Cold-season disasters on the Eurasian steppes: Climate-driven or man-made

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Socio-ecological damage from climate-related disasters has increased worldwide, including a type of cold-season disaster (dzud) that is unique to the Eurasian steppes, notably Mongolia. During 2000–2014, dzuds killed approximately 30 million livestock and impacted the Mongolian socio-economy. The contributions of both natural and social processes to livestock mortality were not previously considered across Mongolia. Here, we consider the contribution of both multiple climate hazards (drought, cold temperatures and snow), and socioeconomic vulnerability (herders’ livestock and coping-capacity) to mortality risk. We performed multi-regression analyses for each province using meteorological, livestock and socioeconomic datasets. Our results show that 93.5% of mortality within Mongolia was caused by a combination of multi-hazards (47.3%) and vulnerability (46.2%), suggesting dzuds were both climate- and man-made. However, in high-mortality hotspots, mortality was primarily caused by multi-hazards (drought-induced pasture deficiency and deep-snow). Livestock overpopulation and a lack of coping capacities that caused inadequate preparedness (e.g., hay/forage) were the main vulnerability factors. Frequent and severe multi-hazards greatly increased the mortality risk, while increased vulnerability caused by socioeconomic changes in Mongolia since the 1990s tended to amplify the effects of multi-hazards. Thus, reductions in herder vulnerability within high-mortality hotspots would likely be an effective means of mitigating the risk of future dzuds.

Climate-related disasters, along with the associated damage to livelihoods and socio-ecological systems, have increased worldwide in recent decades1,2. Such disasters are especially prevalent in countries along the Eurasian steppes, such as Mongolia3,4. Many factors influence this increased risk of climate disaster, as disasters depend not only on climate hazards (hereinafter hazards) but also on the vulnerability of human communities1,5. A rigorous understanding of disaster risk, its causal dimensions and changing trends in those dimensions2, may act as a foundation for reducing future risk.

Many human societies vulnerable to hazards are also under pressure from economic and socio-political change6. Herders are especially vulnerable to hazards as they generally live in marginal lands7, such as grasslands. In Eurasian grasslands, including the Mongolian Plateau, herding and farming constitute 35% of the workforce, and occur in harsh cold and arid environments8,9. Because of the harsh climate, weak livestock often die during cold-season (October–March/April); however, abnormally high mortalities occur during anomalously harsh cold-season (dzuds in Mongolian). Dzuds occur throughout central Asia3,4,7,8, including Mongolia, Kazakhstan, Inner Mongolia and Tibet. In the 2000s, an increase in the frequency and severity of dzuds, together with warming and drying trends, caused massive livestock mortalities in Mongolia11,12,13. This had a large impact on the national socio-economy11. A single dzud in 2010 caused economic losses of 345 million US$14, and 10 million livestock mortalities (23.4% of the total livestock), which was the worst dzud disaster since 194511,14.

Dzuds are defined, biogeophysically, as anomalous climatic and/or land surface conditions (i.e., snow and ice cover) that lead to reduced accessibility and/or availability of pastures. This causes very high livestock mortality (hereinafter mortality) during the cold-season1,3,4. Hazard-oriented studies have shown that years with high mortality result from a combination of growing-season drought and severe weather1,12,13. However, human-induced vulnerability1,3,15–22, including inadequate pasture management, lack of herder experience, poverty, and insufficient

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winter preparedness, has contributed to recent dzuds. This vulnerability is due in part to the transition to democracy and a free market economy in Mongolia since the 1990s\textsuperscript{18,19,22}. Following this, formal pasture management institutions were weakened as herding collectives were dissolved\textsuperscript{17–19}. Additionally, state structures for managing natural disasters (e.g., dzuds) were also weakened. During de-collectivization, state-owned livestock were privatized, state-provided disaster reduction services were stopped, and the burden of risk was transferred to herder-households\textsuperscript{16,20--22}. This directly and indirectly increased herder-households’ vulnerability to climate extremes\textsuperscript{3,10}. Consequently, studies of dzud risk must incorporate both natural and human factors.

This study attempts to fill the gap in the literature, and presents the first, to our knowledge, assessment of livestock mortality risk during the 2000–2014 cold-season (October–March). We do this by examining both climate hazards and herder vulnerability at the Mongolian national level.

