

Digital Transformation: AI-Powered Bot Solutions and Automation for Customer Services

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Abstract—Recent years have witnessed a growth in the adoption of AI-Powered bot solutions in many industries, including healthcare, education, manufacturing, and energy. The drive behind this adoption is to increase productivity and improve customer engagement by enhancing the communication between customers and customer service support. These solutions are at the heart of Digital Transformation, which is focused on introducing new ways for value creation, automation, interaction, transparency, and control. Furthermore, delivers customer benefits: convenience, relevance, experience, empowerment, and savings. This drive has led the IT industry to deliver digital systems for providing customer services aimed at increased customer satisfaction. Virtual Assistant agents are used by organizations to fulfill different tasks related to customer services for solving problems and providing recommendations. Among AI-Powered bot solutions, an EmailBot is an NLP-based tool used to automate routine Email conversations and enhance positive interaction with customers. This paper explores the ability to implement an EmailBot solution on a private cloud environment to automate frequently performed tasks such as application installation and upgrades, how-to, obtaining application links, permission requests, and inquiries related to request statuses. An evaluation conducted compared the automated solution capabilities of infrastructure high-availability and with traditional customer service support that relied wholly on humans for defined tasks. According to resulting measures, the automated model outperformed the traditional model in terms of response time and accuracy, with drastic improvement in the former and slight improvement in the latter.

Keywords—Digital Transformation, AI-Powered Bot, EmailBot, Cloud, Customer Service Automation

I. INTRODUCTION

Oil and Gas (O&G) industry is capital-intensive and involves the exploitation of hydrocarbons, including exploration, field development, production, and transportation. It is highly regulated and requires significant investment in infrastructure and technology to overcome challenges faced by operators, service companies, equipment manufacturers, technology providers, and stakeholders.

Recently, the O&G industry has undergone a significant transformation, driven by embracing digital technologies and strategies such as the Internet of Things (IoT), Artificial Intelligence (AI), Cloud, Edge Computing, Cybersecurity, and Digital Twin. These technologies offer superior agility and reliability in O&G industry operations and services to meet energy demands and discovery by adopting AI tools and solutions. DT has revolutionized the industry by enhancing operational efficiency, effective decision-making processes, availing a platform for innovation and growth, reducing costs, improving safety, exploring new opportunities and business

models, and enhancing customer service response [1]. For example, employing cloud technology has led to maximizing the availability of resources, allowing virtualization to achieve optimal resource utilization and significantly enhance infrastructure scalability. Thus, making a compelling case to incorporate cloud technology to provide exceptional services and effectively oversee virtualized data centers.

Overall, the impact of digital transformation in the O&G industry is noteworthy, particularly in the area of customer service. O&G industry's customer service is a crucial aspect requiring a high level of expertise and knowledge to manage customer inquiries, complaints, and requests for information. Using AI-powered bot solutions and automation has increased efficiency and improved customer satisfaction.

A. Customer service support challenges

In today's competitive global market, organizations retain customer loyalty by setting excellent customer service experience as a strategic objective. Customer service support usually handles repetitive tasks and resolves issues that have been previously addressed [2]. The primary mode of interaction between customers and customer service support remains human-to-human, either through phone calls, emails, or face-to-face communication. This leads to more personalized and empathetic conversations. This approach relies on the availability of customer support, which is typically limited to normal working hours, preventing the customers from gaining provision of services outside business days/hours. Organizations have different applied models of customer service support, for example, having multi-levels of customer support based on domain and technology [3]. Traditional customer service support struggles to handle a considerable number of inquiries without incurring resolution time in addition to a limited knowledgebase (one-stop-shop) relying on experts to manually update for any newly reported issue [2][4]. These approaches are leading to further delays in issue resolution [3], Fig 1 illustrates the traditional customer support model. Customer service is expensive and necessitates significant investment in training and resources [4]. Organizations are concentrating on raising the level of customer services through enhancing interaction with the customers. They aim to provide exceptional customer service that exceeds expectations [3][4].

