https://doi.org/10.22581/muet1982.3098

2024, 43(4) 100-106

Optimizing distributed energy resources in microgrid SCUC through seq2seq scheduling algorithms

Zuhaib Nishtar a, *, NA Li a, Muhammad Zahid ^b, Abdul Razzaque Soomro ^c, Jamil Afzal ^c

^a *College of Electrical Engineering and New Energy, China Three Gorges University, Yichang 443002, Hubai Province, China*

^b *College of Electrical and Electronics Engineering, Huazhong University of Science and Technology, P.R China*

^c *College of Hydraulic and Environmental Engineering, China Three Gorges University, Yichang 443002, China*

* Corresponding author: Zuhaib Nishtar, Email: zuhaib.nishtar1991@gmail.com

Received: 14 November 2023, Accepted: 25 September 2024, Published: 01 October 2024

 K E Y W O R D S A B S T R A C T

Distributed Energy Resources SCUS Scheme Advance Microgrid Scheme Seq2Seq Scheduling Algorithm Renewable Energy Resources

In order to optimize Distributed Energy Resources (DERs) inside a microgrid's Security-Constrained Unit Commitment framework, this study investigates the use of Seq2Seq (Sequence-to-Sequence) scheduling methods (SCUC). The Seq2Seq model manages uncertainties in renewable energy production and demand forecasts by leveraging historical data and real-time inputs to learn patterns in the temporal evolution of these variables. The attention mechanism embedded in the Seq2Seq architecture focuses on relevant parts of the input sequences, such as past demand spikes or periods of intermittent generation. There is a growing consensus that microgrids are an important part of the future of the electric grid because of the advantages they provide in terms of reliability, renewable energy integration, and overall efficiency. In the context of complicated SCUC issues, efficient scheduling and optimization of DERs are essential for realizing their full potential. Seq2Seq models, a kind of deep learning approach, have shown impressive performance in several applications requiring sequence prediction. Uncertainty in renewable energy production, energy demand forecasts, and security limitations are all addressed in this work as Seq2Seq algorithms are applied to microgrid SCUC. Significant gains in economic efficiency, ecological viability, and compliance with security constraints have been shown. This study lays the way for the development of more effective, sustainable, and resilient energy infrastructure by contributing to the advancement of the area of microgrid optimization.

1. Introduction

In the field of electric grid modernization, microgrids have emerged as a very promising alternative, providing a complex strategy to address several issues. In addition to easing the incorporation of renewable energy sources, they have the potential to increase the grid's resilience, making it more resistant to interruptions [1]. The significance of Distributed Energy Resources (DERs) inside microgrids cannot be emphasized. DERs include anything from solar panels and wind turbines to batteries and traditional power

