



JURNAL MANTIK

[INSTITUTE OF COMPUTER SCIENCE](#)

P-ISSN : <> E-ISSN : 26854236 Subject Area : Economy, Science, Education



1.10495
Impact Factor



1411
Google Citations



Sinta 4
Current
Accreditation

[Google Scholar](#) [Garuda](#) [Website](#) [Editor URL](#)

History Accreditation

2018 2019 2020 2021 2022 2023 2024

[Garuda](#) [Google Scholar](#)

[The Effectiveness of Giving Red Spinning Juice on Increasing Hb Levels for Pregnant Anemia in the Work East Lahewa Health Center Year 2020](#)

Institute of Computer Science (IOCS) [Jurnal Mantik Vol. 5 No. 4 \(2022\): February: Manajemen, Teknologi Informatika dan Komunikasi \(Mantik\) 2145-2150](#)

2022 DOI: - [Accred : Sinta 4](#)

[Optimization of Virtual Reality-Based Flood Simulation by Manipulating Particle System](#)

Institute of Computer Science (IOCS) [Jurnal Mantik Vol. 5 No. 4 \(2022\): February: Manajemen, Teknologi Informatika dan Komunikasi \(Mantik\) 2117-2126](#)

2022 DOI: - [Accred : Sinta 4](#)

[Implementation of Elo Rating System and Player Clustering for Competitive Matchmaking in Trivia Education Game](#)

Institute of Computer Science (IOCS) [Jurnal Mantik Vol. 5 No. 4 \(2022\): February: Manajemen, Teknologi Informatika dan Komunikasi \(Mantik\)](#)

2022 DOI: - [Accred : Sinta 4](#)

[Database Forensics in Software as A Service Service using Stored Procedure](#)

Institute of Computer Science (IOCS) [Jurnal Mantik Vol. 5 No. 4 \(2022\): February: Manajemen, Teknologi Informatika dan Komunikasi \(Mantik\) 2127-2135](#)

2022 DOI: - [Accred : Sinta 4](#)

[Automation Simulation of Plant Watering System using Moisture Sensor Based on Mobile Devices Urban Farming Method](#)

Institute of Computer Science (IOCS) [Jurnal Mantik Vol. 5 No. 4 \(2022\): February: Manajemen, Teknologi Informatika dan Komunikasi \(Mantik\) 2136-2144](#)

📅 2022 🗨️ DOI: - 🏆 [Accred : Sinta 4](#)

[Disease in Corn Leaf Using Gabor Wavelet and K-Means Clustering Algorithm](#)

Institute of Computer Science (IOCS) 📖 [Jurnal Mantik Vol. 5 No. 4 \(2022\): February: Manajemen, Teknologi Informatika dan Komunikasi \(Mantik\) 2152-2156](#)

📅 2022 🗨️ DOI: - 🏆 [Accred : Sinta 4](#)

[Optimization of Tree Algorithms by Resampling and Ensembling in Deffect Prediction Software:](#)

Institute of Computer Science (IOCS) 📖 [Jurnal Mantik Vol. 5 No. 4 \(2022\): February: Manajemen, Teknologi Informatika dan Komunikasi \(Mantik\) 2177-2181](#)

📅 2022 🗨️ DOI: - 🏆 [Accred : Sinta 4](#)

[Analysis of Service Quality and Product Quality on Customer Purchase Decisions in CV. Surya Indah Jaya Medan with Price as a Variable Intervention](#)

Institute of Computer Science (IOCS) 📖 [Jurnal Mantik Vol. 5 No. 4 \(2022\): February: Manajemen, Teknologi Informatika dan Komunikasi \(Mantik\) 2156-2164](#)

📅 2022 🗨️ DOI: - 🏆 [Accred : Sinta 4](#)

[Analysis of Factors Affecting the Work Discipline of Employees CV. Golden Seafresh Medan](#)

Institute of Computer Science (IOCS) 📖 [Jurnal Mantik Vol. 5 No. 4 \(2022\): February: Manajemen, Teknologi Informatika dan Komunikasi \(Mantik\) 2165-2170](#)

📅 2022 🗨️ DOI: - 🏆 [Accred : Sinta 4](#)

[The Effect of Pricing and Product Quality on Purchase Decisions on PT. Gunung Sentosa Sumatera Utama Industri](#)

Institute of Computer Science (IOCS) 📖 [Jurnal Mantik Vol. 5 No. 4 \(2022\): February: Manajemen, Teknologi Informatika dan Komunikasi \(Mantik\) 2171-2176](#)

📅 2022 🗨️ DOI: - 🏆 [Accred : Sinta 4](#)

[View more ...](#)



1819 / Pahrul Rodji et al. / Development of Draina...

