

1   **Empirical prediction for travel distance of channelized**  
2   **rock avalanches in the Wenchuan earthquake area**

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9   **Abstract.** Rock avalanches are extremely rapid, massive flow-like movements of fragmented rock. The  
10   travel path of the rock avalanches may be confined by channels in some cases, which were named as  
11   the channelized rock avalanches. Channelized rock avalanches are potentially dangerous due to their  
12   hardly predictable travel distance. In this study, we constructed a dataset with detailed characteristic  
13   parameters of 38 channelized rock avalanches triggered by the 2008 Wenchuan earthquake using the  
14   visual interpretation of remote sensing imagery, field investigation, and literature review. Based on this  
15   dataset, we assessed the influence of different factors on the runout distance and developed prediction  
16   models of the channelized rock avalanches using the multivariate regression method. The results  
17   suggested that the movement of channelized rock avalanche was dominated by the landslide volume,  
18   total relief, and channel gradient. The performance of both models was then tested with an independent  
19   validation dataset of 8 rock avalanches that induced by the 2008 Wenchuan, the Ms7.0 Lushan  
20   earthquake, and heavy rainfall in 2013, showing acceptable good prediction results. Therefore, the  
21   travel distance prediction models for channelized rock avalanches constructed in this study is  
22   applicable and reliable for predicting the run out of similar rock avalanches in other regions.

23  
24   Keywords: channelized rock avalanches; travel distance; empirical prediction; multivariate regression  
25   model; Wenchuan earthquake

26   **1 Introduction**

27   Rock avalanches are extremely rapid, massive flow-like movements of fragmented rock from a very  
28   large rock slide or rock fall (Hung et al. 2014). Hundreds of rapid and long run-out rock avalanches  
29   were triggered by 2008 Wenchuan earthquake in Sichuan Province (Zhang et al. 2013), with

30 catastrophic consequences for residents in the affected areas. For instance, the  $1.5 \times 10^7$  m<sup>3</sup> Donghekou  
31 rock avalanche in Qingchuan County, near the seismogenic fault, traveled 2.4 km, killing about 780  
32 persons and destroying four villages (Zhang et al. 2013). Rock avalanches can cause incredible damage  
33 due to their characteristics of high-speed and unexpectedly long runout, while their transport  
34 mechanisms are still considered to be controversial among many researchers (Hung et al. 2001).

35 Therefore, constructing prediction models for rock avalanche travel distance is meaningful in terms of  
36 not only theoretical research on motion mechanisms but also in practical application for mitigation of  
37 rock avalanche risk.

38 Methods for determining the travel distance of landslides can be divided into two categories: dynamic  
39 modeling (Heim 1932; Hung et al. 2009; Lo et al. 2011; Pastor et al. 2009; Sassa 1988), and empirical  
40 modeling (Scheidegger 1973; Lied et al 1980; Finlay et al. 1999; Westen et al. 2006; Guo et al. 2014).  
41 The dynamic models provide information on landslide intensity, such as velocity, affected area and  
42 deposition depth, in addition to travel distance. Nonetheless, dynamic models require accurately  
43 quantified input parameters that are difficult to obtain before the events, and many simplified  
44 assumptions that are not applicable to the actual situation. Empirical models considering the  
45 correlations between observational data provide an effective technique to aid in understanding  
46 mechanisms of rock-avalanche motion and to develop practical models for predicting rock-avalanche  
47 travel distance. However, the empirical-statistical models set up from samples with different  
48 geomorphological and geological surroundings, trigger conditions, or failure modes are not very  
49 sufficient to be applied to Wenchuan earthquake area.

50 In this study, we compiled a dataset of 38 rock avalanches with flow paths confined by channels (this  
51 kind of landslide is hereinafter termed as channelized rock avalanche) from interpretation of remote  
52 sensing, field investigations and literature review (see Section 3.1). Statistical correlations were used to  
53 determine the principle factors affecting the mobility of the channelized rock avalanches. Then a  
54 stepwise multivariate regression model was developed to build a best-fit empirical model for the travel-  
55 distance prediction of this kind of rock avalanches in the Wenchuan earthquake area. A derivative  
56 multivariate regression model was also constructed. The performance of both models was then tested  
57 with an independent validation dataset of 8 rock avalanches in the same area.

58

59      **2 Rock avalanches in study area**

60      The study area (see Figure 1) is on the northeast-trending Longmenshan thrust fault zone between the  
61      Sichuan basin and the Tibetan plateau. Three major sub-parallel faults are: the Wenchuan-Maowen  
62      fault, the Yingxiu-Beichuan fault and the Pengguan fault (Fan et al., 2014). With highly developed  
63      stream systems, this region is characterized by high mountains and deep valleys and extreme rates of  
64      erosion (Fu et al 2009; Qi et al 2011).

65      This study selected 38 channelized rock avalanches induced by the Wenchuan earthquake to study the  
66      relations between travel distance and influential factors. These rock avalanches occurred along the  
67      seismogenic Yingxiu-Beichuan fault; the distance to the fault ranged from 0 m ~21,300 m with a mean  
68      value of 3,895 m. Another distribution characteristic was that these rock avalanches mainly clustered  
69      on the step-overs, bends and distal ends of the seismogenic fault. These distribution characteristics of  
70      the large rock avalanches suggested that the occurrence of rock avalanches was associated with very  
71      strong earthquake ground motion. The Wolong Station recorded the highest seismic acceleration with  
72      the peak ground acceleration reaching 0.948g vertically and 0.958g horizontally (Yu et al., 2009).

73      Locally, the ground motion was high enough to throw rocks into the air.

74      The lithology of outcropping rock in source areas can be divided to four types: carbonaterock, phyllite,  
75      igneous rock and sandstone. The landslide deposit of the rock avalanches in the study area structure  
76      was usually debris, which suggests that the sliding masses were intensively fragmented during their  
77      movement.

78      The influence of the local geomorphology on the paths of the rock avalanches was obtained from  
79      remote-sensing images after the events. Although the rock avalanches we chose all had flow paths  
80      confined by channels, some topographic differences were found to be significant in affecting present  
81      that had affected the shape morphology of the rock avalanche deposits. The source areas had well-  
82      defined boundaries. When the source mass was detached from the slide bedrock, it may directly  
83      move into the channel down slope (see Figure 2b), or access the channel with enter it at some impact  
84      transition angle of movement direction (see Figure 2a). The channel itself may have changes in  
85      direction and inclination. The distal end of the landslide may lie stop in the channel (see Figure 2a) or  
86      may reach to wide valley or plain (see Figure 2b).

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112 reach to wide valley or plain (see Figure 2b).