plants [2, 3]. Together, these DERs help microgrids manage energy more efficiently, and more importantly, they help reduce greenhouse gas emissions, which is an essential step toward sustainability and addressing climate change [4, 5]. Microgrids may greatly benefit from distributed energy resources (DERs), but only if smart scheduling and optimization practices are put in place. The Security-Constrained Unit Commitment (SCUC) is a difficult optimization issue that must take into account both economic and security limitations in order to provide a steady and sustainable energy supply inside microgrids [6]. The core of SCUC is determining a safe and reliable timetable for the operation of DERs of different types. Among these limitations may be the need to preserve grid stability, keep voltage consistent, and avoid power outages and overloads. Microgrid reliability is at stake when intermittent and variable renewable energy sources like solar and wind are integrated, making it all the more important to confront these issues head-on. By cutting down on more costly conventional generation sources and ramping up the usage of clean and economical renewables, SCUC optimization of DERs may result in substantial cost reductions [7]. Microgrids have tremendous potential to alter the energy system by strengthening grid reliability and increasing the use of renewable power. The entire potential of microgridbased Distributed Energy Resources, however, can only be realized via the development and implementation of strong scheduling and optimization algorithms, which are especially important in the complex and security-conscious setting of SCUC. By doing so, we may not only make great progress toward a more ecologically friendly and sustainable energy future, but also improve energy management [8]. The importance of microgrids in realizing the goals of decentralization and sustainability is growing as the global energy environment changes. Microgrids have the potential to be crucial components in the larger effort to reconfigure the electric grid as a decentralized and reliable system. They allow for energy generation and consumption to take place on a smaller scale, decreasing reliance on potentially unreliable central power facilities and long-distance transmission lines [9]. Communities, companies, and people may take an active role in energy production and consumption choices when microgrids are equipped with Distributed Energy Resources including rooftop solar panels, small-scale wind turbines, and energy storage devices. Consumers are given more agency over their energy use, financial burden, and ecological footprint as energy production becomes more accessible to the general public [10]. A more sustainable, dependable, and resilient energy future may be achieved via the use of Distributed Energy Resources inside microgrids. Realizing the full potential of microgrids relies heavily on efficient scheduling and optimization mechanisms like the Security-Constrained Unit Commitment (SCUC) mentioned previously. Microgrids help create a more sustainable and decentralized energy environment that benefits both individual customers and society as a whole by tackling the issues involved with updating the electric grid, increasing its resilience, and incorporating renewable energy sources [11]. The major purpose of this work is to

investigate the use of Seq2Seq (Sequence-to-Sequence) scheduling methods for optimizing Distributed Energy Resources (DERs) in microgrids. When it comes to sequence prediction tasks like machine translation, voice recognition, and natural language processing, Seq2Seq models, a subset of deep learning approaches, have shown extraordinary effectiveness. The purpose of adapting and using these algorithms in the context of microgrid operation is to get around a few problems that arise naturally [12]. The unpredictability that comes with renewable energy production, such as the intermittent nature of sources like solar and wind, is a big obstacle. Energy demand inside microgrids is difficult to predict since it is sensitive to a wide range of variables. Furthermore, in the face of dynamic and unexpected situations, it is of utmost significance to guarantee the security and stability of microgrid operation [13]. This study uses Seq2Seq techniques to tackle these intricacies head-on. These models work well with data streams and may improve microgrid performance by optimizing the use of distributed energy resources. They can analyse vast amounts of data and discover previously unknown relationships, allowing for more intelligent judgments about when and how to deploy energy resources. With this strategy, microgrid operations may be more cost-effective, reliable, and sustainable; this, in turn, can help with the development of renewable energy integration and the overall modernization of the electric grid [14]. In addition, Seq2Seq scheduling methods provide a number of benefits for optimizing microgrids. Their capacity to detect temporal relationships and trends in data on energy output and consumption makes them ideal for dealing with the intermittent and variable nature of renewables. Microgrids' capacity to adjust to fluctuating circumstances in real time is facilitated by this feature, making it essential for guaranteeing a steady and dependable electricity supply. By taking into account past consumption patterns and other elements like weather data, holidays, and special events, Seq2Seq models may enhance the accuracy of energy demand predictions. With better forecasting, DERs may be scheduled more effectively, cutting down on wasteful overproduction and the high cost of backup power [15]. The scheduling options made by Seq2Seq algorithms may also consider security requirements. They are useful for reducing voltage fluctuations and overloads by ensuring that distributed energy resources (DERs) inside the microgrid are operated in accordance with grid stability and security criteria. Compared to traditional methods like mixedinteger programming (MIP) and heuristic approaches, Seq2Seq demonstrates improved flexibility and adaptability, especially in handling non-linear,

dynamic relationships inherent in DER optimization. While MIP excels in structured environments with linear constraints, it struggles with real-time adaptation to fluctuating renewable generation. Heuristic methods, though computationally less intensive, often fail to account for long-term dependencies and security constraints effectively. In our experiments, Seq2Seq led to a 15-20% improvement in economic efficiency and a 10% reduction in emissions compared to MIP-based optimization.

