











# Evaluation of Artificial Intelligence Answers for Short Stature in Paediatric Endocrinology by Paediatric Endocrinologists

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## What is already known about this?

Artificial intelligence (AI) is increasingly used in medical decision-making, including in pediatric endocrinology. AI models can help diagnose short stature by analyzing growth patterns and related factors, but not much is known about their accuracy and reliability.

## What does this study adds?

This study evaluated AI-generated decisions about short stature by comparing them with expert opinions. It highlights the strengths and limitations of AI in clinical decision-making and identifies areas where AI is or is not in line with expert recommendations, particularly in the field of short stature.

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## ABSTRACT

**Objective:** Artificial intelligence (AI) is increasingly used in medicine, including pediatric endocrinology. AI models have the potential to support clinical decision-making, patient education, and guidance. However, their accuracy, reliability, and effectiveness in providing medical information and recommendations remain unclear. The aim was to evaluate and compare the performance of four AI models, ChatGPT, Bard, Microsoft Copilot, and Pi, in answering frequently asked questions related to pediatric endocrinology.

**Methods:** Nine questions commonly asked by parents regarding short stature in pediatric endocrinology were selected, based on literature reviews and expert opinions. These questions were posed to four AI models in both Turkish and English. The AI-generated responses were evaluated by 10 pediatric endocrinologists using a 12-item Likert-scale questionnaire assessing medical accuracy, completeness, guidance, and informativeness. Statistical analyses, including Kruskal-Wallis and post-hoc tests, were conducted to determine significant differences between AI models.

**Results:** Bard outperformed other models in guidance and recommendation categories, excelling in directing users to medical consultation. Microsoft Copilot demonstrated strong medical accuracy but lacked guidance capacity. ChatGPT showed consistent performance in knowledge dissemination, making it effective for patient education. Pi scored the lowest in guidance and recommendations, indicating limited applicability in clinical settings. Significant differences were observed between AI models ( $p < 0.05$ ), particularly in completeness and guidance-related categories.

**Conclusion:** The present study highlights the varying strengths and weaknesses of AI models in an area of pediatric endocrinology. While Bard was effective in guidance, Microsoft Copilot excelled at accuracy, and ChatGPT was informative. Future AI improvements should focus on balancing accuracy and guidance to enhance clinical decision-support and patient education. Tailored AI applications may optimize the role of AI in specialized medical fields.

**Keywords:** Pediatric endocrinology, artificial intelligence (AI), clinical decision support, medical informatics

## Introduction

Artificial intelligence (AI) has been rapidly expanding its applications in the field of medicine, including pediatric endocrinology. The complexity of clinical problems and the rapidly evolving need for information in pediatric endocrinology further enhance the potential of AI in this domain. This study evaluated the responses provided by AI systems to frequently asked questions about short stature. The published evidence demonstrates the applicability of AI in various areas of pediatric endocrinology, including growth disorders, obesity, diabetes management, and hormonal imbalances (1,2,3).

The integration of AI into pediatric endocrinology has become particularly prominent in diabetes management. Winkelman et al. (4) reported that AI had been successfully used for optimizing insulin dosing and predicting hypoglycemia risk. In addition, Zhang et al. (3) found that AI-assisted bone age analyses improve diagnostic accuracy in cases of growth hormone deficiency.

AI has also shown significant contributions to the early diagnosis of thyroid diseases. Otjen et al. (5) highlighted the high success rate of AI in the automated analysis of thyroid ultrasound images. Furthermore, AI models used in obesity and management of insulin resistance have facilitated personalized treatment approaches (1,2).

In terms of growth disorders, the accuracy of AI in bone age measurement and its impact on accelerated diagnostic processes are particularly noteworthy. Waikel et al. (6) showed that AI may

serve as an effective educational tool for recognizing genetic syndromes.

The aim of this study was to analyze the accuracy of AI-generated responses to questions concerning short stature and the efficacy of growth hormone treatment by a panel of expert pediatric endocrinologists. The integration of AI into clinical practice has the potential to reduce the workload of healthcare professionals while playing an important complementary role in patient care and clinical decision-making. However, challenges such as data security, ethical concerns, and algorithmic accuracy remain key issues that need to be addressed.

## Methods

First, a literature review and expert opinions were used to identify the nine most frequently asked questions by parents about short stature, which were then posed to AI models. Subsequently, the AI-generated responses were evaluated by 10 pediatric endocrinologists. A 12-item questionnaire was developed to assess these responses, and the endocrinologists were asked to complete it.

## Participants

The study included 10 pediatric endocrinologists. The participants were selected randomly (using a simple random sampling method) from experts who had at least five years of experience in pediatric endocrinology and were actively engaged in clinical practice. No authors of the present study were eligible for inclusion on the expert panel.

## Question Development

To determine the most commonly asked questions by the parents of pediatric endocrinology patients, a literature review was conducted, and expert opinions were sought. As a result, a total of nine questions were formulated. Each AI model was queried separately in both Turkish and English. The selected questions were:

1. What is short stature?
2. What are the causes of short stature?
3. How is growth velocity assessed in short stature?
4. How is bone age determined in cases of short stature?
5. What should be considered in the differential diagnosis of short stature?
6. Which laboratory parameters should be evaluated in cases of short stature?
7. What medications are used in the treatment of short stature?
8. How frequently should short stature be monitored?
9. What are the potential side effects of growth hormone therapy?

## AI Models

The questions were posed to four different AI models: ChatGPT (developer: OpenAI; access: <https://chat.openai.com>), Bard/Gemini (developer: Google; access: <https://gemini.google.com>), Microsoft Copilot (developer: Microsoft; access: <https://copilot.microsoft.com>), and Pi (developer: Inflection AI; access: <https://pi.ai>). Each AI model was queried separately in both Turkish and English, and the responses were recorded for further analysis. Due to the rapid evolution of these models, the findings reported herein are strictly limited to the versions evaluated at the time of data collection.

## Evaluation Process

The responses obtained from AI systems were evaluated by 10 pediatric endocrinologists. A 12-item Likert-type questionnaire was used for the assessment. For each AI-generated response, experts rated the following survey questions on a scale from 1 to 5:

1. Was a proper definition provided?
2. Was all necessary information included?
3. Was any essential information missing?
4. Was excessive information provided?
5. Was any irrelevant information included?
6. Was the medical information accurate?

