

IMPACT ON EFFECTIVENESS OF USER PERCEPTION ON USING CHATGPT

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ABSTRACT

The advent of AI technologies like ChatGPT has ushered in a new era of human-computer communication. This paper delves into a critical aspect of this interaction the impact of user perception on the effectiveness of ChatGPT. While the technical prowess of ChatGPT is evident, its success hinges on how users perceive its responses, abilities, and limitations. The abstract explores the intricate relationship between user perception and the system's performance, discussing how accurate responses, contextual understanding, transparency, and ethical considerations play pivotal roles. The study emphasizes the significance of user satisfaction, trust, and acceptance, while also acknowledging the challenges posed by system limitations and potential misunderstandings. By examining the correlation between user perception and ChatGPT's effectiveness, this abstract provides valuable insights into optimizing human-AI interactions for enhanced user experiences.

Keywords- ChatGPT; Chatbot; AI; RLHF; Natural Language; NLP; Open AI; Bard; SFT Model; RM Model; PPO.

1. INTRODUCTION

In the rapidly evolving landscape of artificial intelligence, ChatGPT has emerged as a prominent example of natural language processing technology. Designed to engage in human-like conversations, ChatGPT has the potential to revolutionize the way we interact with machines. However, its effectiveness hinges not only on its technical capabilities but also on how users perceive and interpret its responses. The user's perception of influences their satisfaction, trust, and willingness to engage with the system. This interplay between user perception and system performance ChatGPT's performance, reliability, and alignment with their needs directly underscores the importance of understanding the impact that user perception can have on the utilization and acceptance of ChatGPT.

2. REVIEW OF LITERATURE

Conversational AI systems that use large language models like ChatGPT have become increasingly popular in recent years. These systems are capable of generating human-like responses to user queries and interactions, but there are concerns about the privacy and security of user data in these systems, particularly given the complexity of the algorithms and machine learning techniques used to generate responses. A survey of 500 users conducted by the consulting firm Capgemini found that 56% of respondents were comfortable interacting with virtual assistants that use large language models, but users also had concerns about the accuracy and transparency of these systems, as well as about the collection and use of their personal data (Capgemini, 2019). A study by researchers at the University of California, Berkeley found that some chatbots using large language models could be exploited by attackers to extract sensitive information from users. The study also identified issues with the transparency and explain ability of these systems, which could make it difficult for users to understand how their data is being used (Li et al., 2020). There are also ethical considerations involved in the use of ChatGPT in conversational AI systems. For example, there are concerns about the potential for these systems to perpetuate biases and stereotypes, particularly if they are trained on datasets that contain biased or incomplete information. There are also concerns about the accountability of these systems, particularly in cases where they make decisions that have real-world consequences (Holstein & Schultz, 2019).

3. METHODOLOGY

This study used a descriptive statistics method to investigate user perceptions and attitudes towards conversational AI systems that use ChatGPT. A survey was administered through google form to a sample of 101 users to gather the data. The quantitative data collected in this study was analyzed using descriptive statistics, such as frequencies and percentages, to summarize the characteristics of the sample and identify patterns in user attitudes towards the use of Chat GPT- Impact on effectiveness of user perception on using chatgpt.

To know the impact of using chatgpt.

To know the awareness of chatgpt among public.

To identify the factors affecting using chatgpt.

4. ANALYSIS

1. PERCENTAGE ANALYSIS

TABLE 1

		Gender			
		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	Male	59	59.0	59.0	59.0
	Female	41	41.0	41.0	100.0
	Total	100	100.0	100.0	

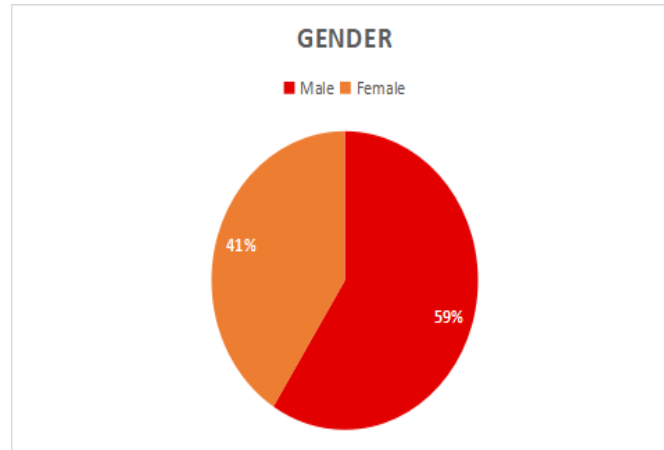


Figure: 1

Interpretation: Form the above table it is interpreted that 59% are male and 41%of respondents are female majority (59) are male.