In summary, these approaches have their fair share of drawbacks, such as slower task completion time and increased complexity of resolving technical issues. There is no guarantee that all customers will receive immediate and accurate service, which leads to frustration and dissatisfaction, and negatively impacts business operations. The challenges

are due to dependencies on human expertise, knowledge, availability, and means of preferred communication [2].

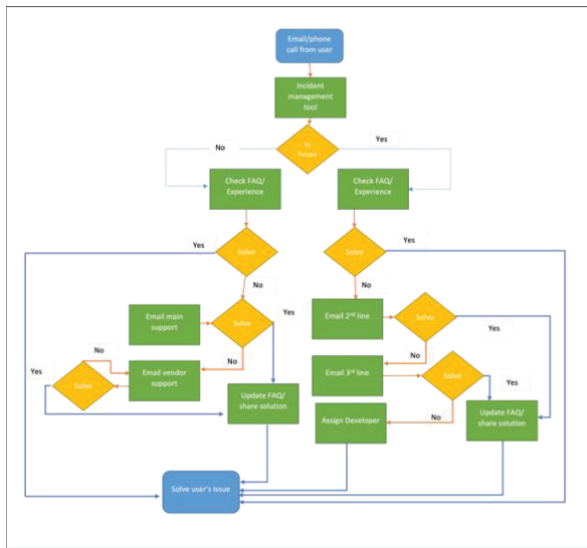


Fig. 1. Example of a figure caption.

The evaluation conducted set out to demonstrate whether AI-Powered bot capabilities can play a crucial role in solving costly IT customer support routine tasks to achieve service-level objectives and meet expected customer satisfaction benchmarks. AI-Powered Bots are used to perform several customer support tasks to answer inquiries, provide recommendations, and solve reported problems. Adopting and deploying customer service AI-Powered Bot automation applications can provide competitive measurements in service accuracy and reduced response time in customer service compared with the traditional customer support model.

The rest of the paper is organized as follows. Section 2 reviews related topics and provides context for business motivation. AI-Powered Bot deployment is proposed in Section 3. Section 4 discusses an experimental setup, results, and analysis. The last section concludes the paper.

II. BACKGROUND

A. Digital Transformation (DT) and Industrial Revolution 4.0 (IR 4.0)

Digital Transformation (DT) and Industrial Revolution 4.0 (IR 4.0) are closely related concepts transforming organizations' operations. DT refers to utilizing disruptive technologies to optimize products and services, sustain economic growth, and enhance the user experience [5]. This process significantly changes an organization's structure, capabilities, business models, and product development methods. The selection and implementation of digital technologies should align with a corporation's strategic goals and consider technology, value creation, organizational structure, and financial abilities [5]. DT often aims to monetizing, and leveraging data-driven technologies through creating new services, products, and platforms that improving customer experience [5].

The benefits of DT enclosed increased profitability, competitiveness, and efficiency by developing new ways of conducting business, creating software and systems, and modifying existing processes [5]. DT requires organizations to adjust their strategies, methods, structures, culture, and

business models, leading to an empowered workforce, innovation, and personalized customer experiences [6].

The Industrial Revolution 4.0 is a new era of industrialization driven by advanced Digital Transformation technologies [7]. It is a complete transformation of the entire value creation process, from the initial design and planning stages to the final production and delivery of goods and services [8]. IR 4.0 leverages cutting-edge technologies such as IoT, AI, Advanced Robotics, Autonomous Vehicles, Cloud Computing, Augmented Reality (AR), Simulated Reality, distributed databases, and Digital Twin technology [5][7]. These technologies are integrated into production to create a seamless and highly efficient system. IR 4.0 technologies enable supply chain automation, employing advanced digital systems to optimize sourcing raw materials and delivering finished products. Automating the many processes allows the system to operate more efficiently, reducing costs and improving performance [10]. Optimizing energy consumption, materials, human resources, and proactive management prevent problems due to manual or flawed systems [8]. Real-time data systems allow for transparency in the production processes which is vital for review and control [10].

