1	Separation of surface flow from subsurface flow in catchments using runoff
2	coefficient
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11 Abstract:

Separating surface flow (SF) from subsurface flow (SSF) based on direct runoff 12 measurements in river gauges is an important issue in hydrology. In this study, we 13 developed a simple and practical method, based on runoff coefficient (RC), for 14 separating SF from SSF. RC depends mainly on soil texture, land use and land cover, 15 but we also considered the effect of slope and rainfall intensity. We assessed our 16 RC-based method for three different soil types by comparing the value obtained with 17 laboratory rainfall simulator data. The correlation coefficient between observed and 18 19 calculated data exceeded 0.93 and 0.63 when estimating SF and SSF respectively. The method was then used to separate SF and SSF in two catchments (Heng-Chi and 20 San-Hsia) in Northern Taiwan, and the results were compared with those produced 21 by the geomorphological instantaneous unit hydrograph (GIUH) model. Test 22 revealed that, if RC is calculated accurately, the proposed method can satisfactorily 23 separate SF from SSF at catchment scale. 24

25 Key Words: Surface flow, Subsurface flow, Separation, Runoff coefficient

1. Introduction

Estimating direct runoff is important in flood risk assessment and in the design of hydraulic structures such as diversion and storage dams. In general, total runoff occurring in streams consists of three components: surface runoff, subsurface flow, and base flow. The sum of surface runoff and subsurface flow is commonly defined as direct runoff.

Surface runoff (SF) is usually the most important of such three components. Many
rainfall-runoff models have been proposed to compute the surface flow of ungauged
catchments (Menberu et al , 2014, Sabzevari, 2017; Keshtkaran et al, 2018; Petroselli
et al, 2020a, b; Dehghanian et al, 2020).

However, in hilly catchments with very permeable soil or dense vegetation cover, the rate of infiltration is high and can lead to rapid subsurface flow. In such catchments, subsurface flow can enter streams at the lower part of hillslopes and contribute effectively to direct flow (Singh, 1988, Sabzevari et al, 2013).

The underground flow can be slow or quick. The quick underground flow is often called saturated subsurface flow (SSF), and it usually occurs near the soil surface, eventually entering the streams. Slow underground flow is generally a source of groundwater recharge. It is formed through infiltration of water into deeper layers of the soil and eventually enters rivers as base flow (BF).

Based on the Dunne-Black runoff mechanism, the lower soil layers are saturated by SSF, which eventually joins surface flow (SF) entering the streams (Chow et al. 1988). To separate SF from SSF, the complicated interactions of saturated and unsaturated zones in soil must be determined. Several previous studies have attempted to separate SF from SSF, but this topic still needs further investigation (Hursh et al. 1941; Wels et al. 1991; Johst et al. 2013).

Harris et al. (1995) proposed a hydrograph separation method for runoff source modeling based on continuous open system isotope mixing, using a variable source area and three isotopic reservoirs. They estimated time-dependent streamflow contributions of SF and SSF in storm rainfall events, and estimated parameters for determining the relationships between saturated area fraction and streamflow, and between saturated area and subsurface water storage (Harris et al. 1995).

A stable environmental isotope was used by Tekeli and Sorman (2003) to investigate the rainfall-runoff relationship and to separate SF from SSF in hydrographs, based on analysis of water samples from rainfall, runoff (total discharge), springs (subsurface flows), and wells (groundwater) in the Guvenc Basin, Turkey. Through this approach, they successfully determined the contribution of SSF originating from various sublayers.

Foks et al. (2019) used an optimal hydrograph separation technique based on a twoparameter recursive digital filter and specific conductance mass-balance constraints
to estimate the base flow contribution to observed flow in river gauges.

Some previous studies of SSF at hillslope scale have used existing methods based on the Dupuit-Forchheimer approach, Boussinesq equation, or numerical solution of complex three-dimensional equations (e.g., Troch et al. 1993; Chen et al. 1994a, 1994b). Numerical methods give good accuracy, but most hydrologists want simpler methods. Some hydrological models have also been used to estimate SSF (Robinson and Sivapalan 1996; Lee and Chang 2005; Sabzevari et al. 2013; Sabzevari and Noroozpour, 2014).

Lee and Chang (2005) developed the geomorphological instantaneous unit
hydrograph (GIUH) model for predicting SSF. Surface and subsurface travel time
are the most important parameters in the GIUH model. Subsurface travel time is a

function of overland length and slope and soil characteristics, e.g., hydraulic
conductivity and porosity. Lee and Chang (2005) used the GIUH model to separate
SF and SSF in the Heng-Chi basin, Taiwan.

Sabzevari et al. (2013) modified the Lee and Chang (2005) model by calculating the
SSF hydrograph of the catchment through convoluting the subsurface GIUH model
in the infiltration hyetograph. In their modified version, a more accurate saturation
model was used to predict SF and SSF according to the Dunne-Black mechanism.
Sabzevari et al. (2013) applied the modified model in the Kasilian catchment, Iran,
to separate SF and SSF.

Sabzevari and Noroozpour (2014) examined the role of hillslope shape and profile
curvature on SF and SSF in complex hillslopes and applied a new complex saturation
model to separate the saturation region. They used the model to estimate SSF in a
small basin, No. 125 in Walnut Gulch, Arizona, USA.

The theory of Sabzevari et al. (2013) was used by Petroselli (2020) that generalized the EBA4SUB rainfall-runoff model (Piscopia et al. 2015; Petroselli and Grimaldi 2018; Petroselli et al. 2020 a, b), originally developed only for SF estimation, introducing within the model the subsurface flow process and allowing its application to both Hortonian and Dunne-Black runoff formation mechanisms, employing the Width Function Based IUH framework.