2. AGE FOR RESPONDENTS

TABLE 2

		Age			
		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	18-24	77	77.0	77.0	77.0
	24-34	9	9.0	9.0	86.0
	35-40	5	5.0	5.0	91.0
	Above40	9	9.0	9.0	100.0
	Total	100	100.0	100.0	

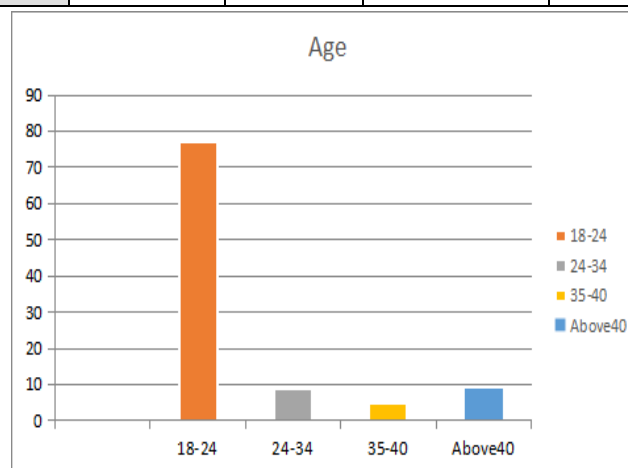


FIGURE: 2

INTERPRETATION: From the above table it is interpreted that 77 are 18-24 years, 9 are 24-34 years, 5 are 35-40 years and 9 are above 40 and above. Majority (77) are 18-24 years.

3. EDUCATIONAL QUALIFICATION

TABLE 3

Education background					
		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	High school	14	14.0	14.0	14.0
	Bachelor's degree	24	24.0	24.0	38.0
	Master's degree	53	53.0	53.0	91.0
	Professional course	9	9.0	9.0	100.0
	Total	100	100.0	100.0	

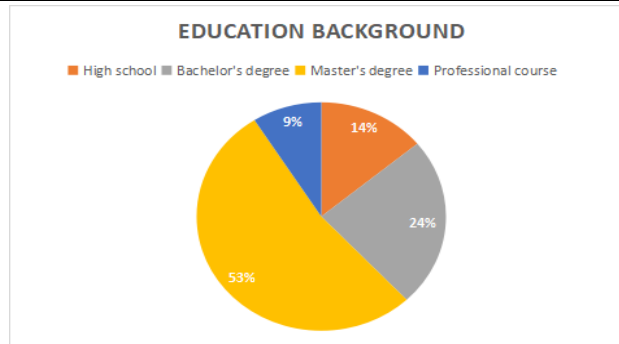


FIGURE:3

INTERPRETATION: From the above table it is interpreted that 9 high school are 24 Bachelor's degree, 14 are Master's degree, 53 are Professional degree, majority (53) master's degree.

4. EMPLOYMENT STATUSES

TABLE 4

Occupation					
		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	Student	68	68.0	68.0	68.0
	Empolyee	20	20.0	20.0	88.0
	Umempolyee	3	3.0	3.0	91.0
	Retried	3	3.0	3.0	94.0
	Homemaker	6	6.0	6.0	100.0
	Total	100	100.0	100.0	

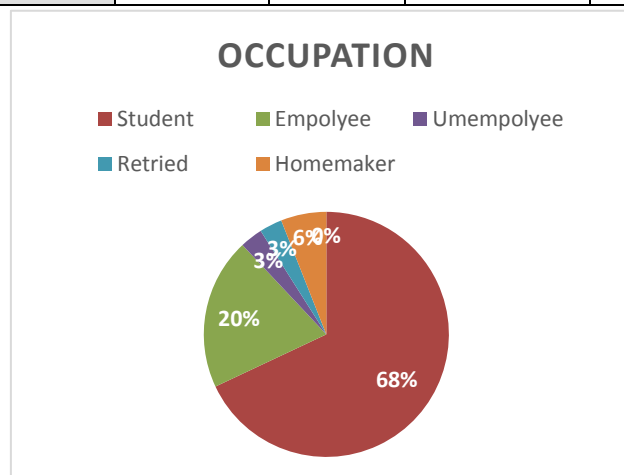


FIGURE:4

INTERPRETATION: From the above table it is interpreted that 68, student, 20, empolyee, 3, unemployed, 3, retried, 6homemaker and, majority (68) student.

