

**OPTIMASI LSTM MENGGUNAKAN MODIFIKASI ALGORITMA
GENETIKA DALAM PERAMALAN CUACA DI KOTA SURABAYA**

SKRIPSI



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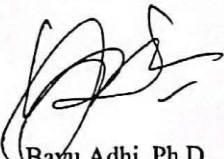
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ABSTRAK

Optimasi Lstm Menggunakan Modifikasi Algoritma Genetika Dalam Peramalan Cuaca Di Kota Surabaya

Oleh :

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Perubahan cuaca yang sering terjadi dapat memberikan dampak signifikan terhadap berbagai aspek kehidupan manusia, termasuk di Kota Surabaya. Untuk mengurangi dampak negatif tersebut, diperlukan sistem prediksi cuaca yang akurat. Metode *Genetic Algorithm-Long Short-Term Memory* (GA-LSTM) diusulkan untuk meningkatkan akurasi prediksi cuaca. Penelitian ini bertujuan untuk mengimplementasikan dan mengevaluasi GA-LSTM dalam meningkatkan akurasi prediksi dibandingkan dengan metode optimasi lainnya. Model prediksi dibangun menggunakan data dari 10 parameter cuaca, dengan *Long Short-Term Memory* (LSTM) sebagai model prediksi utama dan *Genetic Algorithm* (GA) sebagai teknik optimasi *hyperparameter*. Hasil penelitian ini menunjukkan bahwa GA-LSTM mampu menghasilkan prediksi cuaca yang lebih akurat, dengan nilai *Mean Squared Error* (MSE) 4.04% lebih rendah dan *Root Mean Squared Error* (RMSE) 2.05% lebih rendah dibandingkan *Bayesian Optimization-Long Short-Term Memory* (BO-LSTM). Penerapan metode GA-LSTM menunjukkan potensi signifikan dalam meningkatkan akurasi prediksi cuaca, yang dapat membantu mitigasi dampak perubahan cuaca di Kota Surabaya.

Kata Kunci : Peramalan, Cuaca, *Long Short Term Memory*, *Genetic Algorithm*, *Tuning Hyperparameter*, GA-LSTM

ABSTRACT

Lstm Optimization Using Modified Genetic Algorithm in Weather Forecasting in Surabaya City

Oleh :

Nafila Rossiana Putri

Frequent weather changes can have quite significant impacts on various aspects of human life, including in the city of Surabaya. To reduce these negative impacts, an accurate weather prediction system is needed. The Genetic Algorithm-Long Short-Term Memory (GA-LSTM) method is proposed to improve the accuracy of weather prediction. This ssptimization methods. The prediction model is built using data from 10 weather parameters, with Long Short-Term Memory (LSTM) as the main prediction model and Genetic Algorithm (GA) as the hyperparameter optimization technique. The results of this study indicate that GA-LSTM is able to produce more accurate weather predictions, with a Mean Squared Error (MSE) value of 4.04% lower and a Root Mean Squared Error (RMSE) of 2.05% lower than Bayesian Optimization-Long Short-Term Memory (BO-LSTM). The application of the GA-LSTM method shows significant potential in improving the accuracy of weather predictions, which can help reduce the impact of weather changes in the city of Surabaya.

Keywords: *Forecasting, Weather, Long Short Term Memory, Genetic Algorithm, Tuning Hyperparameter, GA-LSTM.*