Laboratory physical models are commonly used to validate the results of SF and SSF estimation models. Essig et al. (2008) devised a laboratory set-up to separate deep flow and surface flow for sloping surfaces. The equipment consisted of a rainfall simulator device with length 1.52 m and width 1.22 m, and a soil box with depth 78 cm, which was equipped to measure SSF and the SF separately by two weirs. In Essig et al. (2008), the separation between SF and SSF was also modeled by the

Hydrus 2D (numerical) model for different slopes up to 10 degrees, and the resultswere compared.

The runoff coefficient (RC) is used to separate the amount of excess rainfall from infiltration in many hydrological models (e.g. the rational method), in doing so trying to express the relationship between SF and SSF. The RC value indicates the ratio of surface runoff depth to total rainfall depth. Based on RC values, the surface runoff depth and infiltration depth can be determined (Kim and Shin 2018; Kim et al. 2016).

RC depends on factors such as soil type and land use, slope and rainfall rate. In this study, we developed a new method for separating SF and SSF in catchments by investigating the effect of slope and rainfall intensity on RC. We verified the method using laboratory data in the hillslope dimension. Finally, we tested the method in separation of SF and SSF for two catchments (Heng-Chi and San-Hsia) in northern Taiwan and compared the modeled results with observed direct runoff.

A number of studies have been presented on the separation of surface and subsurface
flow from runoff hydrograph(Hursh and Brater, 1941; Wels et al, 1991; Johst et al,
2013).

Lee et al (2015) introduced a new method to estimate the runoff coefficient through the infiltration analysis based on the comparative results of the existing runoff coefficient method. The effect of rainfall intensity and soil characteristics to runoff coefficient was also analyzed by the FFC-COBRA model and effective rainfall separation method based on NRCS CN. This result showed that runoff coefficient in this study is not only in the range of runoff coefficient, but also over the upper limit of 0.10~0.22 at 'forest, etc' from ASCE.

Johst et al (2013) studied in 31 ha headwater basin in Western Germany to separate the surface flow and subsurface flow from runoff hydrograph. In this study, the contribution of infiltration excess and saturation overland flow as well as matrix and preferential flow has been assessed along a deeply incised channel of 300 m length. Measurable parameters and simple algorithms were used to assess the flow rate of the different runoff components. The results showed that during wet conditions the subsurface flow rates exceed the surface flow rates tremendously.

The main classification of the sections of this article is as follows: In the first part, the equations of separation of surface and subsurface flow are presented, then the effect of rainfall intensity and slope on surface flow is investigated. In the next section, the results of two laboratory models for measuring surface and subsurface flow are presented and the observed runoff coefficients and the calculated runoff coefficient are evaluated. Finally, the proposed method for two catchments in Taiwan is evaluated.

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2. Materials and Methods

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141 **2.1.** Separation of surface flow from subsurface flow

The amount of rainfall or liquid precipitation (P) falling on a hillslope (Fig. 1) can
be calculated from the sum of surface runoff (R) and infiltration (F):

(1)

144 P=F+R

Introducing RC (R=RC×P) and substituting P with R/RC in Eq. 1, we can calculate the ratio of surface runoff depth to infiltration (subsurface runoff) depth as a function of RC:

148 R/F = RC/(1-RC) (2)

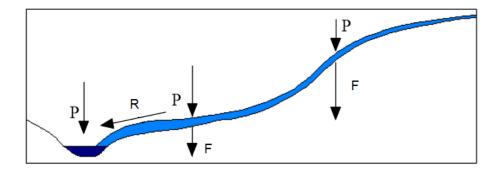


Fig. 1: Schematic diagram of the rainfall-runoff process in a hillslope, where P is precipitation, F
 is infiltration, and R is surface runoff (Tarboton, 2003)

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In this study, we assumed that the bedrock is close to the surface and that all infiltrated water is SSF and does not contribute to groundwater. In the steady state condition with excess rainfall intensity (I_e) on a hillslope, the maximum surface and subsurface flow (Q_s and Q_{sub} , respectively) can be calculated as (Akan and Houghtalen 2003):

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$$Q_s = I_e \times A$$
 (3)

- 159 and
- $160 \quad Q_{sub} = I_f \times A \tag{4}$

where I_f is the recharge rate into the soil layer and *A* is the contributing area of the hillslope. The ratio (m) of the SF peak to the SSF peak can be calculated as:

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$$m = \frac{Q_s}{Q_{sub}} = \frac{I_e \times A}{I_f \times A} = \frac{I_e}{I_f}$$
(5)

- 164 or:
- 165 $m = \frac{I_e}{I_f} = \frac{R}{F}$ (6)

where R is surface runoff depth and F is infiltration depth. From Eq. 2, we have ratioof the SF peak to the subsurface flow peak as a function of RC, so:

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$$m = \frac{Q_s}{Q_{sub}} = \frac{RC}{(1-RC)}$$
 (7)

Based on Eq. 3, we can calculate the coefficients *m* and *RC* if we know peak discharge as SSF and SF. In the next step, we need to validate Eq. 3 to investigate the relationship between RC and SF and SSF.

