

# The spatial distribution of injuries in need of surgical intervention in Nepal

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## Abstract

Geographic information system modelling can accurately represent the geospatial distribution of disease burdens to inform health service delivery. Given the dramatic topography of Nepal and a high prevalence of unmet surgical needs, we explored the consequences of topography on the prevalence of surgical conditions. The Nepalese Surgeons OverSeas Assessment of Surgical Need (SOSAS) is a validated, countrywide, cluster randomised survey that assesses surgical need in low- and middle-income countries; it was performed in Nepal in 2014. Data on conditions potentially affected by topography (e.g. fractures, hernias, injuries, burns) were extracted from the database. A national digital elevation model was used to determine altitude, aspect, slope steepness and curvature of the SOSAS survey sites. Forward stepwise linear regression was performed with prevalence of each surgical con-

dition as the response variable and topographic data as explanatory variables. The highest correlation coefficient was for models predicting hernias and fractures, both explaining 21% of the variance. The model fitted to death due to fall would become significant when an outlier was excluded ( $P < 0.001$ ;  $R^2 = 0.27$ ). Excluding the outlier yielded a better-fitted model to burn injury (stepwise regression) without any explanatory variables. Other models trended towards a correlation, but did not have sufficient power to detect a difference. This study identified slight correlation between elevation and the prevalence of hernias and fall injuries. Further investigation on the effects of topography and geography on surgical conditions is needed to help determine if the data would be useful for directing allocation of surgical resources.

## Introduction

With increasing empirical data globally, surgical care is gaining momentum as a public health priority (Groen *et al.*, 2012; Petroze *et al.*, 2013; Gupta *et al.*, 2015; Bae *et al.*, 2011; Meara *et al.*, 2014). Global estimations conclude that 73.6% of all surgical operations are carried out in the wealthiest nations, while only 3.5% of the operations are performed in the world's poorest nations (Weiser *et al.*, 2008). Approximately 2 billion people lack access to an operating room (Funk *et al.*, 2010). Despite the recent interest in global surgical needs, there is still a paucity of objective data needed to define which barriers make access to surgical care so difficult in the world's poorest nations, and if surgical conditions vary depending on geospatial considerations globally. Geographic information systems (GIS) are increasingly used to help model the geospatial distribution of diseases (Rocha *et al.*, 2014; Faierman *et al.*, 2014; Tollefson *et al.*, 2015). Such modelling can provide an accurate, geographic representation of the spatial variation of disease burdens, which can help to guide policymakers in developing future service plans, tackling this aspect of healthcare access. In addition to planning for the location of healthcare services with the aim of providing support where the need is greatest, geospatial modelling can also be used to investigate the extent to which biophysical conditions, such as the presence of parasite reservoirs or how the topography of a country affects the prevalence of medical conditions (Osei *et al.*, 2010). Nepal is a South Asian country with just over 27 million people. Situated between China in the North and India in the South, West and East, Nepal is a landlocked nation with diverse geography, housing eight of the world's ten largest mountains in addition

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to numerous lower hills as well as lakes (World Bank, 2013). The terrain rises from a low of 59 m above the mean sea level in the Terai Region at the northern rim of the Gangetic Plain to 8848 m, representing Earth's highest peak Mt. Everest (World Bank, 2013). This wide range of altitudes, and the associated variation in topographic conditions (from gentle flat areas to steep, rugged terrain), provides a useful case study to explore the possible consequences of topography on the prevalence of surgical conditions, such as fractures, hernias and burns. This is an important aspect as was vividly brought home by the devastation of the country in April 2015 through the Gorkha Earthquake, one of the worst natural disasters in its history, which killed more than 8000 people and injuring more than 23,000 (Central News Network, 2015). Such an event in an already low-income and politically fragile country such as Nepal further complicates access to surgical care.

Prior to the 2015 earthquake, the Surgeons Overseas Assessment of Surgical Need (SOSAS) survey was used in Nepal in 2014 to assess the prevalence of surgical conditions and avoidable deaths. The objective of the present study is to use the data to explore the effect of topography on prevalence of surgical conditions in Nepal.

