across
 184 professional levels (resident, specialist, consultant) were
 185 performed using one-way analysis of variance (ANOVA).
 186 Associations between physicians’ attitudes and various
 187 sociodemographic and professional factors were
 188 evaluated using Pearson’s chi-square test (χ^2) or Fisher’s
 189 exact test when appropriate. Relationships between
 190 attitudes and perceived benefits or challenges of AI use
 191 were also examined using chi-square tests. Furthermore,
 192 a multivariate logistic regression analysis was performed
 193 to identify independent predictors of having a positive
 194 attitude toward AI. Results were expressed as adjusted
 195 odds ratios (AORs) with corresponding 95% confidence
 196 intervals (CIs). A *p*-value of < 0.05 was considered
 197 statistically significant throughout the analysis. All
 198 figures and visualizations were generated using Microsoft
 199 Excel 2019.

200 **Results**

201 The study included 352 physicians, with roughly equal
 202 representation of male (*N* = 180, 51.1%) and female (*N*
 203 = 172, 48.9%) participants. The bulk of participants were
 204 young physicians, with more than 60% (*N* = 212, 60.2%)
 205 falling between the ages of 20 and 30. Emergency
 206 medicine residents made up the biggest group (*N* =
 207 234, 66.5%), followed by specialists (*N* = 82, 23.3%)
 208 and consultants (*N* = 36, 10.2%). Geographically, the
 209 Western Region received the most replies (*N* = 116,
 210 33.0%), followed by the Central Region (*N* = 83, 23.6%)
 211 and the Eastern Region (*N* = 77, 21.9%). Finally, the vast
 212 majority of the physicians were Saudi (*N* = 317, 90.1%)
 213 (Table 1).

214 Table 2 summarizes clinicians’ thoughts on the role of
 215 AI in ED diagnosis. Physicians were most optimistic
 216 about AI’s ability to increase diagnostic accuracy and
 217 minimize diagnosis time (mean = 2.3 ± 1.2). However, in
 218 both cases, less than 35% of respondents strongly agreed
 219 (accuracy: *N* = 123, 34.9%; time reduction: *N* = 116,
 220 33.0%).

Table 1. Socio-demographic characteristics and professional data of study participants (*N* = 352).

Data	No	%
Region		
Central region	83	23.6%
Northern region	24	6.8%
Eastern region	77	21.9%
Western region	116	33.0%
Southern region	52	14.8%
Age in years		
20-30	212	60.2%
30-40	66	18.8%
40-50	46	13.1%
50-60	21	6.0%
> 60	7	2.0%
Gender		
Male	180	51.1%
Female	172	48.9%
Nationality		
Saudi	317	90.1%
Non-Saudi	35	9.9%
Profession		
Emergency Medicine Resident	234	66.5%
Emergency Medicine Specialist	82	23.3%
Emergency Medicine Consultant	36	10.2%

In terms of reliability and comfort (negative effects), the highest mean scores were observed for relying on AI in emergency decision-making (mean = 2.8 ± 1.3) and considering AI a reliable substitute to human physicians (mean = 2.8 ± 1.4). In the latter case, a significant proportion of respondents (*N* = 91, 25.9%) chose a neutral response, indicating uncertainty rather than outright rejection. Difficulty working with AI systems had a high mean score (2.8 ± 1.3), indicating ongoing hurdles to process integration.

Only 11.6% (*N* = 41) of physicians reported regular use of AI approaches in their job, resulting in a neutral mean score (mean = 2.7 ± 1.3). Furthermore, the majority of respondents were ambivalent when comparing AI accuracy to their own clinical judgment (*N* = 117, 33.2%), indicating that AI is currently viewed as an unconfirmed or non-integrated technology rather than a trusted clinical aid in emergency circumstances.

Figure 1 depicts clinicians’ thoughts on the benefits and challenges of employing AI in ED diagnosis. The three most often stated benefits were improved diagnostic accuracy (*N* = 186, 52.8%), supporting clinicians with rapid decision-making (*N* = 184, 52.3%), and shorter patient wait times (*N* = 184, 52.3%). These findings suggest that AI is largely viewed as a tool for improving the essential operations of emergency care, including speed and precision. In contrast, the most generally stated problems were a lack of training in AI technologies (*N* = 191, 54.3%), system inaccuracy in some circumstances (*N* = 148, 42.0%), and difficulty interpreting AI-generated findings (*N* = 113, 32.1%), all of which have a negative

Table 2. Physicians' perspectives on the effectiveness, utility, and reliability of AI in emergency diagnosis (N = 352).

