Exploratory Research of New Curve Number System

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Abstract. In the past, the *CN* was determined through SCS handbook. In order to determine runoff prediction using SCS-CN model, selection of *CN* is important. However, the conventional *CN* methodology with inappropriate *CN* selection often produces inconsistent runoff estimation. Thus, the new direct curve number derivation technique based on rainfall-runoff datasets with supervised numerical optimization technique under the guide of inferential statistics was developed to improve the accuracy of surface runoff prediction. Furthermore, the two decimal point *CN* system was proposed in this study. The optimum *CN* of Melana site is 90.45 at alpha 0.01 with BCa 99 % confidence interval range from 90.45 to 95.12. The regional specific calibrated SCS-CN model with two decimal point *CN* derivation technique is out-performed the runoff prediction of conventional SCS-CN model and the asymptotic curve number fitting method.

1 Introduction

Recent year, due to rapid urbanization, the flooding scenarios occur more frequently in main city of Malaysia thus the rainfall-runoff model plays an important role in planning and managing the water resource system and flood control. There are various types of rainfall-runoff models available. Since 1954, the Soil Conservation Service Curve Number (SCS-CN) is one of the most popular rainfall-runoff model to predict direct runoff amount from agricultural site and then extended to urban watershed due to its simplicity. Nowadays, the SCS-CN model is incorporated into various software such as ArcGIS [15, 54, 55], remote sensing [15, 29, 35] and SWAT [24, 41]. The SCS-CN method also adopted by Malaysian government agencies and used as teaching purpose in engineering hydrology textbook.

Unfortunately, the accuracy and consistency of surface runoff prediction by using the curve number method derived from SCS had been questioned by some hydrologist from various countries [1, 8-9, 13, 16-18, 25, 27, 30, 33, 37-39, 50, 53, 56]. There was a study proved that by using tabulated *CN* values had a tendency to over-predict runoff amount [37]. Furthermore, a study in South Korea concluded that by referring the tabulated *CN* unable obtained satisfied runoff estimation [24]. The tabulated *CN* groups the effect of land use and land cover condition, hydrologic soil group (HSG), and antecedent runoff condition (ARC) based on watershed characteristics into a single coefficient [3, 42-45, 51]. As a

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result, CN is the most important parameter in SCS-CN model. Wrong selection of CN will lead to inappropriate runoff prediction [17]. In 1954, the SCS-CN model was used CN handbook to search the suitable CN value to represent the watershed. The CN value was then substituted into the CN formula as shown in Eq. 1 to find the maximum potential water retention of watershed (S). By fixing $\lambda=0.2$ with S value and event rainfall (P) substituted into the base SCS-CN model as shown in Eq. 2, the surface runoff (Q) could be determined.

$$CN = \frac{25,400}{S + 254} \quad (1)$$

$$Q = \frac{(P - I_a)^2}{P - I_a + S}; I_a = \lambda S; P > I_a, Q \neq 0; P < I_a, Q = 0 \quad (2)$$

Where,

CN = Curve Number Value

Error! Reference source not found. = Runoff depth (mm)

P = Rainfall depth (mm)

I_a = the initial abstraction (mm)

S = maximum potential water retention of a watershed (mm)

Basically, the CN is dimensionless watershed index which ranged from 0 to 100 to represent a watershed from high infiltration to fully permeable respectively. All the tabulate CN and charts of CN available in the Soil Conservation Service (SCS) National Engineering Handbook Section 4: Hydrology (NEH-4) [42-45] for agricultural areas, as well as Technical Release 55 (TR-55) [51] and applied in Technical Release 20 (TR-20) software for small and big urban catchments respectively were derived based on approximately 199 experimental watersheds in United States of America (USA) that ranged in size from 0.0971 ha to 18,600 ha which were located at 23 locations nationwide, using measurements of annual maximum rainfall and runoff collected between 1928 and 1954 and thousands of infiltrometer tests [49]. However, the information about the initial development of the curve number method has not been preserved [19, 50, 51].