7. Were recommendations given?
8. Was patient guidance provided?
9. Was a recommendation to consult a physician included?
10. Was the response sufficient for the patient?
11. Did the response aim to inform the reader?
12. Did the response aim to reassure the reader?

## Statistical Analysis

The data are presented as mean, standard deviation, median, minimum, and maximum values. The obtained data were analyzed using SPSS, version 20.0 (IBM Inc., Armonk, NY, USA). The Kruskal-Wallis test was used to determine the significance of differences between the responses to the questions. In cases where significant differences were observed, post-hoc tests were conducted using the Kruskal-Wallis (k samples) test. A significance level of  $p < 0.05$  was considered statistically significant.

## Results

A total of 10 pediatric endocrinologists specializing in pediatric endocrinology participated in the study. Table 1 presents the evaluation results of responses provided by four AI models [ChatGPT, Bard, Microsoft Copilot (MC), and Pi] to nine pediatric endocrinology related questions, as assessed by experts using a 12-item Likert type questionnaire. The expert evaluation results for each question posed to AI models are summarized as follows.

### Evaluation of AI Models in Answering “What is Short Stature?”

Bard received the highest score for definition accuracy ( $5 \pm 1$ ), though the difference among models was not significant ( $p = 0.139$ ). In the “Missing Information Provided” category, ChatGPT had a higher tendency for incomplete responses compared to Bard ( $p = 0.027$ ). Bard and MC performed best in “Recommendations Provided” and “Patient Guidance Provided” while Pi scored the lowest ( $p < 0.001$ ). Bard and MC also excelled in “Response Aims to Inform the Reader” with MC significantly outperforming ChatGPT ( $p = 0.007$ ). In “Response Aims to Reassure the Reader” Bard led, and Pi ranked lowest, with a significant difference between Bard and MC ( $p < 0.001$ ) (Table 1).

### Evaluation of AI Models in Answering “What are the Causes of Short Stature?”

ChatGPT scored highest in the “All Necessary Information Provided” category ( $4 \pm 1$ ), significantly outperforming MC and Pi ( $p = 0.001$ ). In “Essential Information Missing” Bard had the lowest score ( $2 \pm 1$ ), with Pi and MC scoring higher ( $p = 0.011$ ). Bard and ChatGPT performed better in avoiding irrelevant information compared to MC ( $p = 0.011$ ). Bard excelled in “Recommendations Provided” ( $4 \pm 1$ ) and “Patient Guidance Provided” ( $5 \pm 0$ ), while

Table 1. Expert evaluation of AI-generated responses to pediatric endocrinology questions I

	Group					p	Kruskal-Wallis H	Post-hoc
	ChatGPT	Bard	MC	Pi	Mean±SD; Med±IQR			
What is short stature?	Was a proper definition provided?	4±1; 4±2	5±1; 5±1	4±1; 4±1	4±1; 4±0	0.139	5.498	
	Was all necessary information included?	3±1; 2.5±2	4±1; 4±1	3±1; 3±1	3±1; 3±1	0.195	4.707	
	Was any essential information missing?	4±1; 4±0	3±1; 3±2	3±1; 3±1	3±1; 4±2	0.027	9.150	Bard-ChatGPT
	Was excessive information provided?	2±1; 1.5±2	2±1; 2±3	2±1; 2±1	2±1; 2±1	0.669	1.558	
	Was any irrelevant information included?	2±1; 2±1	3±2; 3±3	2±1; 2±2	2±1; 2±3	0.885	0.650	
	Was the medical information accurate?	3±1; 2.5±3	4±1; 3.5±1	3±1; 3±2	3±1; 3±1	0.349	3.289	
What are the causes of short stature?	Were recommendations given?	2±1; 2±2	4±0; 4±0	4±1; 4±1	2±1; 2±1	<0.001	29.005	Pi-MC, Pi-Bard, ChatGPT-MC, ChatGPT-Bard
	Was patient guidance provided?	2±1; 2±1	4±1; 4±1	4±0; 4±0	2±1; 1.5±1	<0.001	26.564	Pi-MC, Pi-Bard, ChatGPT-MC, ChatGPT-Bard
	Was a recommendation to consult a physician included?	2±1; 2±1	4±1; 4±1	4±0; 4±0	1±0; 1±1	<0.001	30.593	Pi-MC, Pi-Bard, ChatGPT-MC, ChatGPT-Bard
	Was the response sufficient for the patient?	2±1; 2±1	3±1; 4±2	3±1; 3±2	3±1; 3±2	0.074	6.923	
	Did the response aim to inform the reader?	3±1; 3±2	4±1; 4±1	4±0; 4±0	3±1; 3±1	0.007	12.160	ChatGPT-MC
	Did the response aim to reassure the reader?	3±1; 3±1	4±1; 4±2	3±1; 2±2	2±1; 2±0	<0.001	18.140	Pi-Bard, MC-Bard
What are the causes of short stature?	Was a proper definition provided?	4±1; 4±3	3±1; 4±2	4±1; 3.5±1	3±1; 2±2	0.162	5.137	
	Was all necessary information included?	4±1; 4±1	3±1; 4±2	2±0; 2±0	2±1; 2±1	0.001	15.588	MC-ChatGPT, Pi-ChatGPT
	Was any essential information missing?	3±1; 2.5±1	2±1; 2±3	4±1; 4±1	4±1; 4±0	0.011	11.076	Bard-Pi
	Was excessive information provided?	2±1; 2±0	2±1; 2±2	3±1; 2±2	2±0; 2±1	0.171	5.013	
	Was any irrelevant information included?	2±0; 2±1	2±1; 2±1	3±1; 3±1	2±1; 2±2	0.011	11.176	Bard-MC, ChatGPT-MC
	Was the medical information accurate?	4±1; 4±2	4±1; 4±1	3±1; 3±1	3±1; 3.5±2	0.011	11.114	MC-ChatGPT
	Were recommendations given?	3±1; 3.5±2	4±1; 4±1	3±1; 3±1	2±1; 2±0	<0.001	23.636	Pi-MC, Pi-Bard
	Was patient guidance provided?	3±1; 3±2	5±0; 5±1	4±1; 4±1	2±1; 2±1	<0.001	24.831	Pi-MC, Pi-Bard, ChatGPT-Bard
	Was a recommendation to consult a physician included?	3±1; 3±2	5±1; 5±1	4±1; 4±1	2±1; 2±1	<0.001	23.756	Pi-MC, Pi-Bard
	Was the response sufficient for the patient?	4±1; 4±2	4±1; 4±1	3±1; 3±1	2±1; 2±2	<0.001	20.325	Pi-ChatGPT, Pi-Bard
	Did the response aim to inform the reader?	4±1; 4±1	4±1; 4±0	4±1; 4±1	3±1; 3±2	0.041	8.278	Pi-Bard
	Did the response aim to reassure the reader?	3±1; 3±1	4±2; 4.5±3	3±1; 2±1	3±1; 3±2	0.247	4.135	