113

114 **3 Data and method**

115 **3.1 General consideration**

116 Various statistical methods have been applied to predict travel distance of landslides. The most  
117 prevalent one is the equivalent friction coefficient model, which only takes account of landslide volume  
118 (Scheidegger 1973). Another well-known model is the statistical  $\alpha$ - $\beta$  model in which the maximum  
119 runout distance is solely a function of geography (Lied et al., 1980; Gauer et al., 2010). Finlay et  
120 al.(1999) developed some multiple regression models containing slope geometric parameters like slope  
121 height and slope angle for the travel distance prediction of landslides on the artificial slopes upon the  
122 horizontal surface. Based on the data of 54 landslides which was relatively open or confined by gentle  
123 lateral slope, Guo et al.(2014) established an empirical model for predicting landslide travel distance in  
124 Wenchuan earthquake area and suggested that rock type, landslide volume, and slope transition angle  
125 play dominant roles on landslide travel distance. And there are increasing sound that the prediction  
126 models of travel distance should adapt to different types of landslides (Corominas 1996; Fan et al,  
127 2014;).

128 Moreover, the local morphology plays an important role on shape and mobility of rock avalanches.  
129 Heim (1932) firstly mentioned the influence of local morphology that the debris masses will undergo  
130 different effects with the angle of impact changing, and rock avalanches has to conform to the local  
131 morphology regardless of their scale. Abele (1974) summarized four different possibilities of  
132 adaptation of the rock avalanche to local morphology. Hsu(1975) noted that a sinuous pathway can  
133 reduced runout distance of rock avalanches. Nicoletti (1991) inferred that local morphology impacts on  
134 landslide motion through changing the rate of total energy dissipation along the travel path. To  
135 determine the influence of specific channels on the travel distances of rock avalanches, we respectively  
136 consider the impacts of gradients of the upper slopes and lower channels.

137 Rock avalanches triggered by Wenchuan earthquake usually initiated from top or the higher part of  
138 slopes possibly due to the altitude amplification effect of earthquake acceleration, therefore the toe of  
139 the rupture surface  commonly found in the source area at the upstream of the channel  (See Figure  
140 3). When the slope failed, the failed mass travelled a long distance down the channel. The 38 rock  
141 avalanches in this study are selected with the criterion that the flow path is partially or fully confined  
142 by channels. The volumes of these rock avalanches ranged from  $0.4\text{--}50\times10^6\text{m}^3$ ; with horizontal travel

143 distances between 0.58 and 4.00 km. The volume is prior to the area to be put into the travel distance  
144 prediction model as it had much more physical meanings. And we introduced total relief as well as the  
145 height of source area to probe the influences of the potential energy difference and altitude difference  
146 of source mass on the travel distance of the rock avalanches.

147 **3.2 Data**

148 The terms and notations of a typical channelized rock avalanche are shown in Figure 3. The local  
149 morphology of a rock avalanche can be divided to three sections: initiate slope (source area), channel  
150 (main travel path or flow area) and valley floor (deposition area). When the mass moves over the slope  
151 section, it is free from lateral constraints, and the moving mass is able to spread laterally. After entering  
152 the channel, the flowing mass is constrained by the two lateral slopes. Finally, the mass may reach to a  
153 wide valley floor, where it spreads laterally and deposits. The average inclination of slope section and  
154 valley section are obtained respectively, while the gradient of valley section is neglected as it has very  
155 little variation. Slope angle ( $\alpha$ ) , denotes the average inclination of the initiated slope section.  
156 Channel angle ( $\beta$ ) , denotes the average inclination of the sectional channel. Source area height (H<sub>s</sub>),  
157 denotes the elevation difference between the crest of the sliding source and the toe of the rupture  
158 surface. Total relief (H) is the elevation difference between the crest of the sliding source and the distal  
159 end of the debris deposit. Travel distance (L) is the horizontal distance between the crest of the sliding  
160 source and the distal end of the debris deposit. Landslide area (A) is the source area of the rock  
161 avalanche obtained from remote sensing image interpretation. An empirical scaling relationship with  
162 different empirical coefficients is frequently used to link the volume and the area of landslides in  
163 different areas or with different types, and we chose the one developed by Parker et al. (2011) in the  
164 same study area. Volume of some rock avalanches with detailed field investigation are replaced by the  
165 data from published literature. The parameters of 38 rock avalanches are listed in Table 1.

166 **3.3 Method**

167 Travel distance is the most desirable prediction in rock-avalanche hazard evaluation in mountainous  
168 areas. Travel distance prediction of rock avalanche is a complicated issue as it is determined by many  
169 different properties of the materials (i.e., grain size distribution and water content), topographical  
170 factors, mobility mechanics of failed mass, the confinement attributes of travel path, and so on (Guo et

171 al., 2014). Empirical-statistical methods have long been used as tools to study the mobility of rock  
172 avalanche since they are easy to develop and apply, and they are not dependent on knowing the  
173 physical processes involved in causing the mobility. Channelized rock avalanches have unique  
174 movement paths involving complex, and possibly little-known physical processes such as grain  
175 collisions, fragmentation and entrainment of bed material from the channel sides and bottom. Existing  
176 empirical models have not produced a favourable prediction. The forecasting index system and the  
177 prediction model of channelized rock avalanches should be discussed first.

178 In this paper, we first selected controlling factors on rock avalanche travel distance through correlation  
179 analysis. Then we fitted a stepwise multivariate regression model using all significant correlation  
180 variables to obtain a best-fit empirical model for landslide travel distance, and explored which factors  
181 were statistically significant at the same time, as expressed in equation (1).

182

$$y = b_0 + b_1 x_1 + b_2 x_2 + b_3 x_3 + \dots + b_n x_n + \varepsilon \quad (1)$$

183 where  $y$  is the predictant ('dependent variable'), e.g. travel distance of rock avalanche, ( $i = 1, 2, \dots, n$ )  
184 are the predictors ('independent variables'),  $b_0$  is the intercept, ( $i = 1, 2, \dots, n$ ) are the regression  
185 coefficients of the corresponding , and  $\varepsilon$  is the residual error, here assumed to be independently and  
186 normally distributed. Predictors were added to the regression equation one at a time until there was no  
187 significant improvement in parsimonious fit as determined by the adjusted  $R^2$ .

188 **4 Results and validation**

189 **4.1 Relationships between travel distance and volume, topographic relief of rock avalanche**

190 Correlation coefficients between different variables and travel distance ( $L$ ) were calculated first,  
191 generating the correlation coefficients matrix shown in Table 2. The significant relevant predictors with  
192 the 95% confidence for travel distance prediction of channelized rock avalanche are landslide area( $A$ ),  
193 landslide volume( $V$ ), total relief( $H$ ), source area height( $H_s$ ), and channel angle( $\beta$ ), with correlation  
194 coefficient of 0.877, 0.866, 0.857, 0.675, -0.467, respectively.

195 Figures 4 illustrates that the travel distance ( $L$ ) varies exponentially with volume ( $V$ ) of rock avalanche  
196 with an exponential exponent of 0.377. Compared with a compilation of world-wide rock-avalanche  
197 data (Legros, 2002), the mobility of rock avalanches in our study area is stronger than other non-  
198 volcanic landslides (power exponent is 0.25), but weaker than volcanic landslides and debris flows

199 (both power exponent is 0.39). The relation between travel distance ( $L$ ) and total relief ( $H$ ) is shown in  
200 figure 5. The result suggests that the mobility (travel distance) of rock avalanche has relatively strong  
201 linear relationship with total relief ( $H$ ). The scale factor is close to 2.4, which means that the apparent  
202 friction coefficient ( $H/L$ ) for the rock avalanches is approximately 0.42. This is significantly lower than  
203 the commonly observed static coefficient of friction of rock material (~0.6).

