

# APPLICATION OF PRINCIPAL COMPONENT ANALYSIS IN GROUPING GEOMORPHIC PARAMETERS OF UTTELA WATERSHED FOR HYDROLOGICAL MODELING

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### Abstract

Geomorphological characteristics of a watershed are very commonly used for developing the regional hydrological models for solving the various hydrological problems of the ungauged watersheds or inadequate data situations. So, in the present study GIS technique has been used to determine the geomorphological parameters of Uttala nala watershed which is tributary of Son river, located in Sahadol district of Madhya Pradesh. Many of the geomorphic parameters are known to be strongly correlated. There is considerable amount of redundancy in the array of geomorphic parameters currently in use. The screening of such large number of interrelated variables for their underlying dimensions is best achieved by multivariate statistical techniques of the principal component Results of principal analysis analysis. of eleven geomorphometric parameters clearly reveals that some of the parameters are strongly correlated with the components but stream frequency does not show correlation with any of the component. So it has been screened out of analysis. The principal component loading matrix obtained using correlation matrix of ten parameters reveals that first three components together account for 93.71 per cent of the total explained variance. Therefore, principal component loading is applied in order to get better correlation and clearly group the parameters in physically significant components. Based on the properties of geomorphic parameters, three components were defined as slope or steepness, drainage and shape component. One parameter each from the significant component may form a set of independent parameter at a time in modeling the hydrological responses such as runoff and sediment yield from small watersheds.

Keywords: Geomorphic Parameters, Principal Component Analysis, GIS

# Introduction

The development of land and water resources on a sustainable basis without deterioration and with constant increase in

productivity is the mainstay of mankind. Ironically, adequate emphasis has not been paid to conserve, develop and judiciously utilize these resources in many parts of country. This is evident from the fact that 175 million ha of land in India constituting 53% of its total geographical area, suffers

from such deleterious effects. It has been estimated that about 16.4 tones/ha of soil is detached annually in our country because of destruction (Singh, 2000). Apart from serious losses of production, the immediate action of the problem is seen through excessive and premature siltation of multipurpose reservoirs constructed with huge investments. Hence, it becomes necessary to protect their lives as they are considered one of the important resources for national economy. To overcome aforesaid conditions at the national level, an integrated watershed management considering crop production, soil and water conservation and management, reclamation of waste land and degraded lands is essential to increase overall efficiency of the watershed.

Soil and water conservation measures on watershed basis can play an important role in formulating long term comprehensive land and water management strategies. There may be various considerations for the implementation of management programmes in the few sub-watersheds only. It is always better to start management measures from the most critical sub-watershed. Sediment yield from a catchment is one of the criterions to find most critical sub-watershed to soil erosion. However, this criterion requires for assessing continuous monitoring of sediment samples at the catchment outlet. Such data are hardly available in India for small watersheds. Although the sediment yield from large basins can be obtained from such observation, it is not possible to ascertain the vulnerability to soil erosion of small watersheds within a basin. In the absence of sediment yield data morphometric parameters may be helpful in assessing most critical sub-watershed. Morphometry is the measurement and mathematical analysis of the configuration of the earth's surface, shape and dimensions of its landforms (Clarke, 1966). This analysis can be achieved through measurement of linear, aerial and relief aspects of basin and slope contributions.



Morphometric analysis of a basin can be better achieved through a latest technology like GIS as conventional measurement of these parameters is laborious and cumbersome. Many researchers have demonstrated the potential of GIS technique for morphometric analysis of watershed (Shrimali et al., 2001; Thakker et al., 2007; Sharma et.al, 2010).