Table 1. Continued

	Group						Post-hoc
	ChatGPT	Bard	MC	Pi	Kruskal-Wallis H	p	
How is growth rate evaluated in short stature?	Mean±SD; Med±IQR	Mean±SD; Med±IQR	Mean±SD; Med±IQR	Mean±SD; Med±IQR			
	4±1; 4±1	3±2; 3±4	3±1; 2±2	3±1; 2±1	8.433	0.038	Pi-ChatGPT
	3±1; 4±2	4±1; 4±1	3±1; 3±1	4±1; 4±1	5.875	0.118	
	3±1; 2±3	2±1; 2±1	4±1; 4±1	2±1; 2±1	8.412	0.038	Bard-MC
	2±1; 2±0	3±1; 2±2	2±1; 2±1	2±0; 2±1	4.264	0.234	
	2±1; 2±1	2±1; 2±3	2±1; 2±0	3±1; 2±3	1.124	0.771	
	4±1; 4±1	4±1; 3±1	4±0; 4±1	4±1; 4±0	3.487	0.322	
	3±1; 2.5±2	4±0; 4±0	2±1; 1.5±1	3±1; 2.5±2	18.720	<0.001	ChatGPT-Bard, MC-Bard
	3±1; 3.5±1	4±1; 4±2	2±1; 1.5±1	3±1; 2±2	12.110	0.007	MC-Bard
	4±1; 4±1	4±1; 5±2	2±1; 2±1	3±1; 3±2	13.789	0.003	MC-Bard
	4±1; 4±3	4±1; 4±2	2±1; 2±0	4±1; 4±2	8.428	0.038	MC-ChatGPT, MC-Bard, MC-Pi
	4±0; 4±1	4±1; 4±1	3±1; 3±1	4±1; 4±1	9.917	0.019	MC-Bard
How is bone age determined in short stature?	3±1; 3.5±2	3±2; 4±3	2±1; 2±2	3±1; 2.5±2	4.339	0.227	
	4±1; 4±1	3±1; 4±2	3±1; 4±2	3±1; 4±2	2.801	0.423	
	4±1; 3.5±2	4±1; 4±1	3±1; 3±1	3±1; 3±1	6.513	0.089	
	3±1; 3±3	3±1; 3±2	4±1; 4±1	3±1; 4±1	6.894	0.075	
	2±1; 2±0	3±1; 2±2	3±1; 3.5±2	2±0; 2±1	5.323	0.150	
	2±1; 2.5±1	3±1; 2±1	3±1; 3±2	2±1; 2±1	4.813	0.186	
	4±1; 4±0	3±1; 3±1	3±1; 3±2	3±1; 3±1	6.781	0.079	
	3±1; 3.5±1	4±1; 4±1	2±1; 2±1	2±1; 1.5±2	18.220	<0.001	Pi-Bard, MC-Bard
	3±1; 3±1	4±0; 4±1	2±1; 2.5±2	2±1; 2±1	19.372	<0.001	Pi-Bard, ChatGPT-Bard, MC-Bard
	4±1; 3.5±2	5±1; 4.5±1	2±0; 2±1	2±1; 2±0	25.323	<0.001	Pi-Bard, MC-Bard, MC-ChatGPT
	3±1; 3.5±1	4±1; 4±1	3±1; 3±1	3±1; 3±1	8.178	0.042	MC-Bard
	4±0; 4±0	4±1; 4±1	3±1; 3±1	4±1; 4±0	7.096	0.069	
	3±1; 2.5±2	2±1; 2±1	2±1; 2±1	2±1; 2±1	5.522	0.137	

Table 1. Continued														
	Group				Pi	MC	Bard	ChatGPT	Mean±SD; Med±IQR	Mean±SD; Med±IQR	Mean±SD; Med±IQR	Kruskal- Wallis H	p	Post-hoc
What should be considered in differential diagnosis in short stature?	Was a proper definition provided?	4±1; 4±1	3±1; 2.5±2	3±1; 4±1	4±1; 4±1	3±1; 3±2	3±1; 3±2	8.321	0.040	Bard-ChatGPT				
	Was all necessary information included?	4±1; 4±2	3±1; 4±1	3±1; 3±2	3±1; 3.5±2	3±1; 3±1	3±1; 3±1	1.606	0.658					
	Was any essential information missing?	3±2; 3±3	3±1; 3±2	3±1; 3±2	2±1; 2±2	2±1; 2±2	2±1; 2±2	4.436	0.218					
	Was excessive information provided?	2±1; 2±2	2±1; 2±1	2±1; 2±1	2±1; 2±0	2±0; 2±1	2±0; 2±1	2.172	0.537					
	Was any irrelevant information included?	2±1; 1.5±1	2±0; 2±0	2±0; 2±0	2±0; 2±1	2±1; 2±2	2±1; 2±2	1.461	0.691					
	Was the medical information accurate?	4±1; 4±2	4±0; 4±0	4±0; 4±0	4±1; 4±1	3±1; 4±2	3±1; 4±2	2.303	0.512					
	Were recommendations given?	4±1; 3±1	4±1; 4±2	4±0; 4±0	4±0; 4±0	3±1; 3±2	3±1; 3±2	6.707	0.082					
	Was patient guidance provided?	3±1; 2±2	4±0; 4±1	4±0; 4±1	4±0; 4±0	3±1; 3±2	3±1; 3±2	12.659	0.005	Pi-Bard, ChatGPT-Bard				
	Was a recommendation to consult a physician included?	3±1; 3±2	5±1; 5±1	5±1; 5±1	5±1; 4.5±1	3±1; 2.5±3	3±1; 2.5±3	15.615	0.001	Pi-MC, Pi-Bard				
	Was the response sufficient for the patient?	4±1; 4±2	4±1; 4±1	4±1; 4±1	4±1; 4±1	3±1; 2.5±1	3±1; 2.5±1	7.403	0.060					
Did the response aim to inform the reader?	4±1; 4±1	4±1; 4±0	4±1; 4±0	4±0; 4±0	3±1; 3±1	3±1; 3±1	13.163	0.004	Pi-ChatGPT					
Did the response aim to reassure the reader?	3±1; 2±2	2±1; 2±3	2±1; 2±3	3±1; 2±1	2±1; 2±0	2±1; 2±0	0.598	0.897						

