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Revolutionizing Sustainable Energy Production with Quantum Artificial Intelligence: Applications in Autonomous Robotics and Data Management

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Abstract: The present commentary delineates the scope of quantum-based artificial intelligence (AI) in confronting the escalating need for eco-friendly sources of energy. Contemporary concerns regarding the detrimental impacts of conventional energy sources on the environment have led to the exploration of alternative and sustainable options. The amalgamation of quantum computing (QC) and AI is a potentially effective remedy for the existing energy crisis. The present analysis explores the theoretical underpinnings of quantum AI and its palpable ramifications in the energy realm. It highlights the capacity of quantum AI to enhance energy efficiency, reduce waste, and minimize the detrimental impact of energy production on the environment. Experimental techniques, such as solar irradiation, wind velocity, and pumped-water energy storage, were utilized to demonstrate the effectiveness of quantum AI in predicting renewable energy generation. Several information models and machine learning techniques have been examined to identify models that yield accurate predictions of renewable energy generation. The results demonstrate that quantum AI can easily perform complex quantum subatomic simulations, making it a valuable tool in the fields of computational science and molecular physics. A quantum AI-powered automated robotic system is proposed to optimize energy management. The implementation of a monitoring system that monitors energy consumption levels, mechanizes the production and distribution processes of energy, and anticipates future energy demands in real time renders it an indispensable device for both energy conglomerates and policymakers. This study ascertains that the amalgamation of QC and AI harbors substantial potential for sustainable energy generation, marked by streamlined energy production, minimal waste generation, and ecoconservation. The present investigation furnishes insightful observations pertaining to the potential benefits of quantum AI for the purpose of ensuring sustainable energy production. Additionally, it presents a roadmap for future research in this area.

Keywords: machine learning, neural networks, quantum artificial intelligence, qubit, robotic systems

Quantum computing can be used for the rapid training of machine learning models and to create optimized algorithms. Quantum artificial intelligence (AI) is one of the most promising solutions for next-generation energy systems.

1. Introduction

The utilization of renewable energy sources presents a remarkable avenue for adequately addressing energy requirements while ensuring sustainability. However, its effective integration is constrained by formidable hurdles such as the erratic and uncertain nature of supply. Numerous scholars have delved into the implementation of prognostic models and artificial intelligence (AI) algorithms to anticipate renewable energy generation when addressing the aforementioned quandaries. Despite recent advancements in renewable energy models, their accuracy remains a critical challenge, necessitating further enhancements to maximize their potential. This study investigates the potential of quantum

computing (QC) to improve the precision of renewable energy forecasting. Specifically, we examine the efficacy of integrating QC with AI algorithms for this objective. The focus of our investigation is to ascertain potential enhancements in the precision of renewable energy prognostications via the convergence of quality control (QC) and AI algorithms. In addition, we explore the implications of this integration on the sustainable utilization of renewable energy resources.

Recently, there has been much focus on the sustainable use of renewable resources, which has prompted extensive research on this topic. Finding sufficient electrical supplies is the main challenge for renewable energy in the near future. By incorporating renewable resources, renewable energy can be found in current or future energy infrastructure. Critical concerns, such as increased supply dependability and addressing localized power shortages, will be addressed with sustainable energy. Because of the unpredictability and inconsistent and randomized characteristics of renewable energy, its production is discontinuous and chaotic. However, the accuracy of renewable energy statistics is yet to be determined. Numerous studies have demonstrated the application of various

