

Determination of Topographic Factors to Initiate Debris Flow Using Statistical Analysis

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Abstract—This study used GIS to determine 19 topographic indexes, four geologic indexes, and two rainfall data derived indexes of the Laonun River Basin, in southern Taiwan. The four topographic factors, including the effective area of basin, elongation ratio of basin, relief energy, and relief volume, were selected among the 19 topographic indexes using the SPSS multi-variable statistical analysis and the principal components analysis. The four topographic factors combine the four geologic factors and two rainfall factors and were selected for estimating debris flow prone creeks. The ten factors were further applied by the Fisher's Discriminant Analysis and Logistic Regression Analysis to evaluate the potentials of debris flow prone creeks in the basin. There were 13 sub-basins initiated debris flows and 41 non-debris flow sub-basins when Typhoon Morakot hit Taiwan in 2009. The validated results show that the correctness of Fisher's model for the samples is 81.48 % and 92.6 % via the Logistic Regression model. Both models showed acceptable accuracy, and the Logistic model had better accuracy herein. The Logistic Regression Analysis was adapted to evaluate the potential of debris flow sub-basins to assist in developing risk management in the basin.

Index Terms—Debris flow, GIS, principal components analysis, linear regression analysis.

I. INTRODUCTION

Typhoon Morakot hit Taiwan in 2009. The typhoon brought over 2,500 mm of torrential rainfall in mountainous southern Taiwan. The rainfall was equivalent to a year of rain in Taiwan and caused serious debris flow hazards in southern Taiwan. Most of the serious debris flow hazards occurred in the Laonun river watershed. An efficient prediction model for debris flow warning is necessary to prevent hazards during climate change-induced torrential rainfall.

The initiation of debris flow is accompanied by abundant losses of slope material in high discharge of steep slope. Topographic indexes are shown to be important in initiating the debris flow. Researchers have proposed numerous represented field conditions of rainfall, geologic, and topographic factors that attributed to the initiation of debris flow [1]–[3].

Investigation of potential debris flow creeks has been

studied by numerous researchers that collected related topographic and hydrologic factors to initiate debris flow using statistical analysis. Lee [4] studied rainfall threshold for debris flow warning using non-parametric statistics of the Mann–Whitney–Wilcoxon test (MWW test). He extracted four significant factors to initiate debris flow, including mean streambed slope, effective watershed area, landslide ratio, and lithologic characteristic, for the topographic factors analysis and debris flow potential analysis. Wu [5] studied 34 debris flow potential creeks by Multivariate Statistical Analysis to extract main topographic factors using Principal Component Analysis for watershed area, mean slope, form factor, river density, landslide area, geologic index, effective cumulative rainfall, and effective rainfall intensity. The eight factors were further analyzed by Fisher's Discriminant Analysis for debris flow potential analysis.

Rainfall is the main factor attributed to debris flows in Taiwan. Numerous debris flows are initiated during the peak rainfall intensity of a rainfall event in Taiwan [6], [7]. Separating antecedent rainfall and a rainfall event is required before estimating the critical rainfall factors that initiate debris flow. The relationships between rainfall intensity and duration, intensity and cumulative rainfall, and intensity and antecedent cumulative rainfall are commonly used climate factors to construct a predictive relationship for debris flow warning [6], [8], [9].

II. STUDY AREA AND METHODOLOGY

A. Study Area

The study area, the Laonun River Basin, is located in Kaoshiun City in southern Taiwan. The Laonun River is a branch of the Gaoping River, the second largest river in Taiwan. Its length is 130 km with a watershed area of 1,370 km² and average gradient of 1/43. Upstream from the watershed is primarily forest with the South Cross-Island Highway cross the valley. Large areas in the middle and downstream areas have been developed for agricultural activities and for hot spring sightseeing. The greater the development of the areas the more vulnerable they are to natural disasters. Its location is shown in Fig. 1.

The Laonun River Basin is located in the south-west area of the Central Mountain Range. Its topography is higher on the north side and lower on the south side, steeper on the east side and flatter on the west side. The main topographic features are piedmont, meander stream channel, valley, terrace, and alluvial fan. Fig. 2 shows sub-basin division and stream network in the basin.