204 **4.2 Multivariate regression model of rock avalanche travel distance**

205 According to the matrix of correlation coefficients (Table 2), the slope angle ( $\alpha$ ) does not have a  
206 significant correlation with travel distance ( $L$ ) at the 95% confidence level. Thus this variable could be  
207 excluded first during development of the best-fit regression model for travel distance prediction. Prior  
208 to the landslide area ( $A$ ), the landslide volume ( $V$ ) has been considered in the models as it has much  
209 more physical meaning. In the end, a stepwise linear multivariate regression technique was applied to  
210 find the best-fit travel distance regression model using the significant relevant predictors including  
211 landslide volume ( $V$ ), total relief ( $H$ ), source area height ( $H_s$ ) and channel angle ( $\beta$ ). The best-fit  
212 regression equation for travel distance prediction were derived from the dataset of Table 1 (see  
213 equation (2)), and the coefficient of the variables with 95% confidence are shown in Table 3.

214 
$$\log(L) = 0.420 + 0.079 \log(V) + 0.718 \log(H) - 0.365 \log(\tan \beta) \quad (2)$$

215 Where  $\log$  is the logarithm of 10;  $L$  is the predicted travel distance (m);  $V$  is the landslide volume ( $m^3$ );  
216  $H$  is the total relief (m);  $\beta$  is the mean gradient of the channel ( $^\circ$ ).

217 Equation (2) can be transformed to equation (3):

218 
$$L = 2.630 V^{0.079} H^{0.718} (\tan \beta)^{-0.365} \quad (3)$$

219 The best-fit travel distance regression equation indicates that the travel distance of channelized rock  
220 avalanche is positively correlated with landslide scale (landslide volume) and potential energy  
221 loss(total relief), and negatively correlated with channel gradient(channel angle),which is coherent with  
222 the results of correlation analysis in table 2.

223 While the total relief ( $H$ ) will be unknown prior to landslide occurrence, the elevation difference of  
224 source area will be available through specific field investigation on a potential rock avalanche area.  
225 Hence, we introduced  $H_s$  and  $\alpha$  in replacement of  $H$  to the regression model as they have relative high  
226 correlation with  $H$  (correlation coefficients are 0.801 and 0.429 respectively). The transformed

227 alternative regression equation is given as equation (4) with the coefficient of the variables with 95%  
228 confidence in table 3.

229

$$L = 3.6V^{0.303}Hs^{0.244}(\tan \alpha)^{-0.115}(\tan \beta)^{0.072} \quad (4)$$

230 Where L is the predicted travel distance (m); V is the landslide volume ( $m^3$ ); Hs is the height of source  
231 area (m);  $\alpha$  is the mean angle of slope segment ( $^\circ$ );  $\beta$  is the mean gradient of the channel segment ( $^\circ$ ).

232 The validity of these two models were evaluated through the significance test leading to the highest  $R^2$   
233 value and the lowest residual standard error. Table 3 shows the significance values for the prediction  
234 model equations. Adjusted  $R^2$  means adjusted multiple correlation coefficient, which represents the  
235 correlation level between the dependent variable and the independent variables. The calculation of  
236 adjusted  $R^2$  considers the number of variables and can be used to compare goodness of fit of different  
237 regression models. Adjusted  $R^2$  of the two regression equations are high, suggesting that the  
238 constructed regression models are reliable. The adjusted  $R^2$  of equation (4) is higher than equation (5),  
239 implying a higher precision for the best-fit regression model. The significance test results on the  
240 regression equation suggest the significance of multiple regression equations ( $F=173.5 > F_{0.05}(2.883)$ )  
241 for equation(2) and  $F=49.5 > F_{0.05}(2.659)$  for equation (4)). Figures 6 and 7 show the distributions of the  
242 residuals in relation to the observed travel distance estimated by using equation (2) and (4). Both plots  
243 illustrates normality, constant variance and absence of trends in the residuals.

244 Figure 8 compares the predicted travel distances estimated by using equations (2) and (4) with the  
245 observed ones. It suggests that the predicted values of the samples are close to the observed ones.  
246 Where L exceeds 2000 m, the predicted travel distance calculated by using two models are lower than  
247 actual one, with relatively large residual error. The largest residual error appears in Wenjia gully rock  
248 avalanche, followed by Hongshi Gully, Niumian Gully and Donghekou rock avalanche. According to  
249 the field investigation, projectile motion was experienced for these four rock avalanches with vertical  
250 drop of 260 m, 150 m, 60 m and 160 m respectively before they flowed along the channel downslope.  
251 Moreover, fluidization characteristics such as super-elevation near curve transitions can be found in the  
252 channel section of these four rock avalanches. These findings manifest the steep micro-geotopography  
253 will enlarge the mobility of rock avalanches as this kind of topography will lead the slide mass to  
254 undergo the projection, collision, fragmentation effects in the early motion stage which will facilitate

255 motion mode transformation from sliding to flowing. This transformation will enhance the motion  
256 mobility of rock avalanche to travel a much longer distance than predicted one.

257 **4.3 Validation**

258 The regression equations were tested using an independent sample validation data set (Table 4) of 8  
259 rock avalanches in the same area induced by three different kinds of triggers: 2008 M<sub>s</sub>7.8 Wenchuan  
260 earthquake, 2013 M<sub>s</sub>7.0 Lushan earthquake, and heavy rainfall. The volume of these samples ranged  
261 from  $8.8 \times 10^4$ – $150 \times 10^4 \text{m}^3$ , and travel distance from 372–1372 m. The background parameters and the  
262 predicted values of each avalanche are listed in Table 4. The relative errors between the predicted  
263 values estimated by using equation (3) and observed values of the travel distance of the rock  
264 avalanches,  $|L_{\text{predicted}} - L_{\text{observed}}| / L_{\text{observed}} \times 100\%$ , are between -14.4% and 17.2%, while the  
265 relative errors are -44.0% and 17.9% for equation (4). On the whole, these two regression models  
266 achieved acceptable prediction accuracy for preliminary forecasting of travel distance of rock  
267 avalanches in rugged mountainous areas. The best-fit regression model appeared to provide greater  
268 precision than the alternative model. Regarding the influence of triggers on the travel distance of the  
269 channelized rock avalanches, those triggered by rainfall and the Lushan earthquake seemed to be more  
270 mobile. It is inferred that the former difference is due to the high water content in failed mass induced  
271 by rainfall. A possible reason why two rock avalanches triggered in the Lushan earthquake travelled  
272 farther may be because of structural weakening of slope rock mass in the 2008 Wenchuan earthquake  
273 in the study area.