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robotic algorithms to forecast renewable energy. Document models can provide reliable renewable energy predictions. The development of AI models to improve the forecasting accuracy of renewable energy is more significant. We will be able to develop an entirely new energy source that will revolutionize the entire circular energy system by using the most recent technique that integrates quantum physics with AI algorithms [1]. Quantum computation (QC) is a scientific field that studies how the quantum behavior of certain subatomic particles (i.e., photons and electrons) can be used to perform computation and eventually large-scale information processing. Quantum computers perform calculations based on the probability of an object's state before it is measured – instead of just 1s or 0s – which means they have the potential to process exponentially more data compared to classical computers. Superposition and entanglement are two features of quantum physics on which these supercomputers are based. This empowers quantum computers to handle operations at speeds exponentially higher than conventional computers and at much lesser energy consumption. QC could contribute greatly in the fields of military affairs and intelligence, drug design and discovery, aerospace designing, utilities (nuclear fusion), polymer design, finance, machine learning, and basically AI. This study is largely focused on the benefits of QC and its relationship with AI algorithms. Namely, in classical computers, information is recorded in macroscopic, two-level systems. Wires conducting electrical current in computers can be in two basic states: when there is no current flowing through, representing a logical "0," or else when there is some current flowing through, representing a logical "1." These two states constitute a small amount of information. All computations are based on the logical manipulation of bits through logical gates acting on wires that represent these bits. Imagine, however, that instead of wires and currents, we use two electronic states of an atom to record the information. These states are called the ground state $|0_i$ and the excited state $|1_i$ (Dirac notation is the most natural for QC). But, since an atom obeys the laws of quantum mechanics, the most general electronic state is a superposition of the two basic states (Equation (1)).

$$|\Psi_{1i} = a|0_i + b|1_i, \tag{1}$$

called quantum bits or qubits. When there are two bits, there are four possibilities: 00, 01, 10, and 11. However, this should be contrasted with two qubits which are in general in a state of the form (Equation (2)).

$$|\Psi_{2i} = a|00_i + b|01_i + c|10_i + d|11_i. \tag{2}$$

One of the first contributions of QC to AI was the production of truly random numbers. A truly random number of N bits can be produced by applying the Hadamard transformation to an N-qubit quantum register, thus producing the superposition of all basis states (Equation (3)).

$$1/\sqrt{2}\,\widehat{}\,N \sum |k> \tag{3}$$

Then, by measuring this state, we obtain a truly random number in the range of [0, 2N-1].

Discourse and language handling have extraordinary arrangements for acquiring QC. In addition to the previously mentioned surmised design matching of the information signal and the undeniable quick quantum search in tremendous lexical datasets, the portrayal issue can be settled exquisitely in a quantum register and more effectively than any other time [2].

Machine learning has had a longstanding application in the field of study associated with data-driven challenges. AI techniques encompass a diverse set of multidisciplinary tools such as analytics, arithmetic, artificial neural networks, information collection, and optimization. Artificially intelligent algorithms attempt to ascertain correlations between incoming and output data by using mathematical forms. Upon completion of the construction of the final machine learning algorithms utilizing the entire dataset, senior managers may incorporate the predicted input data into the aforementioned models to yield outcomes that fall within the targeted range. The acquisition of prior knowledge is of significant importance in the domain of computer science and can considerably enhance the efficacy of the machine learning process. Machine learning technology predominantly employs three methodologies: learning algorithms, unsupervised classification, and supervised learning. Supervision involves the utilization of annotated training data over the course of the observed duration. The training methodology involved leveraging unannotated data to effectively categorize novel data points based on pre-established standards, thereby facilitating system understanding (Figure 1).

The number of nodes exhibits variability owing to the reliance on the grouping criteria that groups typically employ. Augmentation learning is used to describe the process of acquiring data from the surrounding environment with the aim of attaining greater potential benefits. The incorporation of these three fundamental learning concepts has led to a multitude of tactics and proffered recommendations regarding methodologies (Figure 1). Moreover, the utilization of a single AI framework engenders diverse modes of analysis by projecting renewable energy generation [3].