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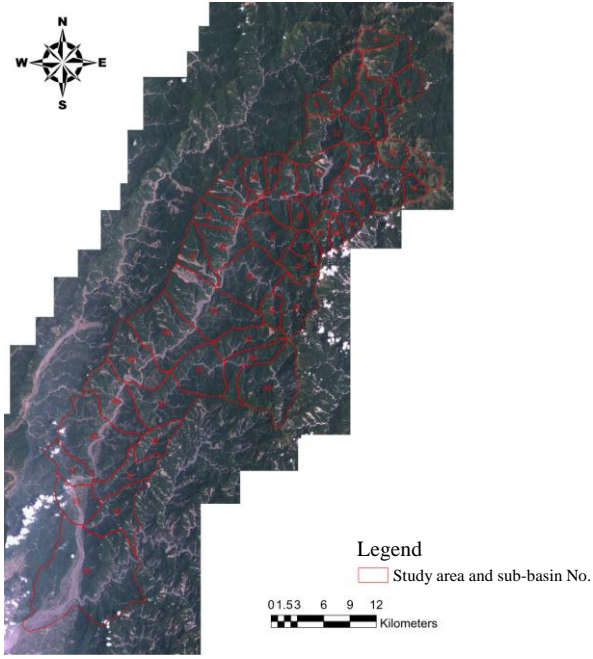


Fig. 1. Study area of the Laonun river basin (satellite image in 2009).

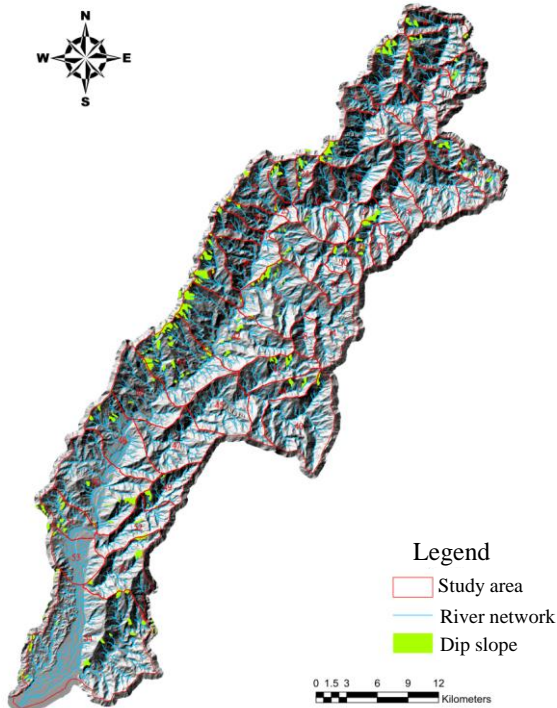


Fig. 2. Basin division, stream network, and dip slope in the study area.

B. Methodology

The study basin is divided into 54 sub-basins to analyze factors that initiate debris flow. Factors that initiate debris flow include climatic, topographic, and geologic conditions. The climatic factors are sourced from rainfall data and topographic factors are calculated from the digital terrain model (DTM) in 40×40 m. The regional characteristics that initiate debris flow are analyzed using statistical analysis to extract factors that trigger debris flows and their potential analysis for the studied Laonun River Basin. The purpose of this study is to establish a precise model to locate potential debris flow creeks for hazard prevention and mitigation. The analysis methodology can be divided into four parts:

1) Digital database collection: Digital database include

satellite images, DTM, rainfall records, and geology maps are collected for the analysis.

- 2) Debris flow initiation factors database setting: The study using GIS spatial analysis to estimate geologic and topographic factors to initiate debris flow and collection of rainfall records during 2006 to 2009 for climate factors estimation.
- 3) Extraction initiation factors of debris flow: The study using Principal Component Analysis to check if the topographic factors have significant differences and show independence. A rainfall event is defined for estimating the climate factors; then, extraction of the main affecting factors, including geologic, topographic, and climatic factors, to determine potential of debris flow for further statistical analysis.
- 4) Test and validation: Statistical Product and Service Solutions package (SPSS) is used for the multi-variable statistical analysis using Fisher's Discriminant Analysis and Logistic Regression Analysis and cross calibration to set a function for debris flow potential assessment.