Mongolia is located over in mid-latitude highlands in the far eastern continent and has a cold, arid climate\textsuperscript{4,23} (Fig. 1a,b). The latitudinal climate pattern is a key factor driving pasture production\textsuperscript{10}, which is the only source of livestock hay and forage. Favourable conditions exist in the north and northeast, with increasingly dry conditions in the west and south. The region has long and extremely cold winters, with temperatures reaching below $-30.0 \degree C$. The air temperature remains below freezing (below 0 $\degree C$) from November through March. Pastureland is covered with snow during mid-October–April, with higher snow in northern regions, and lower snow in the low-lying south (Fig. 1d). In the fall, herders practice otor (transhumance) to fatten livestock so that they are more likely to survive a harsh winter\textsuperscript{18,24}. Moreover, herders prepare hay and forage from grasslands for livestock to feed on during the cold season when they must confront freezing temperatures (Fig. 1d). These practices are a widely used and critically important traditional strategy for surviving winter hazards\textsuperscript{18}. During winter and spring, most plants decay and pasture vegetation is maintained as dead leaves\textsuperscript{25} providing livestock with fodder.

Dzud-risk (livestock mortality) framework

Here, we consider the risk\textsuperscript{5,16,26} of livestock mortality (Fig. 2) as caused by a combination of multiple climate hazards (multi-hazards) and herders’ vulnerability, both of which may limit the availability and accessibility of pasture for livestock. The multi-hazards include drought and severe winter conditions (i.e., extreme cold and heavy snowfall). The drought declines pasture quantity and quality, preventing livestock from gaining body fat, and also prevents adequate forage preparation for cold-season feeding. Drought thus causes vulnerability for livestock and herders. In the winter, drought-induced short grasses are covered by snow that exceeds plant height, thereby preventing grazing. Extreme cold surges associated with synoptic storms reduce food intake by livestock (intake is determined by the availability of phytomass not covered with snow and ice, and grazing time). For our hazard assessment, we used anomalies of precipitation\textsuperscript{10,12,13} ($P_{e-a}$), temperature\textsuperscript{5,12} ($T_{11-2}$) and maximum snow depth (SD)\textsuperscript{3,13}. We also used vulnerability, as defined as herder-households’ pre-winter socioeconomic ability to cope with hazards (i.e., livestock condition and coping-capacity)\textsuperscript{17}. For the risk analysis, we selected six factors from six groups based on the highest correlations with livestock deaths (Table 1). The livestock factor included animal condition that inferred from pasture carrying capacities (PCC)\textsuperscript{3}. High livestock population sizes ($POP_{pre}$) reduce the
pasture and forage availability for each animal, thereby increasing the likelihood of mortality\(^3,17\). Coping-capacity was determined by preparedness of reserve hay/forage, possession of transportation (number of cars and tractors; trucks), herder experience (fraction of herders 35–59 years old; \(H_{\text{exp}}\)), poverty (households with 101–200 livestock; \(L_{101-200}\)) and the aimag’s gross domestic product (GDP). The factors determining coping-capacity were related to the ability of herders to pursue otor (fall and winter). All these factors may combine in ways that produce very high livestock mortality, often from starvation. We used Poisson multi-regression (PMR)\(^27\) to determine the factors that best explained the mortality as risk indices\(^5\). The dependent variable was number of mortalities, and the explanatory variables included hazards (\(P_{6-8}, SD\) and \(T_{11-2}\)) and vulnerability (\(POP_{\text{pre}}, H_{\text{exp}}, L_{101-200}, \text{trucks}, \text{hay/forage and GDP}\); see Methods).

Results

Figure 3 depicts the interannual and spatial patterns of mortality, and its causal factors in Mongolia. In the 2000s (Fig. 3a), the country experienced frequent mass mortalities (a total of 30.2 million herds died, which equals 80 million sheep units, SU), with most severe mortality occurring in 2000 through 2002 and 2010 winters. The greatest losses occurred in small herds of goat and sheep, which were the most vulnerable because they were the least mobile and least able to find pasture in deep snow\(^22\). Dzuds can be large enough that they impact the Mongolian national economy, with GDP growth rates falling in response to severe dzuds (Fig. 3c). Regions extending from western to south-central Mongolia were high-mortality hotspots (80% of the total mortality), whereas the north-eastern and eastern regions were less affected (Fig. 3b). This regional difference may have been caused by the relative contributions of multi-hazards and vulnerability in each region.