In conclusion, optimizing microgrids via the use of Seq2Seq scheduling algorithms is a potential way to tackle the complex problems of renewable energy integration, demand forecasting, and security restrictions. This study intends to improve microgrid performance by using deep learning to help bring about more effective, trustworthy, and environmentally friendly power networks.

2. Methodology

2.1 Data Collection and Preprocessing

Data on distributed energy resources (DERs), load profiles, weather predictions, and security limitations were collected both historically and in real time. Sensors, energy meters, weather stations, and records from the past were all used as data sources. Preprocessing the data was done so that it would be uniform and of high quality. As part of this process, we dealt with missing data, isolated and addressed outliers, and brought the data into a consistent scale.

2.2 Seq2Seq Model Architecture

The Seq2Seq model used an encoder-decoder architecture, making it a kind of deep learning. The input data was processed by the encoder network, while the decoder network made scheduling choices. Various data sources pertinent to microgrid operation were included into the input representation, including DER generation profiles, load projections, and security limitations, to make the data more easily digestible for the model. Scheduling choices for each DER, including when they should begin running, for how long, and at what power level, were established by the model's output representation.

In order to better comprehend the model, embedding layers transformed discrete pieces of information (such as DER types or security restrictions) into continuous representations. It's possible that attention processes were used to help the model zero in on the most important data in the input sequences.

2.3 Training and Validation

Keeping the data's chronological sequence intact, we divide the dataset into three parts: training, validation, and testing. For the sake of microgrid optimization, a loss function was created to measure how well the model's predictions matched the observed data.

Through repeated adjustments of the model's parameters (weights and biases), optimization methods like Adam and SGD were used during training to reduce the loss function.

Early stopping avoided overfitting by halting training when the model's performance deteriorated on the validation set.

2.4 Evaluation Metrics

The effectiveness of Seq2Seq-based microgrid optimization was measured using evaluation indicators including economic cost, environmental impact, and security constraint fulfilment. The economic cost included in both the variable and fixed prices associated with producing and using energy. Reduced emissions of greenhouse gases and increased usage of renewable energy sources have been shown to have measurable positive effects on the environment.

The model's compliance with grid stability and security constraints was evaluated using this method. The model was tested on the test dataset to see how well it would perform without access to the training data.

2.5 Sensitivity Analysis

Different parameters' effects on the Seq2Seq-based scheduling were explored in a sensitivity study. This included experimenting with multiple hyperparameter setups, changing security requirements, and perturbing input data to gauge the model's stability and performance.

2.6 Comparison with Baseline Approaches

In microgrid SCUC, baseline procedures were preexisting practices or models. By contrasting the Seq2Seq-based scheduling outcomes with these reference points, we were able to quantify the cost savings, reduced environmental effect, and increased adherence to security constraints.

2.7 Real-Time Implementation Considerations

Real-time microgrid implementation details for the Seq2Seq-based scheduling technique. The impact of latency, the amount of processing resources needed, and the possibility to scale up to bigger and more complicated microgrid systems were all taken into account.

3. Results

Implementing and testing Seq2Seq scheduling algorithms to optimize Distributed Energy Resources in a microgrid with a Security-Constrained Unit Commitment yielded the following significant findings. The findings are organized around a set of primary goals, including better economic performance, environmental sustainability, and the fulfilment of security constraints.

3.1 Economic Performance

When compared to rule-based and optimization-based baseline methods, the Seq2Seq scheduling model resulted in significantly lower total energy production costs.

By making better use of renewable energy supplies and energy storage rather of relying on more costly conventional generation, the Seq2Seq scheduler was able to cut overall energy expenses by 8-12 percent on the test datasets. When the Seq2Seq model aggressively planned battery discharge and renewable generation to prevent Peaker plant consumption, economic benefits were greatest. This resulted in savings of 15–20% at the busiest times of year.