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DAFTAR PUSTAKA

- Scientific Reports*, 11(1), 20742. <https://doi.org/10.1038/S41598-021-99999-Y>
- Ahmad, F. (2020). Penentuan Metode Peramalan Pada Produksi Part New Granada Bowl St Di Pt.X. *Jisi: Jurnal Integrasi Sistem Industri*, 7(1), 31. <https://doi.org/10.24853/Jisi.7.1.31-39>
- Ahmed, D. M., Hassan, M. M., & Mstafa, R. J. (2022). A Review On Deep Sequential Models For Forecasting Time Series Data. *Applied Computational Intelligence And Soft Computing*, 2022, 1–19. <https://doi.org/10.1155/2022/6596397>
- Alibrahim, H., & Ludwig, S. A. (2021). Hyperparameter Optimization: Comparing Genetic Algorithm against Grid Search and Bayesian Optimization. 2021 *IEEE Congress on Evolutionary Computation (CEC)*, 1551–1559. <https://doi.org/10.1109/CEC45853.2021.9504761>
- Al Kiramy, R., Permana, I., Marsal, A., Munzir, M. R., & Megawati, M. (2024). Perbandingan Performa Algoritma Rnn Dan Lstm Dalam Prediksi Jumlah Jamaah Umrah Pada Pt. Hajar Aswad: Comparison Of Rnn And Lstm Algorithm Performance In Predicting The Number Of Umrah Pilgrims At Pt. Hajar Aswad. *Malcom: Indonesian Journal Of Machine Learning And Computer Science*, 4(4), 1224–1234. <https://doi.org/10.57152/Malcom.V4i4.1373>
- Alkafaween, E., Hassanat, A. B. A., & Tarawneh, S. (2021). Improving Initial Population For Genetic Algorithm Using The Multi Linear Regression Based Technique (Mlrbt). *Communications - Scientific Letters Of The*

University Of Zilina, 23(1), E1–E10.

<https://doi.org/10.26552/Com.C.2021.1.E1-E10>

Alkhairi, P., Windarto, A. P., & Efendi, M. M. (2024). *Optimasi Lstm Mengurangi Overfitting Untuk Klasifikasi Teks Menggunakan Kumpulan Data Ulasan Film Kaggle Imdb*. 6(2).

Alzubaidi, L., Zhang, J., Humaidi, A. J., Al-Dujaili, A., Duan, Y., Al-Shamma, O., Santamaría, J., Fadhel, M. A., Al-Amidie, M., & Farhan, L. (2021). Review Of Deep Learning: Concepts, Cnn Architectures, Challenges, Applications, Future Directions. *Journal Of Big Data*, 8(1), 53.
<https://doi.org/10.1186/S40537-021-00444-8>

Arkadia, A., Hananto, B., & Prasvita, D. S. (2022). *Optimasi Long Short Term Memory Dengan Adam Menggunakan Data Udara Kota Dki Jakarta*.

Bacanin, N., Jovanovic, L., Zivkovic, M., Kandasamy, V., Antonijevic, M., Deveci, M., & Strumberger, I. (2023). Multivariate Energy Forecasting Via Metaheuristic Tuned Long-Short Term Memory And Gated Recurrent Unit Neural Networks. *Information Sciences*, 642, 119122.
<https://doi.org/10.1016/J.Ins.2023.119122>

Cai, M., Pipattanasomporn, M., & Rahman, S. (2019). Day-Ahead Building-Level Load Forecasts Using Deep Learning Vs. Traditional Time-Series Techniques. *Applied Energy*, 236, 1078–1088.
<https://doi.org/10.1016/J.Apenergy.2018.12.042>

Chatfield, C. (1988). What Is The ‘Best’ Method Of Forecasting? *Journal Of Applied Statistics*, 15(1), 19–38.
<https://doi.org/10.1080/02664768800000003>