The concepts of DT and IR 4.0 aim to promote innovation and change in business through technology. However, DT is more focused on using digital technologies to enhance existing processes and operations. IR 4.0 offers a broader perspective when analyzing the application and influence of these technologies on businesses. Moreover, DT is geared toward organizations and their efforts to stay competitive and adjust to market shifts, while IR 4.0 has the potential to transform entire industries.

DT fueled by IR 4.0 technologies envision an intelligent factory of the future where a self-improving system operates autonomously under conditions in real-time or close to real-time. It has the potential to introduce entire production processes autonomously, optimizing operations and minimizing human intervention [9].

Organizations that incorporate disruptive technologies such as big data, cloud computing, mobility, and social technologies into their infrastructure have witnessed significant revenue and market value growth compared to their counterparts [9]. This growth can be attributed to their strategic vision. Therefore, the trend of adopting these technologies is expected to continue and have a significant impact on organizations that embrace DT toward IR 4.0 [6].

B. AI-Powered Bots

An AI-Powered Bot is a software application that runs automated tasks which is empowered by a combination of Artificial Intelligence techniques such as Natural Language Processing (NLP), Machine Learning (ML), and Deep Learning (DL) to interact with users and provide them with personalized responses [4][11].

AI-Powered Bots have experienced a surge in popularity over recent years, proffering numerous advantages for businesses and consumers [11][12]. They are designed to perform specific tasks, such as answering FAQs, scheduling appointments, and processing orders. They can provide round-the-clock customer service, improve response time, elevate user experience, and alleviate the workload of human resources. With the evolution of AI technology, it is

anticipated that the future will witness the emergence of more sophisticated categories of AI-Powered Bots [11]. AI-Powered Bots encompass a variety of types, including ChatBots, EmailBots, and Voice Recognition systems.

1) ChatBot

ChatBots are AI-powered bots that use NLP and ML algorithms to converse with humans naturally [13][14]. In addition, it utilizes textual or text-to-speech modalities to emulate human conversation within online chat interfaces [14]. They can handle various customer service inquiries, such as answering frequently asked questions, resolving technical issues, and providing personalized recommendations [14]. ChatBots can engage in natural language conversations with users, interpret their intentions, and generate responses utilizing predetermined rules and data [13][14]. The development of ChatBot systems requires ongoing refinement and experimentation to emulate natural human conversation accurately. A significant number of individuals in the production process are unable to communicate verbally [14]. ChatBots are utilized in dialog systems to provide customer support, facilitate request routing, and gather information [14].

2) EmailBot

EmailBot is an AI-Powered Bot tool that shares similarities with ChatBots. It focuses on email automation tool that utilizes AI to enhance the efficiency and effectiveness of communication [14]. Using AI and contextual analysis achieves delivering personalized email responses through an automated process using a classifier. These Emailbots rely on scanning for general keywords and generating responses using common phrases stored in a library or database. EmailBot resembles a ChatBot, employing NLP and ML algorithms to comprehend the context and intent of incoming emails and subsequently furnish appropriate responses [14]. EmailBot can schedule meetings, respond to commonly asked inquiries, and dispatch pre-programmed follow-up messages [14].

3) Voice Recognition

Voice Recognition is an AI-Powered Bot tool that analyzes and interprets human speech [15]. Advanced algorithms and software interpret spoken words, phrases, and sentences. The system analyzes phonemes and syllables in speech to identify what is said [15]. AI systems match sounds and patterns to a massive library of words and phrases to correctly detect and transcribe spoken language [15]. Virtual assistants, voice-enabled home automation systems, speech-to-text transcription employ voice recognition technology. As AI technologies evolve, they contribute to changing computers and other devices' uses [15].