1.1. Quantitative methodology/state-of-the-art models

The success of QC holds the potential to prompt a reevaluation of how an AI challenge is currently conceptualized. The game of chess is commonly acknowledged in academic discourse as a testbed for AI. The veracity of this statement lies in the fact that the brute-force algorithmic methodology in the realm of conventional computing involves considerable computational expenditure. Nonetheless, the aforementioned concern does not pose an overly arduous obstacle in the realm of quantum mechanics. If adequate hardware is accessible, a quantum algorithmic technique can resolve the problem within a rational timeframe. The potential benefits of incorporating AI within the realm of quantum physics can be further evaluated by examining its prospective application for predicting molecular wave functions (Equation (4)):

$$H_{mo} = -\hbar_{22}me\nabla_2 - \sum aZae24\pi\varepsilon0|\overrightarrow{r}-\overrightarrow{ra}| \tag{4}$$

$$y(x,t) = A_{sin}(k_x - \omega_t + \varphi)$$

$$T = 2\pi\omega$$

and the electronic properties of the molecules. The conduction of electrons within a given molecule can be observed, and the gathered data may be utilized as input for an AI algorithm, thereby enabling prognostications regarding the consequential future behaviors of the electrons present within the molecule. Through quantum mechanics, the conduction of an electron in a molecule remains explicable through a wave function ($\Psi(x,t) = A_{\cos}(k_x-\omega_t)$), which bears resemblance to its conduction in an atom. In general, AI can detect and predict consistent patterns of natural behavior [4]. AI is operationalized in anticipation of individuals' shopping behavior by observing their shopping patterns over time. In light of these considerations, AI is a viable

Figure 1
The integration model
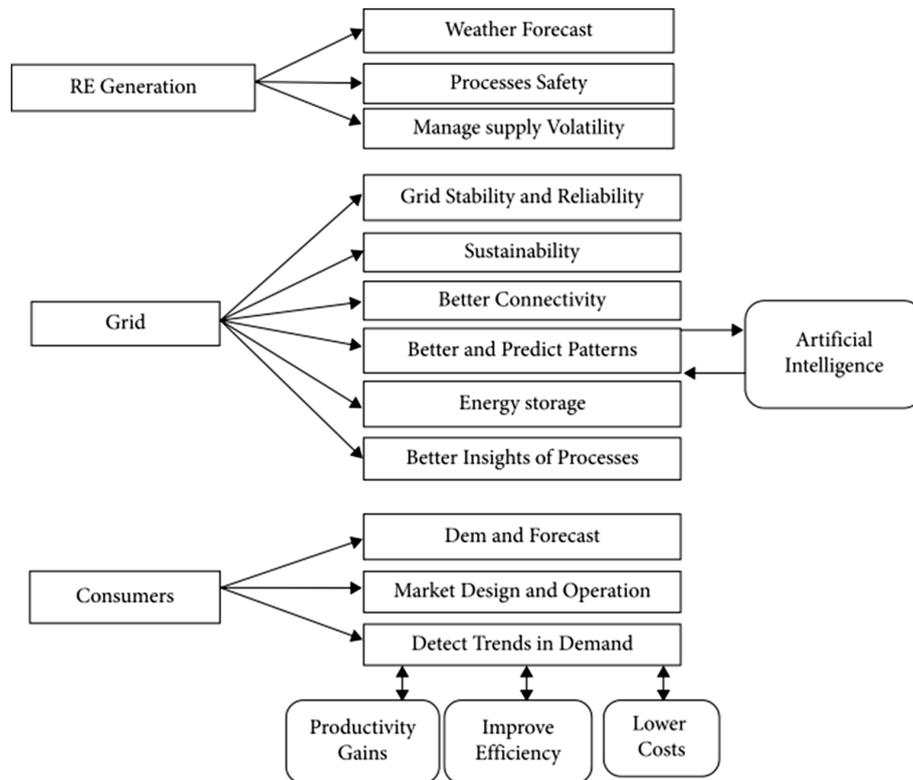
tool for predicting the quantum states of particles, commonly known as wave functions, which play a crucial role in determining all atomic characteristics. The mechanism of AI can be deployed to deduce optimal solutions to critical quantum mechanical quandaries. The traditional approach to accomplishing this task requires a substantial amount of high-performance computing resources, which is an obstacle that typically limits the ability to execute computational designs of novel artificial entities for application in medical and industrial contexts. However, this recently developed AI algorithm can provide precise predictions in mere seconds on both computers and mobile devices. Whereas the study investigates convolution neural networks for sustainable energy predictions, there are numerous existing models that forecast how much energy will be supplied in the system under observation. Four categories – deep faith network, vehicle stack, profound recurring machine learning, and others – were used to group the prediction methodologies. To improve the prediction accuracy, a variety of data pretests and posttests were applied. Studies examined the effectiveness and dependability of adaptive intelligence prediction systems. In this study, energy sources such as solar, hydroelectric, and wind are considered. The benefits of neural networks in terms of the strength and dependability of forecasts have been demonstrated in a large number of cases. Showed how information on solar energy forecasts was assessed using equipment, pattern recognition, AI algorithms, and hybrid approaches. According to the analysis, even data on solar photovoltaic reliance can be estimated using mathematical weather forecasts with features and intelligent systems to carry out prolonged solar energy prognostication,