## Materials and Methods

The SOSAS survey, a validated, countrywide, cluster randomised, cross-sectional survey, was performed in Nepal from May 25 to June 12, 2014. It consists of two parts, the first of which concerns household demographic data, access to healthcare and household member deaths within the past year. The second part randomly selects two household members for a verbal *head-to-toe* examination focusing on six anatomical regions: i) face, head and neck; ii) chest and breast; iii) abdomen; iv) groin and genitalia; v) back; and vi) extremities. Each respondent verbally elicits symptoms or experiences associated with a general spectrum of surgical conditions, such as wounds, swellings, deformities, burns and injuries. The SOSAS survey has been described in more detail previously (Groen *et al.*, 2012).

### Sampling

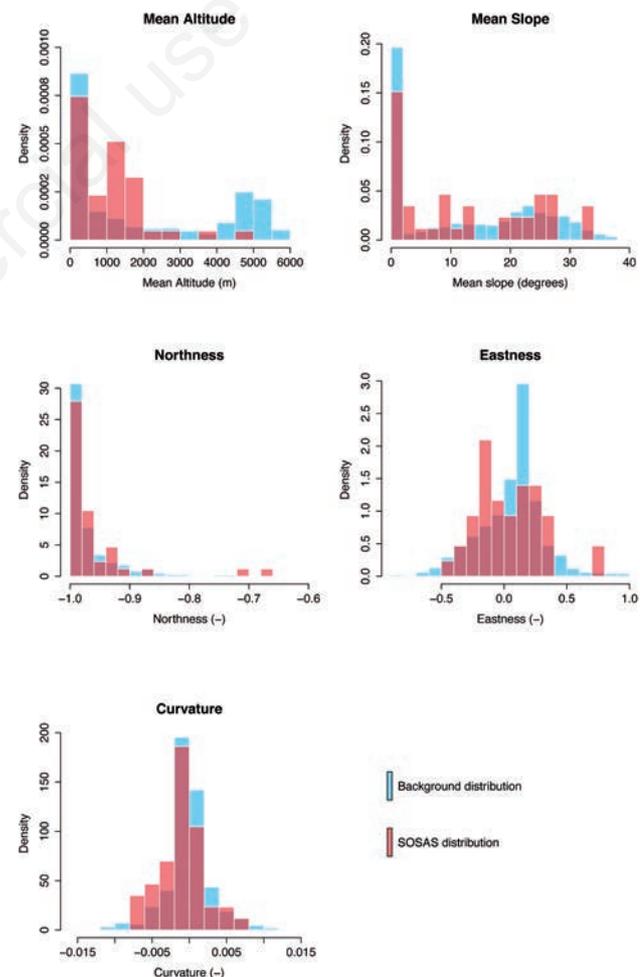
A two-stage cluster sampling was performed. First, 15 of the 75 districts proportional to the population were randomly selected, after which 45 Village Development Committees (VDCs) were randomly selected (Figure 1), three for each district after stratification for the urban to rural population distribution (two rural to one urban). This methodology was similar to that used by the Demographic and Health Surveys normally carried out in Nepal (Nepal Ministry of Health and Population, Department of Health Services, 2012). A total sample size of 1350 households was used. The sample size estimation was calculated from a prevalence of *unmet surgical needs* of 5%, reported by a pilot study of SOSAS in Nepal in January 2014 (Gupta *et al.*, 2014). All surveys were administered in Nepalese and the responses recorded in English via paper forms. Appropriate ethical approval was obtained prior to study execution. Verbal consent was obtained from all respondents prior to survey (parental consent, oral assent and/or parental permission was obtained for individuals younger than 18 years). Individuals noted to be cognitively impaired by household members were excluded from the study.

### The Surgeons Overseas Assessment of Surgical Need database

By consensus among the authors, surgical conditions that might be

affected by altitude were chosen for inclusion in this study. Given the increased strenuous exercise needed at higher altitudes, the augmented risk of falling from heights and lower temperatures (leading to the necessity of fire for warmth), data on fractures, hernias, fall injuries (both fatal and non-fatal), burn injuries, and unmet surgical needs in general were extracted from the SOSAS database and used for the analysis. Unmet surgical need was defined as the affliction of an individual, who had reported a current (within one month) condition that he or she perceived needed at least a surgical consultation, which could not be accessed. Prevalence was calculated as the number of recordings of a medical condition relative to the total number of people surveyed.

