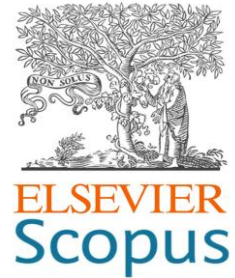




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## Modification of Snyder Synthetic Unit Hydrograph Model

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**Abstract:** This study intends to modify the Snyder Synthetic Unit Hydrograph (SUH) model. There are two physical parameters,  $C_t$  and  $C_p$  in Snyder SUH, which are coefficients that depend on watershed characteristics. In the original model, the values of  $C_t$  and  $C_p$  must be empirically determined according to the watershed condition. However, this research was conducted in 10 watersheds in Indonesia, which must have an Automatic Rainfall Recorder (ARR) and an Automatic Water Level Recorder (AWLR). The methodology consists of 1) finding the relation between input variables for determining the multi-co-linearity between the variables; 2) to find the relation between input and output variables; 3) to analyze by using multiple linear regression by using Partial Least Square-Structural Equation Model (PLS-SEM) and Artificial Neural Network (ANN); 4) to evaluate the regression value for determining whether the regression equation has good performance in presenting the input variables; 5) to obtain the model with a new constant in accordance with the watershed condition in the research location; and 6) to carry out model validation and verification. The modification result of the Snyder SUH model is as follows:  $T_p = C_t \cdot L_c^{0.325}$  ( $R^2 = 89.3\%$ );  $C_t = 0.680 (W_r^{0.153} / S^{0.143})$ ;  $Q_p = C_p \cdot (1 / T_p^{2.969})$  ( $R^2 = 91.6\%$ );  $C_p = 0.537 K^{1.624}$ ;  $T_b = 3.919$



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$$A^{0.569} / Q_p^{0.562} (R^2 = 72.7\%)$$

**Keywords:** Snyder synthetic unit hydrograph; regression; partial least square-structural equation model; artificial neural network

## 斯奈德合成单位线模型的修正

**摘要:** 本研究旨在修改Snyder合成单位过程线(SUH)模型。Snyder SUH中有两个物理参数 $C_t$ 和 $C_p$ , 它们是取决于流域特征的系数。在原始模型中,  $C_t$ 和 $C_p$ 的值必须根据分水岭条件凭经验确定。然而, 这项研究是在印度尼西亚的10个流域进行的, 这些流域必须有自动降雨记录仪 (ARR) 和自动水位记录仪 (AWLR)。该方法包括: 1) 寻找输入变量之间的关系, 以确定变量之间的多重共线性; 2) 找出输入变量和输出变量之间的关系; 3) 利用偏最小二乘结构方程模型 (PLS-SEM) 和人工神经网络 (ANN) 进行多元线性回归分析; 4) 评估回归值, 以确定回归方程是否具有好的表现输入变量的性能; 5) 根据研究地点的流域条件, 得到具有新常数的模型; 6) 进行模型验证和验证。Snyder SUH模型的修正结果如下:  $T_p=C_t \cdot L_c \cdot 0.325$  ( $R^2=89.3\%$ );  $C_t=0.680$  ( $W_r=0.153/S_0.143$ );  $Q_p = C_p \cdot (1 / T_p \cdot 2.969)$  ( $R^2 = 91.6\%$ );  $C_p=0.537K^{1.624}$ ;  $T_b = 3.919 A^{0.569} / Q_p^{0.562}$  ( $R^2 = 72.7\%$ )

**关键词:** 斯奈德合成单位线; 回归; 偏最小二乘结构方程模型; 人工神经网络

### 1. Introduction

Hydrological analysis is a transformation process of rainfall to discharge [1] in a watershed until the outlet of the structure plan. However, the design flood information is very important in relation to the design of the structure dimension [2], which influences the structure performance, development cost, and loss level caused by structures that experience failure due to design limits. Basically, this problem can be solved if there is available discharge data with good quality in a watershed [2-3], in determining the design flood, it is not necessary to use a model [4] to carry out the transformation analysis of rainfall to discharge. However, it is often found in the location of water structure design where discharge data are only available in a short time/period, the data quality is not too good, and there is no available data [4].

In the development of the synthetic unit hydrograph method [5-6], some empirical equations have been used to express the relationship between the characteristics of watershed morphometric [7-8] and the uncertainty of the hydrograph former, that is, peak discharge ( $Q_p$ ), time to peak ( $T_p$ ), and time base ( $T_b$ ). The characteristic of watershed morphometry that is almost different in every region [9] causes a very accurate Synthetic Unit Hydrograph (SUH) model to be used if the SUH model is applied in the watershed where it is developed. However, if the SUH model is applied to other watersheds, there will be more deviation. This deviation produces inaccurate information about the design flood, which can be over- or underestimated. If a structure is designed by over-estimation, it will cause the dimension and development cost to be inefficient,

and if a structure is designed by under-estimation, it will cause the failure risk of high structure.

In Snyder SUH, there are two non-physical parameters,  $C_t$  and  $C_p$ , which are dependent on the unit and characteristics of the watershed [10-11]. The  $C_t$  and  $C_p$  values were obtained from several watersheds in the Appalachian Highland (USA). However, the coefficients of  $C_t$  and  $C_p$  must be empirically determined because the values change between regions. Therefore, it is difficult to use this method for Indonesian rivers [12-13]. Therefore, it can be accepted in research locations in Indonesia, and the Snyder SUH model must be customized with local characteristics and conditions.

According to [14], in watersheds in East Java, Indonesia, the Snyder SUH produced an accuracy of  $T_p$  of 96.90%,  $Q_p$  of 82.72%, and  $T_b$  of 85.44%. Until now, the other islands in Indonesia have not had an SUH model that is used for the design and management of water resources, which is developed based on the characteristics of hydrology and watershed. Therefore, it is necessary to adjust the parameters to increase the performance of the Snyder SUH model for use in Indonesia. Based on the reasons above, it is necessary to carry out research related to the coefficients of  $C_t$  and  $C_p$  in watersheds in Indonesia, as well as  $T_p$ ,  $T_b$ , and  $Q_p$ ; therefore, it can be applied to other watersheds outside of the research location.