C. Cloud Computing

Cloud computing combines computer services, servers, databases, networks, software, storage, and analysis in its most basic form [7]. The cloud is an essential enabler of IR 4.0 [7]. Cloud computing is a crucial enabler of digital transformation, leveraging digital technologies to transform business operations, improve customer experiences, and drive innovation [8][9]. In digital transformation, cloud computing and its services provide organizations with scalability, flexibility, high availability, competitive advantage, perceived cost-effectiveness, and service quality as an essential part of the modern IT infrastructure for businesses [7]. Cloud system uses virtual machine technology hosted on virtualized

physical resources [7]. This concept provides self-contained computer resources that ease the operation of virtualized networks, servers, and data processing [7].

Three main types of cloud computing can support DT [7][9] [16]:

Public Cloud: Public cloud services are offered by third-party cloud service providers. These services are accessible online and can be used by any organization or individual. Public cloud services are ideal for organizations that need scalable, flexible, and cost-effective computing resources but want to avoid investing in and managing their own IT infrastructure.

Private Cloud: Private cloud services are hosted and managed by internal resources in the organization itself or a third-party service provider, designed for the exclusive use of a single organization. Private cloud services offer greater control, security, and customization than public cloud services, but they require more investment in IT infrastructure.

Hybrid Cloud: Hybrid cloud services combine the benefits of public and private cloud services, enabling organizations to use a mix of on-premises IT infrastructure, private cloud services, and public cloud services to meet their specific needs. Hybrid cloud services offer greater flexibility and scalability than private cloud services and provide greater control and security than public cloud services.

Cloud computing services can play a vital role DT by providing organizations with the computing resources needed to develop and deliver innovative digital services and applications. There are three main types of cloud computing services:

Infrastructure as a Service (IaaS): IaaS provides users access to virtualized computing resources, such as servers, storage, and networking infrastructure, hosted by a cloud service provider. Users can deploy and run their applications and services on this infrastructure while the cloud service provider manages the underlying hardware.

Platform as a Service (PaaS): PaaS provides users with a complete platform for developing, testing, and deploying applications, without worrying about managing the underlying infrastructure. The cloud service provider takes care of the operating system, runtime environment, and other middleware components while the user focuses on developing and deploying their applications.

Software as a Service (SaaS): SaaS provides users access to fully functional applications hosted and managed by a cloud service provider. Users can access the applications online, usually through a web browser or a mobile app, without worrying about installing, maintaining, or updating the software.

D. Customer Service Automation

Customer service automation uses technologies like AI-Powered Bots to automate and streamline customer service interactions [17]. Customer services support routine tasks which encompass responding to customer inquiries, providing support, and resolving issues [17]. Customer service automation aims to expedite and optimize the process of customer assistance while concurrently alleviating the workload for customer service support [17] [18]. Organizations can free up their human resources by automating routine tasks and processes to focus on more complex and high-value customer interactions [17]. The

ability of organizations to assist their consumers 24/7 is one of the primary advantages of implementing automation into their customer service processes [17]. EmailBot, ChatBots, and Voice Recognition are used to answer frequently asked questions and handle problems outside regular business hours. This assists in enhancing customer satisfaction and minimize the time it takes for a response [17].

Overall, customer service automation helps companies improve their customer service operations' efficiency and effectiveness while improving the customer experience [17] [18].

III. PROPOSED CUSTOMER SERVICE AUTOMATION

To test the capabilities of customer service automation, an EmailBot was selected among the other AI-Powered Bots tools due to the reliance on emails as the official communication means of day-to-day interaction within an organization. The EmailBot application was deployed via a PaaS environment hosted on a private cloud in an enterprise data center.

A. EmailBot Infrastructure

For the purposes of this research, a private cloud environment was setup according to industry standards based on full-stack enterprise architecture. This design was constructed on a multi-layer infrastructure, including a cloud manager, monitoring, access management, virtual network, virtual storage, virtualization, and physical hosts. Each of these layers works together to complement the others.