memory channels, coevolutionary neural connections, and recurrent artificial neural networks. A transportation vehicle is introduced where decisions are made automatically using best control techniques to investigate the impact of applied optimization in the energy management systems of various devices. A learning model employs earlier datasets for identification scenarios when the vehicle is tested with two different energy cases, low and high, and as a result, an energy intersection line is visible. Additionally, a customized metering system that operates in automatic mode is integrated with a home management system that uses AI algorithms.

The projected model was used to track the energy consumption of each appliance and to reduce energy consumption while still guaranteeing the same level of performance. An AI algorithm was used to identify and manage the presence of renewable energy sources, thus boosting the efficiency of the proposed technique. Additionally, if the energy level drops below a predetermined threshold, an error detection process will be used to identify it, and it will also be framed in an analytical form. Additionally, the proposed approach guarantees a precise energy monitoring system that controls the energy by keeping track of all monitored variables in a different cloud management system [5–8].

2. Objectives

The purpose of the research is to ascertain the significance of quantum AI and self-driving robots in the upcoming industrial revolution and how they can generate sustainable energy.

3. Research Methodology

3.1. Research design

The research design for this work incorporates independent analysis and experiments as well as theoretical concerns. In order to explore how quantum mechanics can be incorporated into AI algorithms to create quantum neural networks, and how AI can be used to create a sustainable free energy source and alter the circular energy system, theoretical considerations were developed through a review of pertinent literature and expert interviews. Parallel to theoretical considerations, independent research and experiments have been conducted to demonstrate how quantum mechanics can be incorporated into AI algorithms to obtain quantum neural networks, how to obtain a sustainable free energy source with the aid of AI, how AI will change the circular energy system, and why this approach is required under conditions of global warming. The practical component of the research design involves the use of digital and physical tools to test and demonstrate the feasibility of the theoretical considerations. For the digital component, a digital server was created to process and teach data, and a server based on personal data was used to increase processing speed. Testing was conducted on a hub module that provided access to almost all data on the internet. Integration of quantum mechanics was achieved using the IBM cloud, where periodic training and improvement of the module were performed.

For the physical components, a robot was assembled using Arduino-Raspberry electronic parts to transfer the digital system. The intended function of constructing a tangible system is attributed to its facile assimilation, in addition to the equipment utilized for validation and prospective advancements.

3.2. Instruments

This study encompasses a range of methodological approaches and methodologies to prognosticate electricity, namely data acquisition and analysis coupled with the application of AI. The present study employed advanced neural networking techniques to undertake data mining and formulated a novel algorithm based on machine learning, serving the purpose of detecting diverse energy supply levels and exerting control over the corresponding electrical connections. The algorithm was deployed through a hardware solution based on Arduino, which effectively retained electrical energy for subsequent utilization. In addition, a robotic hardware architecture was conceptualized and developed to facilitate the acquisition and retention of electric potential. In addition, a platform was devised employing the principles of quantum machine learning, imbued with AI capabilities, with the aim of mitigating extraneous consumption of electricity and overseeing actions within the vicinity, including that of the computer and within the household setting. Quantum data storage and control pose significant challenges for researchers. However, the application of AI techniques has demonstrated potential for addressing the complex aspects of quantum molecular simulations. This topic is also discussed.

The ensuing discourse expounds upon the theoretical and technical facets, with meticulous attention paid to detail.

