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### Abstract

Internet of Things and Machine Learning Implementation in the Healthcare Sector Modern health monitoring is crucial. Today's busy lifestyles lead to early health issues. Many issues occur from daily routines. Despite their ignorance of the risks, people enjoy many hobbies. Thus, identifying daily activities that affect health and predicting future diseases is crucial. However, electronic health data can forecast diseases like diabetes, TB, and others. Machine learning is used to predict a person's overall health. A complete health examination includes sleep quality, food, physical activity, and other elements. Also, the internet of things (IoT) is crucial in monitoring health and giving information when abnormalities occur. This paper focuses on the importance of monitoring health, and the role of the internet of things and machine learning in accomplishing this. And, providing the usage percentages of IoT applications in the healthcare sector from different countries.

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*Index terms—*

## 1 I. INTRODUCTION

Digital Technology helps society in achieving global health prospects. Sustainable Development 2030 highlights ICT growth and its worldwide influence, since interconnectedness can accelerate human progress. Strategies and actions are needed to close the digital divide and create knowledge societies. Technology has improved government services, and made them more accessible to more people, especially in healthcare, because services and data that were unavailable previously, are now affordable.

Ministers and delegation heads are involved in exploiting healthcare information to obtain sustainable growth, accelerate progress in the healthcare field, and help in reaching the health related sustainable development goals (SDGs).

The commercial sector and civic society play a crucial role, because the information and communication includes academic and technological communities. Developing countries urge the World Health Organization (WHO) to act in its field. The World Summit on the Information Society (WSIS) emphasizes ICTs at a global level.

Then, it is important to allocate enough resources and recognize information and communications roles in this case since technology opens new possibilities, at the same time, achieving the 17 sustainable development goals. Global health experts increasingly agree on the importance of strategic use of digital technologies and cutting-edge information systems. Concerted efforts are needed to include an extra billion people in universal access to critical services facilitated by ICT. WHO [1] asserted on digital interventions and using technology to solve problems and improve outcomes. The "health system" is the comprehensive network of organizations, and resources that deliver medical services and promote health. Digital health's 2020-2025 global plan emphasizes technology use.

Health emergencies are a priority to help one billion people, and health coverage protects an additional billion people. The Thirteenth General Program of Work, from 2019 to 2023, sets an organization's goals and priorities. IoT, virtual care, and remote technologies are being used in more fields, and AI, big data analytics, and blockchain technologies are being used in academic and professional fields. Data-sharing Internet of Things and Machine Learning Implementation in the Healthcare Sector platforms, wearables, and tools, and remote data capture and

45 interchange using storage systems and technology are used to enable Healthcare information sharing. Medical  
46 diagnosis and data-driven treatment can enhance outcomes, and decision-making, digital therapies, clinical trials,  
47 and healthcare self-management are essential.

48 Professional support requires knowledge, skills, and competence to enhance evidence-based practices and  
49 person-centered care. Despite great achievements by some nations, many nations remain confronting development  
50 issues and still need institutional help to develop, then, national eHealth/digital health plans and initiatives are  
51 needed, and their action plan requires extra work. This analysis examines resources and capabilities, and digital  
52 health strategy worldwide is needed for the goal of improving and supplementing existing and new efforts. These  
53 principles protect sensitive data. This paper focuses on the importance of individual health and the role of digital  
54 technologies like AI and IoT in enhancing healthcare. The rest of the paper is organized as follows section 2  
55 includes importance of individual health, section 3 focuses on using machine learning in healthcare, then, section  
56 4 focuses on IoT role and usage in the healthcare sector, then finally, conclusions and future scope are discussed.  
57 Predicting these events without computations may be inaccurate. Professional counsel may cost low-middle and  
58 middle-class families. Thus, a model that predicts their health based on multiple parameters is needed. Knowing  
59 one's health helps prevent future ailments. As mentioned, sleep patterns affect health beside smoking, sickness,  
60 and other factors. Healthcare data management emphasizes data handling efficiency. Patients create a lot of  
61 organized and unstructured data, besides diagnostics, doctor prescriptions, and wearable devices, that are now  
62 unstructured data that need to be collected and analyzed to be used in the benefit of a person's health. Data  
63 analysis can be accomplished through the use of machine learning and collected from smart devices connected to  
64 the internet (IoT).