TABLE I: DEFINITION OF USED TOPOGRAPHIC FACTORS FOR THE ANALYSIS

| Factors | Symbol | Unit | Definition |
|------------------------|--------|---------------|---|
| Effective basin area | A_s | km^2 | cell area for slope over 15° in the basin |
| Perimeter of basin | P | km | the length of a map line that encloses the catchment area of a drainage basin |
| Basin length | L | km | length in a straight line from the mouth of a stream to the farthest point on the drainage divide of its basin |
| Length of main channel | L_o | km | the length of the longest succession of segments that connect a source to the outlet of the basin |
| Total stream length | L_T | km | summation of main stream's and branches' length in the basin |
| Mean basin width | W | km | the ratio between the area of the basin and the length of the main channel |
| Sinuosity | J | - | the ratio of channel length and down valley path length |
| Number of streams | N | - | total numbers of main and branch rivers in the basin |
| Mean basin elevation | H | m | the mean of difference of maximum and minimum elevation in the basin |
| Relief energy | R_f | m | maximum difference of elevation |
| Relief ratio | R_d | - | drop in elevation between the river's source and the river's mouth divided by the total length of the river |
| Mean basin slope | S | % | the mean of the basin slope |
| Mean river slope | S_t | % | the ratio of drop in elevation of a stream per unit horizontal distance |
| Form factor | F | - | the ratio between the area of the basin and the square of the length of the main channel |
| Compactness | C | - | the ratio between the perimeter of the basin and the diameter of the circle having the same area of the basin |
| Circularity ratio | M | - | the ratio between the area of the basin and the area of the circle having the same perimeter of the basin |
| Elongation ratio | E | - | the ratio between the diameter of the circle having the same area of the basin and the length of the main channel |
| Drainage density | D_s | - | the ratio between the total length of the river network and the basin area |
| Stream frequency | F_s | - | the ratio between the number of the branches and the area of the basin |

III. SPATIAL DATABASE SETTING AND STATISTICAL SCREENING ANALYSIS

Topographic factors of every sub-basin are estimated using spatial analysis in GIS. There are 19 topographic factors in the database including: (1) effective basin area, (2) basin perimeter, (3) basin length, (4) length of main stream, (5) total stream length, (6) mean basin width, (7) sinuosity, (8) number of streams, (9) mean basin elevation, (10) relief energy, (11) relief ratio, (12) mean basin slope, (13) mean river slope, (14) form factor, (15) compactness, (16) circularity ratio, (17) elongation ratio, (18) drainage density, (19) stream frequency (see Table I).

The database of geological factors is sourced from the Central Geological Survey (<http://www.moeacgs.gov.tw/>) and includes landslide and dip slope area in the basin and Formosa-2 satellite images before and after Typhoon Morakot. The digitalized database includes geological indexes (E_d), fault length, landslide ratio, and ratio of dip slope area (see Table II).

Rainfall isohyets distribution during years 2006 to 2009 showed that downstream area had the higher frequency of debris flows. The results show that high cumulative rainfall is one of the attributing factors to initiate debris flow. The effective cumulative rainfall and effective rainfall intensity are chosen as the climate factors for the analysis (see Table II).

TABLE II: DEFINITION OF GEOLOGICAL AND CLIMATIC FACTORS USED FOR THE ANALYSIS

| Factors | Symbol | unit | Definition |
|-------------------------------|--------|-----------------|--|
| Geologic index | E_d | km ² | $E_d = (\sum A_i \times W)/A$ A_i : area of rock type i (km ²) A : sub-basin area (km ²) W : weight (unit-free) |
| Fault length ratio | F_L | km | $F_L = L/P$ L : fault length (km) P : premium of basin (km) |
| Landslide ratio | G | km ² | $G = g/A$ g : landslide area (km ²) A : basin area (km ²) |
| Dip slope area ratio | D_A | km ² | $D_A = D_a/A$ D_a : dip slope area (km ²) A : basin area (km ²) |
| effective cumulative rainfall | R_w | mm | $R_e = \alpha_1 d_1 + \alpha_2 d_2 + \dots + \alpha_{14} d_{14} = \sum_{t=1}^{14} \alpha_t d_t$ $\alpha_t = 0.5^{t/T}$ α_t : attenuation coefficient, d_t (mm) = daily rainfall of t day, T = half-life, used one day |
| effective rainfall intensity | I_w | mm/hr | $I_w = R_w / T_w$ T_w = effective rainfall duration by average of the three nearest rain gauge stations during the initiation of debris flow |

IV. RESULTS AND DISCUSSION

We first used the Kolomogorov-Smironv test to check if the topographic factors show normal distribution. Then, we adopted one of the two methods for “Test for Difference between Means” for the different factors. One-Way ANOVA test (analysis of variance, ANOVA) was used if the factors showed normality; and, the non-parametric Mann-Whitney-Wilcoxon test (MWW test) was used if the

factors exhibited non-normality and to find significant effects among the different factors.