Library

Submissions

Workflow Publication

Submission Review Copyediting

Production

Submission Files Search

▶	5197-1	alikhumaidi,	November	Article
	Mantik Pahrul.doc		25, 2021	Text

[Download All Files](#)

Pre-Review Discussions Add discussion

Name	From	Last Reply	Replies	Closed
<i>No Items</i>				



1819 / Pahrul Rodji et al. / Development of Draina...

Library

Submissions

Workflow | **Publication**

Submission | **Review** | **Copyediting**

Production

Round 1

Round 1 Status
Submission accepted.

Notifications

[Jurnal Mantik] Editor Decisi...	2021-11-26 10:13 PM
[Jurnal Mantik] Editor Decisi...	2021-11-26 10:13 PM

Reviewer's Attachments [Search](#)

No Files

Revisions [Search](#) [Upload File](#)

No Files

Name	From	Last Reply	Replies	Closed
<i>No Items</i>				

Platform &
workflow by
OJS / PKP



[Jurnal Mantik] Editor Decision

2021-11-26 10:13 PM

Ali Khumaidi, Ahmad Pahrul Rodji1:

We have reached a decision regarding your submission to Jurnal Mantik, "DEVELOPMENT OF DRAINAGE STATUS PREDICTION MODEL BASED ON INTERNET OF THINGS AND LONG SHORT TERM MEMORY ALGORITHM".

Our decision is to: Accept Submission

[Jurnal Mantik](#)



[Jurnal Mantik] Editor Decision

2021-11-26 10:13 PM

Ali Khumaidi, Ahmad Pahrul Rodji1:

We have reached a decision regarding your submission to Jurnal Mantik, "DEVELOPMENT OF DRAINAGE STATUS PREDICTION MODEL BASED ON INTERNET OF THINGS AND LONG SHORT TERM MEMORY ALGORITHM".

Our decision is to: Accept Submission

[Jurnal Mantik](#)



[Jurnal Mantik] Editor Decision

2021-11-26 10:13 PM

Ali Khumaidi, Ahmad Pahrul Rodji1:

The editing of your submission, "DEVELOPMENT OF DRAINAGE STATUS PREDICTION MODEL BASED ON INTERNET OF THINGS AND LONG SHORT TERM MEMORY ALGORITHM," is complete. We are now sending it to production.

Submission URL:

<https://iocscience.org/ejournal/index.php/mantik/authorDashboard/submission/1819>

[Jurnal Mantik](#)



[Jurnal Mantik] Editor Decision

2021-11-26 10:13 PM

Ali Khumaidi, Ahmad Pahrul Rodji1:

The editing of your submission, "DEVELOPMENT OF DRAINAGE STATUS PREDICTION MODEL BASED ON INTERNET OF THINGS AND LONG SHORT TERM MEMORY ALGORITHM," is complete. We are now sending it to production.

Submission URL:

<https://iocscience.org/ejournal/index.php/mantik/authorDashboard/submission/1819>

[Jurnal Mantik](#)

[Jurnal Mantik] Submission Acknowledgement

1 message

Hengki Tamando Sihotang <editor.mantik@iocscience.org>
To: Ali Khumaidi <alikhumaidi@unkris.ac.id>

Thu, Nov 25, 2021 at 9:30 PM

Ali Khumaidi:

Thank you for submitting the manuscript, "DEVELOPMENT OF DRAINAGE STATUS PREDICTION MODEL BASED ON INTERNET OF THINGS AND LONG SHORT TERM MEMORY ALGORITHM" to Jurnal Mantik. With the online journal management system that we are using, you will be able to track its progress through the editorial process by logging in to the journal web site:

Submission URL: <https://iocscience.org/ejournal/index.php/mantik/authorDashboard/submission/1819>

Username: alikhumaidi

If you have any questions, please contact me. Thank you for considering this journal as a venue for your work.