The overall strategy for forecasting electricity comprises two components. The discussion of data collection and analysis comes first. Second, AI approaches using AI are used for data or information. The gathering and interpretation of data are the most important components in the prediction of sustainable energy.

Standardization, undesirable/false data export, data grouping, and data inferential statistics should be entered into the data packet for the right energy projection. Regarding the training modeling and assessment of specific forecasts of renewable radiation, renewable energy, and hydroelectricity, various renewable energy input variables were used (Figure 2).

Figure 2 summarizes the approach that is based on the machine intelligence (MI) paradigm [9–11].

The best data mining algorithms use neural networking techniques because they have the ability to learn, store, and connect nonlinear information. Although sensors can monitor data in predictable ways, they are not interchangeable. Algorithms are more capable of approximating any linear combination and are fault-tolerant. As a result, they can manage noisy, fragmented, quadratic, and quasi-datasets [12].

The simple two-wheeled robot used in this study was modeled using the lpzrobot simulation tool and was based on a physical motor ODE. The motor values are provided via vector yt R2, which is the controller's output. xt R2 is the vector of the observed rotational speed of the wheels, and the only sensors are wheel counters that measure each wheel's actual speed. The ODE (Open Dynamics Engine) physics engine realistically replicates the effects of the robot.

Owing to the principle of inertia, various outcomes arise from wheel slip, ground friction, and collisions. The present system embodies a fully realized robotic entity given that the robot operates at a velocity sufficient to generate destabilization upon impact. A diverse range of recent contributions has suggested the renewed exploration of novel approaches to autonomous driving. In the context of Markov systems, extant research has evinced the capacity of predictive information to diminish the quantity of reciprocity information (MI) observable between temporal points. These findings suggest that MI can function as an initial methodological approach for achieving comprehensive predictive information. Empirical analysis was employed to determine the measurement variability of the sensor values over a given period of time during robot investigations, which indicates that the maximum mutual information (MI) in the work mode is attained by the robot, which could be inferred to have effectively devised a navigation strategy, as perceived by an external observer. Because the electrical connection is not constant and the current is not constant based on previous studies, the neural networks integrated in the given algorithm can regulate the given connection and further accumulation. This physical experiment assisted me in developing a fully predictive machine learning-based algorithm that can identify the energy supply level with the help of a network connection [13, 14].

An Arduino-based hardware solution was developed to fully utilize the algorithm. It stores the electrical energy gathered by the algorithm for later use. Using the Arduino chipset, this hardware system transmits the signal to the brain panel, from which it is redirected to the power supply unit, which retains the accumulated electrical source (Figure 3).

The photo shows a hardware system designed in the form of a robot (Figures 3 and 4).

Drawing on quantum machine learning, a platform directed by advanced AI has been developed. By integrating electrical energy management, this system establishes a solitary decentralized network for users. The aforementioned AI effectively restricts or eradicates the surplus electrical consumption. In addition, the network is continuously monitored by quantum neural networks, which assess activity within the surrounding environment, including areas such as computers and homes. In the absence