A. Kolomogorov-Smironv Test (K-S Test)

The study used the Kolomogorov-Smironv test (K-S test) to test if the 19 topographic factors showed normal distribution (normality). The level of significance in the K-S test was set as 0.01; a level of significance greater than 0.01 is accepted and rejected if smaller than 0.01. The test results show that the three factors for mean basin width (W), mean river slope (S), and form factor (F) showed a level of significance smaller than 0.01 (see Table III). The overall factors showed non-normality for the 19 geologic factors. The MWW test is adopted for the test of difference between means.

TABLE III: K-S TEST DATA TABLE

| Factors | mean | std dev | min | max | Level of significance |
|------------|----------|----------|----------|----------|-----------------------|
| A_s (km) | 16.4681 | 12.2543 | 3.4427 | 68.0326 | 0.400 |
| P (km) | 18.9306 | 7.9454 | 8.4635 | 52.3784 | 0.480 |
| L (km) | 5.9840 | 2.2523 | 2.5063 | 15.8583 | 0.217 |
| L_o (km) | 6.3666 | 2.8442 | 2.0919 | 17.7624 | 0.693 |
| L_T (km) | 28.9432 | 29.0145 | 5.3540 | 191.2645 | 0.190 |
| W (km) | 2.9013 | 2.0247 | 1.1121 | 12.7360 | 0.005 |
| J | 1.3681 | 0.3018 | 0.4430 | 2.2423 | 0.235 |
| N | 7.8704 | 3.4644 | 2.0000 | 16.0000 | 0.498 |
| H (m) | 1993.509 | 724.5209 | 517.3529 | 3205.040 | 0.245 |
| R_f (m) | 1643.333 | 396.0703 | 720.0000 | 2380.000 | 0.972 |
| R_d | 0.3396 | 0.1010 | 0.1507 | 0.5448 | 0.968 |
| S (%) | 32.0422 | 5.1678 | 14.4800 | 38.9600 | 0.053 |
| S_r (%) | 1.5461 | 0.0419 | 1.3424 | 1.5684 | 0.000 |
| F | 0.5885 | 0.7366 | 0.1829 | 4.4621 | 0.000 |
| C | 0.7702 | 0.0772 | 0.5494 | 0.9012 | 0.665 |
| M | 0.5990 | 0.1160 | 0.3019 | 0.8122 | 0.883 |
| E | 0.3841 | 0.0558 | 0.2730 | 0.5055 | 0.988 |
| D_s | 0.0015 | 0.0002 | 0.0010 | 0.0022 | 0.505 |
| F_s | 5.57E-07 | 2.61E-07 | 1.38E-07 | 1.15E-06 | 0.799 |

B. Non-Parametric Mann-Whitney-Wilcoxon Test (MWW Test)

The MWW test assumed samples show non-normality to compare two samples' differences. The process of the test is to separate the 54 sub-basins into two parts of with and without debris flows initiated by historical records. The basins without debris flow creeks is identified (ID) as 「0」 (validation set), and 「1」 for those with debris flow creeks (training set). The significance level is defined as 0.01, smaller than 0.01 is rejected as no significant difference among these factors. Results of the MWW test (Table IV) for the 19 topographic factors show that the following 12 topographic factors showed significant differences (a significance level greater than 0.01) including effective basin area (A_s), length of main stream (L), width of basin (W), sinuosity (J), relief energy (R_f), relief ratio (R_d), form factor (F), compactness (C), circularity ratio (M), elongation ratio (E), drainage density (D), and stream frequency (F_s).

C. Topographic Factors Screening Analysis Using Principal Component Analysis

There are up to 12 sub-basins' topographic factors that showed significant differences following mutual verification by comparing the K-S test (see Table III) and MWW test (see