Hengki Tamando Sihotang

[Jurnal Mantik](#)



Development of Drainage Status Prediction Model Based on Internet of Things and Long Short Term Memory Algorithm

Ahmad Pahrul Rodji¹, Wargijono Utomo², Ali Khumaidi³, Hudzaifah Al Jihad⁴

¹Civil Engineering, Faculty of Engineering, Krisnadwipayana University, Indonesia

^{2,3,4}Informatics, Faculty of Engineering, Krisnadwipayana University, Indonesia

E-mail: pahrul_rodji@yahoo.com¹, alikhumaidi@unkris.ac.id³, al.ghifari22@gmail.com⁴

ARTICLE INFO

ABSTRACT

Article history:

Received: September 11, 2021

Revised: October 18, 2021

Accepted: November 24, 2021

Keywords:

Drainage
Hydrology
Internet of Things
Prediction Model
RNN LSTM

The capacity of drainage can overflow due to inadequate conditions and high rainfall intensity. Several incidents in Bekasi City due to poor drainage resulted in inundation of water on the roads which resulted in damaged roads and flooding in residential areas. Several previous studies have discussed the evaluation of the drainage system using the analytical method hydrology in modeling water discharge. In most cases, the minimum capacity of the drainage canal is caused by the high intensity of rain, so the research focuses on the volume of drainage and the intensity of the rain. However, based on observations and interviews with the cleaning service, it turns out that many drainage channels are in a non-optimal condition, where there is a lot of garbage and sedimentation that hinders the flow of water when it rains. This study combines hydrological analysis modeling with drainage channel conditions whose real time data is obtained by using sensors through the internet of things (IoT). IoT devices have been able to send data well in the cloud, by combining rainfall data and then predictive modeling using RNN LSTM with training model parameters used are two layers and 20 cells with each layer given a Dropout layer with a probability of 10%. In the metric evaluation, four functions are used, namely mean squared error, Mean absolute, Nash-Sutcliffe Efficiency and Coefficient of Determination. The model has been able to see the occurrence of an increase or decrease in height and discharge. However, if you look at the results of metric calculations, the predictions generated by the model are not very good.

Copyright © 2021 Jurnal Mantik.
All rights reserved.

1. Introduction

Rapid population growth has resulted in urban areas experiencing very significant development, causing the need for residential land to increase. This shift in land use function from open areas to residential areas causes rainwater catchment areas in residential areas to decrease, causing puddles and flooding [1]. Based on data from the Bekasi medium-term development plan (RPJMD) for 2018-2023, the population of Bekasi this year has reached 3.2 million, bringing the population density to 16,500 per kilometer out of a total area of 210 thousand kilometers. The National Development Planning Agency (BAPPENAS) said that Bekasi City is a metropolitan city with the highest population in Indonesia after DKI Jakarta and Surabaya City [2].

Every year there are cases of flooding on roads and flooding in residential areas. On January 21, 2020 there was a flood as high as 80cm in residential areas due to poor drainage [3]. This is because the water from the culverts comes from three directions, but the drain is narrow and full of mud. Based on field observations and interviews with officers at the Bekasi City cleaning service, every rainy season the streets are often flooded with water. Inundation points usually occur at the fork in the road and inundation occurs because the rainwater catchment area is not balanced with the rapid settlement area. In addition, drainage channels along the road are inadequate and cannot function properly to accommodate rainwater. So that puddles are inevitable. Therefore we need a means to drain rainwater, one of which is a drainage system.



The drainage system is an important part in planning the development of an urban area. Judging from the decrease in existing land, the drainage system is one of the efforts to overcome and reduce the problem of waterlogging in residential areas. Generally, residential areas are unable to accommodate the runoff discharge because the drainage channels cannot function properly. Some channels that are not able to accommodate the runoff discharge optimally need to be repaired and some changes to the dimensions and shape of the channel [4]. Meanwhile, according to Evid Zulhaqi (2013), the channel has decreased storage capacity due to the absence of regular cleaning of sedimentation, garbage and plants in the drainage channel, resulting in obstacles for water to the outlet channel [5].