The method of quantitative analysis of watershed was developed by Horton (1945) and was further modified by Strahler (1964). Sufficient works on the quantitative analysis of geomorphological parameters of watersheds have been done in India and abroad (Ghose et al., 1969). However, a very little work on the interrelationship of morphological been out. То determine parameters has carried interrelationship of these geomorphological parameters is very important to develop sediment yield regression models (Hydrological modeling). Statistical methods are applied in a variety of fields in hydrological research. Factor analysis is useful for interpreting morphometric parameters and relating the same to specific hydrological processes. Multivariate analysis is simply a collection of procedures for analyzing the associations between two or more sets of data that have been collected on each object in one or more samples of object. Synder (1962) introduced some solutions, possibilities of multivariate statistics in hydrological modeling. Wong (1979) utilized a multivariate statistical technique component analysis in analyzing the effects of twelve basins and climatological parameters. Wallis (1965) in discussion of multivariate statistical methods in hydrology recommends, for multifactor hydrological problems, the use of principle component analysis with varimax rotation of the factor weight matrix. Hann and Allen (1972), Decoursey and Deal (1974) and have also demonstrated the use of multiple regression analysis for development of hydrological predictions equations involving geomorphic parameters. Mishra and Satyanarayana (1988) carried out principal component analysis with varimax rotation on ten geomorphic parameters at Damoder valley catchment of India and concluded that nine parameters could be significantly grouped into three components. Singh et.al (2009) carried out principal component analysis to thirteen geomorphic parameters collected for sixteen watersheds of Chambal catchment of Rajasthan. The parameters are grouped into three components. Therefore, in this study an attempt has been made to determine geomorphological parameters and to study the intercoorelationship (multicollinearity) among variables in order to screen out the less significant variables out of the analysis and to arrange the remaining into physically significant groups by applying principal component analysis for better interpretability.

# Materials and methods

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Uttala Nala watershed, a tributary of Son river falling in Sahadol district of Madhya Pradesh lying between  $23^{\circ}8'21.6''$  to  $23^{\circ}13'19.2''$  N latitude and  $89^{\circ}19'54.7''$  to  $81^{\circ}28'48.76''$  E longitude has been selected as the study area. The elevation of the watershed range from 460 to 800 m above mean sea level. The location map of the study watershed is shown in Fig. 1. The total area of the watershed is 4763 ha. The watershed receives an average annual rainfall of 1099.9 mm and more than 80 % of rainfall is received during the monsoon season (June to October). The minimum and maximum temperature varies in the range of  $2.6^{\circ}$ C to  $46^{\circ}$ C.



Figure 1. Location map of Uttala nala watershed

A. Geomorphometric Parameters The watershed boundary of the study area was delineated using Survey of India Toposheet (64 E/8) on 1:50000 scale. The delineated watershed boundary was further sub divided into sub-watersheds. The input parameters for present study such as area, perimeter, stream order, number of streams, stream length, elevation and basin length were derived from digitized stream network and contour map in GIS environment. The morphometric parameters for the delineated watershed area were calculated using formula suggested by Horton (1945), Strahler (1964), Schumm (1956) and Miller (1953) given in Table 1.

Table 1.	Formula for Computation of Morphometric					
Parameters						

i uruneters							
Morphometric Parameters	Formula	Refrence					
Bifurcation	$R_b = N_u / N_{u+1}$	Schumn					
Ratio (R <sub>b</sub> )	Where, R <sub>b</sub> = Bifurcation Ratio	(1956)					
	N <sub>u</sub> = Total number of stream of						
	segment of order u						
	$N_{u+1} = Total$ number of stream of						
	segment of next higher order						



Drainage	$D_d = L_u / A$	Horton (1945)
density (D <sub>d</sub> )	Where, $D_d = Drainage$ density	
	$L_u = Total stream length of$	
	order u	
	A = Area of basin	
Texture ratio	$\mathbf{T} = \mathbf{N}_1 / \mathbf{P}$	Horton (1945)
(T)	Where, $T = Texture ratio$	
	$N_1 = Total$ number of streams of	
	first order	
	$\mathbf{P} = \mathbf{Perimeter}$	
Stream	$F_{n} = N_{n}/A$	Horton (1945)
Frequency (F.,)	Where, $N_{y} = \text{Total number of streams}$	
riequency (ru)	of all order	
	A = Area of basin	
Circulatory	$R = 4 \prod A / P^2$	Miller (1953)
ratio (R)	Where $\mathbf{R} = \mathbf{Circulatory ratio}$	Willer (1955)
radio (R <sub>c</sub> )	A = Area of basin	
	P = Perimeter	
Form factor (P.)	$\mathbf{P}_{1} = \mathbf{A} / \mathbf{L}_{2}^{2}$	Horton (1045)
Torin factor (Rf)	$\mathbf{R}_{\mathbf{f}} = \mathbf{A}_{\mathbf{f}} \mathbf{L}_{\mathbf{b}}$ Where $\mathbf{P}_{\mathbf{f}} = \mathbf{Form}$ factor	11011011 (1945)
	where, $R_1 = 10$ m factor	
	A = Area or basin	
Elemention notio	$L_b = \text{Length of basin}$ $P = (2/L_b) * (A/H)^{0.5}$	Sahuma
Elongation ratio	$\mathbf{R}_{e} = (2/\mathbf{L}_{b}) \cdot (\mathbf{A}/\mathbf{\Pi})$	
$(\mathbf{K}_{e})$	where, $R_e = Form factor$	(1950)
	A = Area of basin	
T d C	$L_b = Length of basin$	
Length of over	$L_0 = \frac{1}{2} D_d$	
land flow $(L_o)$	Where, $L_0$ = Length of over land flow	
	$D_d = Drainage density$	
Relative relief	R <sub>r</sub> = H/P	
$(R_r)$	Where, $R_r$ = Relative relief	
	H=Maximum watershed relief	
	P= Perimeter of basin	
Relief ratio (R <sub>h</sub> )	$R_{h}=H/L_{b}$	
	Where, $R_h$ = Relief ratio	
	H=Maximum watershed relief	
	$L_b = Length of basin$	
Ruggedness	$R_N = H^* D_d$	
number (R <sub>N</sub> )	Where, R <sub>N</sub> = Ruggedness number	
	H=Maximum watershed relief	
	$D_d = Drainage density$	
	$D_d = Dramage density$	