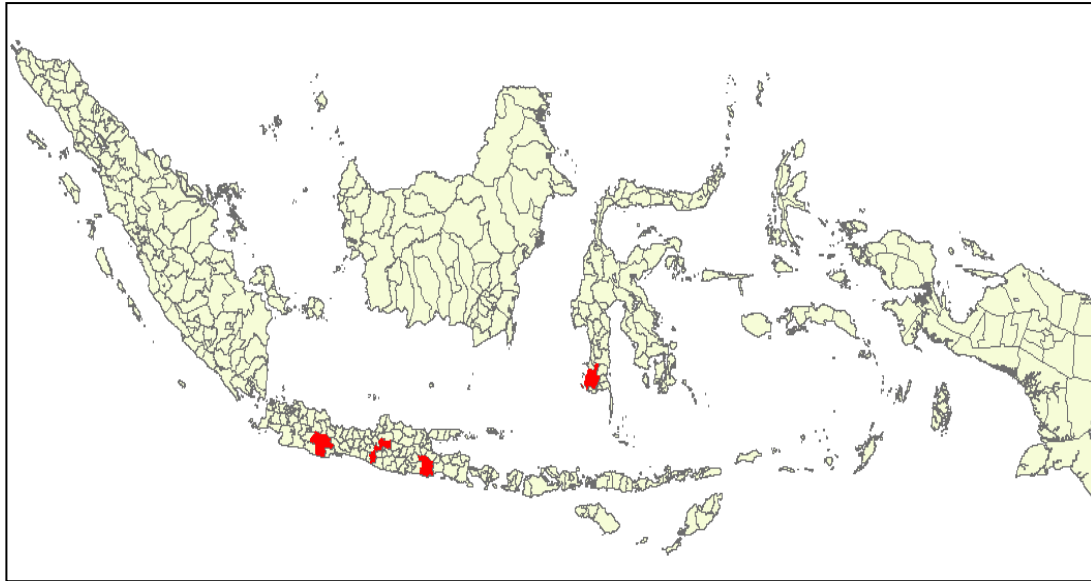
## 2. Materials and Methods

### 2.1. Research Location

The research locations are some watersheds in

Indonesia that are completed with an Automatic Water Level Recorder (AWLR) and Automatic Rainfall

Recorder (ARR) with complete and long enough data recording, as presented in Figure 1 and Table 1.



**Figure 1. Watershed Map of Research Location**  
Source: Riman et.al [15]

**2.2. Snyder SUH (Synthetic Unit Hydrograph) [15]**

Snyder has developed a formulation using empirical coefficients that relate the uncertainty of the unit hydrograph with the specific characteristics of the watershed and to analyze the amount of watershed hydrographs in the Appalachian Mountain Region (USA) by the watershed area between 25 km<sup>2</sup> and 25,000 km<sup>2</sup>.

The unit hydrograph is sufficient to determine the rainfall depth (d) = 1 cm, and the three other uncertain values are Q<sub>p</sub> (m<sup>3</sup>/s), T<sub>b</sub> (hours), and T<sub>r</sub> (hours). The uncertainty of the hydrograph is related to A

(watershed area, km<sup>2</sup>), L (length of the main river, km), and L<sub>c</sub> (the length of the river until the point nearest to the weight point of the catchment area, km).

Based on this uncertainty, Snyder built the following formulations [1-20]:

$$t_p = C_t (L_c \times L)^{0.3} \tag{1}$$

$$t_r = \frac{t_p}{5.5} \tag{2}$$

$$Q_p = 2.78 \frac{C_p \times A}{t_p} \tag{3}$$

$$T_b = \frac{72 + 3T_p}{24} \tag{4}$$

**Table 1 Location of Research Watershed, AWLR, and ARR (Source: Riman et.al [15])**

No.	Research Location	Watershed	AWLR	ARR
1	Jawa Barat	Pataruman, WS Citanduy	AWLR Pataruman	St. Cibereum, St. Pataruman, St. Cihonje, St. Kawali, St. Cineam, and St. Pagerageung
2	Jawa Timur	Lesti, WS Brantas	AWLR Tawang Rejeni	St. Poncokusumo, St. Dampit
3	Jawa Timur	Gadang WS Brantas	AWLR Gadang	St. Jabung, and St. Malang
4	Jawa Timur	Dungus, WS B.Solo	AWLR Dungus	St. Ahmad Yani, St. Jatirogo
5	Jawa Tengah	Girimargo, WS Pemali Juana	AWLR Kunthi	St. Kunthi
6	Jawa Tengah	Jurug WS B.Solo	AWLR Jurug	St. Klaten, St. Jatipuro, and St. Gunungan
7	Daerah Istimewa Yogyakarta	Code, WS Serayu Opak	AWLR Kaloran	St. Prumpung, St. Kempud, and St. Angin-angin
8	Sulawesi Selatan	Sombala, WS Pompengan Jeneberang	AWLR Sombala	St. Janelata,

No.	Research Location	Watershed	AWLR	ARR
				St. Bonto Jai, and St. Jonggoa
9	Sulawesi Selatan	Kampili WS Pompengan Jeneberang	AWLR Kampili	St. Janelata, St. Bonto Jai, and St. Jonggoa
10	Sulawesi Selatan	Jonggoa WS Pompengan Jeneberang	AWLR Jonggoa	St. Jonggoa, and St. Daraha

However, the coefficients  $C_t$  and  $C_p$  must be empirically determined because the dimensions are changed between one area and another.  $C_t$  is a coefficient that is dependent on the slope-basin (0.75 – 3.00), and can be used in the approach of Taylor and Schawarz as follows:

$$C_t = \frac{0.6}{\sqrt{S}} \tag{5}$$

$C_p$  is a coefficient that is dependent on watershed characteristic (0.40 – 0.80).