Perspectives on the Effectiveness	1 (No, %)	2 (No, %)	3 (No, %)	4 (No, %)	5 (No, %)	Mean ± SD
Can AI improve diagnostic accuracy in emergency cases?	123 (34.9%)	88 (25.0%)	77 (21.9%)	40 (11.4%)	24 (6.8%)	2.3 ± 1.2
Does AI reduce the time to diagnose emergencies compared to traditional methods?	116 (33.0%)	97 (27.6%)	78 (22.2%)	32 (9.1%)	29 (8.2%)	2.3 ± 1.2
Can AI be a reliable alternative to human doctors in emergency diagnosis?	90 (25.6%)	60 (17.0%)	91 (25.9%)	63 (17.9%)	48 (13.6%)	2.8 ± 1.4
Do you regularly use AI techniques in emergency department work?	86 (24.4%)	72 (20.5%)	99 (28.1%)	54 (15.3%)	41 (11.6%)	2.7 ± 1.3
How do you evaluate AI diagnostic accuracy versus your own?	82 (23.3%)	80 (22.7%)	117 (33.2%)	44 (12.5%)	29 (8.2%)	2.6 ± 1.2
Do you find it difficult to interact with AI systems during diagnosis?	68 (19.3%)	84 (23.9%)	88 (25.0%)	59 (16.8%)	53 (15.1%)	2.8 ± 1.3
Is AI particularly useful in rapid diagnosis (e.g., stroke, heart attack)?	81 (23.0%)	87 (24.7%)	95 (27.0%)	53 (15.1%)	36 (10.2%)	2.6 ± 1.3
Do you feel comfortable relying on AI in emergency decisions?	79 (22.4%)	66 (18.8%)	107 (30.4%)	60 (17.0%)	40 (11.4%)	2.8 ± 1.3

impact on trust and adoption. Resistance from physicians or medical teams was also recorded ($N = 94, 26.7\%$).

Figure 2 demonstrates that the vast majority of physicians (65.3%) had a negative view about the use of AI in emergency diagnosis. Only 6.8% had a positive attitude, while 27.8% had neutral or moderately supportive views.

Figure 3 illustrates the differences in mean perspective scores between male and female physicians.

Table 3 shows the mean scores of physicians' thoughts on AI use in emergency diagnosis, broken down by professional level. Across most items, emergency medicine consultants consistently reported higher mean scores, reflecting more positive sentiments toward AI than specialists and residents. Statistically significant differences were seen across multiple areas. Consultants were more likely to agree that AI can minimize diagnostic time (mean = 2.6 ± 1.3) than residents (2.2 ± 1.2 ; $p = 0.049$). Consultants assessed AI diagnostic accuracy much higher (3.2 ± 1.4) compared to residents (2.5 ± 1.2) and specialists (2.6 ± 1.1 ; $p = 0.006$). Consultants agreed more on the utility of AI in rapid diagnosis (3.2 ± 1.3 ; $p = 0.003$) and were more confident in using AI for emergency judgments (3.2 ± 1.5 ; $p = 0.0001$). They also reported considerably greater ease in engaging with AI systems (3.2 ± 1.3) than locals (2.7 ± 1.3 ; $p = .017$).

Table 4 outlines factors influencing physicians' attitudes about AI use in emergency diagnosis. Attitude was substantially linked with region ($p = 0.001$), with physicians from the Eastern (79.2%) and Western Regions (70.7%) having the most negative attitudes, while physicians from the Southern Region had the most positive views (17.3%). Age was also substantially associated with viewpoint ($p = 0.0001$), with physicians aged 20-30 years more likely to have unfavorable attitudes (68.4%) than those older than 60 years, who had the largest proportion of positive views (57.1%). Gender was strongly linked with viewpoint ($p = 0.010$), with females having higher negative views (73.3%) than males (57.8%). Nationality also indicated a significant correlation ($p = 0.004$), with Saudi physicians more likely to have negative attitudes (67.8%) than non-Saudi physicians (42.9%), who had higher positive attitudes (17.1%). Professional level was also strongly associated with viewpoint ($p = 0.022$), with residents having the most unfavorable opinions (70.1%), while consultants had more positive attitudes (16.7%).