5. ANNUA-LINCOME

TABLE 5

Income (per month)					
		Frequency	Percent	Valid Percent	Cumulative Percent
Valid	Below 10000	50	50.0	50.0	50.0
	10000-30000	38	38.0	38.0	88.0
	30000-50000	9	9.0	9.0	97.0
	Above 50000	3	3.0	3.0	100.0
	Total	100	100.0	100.0	

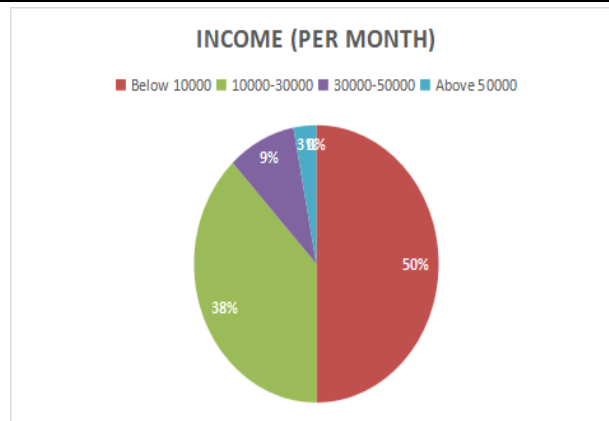


FIGURE:5

INTERPRETATION: From the above table it is interpreted that Below 100000, Are 50 Rs100000-300000,38 are Rs 300000-500000 are 9 and 500000 Above are .3 Majority (50) are below10000.

ONEWAY ANOVA

One-Way Analysis of Variance (ANOVA) is a statistical technique used to compare means across multiple groups. It's commonly used when you have one independent variable (also known as a factor) with more than two levels , and you want to determine if there are any significant differences in the means of a dependent variable among those levels.

Inference

To find out the difference between the frequency of using chat gpt and how respondents came to know about chat gpt.

Null Hypothesis

Ho: There is no difference between the frequency of using chat gpt and how respondents came to know about chat gpt

Alternative Hypothesis

H1: There is difference between the frequency of using chat gpt and how respondents came to know about chat gpt

TABLE 6

ANOVA					
How did you come to know about chatgpt					
	Sum of Squares	df	Mean Square	F	Sig.
Between Groups	7.376	4	1.844	1.976	.104
Within Groups	88.664	95	.933		
Total	96.040	99			

Chi-Square Tests

The Chi-Square test is a statistical method used to determine if there is a significant association or relationship between categorical variables. It's commonly used to analyse data in the form of frequency counts or proportions in contingency tables. The Chi-Square test assesses whether the observed frequencies in the table differ significantly from the expected frequencies, assuming that there is no association between the variables.

Inference

To find out the association between the satisfaction level of the respondent and recommend of chat Gpt

Null Hypothesis

H0: there is no the satisfaction level of the respondent and recommment of chat Gpt

Alternative Hypothesis

H1: there is the satisfaction level of the respondent and recommment of chat Gpt

TABLE 7

Chi-Square Tests			
	Value	df	Asymptotic Significance (2-sided)
Pearson Chi-Square	4.698 ^a	3	.195
Likelihood Ratio	5.014	3	.171
Linear-by-Linear Association	1.514	1	.219
N of Valid Cases	100		

a. 2 cells (25.0%) have expected count less than 5. The minimum expected count is .64.

5. FINDINGS

According to the study, 59% are male and 41% of respondents are female. Majority (59) are male. According to the study, 77 are 18-24 years, 9 are 24-34 years, 5 are 35-40 years and 9 are 40 above. Majority (77) are 18-24 years. According to the study, 9 high school, 24 Bachelor's degree, 14 are Master's degree, 53 are Professional degree, majority (53) master's degree. According to the study, 68, Student, 20, Employee, 3, Unemployed, 3, Retired, 6 Homemaker and, majority (68) Below 100000, 38 are 50 Rs 100000-300000, 9 are Rs 300000-500000 and 3 are 500000 Above. Majority (50) are below 10000.

6. CONCLUSION and SUGGESTIONS

Chat GPT is all that was used under the age group of 18-24 with a percentage of 77.0% from this we can must use the age group of between 18 and 24 can look for Chat GPT and mainly user on Chat GPT and 59% are male and 41% of female use on Chat GPT use and student 68% that was must and more use on chat gpt and this is user free Chat bot. The advent of AI technologies.

7. REFERENCE

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