- Chung, H., & Shin, K. (2018). Genetic Algorithm-Optimized Long Short-Term Memory Network For Stock Market Prediction. *Sustainability*, 10(10), 3765. <https://doi.org/10.3390/Su10103765>
- Dean, J. (2022). A Golden Decade Of Deep Learning: Computing Systems & Applications. *Daedalus*, 151(2), 58–74. https://doi.org/10.1162/Daed_A_01900
- Deswal, V., & Kumar, D. (2024). Hyperparameter Tuning Of The Lstm Model For Stock Price Prediction. *International Journal Of Intelligent Systems And Applications In Engineering*.
- Ekundayo, O. S., & Ezugwu, A. E.-S. (2024). Deep Learning: Historical Overview From Inception To Actualization, Models, Applications And Future Trends. <https://doi.org/10.36227/techrxiv.171504655.52310452/V1>
- Elmousaid, R., Drioui, N., Elgouri, R., Agueny, H., & Adnani, Y. (2024). Accurate Short-Term GHI Forecasting Using A Novel Temporal Convolutional Network Model. *E-Prime - Advances In Electrical Engineering, Electronics And Energy*, 9, 100667. <https://doi.org/10.1016/J.Prime.2024.100667>
- Essai Ali, M. H., Abdel-Raman, A. B., & Badry, E. A. (2022). Developing Novel Activation Functions Based Deep Learning Lstm For Classification. *Ieee Access*, 10, 97259–97275. <https://doi.org/10.1109/Access.2022.3205774>
- Fadholi, A. (2013). Analisis Data Arah Dan Kecepatan Angin Landas Pacu (Runway) Menggunakan Aplikasi Windrose Plot (Wrplot). 9.
- Farhah, A., Prasasti, A. L., & Paryasto, M. W. (2021). Implementasi Recurrent Neural Network Dalam Memprediksi Kepadatan Restoran Berbasis Lstm. *Jurnal Media Informatika Budidarma*, 5(2), 524.

<https://doi.org/10.30865/Mib.V5i2.2916>

Ghozali, I. (2018). Aplikasi analisis multivariate dengan program IBM SPSS 23.

Gülmez, B. (2021). *Prediction Of Retail Prices Of Roasted Coffee By Time Series Analysis*. <https://doi.org/10.5281/Zenodo.7915779>

Gülmez, B. (2023). Stock Price Prediction With Optimized Deep Lstm Network With Artificial Rabbits Optimization Algorithm. *Expert Systems With Applications*, 227, 120346. <https://doi.org/10.1016/J.Eswa.2023.120346>

Hariz, F. A., Yulita, I. N., & Suryana, I. (2022). Human Activity Recognition Berdasarkan Tangkapan Webcam Menggunakan Metode Convolutional Neural Network (Cnn) Dengan Arsitektur Mobilenet. *Jitsi : Jurnal Ilmiah Teknologi Sistem Informasi*, 3(4), 103–115. <https://doi.org/10.30630/Jitsi.3.4.97>

Hassanat, A., Almohammadi, K., Alkafaween, E., Abunawas, E., Hammouri, A., & Prasath, V. B. S. (2019). Choosing Mutation And Crossover Ratios For Genetic Algorithms—A Review With A New Dynamic Approach. *Information*, 10(12), 390. <https://doi.org/10.3390/Info10120390>

Hewage, P., Trovati, M., Pereira, E., & Behera, A. (2021). Deep Learning-Based Effective Fine-Grained Weather Forecasting Model. *Pattern Analysis And Applications*, 24(1), 343–366. <https://doi.org/10.1007/S10044-020-00898-1>

Hochreiter, S., & Schmidhuber, J. (1997). Long Short-Term Memory. *Neural Computation*, 9(8), 1735–1780. <https://doi.org/10.1162/Neco.1997.9.8.1735>