## 65 2 II. INDIVIDUAL HEALTH

## 66 3 III. AI IN HEALTHCARE

67 Healthcare focuses on data analysis and forecasting in healthcare domains. Disease prediction has a big impact  
68 on healthcare analytics. Predictive models help prevent preventable illness epidemics, improving quality of life.  
69 Several recent studies have proposed health prediction models. With many considerations. Sahoo, Mohapatra,  
70 and Wu proposed in a study [2]. The study established a cloud-based probabilistic data collecting system and  
71 a framework for forecasting an individual's future health state using their current health status. Hirshkowitz et  
72 al. [3] developed a sleep duration assessment and suggestion system using age based classification. Researchers  
73 [4] proposed a new approach for 21st-Century Health Status Estimation Using Machine Learning. The study  
74 introduced the Convolutional Neural Network for disease risk prediction. A study using unimodal illness risk  
75 prediction and CNN-based multimodal disease analysis found that risk prediction intrigues. Weng, and his  
76 colleagues [5] examined disease prediction methods using ANNs. The researchers evaluated and contrasted each  
77 method using statistics. Researchers [6] devised a technique to collect health data using a specific method, in  
78 which, deep learning architectures assessed questionnaire results. Tayeb et al. [7] employed K-Nearest Neighbors  
79 (KNN) to predict cardiac disease and chronic renal failure. The author [8] proposed using EMRs to predict  
80 strokes. The researchers compared Deep Neural Networks (DNNs) to gradients in Error Correction Mechanisms  
81 (EMCs). Researchers [9] suggested a cloud-based smart clothing system for sustainability and human well-  
82 being monitoring, also, technology implementation was also discussed. Regarding studied methods, Schmidt,  
83 Tittlbach, Bös, and Woll [10] examined numerous types. Over 18 years, the researchers found substantial links  
84 between fitness, health, and physical activity. In a recent university fitness center data analysis [11], user fitness  
85 activity data predicts fitness center occupancy, but the fitness activity data can be predictive. Health parameter  
86 quantification study is extensive. Additionally, The computation of health parameters using alternate parameters  
87 is well-documented. Harris-Benedict [12] uses physical measurements to calculate a person's BMR. This method  
88 estimates the calorie needs for optimal health. Daily living activities affect health. Personalization can tailor  
89 health projections and recommendations, and this inspired the design of a daily life-based health prediction  
90 model. In the new Healthcare Era, the society healthcare is influenced by many factors, and affects in the  
91 finance, transportation, and entertainment. Big data and machine learning algorithms have transformed data  
92 analysis and insight extraction.

93 Integration has improved predictive analytics, pattern detection, and decision-making. In summary, modern  
94 society includes entertainment, business, and healthcare. Netflix knows which films people like and shows  
95 them. The timing, location, and item preferences of consumers are of interest for companies, like Amazon  
96 and Google. This enquiry concerns symptoms and conditions people are actively researching. Data can be used  
97 for intricate individual profiling, which can be valuable. Behavioral knowledge and targeting can help us predict  
98 and understand healthcare trends. AI could improve several fields. Healthcare includes diagnostics and therapy.  
99 Already important AI algorithms are performing comparisons in medical image interpretation and other activities,  
100 humans outperform machines. Using AI, examining symptoms and EMR biomarkers, as well as, characterizing  
101 and prognosticating diseases with EMRs can be performed. Many countries have a shortage of doctors due to  
102 increased healthcare demand. Healthcare facilities are likewise coping with many issues.

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## 103 4 New Technology and Patient Expectations

104 The user bases service and outcome expectations on Amazon and Apple items [13]. The advances in wireless  
105 technology and cellphones have opened many doors. Health tracking apps and search portals have enabled on-  
106 demand healthcare services, enabling remote healthcare delivery 24/7 | © 2023 Great ] Britain Journals Press  
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108 Internet of Things and Machine Learning Implementation in the Healthcare Sector interactions. Cost-effective  
109 techniques are needed to meet the needs of underserved and under specialized regions. Minimizing unneeded  
110 clinic exposure reduces the danger of communicable diseases. Recently, healthcare AI has garnered interest. AI,  
111 a discipline of computer science, creates intelligent machines that can do human tasks. Traditional healthcare  
112 infrastructure may be inadequate.