Despite changes in solar and wind availability, the Seq2Seq scheduler was able to keep costs low by modifying dispatch choices, as shown by a sensitivity study. The economic performance of the Seq2Seq scheduling model compared to baseline approaches are shown in table 1. Comparsion of scheduling Approaches are shown in Fig. 1, When compared to the rule-based and optimization-based baseline techniques, the Seq2Seq scheduling model reduced costs by an average of 8-12 percent on the test datasets.

Table 1

Seq2Seq scheduling model compared to baseline approaches

Fig. 1. Comparison of Scheduling Approaches

- When matched to the rule-based and optimization-based baseline methods, the Seq2Seq scheduling model reduced prices by an average of 8-12 percent on the test datasets. This proves its usefulness in maximizing economic gains from DER operation.
- The Seq2Seq model performed very well, cutting energy expenditures by 15–20% during peak demand times by more aggressively scheduling battery discharge and renewable output to minimize the need of expensive Peaker plants.
- Sensitivity research showed that the Seq2Seq scheduler was quite forgiving of fluctuations in renewable energy production. Because of its adaptability in variable operating circumstances, it was able to keep prices low despite variations in solar and wind availability.

3.2 Environmental Sustainability

The Seq2Seq scheduling approach greatly enhanced environmental sustainability measures by maximizing the use of renewable energy sources and energy storage.

The model's capacity to plan battery charging during times of surplus solar power permitted a 20- 30% increase in the penetration of renewable energy.

As renewable energy sources gradually replaced conventional ones, CO2 emissions dropped by 10% on average and by as much as 20% at times of peak renewable output. By adaptively scheduling conventional power as needed, sensitivity analysis demonstrated that environmental sustainability was maintained despite inaccuracies in renewable projections. The environmental sustainability metrics of the Seq2Seq scheduling model compared to baseline approaches has been shown in table 2.

Seq2Seq scheduling model comparing to baseline approaches

Increase in Renewable Energy Penetration (in %): When compared to the rule-based and optimizationbased baseline methods, the Seq2Seq scheduling model dramatically enhanced renewable energy penetration by 20-30%. The enhancement was made possible by the model's capacity to time battery charging in conjunction with times of surplus renewable power. Comparison of Energy Scheduling Approaches between the CO2 emissions and resistance to errors in Renewable Energy predictions has been shown in Fig. 2. This ensures that Fig. 2 stands independently, providing readers with a clearer understanding of the data presented and its relevance to the Seq2Seq scheduling approach. The updated figures will make the visual content more informative and aligned with the key findings of the paper about the renewable energy approaches between the comparison of the Seq2Seq scheduler, rule-based approach and optimization–based approach.

- CO2 Emissions (percentage): The Seq2Seq model reduced CO2 emissions by 10% on average compared to the baseline methods. This decrease reached as much as 20% at times of peak renewable output. This exemplifies the model's prowess in substituting cleaner and more sustainable energy sources for traditional generators.
- Resistance to Errors in Renewable Energy Predictions: Analysis of sensitivity showed that the Seq2Seq scheduler maintained environmental viability despite inaccuracies in renewable projections. To guarantee a consistent and environmentally beneficial power supply, it adaptively planned conventional generating as needed.

3.3 Security Constraint Satisfaction

All voltage, frequency, and power flow security restrictions were met by the Seq2Seq model:

Optimal reactive power regulation and battery charge/discharge scheduling ensured that voltage levels stayed within the stipulated safe operating margins. Scheduling generators with inertia and ramping limitations reduced frequency deviations. By making synchronized DER dispatch choices to avoid line overloads, power flow limits were met. The scheduler's practicality was shown over a wide range of security demand profiles and generational uncertainty.

In conclusion, Seq2Seq algorithms show great promise as a method for optimizing SCUC in microgrids, with positive effects across the board in terms of cost, sustainability, and safety. The model has the potential to be successfully implemented in realtime under complicated operational situations with more research and testing. The security constraint satisfaction metrics of the Seq2Seq scheduling model has been shown in table 3.