Husein, M., & Chung, I.-Y. (2019). Day-Ahead Solar Irradiance Forecasting For

- Microgrids Using A Long Short-Term Memory Recurrent Neural Network:
A Deep Learning Approach. *Energies*, 12(10), 1856.
<https://doi.org/10.3390/en12101856>
- Iman, F. N., & Wulandari, D. (2023). *Prediksi Harga Saham Menggunakan Metode Long Short Term Memory*. 1(3).
- Indarwati, S., Respati, S. M. B., & Darmanto, D. (2019). Kebutuhan Daya Pada Air Conditioner Saat Terjadi Perbedaan Suhu Dan Kelembaban. *Jurnal Ilmiah Momentum*, 15(1). <https://doi.org/10.36499/Jim.V15i1.2666>
- Kara, A. (2021). Multi-Step Influenza Outbreak Forecasting Using Deep Lstm Network And Genetic Algorithm. *Expert Systems With Applications*, 180, 115153. <https://doi.org/10.1016/j.eswa.2021.115153>
- Karevan, Z., & Suykens, J. A. K. (2020). Transductive Lstm For Time-Series Prediction: An Application To Weather Forecasting. *Neural Networks*, 125, 1–9. <https://doi.org/10.1016/j.neunet.2019.12.030>
- Katoch, S., Chauhan, S. S., & Kumar, V. (2021). A Review On Genetic Algorithm: Past, Present, And Future. *Multimedia Tools And Applications*, 80(5), 8091–8126. <https://doi.org/10.1007/s11042-020-10139-6>
- Khairunisa, N. K., & Hendikawati, P. (2024). Long Short-Term Memory And Gated Recurrent Unit Modeling For Stock Price Forecasting. *Jurnal Matematika, Statistika Dan Komputasi*, 21(1), 321–333. <https://doi.org/10.20956/j.v21i1.35930>
- Kilichev, D., & Kim, W. (2023). Hyperparameter Optimization For 1d-Cnn-Based Network Intrusion Detection Using Ga And Pso. *Mathematics*, 11(17), 3724. <https://doi.org/10.3390/math11173724>

- Letta, M., & Tol, R. S. J. (2019). Weather, Climate And Total Factor Productivity. *Environmental And Resource Economics*, 73(1), 283–305.
<https://doi.org/10.1007/S10640-018-0262-8>
- Li, Y., Tang, F., Gao, X., Zhang, T., Qi, J., Xie, J., Li, X., & Guo, Y. (2022). Numerical Weather Prediction Correction Strategy For Short-Term Wind Power Forecasting Based On Bidirectional Gated Recurrent Unit And Xgboost. *Frontiers In Energy Research*, 9, 836144.
<https://doi.org/10.3389/Fenrg.2021.836144>
- Lin, Z., Shi, Y., Chen, B., Liu, S., Ge, Y., Ma, J., Yang, L., & Lin, Z. (2022). Early Warning Method For Power Supply Service Quality Based On Three-Way Decision Theory And Lstm Neural Network. *Energy Reports*, 8, 537–543.
<https://doi.org/10.1016/J.Egyr.2022.02.243>
- Liu, R., & Liu, L. (2019). Predicting Housing Price In China Based On Long Short-Term Memory Incorporating Modified Genetic Algorithm. *Soft Computing*, 23(22), 11829–11838. <https://doi.org/10.1007/S00500-018-03739-W>
- Luo, X. J., & Oyedele, L. O. (2021). Forecasting Building Energy Consumption: Adaptive Long-Short Term Memory Neural Networks Driven By Genetic Algorithm. *Advanced Engineering Informatics*, 50, 101357.
<https://doi.org/10.1016/J.Aei.2021.101357>
- Maricar, M. A. (N.D.). *Analisa Perbandingan Nilai Akurasi Moving Average Dan Exponential Smoothing Untuk Sistem Peramalan Pendapatan Pada Perusahaan Xyz*. 13(2).
- Melanie, M. (1998). *An Introduction To Genetic Algorithms*.
- Meliuwati, P., & Kurniati, E. (2022). Ekstraksi Data Digital Menggunakan Teknik

