

# Inferring potential landslide damming using slope stability, geomorphic constraints and run-out analysis; case study from the NW Himalaya

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## 1 ABSTRACT

2 Prediction of potential landslide damming has been a difficult process owing to the  
3 uncertainties related to landslide volume, resultant dam volume, entrainment, valley  
4 configuration, river discharge, material composition, friction, and turbulence associated with  
5 material. In this study, instability pattern of landslides, geomorphic indices, post failure run-  
6 out predictions, and spatio-temporal patterns of rainfall and earthquakes are explored to predict  
7 the potential landslide damming sites. The Satluj valley, NW Himalaya is chosen as a case  
8 study area. The study area has witnessed landslide damming in the past and incurred losses of  
9 \$ ~30M and 350 lives in the last four decades due to such processes. Forty-four active  
10 landslides that cover a total  $\sim 4.81 \pm 0.05 \times 10^6 \text{ m}^2$  area and  $\sim 34.1 \pm 9.2 \times 10^6 \text{ m}^3$  volume are  
11 evaluated to identify those landslides that may result in potential landslide damming. Out of  
12 forty-four, five landslides covering a total volume of  $\sim 26.3 \pm 6.7 \times 10^6 \text{ m}^3$  are noted to form the  
13 potential landslide dams. Spatio-temporal variations in the pattern of rainfall in the recent years  
14 enhanced the possibility of landslide triggering and hence of potential damming. These five  
15 landslides also revealed  $24.8 \pm 2.7\text{m}$  to  $39.8 \pm 4.0\text{m}$  high debris flows in the run-out predictions.

16 **Key words:** Landslide damming, Slope stability; Run-out; Himalaya

17 **1.0 INTRODUCTION**

18 Landslide damming is a normal geomorphic process in narrow river valleys and poses  
19 significant natural hazard (Dai et al. 2005; Gupta and Sah 2008; Delaney and Evans 2015; Fan  
20 et al. 2020). Many studies have explored damming characteristics (Li et al. 1986; Costa and  
21 Schuster 1988; Takahashi and Nakawaga 1993; Ermini and Casagli 2003; Fujisawa et al. 2009;  
22 Stefanelli et al. 2016; Kumar et al. 2019a). However, studies concerning the prediction of  
23 potential landslide dams and their stability at regional scale have been relatively rare,  
24 particularly in Himalaya despite a history of landslide damming and flash floods (Gupta and  
25 Sah 2008; Ruiz-Villanueva et al. 2016; Kumar et al. 2019a). In order to identify the landslides  
26 that have potential to form dams, the following factors have been main requisites; (i) pre- and  
27 post-failure behaviour of landslide slopes, and (ii) landslide volume, stream power, and  
28 morphological setting of the valley (Kumar et al. 2019a).

29 To understand the pre-failure pattern, Finite Element Method (FEM) based slope stability  
30 evaluation has been among the most widely used approaches for complex slope geometry  
31 (Griffiths and Lane 1999; Jing 2003; Jamir et al. 2017; Kumar et al. 2018). However, the  
32 selection of input parameters in FEM analysis and the set of assumptions used (material model,  
33 failure criteria, and convergence) may also result in uncertainty in the final output (Wong 1984;  
34 Cho 2007; Li et al. 2016). Uncertainty from input parameters can be resolved by performing  
35 parametric analysis, whereas the utilization of most appropriate criteria can minimize the  
36 uncertainty caused by assumptions. Post-failure behavior of landslides can be understood using  
37 run-out analysis (Hung et al. 1984; Hutter et al. 1994; Rickenmann and Scheidl 2013). These  
38 methods could be classified into empirical/statistical and dynamical categories (Rickenmann  
39 2005). Owing to the flexibility in rheology, solution approach, reference frame, and  
40 entrainment, dynamic models have been relatively more realistic for site-specific problems  
41 (Corominas and Mavrouli 2011). Though the different numerical models have different  
42 advantages and limitations, Voellmy rheology (friction and turbulence) (Voellmy 1955; Salm  
43 1993) based Rapid Mass Movement Simulation (RAMMS) (Christen et al. 2010) has been used  
44 widely owing to the inclusion of rheological and entrainment rate flexibility.

45 Apart from the pre and post-failure pattern, landslide volume, stream power and morphological  
46 setting of the valley are crucial to infer the potential landslide damming. The Morphological  
47 Obstruction Index (MOI) and Hydro-morphological Dam Stability Index (HDSI) have been

48 widely used geomorphic indices to infer the potential of landslide dam formation and their  
49 temporal stability (Costa and Schuster 1988; Ermini and Casagli 2003; Stefanelli et al. 2016).

50 The NW Himalaya has been one of most affected terrains by the landslides owing to the active  
51 tectonics and multiple precipitation sources i.e., Indian Summer Monsoon (ISM) and Western  
52 Disturbance (Dimri et al. 2015). The NW Himalaya has accommodated ~51 % of all the  
53 landslides in India during yrs. 1800-2011 (Parkash 2011). The Satluj River valley, NW  
54 Himalaya is one such region where landslides and associated floods have claimed ~350 lives  
55 and resulted in the loss of minimum 30 million USD in the last four decades. This region holds  
56 a high potential for future landslide damming and resultant floods (Ruiz-Villanueva et al. 2016;  
57 Kumar et al. 2019a). Therefore, the Satluj valley is taken as a case study area, and 44 active  
58 landslides belonging to the different litho-tectonic regimes are modeled using the FEM  
59 technique. Multiple slope sections and a range of values of different input parameters are used  
60 to perform the parametric study. In order to determine the human population that might be  
61 affected by these landslides, census statistics are also used. The MOI and HDSI are used to  
62 determine the potential of landslide dam formation and their stability, respectively. In view of  
63 the role of rainfall and earthquakes as main landslide triggering factors, the spatio-temporal  
64 regime of these two factors is also discussed. Run-out prediction of certain landslides is also  
65 performed to understand the role of run-out in the potential landslide damming. This study  
66 provides detailed insight into the regional instability pattern, associated uncertainty, and  
67 potential landslide damming sites and hence it can be replicated in other hilly terrain witnessing  
68 frequent landslides and damming.

## 69 **2.0 STUDY AREA**

70 The study area is located between the Moorang (31°36'1" N, 78°26' 47" E) and Rampur town  
71 (31°27'10" N, 77°38' 20" E) in the Satluj River valley, NW Himalaya (Fig. 1). The Satluj River  
72 flows across the Tethyan Sequence (TS), Higher Himalaya Crystalline (HHC), Lesser  
73 Himalaya Crystalline (LHC), and Lesser Himalaya Sequence (LHS). The TS in the study area  
74 comprises slate/phyllite and schist and has been intruded by the biotite-rich granite i.e.,  
75 Kinnaur-Kailash Granite (KKG) near the Sangla Detachment (SD) fault (Sharma 1977; Vannay  
76 et al. 2004). The SD fault separates the TS from the underlying crystalline rockmass of the  
77 HHC. Migmatitic gneiss marks the upper part of the HHC, whereas the base is marked by the  
78 kyanite-sillimanite gneiss rockmass (Sharma 1977; Vannay et al. 2004; Kumar et al. 2019b).  
79 The Main Central Thrust (MCT) fault separates the HHC from the underlying schist/gneissic

80 rockmass of the LHC. The LHC comprises mica schist, carbonaceous schist, quartzite, and  
81 amphibolite. A thick zone of gneiss i.e., Wangtu Gneissic Complex (WGC) is exposed in the  
82 LHC, which comprises augen gneiss and porphyritic granitoids. The LHC is delimited at the  
83 base by the Muniari Thrust (MT) fault that is thrusted over the Lesser Himalaya Sequence  
84 (LHS) rockmass. The MT contains breccia, cataclastic, and fault gouge (Sharma 1977; Vannay  
85 et al. 2004; Kumar et al. 2019b). The LHS in the study area consists of quartz-arenite (Rampur  
86 Quartzite) with bands of phyllite, meta-volcanics, and paragneiss (Sharma 1977).

87 The present study covers forty-four active landslides (20 debris slides, 13 rock falls, and 11  
88 rock avalanches) along the study area (Table 1) that have been mapped recently by Kumar et  
89 al. (2019b). Field photographs of some of these landslides are presented in Fig. 2. The TS and  
90 LHS in the study area have been subjected to relative tectonic tranquility with exhumation rates  
91 as low as 0.5 - 1.0 mm/yr, whereas the HHC and LHC region have undergone 1.0 - 4.5 mm/yr  
92 rate of exhumation (Thiede et al. 2009). The MCT fault region and the WGC are noted to have  
93 maximum exhumation rate (i.e., ~4.5 mm/yr) that is evident from the deep gorges in these  
94 regions (Fig. 2c, 2e). A majority of the earthquake events in the study area in the last 7 decades  
95 have been related to the N-S oriented Kaurik - Chango Fault (KCF) (Kundu et al. 2014;  
96 Hazarika et al. 2017; International Seismological Centre Catalogue 2019). The climate in the  
97 study area shows a spatial variation from humid (~800 mm/yr mean annual precipitation) in  
98 the LHS to the semi-arid (~200 mm/yr) in the TS (Kumar et al. 2019b). The HHC acts as a  
99 transition zone where climate varies from semi-humid to semi-arid in the SW-NE direction.  
100 This transition has been attributed to the ‘orographic barrier’ nature of the HHC that marks the  
101 region in its north as ‘orographic interior’ and the region to its south as the ‘orographic front’  
102 (Wulf et al. 2012; Kumar et al. 2019b).

103 Landslides in the study area have been a consistent threat to the socio-economic condition of  
104 the nearby human population (Gupta and Sah 2008; Ruiz-Villanueva et al. 2016; Kumar et al.  
105 2019a). Therefore, the human population in the vicinity of each landslide was also determined  
106 by considering the nearby villages/town. Notably, a total of 25,822 people reside within 500 m  
107 extent of the 44 landslide slopes, and about 70 % of this population is residing in the reach of  
108 debris slide type landslides. Since the Govt. of India follows a 10 year gap in census statistics,  
109 the human population data was based on last official data i.e., Census-2011. The next official  
110 census is due in year 2021. The population density in the Indian Himalayan region was  
111 estimated to be 181/km<sup>2</sup> in the year 2011 that might grow to 212/km<sup>2</sup> in 2021 with a decadal

112 growth rate of 17.3% (<https://censusindia.gov.in>, retrieved on 02 Sep 2020;  
113 <http://gbpihedenvis.nic.in>, retrieved on 02 Sep 2020).

114 **3.0 METHODOLOGY**

115 The methodology involved field data collection, satellite imagery analysis, laboratory analyses,  
116 slope stability modelling, geomorphic indices, rainfall/earthquake pattern and run-out  
117 modelling. Details are as follows;

118 *3.1 Field data, satellite imagery processing, and laboratory analyses*

119 The field work involved rock/soil sample collection from each landslide location, rockmass  
120 joint mapping, and N-type Schmidt Hammer Rebound (SHR) measurement. Joints were  
121 included in the slope models for the FEM based slope stability analysis. The dataset involving  
122 the joint details is available in the data repository (Kumar et al. 2021). The SHR values were  
123 obtained as per International Society of Rock Mechanics (ISRM) standard (Aydin 2008).  
124 Cartosat-1 satellite imagery and field assessment were used to finalize the location of slope  
125 sections (2D) of the landslides. Cartosat-1 imagery has been used widely for the landslide  
126 related studies (Martha et al. 2010). The Cartosat-1 Digital Elevation Model (DEM) having  
127 10m spatial resolution, prepared using the Cartosat-1 stereo imagery, was used to extract the  
128 slope sections of the landslides using the Arc GIS-10.2 software. Details of the satellite imagery  
129 are mentioned in Table 2.

130 The rock/soil samples were analyzed in the National Geotechnical Facility (NGF) and Wadia  
131 Institute of Himalayan Geology (WIHG) laboratory, India. The rock samples were drilled and  
132 smoothed for Unconfined Compressive Strength (UCS) (IS: 9143-1979) and ultrasonic tests  
133 (CATS Ultrasonic (1.95) of Geotechnical Consulting & Testing Systems). The ultrasonic test  
134 was conducted to determine the density, elastic modulus, and Poisson's ratio of rock samples.  
135 The soil samples were tested for grain size (IS: 2720-Part 4-1985), UCS test (IS: 2720-Part 10-  
136 1991), and direct shear test (IS: 2720-Part 13- 1986). If the soil samples contained < 5% fines  
137 (< 75 mm), the hydrometer test was not performed for the remaining fine material. In the direct  
138 shear test, soil samples were sheared under the constant normal stress of 50, 100 and 150  
139 kN/m<sup>2</sup>. The UCS test of soil was performed under three different rates of movements i.e., 1.25  
140 mm/min, 1.50 mm/min and 2.5 mm/min.

141 *3. 2 Slope stability modelling*

142 The Finite Element Method (FEM) was used along with the Shear Strength Reduction (SSR)  
143 technique to infer the critical Strength Reduction Factor (SRF), Shear Strain (SS), and Total  
144 Displacement (TD) in the 44 landslide slopes using the RS2 software. The SRF has been  
145 observed to be similar in nature to the Factor of Safety (FS) of the slope (Zienkiewicz et al.  
146 1975; Griffiths and Lane 1999). To define the failure in the SSR approach, non-convergence  
147 criteria were used (Nian et al. 2011). The boundary condition with the restraining movement  
148 was applied to the base and back, whereas the front face was kept free for the movement (Fig.  
149 3). In-situ field stress was adjusted in view of dominant stress i.e., extension or compression,  
150 by changing the value of the coefficient of earth pressure ( $k$ ). A value of  $k = \sigma_h/\sigma_v = 0.5$  was  
151 used in extensional regime, whereas  $k = \sigma_h/\sigma_v = 1.5$  was used in compressional regime. The  
152 Tethyan Sequence has been observed to possess the NW-SE directed extensional regime. The  
153 structures in the upper part of the HHC are influenced by the east directed extension along the  
154 SD fault. The lower part, however, is characterized by the SW directed compression along the  
155 Main Central Thrust. In contrast to the HHC, structures in the Lesser Himalaya Crystalline and  
156 Munsiari Thrust region are influenced by the compressional regime. In the Lesser Himalaya  
157 Sequence region, the SW directed compressional regime has been observed on the basis of the  
158 SW verging folds, crenulation cleavage, and other features (Vannay et al. 2004).

159 The soil and rock mass were used in the models through the Mohr-Coulomb (M-C) failure  
160 criterion (Coulomb 1776; Mohr 1914) and Generalized Hoek-Brown (GHB) criterion (Hoek et  
161 al. 1995), respectively. The parallel- statistical distribution of the joints with normal-distributed  
162 joint spacing in the rock mass was applied through the Barton-Bandis (B-B) slip criterion  
163 (Barton and Choubey 1977; Barton and Bandis 1990). Plane strain triangular elements having  
164 6 nodes were used through the graded mesh in the models. Details of the criteria used in the  
165 FEM analysis are mentioned in Table 3. The dataset of input parameters used in the FEM  
166 analysis is available in the data repository (Kumar et al. 2021). It is to note that the FEM  
167 analysis was performed under the static load i.e., field stress and body force. The dynamic  
168 analysis was not performed, at present, in absence of any major seismic events in the region in  
169 the last 4 decades (sec. 4.3) and lack of reliable dynamic load data of nearby major seismic  
170 events.

171 To understand the uncertainty caused by the selection of 2D slope section, multiple slope  
172 sections were taken, wherever possible. More than one slope section was modeled for each  
173 debris slide, whereas for the rock falls/ rock avalanche only one slope section was chosen due  
174 to the limited width of the rock falls/rock avalanche in the study area. To find out the relative

175 influence of different input parameters on the final output, a parametric study was performed.  
176 In the parametric study for debris slides, Akpa landslide (S.N.5 in Fig. 3), Pangi landslide  
177 (S.N.13 in Fig. 3), and Barauni Gad landslide (S.N.38 in Fig. 3) were chosen, whereas Tirung  
178 khad (S.N.2 in Fig.3) and Chagaon landslide (S.N.21 in Fig. 3) were considered to represent  
179 rock fall. Baren Dogri (S.N.7 in Fig. 3) landslide was used to represent the rock avalanches.  
180 The selection of these landslides for the parametric study was based on the following two  
181 factors; (1) to choose the landslides from different litho-tectonic regime, and (2) to represent  
182 varying stress regime i.e., extensional, compressional, and relatively stagnant. The Parametric  
183 study of the debris slide models involved following 9 parameters; field stress coefficient,  
184 stiffness ratio, cohesion and angle of friction of soil, elastic modulus and Poisson's ratio of soil,  
185 rockmass modulus, Poisson's ratio and uniaxial compressive strength of rock. For the  
186 rockfalls/rock avalanche, the following 6 parameters were considered; uniaxial compressive  
187 strength of rock, rockmass modulus of rock, Poisson's ratio of rock, 'm<sub>i</sub>' parameter, stiffness  
188 ratio, and field stress coefficient. The 'm<sub>i</sub>' is a Generalized Hoek-Brown (GHB) parameter  
189 that is equivalent to the angle of friction of Mohr-coulomb (M-C) criteria.