The spatial database was constructed as follows. A digital elevation model (DEM) for the whole of Nepal with a resolution of 90 by 90 m (Figure 2A) was downloaded from the United States Geological Survey (USGS), created by the shuttle radar topography mission (SRTM) (<http://srtm.usgs.gov/>). Next to altitude, aspect (degrees from north), slope steepness (in degrees) and curvature (the rate at which the slope changes) were calculated from the DEM. For every location where



**Figure 1.** Distribution of topographical conditions that were used in this study from a random selection of 1000 locations in Nepal (*i.e.* the background) and sites sampled by the Surgeons Overseas Assessment of Surgical Need (SOSAS). Darker red represents overlapping between background and SOSAS distribution.

SOSAS data had been collected, average and standard deviation of these factors were calculated within a 5 km radius from each site. Aspect was then converted to *northness* and *eastness*, which converts the degrees from north to a factor with a more straightforward interpretation. Thus, minimum northness (*i.e.* straight south) becomes -1, while maximum (*i.e.* straight north) = +1. Similarly, west signifies low levels of eastness (*i.e.* straight west is minimum eastness = -1), and high values indicate east (*i.e.* straight east = +1).

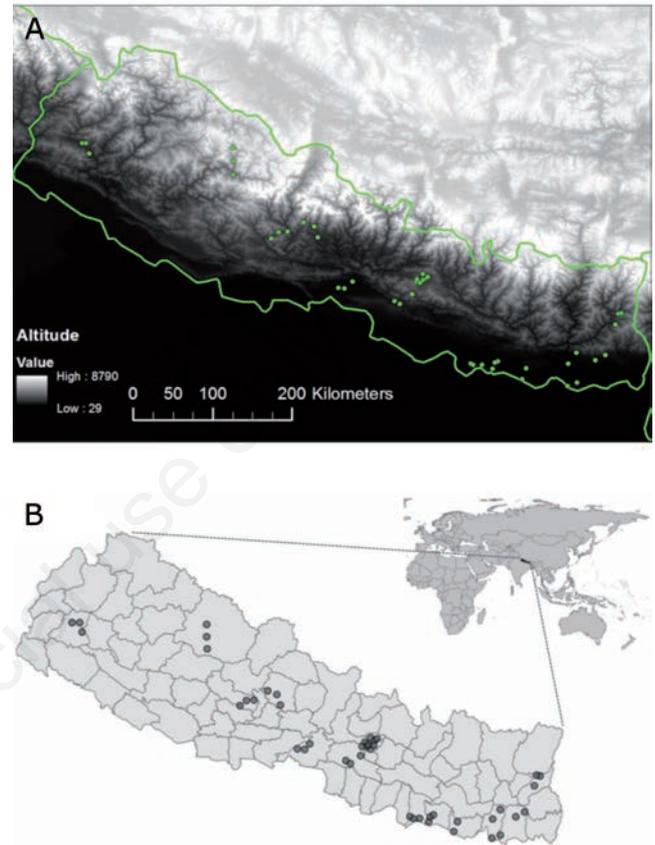
### Statistical analyses

Simple forward stepwise linear regression was performed in R (R Development Core Team, 2014) with prevalence of surgical conditions as response and topographic explanatory variables. Model selection was based on the Akaike Information Criterion (AIC) (Akaike, 1978). The AIC gives a goodness of fit indication based on the number of explanatory variables included and sample size of the dataset. Preference is given to models with fewer explanatory variables. Before fitting the models, collinearity was checked by means of variance inflation factors (VIFs). A VIF upper threshold of 10 was used as criterion to identify too strong collinearity (Quinn and Keough, 2002). After final models were fitted, normality of residuals was checked to ensure that distribution of data was in accordance with model assumptions. Additionally, the degree of representativeness of SOSAS clusters for the topography in the region was made visible by creating histograms of the tested topographic determinants against a background set of 1000 random points in the same area.