Assuming that base flow is zero ($Q = Q_s + Q_{sub}$), based on total observed flow (direct runoff) Q_s and Q_{sub} are calculated as follows:

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$$Q_s = RC \times Q$$

175 $Q_{sub} = (1 - RC) \times Q$ (8)

Thus using Eq. 8, SF and SSF can be calculated separately. In this study, the results
obtained using Eq. 8 were validated using the results of laboratory rain simulations
on artificial slopes.

The most important innovation of this study is that the separation of surface flow from subsurface flow according to Eq. 8 based on runoff coefficient. RC was calculated only from the observed surface flow. In this research, two laboratory models and observed subsurface flow and observed surface flow were used to evaluate the Eq.8.

184 **2.2. Calculation of runoff coefficient (RC)**

. Runoff coefficient is the percentage of rainfall that is converted to runoff.
Calculation of RC is complex due to the heterogeneity of infiltration across
catchments, and in practice it is impossible to provide an average RC for a
catchment. For small hillslopes, we can calculate the average RC by measuring total
runoff from the hillslope, using one of the following two methods:

190 Method 1) RC is calculated as:

 $191 \quad RC = V/(P \times A) \tag{9}$

- where *V* is runoff volume (i.e. the area below the graph of surface runoff hydrograph)and P is rainfall depth. This method is more accurate than method 2.
- Method 2) The RC value is obtained by the rational method, used to predict the runoff peak of small basins, and it is calculated as:
- 196 $RC = Q_p / (0.278 \times i \times A)$ (10)
- where Q_p is peak surface runoff (m³ s⁻¹), *i* is rainfall intensity (mm h⁻¹), and A is basin area (km²).

199 2.2.1. Relationship between rainfall and RC

In general, greater amounts of rainfall and lower infiltration rates lead to higher surface runoff or higher RC values.

- The SCS-CN infiltration method calculates RC (= R/P) using the following equation (Mishra and Singh 2013):
- 204 $RC = R/P = [(P 0.2 \times S)^2 / (P \times (P + 0.8 \times S))]$ (11)

where P is rainfall depth in inches and S is potential maximum retention, which is
equal to (1000/CN-10), where CN is the selected curve number based on land use,
group (from A, sand, to D, clay) and antecedent moisture conditions (from I, dry
soil, to III, wet soil) (Chow et al. 1962).

Figure 2 shows the change in RC as a function of change in rainfall intensity from 31.73 to 63.46 mm h⁻¹ for a 3-hour rainfall event for different values of CN based on Eq. (11).

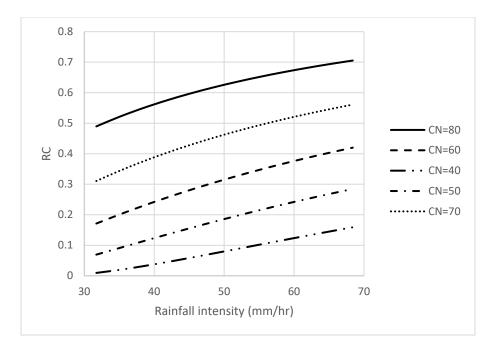




Fig. 2: Relationship between rainfall intensity and RC for different CN values.

The CN range for soils with high, medium, and low permeability is 10-30, 40-60, and 70-90, respectively, which directly influences RC. For example, a 20 mm increase in rainfall leads to an increase of around 25%, 15%, and 3% in RC for high, medium, and low permeability soils, respectively.

219 **2.2.2. Effect of slope on RC**

Slope is another influential parameter on surface runoff and infiltration (Ribolzi et al. 2011; Morbidelli et al. 2015; Morbidelli et al. 2018). In general, with steeper ground slope, the potential for infiltration is lower and consequently the amount of surface runoff generated will be higher (RC increase).

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Table 1 presents the RC values for different types of soils and land uses on different slopes (Liu and De Smedt, 2004).

The runoff coefficient for different slopes can be calculated as (Liu and De Smedt, 2004):

230 $C = C_0 + (1 - C_0) \times (S/(S + S_0))$ (12)

where *C* is RC for slope *S* % and C_0 is RC for horizontal slope $S_0(0\%)$, which is calculated from Table 2.

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Table 1. Effect of slope and land use on RC

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(Liu and De Smedt, 2004)

Land use	Slope (%)	Sand	Loamy sand	Sandy loam	Loam	Silt loam	Silt	Sandy clay	Clay	Silty clay	Silty clay	Clay
E	-0.5	0.02	0.07	0.10	0.12	0.17	0.20	loam	loam	loam	0.27	0.40
Forest	<0.5	0.03	0.07	0.10	0.13	0.17	0.20	0.23	0.27	0.33	0.37	0.40
	0.5-5	0.07	0.11	0.14	0.17	0.21	0.24	0.27	0.31	0.37	0.41	0.44
	5-10	0.13	0.17	0.20	0.23	0.27	0.30	0.33	0.37	0.43	0.47	0.50
	>10	0.25	0.29	0.32	0.35	0.39	0.42	0.45	0.49	0.55	0.59	0.62
Grass	< 0.5	0.13	0.17	0.20	0.23	0.27	0.30	0.33	0.37	0.43	0.47	0.50
	0.5-5	0.17	0.21	0.24	0.27	0.31	0.34	0.37	0.41	0.47	0.51	0.54
	5-10	0.23	0.27	0.30	0.33	0.37	0.40	0.43	0.47	0.53	0.57	0.60
	>10	0.35	0.39	0.42	0.45	0.49	0.52	0.55	0.59	0.65	0.69	0.72
Crop	< 0.5	0.23	0.27	0.30	0.33	0.37	0.40	0.43	0.47	0.53	0.57	0.60
-	0.5-5	0.27	0.31	0.34	0.37	0.41	0.44	0.47	0.51	0.57	0.61	0.64
	5-10	0.33	0.37	0.40	0.43	0.47	0.50	0.53	0.57	0.63	0.67	0.70
	>10	0.45	0.49	0.52	0.55	0.59	0.62	0.65	0.69	0.75	0.79	0.82
Bare	< 0.5	0.33	0.37	0.40	0.43	0.47	0.50	0.53	0.57	0.63	0.67	0.70
Soil	0.5-5	0.37	0.41	0.44	0.47	0.51	0.54	0.57	0.61	0.67	0.71	0.74
	5-10	0.43	0.47	0.50	0.53	0.57	0.60	0.63	0.67	0.73	0.77	0.80
	>10	0.55	0.59	0.62	0.65	0.69	0.72	0.75	0.79	0.85	0.89	0.92