274 **5 Discussion**

275 **5.1 Prediction for travel distance of channelized rock avalanche**

276 The results of our analysis of the data set, indicates that the mobility (travel distance) of channelized  
277 rock avalanche is positively correlated with landslide volume and total relief but negatively correlated  
278 with channel angle. It is inferred that the movement of channelized rock avalanche was strictly  
279 constrained by the local geomorphology. As Figure 3 shows, the travel distance of channelized rock  
280 avalanche would rapidly increase with volume of rock avalanche enlarged. Such a high correlation  
281 between landslide volume and travel distance implies that the travel distance of channelized rock  
282 avalanche is dominated by the spreading of the slide mass (Davies, 1982; Staron, 2009). The high

positive correlation between total relief and travel distance is for two reasons: the larger the total relief is, the more kinetic energy the slide mass could obtain and the further distance could it travel; another contribution is the geometrical similarity of hillslope geomorphology in the study area (Legros, 2002).

Regarding the medium negative correlation between travel distance and channel angle, it is inferred that when the slide mass rushed into the channel after the acceleration movement on the upper hillslope, it had relatively high velocity and extremely low frictional coefficient among the rock fragments, and the channel could not stop the rock avalanche until it lost fragment flow discharge. Hence, the travel distance of channelized rock avalanche would increase with the channel angle cut down given the same flow discharge (landslide volume), relative stable flow velocity, and similar flow capacity. However, it is still difficult to evaluate the flow capacity of the channels due to difficulty of quantifying its cross-section shape (width and depth of channels), resistance to the rock avalanche and even the shape changing induced by entrainment process of rock avalanche.

The residual analysis result demonstrates that the projection process in the early motion stage will significantly enlarge the travel distance of rock avalanches. The nature of this phenomenon is suggested to be involved with transformation of motion mode from sliding to flowing due to collision and fragmentation effects after the projection (Davies et al, 1999). Furthermore, the degree of fragmentation of failed mass should have remarkable influence on the travel distance of rock avalanche, and other factors changing the fragmentation degree should be further study, such as earthquake effect, geologic structure and rock type.

### 5.2 Conceptual model for transportation of channelized rock avalanche

The statistical results imply that the travel distance of channelized rock avalanche is highly correlated with landslide volume, total relief and channel angle. As the total relief and channel angle act as external factors for the motion of rock avalanche, it seems like it is in essence landslide volume that control the rock avalanche movement. Actually, a good fitting result between travel distance and landslide volume appears on our data set (Figure 4). So we propose a conceptual model for channelized rock avalanche transportation: An initial failed mass rushes into the channel with certain velocity after acceleration and fragmentation effects over the upper slope. Then the failed mass will “forget” the initial fall height and flow down in the channel like unsteady flow. The flow discharge (including

312 initial landslide volume and entrainment volume) and the flow capacity of the channel control the  
313 travel distance of channelized rock avalanche without considering the motion mechanism.  
314 However, the flow capacity varies along the channel. Some local depression can store a mass of the  
315 moving rock debris, causing a lack of flow discharge for the downstream channel and a considerable  
316 decrease of travel distance. Taking Wenjia Gully rock avalanche for an example, almost a half of total  
317 volume of the landslide deposit on the beginning of the channel (red dash circle area in Figure 9),  
318 leading to that the distal deposition appeared in the channel instead of the valley. Thus assessing the  
319 flow capacity of the channel for rock avalanche motion will assist in future forecast of potential rock  
320 avalanche hazard in mountainous areas.

## 321 **6 Conclusion**

322 Channelized  rock avalanche refers to a rock avalanche with a flow path confined between valley walls.  
323 Relevant  Detailed data on thirty-eight channelized rock avalanches triggered by Wenchuan earthquake  
324 were collected by remote sensing, field investigation and literature review. The results of correlation  
325 and regression analysis revealed that the movement of channelized rock avalanches is dominated by  
326 spreading of the failed mass. Landslide volume ( $V$ ), total relief ( $H$ ) and channel angle ( $\beta$ ) had  
327 predominant effects played a dominating role in the on travel distance of channelized rock avalanches.  
328 Stepwise multivariate regression was used to develop a nonlinear best-fit  travel distance prediction  
329 model for the channelized rock avalanches. An alternative multivariate regression model was also built.  
330 The reliability of the two models was tested on by an independent validation dataset of 8 rock  
331 avalanches in the same area and produced good results, meeting the requirements for preliminary  
332 evaluation of travel distance for channelized rock avalanches in the Wenchuan earthquake area.

## 333 **Acknowledgement**

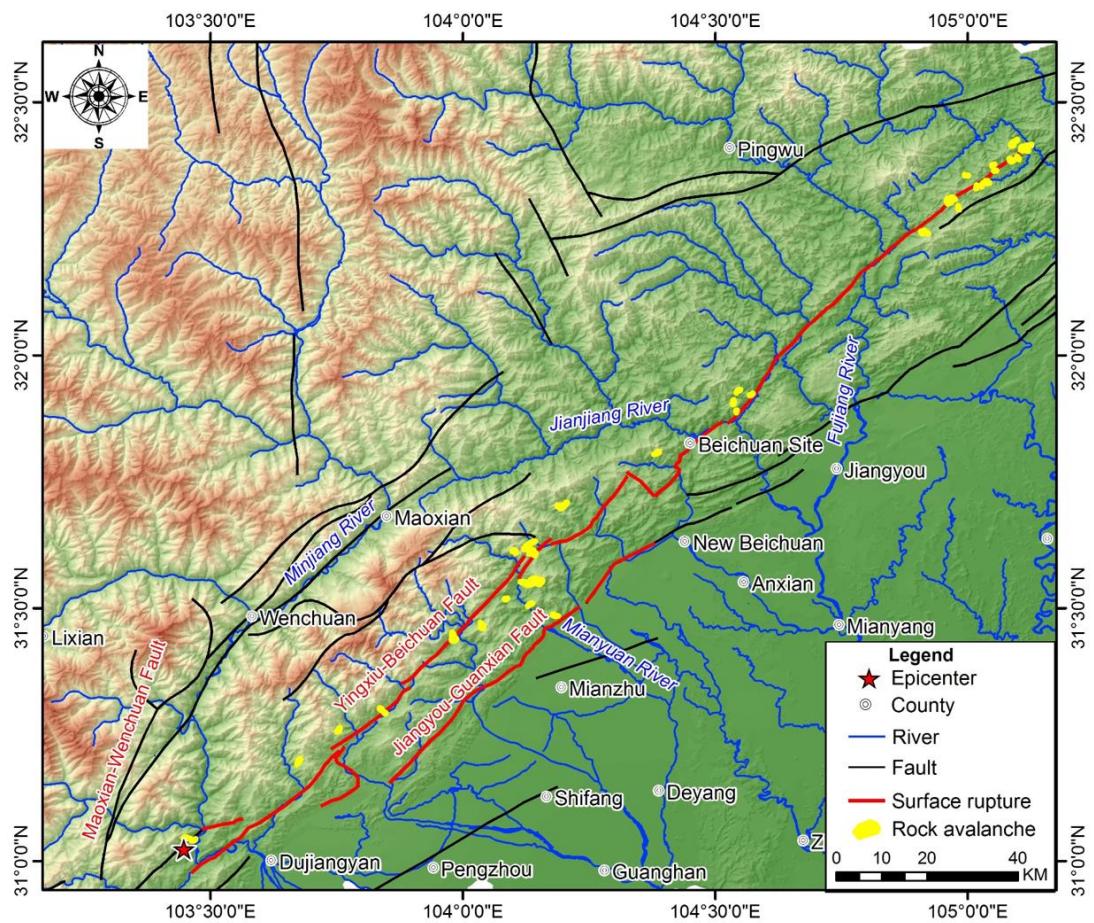
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339