- Max Pooling Dan Average Pooling. *Jurnal Riset Matematika*, 137–144.
<https://doi.org/10.29313/Jrm.V2i2.1338>
- Mentari, S. A., Susanto, T., & Budipriyanto, A. (2022). *Analisis Forecasting Erratic Demand Dan Safety Stock Laptop Dan Notebook User (Studi Kasus Di Pt Bank Xyz)*.
- Muhuri, P. S., Chatterjee, P., Yuan, X., Roy, K., & Esterline, A. (2020). Using A Long Short-Term Memory Recurrent Neural Network (Lstm-Rnn) To Classify Network Attacks. *Information*, 11(5), 243.
<https://doi.org/10.3390/Info11050243>
- Nurhakiki, J., & Yahfizham, Y. (2024). *Studi Kepustakaan: Pengenalan 4 Algoritma Pada Pembelajaran Deep Learning Beserta Implikasinya*.
- Nurhambali, M. R., Angraini, Y., & Fitrianto, A. (2024). Implementation Of Long Short-Term Memory For Gold Prices Forecasting. *Malaysian Journal Of Mathematical Sciences*, 18(2), 399–422.
<https://doi.org/10.47836/Mjms.18.2.11>
- Öztürk, M. M. (2023). Hyperparameter Optimization Of A Parallelized Lstm For Time Series Prediction. *Vietnam Journal Of Computer Science*, 10(03), 303–328. <https://doi.org/10.1142/S2196888823500033>
- Pahlevi, M. R., Kusrini, K., & Hidayat, T. (2023). Comparison Of Lstm And Gru Models For Forex Prediction. *Sinkron*, 8(4), 2254–2263.
<https://doi.org/10.33395/Sinkron.V8i4.12709>
- Patil, J., & Shinde, N. (2022). *Time Series Weather Forecasting Techniques: Literature Survey*. 09(06).
- Probst, P., Bischl, B., & Boulesteix, A.-L. (2018). *Tunability: Importance Of*

Hyperparameters Of Machine Learning Algorithms (Arxiv:1802.09596).
Arxiv. [Http://Arxiv.Org/Abs/1802.09596](http://arxiv.org/abs/1802.09596)

Rajwar, K., Deep, K., & Das, S. (2023). An Exhaustive Review Of The Metaheuristic Algorithms For Search And Optimization: Taxonomy, Applications, And Open Challenges. *Artificial Intelligence Review*, 56(11), 13187–13257. [Https://Doi.Org/10.1007/S10462-023-10470-Y](https://doi.org/10.1007/S10462-023-10470-Y)

Rawat, B., Duwal, D., Phuyal, S., & Pant, A. (2022). A Comparative Review Between Various Selection Techniques In Genetic Algorithm For Finding Optimal Solutions. *International Journal Of Computer Sciences And Engineering*.

Rochmawati, N., Hidayati, H. B., Yamasari, Y., Tjahyaningtijas, H. P. A., Yustanti, W., & Prihanto, A. (2021). Analisa Learning Rate Dan *Batch size* Pada Klasifikasi Covid Menggunakan Deep Learning Dengan Optimizer Adam. *Journal Of Information Engineering And Educational Technology*, 5(2), 44–48. [Https://Doi.Org/10.26740/Jieet.V5n2.P44-48](https://doi.org/10.26740/Jieet.V5n2.P44-48)

Ruswanti, D. (2020). *Pengukuran Performa Support Vector Machine Dan Neural Netwok Dalam Meramalkan Tingkat Curah Hujan*. 13.

Saputra, R., Sunardiyo, S., Nugroho, A., & Subiyanto, S. (2023). Analisis Arsitektur Jaringan Syaraf Tiruan-Multilayer Perceptron Untuk Efektivitas Estimasi Beban Energi Listrik Pt. Pln (Persero) Up3 Salatiga. *Elkomika: Jurnal Teknik Energi Elektrik, Teknik Telekomunikasi, & Teknik Elektronika*, 11(3), 664. [Https://Doi.Org/10.26760/Elkomika.V11i3.664](https://doi.org/10.26760/Elkomika.V11i3.664)

Sari, K. P. (2021). Analisis Perbedaan Suhu Dan Kelembaban Ruangan Pada Kamar Berdinding Keramik. *Jurnal Inkofar*, 1(2).