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Table 2. RC values for different land uses and soil types on land with zero slope (source: Liu and De Smedt, 2004)

Land use	Sand	Loamy sand	Sandy loam	Loam	Silt Ioam	Silt	Sandy clay loam	clay loam	Silty clay loam	Sandy clay	Silty clay	Clay
Forest	0.680	0.650	0.620	0.590	0.560	0.530	0.500	0.470	0.440	0.410	0.380	0.350
Grass	0.580	0.551	0.522	0.493	0.464	0.435	0.405	0.376	0.347	0.318	0.289	0.260
Crop	0.500	0.471	0.442	0.413	0.384	0.355	0.325	0.296	0.267	0.238	0.209	0.180
Bare soil	0.420	0.393	0.365	0.338	0.311	0.284	0.256	0.229	0.202	0.175	0.147	0.120

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240 **3. Physical model description**

Laboratory tests were conducted using an experimental set-up at the Hydraulic Laboratory of the Civil Engineering Department at Estahban Azad University, Iran (Fig. 3). It consists of a rainfall simulator over a soil box (length 1.92 m, width 1 m, depth 35 cm, which was filled with loamy sand soil and sandy clay soil. Tests were run with four slopes (0, 3, 6, and 9 degrees) and three rainfall intensities (31.73, 47.6,

and 63.46 mm/ h). The rainfall duration in most event has been about 300 minutes.

- Each test has been tested after drying the soil that soil moisture error does not affect
- measurements. The intensity of rainfall was tested by nozzles before each test.
- 249 SF and SSF were measured by two separate weirs.

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- Fig. 3. Physical soil model and rainfall simulator used in laboratory tests.
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4. Results and Discussion

4.1 Hydrograph produced by the physical model

For loamy sand soil, the maximum SF measured at the outlet of the physical model varied between 0.78-0.89, 1.31-1.39, and 1.76-1.89 l min⁻¹ for the 31.73, 47.6, and 63.46 mm h⁻¹ rainfall events, respectively (Figs. 4a-4c). The maximum SSF ranged between 0.112 and 0.228 l min⁻¹ (Table 3). Substituting the maximum values of observed SF and SSF into Eq. 7 allowed us to calculate RC of the loamy sand (Table 3) for different rainfall events and slopes (Table 4). The observed and calculated runoff coefficient showed a significant positive correlation ($R^2 = 0.93$) (Fig. 5a).

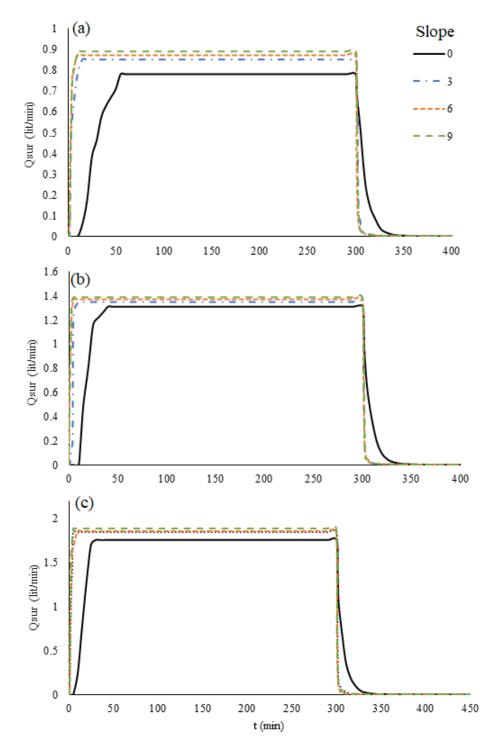
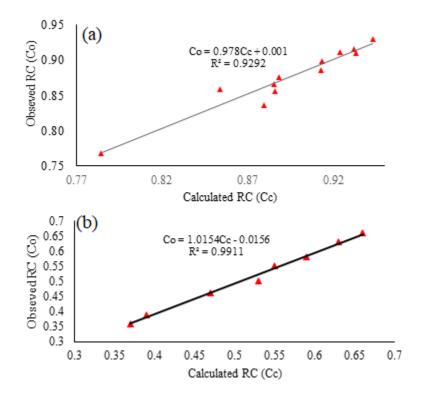


Fig. 4. Observed SF from the physical model with loamy sand soil with different land slope (0-9 degrees) at rainfall intensity of a) 31.73 mm h⁻¹, b) 47.6 mm h⁻¹, and c) 63.46 mm h⁻¹.

Table 3 shows the maximum values of SF and SSF, SF to SSF ratio, calculated RC (Cc) and observed RC (Co) according to the surface runoff volume method.