B. Principal Component Analysis The method of principal components or component analysis is based upon the early work of Pearson with specific adaption to principal component suggested by the Hotelling (1933). analysis The geomorphometric parameters are usually many times correlated. The correlation indicates that some of the information contained in one variable is also contained in some of the other remaining variables. More specifically, the first principal component is that linear combination of the original variables which contributes a maximum to their total variance; the second principal component, uncorrelated with the first, contributes a maximum to the residual variance, and so on until the total variance is analyzed. Since the method is so dependent on the total variance of the original variables, it is most suitable when all the variables are measured in the same units. Hence, it is customary to express the variables in standard form, i.e., to select the unit of measurement for each variables so that its International Journal of Remote Sensing & Geoscience (IJRSG) www.ijrsg.com

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sample variance is one. Then, the analysis is made on the
correlation matrix, with the total variance equal to n. The
objectives are achieved in three steps:
Step 1 Calculate the correlation matrix, R
Step 2 Calculate the principal component loading matrix by
principal component analysis.
Step 3 In the principal component (PC) Loading matrix Eigen
values greater than one indicates significant PC
loading
Figen value indicated how well each of the identified factors
fit the data from all the geomorphic parameters on all the
principal components
principal components.
C. Correlation Matrix The inter-correlation matrix of the
geomorphic parameters is obtained by using the following
procedure:
1. The parameters are standardized
$X = (x_{ii} - x_i) / S_i $ (1)
Where.
X denotes the matrix of standardized parameters
$x_{ii}$ ith observation on ith parameters
$i_{j}$ in order values of $j$ and $j$
<i>i</i> 1 P (Number of observation)
$r_{i}$ mean of the <i>i</i> th parameters
S standered deviation of the <i>i</i> th parameters
The correlation matrix of parameters is the minor
product moment of the standardized predictor
measures divided by N and is given by
$\mathbf{P} = (\mathbf{v}^2 \times \mathbf{v}) / \mathbf{N} $ (2)
$\mathbf{R} = (\mathbf{X} \wedge \mathbf{X}) / \mathbf{I} \mathbf{N} \qquad (2)$ where $\mathbf{x}^2$ denotes the transmiss of the star densities $\mathbf{J}$
where, x denotes the transpose of the standardized
matrix of predictor parameters

*D. Principal Component Loading Matrix* The principal component loading matrix which reflects how much a particular parameter is correlated with different factors, is obtained by premultiplying the characteristics vector with square root of the characteristics values of the correlation matrix.

Thus, 
$$A = Q \times D^{0.5}$$
 (3)

where

A principal component loading matrix,

- Q characteristics vector of the correlation matrix
- *D* characteristics value of the correlation matrix

# Results and discussion

The drainage and sub-watershed map of Uttala Nala watershed is presented in Fig 2.





Figure 2. Stream network with sub watershed boundaries

Morphometric parameters of sub watersheds were calculated in GIS environment and are presented in Table 2 and computed geomorphometric parameters are presented in Table 3.