Table IV), and therefore need to be further reduced. The study used linear combination of statistical model to screen debris flow potential factors by Factor Analysis in multi-variable statistics of Principal Component Analysis (PCA) and fit Correlation Analysis using a correlation coefficient matrix. The purpose of performing PCA was to screen out the factors that show the largest differences or are independent of others. A Pearson Correlation Matrix was used to standardize and transfer the data from different units of factors. The factor of standardized correlation coefficient is between -1 and +1 and shows the lowest correlation with other factors as the correlation coefficient close to 0, and vice versa as the absolute value close to 1. The selection of principal components follows the Kaiser rule [10] for the factor's corresponding initial eigenvalue (λ) with $\lambda \geq 1$ and the level of significance is defined as the absolute value of eigenvectors over 0.7 [11]. In general, a higher absolute value of eigenvector shows a close correlation between the two factors and high absolute value of eigenvector of factor is chosen by Pearson correlation coefficient matrix. The order of first principal component is stream frequency, effective basin area, circularity ratio, compactness, and relief ratio by the ranking of the absolute value of eigenvectors (Tables V and VI). It is found that the effective basin area has a higher significant (absolute value of eigenvector) than others and is selected as the first principal component.

TABLE IV: MANN-WHITNEY-WILCOXON TEST DATA TABLE

| Factors | ID | No. | mean | Sum | U value | Wilcoxon W | Z value | Level of significance |
|------------|----|-----|---------|-------|---------|------------|---------|-----------------------|
| A_s (km) | 0 | 41 | 24.3659 | 999 | 138 | 999 | -2.5999 | 0.010 |
| | 1 | 13 | 37.3846 | 486 | | | | |
| P (km) | 0 | 41 | 24.2927 | 996 | 135 | 996 | -2.6606 | 0.008 |
| | 1 | 13 | 37.6154 | 489 | | | | |
| L (km) | 0 | 41 | 23.9268 | 981 | 120 | 981 | -2.9640 | 0.004 |
| | 1 | 13 | 38.7692 | 504 | | | | |
| L_0 (km) | 0 | 41 | 25.7805 | 1057 | 196 | 1057 | -1.4264 | 0.154 |
| | 1 | 13 | 32.9231 | 428 | | | | |
| L_T (km) | 0 | 41 | 23.5854 | 967 | 106 | 967 | -3.2473 | 0.002 |
| | 1 | 13 | 39.8462 | 518 | | | | |
| W (km) | 0 | 41 | 24.4146 | 1001 | 140 | 1001 | -2.5594 | 0.011 |
| | 1 | 13 | 37.2308 | 484 | | | | |
| J | 0 | 41 | 27.3659 | 1122 | 261 | 1122 | -0.1113 | 0.912 |
| | 1 | 13 | 27.9231 | 363 | | | | |
| N | 0 | 41 | 24.1829 | 991.5 | 130.5 | 991.5 | -2.7652 | 0.006 |
| | 1 | 13 | 37.9615 | 493.5 | | | | |
| H (m) | 0 | 41 | 32.1220 | 1317 | 77 | 168 | -3.8340 | 0.001 |
| | 1 | 13 | 12.9231 | 168 | | | | |
| R_f | 0 | 41 | 29.0488 | 1191 | 203 | 294 | -1.2851 | 0.199 |
| | 1 | 13 | 22.6154 | 294 | | | | |
| R_d | 0 | 41 | 30.0488 | 1232 | 162 | 253 | -2.1143 | 0.035 |
| | 1 | 13 | 19.4615 | 253 | | | | |
| S (%) | 0 | 41 | 31.2683 | 1282 | 112 | 203 | -3.1261 | 0.002 |
| | 1 | 13 | 15.6154 | 203 | | | | |
| S_L (%) | 0 | 41 | 31.2927 | 1283 | 111 | 202 | -3.1461 | 0.002 |
| | 1 | 13 | 15.5385 | 202 | | | | |
| F | 0 | 41 | 25.5366 | 1047 | 186 | 1047 | -1.6287 | 0.104 |
| | 1 | 13 | 33.6923 | 438 | | | | |
| C | 0 | 41 | 27.5122 | 1128 | 266 | 357 | -0.0101 | 0.992 |
| | 1 | 13 | 27.4615 | 357 | | | | |
| M | 0 | 41 | 27.5122 | 1128 | 266 | 357 | -0.0101 | 0.992 |
| | 1 | 13 | 27.4615 | 357 | | | | |
| E | 0 | 41 | 27.3415 | 1121 | 260 | 1121 | -0.1315 | 0.896 |
| | 1 | 13 | 28.0000 | 364 | | | | |
| D_s | 0 | 41 | 24.5610 | 1007 | 146 | 1007 | -2.4380 | 0.015 |
| | 1 | 13 | 36.7692 | 478 | | | | |
| F_s | 0 | 41 | 30.0488 | 1232 | 162 | 253 | -2.1143 | 0.035 |
| | 1 | 13 | 19.4615 | 253 | | | | |

The second principal components for absolute value of eigenvectors over 0.7 include effective basin area and mean basin width. The two factors have close correlation, considering the fact that the form factor is more commonly used and is higher in significance than the mean basin width. Form factor is selected as the second principal component. The third principal component is elongation ratio and the fourth principal component is Relief energy. In summary, the four factors for effective basin area, form factor, elongation ratio, and relief energy are chosen as the effective principal components for the following analysis.