The adoption of a private cloud computing type, which rigged our infrastructure with main qualities due to its built-in features that avail better resource utilization, managed traffic, monitoring tools, secured, controlled data access, high availability, and recovery for the hosted environments [19][20]. EmailBot is a cloud-native application deployed based on PaaS cloud services. PaaS design includes IaaS servers as well. IaaS is an abstraction of underlay infrastructure consisting of a pool of compute resources (servers, storage, and networks). PaaS abstracts the deployed cloud-native applications from their dependencies [16] [20]. Additionally, IaaS and PaaS allow customers to automatically have their applications and resources assigned due to the extensive integration capabilities. EmailBot infrastructure leveraged the benefits of cloud implementation design, which uses a secured connection to the internal database to store the trained data sets. In addition, the EmailBot application has high availability due to cloud infrastructure redundancy [19]. The deployed infrastructure is illustrated in Fig 2.

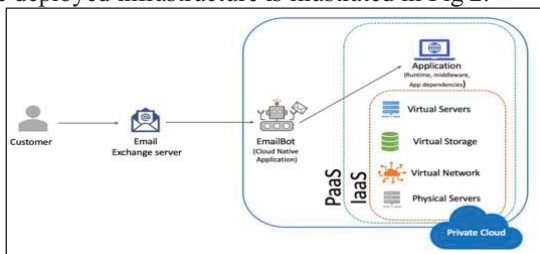


Fig. 2. Deployed EmailBot Infrastructure.

B. EmailBot Methodology

Email is a ubiquitous and essential way of communication, as proven by its widespread use [21]. An inbox is often filled

with different emails, making it challenging to keep an eye on critical emails and organize them in the proper classification. In 2019, 246 billion emails were exchanged in one day [22]. Of these emails, 128.8 billion were business emails, while 117.7 billion were consumer emails [23].

NLP is the field of computer science focusing on understanding the intent, sentiment, and context of human language using computational methods [24][23]. NLP's efficacy in classifying emails depends on the number of emails utilized to train the ML model [25]. The number of emails used as a trained dataset can be attributed to an improved comprehension of the purpose and contexts surrounding the human language utilized within the organization.

ML algorithms classify the emails using the trained dataset to develop a language model. Consequently, the trained datasets must cover the most frequent inquiries to train the deep learning model. As the number of trained datasets increases, the quality of the model improves and enhances the accuracy of the classification, as illustrated in Fig 3.

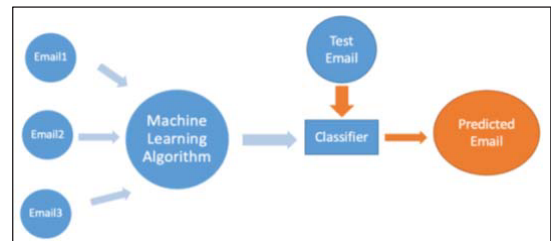


Fig. 3. ML Training Algorithm for Email Classification.

C. Traditional Support Infrastructure

The comparative traditional customer support showed a significant dependency on first-line support, where support personnel access multiple systems and knowledgebases deployed independently. Each system has its own design and access requirements.

IV. EXPERIMENTAL RESULTS

In this section, we discussed experimental measurements and results.

A. Testing Methodology

Experiments were designed to evaluate two aspects: email response time and response accuracy. The response time metric is essential for customer service support, where the response email to customer inquiries/problems must be delivered swiftly. However, the response accuracy metric is important to provide the right response and gain customers' trust and satisfaction. These two measures are critical to customer service support to reduce the man-hours through automation compared to the traditional. Each test is conducted over twelve weeks. There are two key motivations for repeating the tests. First, to evaluate the two support models focused on installing a new application and obtaining the correct and appropriate accesses/roles for the customers' authority to use the application. It is noteworthy that these two inquiries are the most received by the support team. Secondly, to ensure that the automated process proves accepted response time and accuracy.