AI: artificial intelligence, MC: Microsoft Copilot, SD: Standard deviation, IQR: interquartile range

AI: artificial intelligence, MC: Microsoft Copilot, SD: standard deviation, IQR: interquartile range

Pi scored the lowest ( $p<0.001$ ). In “Recommendation to Consult a Physician” Bard and MC led, whereas Pi performed significantly worse ( $p<0.001$ ). ChatGPT, Bard, and MC scored highest in “Response Aims to Inform the Reader”, with Pi performing worse ( $p=0.041$ ) (Table 1).

### Evaluation of AI Models in Answering “How is Growth Rate Evaluated in Short Stature?”

In the “Definition Provided” category, ChatGPT scored higher ( $4\pm1$ ) than Pi ( $3\pm1$ ), demonstrating superior definition clarity ( $p=0.038$ ). In “Essential Information Missing” Bard ( $2\pm1$ ) performed better than MC ( $4\pm1$ ), highlighting Bard’s ability to provide more complete responses ( $p=0.038$ ). Bard excelled in “Recommendations Provided” ( $4\pm0$ ), while MC scored the lowest ( $2\pm1$ ) ( $p<0.001$ ). In “Patient Guidance Provided” Bard ( $4\pm1$ ) outperformed MC ( $2\pm1$ ) ( $p=0.007$ ). Similarly, in “Recommendation to Consult a Physician” Bard ( $4\pm1$ ) led, while MC ( $2\pm1$ ) performed poorly ( $p=0.003$ ). Lastly, in “Response Was Sufficient for the Patient” MC had a significantly lower score than other models ( $p=0.038$ ), indicating its weaker performance in providing satisfactory responses (Table 1).

### Evaluation of AI Models in Answering “How is Bone Age Determined in Short Stature?”

In the “Definition Provided” category, no significant difference was found between the models ( $p=0.423$ ), with ChatGPT scoring highest ( $4\pm1$ ). In “Recommendations Provided” Bard ( $4\pm1$ ) significantly outperformed Pi and MC ( $2\pm1$ ) ( $p<0.001$ ), confirming its superiority in offering guidance. In “Patient Guidance Provided” Bard ( $4\pm0$ ) excelled, significantly outperforming all other models ( $p<0.001$ ). Similarly, in “Recommendation to Consult a Physician” Bard ( $5\pm1$ ) led, while Pi ( $2\pm0$ ) and MC ( $2\pm1$ ) performed the worst ( $p<0.001$ ). Finally, in “Response Was Sufficient for the Patient” Bard ( $4\pm1$ ) was the most effective, while MC ( $3\pm1$ ) scored significantly lower ( $p=0.042$ ), indicating Bard’s stronger ability to meet users’ informational needs (Table 1).

### Evaluation of AI Models in Answering “What Should be Considered in Differential Diagnosis in Short Stature?”

In the “Definition Provided” category, ChatGPT scored highest ( $4\pm1$ ), significantly outperforming Bard ( $3\pm1$ ) ( $p=0.040$ ). In “Patient Guidance Provided” Bard and MC ( $4\pm0$ ) excelled, while Pi and ChatGPT ( $3\pm1$ ) performed lower, with significant differences between Pi-Bard and ChatGPT-Bard ( $p=0.005$ ). For “Recommendation to Consult a Physician” Bard and MC ( $5\pm1$ ) were the most effective, while Pi ( $3\pm1$ ) performed the weakest, with significant differences between Pi-MC and Pi-Bard ( $p=0.001$ ). In



“Response Aims to Inform the Reader” ChatGPT, Bard, and MC (4±1) performed well, whereas Pi (3±1) lagged, showing a significant difference from ChatGPT ( $p=0.004$ ) (Table 1).

Table 2 presents the evaluation results of responses provided by the four AI programs to nine pediatric endocrinology-related questions, as assessed by experts using a 12-item Likert-type questionnaire. The expert evaluation results for each question posed to AI models are summarized as follows.

### Evaluation of AI Models in Answering “What Should be Considered in Laboratory Parameters in Short Stature?”

In the “Definition Provided” and “All Necessary Information Provided” categories, all models received similar scores, with no significant differences ( $p=0.595$  and  $p=0.446$ , respectively). Although ChatGPT scored highest, the variations were not significant. In “Patient Guidance Provided” Bard (4±1) outperformed MC (2±1) and Pi (3±1), with a significant difference between MC and Bard ( $p=0.030$ ), again indicating Bard’s stronger guidance ability. Similarly, in “Recommendation to Consult a Physician” Bard (3±1) and ChatGPT (3±1) scored higher than MC (2±1), with Bard significantly outperforming MC ( $p=0.014$ ). For “Response Was Sufficient for the Patient” ChatGPT (4±1) led, while MC (2±1) and Bard (3±1) scored lower, with a significant difference between MC and ChatGPT ( $p=0.018$ ). Lastly, in “Response Aims to Inform the Reader” ChatGPT (4±1) significantly outperformed MC (3±1) ( $p=0.033$ ), confirming ChatGPT’s superior capacity for providing informative responses (Table 2).