The development of sensor technology and the internet of things has been able to monitor and sending data in real time [6][7]. Hydrological analysis modeling has been quite good in measuring channel discharge and capacity. Research related to the prediction of water flow and rain discharge using a neural network produces a fairly good accuracy [8][9]. This study focuses on evaluating waterlogging based on the existing condition of drainage channels and the factors that influence the occurrence of inundation using hydrological modeling associated with monitoring data for internet of things devices. The combination of the two is expected to provide more updated information in creating an early warning system for preventing puddles and flooding.

2. Method

Figure 1 describes the stages of research in building an early warning system for drainage. The study began with observing the existence of the existing canal by checking the condition of the research location, in this case looking at the problems (inundation/flood) that occurred at the research location. The existing inundation problem is expected to be overcome by structuring the channel system in the form of changing the channel system, increasing the number of channels or adding channel capacity. The arrangement of the canal system has resulted in a new drainage network system that will be used as a benchmark for further analysis. Hydrological analysis was carried out to determine the value of the design discharge and capacity discharge. Hydrological analysis includes analysis of rainfall data and analysis of planned discharge. Rainfall data analysis aims to determine the rainfall data that will be used for calculations. The steps that must be taken are to identify rain gauge stations both within the research location and around the research location, then collect data from all selected stations and select the data to be analyzed. To find statistical data that deviates from the data set, an analysis of the quality of the data is carried out in the form of outlier analysis. Looking at the condition of the research location, the method used to calculate the average rainfall of the area was determined.

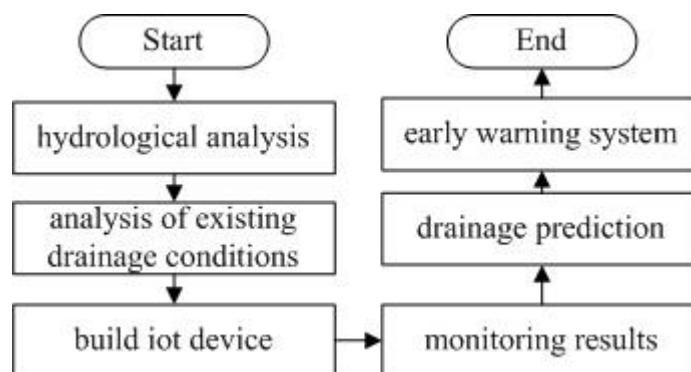


Fig. 1. Research Stages

The next stage is to build an internet of things device that requires system architecture design and sensor calibration. The device is built using an ultrasonic sensor HCSR-04, Humidity and Rain Drop Sensor, NodeMCU Lua ESP8266 and Powerbank 10000mah. The results of the hydrological analysis will be combined with data on the condition of the drainage channel and then predictive modeling will be made using the RNN LSTM algorithm so that an early warning system can be produced to prevent waterlogging and flooding.

3. Result and Analysis

The development of an early warning system by making prediction models and recommendations for action on drainage channels begins with making Internet of Things (IoT) devices with the architecture and devices in Figure 2. Figure 3 describes the network architecture starting from the IoT devices that are built to capture water level data, humidity and raindrops and sends it to the cloud server. Rainfall data in the related area and data from IoT devices will be processed for predictive modeling. The results of the prediction modeling are then made an early warning system to notify the condition of the drainage system in preparation for the rainy season.