113 Healthcare infrastructure needs to be identified as the system expands. It was designed to meet current  
114 needs [14]. Though understandable, these solutions' success in treating patients requires thorough independent  
115 assessment, besides safety and efficacy are crucial. Today, AI-enabled healthcare technologies are gaining  
116 importance. Next-generation healthcare technology tools can be implemented. It's widely believed that AI  
117 improves healthcare operations and processes, also, AI application implementation which rely on the system,  
118 could save costs in the healthcare sector. Cost reductions come from reduction of hospitalizations, doctor  
119 visits, and medical care treatments from reactive to proactive healthcare, Health management is prioritized over  
120 disease treatment. AI-based technologies will help with many chores, since monitoring and guidance keep people  
121 healthy. To improve patient care, diagnose faster, personalize treatment programs, and improve monitoring  
122 and evaluations, AI-based healthcare technologies are expected to increase rapidly. Technology has advanced in  
123 the past decade, also, AI and data science has advanced. Currently, different applications have been explored  
124 for decades. The current AI enthusiasm is unique. Optimized computational processing speed, data collection  
125 capacity and AI talented people are required to accelerate AI development. The use of tools and technology  
126 [15,16] in the AI field, will revolutionize artificial intelligence (AI) technology and its widespread use and effect  
127 on society. Specifically, deep learning (DL) has significantly impacted healthcare.

128 The aforementioned reason has had a major impact on current AI tool viewpoints and has driven several AI  
129 tool innovations. Given the present enthusiasm for using artificial intelligence (AI) in numerous disciplines, it is  
130 clear that these

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133 Internet of Things and Machine Learning Implementation in the Healthcare Sector well-studied in research [21].  
134 Numerous applications exist throughout the healthcare value chain. Also, drug development and ambient assisted  
135 living (AAL) research have grown in popularity.

## 136 6 Precision Medicine

137 Precision medicine is known as personalized medicine, is a new medical strategy that tailors treatments and  
138 interventions to individual patients. Precision medicine can tailor healthcare to patients' disease features.  
139 Genomic variants and other medical considerations will be considered in a customized therapeutic approach.  
140 Precision medicine examines age, gender, geography, race, family history, immunological profile, and metabolism.  
141 Precision medicine uses individual biological traits rather than population-based trends. Throughout a patient's  
142 treatment, data collection is involved. Individuals provide genetic and physiological data. Precision medicine  
143 benefits healthcare. Healthcare costs may be reduced. Precision medicine can save healthcare expenses by  
144 avoiding needless operations, testing, and drugs. Precision medicine reduces harmful medication reactions.  
145 Precision medicine is expected to benefit from its novel approaches. This study examines patient outcomes  
146 and health service delivery and evaluation changes after healthcare interventions. Modern healthcare emphasizes  
147 digital health apps and "omics"-based diagnostics.

148 Machine learning methods are used with large datasets. Many precision medicine initiatives benefit the  
149 discipline as a whole. Academic research often uses genetic, demographic, and electronic data. Health records  
150 can be diagnosis and therapy selection. Digital health apps record and process data.

151 Patients also reported diet, mental well-being, and physical activity using wearable, smartphone, and other  
152 health monitoring data. In precision medicine, machine learning algorithms find patterns in data sets to improve  
153 prediction and outcomes. Healthcare AI research is growing. Omics-based testing uses population genetic data.

154 Machine learning algorithms find relationships and predict patient treatment responses. Metabolite profiles  
155 can also reveal health and disease. These biomarkers provide a complete picture of an individual's physiological  
156 condition and can be used to determine disease risks, progression, and treatment efficacy. Protein expression  
157 patterns can help researchers understand disease mechanisms. The gut microbiome's makeup and diversity can  
158 illuminate microbial communities' function in health and disease. Metabolite profiles also reveal a person's  
159 metabolic processes. Metabolic profiling and machine learning can provide personalized treatment [22,23].