Figure 2
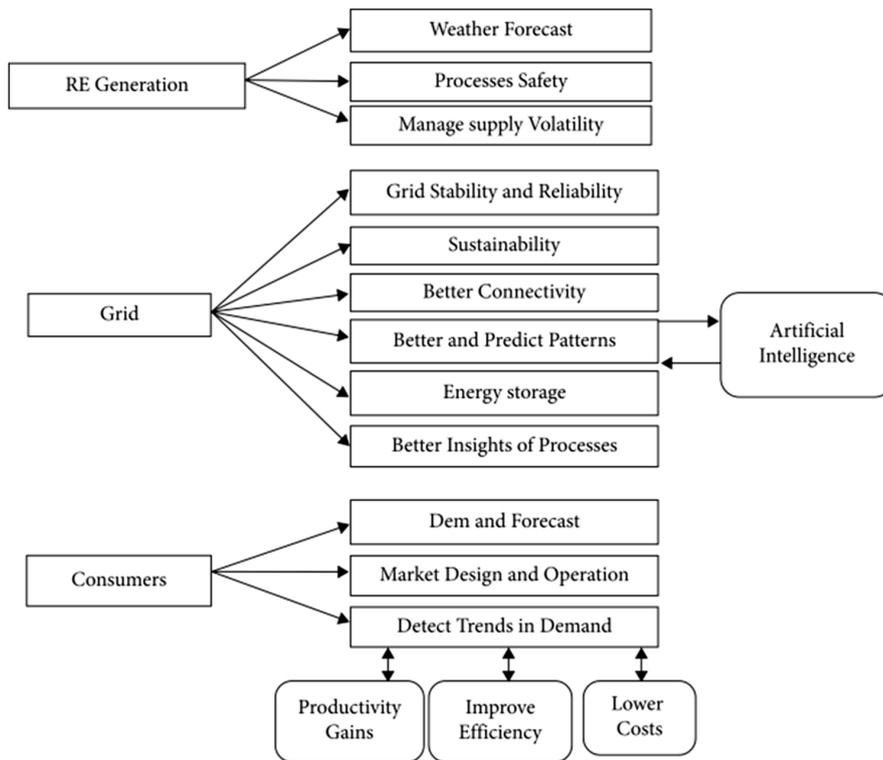**Machine intelligence (MI) paradigm**

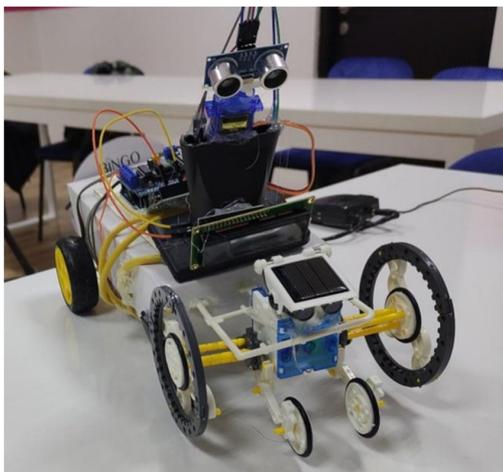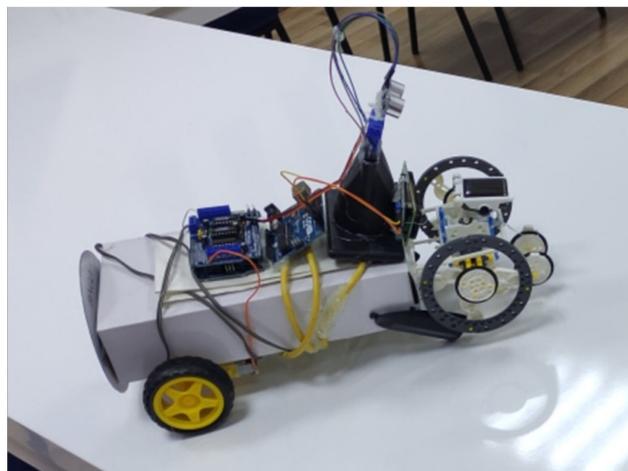**Figure 3**
Hardware prototyping/production

Figure 4
A robot with an integrated algorithm

of seamless integration with a given object or space, AI subsequently reduces the energy supply to the identified block after a specific duration. The present invention offers a valuable solution for mitigating excessive electricity consumption and establishing a singular sustainable energy system applicable across diverse industries. In addition to environmental advantages, the incorporation of this technology also takes into account human necessities through the hardware system that procures and recycles energy, leading to cost reduction of power for individuals, enterprises, or a country in its entirety.