The Snyder formulation, which has been changed, is used more frequently. The change is in the exponential of 0.30 is changed with n as follows:

$$t_p = C_t (L_c \times L)^n \tag{6}$$

$t_r$  in formulation-2 is changed to  $t_e$ , which is the duration of the effective rainfall; however,  $t_r = 1$  h. The duration of effective rainfall ( $t_e$ ) is influenced by the time lag; this relationship can be seen in the formulation below:

$$t_e = \frac{t_p}{5.5} \tag{7}$$

The relation between  $t_e$ ,  $t_p$ ,  $t_r$ , and  $T_p$  is as follows:

If  $t_e > t_r$

$$t'_p = t_p + 0.25 (t_r - t_e) \tag{8}$$

Therefore, there is obtained time to peak as follows:

$$T_p = t'_p + 0.50t_r \tag{9}$$

$$\text{If } T_p = t_p + 0.50t_r \tag{10}$$

$$\text{Thus, } q_p = 0.278 \frac{C_p}{t_p} \tag{11}$$

where:

$t_e$  : duration of effective rainfall (hour)

$t_r$ : duration of 1-hour effective rainfall (hour)

$t_p$ : time lag (hour)

$T_p$ : time to peak (hour)

$n$  : coefficient of proportional (0.10 – 0.38)

$Q_p$ : peak discharge ( $\text{m}^3/\text{s}/\text{mm}$ )

$Q$ : unit hydrograph peak ( $\text{m}^3/\text{s}/\text{mm}/\text{km}^2$ )

$S$  : average slope of watershed

$H$ : rainfall depth= 1 mm

The other functions that are used is as follows:

$$T_p = C_t \left\{ \frac{L \times L_c}{S} \right\}^n \tag{13}$$

$$T_b = 5 + (T_p + t_r/2) \tag{14}$$

Snyder built an empirical formula for the peak discharge ( $Q_p$ ) and time to peak ( $T_p$ ) of a hydrograph. To obtain a hydrograph curve, time was required to calibrate the parameters. To accelerate the process, the Alexeyev formulation is used, which provides the unit hydrograph form. The form of the unit hydrograph determined using the Alexeyev formulation is as follows:

$$Q = f(t) \tag{15}$$

$$Y = \frac{Q}{Q_p} \tag{16}$$

$$x = \frac{t}{T_p} \tag{17}$$

$$Y = 10^{-a \frac{(1-x)^2}{x}} \tag{18}$$

$$a = 1.32 \lambda^2 + 0.15 \lambda + 0.045 \tag{19}$$

$$\lambda = \frac{Q_p T_p}{h.A} \tag{20}$$

Figure 2 shows the Synthetic Unit Hydrograph of Snyder with Alexeyev formulation.

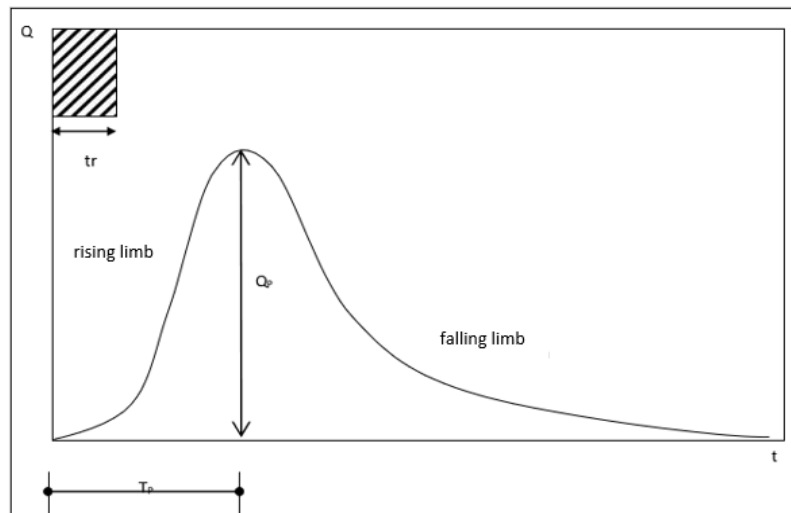


Figure 2. Synthetic Unit Hydrograph of Snyder with Alexeyev Formulation [8]

Source: own study

## 2.2. Methodology

The modification of the Snyder SUH model is based on the independent variable input, which is determined based on the characteristics of the watershed morphometric and dependent variable output, which is determined based on the average unit hydrograph. The steps for building the model are as follows.

(1) Correlation analysis was used to find the relationship between input variables for determining the multicollinearity between variables. If the correlation between variables is high, it means that both variables influence each other; therefore, in determining the regression variable, one of both variables is sufficient.

(2) To determine the relationship between each input variable and each output variable, an input variable with a high correlation can be used as the input variable of the model.

(3) Analysis of multiple linear regressions using the partial least squares structural equation modeling (PLS-SEM) method was carried out to modify the SUH model. This analysis is carried out to show the relationship between two or more variables; in this analysis, there are two variables: independent variables (VB) and dependent variables (VTB). However, the independent variable is variable that has been known as the value (watershed parameter), and the dependent variable is that will be found the value.

(4) The regression test was used to determine whether the regression equation performed well in presenting input variables. This test is carried out to evaluate the best form of the model; it indicates whether the model has shown good performance, and it is seen from the values of the correlation coefficient, determination coefficient, and variance analysis. The model performance can also be analyzed based on the

residual test, which is a normally distributed residual, non-auto-correlation residual to each other, and homogeneous residual variance. The arrangement and evaluation of the model was carried out using the graded residual gradient (GRG) method with an Artificial Neural Network (ANN).

(5) Based on the regression analysis using PLS-SEM, the model was obtained with a new constant (parameter) in accordance with the watershed condition in the research location, and it is expressed by the formula.

(6) Validation and verification of the model to determine the model performance are then carried out to validate and verify the modification of the Snyder SUH model that has been built using ANN-based GRG. Validation is intended to evaluate the enforceability of the model using other data. Validation was carried out by measuring the model performance using rainfall and discharge data in other periods (years) that were not used in the model building. Verification was then carried out using data from other watersheds on the islands of Indonesia that were not used for building the model. The evaluation results of the validation and verification of the model can be observed from the error indicator of the parameter.

## 3. Results and Discussion

### 3.1. Time to peak ( $T_p$ )

The time to peak is explained by  $L_c$ ,  $W_r$ , and  $S$  by the nonlinear equation as follows:

$$T_p = b_0 \cdot L_c^{b_1} \cdot W_r^{b_2} \cdot S^{b_3}$$

Table 2 presents the test results of the regression coefficient for the time to peak, and Table 3 presents the coefficient of ANN for the time to peak.