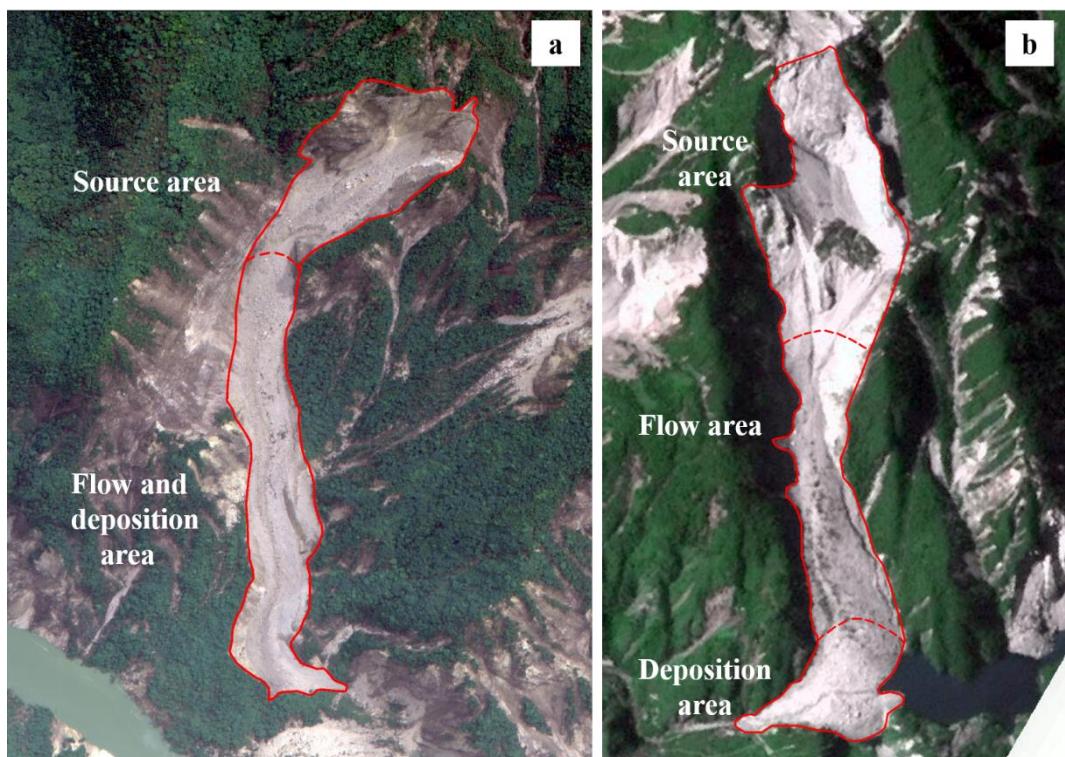
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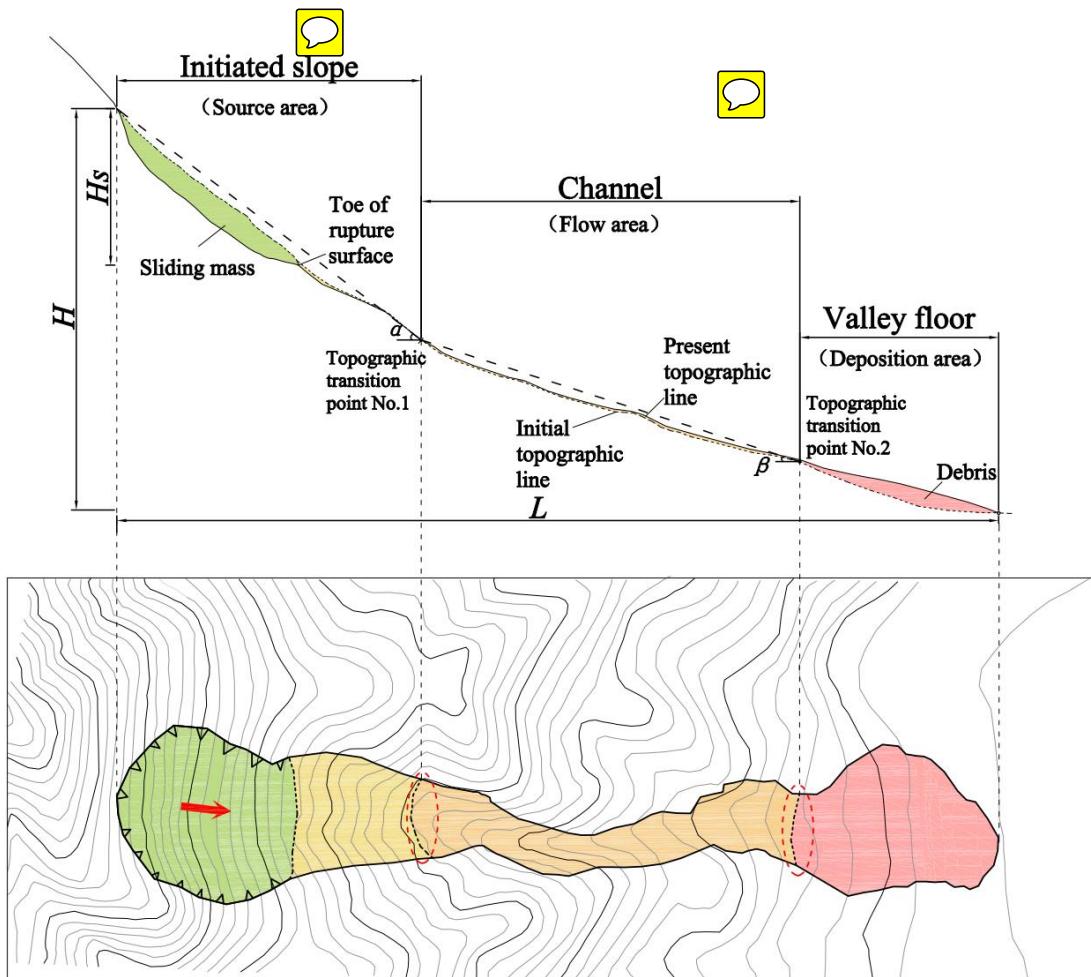
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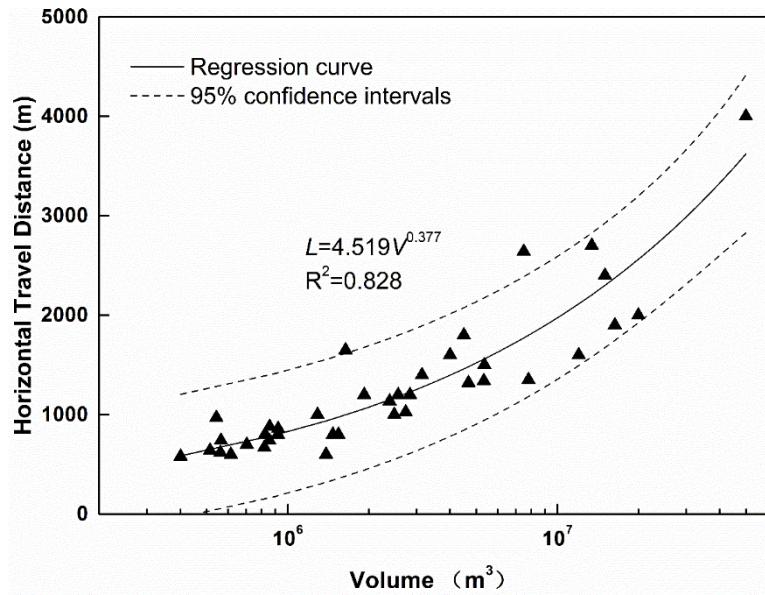
393