190 *3. 3 Geomorphic indices*

191 Considering the possibility of landslide dam formation in case of slope failure, the following  
192 geomorphic indices were also used;

193 (i) Morphological Obstruction Index (MOI)

$$194 \quad \text{MOI} = \log (V_l/W_v) \quad \text{Eq. 1}$$

195 (ii) Hydro-morphological Dam Stability Index (HDSI)

$$196 \quad \text{HDSI} = \log (V_d/A_b \cdot S) \quad \text{Eq. 2}$$

197 Where, V<sub>d</sub> (dam volume)= V<sub>l</sub> (landslide volume), m<sup>3</sup>; A<sub>b</sub> is upstream catchment area (km<sup>2</sup>); W<sub>v</sub>  
198 is width of the valley (m) and S is local slope gradient of river channel (m/m). Though the  
199 resultant dam volume could be higher or lower than the landslide volume owing to slope  
200 entrainment, rockmass fragmentation, retaining of material at the slope, and washout by the  
201 river (Hung and Evans 2004; Dong et al. 2011), dam volume is assumed to be equal to  
202 landslide volume for the worst case. By utilizing the comprehensive dataset of ~300 landslide  
203 dams of Italy, Stefanelli et al. (2016) have classified the MOI into (i) non-formation domain:

204 MOI <3.00, (ii) uncertain evolution domain: 3.00 <MOI >4.60, and (iii) formation domain:  
205 MOI >4.60. By utilizing the same dataset, Stefanelli et al. (2016) defined the HDSI into  
206 following categories (i) instability domain: HDSI <5.74, (ii) uncertain determination domain:  
207 5.74<HDSI >7.44, and (iii) Stability domain: HDSI>7.44.

208 *3. 4 Rainfall and Earthquake regime*

209 Precipitation in the study area is related primarily to the Indian Summer Monsoon (ISM) and  
210 Western Disturbance (WD) and varies spatially-temporally due to various local and regional  
211 factors (Gadgil et al. 2007; Hunt et al. 2018). Therefore, we have taken the TRMM\_3B42  
212 (Huffman et al. 2016) daily rainfall data of years 2000-2019 at four different locations;  
213 Moorang, Kalpa, Nachar, and Rampur (Locations mentioned in Fig. 1). The dataset of  
214 earthquake events (2<M<8) in and around study area during the years 1940-2019 was retrieved  
215 from the ISC catalogue (<http://www.isc.ac.uk/iscbulletin/search/catalogue/>, retrieved on 02  
216 March 2020) to determine the spatio-temporal pattern.

217 *3. 5 Run-out modelling*

218 Since the study area has witnessed many disastrous landslides, mostly rainfall triggered, and  
219 flash floods in past (Gupta and Sah 2008; Ruiz-Villanueva et al. 2016), run-out analysis was  
220 performed to understand the post-failure scenario. Such run-out predictions will also be helpful  
221 to ascertain the possibility of damming because various studies have noted river damming by  
222 the debris flows (Li et al. 2011; Braun et al. 2018; Fan et al. 2020). The landslides that have  
223 potential to form dams based on the indices (sec. 3.3) are evaluated for such run-out analysis.

224 In this study, Voellmy rheology (Voellmy 1955; Salm 1993) based Rapid Mass Movement  
225 Simulation (RAMMS) (Christen et al. 2010) model was used to understand the run-out pattern.  
226 The RAMMS for debris flow uses the Voellmy friction law and divides the frictional resistance  
227 into a dry-Coulomb type friction ( $\mu$ ) and viscous-turbulent friction ( $\xi$ ). The frictional resistance  
228  $S$  (Pa) is :

229 
$$S = \mu N + (\rho g u^2)/\xi \quad \text{Eq. 3}$$

230 where  $N = \rho h g \cos(\phi)$  is the normal stress on the running surface,  $\rho$  is density,  $g$  is gravitational  
231 acceleration,  $\phi$  is slope angle,  $h$  is flow height and  $u = (u_x, u_y)$ , consisting of the flow velocity  
232 in the x- and y-directions. In this study, a range of friction ( $\mu$ ) and turbulence ( $\xi$ ) values, apart  
233 from other input parameters, are used to evaluate the uncertainty in output (Table 4). Generally,

234 the values for  $\mu$  and  $\xi$  are determined using the reconstruction of real events through the  
235 simulation and subsequent comparison between the dimensional characteristics of real and  
236 simulated event. However, the landslides in the study area merge with the river floor and/or  
237 are in close proximity and hence there is no failed material left from the previous events to  
238 reconstruct. Therefore, the  $\mu$  and  $\xi$  values were taken from a range in view of topography of  
239 landslide slope and run-out path, landslide material, similar landslide events/material, and  
240 results from previous studies/models (Hürliemann et al. 2008; Rickenmann and Scheidl 2013;  
241 RAMMS v.1.7.0). Since these landslides are relatively deep in nature and during the slope  
242 failure, irrespective of type of trigger, entire loose material might not slide down, the depth of  
243 landslide is taken as only  $\frac{1}{4}$  (thickness) in the run-out calculation. Further, a release area  
244 concept (for unchanneled flow or block release) was used for the run-out simulation. During  
245 the field visits, no specific flow channels (or gullies) were found on the landslide slopes except  
246 a few centimeters deep seasonal flow channels for S. N. 5 and S.N. 15 landslides (Table 1).  
247 However, the data pertaining to the spatial-temporal pattern of discharge at these two landslides  
248 was not available. Therefore, the release area concept was chosen because it has been more  
249 appropriate when the flow path (e.g. gully) and its possible discharge on the slope is uncertain  
250 (RAMMS v.1.7.0).

## 251 **4.0 RESULTS**

### 252 *4.1 Slope instability regime and parametric output*

253 Out of the 44 landslides studied here, 31 are in meta-stable state ( $1 \leq FS \leq 2$ ) and 13 in unstable  
254 state ( $FS < 1$ ) (Fig. 4). Most of the unstable landslides are debris slides, whereas the majority  
255 of the meta-stable landslides are rock fall/rock avalanche. Debris slides constitute  $\sim 90\%$  and  
256  $\sim 99\%$  of the total area and volume, respectively, of the unstable landslides. Notably, about  
257  $\sim 70\%$  of the total human population along the study area resides in the vicinity ( $\sim 500$  m) of  
258 these unstable debris slides (Fig. 4). Rock falls/Rock avalanches constitute  $\sim 84\%$  and  $\sim 78\%$   
259 of the area and volume, respectively, of the meta-stable landslides. Out of total 20 debris slides,  
260 12 debris slides are found to be in unstable stage, whereas 8 in the meta-stable condition (Fig.  
261 4). These 20 debris slides occupy  $\sim 1.9 \pm 0.02 \times 10^6$  m<sup>2</sup> area and  $\sim 26 \pm 6 \times 10^6$  m<sup>3</sup> volume. When  
262 comparing the Factor of Safety (FS) with the Total Displacement (TD) and Shear Strain (SS),  
263 nonlinear poor correlation is achieved (Fig. 5). Since the TD and SS are a relatively good  
264 correlation (Fig. 5), only the TD is used further along with the FS. The TD ranges from  $7.4 \pm$   
265  $8.9$  cm to  $95.5 \pm 10$  cm for the unstable debris slides and  $\sim 18.8$  cm for meta-stable landslides

266 (Fig. 4). Out of 13 rockfalls, 1 belongs to the unstable state and 12 to the meta-stable state (Fig.  
267 4). The TD varies from 0.4 to 80 cm with the maximum for Bara Kamba rockfall (S.N. 31).  
268 Out of 11 rock avalanches, 1 belongs to the unstable state and 10 to the meta-stable state (Fig.  
269 4). The TD varies from 6.0 to 132.0 cm with the maximum for the Kandar rock avalanche (S.N.  
270 25). Relatively higher TD is obtained by the rock fall and rock avalanche of the Lesser  
271 Himalaya Crystalline region (Fig. 4). The landslides of the Higher Himalaya Crystalline  
272 (HHC), Kinnaur Kailash Granite (KKG) and Tethyan Sequence (TS), despite being only 17  
273 out of the total 44 landslides, constituted ~ 67 % and ~ 82 % of the total area and total volume  
274 of the landslides.

275 The Factor of Safety (FS) of debris slides is found to be relatively less sensitive to the change  
276 in the value of input parameters than the Total Displacement (TD) (Fig. 6). In case of Akpa  
277 (Fig. 6a) and Pangi landslide (Fig. 6b), soil friction and field stress have more influence on the  
278 FS. However, for TD, field stress, elastic modulus and Poisson's ratio of the soil are relatively  
279 more controlling parameters. The FS and TD of the Barauni Gad landslide (Fig. 6c) are  
280 relatively more sensitive to soil cohesion and 'mi' parameter. Therefore, it can be inferred that  
281 the FS of debris slides is more sensitive to soil friction and field stress, whereas TD is mostly  
282 controlled by the field stress and deformation parameters i.e, elastic modulus and Poisson's  
283 ratio. Similar to the debris slides, the FS of rock falls and rock avalanche are found to be  
284 relatively less sensitive than TD to the change in the value of input parameters (Fig. 7). Tirung  
285 Khad rock fall (Fig. 7a) and Baren Dogri rock avalanche (Fig. 7b) show dominance of 'mi'  
286 parameter and field stress in the FS as well as in TD. In case of Chagaon rock fall (Fig. 7c),  
287 Poisson's ratio and UCS have relatively more influence on FS and TD. Thus, it can be inferred  
288 that the rock fall/rock avalanche are more sensitive to 'mi' parameter and field stress.

#### 289 *4.2 Potential landslide damming*

290 Based on the MOI, out of total 44 landslides, 5 (S.N. 5, 7, 14, 15, 19) are observed to be in the  
291 formation domain, 15 in uncertain domain, and 24 in non-formation domain (Fig. 8a). The five  
292 landslides that have potential to dam the river in case of slope failure comprise  $\sim 26.3 \pm 6.7 \times$   
293  $10^6 \text{ m}^3$  volume (Fig. 9 a-e). In terms of temporal stability (or durability), out of these five  
294 landslides, only one landslide (S.N. 5) is noted to attain the 'uncertain' domain, whereas the  
295 remaining four show 'instability' (Fig. 8b,d). The lacustrine deposit in the upstream of Akpa  
296 landslide (S.N. 5) in Fig. 9a shows signs of landslide damming in the past (Fig. 10). The  
297 'uncertain' temporal stability indicates that the landslide dam may be stable or unstable

298 depending upon the stream power and landslide volume, which in turn are dynamic factors and  
299 may change owing to the changing climate and/or tectonic event. The landslides that have been  
300 observed to form the landslide dam but are noted to be in temporally unstable category (S.N.  
301 7, 14, 15, 19) are still considerable owing to the associated risks of lake-impoundment and  
302 generation of secondary landslides. Urni landslide (S.N. 19) (Fig. 9e) that damaged the part of  
303 National Highway road (NH)-05 has already partially dammed the river since year 2016 and  
304 holds potential for the further damming (Kumar et al. 2019a). Apart from the S.N. 5 and S.N.  
305 19 landslides, remaining landslides (S.N. 7, 14, 15) belong to the Higher Himalaya Crystalline  
306 (HHC) region that has been observed to accommodate many landslide dams and subsequent  
307 flash floods events in the geological past (Sharma et al. 2017).

308 *4.3 Rainfall and Earthquake regime*

309 In order to explain the spatio-temporal variation in rainfall, the topographic profile of the study  
310 area is also plotted along with the rainfall variation (Fig. 11a). The temporal distribution of  
311 rainfall is presented at annual, monsoonal i.e., Indian Summer Monsoon (ISM): June-  
312 September and non-monsoonal i.e., Western Disturbance (WD): Oct-May (Fig. 11b-d) level.  
313 Rainfall data of the years 2000-2019 revealed a relative increase in the annual rainfall since the  
314 year 2010 (Fig. 11b). The Kalpa region (orographic barrier) received relatively more annual  
315 rainfall than the Rampur, Nachar and Moorang regions throughout the time period, except the  
316 year 2017. The rainfall dominance at Kalpa is more visible in the non-monsoonal season (Fig.  
317 11d). This difference may be due to the orographic influence on the saturated winds of the WD  
318 (Dimri et al. 2015). Further, the rainfall during the monsoon season that was dominant at the  
319 Rampur region till year 2012 gained dominance at Kalpa region since the year 2013 (Fig. 11c).

320 Extreme rainfall events of June 2013 that resulted in the widespread slope failure in the NW  
321 Himalaya also caused landslide damming at places (National Disaster Management Authority,  
322 Govt. of India, 2013; Kumar et al. 2019a). Similar to the year 2013, the years 2007, 2010 and  
323 2019 also witnessed enhanced annual rainfall and associated flash floods and/or landslides in  
324 the region (hpenvis.nic.in, retrieved on March 1, 2020; sandrp.in, retrieved on March 1, 2020).  
325 However, the contribution of the ISM and WD associated rainfall was variable in those years  
326 (Fig. 11). Such frequent but inconsistent rainfall events that possess varied (temporally)  
327 dominance of the ISM and WD are noted to owe their occurrence to the El-Nino Southern  
328 Oscillation (ENSO), Equatorial Indian Ocean Circulation (EIOC), and planetary warming  
329 (Gadgil et al. 2007; Hunt et al. 2018). The orographic setting is noted to act as a main local

330 factor as evident from the relatively more rainfall (total precipitation=1748±594 mm/yr.) at  
331 Kalpa region (orographic barrier) in the non-monsoon and monsoon season from the year 2010  
332 onwards (Fig. 11). Prediction of the potential landslide damming sites in the region revealed  
333 that four (S.N. 7, 14, 15, 19) out of five landslides that are predicted to be able to form dams  
334 belong to this orographic barrier region. Therefore, in view of the prevailing rainfall trend since  
335 the year 2010, regional factors, discussed above, and orographic setting, precipitation triggered  
336 slope failure events may be expected in the future. Such slope failure events, if they occur, at  
337 the predicted landslide damming sites may certainly dam the river.

338 The seismic pattern revealed that the region has been hit by 1662 events during the years 1940-  
339 2019 with the epicenters located in and around the study area (Fig. 12a). However, ~99.5 % of  
340 these earthquake events had a magnitude of less than 6.0 and only 8 events are recorded in the  
341 range of 6.0 to 6.8 M<sub>s</sub> (International Seismological Centre 2019). Out of these 8 events, only  
342 one event i.e., at 6.8 M<sub>s</sub> (19<sup>th</sup> Jan. 1975), has been noted to induce widespread slope failures in  
343 the study area (Khattri et al. 1978). The majority of the earthquake events in the study area  
344 occurred in the vicinity of the N-S oriented trans-tensional Kaurik - Chango Fault (KCF) that  
345 accommodated the epicenter of 19<sup>th</sup> Jan. 1975 earthquake (Hazarika et al. 2017;  
346 <http://www.isc.ac.uk/iscbulletin/search/catalogue/>, retrieved on 02 March 2020). About 95%  
347 of the total 1662 events had their focal depth within 40 km (Fig. 12b). Such a relatively low  
348 magnitude - shallow seismicity in the region has been related to the Main Himalayan Thrust  
349 (MHT) decollement as a response to the relatively low convergence (~14±2 mm/yr) of India  
350 and Eurasia plates in the region (Bilham 2019) (Fig. 12c). Further, the arc (Himalaya)-  
351 perpendicular Delhi-Haridwar ridge that is under thrusting the Eurasian plate in this region has  
352 been observed to be responsible for the spatially varied *low* seismicity in the region (Hazarika  
353 et al. 2017). Thus, though the study area has been subjected to frequent earthquakes, chances  
354 of earthquake-triggered landslides have been relatively low in comparison to rainfall-triggered  
355 landslides and associated landslide damming. For this reason and the lack of reliable dynamic  
356 load of major earthquake event, we have performed the *static* modelling in the present study.  
357 However, we intend to perform the *dynamic* modelling in the near future if the reliable dynamic  
358 load data will be available.

359 *4.4 Run-out analysis*

360 All five landslides (S.N. 5, 7, 14, 15, 19 in Fig. 9) that are predicted to form potential landslide  
361 dams in case of slope failure were also used for the run-out analysis to evaluate expected  
362 runout distances in the event of reactivation and failure in the future. Results are as follows;

363 *4. 4.1 Akpa landslide (S.N. 5)*

364 Though it is difficult to ascertain the extent to which the predicted debris flow might contribute  
365 in the river blockage, it will certainly block the river in view of ~38 m high debris material  
366 with ~50 m wide run-out across the channel in this narrow part of river valley (Fig. 9a) even at  
367 maximum value of coefficient of friction (i.e.,  $\mu = 0.3$ ) (Fig. 13a). Notably, not only the run-out  
368 extent but flow height also decreases on increasing the friction value (Fig. 13a.1-13.a.3). The  
369 maximum friction takes into account the shear resistance by slope material and the bed-load on  
370 the river channel. However, apart from the frictional characteristics of run-out path, turbulence  
371 of a debris flow also controls its dimension and hence consequences like potential damming.  
372 Therefore, different values of turbulence coefficient ( $\xi$ ) were used (Table 4). The resultant flow  
373 height (representing 9 sets of modeled debris flows obtained using  $\mu = 0.05, 0.1$  and  $0.3$  and  $\xi =$   
374  $100, 200$  and  $300 \text{ m/s}^2$ ) attains its peak value i.e.,  $39.8 \pm 4.0 \text{ m}$  at the base of central part of  
375 landslide (Fig. 14a).

376 *4.4.2 Baren dogri landslide (S.N. 7)*

377 At the maximum friction value ( $\mu = 0.4$ ), the Baren dogri landslide would attain a peak value  
378 of flow height i.e., ~30 m at the base of central part of landslide (Fig. 13b). Similar to the valley  
379 configuration around the Akpa landslide (sec 4.4.1), the river valley attains a narrow/deep  
380 gorge setting here also (Fig. 9b). The maximum value of debris flow height obtained using the  
381 different  $\mu$  and  $\xi$  values is  $25.6 \pm 2.1 \text{ m}$  (Fig. 14b). Flow material is also noted to attain more  
382 run-out in upstream direction of river (~1100 m) than in the downstream direction (~800 m).  
383 This spatial variability in the run-out length might exist due to the river channel configuration  
384 as river channel in upstream direction is relatively narrower than the downstream direction.