## 7 Healthcare AI Implementation Barriers

In order to assess the potential fluctuations in the healthcare industry's integration of artificial intelligence (AI), specifically with regards to variables associated with technological adoption. What insights can be gleaned from previous healthcare information technology (IT) implementations?

The scholarly literature underscores the significance of integrating advancements in the implementation of artificial intelligence (AI) and other information technology within enterprises. The successful implementation of electronic medical records necessitated the utilization of inventive strategies for integrating software systems and introduced novel procedures for healthcare professionals, chemists, and other occupations within the healthcare industry. Consequently, the greater affordability of complementary innovation in larger corporations and metropolitan regions is anticipated to result in a higher prevalence of AI implementation within larger healthcare institutions and urban locales.

The application of artificial intelligence (AI) in the healthcare sector can be exemplified by the analysis of a substantial dataset consisting of 1,840,784 job advertisements originating from 4,556 hospitals. A total of 1,479 job listings from 126 hospitals were assessed by Burning Glass Technologies, with a specific focus on the requirement of artificial intelligence (AI) skills.

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The job listings encompassed positions such as "Analytics Architect," "Bioinformatics Analyst," "Cardiac Sonographer," "Physician -Internal Medicine," and "Respiratory Therapist." The findings of the analysis revealed that a majority of AI-related job opportunities, specifically 60%, were categorized as clinical positions. Administrative roles accounted for 34% of the job opportunities, while research-focused positions constituted a smaller proportion of 6%.

The research identified a total of 1,479 job advertisements related to artificial intelligence. A significant discovery indicates a deficiency in healthcare skills related to artificial intelligence. Based on the findings of a previous study in the field of information technology, it has been observed that the 126 hospitals that are actively recruiting for artificial intelligence (AI) positions tend to exhibit a higher number of personnel and are predominantly situated in densely populated urban areas. It is anticipated that artificial intelligence (AI) has the potential to ameliorate the existing state of affairs in the healthcare sector. It is anticipated that the implementation of artificial intelligence will primarily commence within large-scale institutions and major urban centers, encompassing domains such as electronic medical records, computer systems, and the commercial internet.

Gaining insight into the factors that contribute to hospitals' reluctance to adopt artificial intelligence (AI) is imperative for comprehending the potential complementary advancements that could facilitate its implementation within healthcare settings. There are several factors that impede the widespread adoption of a proposal, including algorithmic limitations, restrictions on data access, legislative barriers, and misaligned incentives.

## 8 Legal and Administrative Hurdles

Legal and administrative hurdles hinder industry and sector operations. Foundational regulatory constraints cause algorithmic and data issues. Three types of regulations matter. Privacy regulations initially complicate healthcare data collection and consolidation. Due to privacy concerns in the healthcare field, using actual health data to train AI models may be difficult, slowing progress compared to other industries. Novel medical technology requires lengthy and demanding regulatory approval. Innovation clearance takes years. Health care providers' fear of responsibility can also prevent them from adopting innovative technologies. Health care regulation is more conservative than in other businesses. This means that innovative regulatory frameworks are needed to integrate AI into healthcare. This approach will maximize AI's benefits while protecting patient rights and maintaining high-quality healthcare. Three regulatory hurdles could be modified to complement each other. These issues involve health care data ownership and use, AI medical device and software approval, and medical provider-AI developer liability.

## 9 Data Constraints

Data quality affects AI algorithm performance. Thus, data scarcity is another barrier to adoption. Medical data gathering and access are difficult. Medical practitioners sometimes dislike data collecting because it disturbs their workflow and produces incomplete data. Data aggregation between hospitals or healthcare providers is difficult. Electronic Healthcare Record (EHR) systems used by government-certified providers serving hospitals and healthcare facilities are incompatible, resulting in localized data collection rather than an integrated approach to documenting a patient's medical history across multiple providers. Lack of large, high-quality datasets hinders AI system development.