**Table 2. Test Result of regression coefficient for time to peak (Source: own study)**

Variable	Parameter	Estimate	Std. Error	t	p
Contant	$b_0$	0.680	0.162	4.198	0.001
$L_c$	$b_1$	0.325	0.068	4.779	0.000
$W_r$	$b_2$	0.153	0.049	3.122	0.008
$S$	$b_3$	-0.143	0.051	-2.804	0.014

The regression equation for time to peak is as follows:

$$T_p = 0.680 L_c^{0.325} W_r^{0.153} S^{-0.143}; R^2 = 0.893$$

or

$$T_p = 0.680 (W_r^{0.153} / S^{0.143}) \cdot L_c^{0.325}$$

$$= Ct \cdot L_c^{0.325}$$

$$Ct = 0.680 (W_r^{0.153} / S^{0.143})$$

**Table 3. ANN coefficient for time to peak (Source: own study)**

Predictor	Predicted								
	Hidden Layer <sup>a</sup>							Output Layer	
	H(1)	H(2)	H(3)	H(4)	H(5)	H(6)	H(7)	$T_p$	
Input Layer	$L_c$	20.146	10.268	33.412	21.552	33.459	28.935	10.909	
	$W_r$	18.466	22.999	24.622	32.031	5.177	43.090	6.595	
	$S$	0.023	0.048	0.016	0.020	0.019	0.009	0.045	
Hidden Unit Width		3.550	2.736	2.736	2.736	2.736	4.411	4.151	

Continuation of Table 3		
	H(1)	5.102
	H(2)	3.861
	H(3)	5.976
Hidden Layer	H(4)	5.402
	H(5)	5.190
	H(6)	7.402
	H(7)	2.998

a. Displays the center vector for each hidden unit.

The validation of the regression equation result for the time to peak ( $T_p$ ) had NSE = 89.2% and MAPE = 9.7%. This means that the predicted value is a very good category. The comparison with the ANN result has the best model with a hidden layer of 7, and the prediction of time to peak has higher validation than the nonlinear model. The validation of the ANN result had an NSE of 92.0% and MAPE of 6.4%.

The verification of the prediction result for  $T_p$  for the four datasets of the watershed produces a regression equation for time to peak ( $T_p$ ) with NSE = 69.5% and MAPE = 5.2%, which explains why the prediction

value of the verification data is a very good category. The verification using the ANN result had an NSE of 72.0% and MAPE of 5.0%, indicating that the prediction value of the verification data is a very good category.

### 3.2. Peak Discharge ( $Q_p$ )

The peak discharge is explained by  $L_c$ ,  $W_r$ , and  $S$  using the following nonlinear equation:

$$Q_p = b_0 \cdot T_p^{b_1} \cdot K^{b_2}$$

Table 4 presents the test results of the regression coefficients for the peak discharge.

**Table 4. Test result of regression coefficient for peak discharge (Source: own study)**

Variable	Parameter	Estimate	Std. Error	t	p
Constant	$b_0$	0.537	0.281	1.911	0.069
$T_p$	$b_1$	-2.969	0.988	-3.005	0.009
K	$b_2$	1.624	0.325	4.997	0.000

The regression equation for peak discharge is as follows:

$$Q_p = 0.537 T_p^{-2.969} K^{1.624}; R^2 = 0.916$$

or

$$Q_p = 0.537 K^{1.624} (1 / T_p^{2.969})$$

$$= C_p (1 / T_p^{2.969})$$

$$C_p = 0.537 K^{1.624}$$

The result validation of the regression equation for the peak discharge ( $Q_p$ ) has an NSE of 91.6% and MAPE of 39.6%, which indicates that the prediction value is a good category. The comparison with the ANN result has the best model with a hidden layer of six, and the prediction of peak discharge has the

validation is almost the same as that of the non-linear model. The result validation of the ANN had an NSE of 81.4% and MAPE of 6.8%.

The result of  $T_p$  verification for the four watershed datasets produces the regression equation of peak discharge ( $Q_p$ ) with NSE = 94.0% and MAPE = 30.5%, which explains that the prediction value of the verification data is a good category. The verification using ANN had an NSE of 79.9% and MAPE of 13.2%, which indicates that the prediction value of the verification data is a good category. Table 5 presents the coefficients of ANN for peak discharge.

**Table 5. ANN Coefficient for peak discharge (Source: own study)**

Predictor	Predicted						
	Hidden Layer <sup>a</sup>						Output Layer
	H(1)	H(2)	H(3)	H(4)	H(5)	H(6)	$Q_p$
Input Layer	$T_p$	4.493	5.082	5.700	8.200	6.600	3.167
	K	92.546	235.874	426.154	1195.397	714.085	17.163
Hidden Unit Width		27.476	41.229	2.361	2.361	2.361	2.361
Hidden Layer	H(1)						14.493
	H(2)						28.228
	H(3)						55.591
	H(4)						97.750
	H(5)						98.780
	H(6)						2.641

The regression equation for time base is as follows:  
 $T_b = 3.919 Q_p^{-0.562} \cdot A^{0.569}; R^2 = 0.727$

or  $T_b = 3.919 \cdot A^{0.569} / Q_p^{0.562}$

**Table 6. Test Result of regression coefficient for time to base (Source: own study)**

Variable	Parameter	Estimate	Std. Error	t	p
Constant	b <sub>0</sub>	3.919	1.332	2.942	0.011
Q <sub>p</sub>	b <sub>1</sub>	-0.562	0.129	-4.357	0.001
A	b <sub>2</sub>	0.569	0.106	5.368	0.000

The regression equation for time base is as follows:

$$T_b = 3.919 Q_p^{-0.562} \cdot A^{0.569}; R^2 = 0.727$$

$$\text{or } T_b = 3.919 \cdot A^{0.569} / Q_p^{0.562}$$

The result validation of the regression equation for the time base (T<sub>b</sub>) has an NSE of 72.7% and MAPE of 16.6%, which indicates that the prediction value is a good category. The comparison with the ANN results

showed that the best model with a hidden layer of 9. However, the prediction of peak discharge has a higher validation than that of the nonlinear model. The result validation of the ANN had an NSE of 80.7% and MAPE of 11.2%. Table 7 presents the coefficients of the ANN for the time base.