395     Figure 2: Remote-sensing images of two channelized rock avalanches triggered by the Wenchuan  
396     earthquake. a is Changtan rock avalanche (No.21 in table 1); b is Laoyingyan rock avalanche, which is  
397     river-blocked.



398  
399     Figure 3: Sketch map of a channelized rock avalanche defining geometric parameters. The red-dashed  
400     ellipse indicates the topographic transition dividing the initiated slope, channel and valley floor.  
401     The red arrow represents sliding direction of source mass.



402      **Figure 4: Relationship between horizontal travel distance and volume of channelized rock avalanches.**

403

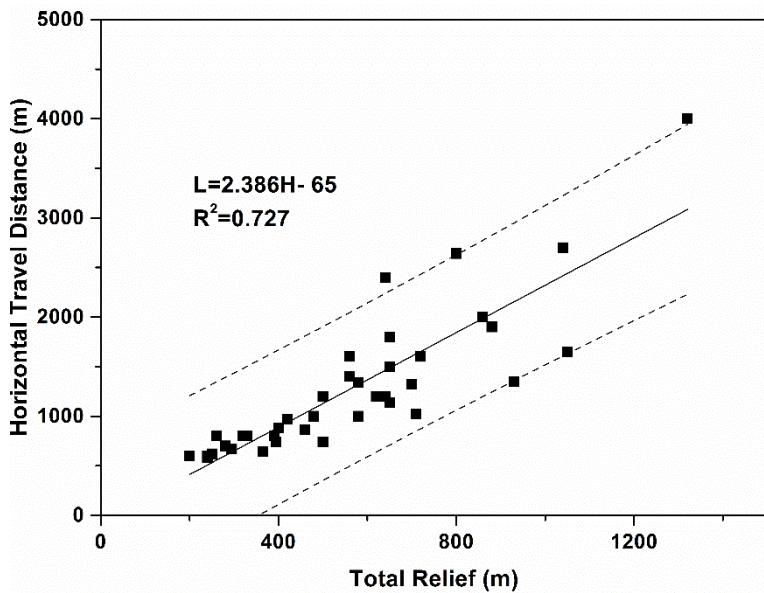


Figure 5: Relationship between horizontal travel distance and total relief of channelized rock avalanche.

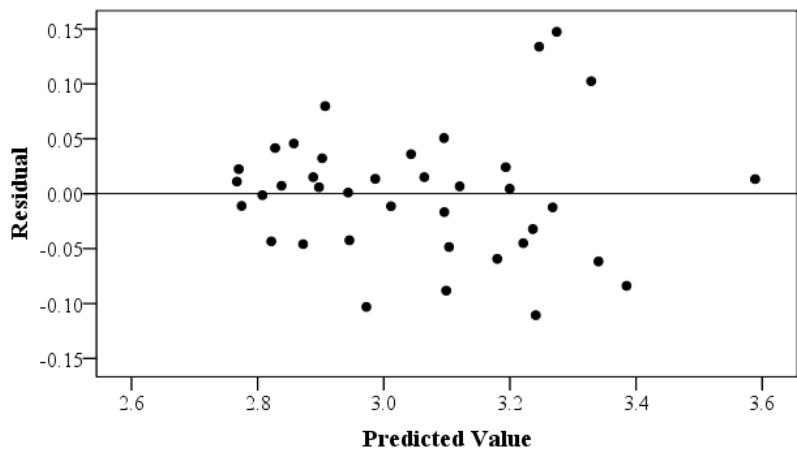


Figure 6: Residual plot for equation (2).

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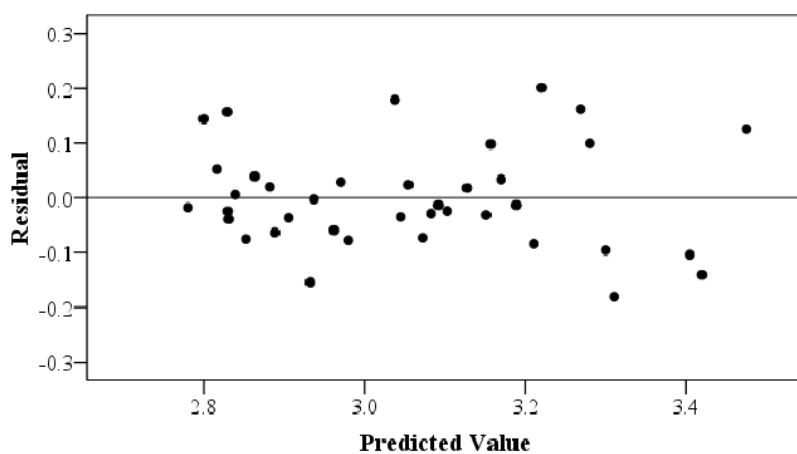
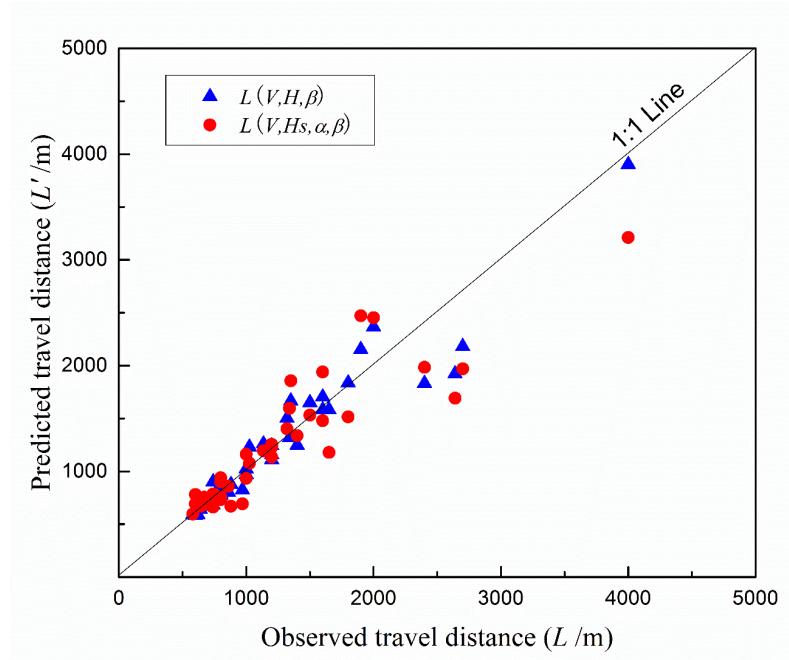


Figure 7: Residual plot for equation (4).