<https://doi.org/10.46846/Jurnalinkofar.V1i2.156>

Sarker, I. H. (2021). Deep Learning: A Comprehensive Overview On Techniques, Taxonomy, Applications And Research Directions. *Sn Computer Science*, 2(6), 420. <https://doi.org/10.1007/S42979-021-00815-1>

Sdgs. (2024). Sdgs Indonesia. <https://sdgs.bappenas.go.id/>

Shahid, F., Zameer, A., & Muneeb, M. (2021). A Novel Genetic Lstm Model For Wind Power Forecast. *Energy*, 223, 120069. <https://doi.org/10.1016/j.energy.2021.120069>

Shirdel, M., Asadi, R., Do, D., & Hintlian, M. (2021). *Deep Learning With Kernel Flow Regularization For Time Series Forecasting* (Arxiv:2109.11649). Arxiv. <https://doi.org/10.48550/Arxiv.2109.11649>

Siami-Namini, S., Tavakoli, N., & Siami Namin, A. (2018). A Comparison Of Arima And Lstm In Forecasting Time Series. *2018 17th IEEE International Conference On Machine Learning And Applications (ICMLA)*, 1394–1401. <https://doi.org/10.1109/ICMLA.2018.00227>

Sianturi, T. B., Cholissodin, I., & Yudistira, N. (2023). *Penerapan Algoritma Long Short-Term Memory (Lstm) Berbasis Multi Fungsi Aktivasi Terbobot Dalam Prediksi Harga Ethereum*.

Siregar, A. M., Faisal, S., Cahyana, Y., & Priyatna, B. (2020). Perbandingan Algoritme Klasifikasi Untuk Prediksi Cuaca. *Jurnal Accounting Information System (Aims)*, 3(1), 15–24. <https://doi.org/10.32627/Aims.V3i1.280>

Sivanandam, S. N., & Deepa, S. N. (2007). *Introduction To Genetic Algorithms*. Springer.

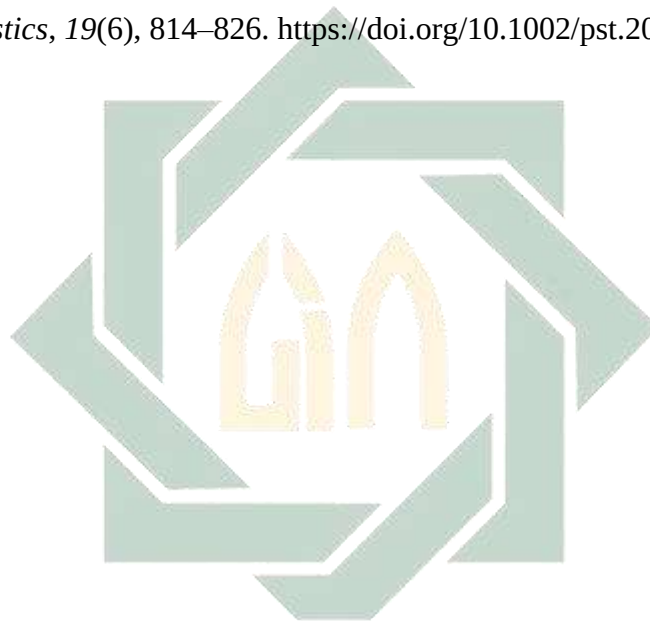
- Sri Rahayu, D. H. N. (2022). *Analisis Peramalan Penjualan Produk Kaos Sablon (Studi Kasus: Home Industry Alva Cloth)*.
<https://doi.org/10.5281/Zenodo.7349799>
- Syarif, I., Prugel-Bennett, A., & Wills, G. (2016). SVM Parameter Optimization using Grid Search and Genetic Algorithm to Improve Classification Performance. *TELKOMNIKA (Telecommunication Computing Electronics and Control)*, 14(4), 1502. <https://doi.org/10.12928/telkomnika.v14i4.3956>
- Talaei Khoei, T., Ould Slimane, H., & Kaabouch, N. (2023). Deep Learning: Systematic Review, Models, Challenges, And Research Directions. *Neural Computing And Applications*, 35(31), 23103–23124.
<https://doi.org/10.1007/S00521-023-08957-4>
- Teixeira, R., Cerveira, A., Pires, E. J. S., & Baptista, J. (2024). Enhancing Weather Forecasting Integrating Lstm And Ga. *Applied Sciences*, 14(13), 5769.
<https://doi.org/10.3390/App14135769>
- Thilakarathne, H., & Premachandra, K. (2017). Predicting Floods In North Central Province Of Sri Lanka Using Machine Learning And Data Mining Methods. *Sri Lanka*.
- Tomar, V., Bansal, M., & Singh, P. (2024). Metaheuristic Algorithms For Optimization: A Brief Review. *Raise-2023*, 238.
<https://doi.org/10.3390/Engproc2023059238>
- Torres, J. F., Hadjout, D., Sebaa, A., Martínez-Álvarez, F., & Troncoso, A. (2021). Deep Learning For Time Series Forecasting: A Survey. *Big Data*, 9(1), 3–21. <https://doi.org/10.1089/Big.2020.0159>
- Uddin, Md. J., Li, Y., Sattar, Md. A., Nasrin, Z. Most., & Lu, C. (2022). Effects Of