The NRCS handbook [42-45] was created based on the hydrological condition and the soil group type in United States of America (USA) thus the soil condition and hydrological situation in Malaysia might not be the same as USA. There is a possibility to obtain unsuitable soil group type for a particulate watershed in Malaysia and thus produces unrealistic runoff volume. As a result, the SCS-CN practitioners should not blindly adopted the CN values in NRCS handbook [42-45], TR-55 [51] and TR-20 software. Although the SCS-CN model had integrated with remote sensing (RS) and geological information sensing (GIS) to produce a high resolution remote sensing imagery, the number of land cover types described in NEH-4 [42-45], TR-55 [51] and applied in TR-20 software was still so enormous that it was hard to classify into appropriate CN categories accurately [6]. As a result, Hawkins (1998) and Canters *et al.* (2006) [4, 20] suggested that CN tables should only be used as a guideline, and the actual CN should be determined based on local and regional data. Many researchers developed the CN calculation methods by incorporating the land cover information and the original CN in TR-55 [14] but the CN still unable to represent the regional specific watershed. In order to improve the surface runoff prediction result, most of the SCS practitioners practiced a common approach by “trial and error” to tweak the CN based on observed data without any statistical justification. By using “trial and error” method cannot obtain a consistent CN to represent particular watershed.

Table 2-2a Runoff curve numbers for urban areas ^{1/}					
Cover description		Curve numbers for hydrologic soil group			
Cover type and hydrologic condition	Average percent impervious area ^{2/}	A	B	C	D
<i>Fully developed urban areas (vegetation established)</i>					
Open space (lawns, parks, golf courses, cemeteries, etc.) ^{3/} :					
Poor condition (grass cover < 50%)		68	79	86	89
Fair condition (grass cover 50% to 75%)		49	69	79	84
Good condition (grass cover > 75%)		39	61	74	80
Impervious areas:					
Paved parking lots, roofs, driveways, etc. (excluding right-of-way)		98	98	98	98
Streets and roads:					
Paved; curbs and storm sewers (excluding right-of-way)		98	98	98	98
Paved; open ditches (including right-of-way)		83	89	92	93
Gravel (including right-of-way)		76	85	89	91
Dirt (including right-of-way)		72	82	87	89
Western desert urban areas:					
Natural desert landscaping (pervious areas only) ^{4/}		63	77	85	88
Artificial desert landscaping (impervious weed barrier, desert shrub with 1- to 2-inch sand or gravel mulch and basin borders)		96	96	96	96
Urban districts:					
Commercial and business	85	89	92	94	95
Industrial	72	81	88	91	93
Residential districts by average lot size:					
1/8 acre or less (town houses)	65	77	85	90	92
1/4 acre	38	61	75	83	87
1/3 acre	30	57	72	81	86
1/2 acre	25	54	70	80	85
1 acre	20	51	68	79	84
2 acres	12	46	65	77	82
<i>Developing urban areas</i>					
Newly graded areas (pervious areas only, no vegetation) ^{5/}					
		77	86	91	94
Idle lands (CN's are determined using cover types similar to those in table 2-2c).					

¹ Average runoff condition, and $I_a = 0.2S$.

² The average percent impervious area shown was used to develop the composite CN's. Other assumptions are as follows: impervious areas are directly connected to the drainage system, impervious areas have a CN of 98, and pervious areas are considered equivalent to open space in good hydrologic condition. CN's for other combinations of conditions may be computed using figure 2-3 or 2-4.

³ CN's shown are equivalent to those of pasture. Composite CN's may be computed for other combinations of open space cover type.

⁴ Composite CN's for natural desert landscaping should be computed using figures 2-3 or 2-4 based on the impervious area percentage (CN = 98) and the pervious area CN. The pervious area CN's are assumed equivalent to desert shrub in poor hydrologic condition.

⁵ Composite CN's to use for the design of temporary measures during grading and construction should be computed using figure 2-3 or 2-4 based on the degree of development (impervious area percentage) and the CN's for the newly graded pervious areas.

Fig. 1. Runoff Curve Number for Urban Area
Source: Table 2.2a. USDA, 1986, pp 2-5 [51])

Fig. 2. Chart of Rainfall versus Direct Runoff and Curve Number.
Source: Figure 2.1. USDA, 1986, pp 2-2 [51])

Recent decades, most of the research studies used the recorded daily rainfall-runoff (P - Q) data pairs from local or nearby watershed to derive the local CN value [17, 21, 22, 46-47]. There was few approaches for CN determination from observed P - Q data had been reported such as least-squares method (LSM) [50] and asymptotic fitting method (AFM) [17, 19]. The asymptotic fitting method created by Professor Hawkins was based on frequency matching concept. The rainfall-runoff data were sorted separately in descending order. The asymptotic fitting method basically classified into three different response behavior which was standard behavior, complacent behavior and violent behavior. For complacent behavior, there was no constant CN can be obtained. The asymptotic fitting method had been applied in some earlier research studies [2, 8, 34, 40, 49] In 2014, there was a study in Sicily by D’Asaro *et al* (2012) [8] discovered that the CN values found by the asymptotic fitting and least-squares fitting methods were all lower than the tabulated CN in handbook table [8]. In summary, there are two issues to be addressed in this article as stated below:

- Issue 1: Limitation of NRCS handbook in choosing an exact curve number.
- Issue 2: The regional specific curve number should be rounded or not.
- As a result, in this study will present a new approach to derive regional specific CN based on direct P - Q data sets under the guide of non-parametric inferential statistics and a new CN system with at least two decimal point is proposed to solve all issues above.