B. Process

1) Traditional Customer Support

The workflow of the traditional customer support team starts from receiving the email by the first-line support, whose task is to view the content of the email and determine the type of inquiry or problem. Thereby determining the path of the inquiry or problem based on the difficulty by opening a ticket through the incident management system, either by addressing it through the first-line support or reassigning the ticket to a specialist support member. Subsequently, the user is contacted to ensure the resolution of the problem or inquiry has been adequately answered, and then the ticket is closed, as shown in Fig 4. Additionally, the traditional customer service workflow was amended based on the selected inquiry due to the need for higher privileges, experience, or the process under the responsibility of other support teams, as presented in Fig 4, 5, 6. Whenever an application installation request email reaches customer support, first-line support will open a ticket and forward it with the installation link to the system admin for approval and granting access. After that, the system admin will confirm and share the link with the requester via email. The demonstrated process involved two layers for approval and confirmation that can be disposed of and streamline the process, as shown in Fig 5.

In case of requesting the appropriate accesses/roles to use an application, first-line support examines access/role requests and verifies the legitimacy of the requester. This is to ensure that only those with permission to use the application are granted access. Additionally, first-line support performs verification on other platforms to find any missing privileges. In case of missing privileges, the request will be forwarded to the access management team for approval and initiate a workflow to obtain the missing privileges. The access management team will update the requester once the requested privileges are permitted, as shown in Fig 6.

2) Automated Customer Support

In automated customer service support, those two processes are managed by the deployed AI-Power EmailBot. The process of requesting the installation of the application begins with receiving the email where EmailBot classifies them, initiates a ticket with the request details through incident management (as the integration has been implemented), and then verifies the customer's authorization to use this application (as the integration has been implemented with the designated platform, where the inquiries have been defining in advance). After the verification, an email has the application installation link is shared with the customer, as shown in Fig 7. This happens in the same vein as if the request is for the grantee role or access permission but with some difference. This process starts from receiving the email where EmailBot, classifies them, opens a ticket of the request, verifies the missing privileges through the integration with the designated platform, and then shares a detailed email with the customer that has the access acquirement process to obtain the required approvals, as shown in Fig 8.

C. Response time

Response time is the time difference from sending the email by the customer until the support team provides the required solution. It is one of the factors considered by any support team, where the shorter response time indicates the strength of the support team. In this section, a comparison of two support methods is shown by calculating the average

response time for both inquiries per week, as shown in Fig 5,6,7,8,9,10 and Tables I, III.

The response time of the traditional support team is affected by many factors, including the number of emails that reach the first-line support in a limited period, sufficient information in the request, the type of request, the experience of the first-line support team, and working days/hours. This explains the vast difference between the average response time per week for the traditional support team. Most of the tasks of the EmailBot are carried out in parallel, significantly improving response time. The average response time of a virtual assistant is greatly affected by the synchronization process between it and mail servers. The difference between the highest and lowest peak of the average time is acceptable and is approximately 8 seconds. The difference between the highest and lowest peak of the average time is acceptable. It is about 8 seconds, compared to the difference between the highest and lowest peak of the average time, the average response time for the same period of the traditional support team, which is approximately 1,140 seconds (~19 minutes).

D. Accuracy

Accuracy is measured the how many times the ML model was correct overall. It is another factor that is considered by any support team. It is a direct process where the higher accuracy rate indicates the strength of the support team. This section compares the two support methods to calculate the average accuracy for both inquiries per week, as shown in Fig 5,6,7,8,9,10, and Tables II, III.

$$\text{Accuracy(per week)} = \frac{\text{Total Number of Correct Predictions}}{\text{Total Number of Predictions}}$$

The accuracy rate for the traditional support team is affected by many factors, as mentioned earlier in the response time section, including the number of emails that reach the first-line support in a limited period, sufficient information in the request, the type of request, the experience of the first-line support team, and working days/hours. This explains the vast difference between the average accuracy rates per week for the traditional support team, as illustrated in Fig 11 and Table II. EmailBot is presenting an improvement in accuracy rate during the period. The average accuracy of a virtual assistant is greatly affected by the number of emails used as a trained dataset. In addition, the trained datasets must cover the most frequent inquiries to train the deep learning model. As the number of trained datasets increases, the quality of the model improves and enhances the accuracy of the classification, as illustrated in Fig 11 and Table II.