385 *4.4.3 Pawari landslide (S.N. 14)*

386 The Pawari landslide attains maximum flow height of ~20 m at the maximum friction of run-  
387 out path ( $\mu = 0.4$ ) (Fig. 13c). The resultant debris flow that is achieved using the different values  
388 of  $\mu$  and  $\xi$  parameters attains a peak value of  $24.8 \pm 2.7 \text{ m}$  and decreases gradually with a run-  
389 out of ~1500 m in upstream and downstream direction (Fig. 14c). This landslide resulted in the  
390 relatively long run-out of ~1500 in the upstream and downstream direction. Apart from the

391 landslide volume that affects the run-out extent, valley morphology also controls it as evident  
392 from the previous landslides. The river channel in upstream and downstream direction from  
393 the landslide location is observed to be narrow (Fig. 9c).

394           4.4.4 *Telangi landslide (S.N. 15)*

395 The Telangi landslide would result in peak debris flow height of ~24 m at the maximum friction  
396 ( $\mu=0.4$ ) (Fig. 13d). On increasing the friction of run-out path, flow run-out decreased along the  
397 river channel but increased across the river channel resulting into possible damming. The debris  
398 flow after taking into account different values of  $\mu$  and  $\xi$  parameters attains a peak value of  
399  $25.0 \pm 4.0$  m (Fig. 14d). Similar to Baren dogri landslide (S.N. 7), material attained more run-  
400 out in the upstream direction of river (~1800 m) than in the downstream direction (~600 m) ;  
401 this difference can be attributed to a narrower river channel in upstream than the downstream  
402 direction. The downstream side attains wider river channel due to the Main Central Thrust  
403 (MCT) fault in the proximity (Fig. 1). Since the Pawari and Telangi landslides (S.N 14 & 15)  
404 are situated ~500 m from each other, their respective flow run-outs might mix in the river  
405 channel resulting into disastrous cumulative effect.

406           4.4.5 *Urni landslide (S.N. 19)*

407 The Urni landslide is predicted to attain a peak value of ~44 m of debris flow height at the  
408 maximum friction value ( $\mu=0.4$ ) (Fig. 13e). After considering different values of  $\mu$  and  $\xi$   
409 parameters, the debris flow would attain a height of  $26.3 \pm 1.8$  m (Fig. 14e). The relatively  
410 wider river channel in the downstream direction (Fig. 9e) results in longer run-out in  
411 downstream direction than in the upstream.

412           **5.0 DISCUSSION**

413 This study aimed to determine the potential landslide damming sites in the Satluj River valley,  
414 NW Himalaya. In order to achieve this objective, 44 active landslides were considered. First,  
415 slope stability evaluation of all the slopes, at these landslides sites was performed alongwith  
416 the parametric evaluation. Then the geomorphic indices, i.e., Morphological Obstruction Index  
417 (MOI) and Hydro-morphological Dam Stability Index (HDSI), were used to predict the  
418 formation of potential landslide dams and their subsequent stability. Rainfall and earthquake  
419 regime were also explored in the study area. Finally, run-out analysis was performed for those  
420 landslides that have been observed to form the potential landslide dam.

421 The MOI revealed that out of 44 active landslides in the Satluj valley, five (S.N. 5, 7, 14, 15,  
422 19) have the potential to form the landslide dam (Fig. 8, 9). On evaluating the stability of such  
423 potential dam sites using the HDSI, one landslide (S.N. 5) is predicted to attain an ‘uncertain’  
424 domain ( $5.74 < \text{HDSI} < 7.44$ ) in terms of dam stability. The uncertain term implies that the  
425 resultant dam may be stable or unstable depending upon the landslide/dam volume, upstream  
426 catchment area (or water discharge) and slope gradient (sec 3.3). Since this landslide (S.N.5)  
427 presents clear signs of having already formed a dam in the past, as indicated by the alternating  
428 fine-coarse layered sediment deposit (or lake deposit) in the upstream region (Fig. 10),  
429 recurrence is expected in the future. Further, run-out analysis of this landslide has predicted a  
430  $39.8 \pm 4.0$ m high debris flow in the event of failure that will block the river completely (Fig.  
431 13a, 14a). However, the durability of the blocking cannot be ascertained as it will depend on  
432 the volume of landslide that will be retained in the channel and river discharge.

433 The remaining four landslides (S.N. 7, 14, 15, 19), though showing instability i.e.,  $\text{HDSI} < 5.74$   
434 at present, may form dams in the near future as the region accommodating these landslides has  
435 been affected by such damming and subsequent flash floods in the past (Sharma et al. 2017).  
436 The last one of these i.e., S.N. 19 (Urni landslide) has already dammed the river partially and  
437 holds potential to completely block the river in near future (Kumar et al. 2019a). Run-out  
438 analysis of these landslides (S.N. 7, 14, 15, 19) has predicted  $25.6 \pm 2.1$ m,  $24.8 \pm 2.7$ m,  $25.0 \pm$   
439  $4.0$ m and  $26.3 \pm 1.8$ m flow height, respectively that will result in temporary blocking of the  
440 river (Fig. 13,14). These findings of run-out indicate the blocking of river in the event of slope  
441 failure, irrespective of durability, despite the conservative depth as input because only  $\frac{1}{4}$  of  
442 landslide thickness is used in the run-out analysis (sec. 3.5).

443 Stability evaluation of these five landslide slopes (S.N. 5, 7, 14, 15, 19) that have potential to  
444 form landslide dams revealed that one landslide (S.N.7) is meta-stable, while the other four  
445 belong to the unstable category (Fig. 4). Further, these four unstable landslide slopes are debris  
446 slides in nature. It is noteworthy to discuss the implications of  $\text{FS} < 1$ . The Factor of Safety (FS)  
447 in the Shear Strength Reduction (SSR) approach is a factor by which the existing shear strength  
448 of material is divided to determine the critical shear strength at which failure occurs  
449 (Zienkiewicz et al. 1975; Duncan 1996). Since the landslide represents a failed slope i.e.,  
450 critical shear strength  $>$  existing shear strength,  $\text{FS} < 1$  is justifiable. Further, the failure state of  
451 a slope in the FEM can be defined by different criteria; the FS of the same slope may vary a

452 little depending upon the usage of failure criteria and the convergence threshold (Abramson et  
453 al. 1996; Griffiths and Lane 1999).

454 The possible causes of instability ( $FS < 1$ ) may be steep slope gradient, rockmass having low  
455 strength, and joints. Three (S.N. 7, 14, 15) out of the five landslides that have potential to form  
456 dams belong to the tectonically active Higher Himalaya Crystalline (HHC). The notion of steep  
457 slope gradient cannot be generalized because the HHC accommodates voluminous ( $\sim 10^5$ - $10^7$   
458  $m^3$ ) landslides (Fig. 4). These deep seated landslides must require smooth slope gradient to  
459 accommodate the voluminous overburden. Further, the HHC comprises gneiss having high  
460 compressive strength and Geological Strength Index (Supplementary Table 2, Kumar et al.  
461 2021), therefore the notion of low strength rockmass also may not be appropriate. However,  
462 the jointed rock mass that owes its origin to numerous small-scale folds, shearing, and faults  
463 associated with the active orogeny process can be considered as the main factor for relatively  
464 more instability of debris slide type landslides. Since, the study area is subjected to the varied  
465 stress regime caused by the tectonic structures (Vannay et al. 2004), thermal variations (Singh  
466 et al. 2015), and anthropogenic cause (Lata et al. 2015), joints may continue to develop and  
467 destabilize the slopes. Apart from this inherent factor like joints, external factors like rainfall  
468 and exhumation rate may also contribute to instability of these landslides. This region receives  
469 relatively more annual rainfall owing to orographic barrier setting (Fig. 11) and is subjected to  
470 relatively high exhumation rate of 2.0-4.5 mm/yr (Thiede et al. 2009).

471 Two landslides (S.N. 5, 19) that are also capable of forming potential landslide dams (Fig. 8,  
472 9a; e) and are also unstable ( $FS < 1$ ) in nature (Fig. 4) do not belong to the HHC. The first  
473 landslide (S.N. 5) exists at the lithological contact of schist of the Tethyan Sequence and  
474 Kinnaur Kailash Granite rockmass. A regional normal fault, the Sangla Detachment (SD),  
475 passes through this contact. Some prior studies suggest that the SD is an outcome of  
476 reactivation of a former thrust fault that has resulted in intense rockmass shearing (Vannay et  
477 al. 2004; Kumar et al. 2019b). Owing to its location in the orographic interior region, hillslopes  
478 receive very low annual rainfall (Fig. 11) and thus have the least vegetation on the hillslopes  
479 in this region. The lack of vegetation on hillslopes has been observed to result in low shear  
480 strength of material and hence in the instability (Kokutse et al. 2016). Thus, lithological  
481 contrast, rockmass shearing, and lack of vegetation are thought to be the main reasons of  
482 instability of S.N. 5 landslide. The second landslide (S.N. 19) belongs to the inter-layered  
483 schist/gneiss rockmass of the Lesser Himalaya Crystalline (LHC) and is situated at the

484 orographic front where rainfall increases suddenly (Fig. 11). Further, this region is also  
485 subjected to the high exhumation rate of 2.0-4.5 mm/yr (Thiede et al. 2009). Therefore,  
486 lithological contrast, high rainfall and high exhumation rate are considered as the main reasons  
487 of instability of this landslide slope.

488 The landslides that are inferred not to result in the river damming are mostly in the LHC and  
489 Lesser Himalaya Sequence (LHS) region. These regions consist of a majority of the rock fall  
490 and rock avalanches that are generally of meta-stable category (Fig. 4). Despite the narrow  
491 valley setting, landslides in these regions may not form the potential landslide dam, at present,  
492 owing to the relatively small landslide volume. The possible causes of their meta-stability may  
493 be high compressive strength and geological strength index of gneiss (Kumar et al. 2021),  
494 dense vegetation on the hillslopes (Chawla et al. 2012), relatively less sheared rock mass in  
495 comparison to the HHC region, and relatively less decrease in land use/landcover (Lata et al.  
496 2015). Maximum Total Displacement (TD) is also associated with the rock fall and rock  
497 avalanche of this region (Fig. 4).

498 In the parametric study, soil friction and in-situ stress are noted to affect the FS most in case of  
499 the debris slide, whereas the FS of rock fall and rock avalanche are mainly controlled by the  
500 ‘ $m_i$ ’ and the in-situ stress. The ‘ $m_i$ ’ is a GHB criteria parameter that is equivalent to the friction  
501 in the M-C criteria. For the TD of the debris slides, field stress, elastic modulus and Poisson's  
502 ratio, whereas for rock falls and rock avalanches, the ‘ $m_i$ ’ parameter and in-situ stress played  
503 the dominant role (Fig. 6,7). Friction has been a controlling factor for the shear strength, and  
504 its decrease has been observed to result in the shear failure of slope material (Matsui and San  
505 1992). Since rainfall plays an important role in decreasing the friction of slope material by  
506 changing the pore water pressure regime (Rahardjo et al. 2005), frequent extreme rainfall  
507 events in the study area since the year 2013 (Kumar et al. 2019a) amplify the risk of hillslope  
508 instability. Furthermore, the in-situ field stress that has been compressional and/or extensional  
509 owing to the orogenic setting in the region may also enhance the hillslope instability (Eberhardt  
510 et al. 2004; Vannay et al. 2004). Deformation parameters, e.g. elastic modulus and Poisson's  
511 ratio, are also observed to affect the displacement in slope models of the debris slides. Similar  
512 studies in other regions have also noted the sensitivity of the elastic modulus and Poisson's  
513 ratio on the slope stability (Zhang and Chen 2006).

514 The study area has been subjected to extreme rainfalls since the year 2010 and received  
515 widespread slope failures and flash-floods (Fig. 11b). Three (S.N. 7,14,15 in Fig. 9) out of five

516 potential landslide dams belong to the Higher Himalaya Crystalline (HHC) that receives  
517 relatively more rainfall (Fig. 11). Contrary to the along ‘Himalayan’ arc distribution of  
518 earthquakes, the study area has received most of the earthquakes around the N-S oriented  
519 Kaurik-Chango Fault (Fig. 12a). However, the only major earthquake event has been  $M_w$  6.8  
520 earthquake on 19<sup>th</sup> Jan. 1975 that resulted in the widespread landslides (Khattri et al. 1978).  
521 The low-magnitude recent seismicity in the region has been attributed to the northward  
522 extension of the Delhi-Haridwar ridge (Hazarika et al. 2019), whereas the shallow nature is  
523 attributed to the MHT ramp structure in the region that allows strain accumulation at shallow  
524 depth (Bilham 2019). Thus, earthquakes have not been a major landslide triggering process in  
525 the region in recent times. Finally, the word “active landslide” refers to the hillslope that is still  
526 subjected to the slope failures caused by the various factors. The word “landslide” can be  
527 perceived in the following three ways; pre-failure deformations, failure itself, and post-failure  
528 displacement (Terzaghi 1950; Cruden & Varnes, 1996; Hungr et al., 2014). Landslide slopes  
529 in this study pertain to the post-failure state that are categorized into “unstable” and “meta-  
530 stable” stages based on their existing FS. If an active landslide is not categorized as “unstable”,  
531 it means that the existing slope geometry provides it a “meta-stable” stage that might transform  
532 into an unstable stage with time due to the stability controlling parameters (Sec. 4.1). Though  
533 the field visits were performed in different seasons to cover all the landslides along the study  
534 area, there might be a possibility of vegetation growth on the failed slopes, particularly in the  
535 LHC and LHS. However, the landslides in the LHC and LHS are mostly rockfall/rock  
536 avalanche type because of the deep gorge setting, whereas the vegetation growth generally  
537 requires the debris laden hillslopes. Nonetheless, such aspect will be explored in the future  
538 prospects. The HHC and the TS region are subjected to the semi-humid to semi-arid climate  
539 and hence the vegetation type is mostly scattered trees/shrubs. Therefore, such possibility  
540 might not exist in these regions.

541 A supplementary table involving all the details like landslides dimensions, factor of safety, and  
542 geomorphic indices output of each landslide is provided in the data repository (Kumar et al.  
543 2021).

544 In view of the possible uncertainties in the predictive nature of the study, the following  
545 assumptions and simplifications were made;

546 • To account the effect the spatial variability in the slope geometry, 3D models have been  
547 in use for the last decade (Griffiths and Marquez 2007). However, the pre-requisite for

548 the 3D models involves the detailed understanding of slope geometry and material  
549 variability in the subsurface that was not possible in the study area considering steep  
550 and inaccessible slopes. Therefore, multiple 2D sections were chosen, wherever  
551 possible. To account the effect of sampling bias and material variability, a range of  
552 values of input parameters was used (sec. 4.1).

553 • Determination of the debris thickness has been a major problem in the landslide volume  
554 measurement particularly in the steep, narrow river valleys of the NW Himalaya.  
555 Therefore, the thickness was approximated by considering the relative altitude of the  
556 ground on either side of the deposit, as also performed by Innes (1983). It was assumed  
557 that the ground beneath the deposit is regular.

558 • The resultant dam volume could be different from the landslide volume due to the  
559 entrainment, rockmass fragmentation, pore water pressure, size of debris particles, and  
560 washout of landslide material by the river (Hung and Evans 2004; Dong et al. 2011;  
561 Yu et al. 2014). Therefore, dam volume is presumed to be equal to landslide volume  
562 for the worst-case scenario (sec. 3.3). Stream power is manifested by the upstream  
563 catchment area and local slope gradient in the geomorphic indices. It may also vary at  
564 temporal scale owing to the temporally varying water influx from glaciers and  
565 precipitation systems i.e., ISM and WD (Gadgil et al. 2007; Hunt et al. 2018). Though  
566 our study is confined to the spatial scale at present, the findings remain subjected to the  
567 change at temporal scale.

568 • The RAMMS model (Voellmy 1955; Salm 1993; Christen et al. 2010) requires the  
569 calibrated friction and turbulence values for the run-out analysis. Though the previous  
570 debris flow events have not left evidence in the study area owing to the convergence of  
571 the landslide toes with the river channel, a range of  $\mu$  and  $\xi$  values were used in the  
572 study in view of the material type and run-out path characteristics.

573 Despite these uncertainties, studies such as this one are required to minimize the risk and  
574 avert the possible disasters in the terrain where human population lives in the proximity of  
575 unstable landslides.

## 576 CONCLUSION

577 Out of forty-four landslides that are studied in the Satluj valley in the NW Himalaya, five  
578 landslides are noted to form the potential landslide dam, if failure occurs. Though the blocking  
579 duration is difficult to predict, upstream and downstream consequences of these damming

580 events should be carefully considered as the region has witnessed many damming and flash  
581 floods in the past. These five landslides comprise a total landslide volume of  $26.3 \pm 6.7 \text{ M m}^3$ .  
582 The slopes of four landslides (debris slides) out of these five are unstable, whereas the  
583 remaining one (rock avalanche) is meta-stable. Field observations and previous studies have  
584 noted the damming events by these landslides (or the region consisting these landslides) in the  
585 past also. Since the area is witnessing enhanced rainfall and flash floods since year 2010,  
586 findings of the run-out analysis that revealed  $24.8 \pm 2.7\text{m}$  to  $39.8 \pm 4.0\text{m}$  high material flow  
587 from these landslides become more crucial. The parametric analysis for the slope stability  
588 evaluation revealed that the angle of internal friction of soil or ' $m_i$ ' (equivalent to the angle of  
589 internal friction) of the rockmass, and *in-situ* field stress are the most controlling parameters  
590 for the stability of slopes.