- Learning Rates And Optimization Algorithms On Forecasting Accuracy Of Hourly Typhoon Rainfall: Experiments With Convolutional Neural Network. *Earth And Space Science*, 9(3), E2021ea002168. <https://doi.org/10.1029/2021ea002168>
- Ulum, D. S. N., & Girsang, A. S. (2022). *Hyperparameter Optimization Of Long-Short Term Memory Using Symbiotic Organism Search For Stock Prediction. International Journal Of Innovative Research And Scientific Studies*, 5(2), 121–133. <https://doi.org/10.53894/Ijirss.V5i2.415>
- Ünlü, R. (2019). A Comparative Study Of Machine Learning And Deep Learning For Time Series Forecasting: A Case Study Of Choosing The Best Prediction Model For Turkey Electricity Production. *Süleyman Demirel Üniversitesi Fen Bilimleri Enstitüsü Dergisi*, 23(2), 635–646. <https://doi.org/10.19113/Sdufenbed.494396>
- Wasil, M., Harianto, H., Universitas Hamzanwadi, Fathurrahman, F., & Universitas Hamzanwadi. (2022). Pengaruh *Epoch* Pada Akurasi Menggunakan Convolutional Neural Network Untuk Klasifikasi Fashion Dan Furniture. *Infotek : Jurnal Informatika Dan Teknologi*, 5(1), 53–61. <https://doi.org/10.29408/Jit.V5i1.4393>
- Widianti, A., & Pratama, I. (2024). Penanganan Missing Values Dan Prediksi Data Timbunan Sampah Berbasis Machine Learning. *Rabit : Jurnal Teknologi Dan Sistem Informasi Univrab*, 9(2), 242–251. <https://doi.org/10.36341/Rabit.V9i2.4789>
- Wiranda, L., & Sadikin, M. (2019). *Penerapan Long Short Term Memory Pada Data Time Series Untuk Memprediksi Penjualan Produk Pt. Metiska*

Farma. 8.

Xie, F., Yan, H., Long, Y., Guo, H., Liu, H., & Yu, P. (2024). Weather Prediction Based On Multivariate Lstm Neural Network Model. In Z. Hou (Ed.), *Advances In Transdisciplinary Engineering*. Ios Press. <https://doi.org/10.3233/Atde231201>

Ye, J., Reaman, G., De Claro, R. A., & Sridhara, R. (2020). A Bayesian approach in design and analysis of pediatric cancer clinical trials. *Pharmaceutical Statistics*, 19(6), 814–826. <https://doi.org/10.1002/pst.2039>



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