2 Data and methodology

2.1 Study site

In this study, Melana watershed is chosen to demonstrate the regional specific urban SCS model and new CN system. Melana watershed is located at Johor, Malaysia between 1°30’N to 1°35’N and 103°35’E to 103°39’E as shown in Fig 3 [5]. The total area of whole Melana watershed is 21.12 km². There are total of twenty-seven data sets of rainfall-runoff event between July 2004 to October 2004 were adopted from Chan (2005) [5] as shown in Table 1 [5]. Due to rapid urbanization in Melana watershed, there was only 20 % of Melana watershed was urbanized and after seven years later, more than 60 % of the area would be residential area [32]. As a result, in order to prevent the rainfall-runoff data sets affected by land cover land use change in Melana watershed due to rapid development, there was only a short period of rainfall-runoff data pairs were used in this study.

Fig. 3. Location of Melana Watershed (Source: Chan (2005) [5])

Table 1. Daily Rainfall-Runoff Event Data of Melana Watershed selected from July to October 2004.
(Source: Chan (2005) [5])

No.	Daily Rainfall (mm)	Daily Runoff (mm)	No.	Daily Rainfall (mm)	Daily Runoff (mm)
1	0.054	0.020	15	7.918	1.300
2	0.192	0.022	16	8.505	1.194
3	0.480	0.029	17	16.976	3.055
4	1.337	0.042	18	19.328	3.284
5	1.501	0.314	19	22.246	4.253
6	2.041	0.112	20	29.371	9.062
7	2.067	0.098	21	33.404	6.131
8	2.129	0.133	22	33.417	7.844
9	2.634	0.198	23	34.665	6.875
10	3.825	0.490	24	36.272	6.922
11	3.939	0.372	25	40.804	11.320
12	4.183	0.420	26	52.373	13.127
13	4.926	0.270	27	61.643	36.212
14	6.948	0.891			

2.2 Methodology

In this study, the non-parametric inferential statistics Bootstrapping technique, Bias corrected and accelerated (BCa) procedure with 2,000 sampling [11-12] was conducted by the help of supervised numerical optimisation technique to calibrate the base SCS-CN model and derived the regional specific *CN* for Melana watershed through the *P-Q* datasets directly. The Bootstrapping BCa statistics was chosen due to its robustness nature and the inferential ability via its confidence interval [7, 10, 52]. At the same study site, the Null hypothesis had been set up and all had been rejected in the previous research study in 2015 and 2017 by Ling [26-28] and also concluded the lambda value should not equal to 0.2. Furthermore, the 99 % confident interval (CI) range for λ and *S* were found out in previous publication too. In this research results having the same conclusion as previous research study [26-28], the lambda value fixed at 0.2 being rejected due to 0.2 was not in the λ confident interval range of (0.0004, 0.0005). The regional specific equivalent *CN*_{0.2} value for Melana site could be derived by using *CN* formula (*CN*=25,400/ (*S*+254)) which was proposed by SCS for *CN* comparison [19, 23]. The *S* _{λ} needed to correlate using general *S* _{λ} formula derived by Ling (2017) [27] as shown in Eq. 3 to substitute back into the *CN* formula.

$$S_{\lambda} = \frac{\left[P - \frac{(\lambda - 1)Q}{2\lambda} \right] - \sqrt{PQ - P^2 + \left[P - \frac{(\lambda - 1)Q}{2\lambda} \right]^2}}{\lambda}$$

(3)

Throughout this study, the derived *CN* value and new *CN* system will be created for Melana watershed.