TABLE V: INITIAL EIGENVALUE AND TOTAL VARIANCES EXPLAINED TABLE

| No. | Initial eigenvalues | | Extraction sums of squared loadings | | | |
|-----|---------------------|--------------|-------------------------------------|-------|--------------|----------------|
| | total | Variance (%) | Cumulative (%) | total | Variance (%) | Cumulative (%) |
| | | | | | | |
| 1 | <u>3.825</u> | 31.872 | 31.872 | 3.825 | 31.872 | 31.872 |
| 2 | <u>2.496</u> | 20.797 | 52.669 | 2.496 | 20.797 | 52.669 |
| 3 | <u>1.936</u> | 16.133 | 68.802 | 1.936 | 16.133 | 68.802 |
| 4 | <u>1.353</u> | 11.278 | 80.080 | 1.353 | 11.278 | 80.080 |
| 5 | 0.885 | 7.378 | 87.458 | 0.885 | 7.378 | 87.458 |
| 6 | 0.708 | 5.904 | 93.362 | 0.708 | 5.904 | 93.362 |
| 7 | 0.348 | 2.898 | 96.260 | 0.348 | 2.898 | 96.260 |
| 8 | 0.246 | 2.052 | 98.313 | 0.246 | 2.052 | 98.313 |
| 9 | 0.140 | 1.169 | 99.481 | 0.140 | 1.169 | 99.481 |
| 10 | 0.051 | 0.421 | 99.903 | 0.051 | 0.421 | 99.903 |
| 11 | 0.010 | 0.082 | 99.985 | 0.010 | 0.082 | 99.985 |
| 12 | 0.002 | 0.015 | 100.000 | 0.002 | 0.015 | 100.000 |

TABLE VI: COMPONENT MATRIX AND VALUE OF EIGENVECTOR OF FACTORS IN PRINCIPAL COMPONENT ANALYSIS

| Factor | Component | | | |
|--------------------------|---------------|--------------|--------------|--------------|
| | 1 | 2 | 3 | 4 |
| A_s (km ²) | <u>-0.778</u> | 0.148 | 0.517 | 0.076 |
| L_0 (km) | -0.677 | -0.515 | 0.352 | 0.100 |
| W (km) | -0.515 | <u>0.796</u> | 0.231 | -0.123 |
| J | -0.289 | -0.691 | 0.349 | 0.155 |
| R_f (m) | -0.238 | 0.255 | -0.120 | <u>0.817</u> |
| R_d | <u>0.702</u> | 0.238 | -0.166 | 0.452 |
| F | -0.190 | <u>0.914</u> | -0.141 | -0.111 |
| C | <u>0.727</u> | 0.116 | 0.612 | -0.041 |
| M | <u>0.735</u> | 0.130 | 0.598 | -0.034 |
| E | 0.156 | 0.257 | <u>0.751</u> | 0.176 |
| D_s | 0.152 | -0.024 | 0.031 | -0.608 |
| F_s | <u>0.843</u> | -0.205 | -0.101 | 0.099 |

There were 10 factors used for the following discriminant and regression analyses for debris flow potential analysis and calculation of its correctness. The selected independent variable includes 4 topographic factors (effective basin area, form factor, elongation ratio, and relief energy), 4 geologic factors (geologic index, fault length, landslide ratio, and ratio of dip slope area), and 2 climatic factors (effective cumulative rainfall and effective rainfall intensity).

The database of 10 factors were selected as dependent variables (outcome variable), and were divided into two categories, the first category 「1」 of 13 sub-basins with records of debris flow and the second category 「0」 of 41 sub-basins without debris flow records.