## Results

The representativeness of the SOSAS sample for topography is shown in Figure 2B. The SOSAS database covers the distribution of topographic conditions well except for the really high (*i.e.* >5000 m) and steep (*i.e.* >34°) areas. Table 1 describes the variables retained by the stepwise procedure and the evidence for association with the selected surgical conditions. Because the variable selection was based on AIC, not every variable necessarily proves significant ( $P < 0.05$ ). Next to variables selected and their significance, overall model performance is expressed by means of the variance explained ( $R^2$ ) and the overall model significance based on an F-test. For a graphical assessment of

model fits, observed versus modelled scatter plots are provided in Figure 3. Overall model significance was low. The overall model performance was highest for the model explaining hernias and fractures (both 21% explained variance). For the models fitted to fall as a cause



**Figure 2.** Digital elevation model of Nepal based on United States Geological Survey shuttle radar topography mission data (A) and location of clusters included in the Surgeons OverSeas Assessment of Surgical Need database (B).

**Table 1.** Model summary for each surgical condition analysed.

	Fractures	Hernias	Death due to fall	Fall	Surgical needs	Burning injury
Intercept	<0.001	0.01	0.07	<0.001	<0.001	0.007
Mean altitude (m asl)	0.01	0.06	0.08	0.03		
Standard altitude (m asl)						
Mean slope (degrees)	0.007	0.01		0.11		0.13
Standard slope (degrees)	0.08					
Mean curvature (ratio)						
Northness (ratio)		0.03				
Eastness (ratio)	0.18					
Model $R^2$	0.21	0.21	0.07	0.11	na	0.05
Overall P	0.05	0.02	0.08	0.10	na	0.13

asl, above the mean sea level; na, not available. Values indicate P values of the partial slopes of every model based on t-tests, except for  $R^2$  (variance explained by the model) and the overall P value, which is based on an F-test (best fit statistical comparison). Italics indicates significance at  $P < 0.05$ .

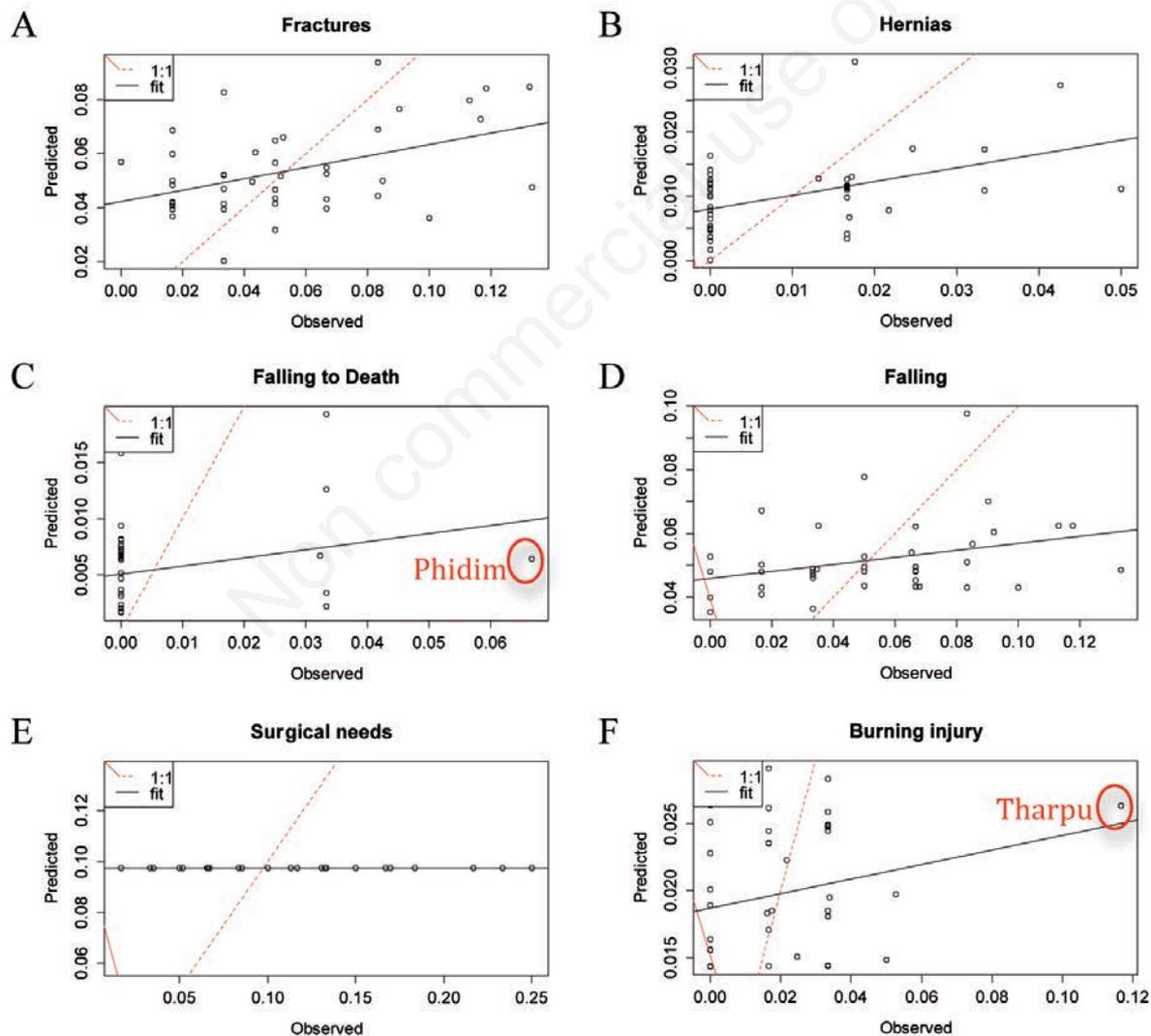
of death (Figure 3C) and burn injury (Figure 3F), one single location had a large leverage on the model. When outliers were excluded, the model fitted to fall injury as a cause of death becomes significant ( $P < 0.001$ ;  $R^2 = 0.27$ ); though excluding the outlier did not make the model fitted to the burn injury model significant.