2.3 Runoff Model Assessment

There are three types runoff model assessment with formulas as stated below:

$$E = 1 - \frac{RSS}{\sum_{i=1}^n (Q_{predicted} - Q_{mean})^2} \quad (4)$$

$$RSS = \sum_{i=1}^n (Q_{predicted} - Q_{observed})^2 \quad (5)$$

$$BIAS = \frac{\sum_{i=1}^n (Q_{predicted} - Q_{observed})}{n} \quad (6)$$

Q = Runoff amount (mm)

n = Total number of datasets

E = Nash-Sutcliffe index

RSS = Residual sum of squares

The three runoff model assessment as shown in Eq. 4, Eq. 5 and Eq. 6 are used to assess the asymptotic curve number model, calibrated SCS-CN model and conventional SCS-CN model. The lower the RSS value, the better the predictive model. Zero value of $BIAS$ indicates the predictive model has zero error. For positive $BIAS$ value means the predictive model has tendency to over-predict and vice versa. E between 0.0 to 1.0 indicates it is a perfect predictive model. When $E < 0$, the predictive model performs worse than using the mean value to predict the dataset [31].

3 Result and discussion

Fig. 4. Graph of Runoff Differences of CN in Hydrologic Soil Group of Urban Districts against Calibrated CN in term of Volumetric Difference vs Precipitation.

There are two issues had been proposed to solve as stated in introduction. First of all, the CN chosen from tabulated handbook is not sensitive due to a big gap in between CN range. The exact CN cannot pick correctly by referring to the NEH-4 handbook [42-45]. However, the accurate regional specific CN for Melana watershed can be calculated by using Eq. (1) under the guide of inferential statistic to exact decimal. The calculated CN of Melana watershed is 90.45 with significant level at alpha 0.01. The CN's 99 % BCa confident interval of Melana watershed is (90.45, 95.12). According to NEH-4 handbook [42-45], the CN of urban district with hydrological soil group A is 89 which is out of the obtained confident interval range of Melana watershed. Therefore, only soil group B, C and D are statistically significant to be considered for Melana watershed. Due to the climate change scenario and rapid urbanization posts a challenge to identify a suitable CN for the Melana

watershed via tabulated handbook. Therefore, by using NRCS handbook [42-45] to select *CN* has a risk of committing Type II error. Based on the graph in Figure 4, it clearly shows that there is uncertainty in runoff volume when the precipitation is changing. When the precipitation is less than 12 mm, the *CN* selected from handbook under urban district for Melana site has the tendency to over-predict runoff amount up to 35 thousands m³ and under-predict up to 15 thousands m³. When the rainfall amount is more than 12 mm, all the runoff volume for *CN*=84, 92, 94, and 95 will over-predict runoff amount up to 500 thousands m³. Moreover, Fig. 4 clearly shown that when the *CN* increases from 85 to 95, the runoff over prediction risk is significant and further magnified toward higher rainfall depth. As a result, by using NRCS handbook [42-45] having inconsistent runoff prediction.

There are some limitations of NEH-4 handbook [42-45]. Often in the tabulated *CN* handbook, there will be sudden jumps in *CN* choice based on four different hydrologic soil groups as shown in Fig 1. Moreover, by referring to the chart in handbook as stated in Fig 2, the selection of *CN* is in inches instead of SI unit and with a condition that λ value in SCS-CN model must fixed at 0.2. *CN* was proposed as whole number and still in use until today. However, rounding up the *CN* will cause a huge differences in runoff volume even it is just different by one *CN* class. This proves that *CN* variation affects the direct runoff estimation more than rainfall variability. Figure 5 shows the graph of runoff volumetric difference when the calculated *CN*=90.45 has been round up to 91 and round down to 90. It is obviously showed that the *CN* should not be round up or round down to whole number *CN*. When *CN*=91, the runoff volume tends to over-predicted up to 300 thousand m³. The maximum over-predicted runoff volume is 255 thousands m³ when the *CN* is 90. Rounding the *CN* causes a large amount of uncertainty runoff volumetric difference in Melana watershed.

Fig. 5. Graph of Runoff Differences of *CN* is 90 and 91 against Calibrated *CN* in term of Volumetric Difference vs Precipitation.

Fig. 7. Graph of Runoff Differences of one decimal point *CN* against Calibrated *CN* in term of Volumetric Difference vs Precipitation.

In this study, the new *CN* system approach with at least two decimal point being proposed instead of one decimal point. This is because there is still over-predicted nearly 273 thousand m³ volumetric difference at rainfall depth of 61.6 mm under *CN*=90.40 as shown in Fig. 7 whereas the volumetric difference for two decimal point *CN*=90.44 is only 270 thousand m³ as shown in Fig. 8. When the rainfall amount increase, the runoff volume differences also increase. The volumetric difference for two decimal point *CN* is smaller than one decimal point *CN*. Thus, the *CN* should be proposed as at least two decimal point.