Table 5 depicts the relationship between physicians' overall attitudes on AI use in emergency diagnosis and their perceived benefits and obstacles. These relationships were statistically significant for both benefits ($p = 0.001$) and problems ($p = 0.008$). Among perceived benefits, "improving diagnostic accuracy" was the most popular; nonetheless, 74.2% of physicians who chose this benefit still had a negative view towards AI. Similarly, individuals with unfavorable or neutral sentiments were more likely to select "assisting physicians in rapid decision-making" and "reducing patient waiting time". Regarding problems, "lack of training in AI technologies" was the most commonly cited issue, with 70.2% of replies coming from physicians with negative sentiments. Other obstacles, such as difficulty interpreting AI results,

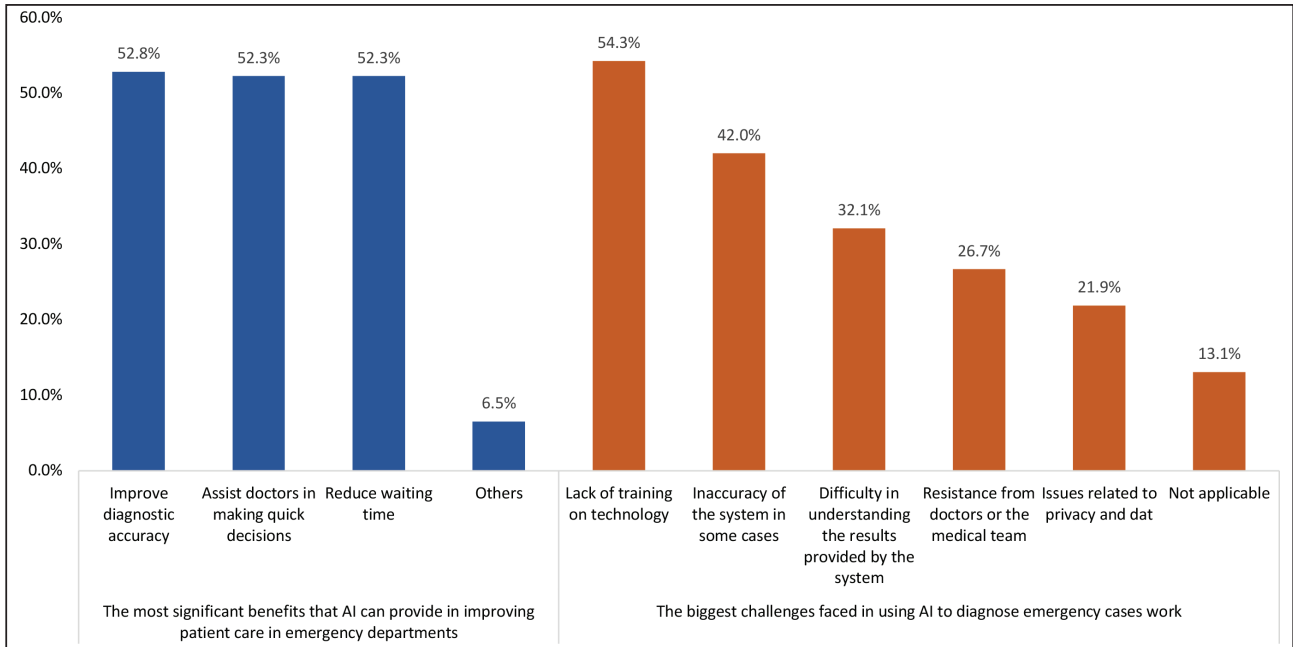


Figure 1. Perceived benefits and significant challenges of using AI in emergency departments (N = 352).