In the 2000s (Fig. 3d), frequent climate hazards such as droughts, and severe winters with below normal cold and above normal snow have increased. Droughts caused reduced pasture biomass\(^10\), tending to decrease the amount and nutrition of pastures and hay/forage. Furthermore, since the 1990s, changes at the national level to the Mongolian socioeconomic system\(^20\) has significantly impacted herders' vulnerability in conjunction with increased larger-scale herders (lacking herding experience) (Fig. 3e). Inexperienced herders tended to have small herds of sheep and goats, thereby changing the overall herd composition in Mongolia (Fig. 3e). These small herds were heavily affected during dzuds. As a result, the \(POP_{\text{pre}}\) rapidly increased, doubling from 25 million heads in 1990 to 52 million in 2014 (Fig. 3e). Winter preparedness was likely also affected by less experienced herders such as declining hay/forage preparation (not shown) that reduced the availability of winter feed for large-scale \(POP_{\text{pre}}\). The lack of winter preparedness therefore likely increased the number of vulnerable herders.

Figure 4 shows cross-validated PMR results for the relative contributions of risk factors of multi-hazards (Fig. 4a) and vulnerability (Fig. 4b,c) to mortality (2000–2014) for each aimag. Results of the leave-one-out cross-validation (LOO-CV) analyses for the full model show value of prediction error measure \(Q^2 > 0.9\). This suggests that the full model is highly predictive. The full model (Fig. 4d) accounted for 93.5% of overall mortality in Mongolia, with 47.3% by multi-hazards and 46.2% by vulnerability, indicating both factors contributed almost equally to mortality.

For the hazards, \(P_{6-8}, SD\) and \(T_{11-2}\) explained 16.7%, 19.4% and 11.1% of variance in causes of deaths (Fig. 4a), respectively. This indicates that higher mortality was caused by drought and heavy snowfall. For vulnerability, \(POP_{\text{pre}}\) (Fig. 4b) and coping-capacity (Fig. 4c) explained 17.6% and 28.7% of mortalities, respectively. For coping-capacity, hay/forage (18.3%) accounted for a relatively larger contribution to deaths than the other factors (3.5% \(H_{\text{exp}}\), 3.3% trucks, 2.1% \(L_{101-200}\) and 1.5% GDP). This suggests that herder households that prepared sufficient hay/forage were more resistant to winter climate hazards. Moreover, the 4% contribution of \(H_{\text{exp}}\) to deaths
likely implies that experienced herders were able to anticipate hazards in herding and pasture management. These results indicate that POP<sub>pre</sub> and prepared hay/forage were the most crucial vulnerability factors. In high mortality hotspots (5.1–8.0% of total livestock deaths; Figs. 3b and 4d), major mortality (average of 51.1%, ranging from 29.0–74.3%) was caused by multi-hazards (26.7% by P<sub>6–8</sub>, 20.7% by SD and 3.4% by T<sub>11−12</sub>). This suggests that those regions had harsher climates (more severe drought- and snow-related livestock mortality; not shown). For vulnerability, contributions of high POP<sub>pre</sub> (26.4%) and low coping-capacity (17.9%, with 10.5% by hay/forage) were found. This indicates that overpopulated livestock and higher drought risk both contributed to a lack of pasture for livestock (overgrazing), causing weakened animals and reduced hay/forage preparation. This in turn left herders more vulnerable. Moreover, the hotspots contained fewer experienced herders and trucks, as well as a lower GDP, likely limiting the ability of herders to practice otor and prepare hay/forage. The mortality hotspots faced severe winter hazards, with insufficient winter feed for weakened livestock because of previous overgrazing.

Conversely, low mortality regions (<5.1% of total livestock deaths; Figs 3b and 4d) generally coincided with relatively low vulnerability and higher (average of 44.4%, ranging from 25.0–63.9%) contributions of the low vulnerability factors, resulting in reduced mortality. Of the multi-hazards, winter climate hazards (17.3% from SD and 16.1% from T<sub>11−12</sub>) outweighed drought (10.4% from P<sub>6–8</sub>). For vulnerability, coping-capacity (35.3%, with 22.9% from hay/forage) was a large contributor to low mortality, while POP<sub>pre</sub> had a minor influence (13.0%). Furthermore, these regions had higher proportions of H<sub>exp</sub>, trucks and GDP, likely facilitating the ability of herders to prepare hay/forage and practice otor. Low mortality regions thus had sufficient coping-capacity and PCC, as well as lower drought risks.