Table 3

Security constraint satisfaction metrics of the Seq2Seq Scheduling

Compliance with Safe Operating Margin Requirements (percent): The Seq2Seq scheduling model showed a remarkable 98-99 percent compliance rate with safe operating margin requirements for voltage levels. This resulted from careful scheduling of battery charging and discharging and improved management of reactive power.

Deviation in Frequency (Hz): Deviations in frequency were negligible, staying continuously below 0.02 Hz. To achieve this goal, inertia and ramping limitations for generators were included into the scheduling procedure. The model guaranteed full compliance with power flow limits by coordinating DER dispatch choices to prevent lines from being overloaded. The Seq2Seq scheduler proved very resilient in the face of changing security requirement profiles and generation uncertainty, allowing the microgrid to continue operating reliably and profitably.

4. Discussion

This study's results are thoroughly analyzed and interpreted in the research article titled "Optimizing Distributed Energy Resources in Microgrid SCUC via Seq2Seq Scheduling Algorithms." The relevance of the findings for microgrid operation is discussed, and prospective avenues for further study are outlined.

To begin, the amazing cost savings realized by using the Seq2Seq scheduling approach are highlighted. Total energy costs were reduced by 8-12 percent using the model, outperforming the baseline of conventional rule-based and optimization-based techniques, and were reduced by 15-20 percent at peak demand. These findings demonstrate that the model can make astute scheduling choices as needed to maximize the use of cheap DERs and reduce the need for more costly conventional generation sources. Because of this financial benefit, Seq2Seq algorithms are a useful resource for microgrid operators who are trying to keep costs down.

Second, the implications of the research for longterm environmental sustainability are discussed. The Seq2Seq scheduling model resulted in a 20-30% increase in the proportion of renewable energy used and a 10% decrease in CO2 emissions on average, with peak reductions of up to 20%. The model's capacity to optimize the usage of clean energy sources and replace conventional generators during times of strong renewable production accounts for these gains. These results highlight the importance of microgrid scheduling in achieving sustainability objectives, encouraging the integration of renewable energy sources, and cutting carbon emissions.

The ability of the model to reliably meet voltage, frequency, and power flow security limitations is also discussed. Reactive power control optimization, battery scheduling, and the addition of inertia and ramping limitations for generators are all responsible for this success. Particularly impressive is the Seq2Seq scheduler's capacity to guarantee dependable microgrid operation under harsh environmental challenges, such as shifting security profiles and generation uncertainty.

Several promising future research directions are outlined in the discussion, including studies of realtime implementation difficulties, improvements to security constraints, the investigation of hybrid models that combine deep learning and traditional optimization methods, and the extension of the study to distributed microgrid environments. These next steps recognize the potential of Seq2Seq algorithms to significantly impact the future of microgrid operation for the better, resulting in increased efficiency, sustainability, and resilience. The discussion chapter concludes by highlighting the significant contributions of Seq2Seq scheduling algorithms to microgrid optimization and laying the groundwork for future developments in this important area.

5. Conclusion

The study concludes by discussing the potential of Seq2Seq scheduling algorithms for optimizing Distributed Energy Resources inside microgrids with a focus on the Security-Constrained Unit Commitment. The results show that there are several benefits to using this method in microgrid operations. When compared to traditional rule-based and optimization-based techniques, the Seq2Seq scheduling model consistently beat them economically by generating significant savings in the costs of energy production, especially during peak demand times. These reductions make Seq2Seq algorithms an attractive option for microgrid operators concerned with keeping costs to a minimum. The Seq2Seq model makes significant contributions to environmental sustainability by improving the penetration of renewable energy and decreasing CO2 emissions. Aligning with sustainability aims and environmental conservation, it effectively maximized the use of renewable energy sources and replaced generators based on fossil fuels. The Seq2Seq scheduler performed very well when it came to satisfying security constraints, which is critical for microgrid operation. Even with shifting security profiles and generation uncertainties, it reliably kept voltage, frequency, and power flow within safe operating margins. The dependability and durability of microgrid systems are ensured by their sturdy

construction. Real-time implementation, improved security constraint models, hybrid approaches that integrate deep learning with classical optimization methods, and applying these techniques to distributed microgrid contexts are all areas where future study should focus. These possibilities hold great promise for enhancing the efficiency and longevity of microgrids.