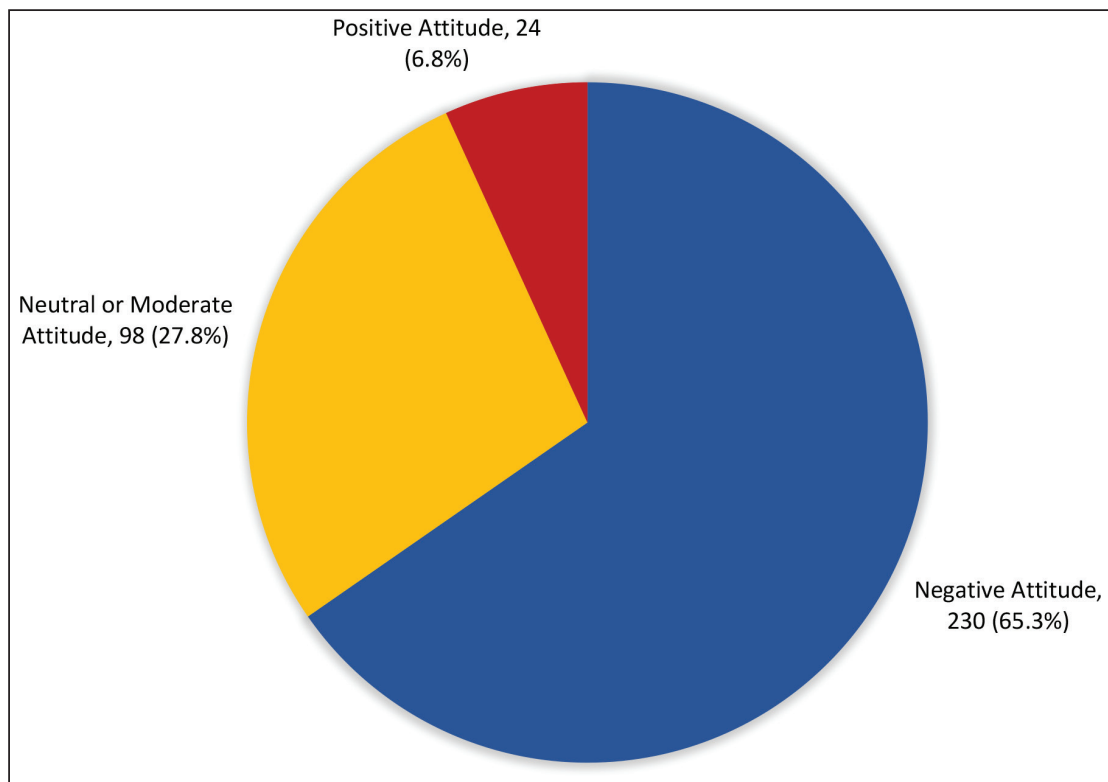


Figure 2. Overall physician perspectives on the use of AI in emergency diagnosis.

315 system inaccuracies, opposition from coworkers, and
 316 privacy or data security issues, followed a similar trend.

317 Table 6 shows the findings of a multivariate logistic
 318 regression analysis, which identified characteristics
 319 linked with physicians' positive attitudes on AI use
 320 in emergency diagnosis. Physicians from the Eastern
 321 Region had a substantially lower likelihood of reporting
 322 a favorable attitude than those from the Central Region

(AOR = 0.40, 95% CI: 0.19-0.84; $p = 0.016$). Gender 323
 was also a significant predictor, with male physicians 324
 reporting a more positive attitude than females (AOR = 325
 1.79, 95% CI: 1.11-2.87; $p = 0.017$). Furthermore, non- 326
 Saudi physicians were more than twice as likely to have 327
 positive sentiments than Saudi physicians (AOR = 2.63, 328
 95% CI: 1.22-5.67; $p = 0.014$). Other variables, such 329
 as age, professional level, and remaining geographical 330
 categories, were not statistically significant ($p > 0.05$). 331

Table 3. Mean scores (\pm SD) of physicians' perspective on AI use in emergency diagnosis, stratified by professional level.

Perspective items	Emergency Medicine Resident	Emergency Medicine Specialist	Emergency Medicine Consultant	p-value
Can AI improve diagnostic accuracy in emergency cases?	2.2 \pm 1.2	2.5 \pm 1.3	2.4 \pm 1.3	0.140
Does AI reduce time to diagnose emergencies versus traditional methods?	2.2 \pm 1.2	2.5 \pm 1.4	2.6 \pm 1.3	0.049*
Can AI be a reliable alternative to human doctors in emergency diagnosis?	2.8 \pm 1.4	2.7 \pm 1.2	3.1 \pm 1.4	0.397
Do you regularly use AI techniques in emergency department work?	2.6 \pm 1.3	2.8 \pm 1.3	3.0 \pm 1.3	0.115
How do you evaluate AI diagnostic accuracy versus your own?	2.5 \pm 1.2	2.6 \pm 1.1	3.2 \pm 1.4	0.006*
Do you find it difficult to interact with AI systems during diagnosis?	2.7 \pm 1.3	3.1 \pm 1.2	3.2 \pm 1.3	0.017*
Is AI particularly useful in rapid diagnosis (e.g., stroke, heart attack)?	2.5 \pm 1.2	2.8 \pm 1.2	3.2 \pm 1.3	0.003*
Do you feel comfortable relying on AI in emergency decisions?	2.6 \pm 1.3	3.0 \pm 1.2	3.2 \pm 1.5	0.001*

P: One-Way ANOVA
 * P < 0.05 (significant).

Table 4. Factors associated with physicians' perspective on AI use in ER diagnosis, Saudi Arabia.