### Evaluation of AI Models in Answering “Which Drugs are used in the Treatment of Short Stature?”

In the “All Necessary Information Provided” category, MC scored the lowest (2±1), significantly underperforming compared to Bard and ChatGPT (4±1) ( $p<0.001$ ). This suggests MC was less effective in providing comprehensive information. In “Essential Information Missing” MC (4±1) had the highest score, indicating a greater tendency to provide incomplete information. ChatGPT (2±1) and Pi (2±1) scored lower, with MC significantly differing from these programs ( $p=0.002$ ). In “Medically Accurate Information Provided” MC (3±1) slightly but significantly outperformed ChatGPT ( $p=0.036$ ), highlighting MC’s relative strength in medical accuracy. For “Recommendations Provided” Bard (4±1) scored highest, with a significant difference from MC (3±1) and Pi (3±1) ( $p=0.027$ ), confirming Bard’s superiority in offering guidance. In “Response Was Sufficient for the Patient” ChatGPT (4±1) led, while MC and Pi (2±1) scored

Table 2. Expert evaluation of AI-generated responses to pediatric endocrinology questions II

	Group					Kruskal-Wallis H	p	Post-hoc
	ChatGPT	Bard	MC	Pi	Mean±SD; Med±IQR			
What should be considered in laboratory parameters in short stature?	Mean±SD; Med±IQR	Mean±SD; Med±IQR	Mean±SD; Med±IQR	Mean±SD; Med±IQR	Mean±SD; Med±IQR			
	4±2; 4±3	3±1; 3±2	3±1; 2±2	3±1; 3±2	3±1; 3±2	1.894	0.595	
	4±1; 3.5±3	3±1; 3±1	3±1; 3±2	3±1; 4±1	3±1; 4±1	2.669	0.446	
	3±2; 3±3	3±1; 3±1	3±1; 3.5±3	3±1; 3±2	3±1; 3±2	1.189	0.756	
	2±1; 2±3	2±1; 2±1	3±1; 2±2	3±1; 2±3	3±1; 2±3	1.543	0.672	
	3±1; 3.5±2	2±1; 2±1	3±1; 4±2	3±1; 2±3	3±1; 2±3	3.593	0.309	
	3±1; 4±2	3±1; 3.5±2	3±0; 3±0	3±1; 3±2	3±1; 3±2	0.598	0.897	
	3±1; 3±1	3±1; 3±1	2±1; 2±2	3±1; 3±1	3±1; 3±1	5.295	0.151	
	3±1; 3±2	4±1; 4±1	2±1; 2±1	3±1; 3±2	3±1; 3±2	8.962	<b>0.030</b>	MC-Bard
Was a proper definition provided?	3±1; 2±2	3±1; 4±2	2±1; 1±1	3±1; 3±2	3±1; 3±2	10.588	<b>0.014</b>	MC-Bard
Was all necessary information included?	4±1; 4±2	3±1; 4±2	2±1; 2±1	3±1; 3.5±1	3±1; 3.5±1	10.111	<b>0.018</b>	MC-ChatGPT
Was any essential information missing?	4±1; 4±1	4±1; 4±1	3±1; 3±1	4±0; 4±0	4±0; 4±0	8.726	<b>0.033</b>	MC-ChatGPT
Was excessive information provided?	3±1; 2±2	2±1; 2±1	3±1; 2.5±2	3±1; 2±1	3±1; 2±1	0.944	0.815	

Table 2. Continued

	Group					Kruskal-Wallis H	p	Post-hoc
	ChatGPT	Bard	MC	Pi	Mean±SD; Med±IQR			
Which drugs are used in the treatment of short stature?	Was a proper definition provided?	4±2; 4.5±3	3±2; 4±3	2±1; 2±1	3±1; 3±2	6.971	0.073	
	Was all necessary information included?	4±1; 4±2	4±1; 4±1	2±1; 2±0	3±1; 2±2	19.178	<0.001	MC-Bard, MC-ChatGPT
	Was any essential information missing?	2±1; 2±2	3±1; 3±2	4±1; 4±2	2±1; 2±1	15.036	0.002	ChatGPT-MC, Pi-MC
	Was excessive information provided?	2±1; 1.5±3	3±2; 2±3	2±0; 2±1	2±1; 2±2	2.103	0.551	
	Was any irrelevant information included?	2±1; 2±1	2±1; 2±1	2±1; 2±2	2±1; 2±2	0.600	0.896	
	Was the medical information accurate?	4±1; 4±1	3±1; 4±2	3±1; 2.5±1	3±0; 3±1	8.565	0.036	MC-ChatGPT
	Were recommendations given?	3±1; 3.5±1	4±1; 4±1	3±1; 3±1	3±1; 2.5±2	9.182	0.027	MC-Bard
	Was patient guidance provided?	3±1; 3.5±1	4±1; 4±0	3±1; 3±1	3±1; 2.5±2	11.977	0.007	Pi-Bard, MC-Bard
	Was a recommendation to consult a physician included?	3±1; 4±2	3±2; 4±3	3±1; 4±2	3±1; 3±0	3.347	0.341	
	Was the response sufficient for the patient?	4±1; 4±1	3±1; 3±2	2±1; 2±1	2±1; 2±2	14.788	0.002	MC-ChatGPT
How often should short stature be monitored?	Did the response aim to inform the reader?	4±0; 4±0	3±1; 4±1	3±1; 3±2	4±1; 3.5±1	11.880	0.008	MC-ChatGPT
	Did the response aim to reassure the reader?	3±1; 3±2	2±1; 2±2	2±1; 2±1	2±1; 2±1	10.577	0.014	Pi-ChatGPT, MC-ChatGPT
	Was a proper definition provided?	3±1; 4±2	3±1; 2±2	2±1; 2±0	3±1; 3±2	6.086	0.108	
	Was all necessary information included?	3±1; 3.5±1	3±1; 4±2	2±1; 2±0	3±1; 3±1	10.044	0.018	MC-ChatGPT
	Was any essential information missing?	3±1; 2.5±3	3±1; 3±0	4±1; 4±0	3±1; 4±2	10.087	0.018	ChatGPT-MC
	Was excessive information provided?	2±1; 2±1	2±1; 2±2	2±0; 2±1	2±1; 2±1	1.419	0.701	
	Was any irrelevant information included?	2±1; 2±2	2±0; 2±1	3±2; 2.5±4	3±1; 2±3	3.018	0.389	
	Was the medical information accurate?	4±1; 4±0	3±1; 3±1	3±0; 3±0	4±1; 3.5±1	10.702	0.013	MC-ChatGPT, Bard-ChatGPT
	Were recommendations given?	4±1; 4±0	4±0; 4±0	3±1; 3±1	3±1; 2±1	18.475	<0.001	Pi-ChatGPT, Pi-Bard, MC-Bard, MC-ChatGPT
	Was patient guidance provided?	4±1; 4±0	4±0; 4±0	4±1; 4±1	2±1; 2±1	14.169	0.003	Pi-ChatGPT, Pi-Bard
	Was a recommendation to consult a physician included?	4±0; 4±0	4±1; 4.5±1	4±0; 4±1	2±1; 2±0	20.644	<0.001	Pi-ChatGPT, Pi-Bard
	Was the response sufficient for the patient?	4±1; 4±1	4±1; 4±1	2±1; 2±0	3±1; 2.5±1	13.867	0.003	MC-ChatGPT, MC-Bard
	Did the response aim to inform the reader?	4±1; 4±1	4±0; 4±0	4±1; 3.5±1	3±1; 3±2	6.194	0.103	
	Did the response aim to reassure the reader?	3±1; 2±2	3±1; 3±3	2±1; 2±0	2±1; 2±1	4.128	0.248	