**Table 7. ANN coefficient for time base (Source: own study)**

Predictor	Predicted										
		Hidden Layer <sup>a</sup>								Output Layer	
		H(1)	H(2)	H(3)	H(4)	H(5)	H(6)	H(7)	H(8)	H(9)	T <sub>b</sub>
Input Layer	Q <sub>p</sub>	98.78	97.75	21.43	32.92	18.22	55.59	3.11	13.76	8.13	
	A	3284.91	3725.05	1257.98	777.36	379.04	1376.75	552.24	157.31	238.24	
Hidden Unit Width		3.96	3.96	3.96	131.08	3.96	3.96	26.01	31.56	16.76	
	H(1)										19.800
	H(2)										31.430
	H(3)										39.716
	H(4)										26.690
Hidden Layer	H(5)										25.635
	H(6)										32.780
	H(7)										16.720
	H(8)										12.944
	H(9)										28.496

a. Displays the center vector for each hidden unit.

The verification of the prediction result for the T<sub>b</sub> to 4 watershed data is as follows. The result of the regression equation for peak discharge (Q<sub>p</sub>) has NSE = 71.1% and MAPE = 4.5%, which indicates that the prediction value of the verification data is a good category. However, the verification using the ANN result had an NSE of 60.1% and MAPE of 4.6%, which indicates that the prediction value of the verification data is a good category.

**3.3. Calibration of Model**

After obtaining the selected regression model for time to peak (T<sub>p</sub>), peak discharge (Q<sub>p</sub>), and time base (T<sub>b</sub>), the next stage was to calibrate the model. The aim of model calibration is to evaluate the accuracy of the modeled unit hydrograph parameters. From the

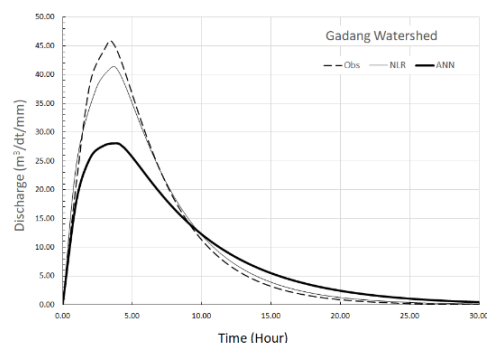
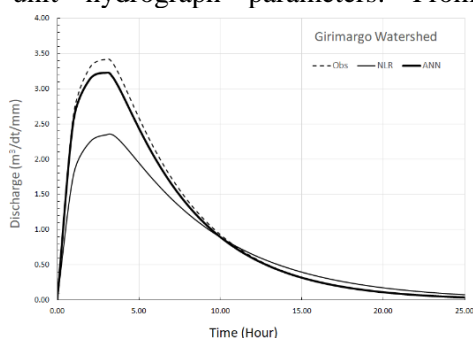
calibration process that is carried out based on the criteria of Nash (KN), the accuracy value for the nonlinear regression result is obtained as follows:

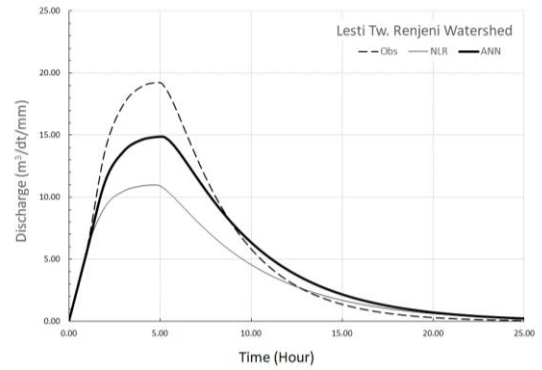
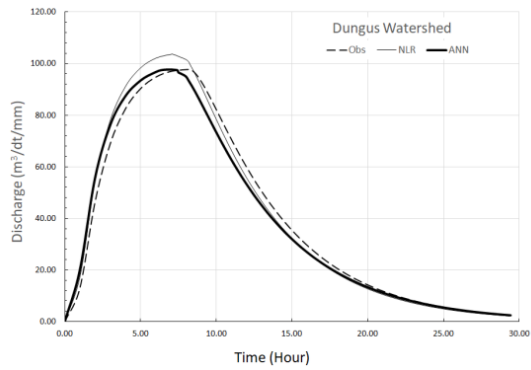
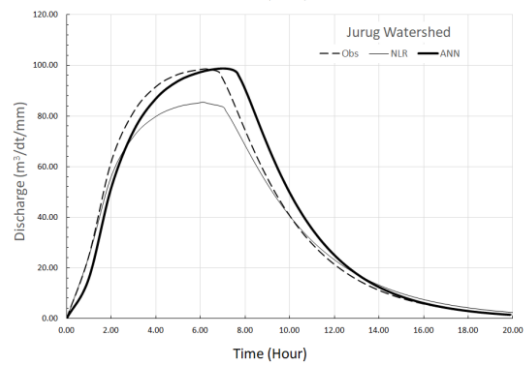
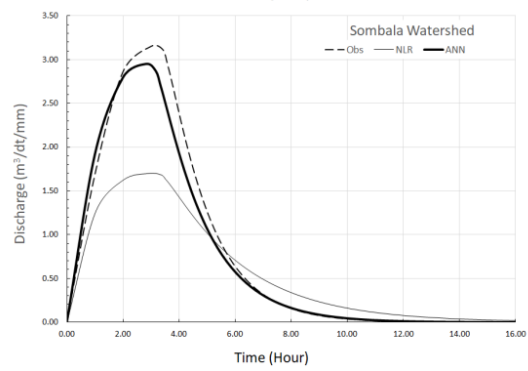
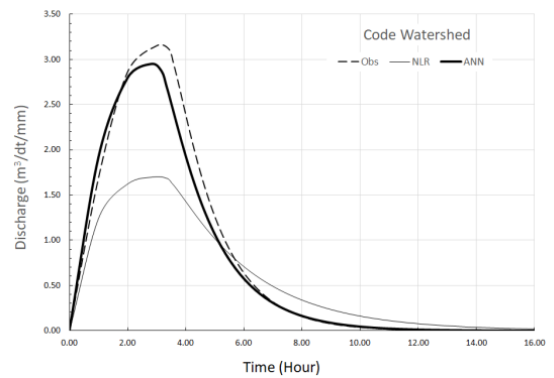
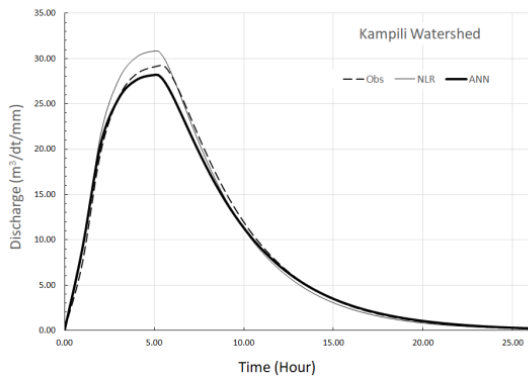
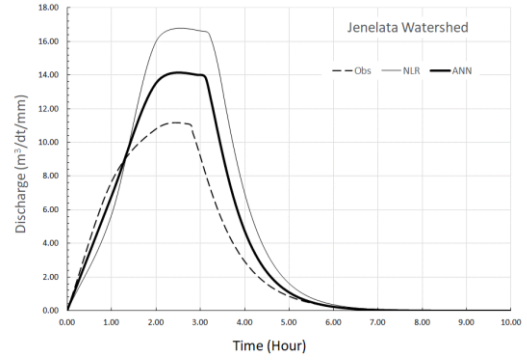
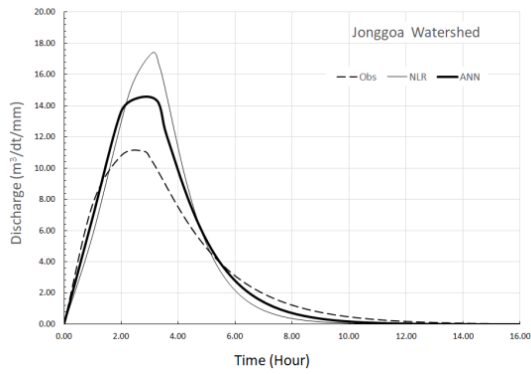
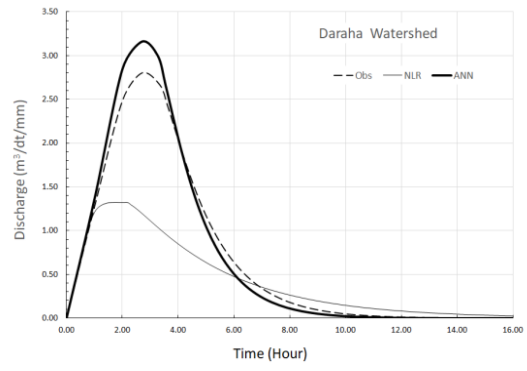
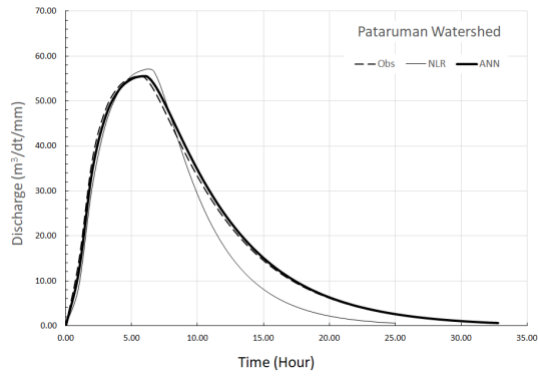
- Model time to peak (T<sub>p</sub>), KN = 89.2%
- Model of peak discharge (Q<sub>p</sub>), KN = 91.6%
- Model time base (T<sub>b</sub>), KN = 72.7%

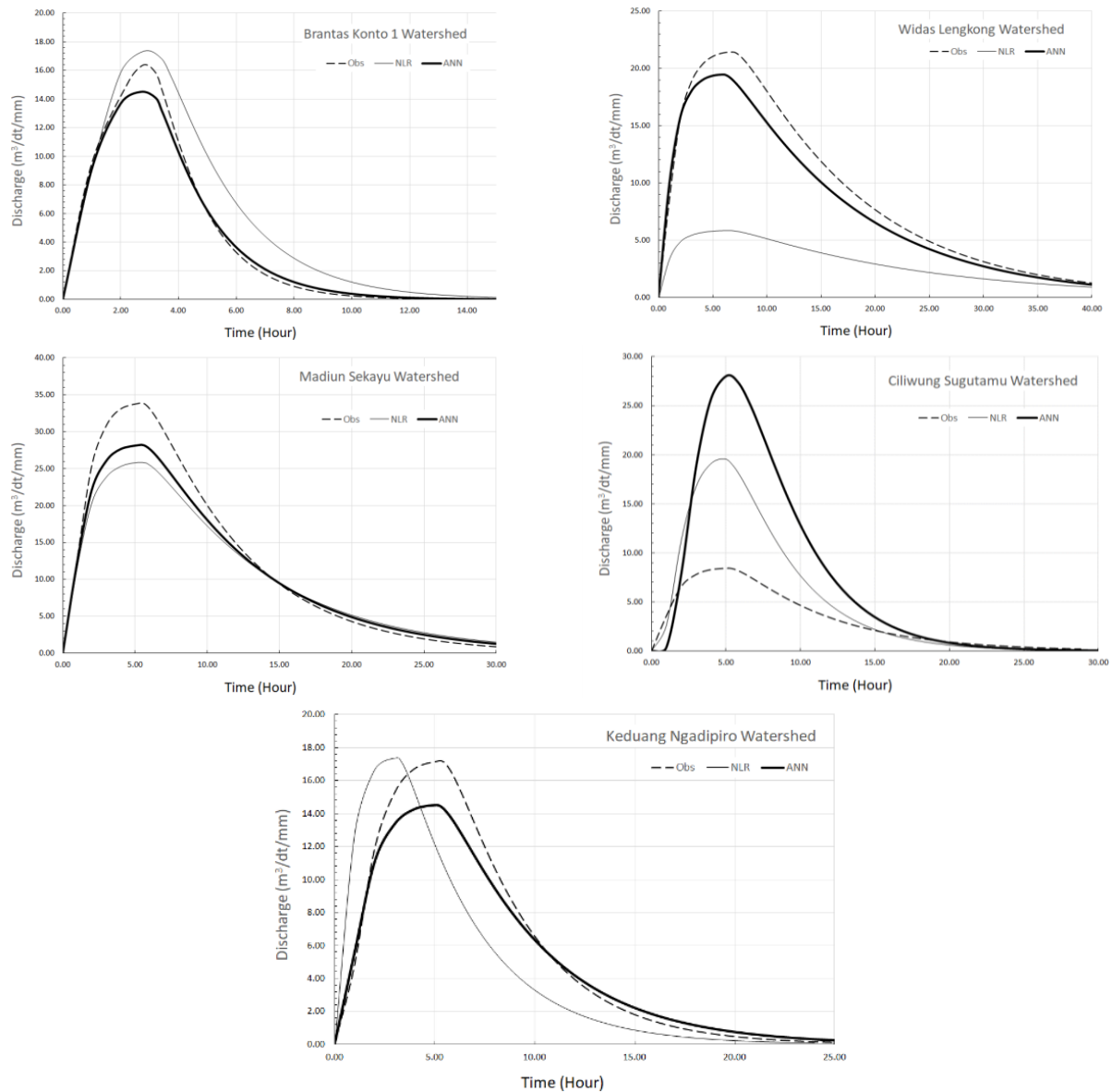
The calibration process that is performed based on the criteria of Nash (KN) produces the accuracy value for the ANN result as follows:

- Model of time to peak (T<sub>p</sub>), KN = 92.0%
- Model of peak discharge (Q<sub>p</sub>), KN = 94.0%
- Model of time base (T<sub>b</sub>), KN = 80.7%

The comparison of hydrographs between the observation data, Snyder SUH modification with NLR, and Snyder SUH modification with ANN is as follows:







**Figure 3. Comparison of hydrograph observation data, modification of Snyder SUH with NLR, modification of Snyder SUH with ANN (Validation) (Source: own study)**

### 3.4. Verification of Model

Verification of the model is carried out to determine whether the Snyder SUH model that is developed based on watershed characteristics in the research location is feasible for use. This model verification is carried out in Lesti Watershed, Bonto Jai Watershed, Brantas Gadang Watershed and Probo Kranggan Watershed. The four watersheds were not used to build the regression model or ANN. Using the equation of the selected regression model, the component of the unit hydrograph for the four watersheds can be determined with the following result.

The analysis results and reliability level of the model in the four watersheds based on the NLR results are as follows:

– Lesti Watershed:  $T_p$  model = 3.89 hours;  $T_p$  measured = 4.11 hour; with the accuracy = 94.6%,  $Q_p$  model =  $15.98 \text{ m}^3/\text{s}$ ;  $Q_p$  measured =  $16.22 \text{ m}^3/\text{s}$ ; with the accuracy = 98.5%;  $T_b$  model = 24.03 hours;  $T_b$

measured = 25.33 hours; with the accuracy = 94.9%.

– Bonto Jai Watershed:  $T_p$  model = 3.98 hours;  $T_p$  measured = 4.24 hours; accuracy = 93.9%;  $Q_p$  model =  $18.17 \text{ m}^3/\text{s}$ ;  $Q_p$  measured =  $19.25 \text{ m}^3/\text{s}$ ; accuracy = 94.4%;  $T_b$  model = 18.24 hours;  $T_b$  measured = 21.95 hours; and accuracy = 83.1%.

– Brantas Gadang Watershed:  $T_p$  model = 5.00 hours;  $T_p$  measured = 5.17 hours; with the accuracy = 96.7%  $Q_p$  model =  $24.84 \text{ m}^3/\text{s}$ ;  $Q_p$  measured =  $29.79 \text{ m}^3/\text{s}$ ; with the accuracy = 83.4%  $T_b$  model = 24.74 hours;  $T_b$  measured = 25.30 hours; with the accuracy = 97.8%.

– Progo Kranggan Watershed:  $T_p$  model = 5.08 hours;  $T_p$  measured = 4.80 hours; accuracy = 94.2%;  $Q_p$  model =  $15.12 \text{ m}^3/\text{s}$ ;  $Q_p$  measured =  $14.63 \text{ m}^3/\text{s}$ ; accuracy = 96.7%  $T_b$  model = 27.35 hours;  $T_b$  measured = 28.10 hours; and accuracy = 97.3%.

The analysis results and reliability level of the model in the four watersheds based on the ANN results

are as follows.

– Lesti Watershed:  $T_p$  model = 4.35 hours;  $T_p$  measured = 4.11 hours; with the accuracy = 94.2%;  $Q_p$  model = 14.61  $m^3/s$ ;  $Q_p$  measured = 16.22  $m^3/s$ ; with the accuracy = 90.1%;  $T_b$  model = 25.65 hours;  $T_b$  measured = 25.33 hours; with the accuracy = 98.7%.