In this audit, we have seen that the laws of quantum mechanics infer an alternate sort of data handling to the conventional one in light of the laws of traditional material science. The focal contrast, as we accentuated, was in the reality that quantum mechanics permits actual frameworks to be in an ensnared express, a peculiarity that is nonexistent in traditional physical science. This prompts a quantum

PC to have the option of addressing specific assignments faster than its traditional partner [15, 16]. All the more significantly, factorization of normal numbers into primes can be performed productively on a quantum PC utilizing Shor's calculation (Equation (5)),

$$O(\log n)\, 2\ (\log\log n)(\log\log\log n) \tag{5}$$

$$q = gcd(x\,r\,2 \pm 1,\ n).$$

However, they are currently viewed as unmanageable for traditional PC. Be that as it may, to understand a quantum PC (or, without a doubt, some other PC), we must have an actual medium in which to store the control data. It is here that quantum data turn out to be exceptionally delicate, and it just so happens that the undertaking of its stockpiling and control requires a ton of exploratory inventiveness. The equipment for quantum registers is currently in the earliest stages because of the obstruction of decoherence, which makes it very difficult to survive. Accordingly, a large portion of the previously mentioned strategies have been tried for either insignificant issues (expecting 3–5 qubits) or QC test systems. But maybe this situation is going to change sooner than expected. The research shows that AI techniques can proficiently play out the most troublesome parts of quantum subatomic reenactments [17–19].

In the following few years, AI techniques will secure themselves as a fundamental part of the disclosure interaction in computational science and subatomic physical science. The interdisciplinary field of joining AI and quantum physical science is developing quickly, and invigorating headways are being made. The above conversations are merely a glimpse of something larger.

Errors incurred during practice sessions are determined through a comparative analysis between the verifiable performance outcomes and the set objectives, resulting in the alteration of the parameters at every stage until an appropriate error threshold is attained. Commonly used indicators for assessing the quality of system networks include the root mean square error (RMSE), coefficient of variation of RMSE (CV-RMSE), mean bias error (MBE), normalized mean bias error, mean absolute error, and mean absolute percentage error (R2). A comparison between the error variance of the average worth (y_k) and the calculated value (x_k) can be made using the number of measurements N, which can be ascertained by employing the RMSE (Equation (6)) and CV-RMSE (Equation (7)).

$$RMSE = \sqrt{\frac{\sum_{K=1}^{N}(y_{k-}x_k)2}{N}} \tag{6}$$

$$CV - RMSE = \frac{RMSE}{x} \times 100\% \tag{7}$$

The MBE measures the average point error, which indicates the general performance of the predicted result with respect to the linear function of the sample (Equation (8)).

$$obj_i = min \sum_{i=1}^{n} MBE, RMSE, E_i, t_i, \tag{8}$$

where E is the appliance energy, t is the temporal complexity function, and the MBE RMSE specifies the error functions. Because AI uses vast quantities of measurements and advanced technologies, it is crucial for both energy industries. In particular, AI may promote the RE market through improved wind industry monitoring, operation, management, and preservation as well as timely system operation and management. Considering the

unpredictability and instability of intermittent renewable energy, the combination of RE and power sources is related to several key AI technologies, including personnel security, accurate predictive forecasting, and weather forecasting, macroeconomic and grid storage techniques [20–22].

4. Conclusions

The present study has undertaken an examination of the urgent matter pertaining to sustainable power in the context of the current climate change and its concomitant distressing phenomenon of melting. The prime aim of this study was to examine the application of AI in forecasting sustainable power in light of the intricate nature of various environmental factors in renewable energy systems. Employing closed mathematical models to represent bioenergy systems is deemed unsuitable due to their complexity. A considerable body of literature has been generated on this subject, illuminating the burgeoning significance of precise computation pertaining to renewable energy. Notwithstanding, the preponderant share of anticipations regarding AI technologies vis-à-vis renewable energy pertains principally to forecasting outcomes related to solar and wind energy. The aforementioned findings imply that forthcoming investigations might elucidate alternative dispatchable projections, such as hydropower, hydroelectric power, wave action, pressured water, and geothermal, as potential aspects worthy of consideration.

The application of quantum-enhanced AI has evinced the efficacy of establishing a novel renewable energy reservoir, engendering a pronounced reduction in superfluous electricity consumption, accumulation, and recycling. Despite the absence of a consolidated quantum learning framework, the mutually beneficial partnership between AI and quantum material science has the potential to revolutionize these disciplines. Continued investment in this particular model may generate the widespread availability of eco-friendly energy, consequently fostering advantageous outcomes for both the environment and mankind. Hence, it is of utmost importance to continue investigating the deployment of AI and quantum physics in the realm of sustainable energy to alleviate the consequences of climate change and endorse sustainable energy for future generations.