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Fig. 5. Correlation between calculated and observed RC for a) loamy sand and b) sandy clay soil.

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Table 3. Observed and calculated value of RC based on observations of SF and SSF obtained in aphysical model with loamy sand and sandy clay soil.

Type of soil	Rainfall intensity	Slope, degrees	SSF max.	SF max.	SF/SSF	Calculated RC	Observed RC
	$mm h^{-1}$	-	1 min ⁻¹	1 min ⁻¹		Eq. 7 (C _C)	Eq. 9 (C _o)
		0	0.214	0.78	3.64	0.784	0.768
Loamy		3	0.116	0.85	7.33	0.880	0.836
sand	31.73	6	0.112	0.87	7.77	0.886	0.856
		9	0.112	0.89	7.95	0.888	0.876
		0	0.224	1.31	5.85	0.854	0.859
Loamy		3	0.129	1.35	10.47	0.913	0.886
sand	47.6	6	0.130	1.37	10.54	0.913	0.899
		9	0.114	1.39	12.19	0.924	0.912
		0	0.228	1.76	7.72	0.885	0.866
Loamy		3	0.132	1.85	14.02	0.933	0.910
sand	63.46	6	0.135	1.86	13.78	0.932	0.915
		9	0.114	1.89	16.58	0.943	0.930
		0	0.500	0.00	0.00	-	-

Sandy clay	31.73	3	0.500	0.00	0.00	0.00	0.00
		6	0.320	0.18	0.56	0.36	0.36
		9	0.270	0.23	0.85	0.46	0.46
		0	1.00	0.00	0.00	-	-
Sandy clay	47.6	3	0.60	0.40	0.67	0.40	0.39
		6	0.47	0.53	1.13	0.53	0.50
		9	0.45	0.55	1.22	0.55	0.55
		0	1.50	0.00	0.00	-	-
Sandy clay	63.46	3	0.62	0.88	1.42	0.59	0.58
		6	0.55	0.95	1.73	0.63	0.63
		9	0.50	1.00	2.00	0.67	0.66

In tests with loamy sandy soil, the observed RC initially increased with increasing 277 slope, e.g. at a slope of 3 degrees above the horizontal (0 degrees), it increased by 278 about 12% on average (Table 3). However, a further increase in slope from 3 to 6 279 degrees and from 6 to 9 degrees gave little change in RC. The average increase in 280 SF with an increase in rainfall intensity from 31.73 mm h⁻¹ to 47.6 and 63.46 mm⁻¹ 281 was between 6.5 and 8.5%. In our results, the RC depended on soil type, slope, and 282 land use, and was weakly related to rainfall intensity in different events. Thus in 283 practice, it was impossible to calculate RC accurately. 284

The observed data for sandy clay soil were similar to those for loamy sand soil (Table 285 3 and Fig. 5b). The calculated and observed RC values for the sandy clay were lower 286 than those for the loamy sand, because of the higher permeability of the sandy clay. 287 At 0 degrees of slope, all rainfall contributed to subsurface flow for the sandy clay, 288 and thus the RC is not shown in Table 3. At 6 degrees of slope, the RC increased by 289 28% and 8% for a rainfall intensity of 47.6 and 63.4 mm h⁻¹, respectively (Table 3). 290 Increasing the rainfall intensity also led to increasing RC for the sandy clay, for 291 instance for a slope of 6 degrees, the RC for a rain intensity of 31.7, 47.6, and 63.46 292 mm h^{-1} was 0.36, 0.53, and 0.59, respectively (Table 4). 293

The results of tests in the physical model for two different soils clearly confirmed that the method developed in this study can be recommended as suitable and simple approach to separate SF and SSF in rainfall-runoff analysis of hillslopes. As shown, different parameters, e.g., soil type, land use, slope, and rainfall intensity, influenced the RC value.

4.2. Verification based on observed and calculated SSF and SF

In this section, for more accurate validation of the proposed method, surface and subsurface flow information of the other two different soils were used. The first soil was clay loam and this soil was evaluated by the device according to Figure 3. The second soil was loamy and SF and SSF information was examined based on Morbidelli et al. (2015) study.

For first verification of the method, we compared the observed and calculated SSF and SF values obtained for different rainfall rates and slopes (Table 4). For this, we filled the soil box in the physical model (Fig. 3) with a clay loam soil and applied three different rainfall intensities (15.63, 31.3, and 46.9 mm/h). For this experiment, the RC for a slope of 3, 5, and 10 degrees was 0.61, 0.67, and 0.79, respectively.

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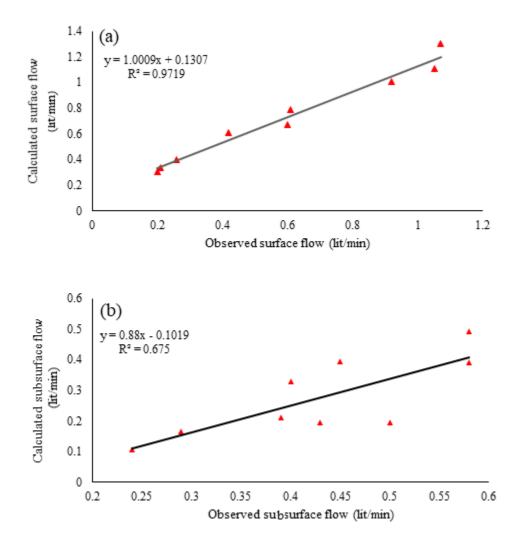
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 Table 4. Observed and calculated SSF and SF for a clay loam soil