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## 598 **Conflict of Interest**

599 The authors declare that they have no conflict of interest.

## 600 **Dataset Availability**

601 The dataset is uploaded in the open access repository (*Mendeley data*) as Kumar et al. (2020).

## 602 **Author contribution**

603 VK conceived the idea and collected the field data. VK and IJ performed the laboratory  
604 analysis. All authors contributed to the dataset compilation, numerical simulation, and  
605 geomorphic interpretations. All authors contributed to the writing of the final draft.

## 606 **REFERENCES**

607 Abramson, L. W., Lee, T. S., Sharma, S. 1996. "Slope stability and stabilization methods".  
608 New York. Wiley.

609 Aydin, A. 2008.“ISRM suggested method for determination of the Schmidt hammer rebound  
610 hardness: revised version”. *In: R. Ulusay (ed) The ISRM Suggested Methods for Rock*  
611 *Characterization, Testing and Monitoring: 2007–2014*. Springer, Cham, pp. 25-33.

612 Barton, N. and Bandis, S. 1982.“Effects of block size on the shear behavior of jointed rock”.  
613 *In: Proc. 23rd US symposium on rock mechanics*. American Rock Mechanics  
614 Association, 739-760.

615 Barton, N. and Choubey, V. 1977.“The shear strength of rock joints in theory and practice”.  
616 *Rock mechanics* 10(1-2):1-54.

617 Barton, N. R. 1972.“A model study of rock-joint deformation. International journal of rock  
618 mechanics and mining sciences”. *Geomech. Abstr.* 9 (5): 579-602.

619 Barton, N. R. and Bandis, S. C. 1990.“Review of predictive capabilities pf JRC-JCS model in  
620 engineering practice”. *In: Barton N and Stephansson O (ed) Rock Joints*, Rotterdam, pp.  
621 603-610.

622 Bhasin, R. and Kaynia, A. M. 2004.“Static and dynamic simulation of a 700-m high rock slope  
623 in western Norway”. *Engineering Geology* 71(3-4): 213-226.

624 Bilham, R. 2019.“Himalayan earthquakes: a review of historical seismicity and early 21st  
625 century slip potential”. *Geological Society, London, Special Publications*. 483(1):423-  
626 82.

627 Bowles, J. E. 1996.*Foundation Analysis and Design*,5<sup>th</sup>ed., McGraw-Hill, New York, pp. 750.

628 Braun, A., Cuomo, S., Petrosino, S., Wang, X. and Zhang, L., 2018. “Numerical SPH analysis  
629 of debris flow run-out and related river damming scenarios for a local case study in SW  
630 China”. *Landslides* 15 (3), 535-550.

631 Cai, M., Kaiser, P. K., Tasaka, Y. and Minami, M. 2007.“Determination of residual strength  
632 parameters of jointed rock masses using the GSI system”. *International Journal of Rock  
633 Mechanics and Mining Sciences* 44(2): 247-265.

634 Chawla, A., Kumar, A., Lal, B., Singh, R.D. and Thukral, A.K. 2012.“Ecological  
635 Characterization of High Altitude Himalayan Landscapes in the Upper Satluj River  
636 Watershed, Kinnaur, Himachal Pradesh, India”. *Journal of the Indian Society of Remote  
637 Sensing* 40(3): 519-539.

638 Cho, S. E. 2007.“Effects of spatial variability of soil properties on slope stability”. *Engineering  
639 Geology* 92(3-4): 97-109.

640 Christen, M., Kowalski, J., Bartelt, P. 2010. “RAMMS: Numerical simulation of dense snow  
641 avalanches in three-dimensional terrain”. *Cold Regions Science and Technology*.63: 1-  
642 14.

643 Corominas, J. and Mavrouli, J., 2011. “Living with landslide risk in Europe: Assessment,  
644 effects of global change, and risk management strategies”. *Documento tecnico, safeland.*  
645 *7<sup>th</sup> Framework Programme Cooperation Theme*, 6.

646 Costa, J. E. and Schuster, R. L. 1988.“The formation and failure of natural dams”. *Geological  
647 society of America bulletin* 100(7): 1054-1068.

648 Coulomb, C. A. 1776.“An attempt to apply the rules of maxima and minima to several  
649 problems of stability related to architecture”. *Mémoires de l'Académie Royale des  
650 Sciences* 7: 343-382.

651 Cruden, D. M. and Varnes, D. J. 1996. Landslides: investigation and mitigation. In: Landslide  
652 types and processes. Transportation Research Board, US National Academy of Sciences,  
653 Special Report, (247).

654 Dai, F.C., Lee, C.F., Deng, J.H. and Tham, L.G. 2005.“The 1786 earthquake-triggered  
655 landslide dam and subsequent dam-break flood on the Dadu River, southwestern China”.  
656 *Geomorphology* 65(3): 205-221.

657 Deere, D. U. and Miller, R. P. 1966.“Engineering classification and index properties for intact  
658 rock”. Illinois University at Urbana, USA.

659 Delaney, K. B. and Evans, S. G. 2015.“The 2000 Yigong landslide (Tibetan Plateau), rockslide-  
660 dammed lake and outburst flood: review, remote sensing analysis, and process  
661 modelling”. *Geomorphology* 246, 377-393.

662 Dimri, A.P., Niyogi, D., Barros, A.P., Ridley, J., Mohanty, U.C., Yasunari, T. and Sikka, D.R.  
663 2015.“Western disturbances: A review”. *Reviews of Geophysics* 53(2): 225-246

664 Dong, J.J., Tung, Y.H., Chen, C.C., Liao, J.J. and Pan, Y.W. 2011.“Logistic regression model  
665 for predicting the failure probability of a landslide dam”. *Engineering Geology* 117(1):  
666 52-61.

667 Duncan, J. M. 1996.“State of art: limit equilibrium and finite element analysis of slopes”.  
668 *Journal of Geotechnical Engineering Division, ASCI* 122 (7): 577-596.

669 Eberhardt, E., Stead, D. and Coggan, J. S. 2004.“Numerical analysis of initiation and  
670 progressive failure in natural rock slopes—the 1991 Randa rockslide”. *International  
671 Journal of Rock Mechanics and Mining Sciences* 41(1): 69-87.

672 Ermini, L. and Casagli, N. 2003.“Prediction of the behaviour of landslide dams using a  
673 geomorphological dimensionless index”. *Earth Surface Processes and Landforms* 28(1):  
674 31-47.

675 Fan, X, Dufresne, A, Subramanian, SS, Strom, A, Hermanns, R, Stefanelli, CT, Hewitt, K,  
676 Yunus, AP, Dunning, S, Capra, L and Geertsema, M. 2020. “The formation and impact  
677 of landslide dams—State of the art”. *Earth-Science Reviews*, 203:103116.

678 Fujisawa, K., Kobayashi, A. and Aoyama, S. 2009.“Theoretical description of embankment  
679 erosion owing to overflow”. *Geotechnique* 59(8): 661-671.

680 Gadgil, S., Rajeevan, M. and Francis, P.A. 2007.“Monsoon variability: Links to major  
681 oscillations over the equatorial Pacific and Indian oceans”. *Current Science*. 93(2):182-  
682 94.

683 Griffiths, D. V. and Lane, P. A. 1999.“Slope stability analysis by finite elements”.  
684 *Geotechnique* 49(3): 387-403.

685 Griffiths, D. V. and Marquez, R. M. 2007.“Three-dimensional slope stability analysis by  
686 elasto-plastic finite elements”. *Geotechnique* 57(6): 537-546.

687 Gupta, V. and Sah, M. P. 2008.“Impact of the trans-Himalayan landslide lake outburst flood  
688 (LLOF) in the Satluj catchment, Himachal Pradesh, India”. *Natural Hazards* 45(3): 379-  
689 390.

690 Gupta, V. Nautiyal, H., Kumar, V., Jamir, I., and Tandon, R. S. 2016.“Landslides hazards  
691 around Uttarkashi Township, Garhwal Himalaya, after the tragic flash flood in June  
692 2013”. *Natural Hazards* 80: 1689-1707

693 Huffman, G.J., Bolvin, D.T., Nelkin, E.J. and Adler, R.F. 2016. “TRMM (TMPA) Precipitation  
694 L3 1 day 0.25 degree x 0.25 degree V7, Edited by Andrey Savtchenko, Goddard Earth

695 Sciences Data and Information Services Center (GES DISC)", Accessed: 05 April 2020,  
696 10.5067/TRMM/TMPA/DAY/7.

697 Hungr, O., Leroueil, S., and Picarelli, L. 2014. "The Varnes classification of landslide types,  
698 an update". *Landslides* 11 (2): 167-194.

699 Hürlimann, M., Rickenmann, D., Medina, V. and Bateman, A. 2008. "Evaluation of approaches  
700 to calculate debris-flow parameters for hazard assessment". *Eng Geol* 102:152–163

701 Havenith, H.B., Torgoev, I. and Ischuk, A. 2018. "Integrated Geophysical-Geological 3D  
702 Model of the Right-Bank Slope Downstream from the Rogun Dam Construction Site,  
703 Tajikistan". *International Journal of Geophysics* 1-17.

704 Hazarika, D., Wadhawan, M., Paul, A., Kumar, N. and Borah, K. 2017. "Geometry of the Main  
705 Himalayan Thrust and Moho beneath Satluj valley, northwest Himalaya: Constraints  
706 from receiver function analysis". *Journal of Geophysical Research: Solid Earth*.  
707 122(4):2929-45.

708 Hoek, E. and Brown, E.T. 1997. "Practical estimates of rock mass strength". *International  
709 Journal of Rock Mechanics and Mining Science* 34(8):1165-1186

710 Hoek, E. and Diederichs, M.S. 2006. "Empirical estimation of rock mass modulus".  
711 *International Journal of Rock Mechanics and Mining Sciences* 43 (2): 203-15.

712 Hoek, E., Kaiser, P.K. and Bawden, W.F. 1995. *Support of Underground Excavations in Hard  
713 Rock*. Rotterdam: A. A. Alkema.

714 Hungr, O. and Evans, S.G. 2004. "Entrainment of debris in rock avalanches: an analysis of a  
715 long run-out mechanism". *Geological Society of America Bulletin* 116(9-10): 1240-1252.

716 Hungr, O., Morgan, G.C. and Kellerhals, R. 1984. "Quantitative analysis of debris torrent  
717 hazards for design of remedial measures". *Can Geotech J* 21:663–677

718 Hunt, K.M., Turner, A.G. and Shaffrey, L.C. 2018. "The evolution, seasonality and impacts of  
719 western disturbances". *Quarterly Journal of the Royal Meteorological Society*.  
720 144(710):278-90.

721 Hutter, K., Svendsen, B. and Rickenmann, D. 1994. "Debris flow modelling: a review".  
722 *Continuum mechanics and thermodynamics* 8: 1-35.

723 Innes, J. L. 1983. "Debris flows". *Progress in Physical Geography* 7(4): 469–501.

724 IS: 2720 (Part 10)–1991. "Method of test for soils: Determination of unconfined compressive  
725 strength". In: *Bureau of Indian Standards, Delhi, India*.

726 IS: 2720 (Part 13)–1986. "Method of test for soils: Direct shear test". In: *Bureau of Indian  
727 Standards, New Delhi, India*.

728 IS: 2720 (Part 4)–1985. "Methods of test for soils: Grain size analysis". In: *Bureau of Indian  
729 Standards, New Delhi, India*.

730 IS: 9143-1979. "Method for the determination of unconfined compressive strength of rock  
731 materials". In: *Bureau of Indian Standards, New Delhi, India*.

732 Jamir, I., Gupta, V., Kumar, V. and Thong, G.T. 2017. "Evaluation of potential surface  
733 instability using finite element method in Kharsali Village, Yamuna Valley, Northwest  
734 Himalaya". *Journal of Mountain Science* 14(8): 1666-1676.

735 Jamir, I., Gupta, V., Thong, G.T. and Kumar, V. 2019. "Litho-tectonic and precipitation  
736 implications on landslides, Yamuna valley, NW Himalaya". *Physical Geography* 41(4)  
737 365-388.

738 Jang, H.S., Kang, S.S. and Jang, B.A. 2014.“Determination of joint roughness coefficients  
739 using roughness parameters”. *Rock mechanics and rock engineering* 47(6): 2061-73.

740 Jing, L. 2003.“A review of techniques, advance and outstanding issues in numerical modelling  
741 for rock mechanics and rock engineering”. *International Journal of Rock Mechanics and*  
742 *Mineral Science* 40:283-353.

743 Khattri, K., Rai, K., Jain, A.K., Sinvhal, H., Gaur, V.K. and Mithal, R.S. 1978.“The Kinnaur  
744 earthquake, Himachal Pradesh, India, of 19 January, 1975”. *Tectonophysics*. 49(1-2):1-  
745 21.

746 Kokutse, N.K., Temgoua, A.G.T. and Kavazović, Z. 2016.“Slope stability and vegetation:  
747 Conceptual and numerical investigation of mechanical effects”. *Ecological Engineering*  
748 86: 146–153.

749 Kumar, V., Gupta, V. and Jamir, I. 2018.“Hazard Evaluation of Progressive Pawari Landslide  
750 Zone, Kinnaur, Satluj Valley, Higher Himalaya, India”. *Natural Hazards* 93: 1029-1047.

751 Kumar, V., Gupta, V., Jamir, I. and Chattoraj, S.L. 2019a.“Evaluation of potential landslide  
752 damming: Case study of Urni landslide, Kinnaur, Satluj valley, India”. *Geosci. Front.*  
753 10(2): 753-767.

754 Kumar, V., Gupta, V. and Sundriyal, Y.P. 2019b.“Spatial interrelationship of landslides, litho-  
755 tectonics, and climate regime, Satluj valley, Northwest Himalaya”. *Geol. J.* 54: 537–551.

756 Kumar, V., Jamir, I., Gupta, V. and Bhasin, R.K. 2021. “Dataset used to infer regional slope  
757 stability, NW Himalaya”. *Mendeley Data*. DOI: 10.17632/jh8b2rh8nz.2

758 Kundu, B., Yadav, R.K., Bali, B.S., Chowdhury, S. and Gahalaut, V.K. 2014.“Oblique  
759 convergence and slip partitioning in the NW Himalaya: Implications from GPS  
760 measurements.” *Tectonics* 33: 2013-2024.

761 Lata, R., Rishi, S., Talwar, D. and Dolma, K. 2015.“Comparative Study of Landuse Pattern in  
762 the Hilly Area of Kinnaur District, Himachal Pradesh, India”. *International Journal of*  
763 *Innovative Science, Engineering & Technology* 2(4): 559-565.

764 Li, D.Q., Qi, X.H., Cao, Z.J., Tang, X.S., Phoon, K.K. and Zhou, C.B. 2016.“Evaluating slope  
765 stability uncertainty using coupled Markov chain”. *Computers and Geotechnics* 73:72-82.

766 Li, T., Schuster, R.L. and Wu, J. 1986.“Landslide dams in south-central China”. In: *Proc.*  
767 *Landslide dams: processes, risk, and mitigation*, ASCE Convention, Washington, 146-  
768 162.

769 Li, M.H., Sung, R.T., Dong, J.J., Lee, C.T., and Chen, C.C. 2011. “The formation and breaching  
770 of a short-lived landslide dam at Hsiaolin village, Taiwan-Part II: Simulation of debris  
771 flow with landslide dam breach”. *Engineering Geology* 123: 60-71.

772 Martha, T.R., Kerle, N., Jetten, V., van Westen, C.J. and Kumar, K.V. 2010.“Landslide  
773 volumetric analysis using Cartosat-1-derived DEMs”. *IEEE Geoscience and remote*  
774 *sensing letters* 7(3): 582-586.

775 Matsui, T. and San, K.C. 1992.“Finite element slope stability analysis by shear strength  
776 reduction technique”. *Soils and foundations* 32(1): 59-70.

777 Mohr, O. 1914.“Abhandlungen aus dem Gebiete der Technischen Mechanik” (2nd ed). Ernst,  
778 Berlin.

779 NDMA: National Disaster Management Authority.2013. Annual report, Government of India,  
780 Delhi.

781 Nian, T.K., Chen, G.Q., Wan, S.S. and Luan, M.T. 2011.“Non-convergence Criterion on Slope  
782 Stability FE Analysis by Strength Reduction Method”. *Journal of Convergence  
783 Information Technology* 6:78-88.

784 O'Brien, J.S., Julien, P.Y. and Fullerton, W.T. 1993.“Two-dimensional water flood and mud  
785 flood simulation”. *J Hydraul Eng* 119:244–260.

786 Parkash, S. 2011.“Historical records of socio-economically significant landslides in India”. *J  
787 South Asia Disaster Studies* 4(2): 177-204.

788 Rahardjo, H., Lee, T.T., Leong, E.C. and Rezaur, R.B. 2005.“Response of a residual soil slope  
789 to rainfall”. *Canadian Geotechnical Journal* 42(2): 340-351.

790 Rickenmann, D. 2005.“Runout prediction methods”. In: *Jakob M, Hungr O (eds) Debris-flow  
791 hazards and related phenomena*. Praxis/Springer, Berlin/Heidelberg, pp. 305–324.

792 Rickenmann, D., and Scheidl, C. 2013.“Debris-Flow Runout and Deposition on the Fan”. In: *M.  
793 Schneuwly-Bollschweiler et al. (eds.), Dating Torrential Processes on Fans and Cones,  
794 Advances in Global Change Research* 47, DOI 10.1007/978-94-007-4336-6 5

795 Ruiz-Villanueva, V., Allen, S., Arora, M., Goel, N. K. and Stoffel, M. 2016.“Recent  
796 catastrophic landslide lake outburst floods in the Himalayan mountain range”. *Progress  
797 in Physical Geography* 41 (1): 3-28.

798 Salm, B., 1993. “Flow, flow transition and runout distances of flowing avalanches”. *Ann.  
799 Glaciol.* 18: 221–226.