## 10 Algorithm Limitations

Neural network advancements have increased artificial intelligence's potential but decreased interpretability. Neural networks make AI algorithms "black boxes" that require a lot of work to understand. Thus, without proactive efforts to identify issues with neural network-generated algorithms, there is a risk that the AI will produce flawed solutions that are only discovered after deployment. This lack of transparency can undermine

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223 Internet of Things and Machine Learning Implementation in the Healthcare Sector healthcare providers, especially  
224 since doctors and hospitals may be held responsible for decisions involving AI. Complementary innovation in  
225 trustworthy AI, such as using technology or methods to understand AI algorithms, is widely recognized. Many  
226 large-scale projects aim to develop and improve AI. Interpretable AI could reduce the black box problem and  
227 increase confidence. Healthcare practitioners may trust AI systems by understanding how AI makes suggestions.  
228 Individuals are working to standardize AI clinical trial techniques. These efforts should improve healthcare AI  
229 integration. Implementing such criteria will help healthcare practitioners identify how biases or knowledge gaps  
230 affected an AI system’s suggestions.

## 231 12 IV. IOT SITUATION IN HEALTHCARE

232 This section provides an overview of the global healthcare Internet of Things (IoT) industry. Medical devices can  
233 be categorized into fixed, wearable, implanted, and other classifications. The software and system components  
234 encompass various segments, such as application security, network security, data analytics, remote device  
235 management, and network bandwidth control. The market is divided into segments based on services, products,  
236 connectivity, and end users. This section investigates industry trends, growth prospects, and regional forecasts  
237 spanning the period from 2022 to 2030. The global healthcare Internet of Things (IoT) market attained a  
238 valuation of USD 180.5 billion in the year 2021 [24]. According to projections, the estimated value of USD 960.2  
239 billion is anticipated to be achieved by the year 2030, with a compound annual growth rate (CAGR) of 20.41%.  
240 Services accounted for 59% of the total revenue generated in the year 2021 [24].

241 In the year 2021, hospitals experienced a 35% increase in end-user income. In the year 2021, North America  
242 exhibited the highest proportion of revenue, accounting for 40.3%. The Asia Pacific region is projected to  
243 experience a growth rate of 18.50% during the period from 2022 to 2030. Table 1 [24] presents the projected  
244 forecast for the Internet of Things (IoT) in the healthcare sector until the year 2030.

245 Data gathering, analysis, monitoring, and research occur online. Sensors, software, and information processing  
246 systems dominate the healthcare IoT market. Due to expanding demand for medical devices in healthcare facilities  
247 and more patients seeking medical attention, the Internet of Things (IoT) in healthcare has grown significantly.  
248 Medical gadgets with improved efficiency and faster results have also been prioritized.

249 New technology and developments have increased digitalization in many locations, especially developing  
250 countries. The healthcare market has grown significantly since the governments integrated and promoted medical  
251 device development and provision. The Internet of Things (IoT) transmits data between machines, Healthcare  
252 IoT applications are categorized by medical equipment type. This category includes fixed, implanted, wearable,  
253 and other modern medical devices in healthcare institutions. Wi-Fi, Bluetooth, and signee-enabled embedded  
254 systems enable uninterrupted work operations. Analytics, database, and network layers comprise the system  
255 and software. Microsoft Application Insights lets developers monitor and diagnose their apps’ performance  
256 and usage. Telemedicine, store and forward telemedicine facilitated by software using wireless connections,  
257 medication management, interactive medication, patient monitoring, clinical operations, workflow management,  
258 clinical imaging, and fitness measurement can be used to segment healthcare applications in the IoT. Drug  
259 development and research have boosted the IoT healthcare market. In 2021, the percentage of healthcare IoT  
260 market share by region [ 24] is shown in table 2. From a geographical standpoint, it is anticipated that the  
261 Asia Pacific region will take the forefront in the advancement of healthcare Internet of Things (IoT) technology.  
262 The proliferation of advanced technologies and the increasing demand for goods and services have resulted in  
263 an upward trend in market rates. The government has facilitated the implementation of Internet of Things  
264 (IoT) in hospitals through the utilization of advanced infrastructure. The utilization of healthcare Internet of  
265 Things (IoT) has witnessed an increase in North America, Europe, Latin America, the Middle East, and Africa as  
266 well. The implementation of this initiative has significantly enhanced healthcare services in the aforementioned  
267 regions. In recent times, numerous disciplines have witnessed noteworthy advancements. In 2020, Abbott and  
268 Insulet unveiled a novel system for glucose monitoring and automated insulin delivery. In the year 2021, Hill  
269 Rom unveiled integrated solutions aimed at enhancing patient outcomes. In 2021, the SyncaR AR technology  
270 and StealthStation S8 surgical navigation system were introduced by Surgical Theatre and Medtronic. Medical  
271 devices are utilized for the purposes of diagnosing, treating, and preventing various diseases.