1) Fisher's discriminant analysis

The results of Fisher's Discriminant Analysis are function coefficient by the classification coefficient of variables (Table VII). The discriminant functions for category 「1」 and category 「0」 are listed as follows:

$$Y_1 = (0.549*A_s) + (1.391*F) + (47.249*E) + (-0.002*R_f) + (-0.923*E_d) + (-124.904*F_L) + (2.949*G) + (42.177*D_A) + (0.068*R_w) + (8.414*I_w) - 142.941 \quad (1)$$

$$Y_0 = (0.318*A_s) + (0.276*F) + (63.941*E) + (0.003*R_f) + (-0.627*E_d) + (-143.363*F_L) + (2.452*G) + (55.851*D_A) + (0.061*R_w) + (7.940*I_w) - 134.263 \quad (2)$$

$$Y = Y_1 - Y_0 = 0.231A_s + 1.115F - 16.692E - 0.005R_f - 0.296E_d + 18.459F_L + 0.497G - 13.674D_A + 0.007R_w - 0.474I_w - 8.678 \quad (3)$$

In the equations, Y_0 is the discriminate function for non-debris flow sub-basins and Y_1 is for debris flow sub-basins. Y is the overall discriminate function for debris flow sub-basins if the value of Y is greater than 0, and for non-debris flow sub-basins if the value is smaller than 0.

A training sample model was adopted to reduce bias (over fitting) and error rate in calculation of the correctness in discriminant analysis in SPSS. The holdout samples method randomly separates the samples into two sets for the training set used for finding the discriminate function of equation and the validation set used for correctness calculation of the discriminate function. The two sets are then changed and the process is repeated for cross validation for the correctness of analysis. The correctness is 88.89 % for the training set, 74.07 % for the validation set, and has overall correctness 81.48 % using Fisher's Discriminant Analysis (see Table VIII).

TABLE VII: COEFFICIENTS OF CLASSIFICATION FUNCTIONS BY FISHER'S DISCRIMINANT ANALYSIS

| Factor | category | | coefficient of function (= $\lceil 1 \rceil - \lceil 0 \rceil$) |
|--------------------------|-------------------|-------------------|---|
| | $\lceil 0 \rceil$ | $\lceil 1 \rceil$ | |
| A_s (km ²) | 0.318 | 0.549 | 0.231 |
| F | 0.276 | 1.391 | 1.115 |
| E | 63.941 | 47.249 | -16.692 |
| R_f (m) | 0.003 | -0.002 | -0.005 |
| E_d | -0.627 | -0.923 | -0.296 |
| F_L | -143.363 | -124.904 | 18.459 |
| G | 2.452 | 2.949 | 0.497 |
| D_A | 55.851 | 42.177 | -13.674 |
| R_w (mm) | 0.061 | 0.068 | 0.007 |
| I_w (mm/hr) | 7.940 | 8.414 | 0.474 |
| constant | -134.263 | -142.941 | -8.678 |

TABLE VIII: RESULTS OF DISCRIMINANT FUNCTION ANALYSIS BY FISHER'S DISCRIMINANT ANALYSIS

| Number of training set (54 data) | | | sum |
|---|-------------|-------------|------------|
| category | 0 | 1 | |
| 0 | 39 (95.1 %) | 2 (4.9 %) | 41 (100 %) |
| 1 | 4 (30.8 %) | 9 (69.2 %) | 13 (100 %) |
| correctness: $[(39+9)/(41+13)] \times 100 \% = 88.89 \%$ | | | |
| Number of validation set (54 data) | | | sum |
| category | 0 | 1 | |
| 0 | 32 (78 %) | 9 (22 %) | 41 (100 %) |
| 1 | 5 (38.5 %) | 8 (61.5 %) | 13 (100 %) |
| correctness: $[(32+8)/(41+13)] \times 100 \% = 74.07 \%$ | | | |
| Number of overall set (108 data) | | | sum |
| category | 0 | 1 | |
| 0 | 71 (86.6 %) | 11 (13.4 %) | 82 (100 %) |
| 1 | 9 (34.6 %) | 17 (65.4 %) | 26 (100 %) |
| correctness: $[(71+17)/(82+26)] \times 100 \% = 81.48 \%$ | | | |

2) Logistic regression analysis

The regression function for debris flow potential analysis was further studied by Logistic Regression Analysis using the ten selected factors. The coefficients of function were obtained from logistic coefficient (B) in classification table of Logistic Regression Analysis (see Table IX). The regression equation was further examined by the Chi-square Test that was significantly (p-value) smaller than 0.05 with acceptable goodness of fit in 95 % of confidence interval. The regression function is listed as follows:

$$Y = 0.165A_s + 0.817F - 14.822E - 0.003R_f - 0.473E_d + 14.150F_L + 0.490G - 29.906D_A + 0.004R_w + 0.347I_w - 3.653 \quad (4)$$

Results of the Logistic Regression Analysis show that the correctness is 100 % for non-debris flow sub-basins (category $\lceil 0 \rceil$), 69.2 % for debris flow sub-basins (category $\lceil 1 \rceil$), with 92.6 % of overall correctness.