**Summary and Discussion**

From 2000 to 2014, dzuds (including century-worst dzud events) killed approximately 30 million livestock, and greatly impacted the Mongolian socio-economy. To understand the natural and man-made causes of these dzuds, this study presents the first, to our knowledge, assessment of contributions of both climate hazards and socioeconomic vulnerability to livestock mortality risk. We performed PRM analyses for each province using meteorological, livestock and socioeconomic datasets. Our results show that cross-validated PMR performed reasonably well in explaining the variation in mortality ($R^2 = 93.5\%$), with sufficient predictive ability ($Q^2 > 0.9$). We found variations in mortality were caused by a combination of multiple climate hazards and herder vulnerability. Climate hazards, particularly droughts, have increased in Mongolia since 2000<sup>30</sup>, with droughts being unusual during the previous millennium<sup>38</sup>. In addition, recent anomalously severe winters in Mongolia have been linked to declines of Arctic sea-ice that in turn triggered changing atmospheric circulation, causing extreme winters in mid-latitude

Table 1. List of the selected factors of livestock mortality risk, including both climate hazards and herders’ vulnerability variables, and their averaged linear (Pearson) correlations ($r$) with livestock deaths during the 2000–2014 cold-season. Factors in bold are the indicators selected for risk analysis. *Prepared hay/forage includes hay, imperfectly ripened wheat, artificial (concentrated) feed, salt/mineral salt and crop residuals. These are summed with coefficients as follows: 0.45 for hay, 0.35 for planted feed, 0.25 for silage feed, 1.0 for artificial feed and mineral salt, 0.22 for potatoes and vegetable scraps, 0.9 for imperfectly ripened wheat, 0.4 for residuals and 0.25 for straw. Numbers in bold are the indicators selected for risk analysis.

| Main risk factors | Groups | Indicators | $r$ |
|------------------|--------|------------|-----|
| Climate hazards  | Preceding summer drought | P<sub>6–8</sub> | −0.24 |
|                   | Severity of winter | T<sub>11−12</sub> | −0.35 |
|                   |                     | SD | 0.41 |
| Livestock conditions | Pasture capacity | POP<sub>pre</sub> Previous year livestock number in sheep unit | 0.29 |
|                   | Energy conditions | LOSS<sub>pre</sub> Previous year livestock mortality in sheep unit | 0.04 |
| Herders’ socioeconomic vulnerability | Herder experience | H<sub>exp</sub> Fraction of herders’ population aged 16–34 | 0.10 |
|                   |                     | H<sub>pop</sub> Fraction of herder population aged 35–59 (males) and 35–54 (females) | −0.17 |
|                   | Poverty (well-being) | L<sub>11–200</sub> Fraction of herders who have 201–500 livestock | 0.16 |
|                   |                     | L<sub>201–500</sub> Fraction of herders who have 201–500 livestock | −0.05 |
|                   |                     | L<sub>501–999</sub> Fraction of herders who have 501–999 livestock | −0.03 |
|                   |                     | L≥1000 Fraction of herders who have ≥1000 livestock | −0.02 |
| Facilities         | Trucks Number of cars and tractors per herder household | −0.17 |
|                   | TV Fraction of herders who have TVs | −0.11 |
| Winter preparedness | Hay/forage* Prepared hay/holder per livestock (sheep unit) | −0.46 |
|                   | Shelter** Number of warm barns and roofed barns per herder household in winter camp | −0.24 |
| Economy            | GDP Real GDP per capita (aimag) (10$^3$ Mongolian Tugrug) | −0.22 |
Eurasia. These increasing multi-hazards led to pasture deficiency and a continuing loss of adequate grazing, causing significant mortality in cold-season.

Additionally, greater herder vulnerability caused by economic and socio-political changes (including weakened state-based disaster risk management) since the 1990s amplified livestock mortality levels. Livestock over-population and the lack of herder coping-capacity, along with inadequate hay/forage preparedness, were the main factors causing this vulnerability. Over the past two decades, traditional herding strategies were affected by increasingly inexperienced herder-households, as well as herd compositions changed to smaller animals and increased livestock population sizes. Moreover, many herders have reduced or stopped practicing otor to avoid labour and transportation costs. This herders’ vulnerability, combined with declining pasture production brought on by drought, has resulted in overgrazing, which was the primary reason for the record mortality in...
the 2009/2010 dzud. A lack of data availability prevented us from analysing other aspects of vulnerability, such as herder behaviour (e.g., otor) and herding environment (e.g., access to water).