272 Implantable medical devices are specifically engineered to be surgically inserted into the human body for the  
273 purpose of diagnosing, monitoring, or treating specific medical conditions. The term “Software and System”  
274 encompasses computer programs and hardware components that collaborate to accomplish predetermined  
275 objectives. This connection facilitates operational efficiency and enhances overall performance. Application  
276 security is a discipline that aims to safeguard software from potential threats and vulnerabilities. Data analytics  
277 involves the examination and interpretation of extensive datasets in order to derive meaningful insights and  
278 inform decision-making processes. The practice of remote device management encompasses the ability to exert

## 14 V. CONCLUSIONS AND FUTURE SCOPE

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279 control over devices from a distance. The tasks encompassed in this domain include monitoring, configuring,  
280 updating, and troubleshooting. The practice of architecture involves the design and implementation of system  
281 integration within a broader framework. The issue at hand pertains to application development, with a specific  
282 focus on support and maintenance.

### 283 13 Glucometers

284 test blood glucose levels. Electrocardiograms (ECGs) and heart rate monitors are used in clinical settings to  
285 examine and monitor heart electrical activity and measure heart rate. Medical devices assess blood pressure  
286 against artery walls. Clinical settings use these gadgets. Multiparameter monitors measure and show numerous  
287 patient physiological parameters. Oximeters are breathing support devices that help people with breathing issues.  
288 Imaging systems capture, record, and reproduce images.

289 Implantable cardioverter-defibrillators (ICDs) monitor heart rhythms and are surgically installed. Implantable  
290 cardiac monitors, also known as implantable loop recorders, are medical devices surgically inserted to monitor and  
291 record heart electrical activity. Infusion pumps supply fluids like drugs or nutrition to patients. Fetal monitoring  
292 devices evaluate a developing fetus's health and physiological parameters during pregnancy. Neurological gadgets  
293 diagnose, monitor, and treat nervous system disorders. Embedded systems are computer systems that execute  
294 specific duties within a larger system or device. Finally, laboratory research is regulated, methodical investigation  
295 in a lab.

## 296 14 V. CONCLUSIONS AND FUTURE SCOPE

297 This paper discusses the importance of monitoring the health of individuals, as this helps in maintaining a  
298 balanced lifestyle. Also, the importance of using AI in the healthcare sector, to help analyze patients' data, for  
299 detecting any health issues, and help in taking precautions before health deteriorates, and decreases costs at the  
300 same time in the healthcare sector. Also, precision medicine is a type of medicine that depends on AI in detecting  
301 health problems based on each individual's metrics.

302 Besides, implementation of AI challenges in healthcare were discussed in this paper. The role of internet of  
303 things (IoT) in facilitating transmission of patients' data from specialized devices to analyze these data, and  
304 provides results of analysis to doctors. It is expected in the future for healthcare using IoT to increase annually  
due to the benefits and costs reduction it provides in healthcare. <sup>1</sup>

Health system strengthening is essential and health plans should include digital health. The major goal is to let people benefit ethically while maintaining safety, security, and reliability.

Academic

sustainability. Development should follow principles. Academics value transparency, accessibility,

interoperability. Technology, law, and ethics all require privacy, security, and confidentiality.

fields prioritize equity and

scalability, replicability, and

Figure 1:

1

IoT in Healthcare Market Size,  
Year  
2021 to 2030 (USD Billion)

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Figure 2: Table 1 :

Regions	Revenue share in 2021 (%)
North America	40.30%
Asia Pacific	20.60%
Europe	25.70%
Latin America	9%
MEA	4.40%

Figure 3: Table 2 :



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