SF(PE) is SF peak error, SFF(PE) is SFF peak error

Rainfall	Slope	Observe	ed (1 min ⁻¹)	Calculated	(1 min^{-1})		
mmhr ⁻¹	Degrees	SSF	SF	SSF	SF	SF(PE)	SFF(PE)
15.63	3	0.5	0.200	0.195	0.305	0.53	0.61
15.63	6	0.29	0.210	0.165	0.335	0.60	0.43
15.63	9	0.24	0.259	0.105	0.395	0.53	0.56
31.3	3	0.58	0.420	0.390	0.610	0.45	0.33
31.3	6	0.4	0.600	0.330	0.670	0.12	0.18
31.3	9	0.39	0.610	0.210	0.790	0.30	0.46
46.9	3	0.58	0.920	0.490	1.000	0.09	0.16
46.9	6	0.45	1.050	0.390	1.110	0.06	0.13
46.9	9	0.43	1.070	0.190	1.300	0.21	0.56

Figure 6 illustrates the estimated versus observed SF and SSF values. The correlation coefficient of predicted SF in this experiment was 0.972, which is good, and the correlation coefficient of predicted SSF was 0.675, which is acceptable.



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Fig. 6. Correlation between a) observed and calculated SSF and b) observed and calculated SFfor a clay loam soil.

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Surface flow measurement is recorded more accurately in laboratory models, but there is more error in measuring subsurface flow due to soil moisture storage and the influence of other factors, and this has reduced the correlation coefficient in subsurface flow.

Moving from laboratory scale to real catchment scale, usually the lack of observed 324 SSF data is the main obstacle to validating SSF forecasting models. Available SSF 325 data in the hillslope dimension are generally used to validate models (Tiefan et al. 326 2005; Brown et al. 1999; Ameli et al. 2015; Fariborzi et al. 2019). For a more 327 accurate validation of the method proposed in this study, rainfall simulator data 328 reported by Morbidelli et al. (2015) were used (Table 5). Their data were obtained 329 used a soil box measuring $152 \times 122 \times 78$ cm in length, width, and thickness, 330 respectively, and containing loamy soil. The slope of the box was adjustable from 0 331 to 10 degrees. Table 5 shows the observed SF and SSF values for different rainfall 332 rates and two slopes, 5 and 10 degrees. For instance, the calibrated values for RC 333 were 0.53 and 0.65 for slopes of 5 and 10 degrees respectively. The SF and SSF 334 values were also calculated using our method (Eq. 8) (Table 6) and the results were 335 compared with observed maximum SF and SFF reported by Morbidelli et al. (2015). 336 The correlation coefficient of SF prediction values in this case was 0.93, which is a 337 very good value, and that of SFF prediction values was 0.64, which is acceptable 338 (Fig. 7). 339

- Table 5. Observed and calculated surface flow (SF) and subsurface flow (SSF) (observed) in tests
 in a physical model (data from Morbidelli et al. 2015)
- 343

SF(PE) is SF peak error, SFF(PE) is SFF peak error

Slope (degrees)	Total flow	Observed SF	Observed SSF	Calculated SF	Calculated SSF		
	$(mm h^{-1})$	$(mm h^{-1})$	$(mm h^{-1})$	$(mm h^{-1})$	$(mm h^{-1})$	SF(PE)	SFF(PE)
5	6.62	4.04	2.58	3.97	2.65	0.02	0.03
5	9.22	5.92	3.3	4.89	4.33	0.17	0.31
5	9.59	5.37	4.22	5.75	3.84	0.07	0.09
5	10.55	5.98	4.57	5.59	4.96	0.07	0.09
5	11.59	7.67	3.92	6.76	4.83	0.12	0.23
10	6.05	3.88	2.17	3.93	2.12	0.01	0.02
10	8.93	6.94	1.99	5.80	3.13	0.16	0.57
10	10.9	8.93	1.97	8.72	2.18	0.02	0.11
10	12.26	10.34	1.92	9.81	2.45	0.05	0.28

10	11.71	9.73	1.98	9.37	2.34	0.02	0.03
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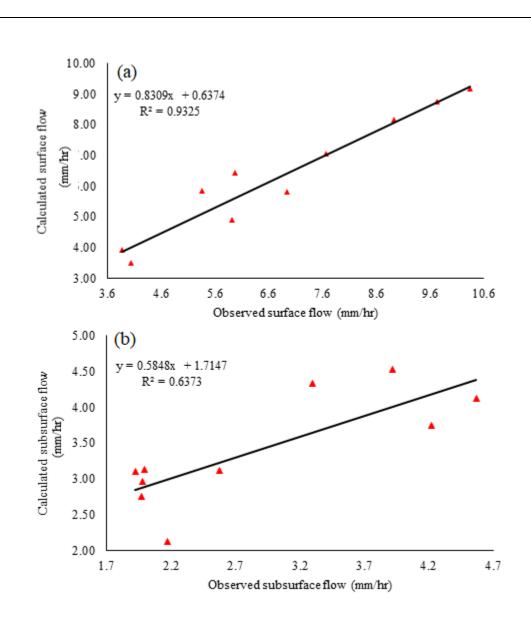


Fig. 7. Correlation between a) observed and calculated surface flow and b) observed and calculated
surface flow for clay loam soil, based on data in Morbidelli et al. (2015).

349

The results showed that the correlation coefficient for predicted SF in this experiment was greater than 0.97, which is very good, and that for predicted SSF was 0.635, which is acceptable.