800 Sharma, S., Shukla, A.D., Bartarya, S.K., Marh, B.S. and Juyal, N. 2017.“The Holocene floods  
801 and their affinity to climatic variability in the western Himalaya, India”. *Geomorphology*  
802 290: 317–334.

803 Sharma, K. K. 1977.“A contribution to the geology of Satluj Valley, Kinnaur, Himachal  
804 Pradesh, India”. *CollquesInternationaux du CNRS* 268: 369-378.

805 Singh, D., Gupta, R.D. and Jain, S.K. 2015.“Statistical analysis of long term spatial and  
806 temporal trends of temperature parameters over Sutlej river basin, India”. *J. Earth Syst.  
807 Sci.* 124 (1): 17-35.

808 Stefanelli, C.T., Segoni, S., Casagli, N. and Catani, F. 2016.“Geomorphic indexing of landslide  
809 dams evolution”. *Engineering Geology* 208: 1-10.

810 Takahashi, T. And Nakagawa, H. 1993.“Flood and debris flow hydrograph due to collapse of  
811 a natural dam by overtopping”. In: *Proc. Hydraulic Engineering, Japan*, 37: 699-704.

812 Terzaghi, K. 1950. “Mechanism of Landslides”. In: *Application of Geology to Engineering  
813 Practice. Geol. Soc. Am* 83-123.

814 Thiede, R.C., Ehlers, T.A., Bookhagen, B. and Strecker, M.R. 2009.“Erosional variability  
815 along the northwest Himalaya”. *Journal of Geophysical Research: Earth Surface* 114  
816 F101015.

817 Vannay, J.C., Grasemann, B., Rahn, M., Frank, W., Carter, A., Baudraz, V. and Cosca, M.  
818 2004.“Miocene to Holocene exhumation of metamorphic crustal wedges in the NW  
819 Himalaya: Evidence for tectonic extrusion coupled to fluvial erosion”. *Tectonics* 23  
820 TC1014.

821 Voellmy, A. 1955. “On the destructive force of avalanches”, *Translation No. 2. Avalanche  
822 Study Center, United States Department of Agriculture, USA.*

823 Wong, F.S. 1984.“Uncertainties in FE modelling of slope stability”. *Computers & structures*  
824 19(5-6): 777-791.

825 Wulf, H., Bookhagen, B. and Scherler, D. 2012.“Climatic and geologic controls on suspended  
 826 sediment flux in the Sutlej River Valley, western Himalaya”. *Hydrology and Earth*  
 827 *System Sciences* 16(7): 2193-2217.

828 Yu, G., Zhang, M., Chen, H., 2014. “The dynamic process and sensitivity analysis for debris  
 829 flow”. In: Sassa, K., Canuti, P., Yin, Y. (Eds.), *Landslide Science for a Safer*  
 830 *Geoenvironment*. Springer.

831 Zhang, P. W. and Chen, Z. Y. 2006.“Influences of soil elastic modulus and Poisson's ratio on  
 832 slope stability”. *YantuLixue (Rock and Soil Mechanics)* 27(2): 299-303.

833 Zienkiewicz, O. C., Humphezon, C. and Lewis, R. W. 1975.“Associated and non-associated  
 834 viscoplasticity and plasticity in soil mechanics”. *Geotechnique* 25(4): 671-689.

835 **LIST OF FIGURES AND TABLES**

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841 **Fig. 2** Field photographs of some of the landslides (a) Khokpa landslide (**S.N.1**); (b) Akpa\_III  
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872 relatively more rainfall resulting into the flash floods, landslides and socio-economic loss  
873 in the region. (i):hpenvis.nic.in, retrieved on March 1, 2020; Department of Revenue,  
874 Govt. of H.P. (ii): hpenvis.nic.in, retrieved on March 1, 2020.(iii): Kumar et al.,  
875 2019a;ndma.gov.in, retrieved on march 1, 2020 (iv):sandrp.in, retrieved on march 1,  
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877 **Fig. 12** Earthquake distribution. (a) Spatial variation of earthquakes. The transparent circle  
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881 of earthquakes within 40 km depth; (c) Cross section view (Based on Hazarika et al. 2017;  
882 Bilham 2019). Red dashed circle represents the zone of strain accumulation caused by the  
883 Indian and Eurasian plate collision (Bilham 2019). ISC: International Seismological  
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885 **Fig. 13** Results of the run-out analysis.  $\mu$  refers to coefficient of friction.

886 **Fig. 14** Results of run-out analysis at different values of  $\mu$  and  $\xi$ .  $\mu$  and  $\xi$  refer to coefficient of  
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888 **Table 1** Details of the landslides used in the study.

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**Table 1** Details of landslides used in the study.

S.N .	Landslide location	Latitude/ Longitude	Type	Area <sup>1</sup> , m <sup>2</sup>	Volume <sup>2</sup> , m <sup>3</sup>	Human population <sup>3</sup>	Litho-tectonic division
1	Khokpa	31°35'18.9"N 78°26'28.6"E	Debris slide	21897± 241	43794± 18361	373	Tethyan Sequence (TS)
2	Tirung Khad	31°34'50.4"N 78°26'20.5"E	Rockfall	28537± 314	14269± 9055	0	
3	Akpa_I	31°34'57.1"N 78°24'30.6"E	Rock avalanche	963051± 10594	1926102± 807515	0	
4	Akpa_II	31°35'2.2"N 78°23'25.4"E	Rock avalanche	95902± 1055	143853± 40734	470	Kinnaur Kailash Granite (KKG)
5	Akpa_III	31°34'54.5"N 78°23'2.4"E	Debris slide	379570± 4175	7591400± 3182681	1617	
6	Rarang	31°35'58.7"N 78°20'39.1"E	Rockfall	4586± 50	4586± 1923	848	
7	Baren Dogri	31°36'23.6"N 78°20'23.1"E	Rock avalanche	483721± 5321	2418605±421561	142	Higher Himalaya Crystalline (HHC)
8	Thopan Dogri	31°36'12.3"N 78°19'50.4"E	Rockfall	55296± 608	165888± 46974	103	
9	Kashang Khad_I	31°36'5.0"N 78°18'44.4"E	Debris slide	113054± 1244	169581± 48019	103	
10	Kashang Khad_II	31°35'58.3"N 78°18'34.0"E	Rockfall	27171± 299	40757± 11541	103	
11	Pangi_I	31°35'36.4"N 78°17'36.4"E	Debris slide	30112± 331	45168± 12790	1389	
12	Pangi_II	31°35'38.9"N 78°17'12.2"E	Debris slide	59436± 654	118872± 49837	1389	
13	Pangi_III	31°34'38.9"N 78°16'55.6"E	Debris slide	75396± 829	188490± 32854	7	
14	Pawari	31°33'49.8"N 78°16'28.6"E	Debris slide	320564± 3526	1602820± 279370	4427	
15	Telangi	31°33'7.0"N 78°16'37.2"E	Debris slide	543343± 5977	13583575± 2367608	6817	

	16	Shongthong	31°31'13.0"N 78°16'17.0"E	Debris slide	5727± 63	11454± 2464	388	
	17	Karchham	31°30'12.4"N 78°11'30.8"E	Rock avalanche	28046± 309	56092± 23516	0	
	18	Choling	31°31'17.0"N 78° 8'4.9"E	Debris slide	20977± 231	20977± 8795	0	
	19	Urni	31°31'8.0"N 78° 7'42.2"E	Debris slide	112097± 1233	1120970± 469965	500	
	20	Chagaon_I	31°30'55.9"N 78° 6'52.0"E	Rockfall	3220± 35	3220± 1350	0	
	21	Chagaon_II	31°30'57.9"N 78° 6'47.7"E	Rockfall	11652± 128	11652± 4885	0	
	22	Chagaon_III	31°31'3.0"N 78° 6'21.4"E	Debris slide	42141± 464	168564± 70670	1085	Lesser Himalaya Crystalline (LHC)
	23	Wangtu_U/s	31°32'4.8"N 78° 3'5.0"E	Rock avalanche	211599± 2328	317399± 89876	17	
	24	Wangtu D/s_1	31°33'27.7"N 77°59'43.7"E	Debris slide	4655± 51	9310± 3903	71	
	25	Kandar	31°33'43.7"N 77°59'54.9"E	Rock avalanche	151128± 1662	302256± 126720	186	
	26	Wangtu D/s_2	31°33'38.9"N 77°59'29.9"E	Debris slide	8004± 88	16008± 6711	71	
	27	Agade	31°33'52.3"N 77°58'3.5"E	Debris slide	9767± 107	14651± 4149	356	
	28	Punaspas	31°33'37.6"N 77°57'31.5"E	Debris slide	3211± 35	3211± 1346	343	
	29	Sungra	31°33'58.8"N 77°56'49.6"E	Debris slide	5560± 61	11120± 4662	2669	
	30	Chota Kamba	31°33'39.2"N 77°54'39.0"E	Rock avalanche	197290± 2170	591870± 167597	401	
	31	Bara Kamba	31°34'10.4"N 77°52'56.7"E	Rockfall	36347± 400	18174± 7619	564	
	32	Karape	31°33'44.9"N 77°53'13.9"E	Debris slide	50979± 561	50979± 21373	1118	
	33	Pashpa	31°34'40.2"N 77°50'53.0"E	Rockfall	16079± 171	8040± 3371	29	
	34	Khani Dhar_I	31°33'43.4"N 77°48'52.5"E	Rock avalanche	218688± 2406	874752± 366738	0	
	35	Khani Dhar_II	31°33'26.3"N 77°48'35.8"E	Rock avalanche	146994± 1617	734970± 248125	0	
	36	Khani Dhar_III	31°33'20.1"N 77°48'27.8"E	Rock avalanche	20902± 230	62706± 17756	0	
	37	Jeori	31°31'58.8"N 77°46'18.2"E	Rock avalanche	93705± 1031	93705± 39286	0	
	38	Barauni Gad_I_S	31°28'56.6"N 77°41'40.4"E	Debris slide	63241± 696	758892± 111620	236	LHC-LHS
	39	Barauni Gad_I_Q	31°29'00.0"N 77°41'38.0"E	Debris slide	59273± 652	711276± 104616	0	Lesser Himalaya

40	Barauni Gad_II	31°28'43.9"N 77°41'24.6"E	Rockfall	6977± 77	3489± 1463	0	Sequence (LHS)
41	Barauni Gad_III	31°29'5.6"N 77°41'23.7"E	Rockfall	33115± 364	33115± 13883	0	
42	D/s Barauni Gad_I	31°28'24.9"N 77°41'8.4"E	Rockfall	19101± 210	19101± 8008	0	
43	D/s Barauni Gad_II	31°28'25.5"N 77°40'56.7"E	Rockfall	21236± 234	21236± 8903	0	
44	D/s Barauni Gad_III	31°28'7.4"N 77°40'42.4"E	Rockfall	15632± 172	15632± 6554	0	

<sup>1</sup>Error (±) caused by GE measurement (1.06 %).

<sup>2</sup>Error (±) is an outcome of multiplication of area ± error and thickness ± error. Thickness error (Std. dev.) corresponds to averaging of field based approximated thickness.

<sup>3</sup>The human population is based on census 2011, Govt. of India. The villages/town in the radius of 500 m from the landslide are considered to count the human population.

**Table 2** Details of the satellite imagery.

Satellite data		Source	Date of data	Spatial resolution
CARTOSAT-1 stereo imagery	524/253	National Remote Sensing Center (NRSC), Hyderabad, India	5 <sup>th</sup> Dec. 2010	~2.5 m
	525/253		16 <sup>th</sup> Dec. 2010	~2.5 m
	526/252		18 <sup>th</sup> Oct. 2011	~2.5 m
	526/253		18 <sup>th</sup> Oct. 2011	~2.5 m
	527/252		24 <sup>th</sup> Nov. 2010	~2.5 m
	527/253		27 <sup>th</sup> Dec. 2010	~2.5 m
	528/252		26 <sup>th</sup> Nov. 2011	~2.5 m

**Table 3** Criteria used in the Finite Element Method (FEM) analysis.

Material Criteria		Parameters	Source
Rockmass	<b>Generalized Hoek &amp; Brown (GHB) Criteria</b> (Hoek et al. 1995)  $\sigma_1 = \sigma_3 + \sigma_{ci} [m_b (\sigma_3 / \sigma_{ci}) + s]^a$	Unit Weight, $\gamma$ (MN/m <sup>3</sup> )	Laboratory analysis (UCS) (IS: 9143-1979)
		Uniaxial Compressive Strength, $\sigma_{ci}$ (MPa)	
	Here, $\sigma_1$ and $\sigma_3$ are major and minor effective principal stresses at failure; $\sigma_{ci}$ , compressive strength of intact rock; $m_b$ , a reduced value of the material constant ( $m_i$ ) and is given by;	Rockmass modulus (MPa)	Laboratory analysis (Ultrasonic velocity test); Hoek and Diederichs (2006).
		Poisson's Ratio	

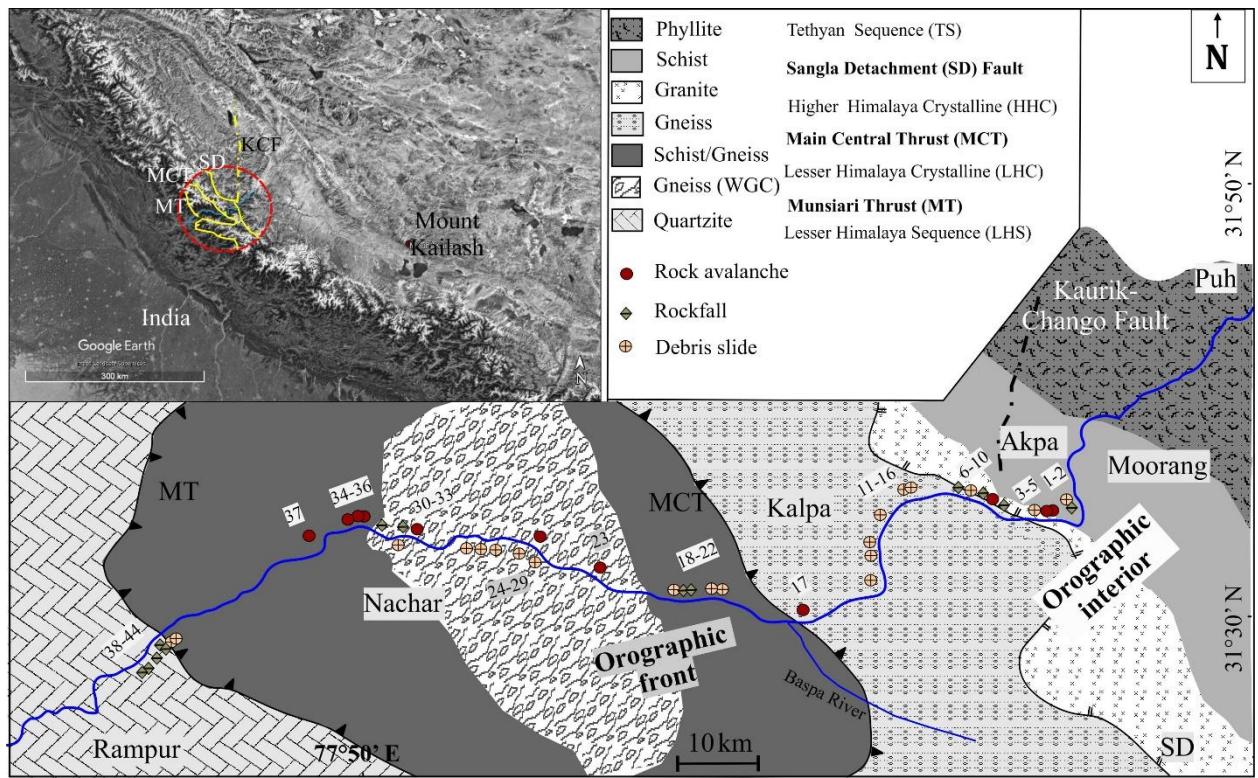
	$m_b = m_i e^{[(GSI-100)/(28-14D)]}$ <p>s and a; constants for the rock mass given by the following relationships;</p> $s = e^{[(GSI-100)/(9-3D)]}$ $a = \frac{1}{2} + \frac{1}{6} \left[ e^{-\left(\frac{GSI}{15}\right)} - e^{-\left(\frac{20}{3}\right)} \right]$ <p>Here, D; a factor which depends upon the degree of disturbance to which the rock mass has been subjected by blast damage and stress relaxation. GSI (Geological Strength Index); a rockmass characterization parameter.</p>	<p>Geological Strength Index</p> <p>Material Constant (<math>m_i</math>)</p> <p><math>m_b</math></p> <p><math>s</math></p> <p><math>a</math></p> <p><math>D</math></p>	<p>Field observation and based on recent amendments (Cai et al. 2007 and reference therein)</p> <p>Standard values (Hoek and Brown 1997)</p> <p>GSI was field depenedent, <math>m_i</math> as per(Hoek and Brown 1997) and D is used between 0-1 in view of rockmass exposure and blasting.</p>
Joint	<p><b>Barton-Bandis Criteria</b> (Barton and Choubey 1977; Barton and Bandis 1990)</p> $\tau = \sigma_n \tan [\phi_r + JRC \log_{10} (JCS/ \sigma_n)]$ <p>Here, <math>\tau</math> is joint shear strength; <math>\sigma_n</math>, normal stress across joint; <math>\phi_r</math>, reduced friction angle; JRC, joint roughness coefficient; JCS, joint compressive strength.</p> <p>JRC is based on the chart of Barton and Choubey (1977); Jang et al. (2014).JCS was determined using following equation;</p> $\log_{10}(JCS) = 0.00088 (R_L)(\gamma) + 1.01$ <p>Here, <math>R_L</math> is Schimdt Hammer Rebound value and <math>\gamma</math> is unit weight of rock.</p> <p>The JRC and JCS were used as <math>JRC_n</math> and <math>JCS_n</math> following the scale corrections observed by Barton and Choubey (1977) and reference therein and proposed by Barton and Bandis (1982).</p> $JRC_n = [JRC(L/L_o)^{-0.02(JRC)}]$ $JCS_n = [JCS(L/L_o)^{-0.03(JRC)}]$ <p>Here, <math>L</math> and <math>L_o</math> are mean joint spacing in field and, respectively. <math>L_o</math> has been suggested to be 10 cm.</p> <p><b>Joint stiffness criteria</b> (Barton 1972)</p> $k_n = (E_i * E_m) / L * (E_i - E_m)$ <p>Here, <math>k_n</math>; Normal stiffness, <math>E_i</math>; Intact rock modulus, <math>E_m</math>; Rockmass modulus L; Mean joint spacing.</p> $E_m = (E_i) * [0.02 + \{1 - D/2\} / \{1 + e^{(60+15*D-GSI)/11}\}]$ <p>Here, <math>E_m</math> is based on Hoek and Diederichs (2006) and reference therein</p>	<p>Normal Stiffness, <math>k_n</math> (MPa/m)</p> <p>Shear Stiffness, <math>k_s</math> (MPa/m)</p> <p>Reduced friction angle, <math>\phi_r</math></p> <p>Joint roughness coefficient, JRC</p> <p>Joint compressive strength, JCS (MPa)</p> <p>Scale corrected, <math>JRC_n</math></p> <p>Scale corrected, <math>JCS_n</math> (MPa)</p>	<p><math>E_i</math> is lab dependent. <math>L</math> and GSI were field depenedent. D is used between 0-1 in view of rockmass exposure and blasting.</p> <p>It is assumed as <math>k_n/10</math>. However, effect of denominator is aslo obtainedthrough parameteric study.</p> <p>Standard values ( Barton and Choubey 1977).</p> <p>Field based data from profilometer and standard values from Barton and Choubey (1977); Jang et al. (2014).</p> <p>Empirical equationof Deere and Miller (1966) relating Schimdt Hammer Rebound (SHR) values, <math>\sigma_{ci}</math> and unit weight of rock. SHR was field dependent.</p> <p>Empirical equation of Barton and Bandis (1982).</p>
	<b>Mohr-Coulomb Criteria</b>	Unit Weight (MN/m <sup>3</sup> )	Laboratory analysis (UCS) (IS: 2720-Part 4-1985; IS: 2720-Part 10-1991)