Factors	The overall physicians' perspective on AI use in ER diagnosis						p-value
	Negative Attitude		Neutral or Moderate Attitude		Positive Attitude		
	No	%	No	%	No	%	
Region							0.001*
Central Region	49	59.0%	33	39.8%	1	1.2%	
Northern Region	12	50.0%	9	37.5%	3	12.5%	
Eastern Region	61	79.2%	11	14.3%	5	6.5%	
Western Region	82	70.7%	28	24.1%	6	5.2%	
Southern Region	26	50.0%	17	32.7%	9	17.3%	
Age in years							0.0001**
20-30	145	68.4%	55	25.9%	12	5.7%	
30-40	42	63.6%	20	30.3%	4	6.1%	
40-50	29	63.0%	15	32.6%	2	4.3%	
50-60	13	61.9%	6	28.6%	2	9.5%	
> 60	1	14.3%	2	28.6%	4	57.1%	
Gender							0.010*
Male	104	57.8%	61	33.9%	15	8.3%	
Female	126	73.3%	37	21.5%	9	5.2%	
Nationality							0.004*
Saudi	215	67.8%	84	26.5%	18	5.7%	
Non-Saudi	15	42.9%	14	40.0%	6	17.1%	
Profession							0.022*
Emergency Medicine Resident	164	70.1%	57	24.4%	13	5.6%	
Emergency Medicine Specialist	48	58.5%	29	35.4%	5	6.1%	
Emergency Medicine Consultant	18	50.0%	12	33.3%	6	16.7%	

P: Pearson X2 test; ^: Exact Probability test.
 * P < 0.05 (significant).

332 **Discussion**

333 This cross-sectional study investigated emergency
 334 physicians' attitudes toward the use of AI in diagnosing
 335 emergency cases throughout Saudi Arabia. The findings
 336 revealed a prevalence of negative sentiments (65.3%),
 337 with only 6.8% of participants expressing a positive
 338 outlook. Despite the low adoption rate, more than half of
 339 physicians recognized AI's potential benefits, particularly
 340 in enhancing diagnostic accuracy (52.8%) and assisting
 341 with speedy decision-making (52.3%). These findings
 342 imply that, while physicians recognize AI's potential, its

practical application in emergency departments (EDs) 343
 is hampered by a number of structural and pedagogical 344
 constraints. 345

The demographics of the respondents had a significant 346
 impact on their perceptions. The bulk of participants were 347
 young physicians aged 20-30 years (60.2%) and residents 348
 (66.5%), both of whom were skeptical of AI. This is due 349
 to insufficient exposure to AI tools and a lack of formal 350
 training programs. Similar findings were observed by 351
 Castelvechhi [9] and Goddard et al. [10], who emphasized 352
 that early-career physicians frequently express distrust 353

Table 5. Association of physicians' perspectives with the benefits and challenges of AI use in emergency department diagnosis.

Items	The overall physicians' perspective on AI use in ER diagnosis						p-value
	Negative Attitude		Neutral or Moderate Attitude		Positive Attitude		
	No	%	No	%	No	%	
In your opinion, what are the most significant benefits that AI can provide in improving patient care in emergency departments?							.001*
Improve diagnostic accuracy	138	74.2%	39	21.0%	9	4.8%	
Assist doctors in making quick decisions	111	60.3%	58	31.5%	15	8.2%	
Reduce waiting time	120	65.2%	56	30.4%	8	4.3%	
Others	8	34.8%	9	39.1%	6	26.1%	
What are the biggest challenges you might face in using AI to diagnose emergency cases in your work?							.008*
Lack of training on technology	134	70.2%	46	24.1%	11	5.8%	
Difficulty in understanding the results provided by the system	79	69.9%	30	26.5%	4	3.5%	
Inaccuracy of the system in some cases	83	56.1%	57	38.5%	8	5.4%	
Resistance from doctors or the medical team	59	62.8%	26	27.7%	9	9.6%	
Issues related to privacy and data	46	59.7%	24	31.2%	7	9.1%	
Not applicable	27	58.7%	14	30.4%	5	10.9%	

P: Pearson χ^2 test; ^: Exact Probability test

* $P < 0.05$ (significant).

Table 6. Multivariate logistic regression analysis of predictors of positive physician attitudes toward AI use in emergency diagnosis.

Predictors	p-value	OR _A	95% CI	
			Lower	Upper
Region				
Northern region	0.637	1.26	0.48	3.29
Eastern region	0.016*	0.40	0.19	0.84
Western Region	0.140	0.63	0.34	1.16
Southern Region	0.285	1.49	0.72	3.07
Age in years	0.352	1.13	0.87	1.47
Male versus Female gender	0.017*	1.79	1.11	2.87
Non-Saudi versus Saudi	0.014*	2.63	1.22	5.67
Profession				
Emergency Medicine Specialist versus Resident	0.492	1.22	0.69	2.17
Emergency Medicine Consultant versus Resident	0.327	1.52	0.66	3.53

OR_A: Adjusted odds ratio; CI: Confidence interval

* $P < 0.05$ (significant).