TABLE IX: CLASSIFICATION COEFFICIENT TABLE BY LOGISTIC REGRESSION ANALYSIS

| Factor | B | S.E. | Wald | degrees of freedom | p-value | Exp (B) |
|--------------------------|---------|--------|-------|--------------------|---------|---------|
| A_s (km ²) | 0.165 | 0.084 | 3.827 | 1 | 0.050 | 1.179 |
| F | 0.817 | 0.595 | 1.881 | 1 | 0.170 | 2.263 |
| E | -14.822 | 12.049 | 1.513 | 1 | 0.219 | 0.000 |
| R_f (m) | -0.003 | 0.002 | 2.064 | 1 | 0.151 | 0.997 |
| E_d | -0.473 | 0.574 | 0.681 | 1 | 0.409 | 0.623 |
| F_L | 14.150 | 20.144 | 0.493 | 1 | 0.482 | 1397007 |
| G | 0.490 | 0.392 | 1.565 | 1 | 0.211 | 1.632 |
| D_A | -29.906 | 22.055 | 1.839 | 1 | 0.175 | 0.000 |
| R_w (mm) | 0.004 | 0.004 | 1.238 | 1 | 0.266 | 1.004 |
| I_w (mm/hr) | 0.347 | 0.360 | 0.931 | 1 | 0.335 | 1.415 |
| constant | -3.653 | 8.240 | 0.196 | 1 | 0.658 | 0.026 |

3) Comparisons of the two analyzed methodologies

The correctness by Fisher's model is 69.23 % for debris flow sub-basins and 95.12 % for non-debris flow sub-basins, with an overall correctness of 81.48 %. The correctness is 69.23 % for debris flow sub-basins and 100 % for non-debris flow sub-basins, with an overall correctness of 92.6 % by the Logistic Regression model. Fisher's model shows the same correctness for Logistic Regression model in category $\lceil 1 \rceil$ (debris flow sub-basin); and lower than Logistic Regression model in category $\lceil 0 \rceil$ (non-debris flow sub-basin). In general, the prediction model by Logistic Regression model has greater correctness than by Fisher's model. The study suggests using the regression function by Logistic Regression model for potential debris flow analysis and is expected to apply to other areas.

V. CONCLUSION

The study used Fisher's Discriminant Analysis and Logistic Regression Analysis for evaluation of debris flow potentials and its correctness by setting up a function of linear equation. The following list summarizes the results:

- 1) There are 54 sub-basins, including 13 with and 41 without debris flow records, in the study area. Screening analysis for 19 topographic factors were analyzed using the Kolmogorov-Smirnov test (K-S test), Mann-Whitney-Wilcoxon test (M-W-W test), Correlation Analysis, and Principal Component Analysis. Four main topographic factors, including effective watershed area, form factor, elongation ratio, and relief energy, were selected for the analysis.
- 2) Ten factors showed significant effects to the initiation of debris flow: Effective basin area, form factor, elongation ratio, relief energy, geologic index, fault length, landslide ratio, ratio of dip slope area, effective cumulative rainfall, and effective rainfall intensity.
- 3) The correctness for the debris flow sub-basins was 88.89 % and 74.07 % for the none-debris flow sub-basins. The overall correctness for the two sets was 81.48 % using Fisher's Discriminant Analysis.
- 4) The correctness for the debris flow sub-basins was 69.23 % and 100.0 % for the none-debris flow sub-basins using Logistic Regression Analysis. The overall correctness for the two sets was 92.6 % using Logistic Regression Analysis.
- 5) The study suggests using Logistic Regression Analysis for debris flow potential analysis in the study basin. The regression equation was set as: $Y = 0.165A_s + 0.817F - 14.822E - 0.003R_f - 0.473E_d + 14.150F_L + 0.490G - 29.906D_A + 0.004R_w + 0.347I_w - 3.653$.

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