4.1. Implications of the study

The results derived from this investigation carry substantial ramifications for the advancement and execution of sustainable energy systems. This study shows the capacity of AI and quantum physics to advance the creation of fresh, eco-friendly energy sources, thereby opening up novel prospects for exploration and innovation in this domain. This emphasizes the imperative requirement for escalated investment along with comprehensive research pertaining to viable and sustainable energy sources, particularly in the face of an exigent need to alleviate the catastrophic consequences of global climate change.

4.2. Limitations and future studies

This exploratory investigation has apprised significant discernments pertaining to the utilization of AI and quantum physics in sustainable energy. However, it is necessary to acknowledge the limitations inherent to this study. The present inquiry prioritized the examination of solar and wind energy prognostications, revealing the possibility of further investigation

into alternative dispatchable projections in future research endeavors. Furthermore, a comprehensive quantum learning theory has not been developed, and various inquiries have been made regarding the utilization of quantum computers for scrutinizing immense collections of quantum information. Prospective investigations could examine these domains to foster a more holistic understanding of the potential of AI and quantum physics in sustainable energy systems. Subsequent investigations could further delve into the conceivable abilities of novel technologies, specifically blockchain and machine learning, in the advancement and execution of sustainable energy systems. In summary, the current investigation provides a strong basis for forthcoming inquiries in this domain and underscores the imperative of sustained ingenuity and financial commitment toward the adoption of renewable energy sources.

Recommendations

By developing a forecast application that allows all users to view the number of resources available at a specific time interval, the energy management system may be used in multiple industries to control various devices in the network. Even the proposed AI method can be used to make the correct judgment regarding the various energies that need to be delivered at the right times in emergency situations.

To foster the advancement of sustainable energy development through the utilization of AI and QC, prominent commercial entities and policymakers are encouraged to undertake a series of definitive actions. It is recommended that businesses and governments form alliances with scholars and authorities specialized in the disciplines of AI and QC to maintain their cutting-edge status with respect to these technologies and exploit the specialized knowledge provided by these professionals. One possible way to achieve this is through the establishment of strategic partnerships with academic institutions, research organizations, or industry entities that possess specialized expertise in the relevant domains.

Efficient energy management systems that employ advanced technologies, such as AI and QC, have demonstrated the capability to optimize energy production, distribution, and consumption. The strategic imperative for businesses and governments is to channel resources toward the development of such systems and, subsequently, to effectively incorporate them into their existing energy infrastructure. The aforementioned systems have the potential to provide real-time data and analysis, which can lead to the optimized allocation of resources and a decrease in wastefulness. AI and QC have been identified as viable approaches for the formulation of forecasting and predictive models, which can potentially enhance the capability of businesses and governments to discern anticipated energy needs and undertake corresponding modifications in terms of production and distribution. This methodology can be used to optimize energy utilization and mitigate waste generation. To maintain a competitive advantage and enhance the efficacy of AI and QC within the energy industry, sustained investments in research and development by both private enterprises and public entities are recommended. This may involve financing the investigation of novel algorithms and methodologies, coupled with investigating the unexplored utilities of these technologies within the energy domain. Collaboration with domain experts, establishment of energy management systems, incorporation of predictive and forecasting models, creation of emergency response plans, and allocation of resources toward research and development are critical measures for endorsing sustainable energy development through the utilization of AI and QC by both businesses and policymakers. The implementation of the following measures has the potential to enhance energy generation efficiency, curtail waste, and mitigate environmental degradation arising from energy production, ultimately fostering a sustainable future.

Conflicts of Interest

The author declares that he has no conflicts of interest to this work.

Data Availability Statement

Data available on request from the corresponding author upon reasonable request.

Author Contribution Statement

Luka Baklaga: Conceptualization, Methodology, Software, Validation, Formal analysis, Investigation, Data curation, Writing – original draft, Writing – review & editing, Visualization.

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