Area and perimeter of sub watersheds varies from 1.33 to  $12.39 \text{ km}^2$  and 5.75 to 15.94 km respectively. After analysis of drainage map it was found that Uttala Nala watershed is of 5<sup>th</sup> order type and drainage pattern is dendritic. The bifurcation ratio ( $R_b$ ) reflects geological and tectonic characteristics of watershed area were calculated for all eight sub-watersheds and are given in Table 3. These values are more or less normal in the sub watersheds 1, 4, 5, 6, and 7 as they range between 1 and 3 (Horton, 1945).

The important parameters that describe the shape of the basin form factor, circulatory ratio and elongation ratio were computed for all eight sub watersheds (Table 3). In the present case sub watersheds have lower  $R_f$  value (0.14 to 0.36) indicating them to be elongated in shape and suggesting flatter peak flow for longer duration. Flood flows of such elongated basins are easier to manage than those of circular basin. In the present case circulatory ratios for sub watersheds are 0.31 to 0.65, indicating that the area is characterized by high relief and the drainage system is structurally controlled. The value of elongation ratio (Re) for sub watersheds to be elongated with high relief and steep slopes.

Relief ratio ( $R_h$ ), and Ruggedness number ( $R_N$ ) values for all the sub watersheds are given in Table 3. Sub watersheds with high  $R_h$  are considered critical from erosion point of view and should be provided with suitable soil and water conservation measures. The Ruggedness number ranges from 0.82 to 2.64 for different sub watersheds. The sub watershed 2 has an overall roughness or unevenness.

Drainage density  $(D_d)$ , Stream frequency  $(F_u)$  and Texture ratio (T) values are computed for all the sub watersheds and are given in Table 3. Drainage frequency values of all the sub watersheds have close correlation with Drainage density

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indicating the increase in stream population with respect to increase in Drainage density. In the present study low value of  $D_d$  for sub watershed 7 indicates that it has highly resistant, impermeable sub soil material with dense vegetation cover and low relief. Texture ratio values for sub watersheds ranges from 0.91 to 3.86.

Length of over land flow values are between 0.30 and 1.33 for all the sub-watersheds. However, compactness coefficient values vary from 1.53 to 3.25 for sub-watershed 7 and sub-watershed 1 respectively.

The correlation matrix (Table 4) of eleven geomorphometric parameters reveals that strong correlations (correlation coefficient more than 0.9) exists between drainage density and length of over land flow, between stream frequency and form factor, between circulatory ratio and elongation ratio, between relief ratio and ruggedness number and relative relief. Also, good correlation (correlation coefficient more than 0.75) exists bifurcation ratio and texture ratio, length of over land flow, ruggedness number, between drainage density and stream frequency, ruggedness number between form factor and elongation ratio. Some more moderately correlated parameters (correlation coefficient more than 0.60) are bifurcation ratio with drainage density, relief ratio, ruggedness number, stream frequency with circulatory ratio, elongation ratio, length of over land flow, circulatory ratio with form factor, form factor with length of over land flow and length of over land flow with ruggedness number. It is very difficult at this stage to group the parameters into components and attach any physical significance. Hence, in the next, the principal component analysis has been applied. The correlation matrix is subjected to the principal component analysis.

The principal component loading matrix obtained from correlation matrix (Table 5) reveals that the first three components whose Eigen values are greater than one, together account for about 89.277 per cent of the total explained variance. The first component is strongly correlated (loading of more than 0.80) with circulatory ratio and elongation ratio and good correlation (loading of more than 0.70) with form factor, which may be termed as shape component. The second component is strongly correlated with bifurcation ratio, texture ratio and length of over land flow and moderately correlated with (loading of more than 0.60) drainage density which may be termed as drainage component. Third component is strongly correlated with relief ratio, relative relief and ruggedness number. It is evident from these results that some of the parameters are highly correlated with some of the component but the parameter texture ratio could not be grouped with any of the component because of its poor correlation with them.