354 in AI due to the “black-box” nature of algorithms and
 355 unclear accountability systems. In contrast, consultants
 356 in this study reported much more comfort and confidence
 357 in using AI for clinical decision-making ($p = 0.001$).
 358 This is consistent with Loh et al. [11] and Alghamdi et
 359 al. [12], who discovered that clinical seniority and past
 360 experience with digital tools are important determinants
 361 of AI trust and adoption.

362 Regional and gender differences were also identified.
 363 Physicians from the Eastern (79.2%) and Western
 364 (70.7%) areas had the most negative sentiments, while
 365 those from the Southern Region had the most positive
 366 attitudes (17.3%). Male physicians were more optimistic
 367 than females ($p = 0.017$), whereas non-Saudi physicians
 368 had higher positive attitudes than Saudi nationals
 369 ($p = 0.014$). These trends could reflect inequalities

in institutional exposure and training possibilities, as
 well as educational backgrounds. Similar trends were
 observed in studies from Australia and Europe, where
 exposure to AI-integrated systems was associated with
 better acceptance rates among clinicians [13,14]. This
 emphasizes the necessity of national measures for
 ensuring fair access to AI education and infrastructure
 across regions and genders.

Participants cited the most significant problems as a
 lack of AI training (54.3%), system inaccuracy (42.0%),
 and difficulty interpreting AI outcomes (32.1%),
 demonstrating that integration barriers remain persistent.
 These findings are consistent with those of Fritsch et al.
 [15] and Chen et al. [16], who noted that insufficient AI
 education, usability problems, and restricted algorithm
 transparency impair clinician trust. Furthermore,

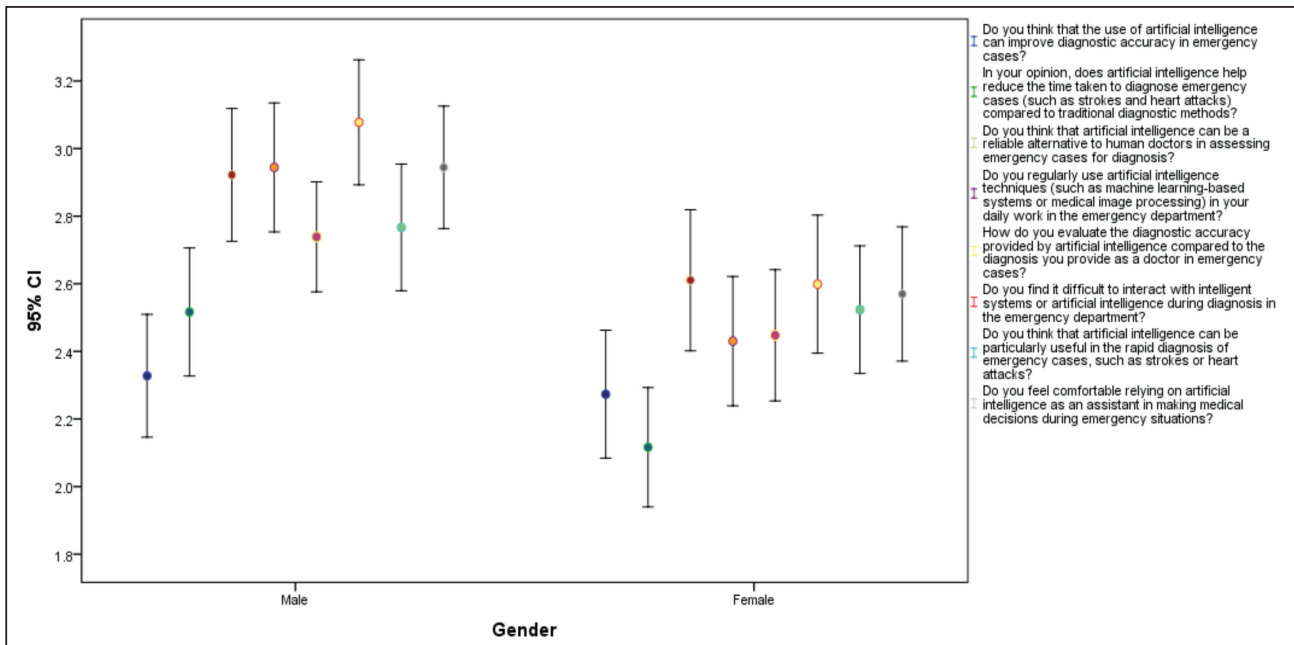


Figure 3. Error bars for Mean Scores (\pm CI) of physicians' perspective on AI use in emergency diagnosis, stratified by gender.

387 ethical concerns about data privacy, accountability, and
 388 automated bias continue to be significant barriers to
 389 clinical use [9,10]. While several studies, such as those
 390 conducted by Stewart et al. [17] and Eastwood et al.
 391 [18], found that emergency physicians regard AI as a
 392 beneficial decision-support tool, they repeatedly stressed
 393 the importance of retaining human oversight to avoid
 394 overreliance on automation.