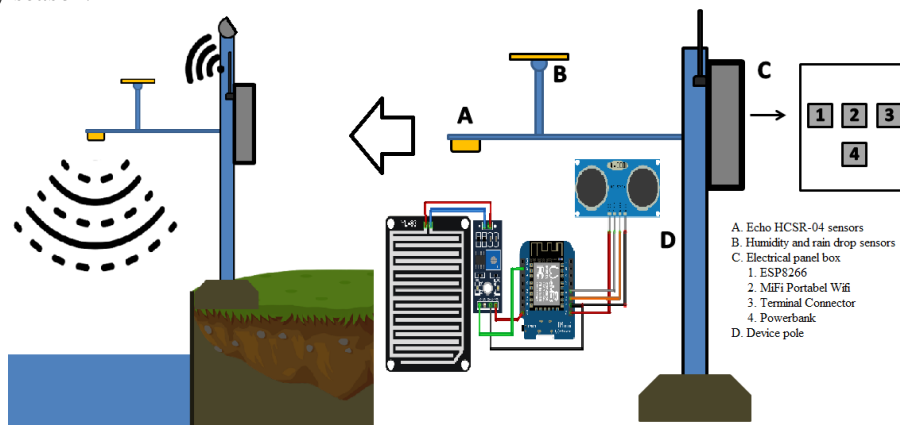


Fig 2. Internet of Things Device Drainage System

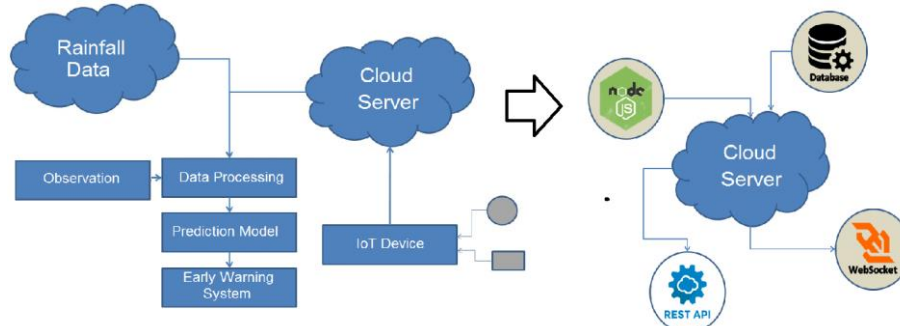


Fig 3. Drainage System Cloud Architecture

Prediction model development using Jupyter Notebook using the Recurrent Neural Networks (RNN) algorithm used is Long Short-Term Memory (LSTM), the RNN LSTM architecture model can be seen in Figure 4. LSTM is one of the developments of neural networks that are capable of storing long time series information, able to overcome the problem of gradient descent when processing long sequence data in conventional RNN [10][11]. The data in the cloud is then processed by checking the invalid data, data missing and then make corrections and adjustments to the completeness of the data.

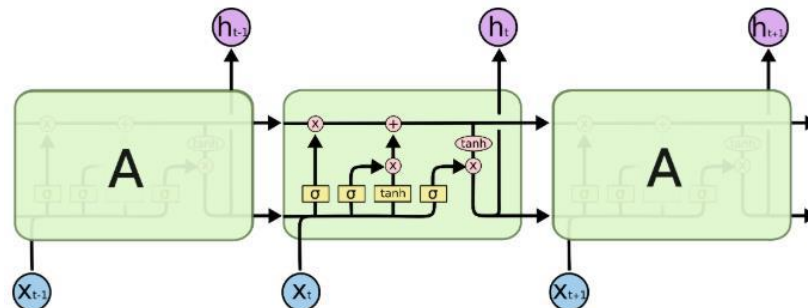


Fig 4. RNN LSTM Architecture

Next, divide the dataset into two parts, namely the train set and the test set. Then normalize the values on the dataset. In the training model the parameters used are two layers and 20 cells with each layer given a Dropout layer with a probability of 10% [12].

In the metric evaluation, four functions are used, namely the mean squared error (MSE) which is used as a loss function loss: 0 MSE , the smaller the better. Mean absolute error (MAE): 0 MAE , the smaller the better. Nash-Sutcliffe Efficiency (NSE): $NSE < 1$, the bigger the better. Coefficient of Determination (R2): 0 R2 1, with a value of 1 indicating perfectly correlated data (prediction data is exactly the same as the actual data) [13]. Since we want to evaluate the metrics for every epoch, a special function was created to be included when compiling the model[14].

TABLE 1
COMPARISON OF TRAIN SET AND TEST SET METRICS

Metrics	Train	Test
Mse	0.408920	0.836735
Mae	0.451762	0.585794
Nse	0.621280	0.339275
R2	0.697100	0.384172

The results of the flow discharge prediction using the test data set are still not good. It can be seen that there are some values that miss the observation data, but the model is able to predict the conditions of altitude and rain that occur in a certain month. Based on Table 1, the metric results show that the predictions from the test data set are not very good, so the model is arguably still under-trained.

4. Conclusion

The development of IoT devices has been successfully built and is able to send data to the cloud server well. In predictive modeling so far the model is still unsatisfactory. The model has been able to see the occurrence of an increase or decrease in height and discharge. However, if you look at the results of metric calculations, the predictions generated by the model are not good, none of the parameters are considered satisfactory.