Fig. 8. Graph of Runoff Differences of two decimal point *CN* against Calibrated *CN* in term of Volumetric Difference vs Precipitation.

Nowadays, most of the research studies used asymptotic *CN* fitting method (AFM) to find CN_{∞} in order to avoid inconsistent runoff prediction based on tabulated *CN* handbook. The CN_{∞} of Melana site is derived as 81.72 based on a previous research study in 2017 [26]. The accuracy of asymptotic *CN* runoff model is benchmark to conventional SCS-CN model and calibrated SCS-CN model by using few predictive model as tabulated in Table 2. Based on Table 2, it obviously stated that the calibrated model has higher Nash-Sucliffe, lower *RSS* and *BIAS* almost near to zero. Furthermore, the CN_{∞} value determined by AFM model did not has statistical significant proof at alpha 0.01.

Table 2. Assessment of Different Runoff Predictive Model.

Predictive Model	AFM Model	Calibrated Model	Conventional SCS Model
<i>p value</i>	Not Significant	0.01	Not Significant
<i>E</i>	0.817	0.863	-0.359
<i>RSS</i>	264.703	198.150	1970.120
<i>BIAS</i>	-0.172	0.187	4.927
$CN_{0.2}$	81.72	90.45	94.47
BCa 99% Confident Interval Range of $CN_{0.2}$	None	[90.45,95.12]	None
Residual: Median	0.618	-0.052	0.473
(Skewness) Residual	-2.399	-1.731	1.287
(Kurtosis) Residual	8.395	9.521	0.761
BCa 99% Confident Interval Range of Median Residual	[-1.043,1.659]	[-0.231,-0.008]	[-0.065,7.831]
Standard Deviation of Model error	3.186	2.754	7.111
Variance (Residual)	10.151	7.584	50.561

Fig.9 shows the graph of three runoff predictive model against the observe runoff amount versus precipitation in Melana watershed. By using the calibrated SCS model, the obtained I_a =0.043mm. The minimum observed rainfall amount is 0.054mm. This showed that the precipitate is more than initial abstraction, thus there is runoff generate. In the other hand, by applying the conventional SCS-CN model, the calculated I_a is 2.971mm. According to SCS, any rainfall amount less than I_a then no runoff generate. However, 1/3 of the rainfall events already violated the SCS constraint. Based on Fig.9 below, when the precipitation is less than 0.5 mm, the runoff amount cannot be more than 0.5 mm rainfall depth while the predicted runoff depth from conventional SCS-CN and AFM model are higher than 0.5 mm rainfall depth. There is a huge different between the predicted runoff against the observed runoff depth when the precipitation reaches 61.6 mm. The SCS-CN model tends to over-predicted up to 326 thousand m³ runoff (equivalent to 130 Olympic size swimming pools) and under-predict up to 6 thousand m³ as compared to calibrated SCS-CN model. In the other hand, the AFM model tends to over-predicted up to 69 thousand m³ runoff (equivalent to 28 Olympic size swimming pools) and under-predict nearly 62 thousand m³ (equivalent to 25 Olympic size swimming pools).

Fig. 9. Graph of Runoff Amount vs Precipitation in Melana Watershed

4 Conclusion

Conclusion, the *CN* should not select from the NRCS handbook. By using the new *CN* derivation approach with supervised numerical optimization technique under the guide of non-parametric inferential statistics can overcome the difficulty of selecting an accurate *CN* from tabulated handbook. Throughout this study, the regional specific calibrated runoff predicted model performs better than SCS-*CN* model and asymptotic curve number fitting method. The derived *CN* of Melana watershed is 90.45 under statistical significant at alpha 0.01 level. The predicted runoff results from conventional SCS-*CN* model tend to 326 thousand m³ which is equivalent to 130 Olympic size swimming pools and under-predict up to 6 thousand m³. Lastly, the new two decimal point *CN* system should be applied instead of rounding up the *CN*. Although the *CN* value derived from asymptotic *CN* method is two decimal point of *CN* value, the new *CN* derivation approach still performs better with the proof of higher *R* value, lower *BIAS* and lesser *RSS* are obtained. The *CN* value determined by asymptotic *CN* method for Melana site is 81.72 which has tendency over-predicted up to 69 thousand m³ runoff (equivalent to 28 Olympic size swimming pools) and under-predict nearly 62 thousand m³ (equivalent to 25 Olympic size swimming pools).

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