395 The findings of this study demonstrated that consultants
 396 scored much higher in their belief that AI decreases
 397 diagnosis time ($p = 0.049$) and increases diagnostic
 398 accuracy when compared to their own judgments.
 399 This suggests that when physicians gain exposure and
 400 practical experience, they get a stronger appreciation
 401 for AI's diagnostic skills. Loh et al. [11] came to similar
 402 conclusions, reporting that prior training and familiarity
 403 dramatically increase doctors' trust and acceptance
 404 of AI in emergency medicine. As a result, structured
 405 educational programs that incorporate workshops,
 406 simulation-based training, and AI certification modules
 407 into residency curricula may be helpful techniques for
 408 increasing competence and confidence among early-
 409 career physicians.

410 Although the current use of AI in EDs is low (11.6%),
 411 increasing awareness of its potential benefits indicates
 412 an opportunity for healthcare change. To achieve
 413 safe and effective integration, AI solutions must be
 414 understandable, transparent, and designed to supplement
 415 rather than replace clinical judgment. Building trust
 416 and accountability requires establishing clear legal
 417 frameworks, developing hospital-level AI policies,
 418 and providing continual professional training [9,12].
 419 Furthermore, joint research between technology
 420 developers and physicians should prioritize improving
 421 the interpretability of AI systems and confirming their
 422 accuracy in real-world emergency circumstances.

Strengths and limitations

423 A notable aspect of this study is the inclusion of a large,
 424 diversified, and geographically representative sample
 425 of emergency physicians from all five Saudi regions,
 426 allowing for a full assessment of national perceptions.
 427 The inclusion of several professional levels—residents,
 428 experts, and consultants—enabled comparisons across
 429 experience groups. However, the study's cross-sectional
 430 design limits causal inference, and the use of self-reported
 431 data may introduce response bias. Furthermore, the study
 432 only included emergency medicine professionals, limiting
 433 the generalizability of the findings to other medical
 434 disciplines. Furthermore, the majority of responders
 435 were from the Western Region, which is most likely due
 436 to the location of the research team and linked university.
 437 This may have contributed to a higher response rate in
 438 that location, limiting the results' applicability to other
 439 regions.
 440

Conclusion

441 In summary, this study reveals a cautious but growing
 442 interest in the use of AI among Saudi emergency
 443 physicians. While the majority recognized AI's promise to
 444 increase diagnostic accuracy and decision-making speed,
 445 real-world adoption remained low due to insufficient
 446 training, a lack of institutional support, and concerns
 447 about reliability. Consultants' higher confidence levels
 448 than locals imply that professional expertise and exposure
 449 play an important role in establishing positive opinions.
 450 As a result, national healthcare authorities should
 451 prioritize structured AI literacy programs, transparent
 452 system development, and policy-level assistance to
 453 ensure the safe incorporation of AI into emergency
 454 care. Future longitudinal and multi-specialty research is
 455 needed to assess changing perceptions and quantify the
 456 long-term impact of AI adoption on patient outcomes and
 457 clinical efficiency.
 458