Soil	(Coulomb 1776; Mohr 1914) $\tau = C + \sigma \tan\phi$	Young's Modulus, $E_i$ (MPa)	Laboratory analysis (UCS); IS: 2720-Part 10-1991.
	Here, $\tau$ ; Shear stress at failure, $C$ ; Cohesion, $\sigma_n$ ; normal strength, $\phi$ ; angle of friction.	Poisson's Ratio	Standard values from Bowles (1996)
		Cohesion, $C$ (MPa)	Laboratory analysis (Direct shear) (IS: 2720-Part 13- 1986)
		Friction angle, $\phi$	

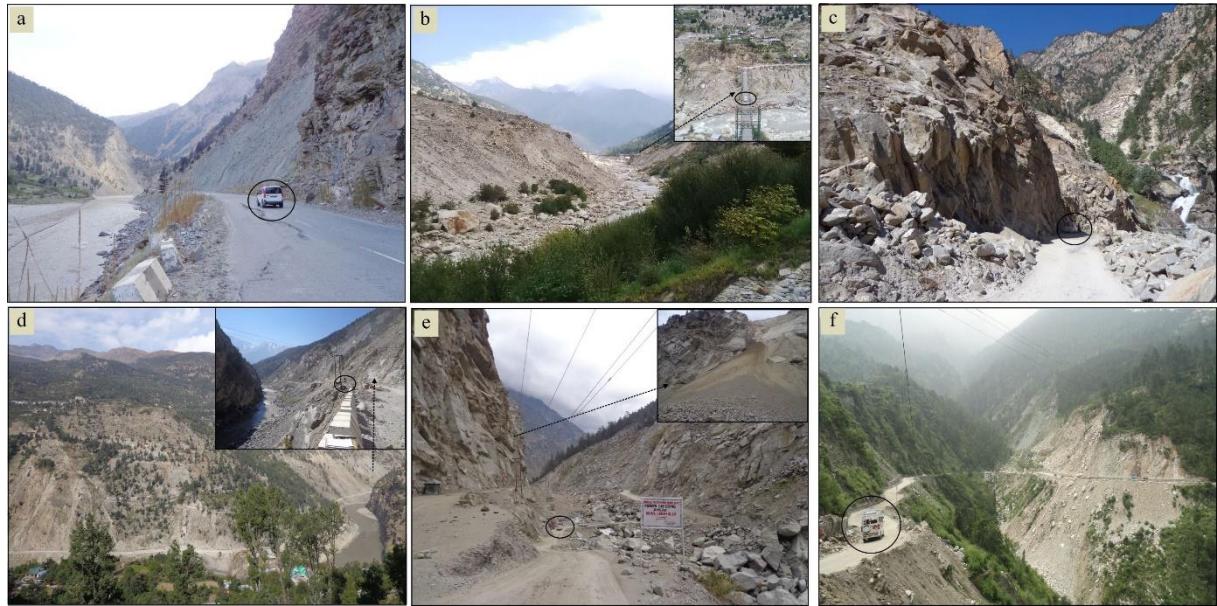
**Table 4** Details of input parameters for run-out analysis. S.N. refers to serial number of landslides in Fig. 1.

Landslide	Material type	Material depth <sup>1</sup> , m	Friction coefficient <sup>2</sup>	Turbulence coefficient <sup>3</sup> , m/sec <sup>2</sup>
Akpa (S.N. 5)	Gravelly sand	5	$\mu = 0.05, 0.1, 0.3$	$\xi = 100, 200, 300$
Baren Dogri (S.N. 7)	Gravelly sand	1.25	$\mu = 0.05, 0.1, 0.4$	$\xi = 100, 200, 300$
Pawari (S.N. 14)	Gravelly sand	1.25	$\mu = 0.05, 0.1, 0.4$	$\xi = 100, 200, 300$
Telangi (S.N. 15)	Gravelly sand	6.25	$\mu = 0.05, 0.1, 0.4$	$\xi = 100, 200, 300$
Urni (S.N. 19)	Gravelly sand	2.5	$\mu = 0.06, 0.1, 0.4$	$\xi = 100, 200, 300$

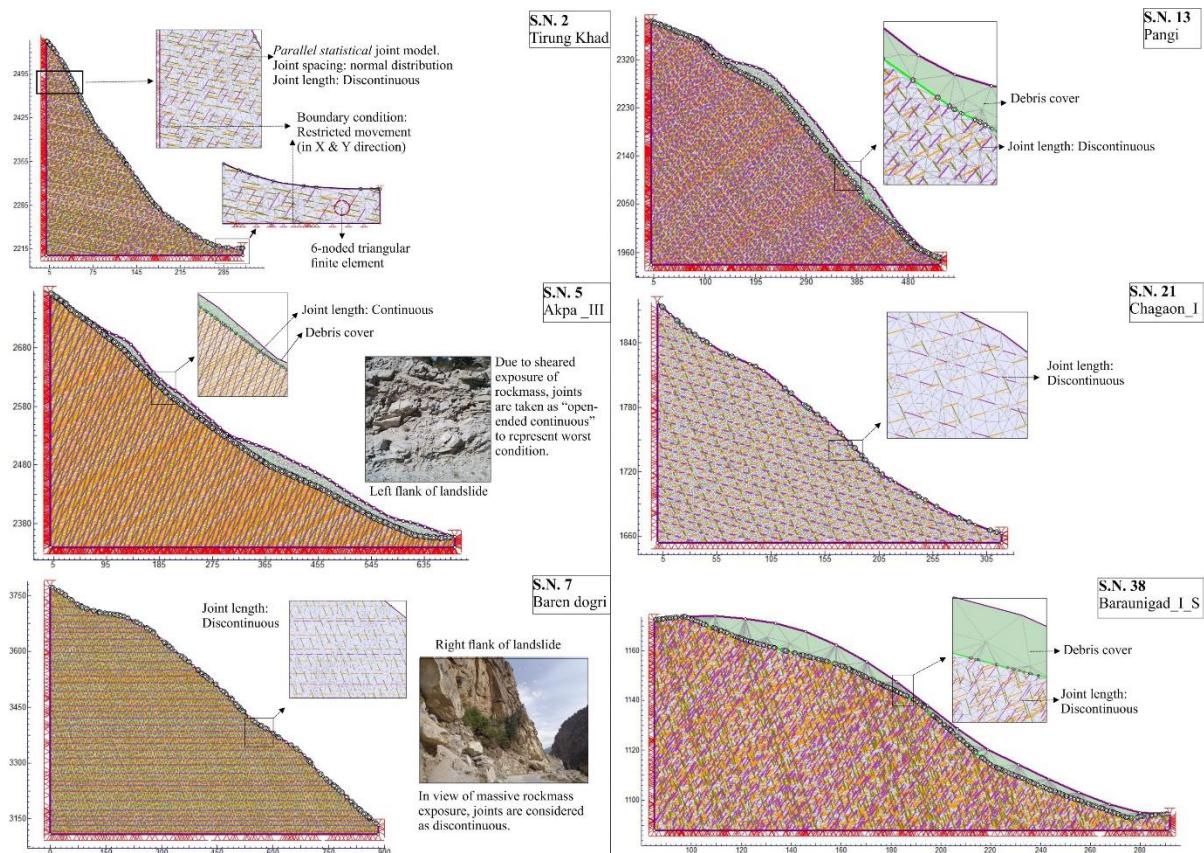
<sup>1</sup>Considering that fact that during the slope failure, irrespective of type of trigger, entire loose material might not slide down, the depth is taken as only  $\frac{1}{4}$  (thickness) in the calculation.<sup>2</sup> Since the angle of run-out track (slope and river channel) varied a little beyond the suggested range  $2.8^\circ$  - $21.8^\circ$  or  $\mu = 0.05$ - $0.4$  (Hung et al., 1984; RAMMS v.1.7.0), we kept out input in this suggested range wherever possible to avoid the simulation uncertainty. <sup>3</sup>This range is used in view of the type of loose material i.e., granular in this study (RAMMS v.1.7.0).



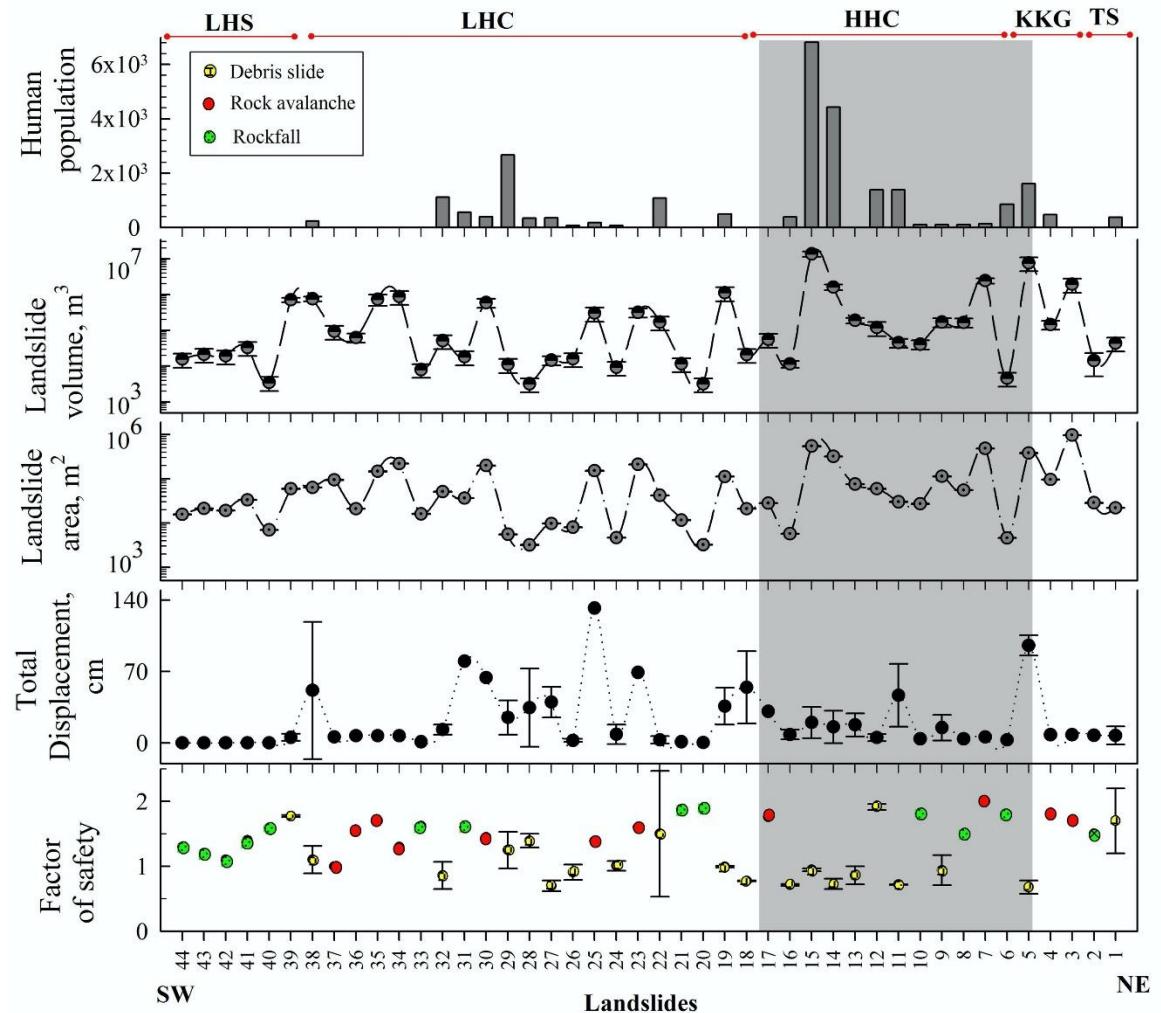
**Fig. 1** Geological setting. WGC: Wangtu Gneissic Complex. The red dashed circle in the inset represents the region within 100 km radius from the Satluj River (marked as blue line) that was used to determine the earthquake distribution in the area. Yellow lines represent the regional faults in the region. KCF in inset refers to Kaurik-Chango Fault. The numbers 1-44 refer to serial number of landslides in Table 1.



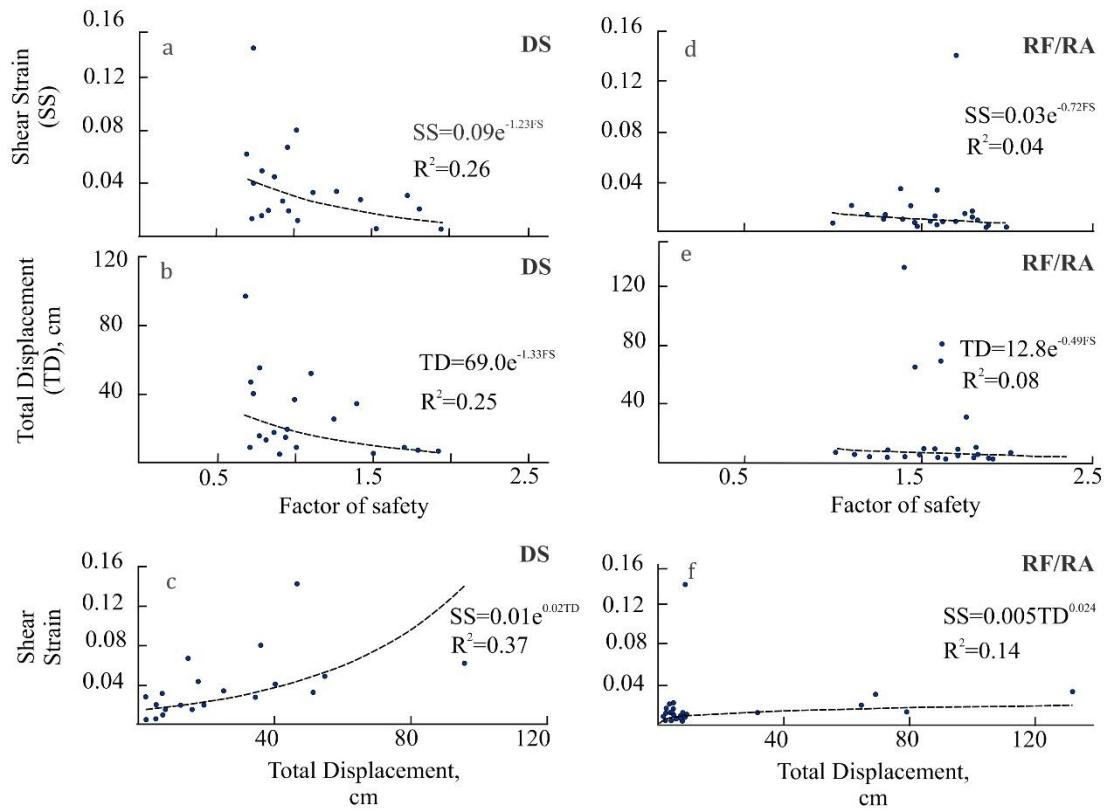
**Fig. 2** Field photographs of some of the landslides (a) Khokpa landslide (S.N.1); (b) Akpa\_III landslide (S.N. 5); (c) Rarang landslide (S.N. 6); (d) Pawari landslide (S.N.14); (e) Urni landslide (S.N.19); (f) Barauni Gad\_I\_S landslide (S.N. 38). Black circle in the pictures that encircles the vehicle is intended to represent the relative scale.