## Discussion

The concept of access to healthcare has been defined as those dimensions that describe the potential and actual entry of a given population to the health care delivery system. The availability of health care is a component in this concept and refers to the volume and distribution of medical resources in an area (Andersen *et al.*, 1983). Much research on global surgery has equated access of adequate surgical care with the availability of trained surgeons and nurses as well as

physical facilities, such as operating theatres, autoclaves, utensils, etc., while other research papers have focused on characteristics of the population, such as demographics, income, perceived need for surgical care, transport to a primary healthcare facility and severity of symptoms (Groen *et al.*, 2012; Kushner *et al.*, 2010; LeBrun *et al.*, 2014; Ologunde *et al.*, 2014). However, apart from patients, staff, buildings, instruments, affordability and cultural acceptability of care there is another important component: *i.e.* geographical accessibility, specifically targeted towards areas with particular high needs for care.

Nepal is a country known for a great topographical variation and the SOSAS survey found a 10% prevalence of unmet surgical needs countrywide, and that was prior to the 2015 earthquake (Gupta *et al.*, 2015). Given the wide range of altitudes and the variation in topographic conditions in Nepal along with a high prevalence of unmet surgical needs, our aim was to explore the possible consequences of topography on the prevalence of surgical conditions. Though not conclusive, our study reveals a correlation (though fairly low) between elevation and the



**Figure 3.** Graphical representations of predicted *vs* observed prevalence rates of medical conditions as functions of topographical conditions. For C) and F), outliers that had a large leverage on the fitted model are identified. Phidim and Tharpu are Village Development Committees within Nepal that were sampled in this study.