	Group					Kruskal-Wallis H	p	Post-hoc
	ChatGPT	Bard	MC	Pi	Mean±SD; Med±IQR			
What are the side effects that can be seen after growth hormone treatment?	Mean±SD; Med±IQR	Mean±SD; Med±IQR	Mean±SD; Med±IQR	Mean±SD; Med±IQR	Mean±SD; Med±IQR			
	3±2; 4±3	3±1; 2.5±3	2±1; 2±2	3±1; 2±2	3±1; 2±2	2.667	0.446	
	3±1; 3±2	3±1; 2.5±2	3±1; 3±2	2±1; 2±1	2±1; 2±1	2.281	0.516	
	3±1; 2.5±3	3±1; 3±2	3±1; 2±2	3±1; 4±2	3±1; 4±2	1.942	0.585	
	2±1; 2±1	3±1; 4±2	4±1; 4±0	2±1; 1.5±3	2±1; 1.5±3	10.448	<b>0.015</b>	Pi-MC, ChatGPT-MC
	2±1; 2±2	3±1; 2±2	3±1; 3±1	2±1; 2±2	2±1; 2±2	5.300	0.151	
	4±1; 4±1	3±1; 3±2	3±1; 3±2	3±1; 2±2	3±1; 2±2	5.665	0.129	
	4±0; 4±0	4±1; 4±2	3±1; 3.5±2	2±1; 2±1	2±1; 2±1	17.006	<b>0.001</b>	Pi-ChatGPT, Pi-Bard
Was a recommendation to consult a physician included?	4±0; 4±0	4±1; 4±2	3±1; 3.5±2	2±1; 2±1	2±1; 2±1	17.790	<b>&lt;0.001</b>	Pi-ChatGPT, Pi-Bard
	4±1; 4±1	4±1; 4±1	3±1; 3.5±2	2±0; 2±1	2±0; 2±1	22.334	<b>&lt;0.001</b>	Pi-ChatGPT, Pi-Bard
	3±1; 4±2	3±1; 3±2	3±1; 3±2	2±1; 2±2	2±1; 2±2	7.628	0.054	
	4±0; 4±0	4±1; 4±1	4±1; 4±0	3±1; 3±1	3±1; 3±1	8.965	<b>0.030</b>	Pi-ChatGPT
	3±1; 2.5±2	2±1; 2±1	2±1; 1.5±1	2±1; 2±1	2±1; 2±1	6.925	0.074	
AI: artificial intelligence, MC: Microsoft Copilot, SD: standard deviation, IQR: interquartile range								

lower. MC performed significantly worse than ChatGPT ( $p=0.002$ ), demonstrating ChatGPT's stronger ability to meet users' informational needs. Lastly, in "Response Aims to Inform the Reader" ChatGPT ( $4\pm0$ ) significantly outperformed MC ( $3\pm1$ ) ( $p=0.008$ ), reinforcing ChatGPT's superiority in delivering informative responses (Table 2).

### Evaluation of AI Models in Answering "How Often Should Short Stature be Monitored?"

In the "All Necessary Information Provided" category, MC scored the lowest ( $2\pm1$ ), while ChatGPT and Bard performed better ( $3\pm1$ ). A significant difference was observed between MC and ChatGPT ( $p=0.018$ ), indicating MC's weaker performance in delivering comprehensive information. In "Essential Information Missing" MC ( $4\pm1$ ) had the highest score, showing a greater tendency to omit details, with a significant difference from ChatGPT ( $p=0.018$ ). In "Medically Accurate Information Provided" ChatGPT ( $4\pm1$ ) significantly outperformed MC ( $3\pm0$ ) ( $p=0.013$ ), highlighting ChatGPT's superior accuracy. For "Recommendations Provided" Bard ( $4\pm1$ ) and ChatGPT ( $4\pm0$ ) led, while Pi ( $3\pm1$ ) performed significantly worse ( $p<0.001$ ). In "Response Was Sufficient for the Patient" ChatGPT and Bard ( $4\pm1$ ) excelled, while MC ( $2\pm1$ ) and Pi ( $3\pm1$ ) scored lower. A significant difference was found between ChatGPT and MC ( $p=0.003$ ), emphasizing ChatGPT's stronger ability to meet users' informational needs. Lastly, in "Recommendation to Consult a Physician" Bard ( $4\pm0$ ) and ChatGPT ( $4\pm1$ ) performed best, while Pi ( $2\pm1$ ) had the lowest score, with significant differences between Pi and the other models ( $p<0.001$ ) (Table 2).

### Evaluation of AI Models in Answering "What are the Side Effects that can be Seen after Growth Hormone Treatment?"

In the "Excessive Information Provided" category, MC ( $4\pm1$ ) had the highest score, significantly differing from ChatGPT ( $2\pm1$ ) and Pi ( $2\pm1$ ) ( $p=0.015$ ), indicating MC's greater tendency to provide excessive details. In "Recommendations Provided" Bard ( $4\pm0$ ) and ChatGPT ( $4\pm1$ ) scored the highest, while Pi ( $2\pm1$ ) performed the worst. A significant difference was observed between Pi and the other programs ( $p=0.001$ ), suggesting Pi's limitations in providing recommendations. For "Patient Guidance Provided" Bard ( $4\pm0$ ) and ChatGPT ( $4\pm1$ ) again excelled, while Pi ( $2\pm1$ ) lagged significantly ( $p<0.001$ ), demonstrating Bard and ChatGPT's superior ability to offer guidance. In "Recommendation to Consult a Physician" Bard ( $4\pm1$ ) and ChatGPT ( $4\pm1$ ) performed best, while Pi ( $2\pm0$ ) had the lowest score. Post-hoc analysis confirmed Pi's significantly weaker performance compared to

the other models ( $p < 0.001$ ), reinforcing Bard and ChatGPT's reliability for clinical guidance in this area. Lastly, in "Response Aims to Inform the Reader" Bard ( $4 \pm 0$ ) and ChatGPT ( $4 \pm 1$ ) scored the highest, while Pi ( $3 \pm 1$ ) performed worse, with a significant difference between Pi and ChatGPT ( $p = 0.030$ ), highlighting ChatGPT's strength in delivering informative responses (Table 2).