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460	AI	Artificial intelligence		
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486	References			
487	1.	Farahmand S, Shabestari O, Pakrah M, Hossein-Nejad		
488		H, Arbab M, Bagheri-Hariri S. Artificial intelligence–		
489		based triage for patients with acute abdominal pain in		
490		emergency department: a diagnostic accuracy study. <i>Adv</i>		
491		<i>J Emerg Med.</i> 2017;1(1):e5. https://doi.org/10.22114/		
492		AJEM.v1i1.11		
493	2.	Seyedhosseini-Davarani S, Nejati A, Hossein-Nejad H,		
494		Mousavi SM, Sedaghat M, Arbab M, et al. Outcome-based		
495		validity and reliability assessment of raters regarding the		
496		admission triage level in the emergency department: a		
497		cross-sectional study. <i>Adv J Emerg Med.</i> 2018;2(3):32.		
498		https://doi.org/10.22114/AJEM.v0i0.76		
499	3.	Mendo IR, Marques G, De La Torre Díez I, López-Coronado		
500		M, Martín-Rodríguez F. Machine learning in medical		
501		emergencies: a systematic review and analysis. <i>J Med</i>		
502		<i>Syst.</i> 2021;45(10):88. https://doi.org/10.1007/s10916-		
503		021-01762-3		
504	4.	Liu N, Zhang Z, Ho AFW, Ong MEH. Artificial intelligence in		
505		emergency medicine. <i>J Emerg Crit Care Med.</i> 2018;2:1–		
506		10. https://doi.org/10.21037/jeccm.2018.10.08		
507	5.	Kijpaisalratana N, Sanglertsinlapachai D, Techaratsami		
508		S, Musikataporn K, Saoraya J. Machine learning		
509		algorithms for early sepsis detection in the emergency		
510		department: a retrospective study. <i>Int J Med Inf.</i>		
		2022;160:104689. https://doi.org/10.1016/j.ijmedinf.2022.104689		511
				512
	6.	Piliuk K, Tomforde S. Artificial intelligence in emergency		
		medicine: a systematic literature review. <i>Int J Med</i>		
		<i>Inf.</i> 2023;179:105274. https://doi.org/10.1016/j.ijmedinf.2023.105274		513
				514
				515
				516
	7.	Cellina M, Cè M, Irmici G, Ascenti V, Caloro E, Bianchi L,		
		et al. Artificial intelligence in emergency radiology: where		
		are we going?. <i>Diagnostics.</i> 2022;12(12):3223. https://doi.org/10.3390/diagnostics12123223		517
				518
				519
				520
	8.	Colakca C, Ergin M, Ozensoy HS, Sener A, Guru S,		
		Ozhasenekler A. Emergency department triaging using		
		ChatGPT based on emergency severity index principles: a		
		cross-sectional study. <i>Sci Rep.</i> 2024;14(1):22106. https://doi.org/10.1038/s41598-024-72793-0		521
				522
				523
				524
				525
	9.	Castelvecchi D. Can we open the black box of AI?. <i>Nature.</i>		
		2016;538(7623):20–3. https://doi.org/10.1038/538020a		526
				527
	10.	Goddard K, Roudsari A, Wyatt J. Automation bias and		
		trust in clinical decision support systems: a systematic		
		review. <i>BMC Med Informat Decis Making.</i> 2012;12:99.		
		https://doi.org/10.1186/1472-6947-12-99		528
				529
				530
				531
	11.	Loh E, Clark J, Sia CH. Clinicians’ trust and acceptance of		
		artificial intelligence in clinical decision-making: a cross-		
		sectional survey. <i>BMJ Health Care Inf.</i> 2023;30(1):e100708.		
		https://doi.org/10.1136/bmjhci-2022-100708		532
				533
				534
				535
	12.	Alghamdi BA, Alkhwaiter TA, Qassim AJ, et al. Perception		
		and awareness of emergency medicine specialists toward		
		artificial intelligence integration in the Riyadh region,		
		Saudi Arabia. <i>Int J Med Dev Ctries.</i> 2024;8(9):2250–9.		
		https://doi.org/10.24911/IJMDC.51-1718105777		536
				537
				538
				539
				540
	13.	Ahun E, Demirci H, Yilmaz S. Perceptions and concerns		
		of emergency medicine clinicians about artificial		
		intelligence use during epidemic triage. <i>Front Public</i>		
		<i>Health.</i> 2023;11:1187456. https://doi.org/10.3389/fpubh.2023.1187456		541
				542
				543
				544
				545
	14.	Stewart J, Eastwood K, Gordon C. Attitudes towards		
		artificial intelligence in emergency medicine: a qualitative		
		study of emergency clinicians in Australia and New		
		Zealand. <i>Emerg Med Australas.</i> 2024;36(2):243–51.		
		https://doi.org/10.1111/1742-6723.14343		546
				547
				548
				549
				550
	15.	Fritsch SJ, Hamet P, Ellahham S, Piper J. Attitudes and		
		perceptions of artificial intelligence in healthcare: a		
		scoping review. <i>Front Digit Health.</i> 2022;4:1024453.		
		https://doi.org/10.3389/fdgth.2022.1024453		551
				552
				553
				554
	16.	Chen M, Hao Y, Cai Y, Li S. Acceptance of clinical artificial		
		intelligence among physicians and medical students: a		
		systematic review. <i>BMC Med Informat Decis Making.</i>		
		2022;22(1):276. https://doi.org/10.1186/s12911-022-		555
		02015-w		556
				557
				558
				559
	17.	Eastwood K, Gordon C, Stewart J, et al. Needs and		
		expectations for artificial intelligence in emergency		
		medicine: a mixed-method survey of physicians. <i>BMC</i>		
		<i>Health Serv Res.</i> 2023;23(1):1156. https://doi.org/10.1186/s12913-023-10148-1		560
				561
				562
				563
				564
	18.	Medscape News. Majority of doctors worry about		
		artificial intelligence driving clinical decisions, survey		
		shows [Internet]. 2023 [cited 2025 Jan 2].		565
				566
				567