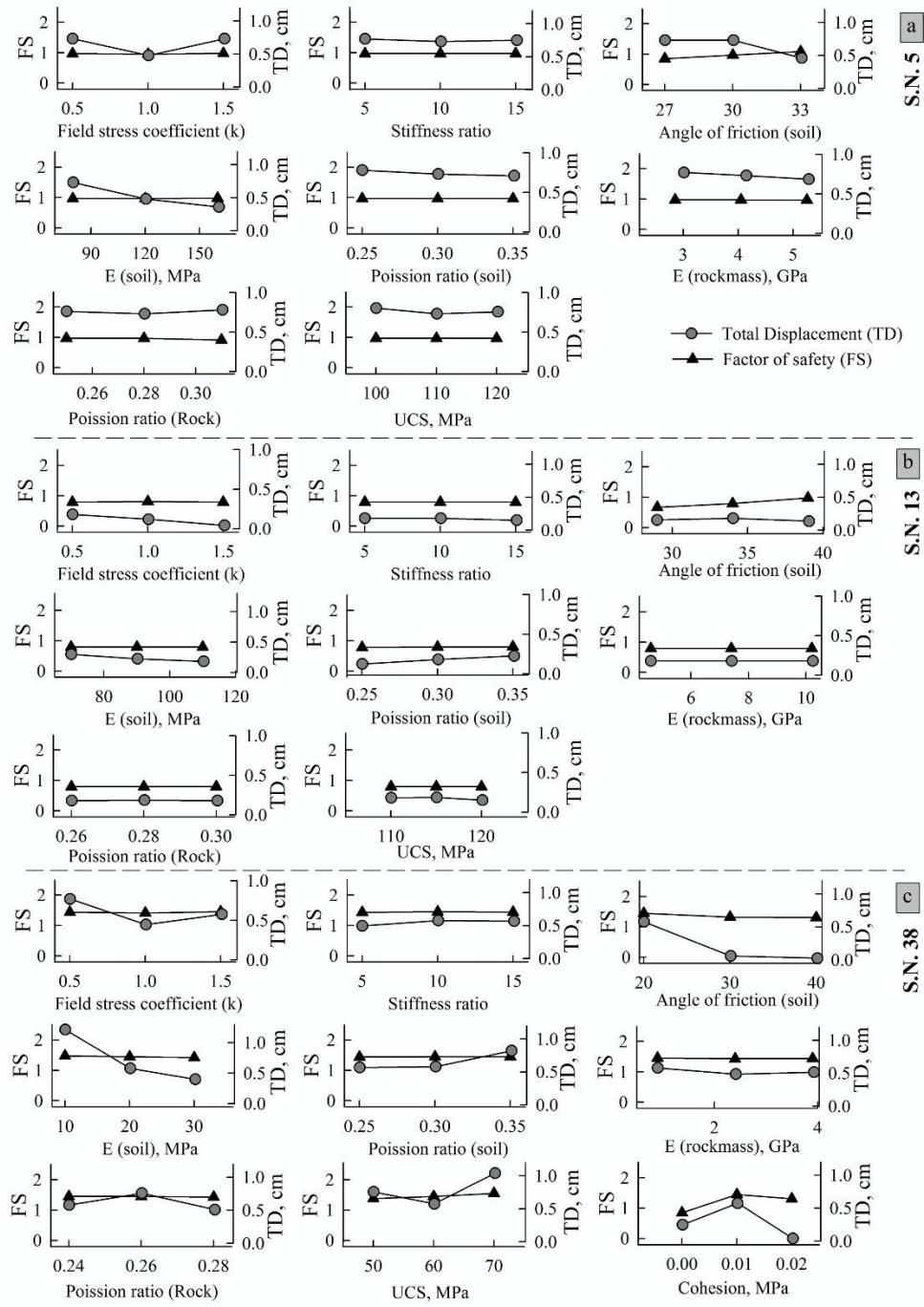
**Fig. 3** The FEM configuration of some of the slope models. S.N. refers to the serial no. of landslides in Table 1. The joint distribution in all the slopes was parallel-statistical with the normal distribution of joint spacing.



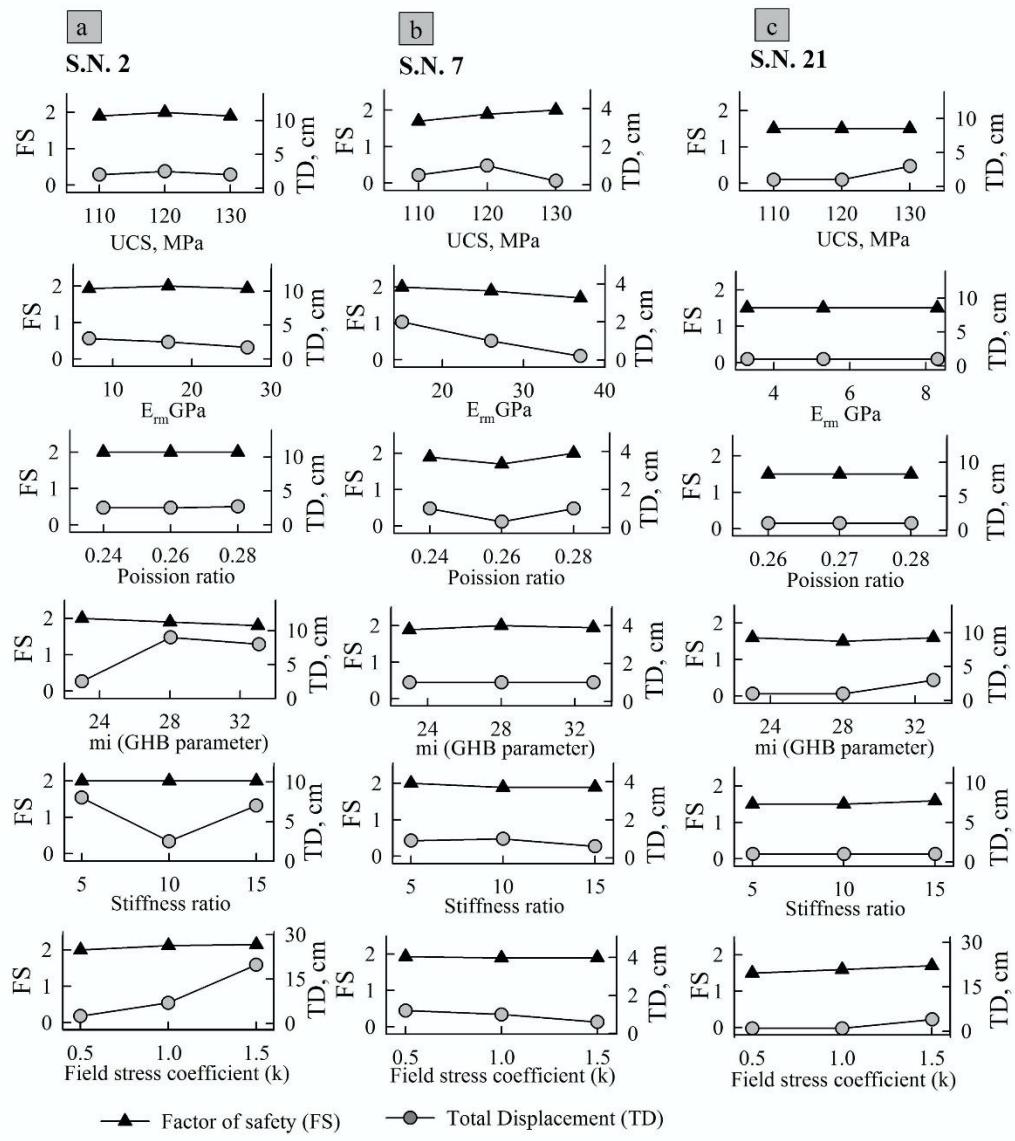
**Fig. 4** The FEM analysis of all forty-four landslides. Grey bar in the background highlights the Higher Himalaya Crystalline (HHC) region that comprises relatively more unstable landslides, landslide volume and human population..TS, KKG, HHC, LHC and LHS are Tethyan Sequence, Kinnaur Kailash Granite, Higher Himalaya Crystalline, Lesser Himalaya Crystalline and Lesser Himalaya Sequence, respectively.



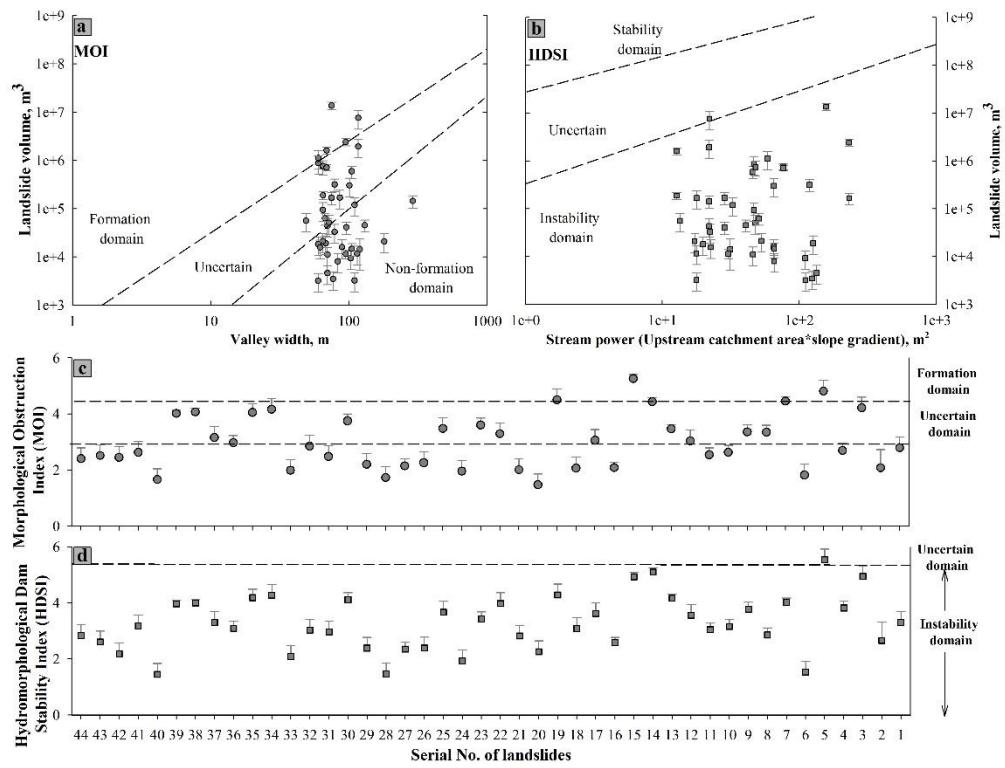
**Fig. 5** Relationship of Factor of Safety (FS), Total Displacement (TD) and Shear Strain (SS). DS, RF, and RA refer to Debris slide, rock fall and rock avalanche, respectively.



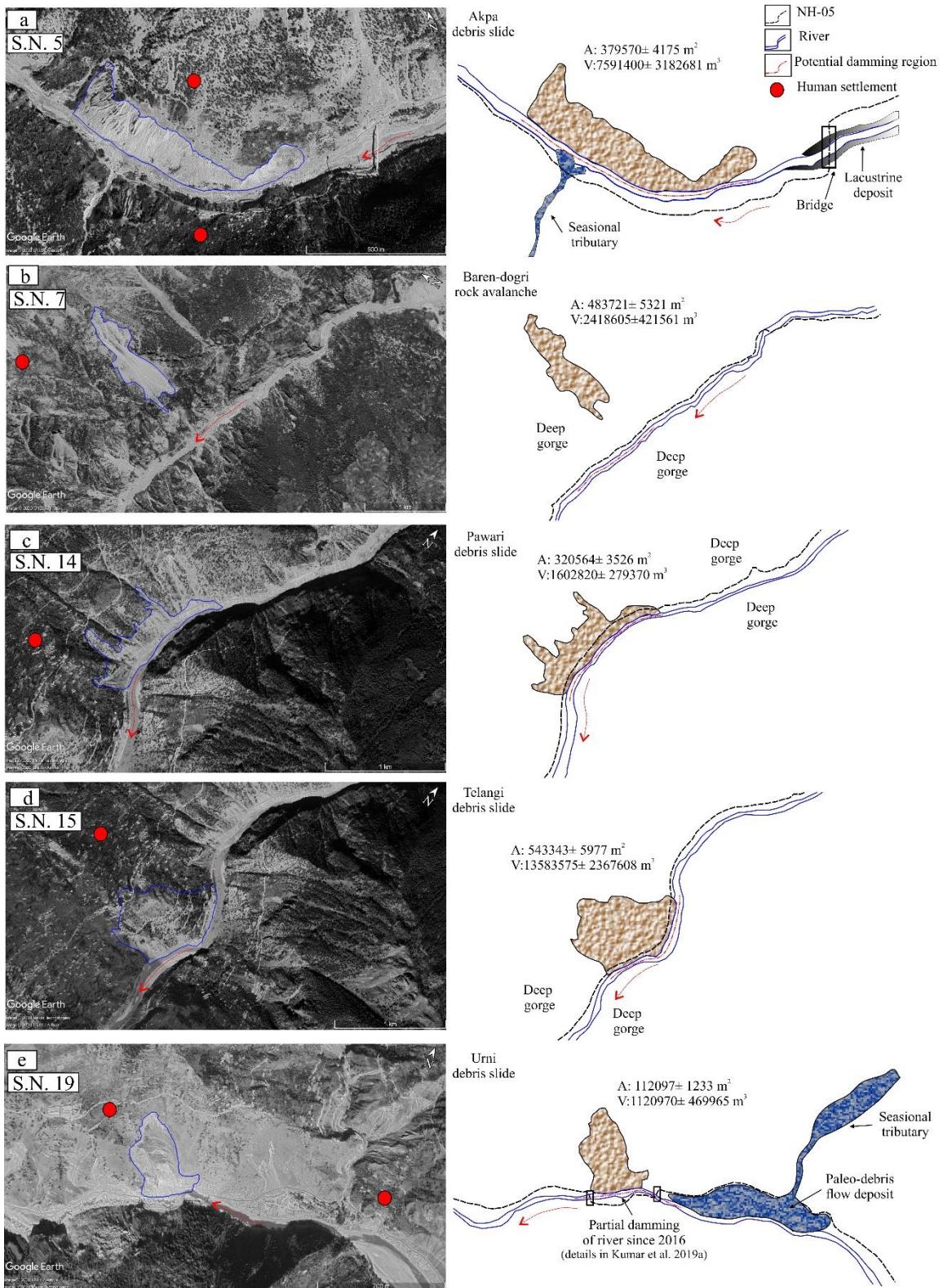
**Fig. 6** Parametric analysis of debris slides. (a) Akpa\_III (S.N. 5); (b) Pangi\_III (S.N. 13); (c) Barauni Gad\_I\_S (S.N. 38). S. N. refers to the serial no. of landslides in Table 1.



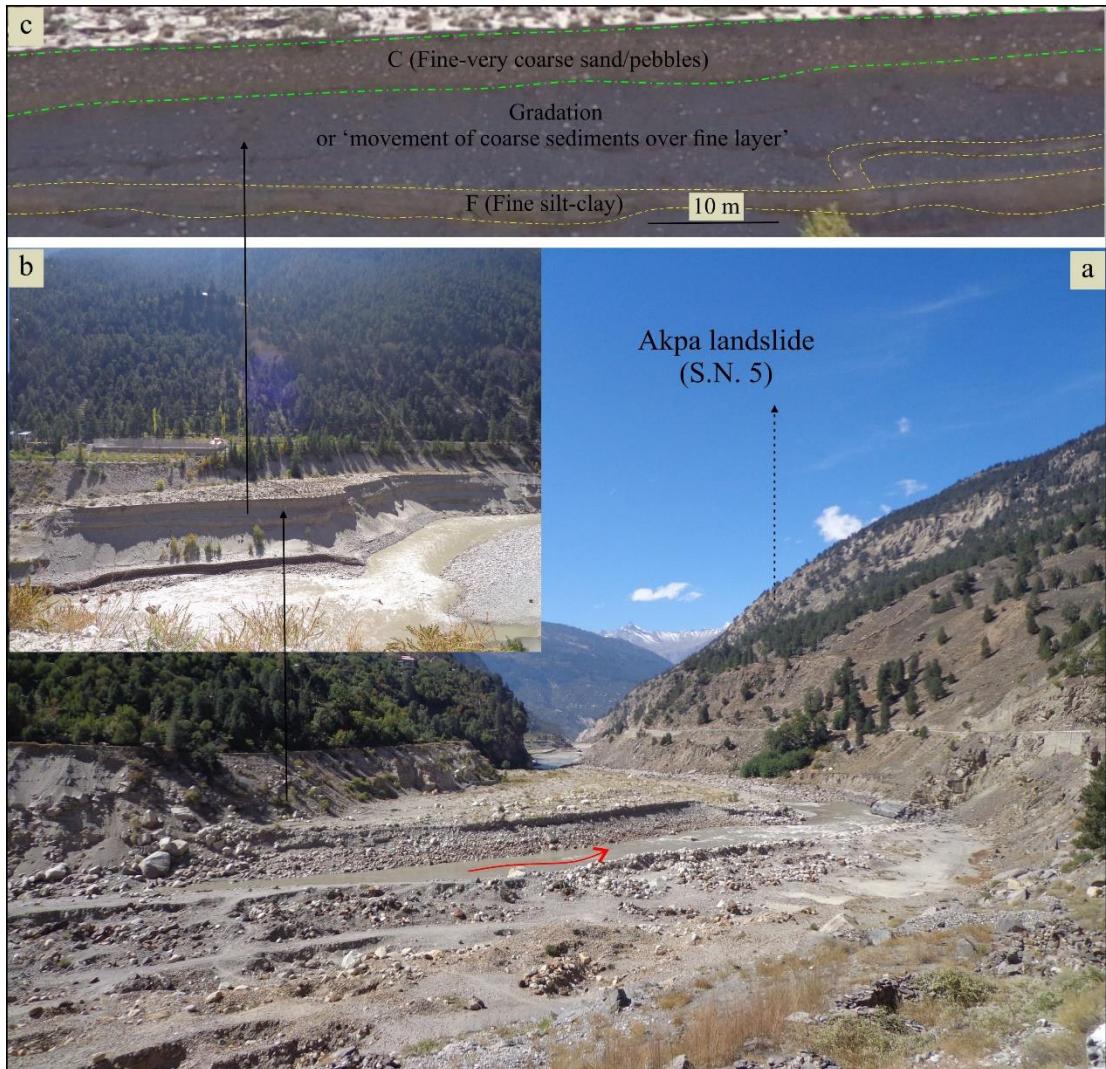
**Fig. 7** Parametric analysis of rockfall/rock avalanche. (a) Tirung khad (S.N. 2); (b) Baren Dogri (S.No. 7); (c) Chagaon\_II (S.N. 21).