In some cases, it was not confirmed that users have the required permissions that would enable them to use the application without any problem, which led to contact the support team again. A few cases were also not appropriately classified due to the lack of information and trained data. The results of the experiment are summarized in Table III, which represents the interaction types and the identified response time for both traditional and automated support methods. It demonstrates the accuracy average, the integration with other systems, and the underlay infrastructure design.

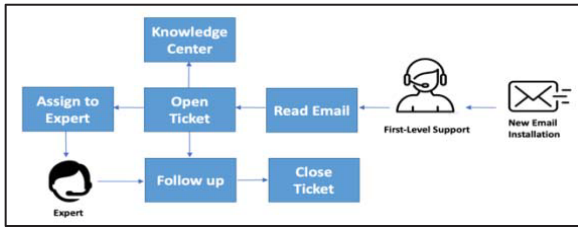


Fig. 4. General Traditional Model Workflow.

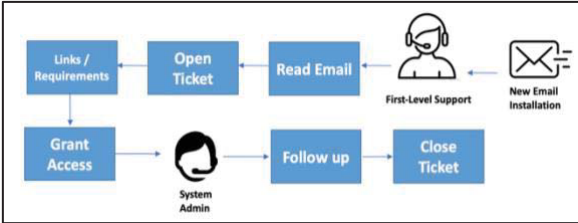


Fig. 5. Traditional Model (Installation Process).

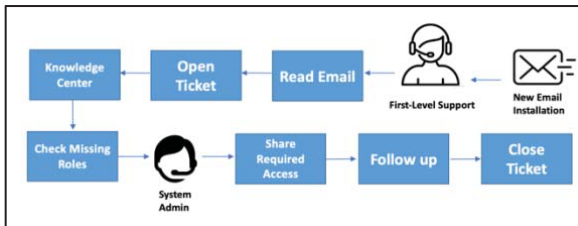


Fig. 6. Traditional Model (Access Process).

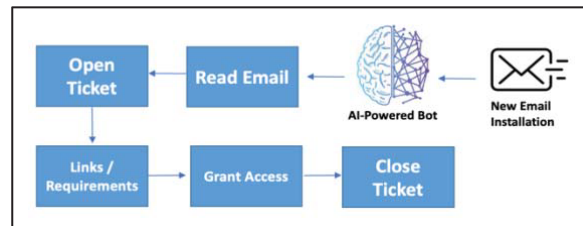


Fig. 7. Automated Model (Installation Process).



Fig. 8. Response Time of Traditional Model (Human).

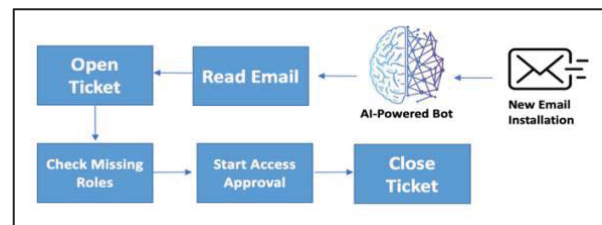


Fig. 9. Automated Model (Access Process).