#### Table 2. Sub watershed wise input Geomorphometric parameters

Sub watershed No.	Area (km²)	Perimeter (km)	Max. Elevation (m)	Min. Elevation (m)	Total Relief (m)	No. of Streams	Max. Length of Watershed (km)	Total Stream Length (km)
1.	2.33	7.19	780	520	260	9	2.92	7.57
2.	5.64	10.42	800	520	280	22	3.98	4.26
3.	12.39	15.90	760	500	260	28	6.95	4.27
4.	5.25	14.54	560	500	60	13	6.14	11.74
5.	4.96	9.66	520	480	40	14	3.77	10.99
6.	1.33	5.75	520	500	20	6	2.55	4.10
7.	8.76	15.94	540	500	40	8	5.13	13.42
8.	6.96	14.61	500	460	40	10	5.38	13.15

#### Table 3. Sub-watershed wise computed Geomorphometric parameters

Sub watershed No.	R <sub>b</sub>	$\mathbf{D}_{\mathbf{d}}$	Т	$\mathbf{F}_{\mathbf{u}}$	R <sub>c</sub>	$\mathbf{R}_{\mathbf{f}}$	Re	Lo	R <sub>r</sub>	$\mathbf{R}_{\mathbf{h}}$	R <sub>N</sub>
1	3.25	0.83	3.86	0.57	0.27	1.33	0.59	0.60	0.036	0.089	0.846
2	3.31	1.63	3.90	0.65	0.36	1.24	0.67	0.30	0.026	0.070	0.931
3	3.01	1.51	2.26	0.62	0.26	1.27	0.57	0.33	0.016	0.037	0.790
4	2.24	0.69	2.48	0.31	0.14	1.79	0.42	0.72	0.004	0.009	0.134
5	2.22	1.04	2.82	0.67	0.35	1.22	0.67	0.48	0.004	0.010	0.088
6	3.09	0.70	5.26	0.51	0.21	1.41	0.51	0.71	0.003	0.007	0.056
7	1.53	0.38	0.91	0.43	0.33	1.52	0.65	1.33	0.002	0.007	0.061
8	1.89	0.55	1.44	0.41	0.24	1.56	0.55	0.99	0.002	0.007	0.075

 Table 4. Intercorrelation matrix of Geomorphometric parameters

	R <sub>b</sub>	$\mathbf{D}_{\mathbf{d}}$	Т	$\mathbf{F}_{\mathbf{u}}$	R <sub>c</sub>	R <sub>f</sub>	R <sub>e</sub>	L	R <sub>h</sub>	R <sub>N</sub>	R <sub>r</sub>
R <sub>b</sub>	1.000	0.682	0.820	0.576	0.011	-0.552	0.004	-0.777	0.731	0.760	0.745
$\mathbf{D}_{\mathbf{d}}$	0.682	1.000	0.320	0.754	0.379	-0.684	0.343	-0.900	0.540	0.768	0.543
Т	0.820	0.320	1.000	0.395	-0.097	-0.377	-0.110	-0.531	0.392	0.293	0.396
$\mathbf{F}_{\mathbf{u}}$	0.576	0.754	0.395	1.000	0.715	-0.988	0.716	-0.701	0.509	0.577	0.509
R <sub>c</sub>	0.011	0.379	-0.097	0.715	1.000	-0.739	0.996	-0.130	0.309	0.300	0.277
$\mathbf{R}_{\mathbf{f}}$	-0.552	-0.684	-0.377	-0.988	-0.739	1.000	-0.750	0.612	-0.501	-0.555	-0.502
Re	0.004	0.343	-0.110	0.716	0.996	-0.750	1.000	-0.103	0.319	0.297	0.290
Lo	-0.777	-0.900	-0.531	-0.701	-0.130	0.612	-0.103	1.000	-0.520	-0.667	-0.534
$\mathbf{R}_{\mathbf{h}}$	0.731	0.540	0.392	0.509	0.309	-0.501	0.319	-0.520	1.000	0.920	0.998
R <sub>N</sub>	0.760	0.768	0.293	0.577	0.300	-0.555	0.297	-0.667	0.920	1.000	0.926
R <sub>r</sub>	0.745	0.543	0.396	0.509	0.277	-0.502	0.290	-0.534	0.998	0.926	1.000



 
 Table 5. Principal component loading matrix of eleven geomorphic parameters

Denometers	Components					
rarameters	1	2	3			
R <sub>b</sub>	-0.013	0.806	0.557			
$\mathbf{D}_{\mathbf{d}}$	0.419	0.663	0.380			
Т	-0.130	0.813	0.154			
$\mathbf{F}_{\mathbf{u}}$	0.774	0.573	0.225			
R <sub>c</sub>	0.969	-0.083	0.138			
$\mathbf{R}_{\mathbf{f}}$	-0.795	-0.513	-0.225			
Re	0.971	-0.108	0.151			
Lo	-0.195	-0.842	-0.317			
$\mathbf{R}_{\mathbf{h}}$	0.193	0.252	0.936			
$\mathbf{R}_{\mathbf{N}}$	0.237	0.359	0.874			
$\mathbf{R}_{\mathrm{r}}$	0.171	0.270	0.937			
Eigen Value	5.477	2.407	1.157			

In order to screen out parameters having less significance in explaining the component variance, the parameter stream frequency is screened out from analysis. Then correlation matrix and principal component matrix are obtained for ten parameters.