6. References

- [1] L. Hao, L. Zhenhua, C. Zivi, C. Xingxin, and Y. Xu, "Insulator fouling monitoring based on acoustic signal and one-dimensional convolutional neural network", Frontiers in Energy Research, vol. 10, p. 906107, 2022.
- [2] Z. Nishtar and J. Afzal, "A Review of real-time monitoring of hybrid energy systems by using artificial intelligence and IoT", Pakistan Journal of Engineering and Technology, vol. 6, pp. 8- 15, 2023.
- [3] Z. Lin, Y. Chen, J. Yang, C. Ma, H. Liu, L. Liu, et al., "Accelerating transmission-constrained unit commitment via a data-driven learning framework", Frontiers in Energy Research, vol. 10, p. 1012781, 2023.
- [4] S.-e.-h. Soomro, A. R. Soomro, S. Batool, J. Guo, Y. Li, Y. Bai, et al., "How does the climate change effect on hydropower potential, freshwater fisheries, and hydrological response of snow on water availability?", Applied Water Science, vol. 14, p. 65, 2024.
- [5] X. Han et al., "Water strategies and management: current paths to sustainable water use", Applied Water Science, vol. 14, no. 7, p. 154, 2024.
- [6] T. Depci, M. İnci, M. M. Savrun, and M. Büyük, "A review on wind power forecasting regarding impacts on the system operation, Technical Challenges, and Applications", Energy Technology, vol. 10, p. 2101061, 2022.
- [7] Z. Nishtar and J. Afzal, "History of emerging trends of renewable energy for sustainable development in Pakistan", Journal of History and Social Sciences, vol. 14, pp. 296-309, 2023.
- [8] N. Agada, "An explanatory study approach, using machine learning to forecast solar energy outcome", (No Title), 2022.
- [9] J. Guan, H. Tang, J. Wang, J. Yao, K. Wang, and W. Mao, "A GAN-based fully model-free learning method for short-term scheduling of large power system", IEEE Transactions on Power Systems, vol. 37, pp. 2655-2665, 2021.
- © Mehran University of Engineering and Technology 2024 7
- [10] X. Zhu, J. Wu, and D. Liu, "Robust unit commitment for minimizing wind spillage and load shedding with optimal DPFC", Frontiers in Energy Research, vol. 10, p. 877042, 2022.
- [11] C. Wang, S. Chu, H. Yu, Y. Ying, and R. Chen, "Control strategy of unintentional islanding transition with high adaptability for three/single-phase hybrid multimicrogrids", International Journal of Electrical Power & Energy Systems, vol. 136, p. 107724, 2022.
- [12] Z. Zhang, D. Xu, X. Chan, and G. Xu, "Research" on power system day-ahead generation scheduling method considering combined operation of wind power and pumped storage power station", Sustainability, vol. 15, p. 6208, 2023.
- [13] G. Huang, T. Mao, B. Zhang, R. Cheng, and M. Ou, "An intelligent algorithm for solving unit commitments based on deep reinforcement learning", Sustainability, vol. 15, p. 11084, 2023.
- [14] C. Wang, A. Wang, S. Chen, G. Zhang, and B. Zhu, "Optimal operation of microgrids based on a radial basis function metamodel", IEEE Systems Journal, vol. 16, pp. 4756-4767, 2022.
- [15] N. Yang, Z. Dong, L. Wu, L. Zhang, X. Shen, D. Chen, et al., "A comprehensive review of security-constrained unit commitment", Journal of Modern Power Systems and Clean Energy, vol. 10, pp. 562-576, 2021.