prevalence of hernias and fall injuries. Geospatial analysis is a useful tool to identify the land cover of particular diseases to: prioritize risk-prone areas; examine healthcare misdistribution; and target areas for disease prevention. For example, a study from Bihar state in India mapped out risk prone areas for visceral leishmaniasis or kala-azar using a GIS approach. This study demonstrated that rural villages surrounded by a higher proportion of transitional swamps and sugarcane plantations had higher sand-fly abundance; and thus, were considered risk-prone areas for man-vector contact (Sudhakar *et al.*, 2006; Salahi-Moghaddam *et al.*, 2010). Another study from sub-Saharan Africa used geospatial modelling to optimize rollout plans for anti-retroviral therapy in South Africa (Gerberry *et al.*, 2014). This study found significant geographic variation in the efficiency of interventions in reducing the transmission of the human immunodeficiency virus using geographic targeting to help maximise geographic equity in access to interventions. Such data provide empirical evidence to plan targeted intervention. Similarly, this study sought to identify risk-prone areas for selected conditions requiring surgery, as such information might contribute to the development of targeted interventions to reduce unmet surgical needs in disproportionately affected populations. Furthermore, geospatial analysis can help explore if changes in altitude have a causal association for surgical conditions from a physiological basis. While this study was not equipped to explore a causal relationship between altitude and surgical conditions, previous studies have explored this relationship with regard to infectious diseases, such as pulmonary tuberculosis (Vargas *et al.*, 2004). A Mexican study reveals that altitude had a strong inverse relationship to pulmonary tuberculosis, perhaps related to the well-known changes in alveolar oxygen pressure at different altitudes. Such techniques can help explore if certain surgical conditions are more likely at higher altitudes from a physiological perspective. Our results show that in Nepal, slope correlated with the numbers of both fractures and hernia prevalence, while altitude was correlated with fractures and death due to fall injuries, though not necessarily all fall injuries (non-fatal included). Other data analysed trended toward a correlation, but this study, as an exploratory analysis, did not have sufficient statistical power to detect a difference with respect to other conditions. The correlations noted may be explained, in part, by: i) physiology of living at higher elevations; ii) a greater relative risk for incurring surgical conditions that populations living in more complex terrain have (*e.g.* falling leading to fracture; repeated strenuous activity leading to hernia); or iii) having less access to appropriate surgical care. However, we did not find evidence for a difference between the prevalence of any unmet surgical need at differing elevations or terrain ruggedness.

Clearly, the large effect that a few observations had on the overall performance of some of the models (Phidim and Tharpu) shows that the spatial spread of these observations and the total sample size was probably not optimal for this study. Nevertheless, these data demonstrate that there may be some relationships between terrain and surgical conditions that need to be better defined. This calls for efforts to collate databases, such as the SOSAS surveys, to analyse these relationships, which then in turn can be used to prioritise the allocation of scarce surgical resources to areas where they are most needed. As geospatial technology continues to improve, such technologies should be included in current data collection strategies and evaluations.

Our study was executed prior to the Gorkha earthquake in 2015, thus many more individuals in mountainous regions without access to healthcare likely exist than were noted in our study. In order to confront this issue, Nepal instituted emergency mobile health clinics post-disaster (World Health Organization, 2015), thus allowing more Nepalese to have access to healthcare. This study has several limita-

tions. Limitations inherent to all cross sectional studies with similar random sampling methodology should be considered. Despite the use of data from a robustly sample-sized survey, resources only allowed for 15 of Nepal's 75 districts to be selected. Because these districts were selected proportional to population, this means that more densely populated areas had a higher chance of being selected resulting in less developed regions and less densely populated areas of Nepal being underrepresented. These more rural areas in Nepal tend to be more mountainous and with higher elevations. Thus, the SOSAS study is a good start to explore such topographical relationships, but it is likely that a dedicated study focusing on both the higher and lower elevations in Nepal is needed to build more conclusive results. SOSAS relies on verbally self-reported data, which may be prone to recall bias. However, as part of SOSAS Nepal, a visual physical examination was performed, which validated these verbal reports (Gupta *et al.*, 2015). Despite those limitations, the present results provide a valuable starting point for future studies to evaluate the effect of topographical difference in the prevalence of surgical conditions.

## Conclusions

Given the wide range of altitudes and the variation in topographic conditions in Nepal, we explored the possible consequences of topography on the prevalence of surgical conditions, such as fractures, hernias and burns. Though not conclusive, the study reveals a correlation between elevation and the prevalence of hernias and fall injuries. Further investigation on the effects of topography and geography on surgical conditions is needed to help direct the allocation of resources for surgical care.

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