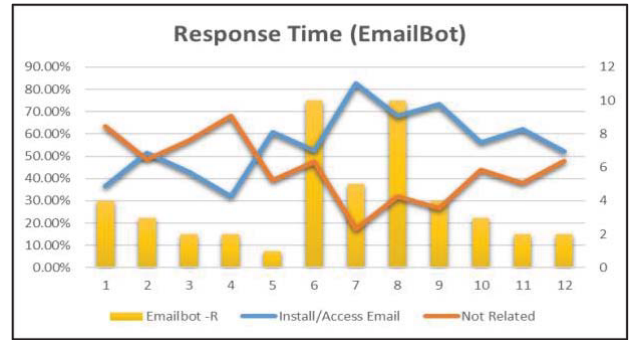


Fig. 10. Response Time of Automated Model (EmailBot).

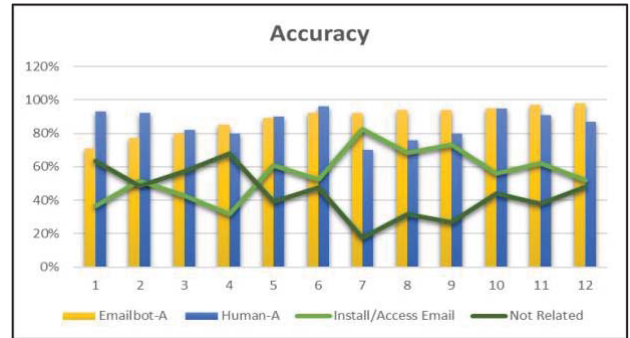


Fig. 11. Accuracy of Traditional and Automated Models (Human vs. EmailBot).

TABLE I. RESPONSE TIME OF TRADITIONAL AND AUTOMATED MODELS (HUMAN VS. EMAILBOT)

Week	Total Emails	Install/Access Emails	Average (Related Emails)	Traditional (Human) (Per Seconds)	EmailBot (Automated) (Per Seconds)
1	63	23	36.51%	1740	4
2	62	32	51.61%	1260	3
3	84	36	42.86%	1320	2
4	94	30	31.91%	1800	2
5	61	37	60.66%	1440	1
6	63	33	52.38%	1500	10
7	63	52	82.54%	3300	5
8	44	30	68.18%	1620	10
9	15	11	73.33%	720	4
10	57	32	56.14%	1080	3
11	50	31	62.00%	1800	2
12	50	26	52.00%	1860	2

TABLE II. ACCURACY OF TRADITIONAL AND AUTOMATED MODEL (HUMAN VS. EMAILBOT).

Week	Total Emails	Install/Access Emails	Average (Related Emails)	Traditional (Human)	EmailBot (Automated)
1	63	23	36.51%	93%	71%
2	62	32	51.61%	92%	77%
3	84	36	42.86%	82%	80%
4	94	30	31.91%	80%	85%
5	61	37	60.66%	90%	89%
6	63	33	52.38%	96%	92%
7	63	52	82.54%	70%	92%
8	44	30	68.18%	76%	94%
9	15	11	73.33%	80%	94%
10	57	32	56.14%	95%	95%
11	50	31	62.00%	91%	97%
12	50	26	52.00%	87%	98%

TABLE III. SUMMARY OF TESTING RESULTS.

Measures	Traditional	Automated
Interaction Method	Human-to-Human	Human-to-Machine
Response Time	10 – 30 minutes	2-10 seconds
Accuracy (Average)	86 %	89%
Integration	Manual access	API
Infrastructure	In dependent on-premises systems	Cloud-based

V. CONCLUSION

Organizations in the O&G industry are facing significant challenges in customer support, mainly related to time constraints and potential financial losses due to delays or incorrect decisions. Industry players increasingly pursue IR 4.0 technology solutions to ensure operational stability and continuity. Technical support is critical in this process, providing the foundation for developing and implementing innovative platforms. In this regard, AI-Powered Bots are gaining popularity in customer service, automating repetitive tasks and streamlining operations. The EmailBot solution is a promising example of this trend. Experimental results demonstrate its efficacy in significantly improving response times and email management, with a response time enhancement of 99.75% and an accuracy improvement of ~4%.

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