References

- [1] M. Shao, G. Zhao, S.-C. Kao, L. Cuo, C. Rankin, and H. Gao, “Quantifying the effects of urbanization on floods in a changing environment to promote water security — A case study of two adjacent basins in Texas,” *J. Hydrol.*, vol. 589, p. 125154, Oct. 2020, doi: 10.1016/j.jhydrol.2020.125154.
- [2] A. S. Pambudi and S. Hidayati, “Analisis Perilaku Sosial Pengguna Moda Transportasi Perkotaan: Studi Kasus Mass Rapid Transit (MRT) DKI Jakarta,” *Bappenas Work. Pap.*, vol. 3, no. 2, pp. 143–156, Sep. 2020, doi: 10.47266/bwp.v3i2.74.
- [3] M. Azzam, “Drainase Buruk, Kampung Buaran Bekasi Sempat Kebanjiran 80 Centimeter Artikel ini telah tayang di WartaKotalive.com dengan judul Drainase Buruk, Kampung Buaran Bekasi Sempat Kebanjiran 80 Centimeter, <https://wartakota.tribunnews.com/2020/01/21/drainase-bu>,” *wartakota*, 21-Jan-2020.
- [4] F. Auzan, “Evaluasi Genangan dan Banjir Jalan PB. Sudirman – Jalan Moch. Serudji Kabupaten Jember,” *Universitas Jember*, 2014.
- [5] Nugraheni, M. Ayu, and Suteki, “Revitalisasi Sungai Oleh Pemerintah Daerahmelalui Kebijakan Normalisasi Sungai Pepe Kota Surakarta (Studi Kasus Normalisasi Sungai Dalam Rangka Peningkatan Kesejahteraan Masyarakat Di Bantaran Sungai),” *Diponegoro University*, 2018.
- [6] B. Harsono, “Sistem Hidroponik Berbasis Internet of Things,” *Dielektrika*, vol. 7, no. 2, p. 82, Aug. 2020, doi: 10.29303/dielektrika.v7i2.240.
- [7] A. Khumaidi, “Prototipe Alat Pengusir Burung Pada Gedung Berbasis Internet of Things Menggunakan Sensor RCWL,” *Ilk. J. Ilm.*, vol. 12, no. 2, pp. 162–167, Aug. 2020, doi: 10.33096/ilkom.v12i2.602.162-167.
- [8] T. S. Megariansyah, “Prediksi Debit Aliran menggunakan Long Short-Term Memory (LSTM),” *Zenodo*, 2019, doi: 10.5281/zenodo.3911043.
- [9] Z. Fang, Y. Wang, L. Peng, and H. Hong, “Predicting flood susceptibility using LSTM neural networks,” *J. Hydrol.*, vol. 594, p. 125734, Mar. 2021, doi: 10.1016/j.jhydrol.2020.125734.



- [10] A. Sherstinsky, “Fundamentals of Recurrent Neural Network (RNN) and Long Short-Term Memory (LSTM) network,” *Phys. D Nonlinear Phenom.*, vol. 404, p. 132306, Mar. 2020, doi: 10.1016/j.physd.2019.132306.
- [11] A. Khumaidi, R. Raafi’udin, and I. P. Solihin, “Pengujian Algoritma Long Short Term Memory untuk Prediksi Kualitas Udara dan Suhu Kota Bandung,” *Telematika*, vol. 15, no. 1, pp. 13–18, 2020.
- [12] F. Kratzert, D. Klotz, C. Brenner, K. Schulz, and M. Herrnegger, “Rainfall–runoff modelling using Long Short-Term Memory (LSTM) networks,” *Hydrol. Earth Syst. Sci.*, vol. 22, no. 11, pp. 6005–6022, Nov. 2018, doi: 10.5194/hess-22-6005-2018.
- [13] R. A. Sowah, K. Bradshaw, B. Snyder, D. Spidle, and M. Molina, “Evaluation of the soil and water assessment tool (SWAT) for simulating E. coli concentrations at the watershed-scale,” *Sci. Total Environ.*, vol. 746, p. 140669, Dec. 2020, doi: 10.1016/j.scitotenv.2020.140669.
- [14] J. B. Pedersen, A. Brodsky, and J. Sampson, “Approximating the buffer allocation problem using epochs,” *J. Parallel Distrib. Comput.*, vol. 68, no. 9, pp. 1263–1282, Sep. 2008, doi: 10.1016/j.jpdc.2008.06.001.