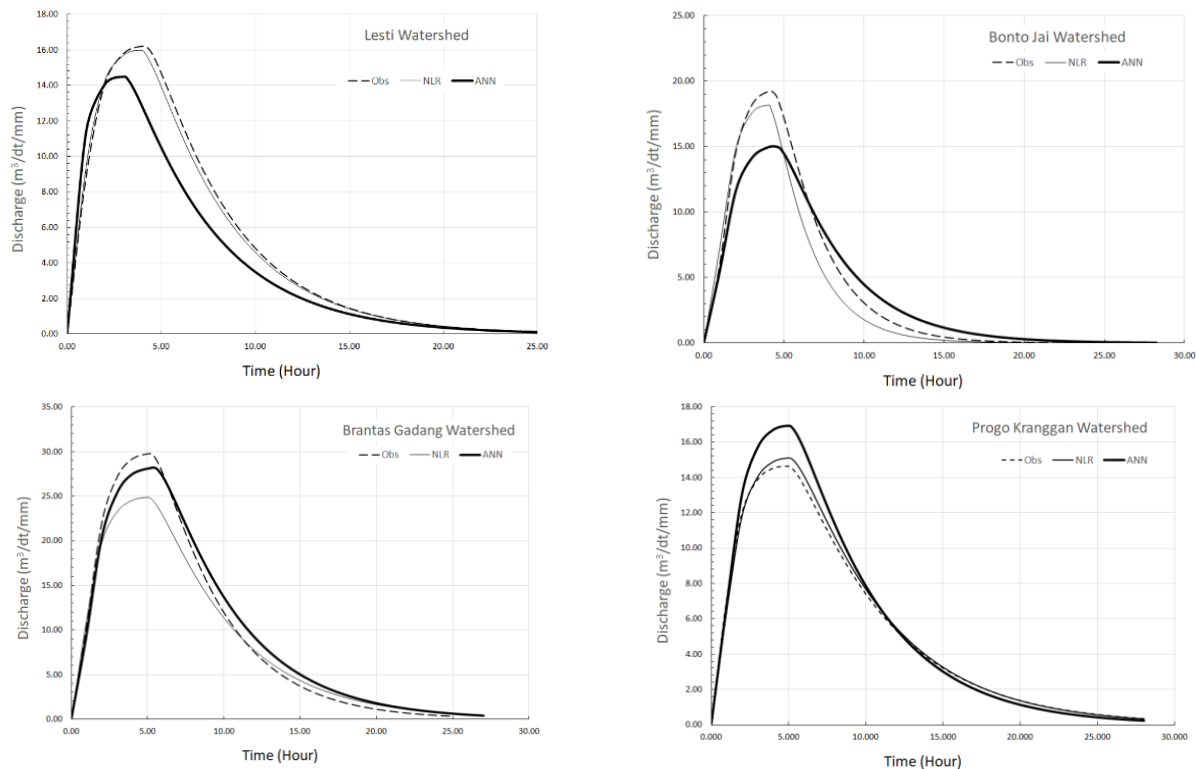
– Bonto Jai Watershed:  $T_p$  model = 4.50 hours;  $T_p$  measured = 4.24 hours; with the accuracy = 93.9%;  $Q_p$  model = 15.04  $m^3/s$ ;  $Q_p$  measured = 19.25  $m^3/s$ ; with the accuracy = 78.1%;  $T_b$  model = 28.26 hours;  $T_b$  measured = 21.95 hours; dengan with the accuracy= 71.3%.

– Brantas Gadang Watershed:  $T_p$  model = 5.38 hours;  $T_p$  measured = 5.17 hours; with the accuracy =

95.9%;  $Q_p$  model = 28.23  $m^3/s$ ;  $Q_p$  measured = 29.79  $m^3/s$ ; with the accuracy = 94.8%;  $T_b$  model = 26.69 hours;  $T_b$  measured = 25.30 hours; with the accuracy = 94.5%.

– Progo Kranggan Watershed:  $T_p$  model = 4.99 hours;  $T_p$  measured = 4.80 hours; with the accuracy = 96.0%;  $Q_p$  model = 16.92  $m^3/s$ ;  $Q_p$  measured = 14.63  $m^3/s$ ; with the accuracy = 84.3%;  $T_b$  model = 26.69 hours;  $T_b$  measured = 28.10 hours; with the accuracy = 95.0%.

The comparison of hydrographs between the observation data, Snyder SUH modification with NLR, and Snyder SUH modification with ANN is as follows:



**Figure 4. Comparison of hydrograph observation data, modification of Snyder SUH with NLR, modification of Snyder SUH with ANN (Verification) (Source: own study)**

## 4. Conclusion

Based on the above analysis and discussion, we can conclude that the modification result of the Snyder SUH model is as follows: a)  $T_p = C_t \cdot Lc^{0.325}$  ( $R^2 = 89.3\%$  and  $C_t = 0.680$  ( $W_r^{0.153} / S^{0.143}$ ); b)  $Q_p = C_p \cdot (1 / T_p^{2.969})$  ( $R^2 = 91.6\%$  and  $C_p = 0.537 K^{1.624}$ ); c)  $T_b = 3.919 A^{0.569} / Q_p^{0.562}$  ( $R^2 = 72.7\%$ )

The reliability of the modification of the Snyder SUH model using the NLR and ANN methods produces the following calibration results: a) model of time to peak ( $T_p$ ), KN = 89.2% and 92.0%; b) model peak discharge ( $Q_p$ ), KN = 91.6% and 94.0%; c) model time base ( $T_b$ ), KN = 72.7% and 80.7%, respectively.

The verification results of the modification of Snyder SUH model by using NLR and ANN methods is as follows: a) The model accuracy of  $T_p$  as the result of NLR: Lesti Watershed (94.6%), Bonto Jai

Watershed (93.9%), Brantas Gadang Watershed (96.7%), Progo Kranggan Watershed (94.2%); b) The model accuracy of  $Q_p$  as the result of NLR: Lesti Watershed (98.5%), Bonto Jai Watershed (94.4%), Brantas Gadang Watershed (83.4%), Progo Kranggan Watershed (96.7%); c) The model accuracy of  $T_b$  as the result of NLR: Lesti Watershed (94.9%), Bonto Jai Watershed (83.1%), Brantas Gadang Watershed (97.8%), Progo Kranggan Watershed (97.3%); d) The model accuracy of  $T_p$  as the result of ANN: Lesti Watershed (94.2%), Bonto Jai Watershed (93.9%), Brantas Gadang Watershed (95.9%), Progo Kranggan Watershed (96.0%); e) The model accuracy of  $Q_p$  as the result of ANN: Lesti Watershed (90.1%), Bonto Jai Watershed (78.1%), Brantas Gadang Watershed (94.8%), Progo Kranggan Watershed (84.3%); f) The model accuracy of  $T_b$  as the result of ANN: Lesti Watershed (98.7%), Bonto Jai Watershed (71.3%),

Brantas Gadang Watershed (94.5%), Progo Kranggan Watershed (95.0%).

This new model can be developed by adding more research locations; therefore, the model can be generally applied to many types of watersheds.

## Declarations

### Author Contributions

Conceptualization, R. and M.B.; methodology, R., M.B. and P.T.J.; formal analysis, R. and M.B.; writing—original draft preparation, all authors contributed equally; writing—review and editing, R.; visualization, A.M.; project administration, R. and M.B. All authors have read and agreed to the published version of the manuscript.

### Data Availability Statement

The data presented in this study are available in this article.

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Funding information is not available.

### Conflicts of Interest

The authors declare that there is no conflict of interests regarding the publication of this manuscript. In addition, the ethical issues, including plagiarism, informed consent, misconduct, data fabrication and/or falsification, double publication and/or submission, and redundancies have been completely observed by the authors.

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