405



**Figure 8:** The comparison between observed and predicted travel distance for the two multivariate regression models.



406

407     Figure 9: Sketch map of flow capacity of channel affecting on the travel distance of Wenjia Gully  
 408     channelized rock avalanche: (a) before the earthquake, (b) after the earthquake, (c) photo taken on  
 409     deposition platform after the earthquake. The red arrow show the sliding direction of source mass. The red  
 410     dotted line in figure.9(a) indicates the original depression on the travel path of the rock avalanche, in where  
 411     debris deposition of about 30 million m<sup>3</sup> was stored after the earthquake (shown in figure.9(b)), and more  
 412     detailed information is shown in the figure.9(c).

**Table 1: Data of various factors for establishment of prediction model of rock avalanche travel distance.**

| Code | Landslide name | Longitude, (°E) | Latitude, (°N) | Landslide area, $A$ ( $m^2$ ) | Landslide volume, $V$ ( $m^3$ ) | Source area height, $H_s$ (m) | Slope angle, $\alpha$ (°) | Channel angle, $\beta$ (°) | Total relief, $H$ (m) | Travel distance, $L$ (m) | Reference       |
|------|----------------|-----------------|----------------|-------------------------------|---------------------------------|-------------------------------|---------------------------|----------------------------|-----------------------|--------------------------|-----------------|
| 1    | Wenjia Gully   | 104.140         | 31.552         | 3000566                       | 50000000                        | 440                           | 26                        | 7                          | 1320                  | 4000                     | Xu et al., 2009 |
| 2    | Shuimo Gully   | 103.981         | 31.442         | 915608                        | 19960000                        | 490                           | 35                        | 10                         | 860                   | 2000                     |                 |
| 3    | Dawuji         | 104.196         | 31.702         | 792190                        | 16330000                        | 540                           | 29                        | 13                         | 880                   | 1900                     |                 |
| 4    | Donghekou      | 105.113         | 32.410         | 1283627                       | 15000000                        | 240                           | 25                        | 11                         | 640                   | 2400                     | Xu et al., 2009 |
| 5    | Hongshigou     | 104.130         | 31.624         | 687520                        | 13410000                        | 290                           | 37                        | 17                         | 1040                  | 2700                     |                 |
| 6    | Woqian         | 104.964         | 32.308         | 695672                        | 12000000                        | 330                           | 30                        | 10                         | 560                   | 1600                     | Xu et al., 2009 |
| 7    | Xiaojiashan    | 104.038         | 31.465         | 465899                        | 7810000                         | 480                           | 48                        | 24                         | 930                   | 1350                     |                 |
| 8    | Niumian Gully  | 103.456         | 31.044         | 527700                        | 7500000                         | 320                           | 32                        | 13                         | 800                   | 2640                     | Xu et al., 2009 |
| 9    | Liqi Gully     | 105.207         | 32.169         | 355113                        | 5360000                         | 360                           | 37                        | 12                         | 650                   | 1500                     |                 |
| 10   | Caocaoping     | 104.139         | 31.607         | 354046                        | 5340000                         | 345                           | 31                        | 17                         | 580                   | 1340                     |                 |
| 11   | Huoshi Gully   | 104.134         | 31.616         | 322155                        | 4680000                         | 270                           | 38                        | 17                         | 700                   | 1320                     |                 |
| 12   | Shibangou      | 105.090         | 32.419         | 496983                        | 4500000                         | 450                           | 34                        | 9                          | 650                   | 1800                     | Xu et al., 2009 |
| 13   | Xiejadianzi    | 103.841         | 31.298         | 294256                        | 4000000                         | 400                           | 34                        | 15                         | 720                   | 1600                     | Xu et al., 2009 |
| 14   | Dashui Gully   | 103.675         | 31.199         | 241874                        | 3150000                         | 320                           | 30                        | 17                         | 560                   | 1400                     |                 |
| 15   | Changping      | 103.754         | 31.259         | 224645                        | 2840000                         | 290                           | 37                        | 16                         | 500                   | 1200                     |                 |
| 16   | Xiaomuling     | 104.102         | 31.613         | 218704                        | 2740000                         | 175                           | 45                        | 26                         | 710                   | 1025                     |                 |
| 17   | Baishuling     | 104.385         | 31.807         | 208968                        | 2570000                         | 335                           | 36                        | 20                         | 620                   | 1200                     |                 |
| 18   | Dawan          | 104.536         | 31.907         | 203959                        | 2480000                         | 220                           | 28                        | 20                         | 480                   | 1000                     |                 |
| 19   | Xiaojiashan    | 104.182         | 31.486         | 198165                        | 2385499                         | 340                           | 44                        | 20                         | 650                   | 1135                     |                 |
| 20   | Shicouzi       | 104.918         | 32.243         | 169540                        | 1920000                         | 260                           | 30                        | 26                         | 640                   | 1200                     |                 |