4.3 Predicting the SF and SSF hydrograph at catchment scale

Separation of SF hydrograph and SSF hydrograph from observed flood hydrograph 354 is very important for hydrologists. In the previous sections, we focused on the 355 separation of SF and SF peaks of hillslopes in the laboratory, but in this section, the 356 proposed RC method was applied to evaluate the separation method in the catchment 357 scale. For further model verification, data on peak SF and SSF from the Heng-Chi 358 and San-Hsia catchments in northern Taiwan were used (Fig. 8 and Table 6). The 359 360 Heng-Chi catchment ranges in elevation from 20 m at the outlet to 970 m, and occupies an area of 53.23 km², which is covered by forest (70%), cultivated land 361 (25%), and urban area (5%). The San-Hsia catchment is similar, with elevation 362 ranging between 30 and 1770 m and area 125.88 km², with 75% forest, 20% 363 cultivated land, and 5% urban land use. 364

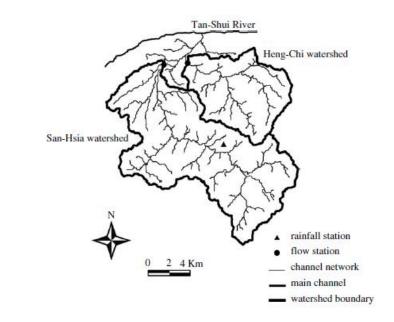


Fig. 8. Location of the Heng-Chi and San-Hsia catchments in Taiwan (after Chang and Lee 2008).

367 **4.3.1. Subsurface GIUH model**

365

Chang and Lee (2005) revised the GIUH model to estimate SSF in catchments. In this model, the Darcy's law was adopted to estimate the runoff travel time in subsurface-flow regions. Based on the Horton-Strahler ordering law, any catchment of order Ω can be divided into a series of runoff states. The catchment hydrologic response can be considered to be functions of the runoff path probabilities and runoff travel time probabilities in different runoff states (Rodriguez-Iturbe and Valdes, 1979).

Let x_{o_i} denotes the *i*th-order overland-flow regions in catchment, denotes the x_{sub_i} *i*th-order subsurface-flow regions, and x_i denotes the *i*th-order channels, in which $i = 1, 2, ..., \Omega$. Ω is maximum order of catchment. The subsurface IUH can be expressed analytically by (Lee and Chang, 2005):

379
$$u_{sub}(t) = \sum_{w_{sub} \in W_{sub}} [f_{x_{sub}}(t) * f_{x_i}(t) * f_{x_j}(t) * \dots * f_{x_{\Omega}}(t)]_{w_{sub}} P(w_{sub})$$
(13)

where $u_{sub}(t)$ is subsurface-flow IUH, W_{sub} is the subsurface flow path space given as $W_{sub} = \langle x_{sub_i}, x_i, x_j, ..., x_{\Omega} \rangle$, $P(w_{sub})$ are the probabilities of a raindrop adopting a subsurface flow path of w_{sub} .

In this study, the subsurface GIUH values for these two case study catchments were

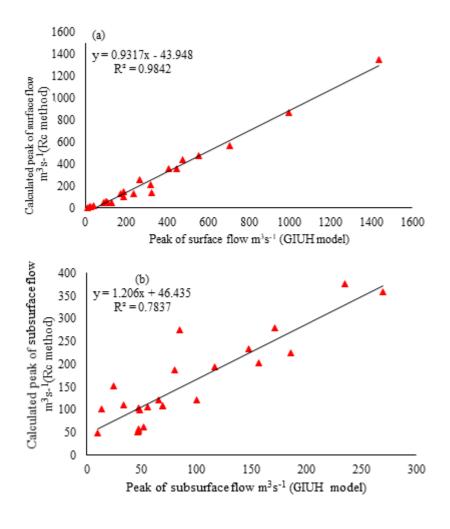
compared with results obtained using the RC-based model developed in this study.

Table 6. Recorded SF, peak SF and SSF estimates obtained using the GIUH method, for the Heng-Chi and San-Hsia catchments in northern Taiwan (source: Chang and Lee 2008)

		DRH (GIUH method)		SF (GIUH method)		SSF (GIUH method)	
Catchment	Event date	$Q_p(m^3s^{-1})$	$T_p(hr)$	$Q_p (m^3 s^{-1})$ (hr)	T_p	$Q_{p} (m^{3}s^{-1})$	$T_p(hr)$
	08/16/1984	157.8	67	88.8	66	69.3	68
	09/16/1985	587.7	8	553.7	7	34.3	9
	09/17/1986	455.9	41	407.9	40	48.2	43
	07/27/1987	161.5	7	105.5	7	55.9	8
	09/08/1987	318.0	36	238	36	80.0	37
	08/18/1990	486.4	32	476.4	31	10.1	32
Heng-Chi	06/05/1993	173.4	11	107	11	66.3	12
	07/10/1994	57.0	12	11	11	45.9	12
	07/30/1996	242.1	30	173	29	68.6	34
	06/22/1997	70.6	5	24	6	47.2	7
	06/18/1999	153.6	4	107	4	47.1	5
	08/22/2000	72.5	18	21	18	52.1	20
	10/31/2000	309.8	18	263	17	46.5	24
San-Hsia	06/03/1983	243.6	14	127	14	116.5	15