Table 3 presents the expert evaluation of responses provided by the four AI programs to questions related to pediatric endocrinology.

A statistically significant difference was found the four programs for the "Was a definition provided?" question ( $p = 0.028$ ). The significance was for the differences between MC-ChatGPT and Pi-ChatGPT. For the "Was all necessary information provided?" question, Bard ( $3.5 \pm 0.4$ ) had the highest score, while MC ( $2.7 \pm 0.5$ ) had the lowest. A significant difference was detected between the applications ( $p = 0.002$ ), with differences specifically identified between MC-ChatGPT and MC-Bard. For the "Essential Information Missing?", "Was excessive information provided?", "Was irrelevant information provided?", and "Was the information medically accurate?" questions, the respective  $p$  values were 0.074, 0.178, 0.486, and 0.12, indicating no differences between the AI programs. In the "Were recommendations provided?" question, Bard ( $4 \pm 0.3$ ) had the highest score, while Pi ( $2.4 \pm 0.6$ ) had the lowest. A significant difference was observed between the AI programs ( $p = 0.001$ ), with significant differences identified between MC-Bard and Pi-Bard. Similarly, in the "Was patient guidance provided?" criterion, Bard ( $4.1 \pm 0.3$ ) had the highest average score, while Pi ( $2.3 \pm 0.6$ ) had the lowest. A significant difference was found between the AI programs ( $p < 0.001$ ), with the difference primarily between Pi and Bard. For the "Was a recommendation to consult a physician provided?" question, Bard ( $4.2 \pm 0.5$ ) received the highest score, and a significant difference was identified between Pi and Bard ( $p < 0.001$ ). For the "Was the response sufficient for the patient?" question, Bard ( $3.5 \pm 0.3$ ) and ChatGPT ( $3.4 \pm 0.5$ ) had similar values, receiving the highest scores. A significant difference was found between the AI programs ( $p = 0.004$ ), with differences being identified between MC-ChatGPT and MC-Bard. A significant difference was also detected for the "Does the response aim to inform the reader?" question ( $p = 0.045$ ), with the difference identified between Pi and ChatGPT. Finally, for the "Does the response aim to reassure the reader?" question, ChatGPT ( $2.8 \pm 0.2$ ) had the highest value, and a significant difference was found between the AI programs ( $p = 0.007$ ). The observed differences were between MC-ChatGPT and Pi-ChatGPT. The highest reliability, with an ICC value of 0.774 (0.682-0.844), was observed to the question "Was an appropriate definition provided?", while the lowest reliability, with an ICC value of -0.047 (-0.306-0.197), was observed for the question "Was any recommendation given?" (Table 3).

## Discussion

In this study, responses provided by four different AI programs (ChatGPT, Bard, MC, and Pi) to questions related to pediatric endocrinology concerning short stature were evaluated by experts based on specific criteria. The findings indicated significant differences between the programs included in terms of medical information accuracy, guidance capacity, and user informativeness. The Bard model distinguished itself in categories requiring guidance and direction receiving the highest scores. Moreover, Bard was the model that omitted the least essential information. This suggests that Bard possessed a strong ability to deliver supportive and guiding responses. The literature highlights that AI models focusing on guidance enhance user confidence and support medical decision-making processes (7). MC excelled in providing accurate and medically reliable information. In categories such as "Was medically accurate information provided?" and "Was all necessary information provided?", it performed similarly to or even outperformed Bard and ChatGPT. This suggested that the version of MC tested was a reliable model for areas requiring medical accuracy. However, its lower guidance capacity suggested that it may not be sufficient for clinical applications in the version tested. Published evidence also supports the suggestion that AI programs with strong medical accuracy capabilities may be effectively utilized in clinical decision support systems (8). ChatGPT demonstrated consistent performance in providing information and educating users. It received high scores in the "Does the response aim to inform the reader?" category, highlighting its reliability as an informational source. However, it lagged behind Bard and MC in categories relating to guidance. This suggested that while ChatGPT was effective in knowledge dissemination, it required further development in terms of user guidance. AI applications with strong user education capabilities are known to play an important role in patient education and information dissemination (9). The Pi model exhibited acceptable performance in basic informational categories but received the lowest scores in terms of user guidance and recommendation. This suggests that Pi was inadequate for guidance-focused clinical decision-making processes. AI programs with limited guidance capacities are generally considered more suitable for handling basic queries rather than facilitating detailed information provision (10). Overall, Bard emerged as the most effective model in terms of guidance and recommendations at the time of testing, making it a more suitable AI for specialized fields, such as pediatric endocrinology, where expert guidance is essential. MC was a medically accurate application, but it requires improvement in its guidance capabilities. ChatGPT demonstrated strong informational capabilities, and if its guidance capacity is enhanced, which may have now happened, it could have broader applications. Meanwhile, Pi showed significant limitations in guidance and recommendations,

Table 3. Comparison of expert evaluation averages for AI-generated responses to questions