| 21 | Changtan       | 104.133 | 31.508 | 151094 | 1640000 | 400 | 33 | 25 | 1050 | 1650 |  |
|----|----------------|---------|--------|--------|---------|-----|----|----|------|------|--|
| 22 | Hongmagong     | 104.962 | 32.301 | 144683 | 1540000 | 195 | 30 | 14 | 330  | 800  |  |
| 23 | Baiguocun      | 105.088 | 32.385 | 139800 | 1470000 | 165 | 26 | 12 | 260  | 800  |  |
| 24 | Qinglongcun    | 105.036 | 32.342 | 134079 | 1390000 | 90  | 21 | 11 | 200  | 600  |  |
| 25 | Pengjiashan    | 104.546 | 31.930 | 127156 | 1290000 | 200 | 33 | 28 | 580  | 1000 |  |
| 26 | Longwancun     | 104.571 | 31.922 | 99821  | 920000  | 205 | 31 | 28 | 460  | 860  |  |
| 27 | Zhangzhengbo   | 105.017 | 32.333 | 99726  | 920000  | 125 | 29 | 15 | 320  | 800  |  |
| 28 | Dujianyan      | 105.028 | 32.336 | 94769  | 860000  | 100 | 33 | 17 | 400  | 880  |  |
| 29 | Madiping       | 104.996 | 32.355 | 94632  | 860000  | 140 | 27 | 31 | 395  | 740  |  |
| 30 | Yandiaowo      | 105.099 | 32.391 | 92128  | 820000  | 145 | 30 | 26 | 390  | 800  |  |
| 31 | Chuangzi Gully | 104.085 | 31.518 | 91717  | 820000  | 185 | 35 | 15 | 295  | 670  |  |
| 32 | Zhaojiashan    | 105.041 | 32.342 | 82329  | 700000  | 115 | 22 | 16 | 280  | 700  |  |
| 33 | Weiziping      | 105.083 | 32.387 | 74661  | 620000  | 135 | 22 | 18 | 240  | 600  |  |
| 34 | Maochongshan 2 | 104.908 | 32.243 | 70251  | 570000  | 160 | 38 | 22 | 500  | 740  |  |
| 35 | Waqianshan     | 105.049 | 32.376 | 70007  | 560000  | 135 | 24 | 18 | 250  | 620  |  |
| 36 | Muhongping     | 104.982 | 32.291 | 68288  | 540000  | 175 | 28 | 20 | 420  | 970  |  |
| 37 | Dapingshang    | 104.542 | 31.889 | 65700  | 520000  | 160 | 34 | 29 | 365  | 640  |  |
| 38 | Liushuping 2   | 105.054 | 32.365 | 54810  | 400000  | 150 | 29 | 16 | 240  | 580  |  |

**Table 2: Correlation coefficients of continuous variables listed in Table 1.**

|                            | <b>A</b> | <b>V</b> | <b>H</b> | <b>Hs</b> | <b><math>\alpha</math></b> | <b><math>\beta</math></b> | <b>L</b> |
|----------------------------|----------|----------|----------|-----------|----------------------------|---------------------------|----------|
| <b>A</b>                   | 1.000    | 0.982    | 0.674    | 0.521     | -0.119                     | -0.524                    | 0.877    |
| <b>V</b>                   | —        | 1.000    | 0.713    | 0.560     | -0.055                     | -0.492                    | 0.866    |
| <b>H</b>                   | —        | —        | 1.000    | 0.801     | 0.429                      | -0.130                    | 0.857    |
| <b>Hs</b>                  | —        | —        | —        | 1.000     | 0.399                      | -0.323                    | 0.675    |
| <b><math>\alpha</math></b> | —        | —        | —        | —         | 1.000                      | 0.264                     | 0.082    |
| <b><math>\beta</math></b>  | —        | —        | —        | —         | —                          | 1.000                     | -0.467   |
| <b>L</b>                   | —        | —        | —        | —         | —                          | —                         | 1.000    |

Note: The number in Italics indicates the two variables are not significantly correlated

**Table 3: The regression coefficients and results of significance tests of two multivariate regression models.**

| Equations                       | Coefficients* | Intercept | Coefficient of log(V) | Coefficient of log(H) | Coefficient of log(tan $\beta$ ) | Coefficient of log(Hs) | Coefficient of log(tan $\beta$ ) | Adjusted R <sup>2</sup> | F-stat | F <sub>0.05</sub> |
|---------------------------------|---------------|-----------|-----------------------|-----------------------|----------------------------------|------------------------|----------------------------------|-------------------------|--------|-------------------|
| Best-fit regression equation    | LCI           | 0.175     | -0.013                | 0.521                 | -0.548                           | —                      | —                                | 0.933                   | 173.5  | 2.883             |
|                                 | Mean          | 0.420     | 0.079                 | 0.718                 | -0.365                           | —                      | —                                |                         |        |                   |
|                                 | UCI           | 0.665     | 0.171                 | 0.914                 | -0.182                           | —                      | —                                |                         |        |                   |
| Alternative regression equation | LCI           | 0.110     | 0.199                 | —                     | -0.165                           | -0.002                 | -0.464                           | 0.840                   | 49.5   | 2.659             |
|                                 | Mean          | 0.561     | 0.303                 | —                     | 0.072                            | 0.244                  | -0.115                           |                         |        |                   |
|                                 | UCI           | 1.012     | 0.407                 | —                     | 0.308                            | 0.489                  | 0.233                            |                         |        |                   |

5 Note: “Coefficients” of each variable has three kinds: LCI is lower bound of the coefficients with 95% confidence; Mean is the mean value of the coefficients; UCI is upper bound of the coefficients with 95% confidence;

**Table 4: Background parameters and predicted values of 8 rock avalanches in the same area used for validation**

| Landslide name | Longitude | Latitude | Triggers * | V /10 <sup>4</sup> m <sup>3</sup> | $\alpha$ /° | B /° | Hs /m | H /m | L /m | L'^(3) ** / % | Error /m | L'^(4) *** / % | Error / % |
|----------------|-----------|----------|------------|-----------------------------------|-------------|------|-------|------|------|---------------|----------|----------------|-----------|
| Pianqiaozi     | 104.370   | 31.822   | WCEQ       | 8.8                               | 35          | 19   | 153   | 205  | 372  | 436           | 17.2     | 373            | 0.3       |
| Yangjiayan     | 104.328   | 31.755   | WCEQ       | 25.4                              | 41          | 23   | 164   | 304  | 518  | 583           | 12.5     | 518            | 0.1       |
| Shanshulin     | 103.508   | 31.181   | WCEQ       | 27.9                              | 34          | 25   | 340   | 433  | 715  | 731           | 2.3      | 660            | -7.6      |
| Fuyangou       | 103.501   | 31.422   | WCEQ       | 71.9                              | 38          | 28   | 385   | 530  | 763  | 869           | 13.8     | 900            | 17.9      |

|            |         |        |      |     |    |    |     |     |      |      |       |     |       |
|------------|---------|--------|------|-----|----|----|-----|-----|------|------|-------|-----|-------|
| Dayanbeng1 | 102.762 | 30.179 | LSEQ | 100 | 53 | 10 | 254 | 424 | 1267 | 1136 | -10.3 | 781 | -38.4 |
| Dayanbeng2 | 102.761 | 30.178 | LSEQ | 110 | 50 | 8  | 237 | 407 | 1372 | 1208 | -12.0 | 787 | -42.6 |
| Ermanshan  | 102.739 | 29.322 | RF   | 100 | 33 | 15 | 148 | 635 | 1370 | 1303 | -4.9  | 767 | -44.0 |
| Wulipo     | 103.567 | 30.919 | RF   | 150 | 30 | 10 | 135 | 377 | 1260 | 1078 | -14.4 | 833 | -33.9 |

Note: “Triggers” is the triggering condition of rock avalanches; “WCEQ” represents the 2008 Wenchuan Ms7.8 earthquake; “LSEQ” represents the 2013 Lushan Ms7.0 earthquake; “RF” represents the rock avalanche was induced by heavy rainfall.  $L'_{(3)}$ ,  $L'_{(4)}$  indicates the predicted travel distance estimated by using equation (3) and (4) respectively.