(09/16/1985	1449.1	8	1435	8	13.4	9
(07/14/1987	142.1	12	42	11	99.6	15
(07/27/1987	336.5	8	189	8	148.0	9
(08/18/1990	1022.1	31	997	31	25.1	31
(08/30/1990	941.8	16	707	15	234.7	16
(09/07/1990	410.0	25	325	25	85.2	26
(07/10/1994	255.8	13	99	12	156.7	13
1	10/09/1994	487.8	20	318	20	170.7	24
(07/30/1996	717.0	31	447	30	270.1	34
(08/17/1997	374.6	32	189	32	185.7	34

In Figure 9a, the SF values for the catchments calculated using Eq. 8 are compared with the values obtained by the GIUH method in the two catchments recorded by Chang and Lee (2008) (column 3, Table 6). The correlation coefficient was 0.98, which is very good. Figure 9b also shows the SSF values for the two catchments calculated using Eq. 8 and those estimated by the GIUH model. The correlation coefficient in this case was lower, 0.78.



406

Fig. 9. Comparison of a) peak SF and b) peak SFF in Heng-Chi and San-Hsia catchments
calculated by the RC method developed in this study and by Chang and Lee (2008) using the GIUH
model.

Furthermore, the SSF and SF hydrographs for Heng-Chi (July 1996) and San-Hsia
(August 1997) calculated using the GIUH model were compared with those
produced using the RC method (Fig. 10). To evaluate model fitness for this purpose,
coefficient of efficiency (CE) and relative error in peak (REP) were calculated
(Chang and Lee 2008):

415
$$CE = 1 - \frac{\sum_{t=1}^{n} [Q_o - Q_s]^2}{\sum_{t=1}^{n} [Q_o - \overline{Q_o}]^2}$$
 (14)

416
$$REP = 100 \times [Q_{p_s} - Q_{p_a}]/Q_{p_a}$$
 (15)

where Q_o is observed discharge at time *t*; Q_s is simulated discharge at time *t*; $\overline{Q_o}$ is average observed discharge during a storm event; *n* is number of discharge records during the storm event; Q_{p_s} is peak discharge of the simulated hydrograph; and Q_{p_o}

421 is observed peak discharge.

The value of CE is between 0 and 1 and CE values above 0.8 are acceptable. The CE was found to be 0.8 and 0.81 for SF, and 0.7 and 0.81 for SSF, in the Heng-Chi and San-Hsia catchment, respectively. Peak error in Heng-Chi was 8% and 80% for SF and SSF, respectively, while it was %18 and %17, respectively, in San-Hsia catchment. Thus, peak error in SSF in Heng-Chi was unacceptably large.

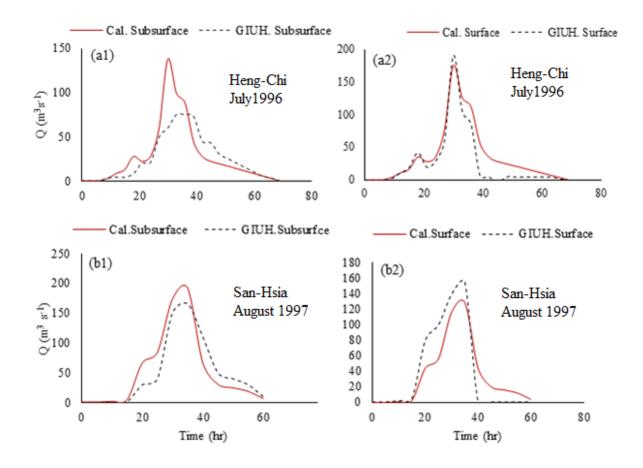


Fig. 10. Hydrographs calculated by the RC method and simulated by the GIUH model for: 1) SFand 2) SSF in a) Heng-Chi catchment, July 1996 and b) San-Hsia catchment, August 1997.

430 5. Conclusions

Separation of surface runoff and subsurface runoff from observed data in catchments 431 is difficult, due to the hydrological complexities of runoff. In many permeable 432 catchments with high vegetation cover, subsurface runoff is of great importance. In 433 this study, we applied the concept of runoff coefficient (RC) to devise a simple and 434 practical method for separating surface and subsurface flow in direct runoff from 435 hillslopes or catchments. The accuracy of the method is directly dependent on the 436 accuracy of RC values. We investigated the effect of slope, rainfall intensity, and 437 soil type on RC. Using the SCS-CN infiltration method, we also tested the effect of 438 rainfall intensity on RC for soils with different curve number. 439

To verify the method, the results were compared with those of laboratory tests on 440 different soils using a rainfall simulator and an adjustable soil box, and with values 441 predicted by the geomorphological instantaneous unit hydrograph (GIUH) model for 442 two watersheds, Heng-Chi and San-Hsia, in Taiwan. Comparison with laboratory 443 values revealed that our RC-based method accurately predicted peak surface flow 444 and subsurface flow in different soils, with correlation coefficient (CE) 0.93 and 445 0.65, respectively. Comparison with surface and subsurface runoff hydrographs for 446 Heng-Chi and San-Hsia catchments, obtained using the GIUH model. Based on 447 results, the CE was found to be 0.8 and 0.81 for SF, and 0.7 and 0.81 for SSF, in the 448 Heng-Chi and San-Hsia catchment, respectively. Peak error in Heng-Chi was 8% 449 and 80% for SF and SSF, respectively, while it was %18 and %17, respectively, in 450 San-Hsia catchment. Thus, peak error in SSF in Heng-Chi was unacceptably large. 451 Thus if RC can be calculated accurately, our method can successfully separate 452 surface and subsurface flow in total runoff. 453

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457

458 **11. References**

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