The principal component loading matrix obtained using the correlation matrix of ten parameters (Table 6) reveals that the first three components now together accounts for 93.71 per cent of the total explained variance showing an increase of about 4.43 per cent.

Danamatana	Components					
r ar anneter s	1	2	3			
R <sub>b</sub>	0.554	0.807	-0.048			
$\mathbf{D}_{\mathbf{d}}$	0.361	0800	0.403			
Т	0.155	0.802	-0.166			
R <sub>c</sub>	0.135	-0.038	0.976			
$\mathbf{R}_{\mathbf{f}}$	-0.228	-0.538	-0.801			
Re	0.151	-0.065	0.975			
Lo	-0.305	-0.856	-0.161			
$\mathbf{R}_{\mathbf{h}}$	0.937	0.264	0.181			
R <sub>N</sub>	0.866	0.381	0.226			
R <sub>r</sub>	0.938	0.280	0.157			
Eigen value	6.475	2.478	1.215			

 Table 6. Principal component loading matrix of ten finally

 screened out
 geomorphic parameters

The principal component loading here also improved considerably in almost all significant parameters. The relief ratio and relative relief have strong correlation (loadings of more than 0.90) with the first component. The ruggedness number has good correlation (loadings of more than 0.80) with first component. The bifurcation ratio, drainage density, texture ratio and length of over land flow have the good correlation with the second component. The circulatory ratio and elongation ratio have strong correlation with third component. The form factor has good correlation with third component.

It is observed that the first component is strongly correlated with relief ratio and relative relief and good correlation with ruggedness number which are grouped under slope or steepness component. The second component has good correlation with bifurcation ratio, drainage density, texture ratio and length of over land flow of watershed and is termed as drainage component. The third component has strong correlation with circulatory ratio and elongation ratio and good correlation with form factor hence is called as shape component.

It can be seen how useful the principal component analysis have been in screening out the parameters or variables of least significance and is regrouping the remaining variables into the physically significant factors. Multiple regression technique can then be applied in modeling the hydrologic response such as runoff and sediment yields from the watersheds. One parameter each from significant component may form a set of independent parameters at a time in modeling the said hydrologic response.

# Conclusion

The quantitative geomorphometric analysis was carried out in eight sub watersheds of Uttala river watershed, using GIS technique for determining the various areal, relief and linear aspects of the watershed. The conventional methods of morphometric analysis are time consuming, tiresome and error prone, while use of GIS technique allows for more reliable and accurate estimation of similar parameters of watersheds. The morphometric analysis of different sub watersheds shows their relative characteristics with respect to hydrologic response of the watershed. The correlation matrix of eleven geomorphometric parameters revealed that strong correlation exist between drainage density and length of over land flow, between stream frequency and form factor, between circulatory ratio and elongation ratio, between relief ratio and ruggedness number and relative relief. The principal component loading matrix obtained from correlation matrix reveals that the first three components, whose Eigen values are greater than one, together accounts for 89.277 per cent of the total explained variance. Based on the results of the principal component analysis, first component is strongly correlated with circulatory ratio and elongation ratio. The second component is strongly correlated with bifurcation ratio, texture ratio and length of over land flow. However, the third component is strongly correlated with relief ratio, relative relief ratio and ruggedness number. The stream frequency



could not be grouped with any of the component because of its poor correlation with them. After screening out the stream frequency the principal component loading matrix of ten parameters indicate that first three component together accounts for 93.71 per cent of the total explained variance. Based on the properties of geomorphometric parameters, three principal components were defined as steepness drainage and shape components. Moreover, it is concluded that in modeling the hydrologic responses such as runoff and sediment yield from small watersheds, the principal component analysis is good tool for screening out the insignificant parameters from the analysis.

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