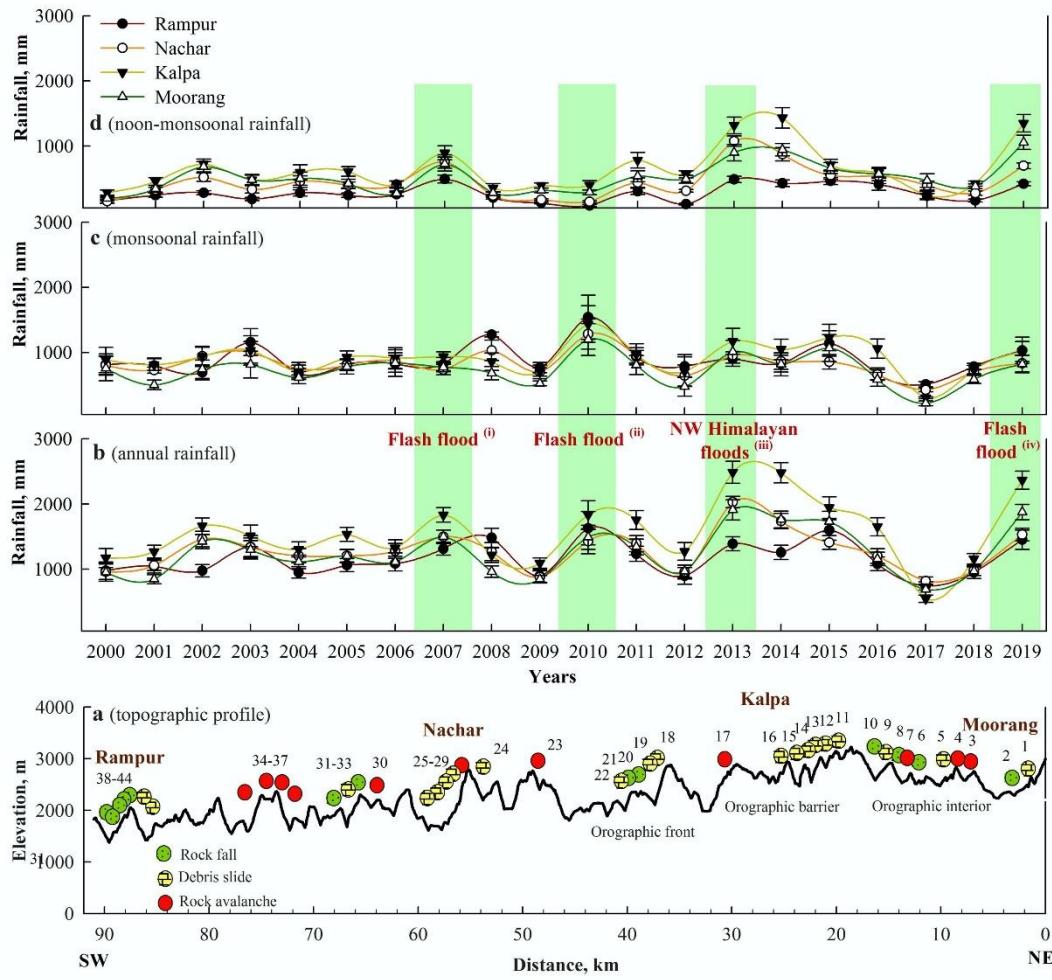
**Fig. 8** Landslide damming indices (a) Morphological Obstruction Index (MOI); (b) Hydro-morphological dam stability index (HDSI); (c) Landslides vs. MOI; (d) Landslides vs. HDSI.



**Fig. 9** Potential landslide damming locations. (a) Akpa III landslide; (b) Baren dogri landslide; (c) Pawari landslide; (d) Telangi landslide; (e) Urni landslide. Red dashed-arrow represents the direction of river flow.

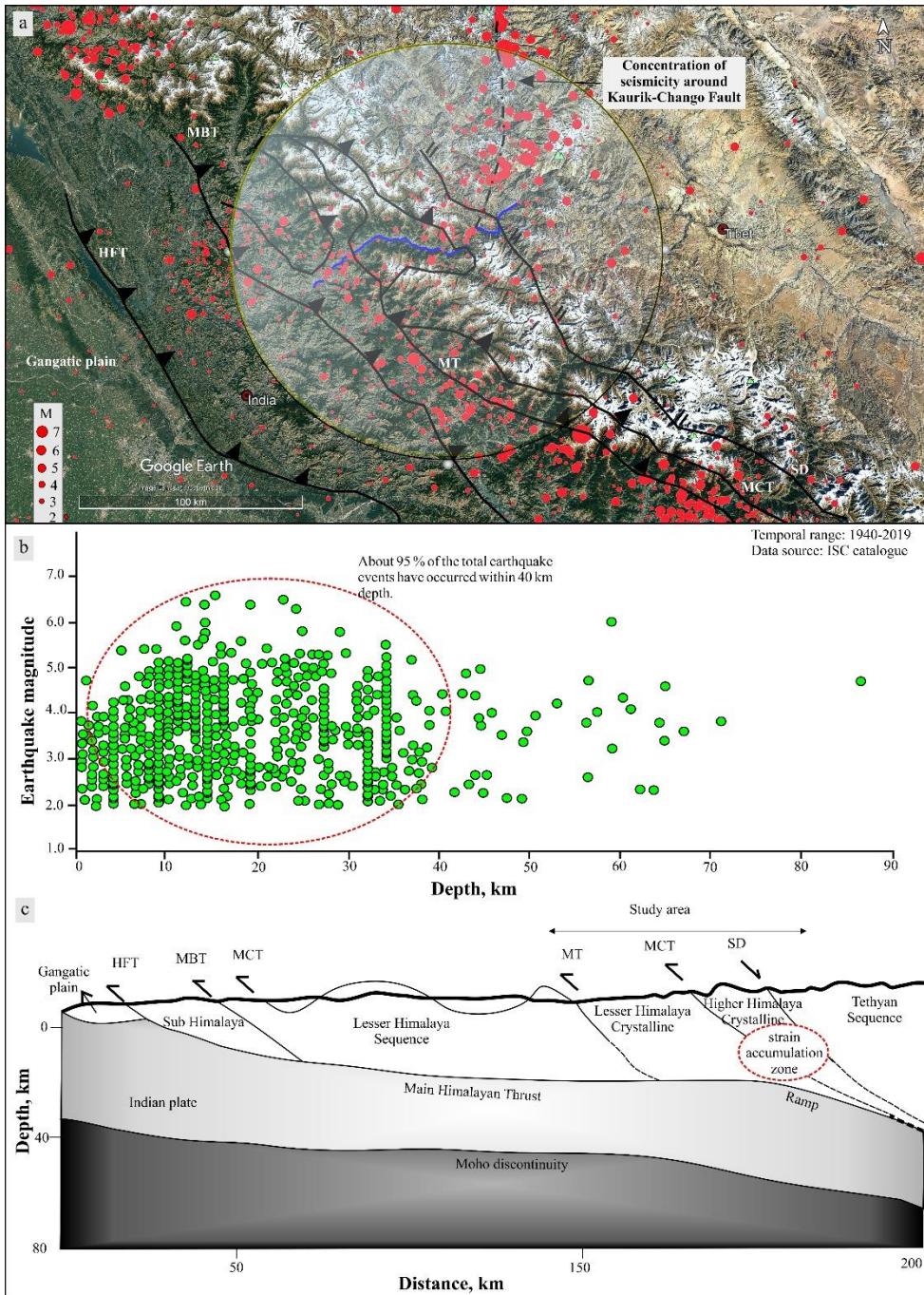


**Fig. 10** Field signatures of the landslide damming near Akpa\_III landslide. (a) Upstream view of Akpa landslide with lacustrine deposit at the left bank; (b) enlarged view of the lacustrine deposit with an arrow indicating the lacustrine sequence; (c) alternating fine-coarse sediments. F and C refer to fine (covered by yellow dashed lines) and coarse (covered by green dashed lines) sediments, respectively.

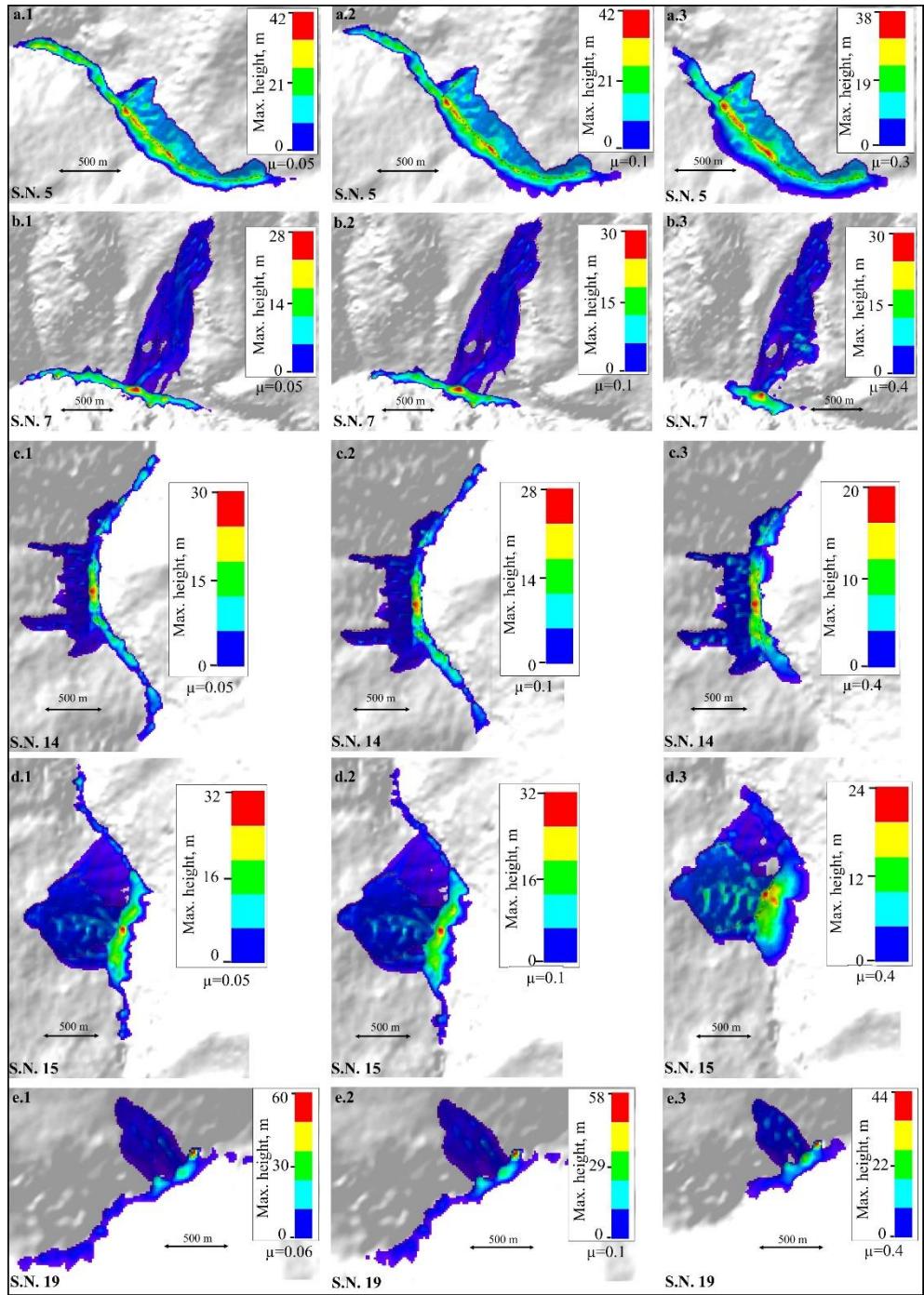


**Fig.**

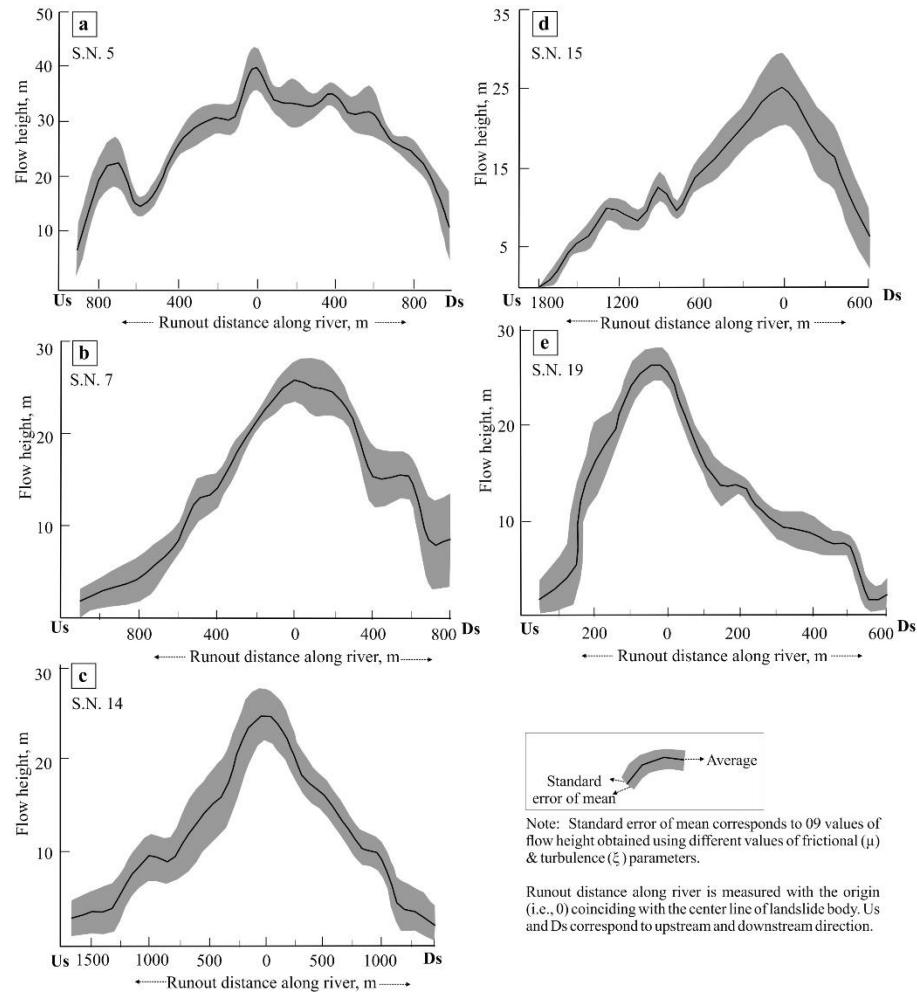
**11** Rainfall distribution. (a) Topographic profile; (b) annual rainfall; (c) monsoonal (June-Sep.) rainfall; (d) non-monsoonal (Oct.-May) rainfall. Green bars represent the years of relatively more rainfall resulting into the flash floods, landslides and socio-economic loss in the region. (i):hpenvis.nic.in, retrieved on March 1, 2020; Department of Revenue, Govt. of H.P. (ii): hpenvis.nic.in, retrieved on March 1, 2020.(iii): Kumar et al., 2019a;ndma.gov.in, retrieved on march 1, 2020 (iv):sandrpr.in, retrieved on march 1, 2020.The numbers 1-44 refer to serial number of the landslides.



**Fig. 12** Earthquake distribution. (a) Spatial variation of earthquakes. The transparent circle represents the region within 100 km radius from the Satluj River (blue line). The black dashed line represents the seismic dominance around the Kaurik-Chango fault; (b) earthquake magnitude vs. focal depth. The red dashed region highlights the concentration of earthquakes within 40 km depth; (c) Cross section view (Based on Hazarika et al. 2017; Bilham, 2019). Red dashed circle represents the zone of strain accumulation caused by the Indian and Eurasian plate collision (Bilham, 2019). ISC: International Seismological Centre. HFT: Himalayan Frontal Thrust.



**Fig. 13** Results of the run-out analysis.  $\mu$  refers to coefficient of friction.



**Fig. 14** Results of run-out analysis at different values of  $\mu$  and  $\xi$ .  $\mu$  and  $\xi$  refer to coefficient of friction and turbulence, respectively.

Note: Standard error of mean corresponds to 09 values of flow height obtained using different values of frictional ( $\mu$ ) & turbulence ( $\xi$ ) parameters.

Runout distance along river is measured with the origin (i.e., 0) coinciding with the center line of landslide body. Us and Ds correspond to upstream and downstream direction.