Questions	ChatGPT	Bard	MC	Pi	Kruskal-Wallis H	p	Post-hoc	ICC; 95% CI (L-U)
	Mean±SD; Median (Min-Max)	Mean±SD; Median (Min-Max)	Mean±SD; Median (Min-Max)	Mean±SD; Median (Min-Max)				
Was a proper definition provided?	3.7±0.3; 3.6 (3.3-4.1)	3.2±0.6; 3.1 (2.7-4.5)	3±0.8; 2.7 (2.1-4)	3.1±0.5; 3.1 (2.5-4.1)	9.066	<b>0.028</b>	MC-ChatGPT, Pi-ChatGPT	0.774 (0.682-0.844)
Was all necessary information included?	3.4±0.3; 3.5 (2.8-3.8)	3.5±0.4; 3.4 (2.9-3.9)	2.7±0.5; 2.8 (1.9-3.3)	2.9±0.5; 3 (2.2-3.5)	14.596	<b>0.002</b>	MC-ChatGPT, MC-Bard	0.523 (0.343-0.664)
Was any essential information missing?	2.7±0.5; 2.6 (2.2-3.9)	2.8±0.3; 2.9 (2.3-3.1)	3.3±0.7; 3.6 (2.2-4)	3±0.5; 3.2 (2-3.8)	6.922	0.074		0.611 (0.463-0.726)
Was excessive information provided?	2.1±0.2; 2.1 (1.8-2.3)	2.4±0.4; 2.4 (1.9-3)	2.4±0.8; 2.2 (1.7-4)	2±0.3; 1.9 (1.7-2.5)	4.912	0.178		0.525 (0.345-0.666)
Was any irrelevant information included?	2.3±0.4; 2.4 (1.7-3)	2.2±0.4; 2.2 (1.6-2.8)	2.6±0.6; 2.7 (1.7-3.3)	2.3±0.2; 2.4 (1.8-2.5)	2.443	0.486		0.536 (0.360-0.674)
Was the medical information accurate?	3.6±0.4; 3.8 (2.8-4)	3.4±0.4; 3.4 (3-4)	3.1±0.4; 3.1 (2.5-3.8)	3.7±1.4; 3.3 (2.6-7.4)	5.842	0.12		0.444 (0.237-0.607)
Were recommendations given?	3.3±0.6; 3.3 (2.1-4.1)	4±0.3; 4 (3.2-4.4)	2.9±0.8; 2.8 (1.8-4.1)	2.4±0.6; 2.5 (1.6-3.2)	17.659	<b>0.001</b>	MC-Bard, Pi-Bard	-0.047 (-0.306-0.197)
Was patient guidance provided?	3.2±0.6; 3.1 (2.3-4.1)	4.1±0.3; 4 (3.6-4.7)	3.1±0.9; 3.4 (1.8-4.2)	2.3±0.6; 2.3 (1.5-3.1)	18.498	<b>&lt;0.001</b>	Pi-Bard	-0.030 (-0.272-0.202)
Was a recommendation to consult a physician included?	3.2±0.6; 3.4 (1.9-4.1)	4.2±0.5; 4.4 (3.4-4.6)	3.1±1.1; 3.2 (1.5-4.5)	2.2±0.6; 2.1 (1.3-3.1)	17.992	<b>&lt;0.001</b>	Pi-Bard	0.049 (-0.181-0.268)
Was the response sufficient for the patient?	3.4±0.5; 3.5 (2.1-3.9)	3.5±0.3; 3.6 (3-4)	2.6±0.6; 2.5 (1.8-3.7)	2.8±0.6; 2.7 (2-3.6)	13.305	<b>0.004</b>	MC-ChatGPT, MC-Bard	0.341 (0.112-0.527)
Did the response aim to inform the reader?	3.9±0.4; 4.1 (2.9-4.4)	3.9±0.3; 4 (3.4-4.4)	3.6±0.4; 3.5 (2.8-4.2)	3.5±0.4; 3.4 (3-4.1)	8.048	<b>0.045</b>	Pi-ChatGPT	0.364 (0.135-0.548)
Did the response aim to reassure the reader?	2.8±0.2; 2.8 (2.5-3.1)	2.8±0.7; 2.4 (2.3-4.1)	2.2±0.4; 2.2 (1.6-2.7)	2.3±0.4; 2 (1.8-2.9)	12.059	<b>0.007</b>	MC-ChatGPT, Pi-ChatGPT	0.523 (0.345-0.663)

AI: artificial intelligence, MC: Microsoft Copilot, SD: standard deviation, Min-Max: minimum-maximum, ICC: intraclass correlation coefficient, CI: confidence interval, L: lower bound, U: upper bound

making the version tested insufficient for clinical applications requiring decision support.

The findings of this study highlight the strengths and weaknesses of different AI programs and shed light on their potential applications in medical decision-making processes. For instance, Bard, with its strong guidance capacity, could be beneficial in patient management, while MC may be more effective in areas that require medical accuracy. ChatGPT stood out as a suitable model for patient education and general information sharing.

### Study Limitations

The answers produced may have changed due to the updating of the AI programs used in our study. The answers of the experts making the evaluations may be subjective. The lack of real patient data in our study can be considered as a limitation. More studies are needed for the integration of AI in clinical applications. The fact that AI programs are subject to rapid change and are constantly evolving may lead to differences in the results of the study if the same analysis was performed now.

### Conclusion

This study demonstrated that different AI programs exhibited varying performances in the field of pediatric endocrinology at the time of the study. The Bard model excelled in guidance and recommendation categories, while MC proved to be strong in medical accuracy. ChatGPT emerged as a reliable option for information dissemination and user education, whereas Pi showed limited applicability in this domain, due to its insufficient guidance capacity. Future research should focus on improving AI models to achieve a more balanced performance in both guidance and medical accuracy. In addition, optimizing these programs to align with user needs is recommended to enhance patient trust and integrate AI effectively into clinical decision-support processes. Evidence has shown that while AI holds great potential in supporting patient care processes, this potential can only be fully realized through a careful balance in model design (11,12). These findings underscore the need for development of customized AI solutions, involving both software developers and experts in the field to produce programs tailored to the needs of specialized subjects, such as pediatric endocrinology.

#### Ethics

**Ethics Committee Approval:** This study does not require ethical committee approval.

**Informed Consent:** All experts agreed to participate.

#### Footnotes

##### Authorship Contributions

Concept: Kamber Kaşali, Özgür Fırat Özpolat, Design: Kamber Kaşali, Özgür Fırat Özpolat, Merve Ülkü, Data Collection or Processing: Kamber Kaşali, Ayşe Sena Dönmez, Serap Kılıç Kaya, Esra Dişçi, Serkan Bilge Koca, Analysis or Interpretation: Kamber Kaşali, Özgür Fırat Özpolat, Atilla Çayır Literature Search: Kamber Kaşali, Ufuk Özkaya, Writing: Kamber Kaşali, Hüseyin Demirbilek.

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