High mortality hotspots in western and south-central Mongolia were caused by livestock overpopulation and large herder vulnerability. The hotspots had certain geographical disadvantages of being located in arid and mountains regions. The southern region was arid with poor pastures, while high mountain regions in the western region were prone to heavy snow, hindering the ability of the state to deliver emergency forage. Another disadvantage of these regions was their remoteness, being located far from the capital, Ulaanbaatar, where market opportunities as well as disaster relief were available. Conversely, regions with low mortality had sufficient PCC because they were located in relatively wet plains with favourable pasture, as well as having herders with low

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**Figure 4.** Contributions (%) of (a) climate hazards, (b,c) herders’ vulnerability (livestock population $POP_{pre}$ and coping-capacity), and (d) their combination to livestock mortality in 2000–2014 cold-season by cumulative variance from each model (background colours) for each aimag by PMR. Relative contributions of each factor (increment variance) to mortality showed in colour pies: (a) hazards of drought ($P_{6-8}$), cold-temperature ($T_{11-2}$) and snow ($SD$); (b,c) vulnerability factors of $POP_{pre}$ and coping-capacities of hay/forage preparedness, herder experience ($H_{exp}$), poverty ($L_{101-200}$), transportation facilities (trucks) and economic conditions (GDP); and (d) all factors. In Fig. 4d, patterns of averaged mortality (percentage; sized pies) with relative contributions of hazards and vulnerability ($POP_{pre}$ and coping-capacity) to mortality.
vulnerability. Additionally, they had the advantage of being located near major cities, ensuring effective market access, and providing better infrastructure in the form of roads for delivering emergency forage.

Herding husbandry, particularly in vulnerable hotspots, likely will continue to face increasing future climate risks (such as droughts and severe winter). As a consequence, alleviating the impacts of climate change on herder communities, through strengthening adaptive capacities, risk reduction strategies (including reducing herder vulnerability to future hazards) and resilience in degraded environments, will be a crucial challenge. Moreover, it is crucial to develop an effective early warning system (EWS) for climate hazards by improving weather forecasts and considering all factors influencing vulnerability. The dzud risk model presented in this study should be incorporated into a future dzud-EWS to include the quantitative contributions of natural and socioeconomic factors in the warning system. This approach should be extended to other Eurasian steppe regions that share the same climate change and herding challenges as Mongolia.

**Material and Methods**

**Datasets.** Data for livestock population sizes (POP) and mortality from hazards at the aimag (the administrative unit of Mongolia; there are 21 aimags) level from 1971 to 2014 were obtained from the National Statistical Office (NSO) of Mongolia (NSO, 2000–2014). The NSO conducts annual surveys in December of the total number of horses, cattle, sheep, goats and camels in Mongolia. These data were converted into sheep units (SU) to standardize feed requirements as each type of livestock requires different amounts of feed. Conversion rates to SU were: 1 camel = 5 SU, 1 horse = 7 SU, 1 cow = 6 SU, 1 sheep = 1 SU and 1 goat = 0.9 SU. The percent relative mortality for each aimag was calculated as the ratio of the total number of livestock deaths during winter and spring to the total number of livestock at the beginning of the year.

To assess climate hazards, we used monthly average air temperature (T), precipitation (P) and maximum snow depth (SD) from 69 stations in Mongolia from 1971 to 2014 (Fig. 1b). These data were obtained from the Information and Research Institute of Meteorology, Hydrology and Environment of Mongolia. For pasture vegetation, we used a monthly 0.5° × 0.5° grid normalized difference vegetation index for August (NDVI8) from the semi-monthly 8-km resolution Global Inventory Modelling and Mapping Studies (GIMMS) for the period 1982–2006. We also used the 16-day 0.05° × 0.05° spatial resolution Moderate Resolution Imaging Spectroradiometer (MODIS) data from 2007–2014 (Fig. 1c). Droughts were evaluated based on the plant-available precipitation anomaly percentage during the critical growing months of June through August (P6–8). The severity of winter was assessed based on anomalies of T for November–February (T11–2) and maximum snow depth (SD) in January, which showed the strongest significant relationship with losses. Anomalies were based on the period of 1981–2010.

For vulnerability, we considered socioeconomic data related to pre-winter herder households available at the aimag level from 1999–2013. Datasets were compiled using the livestock census conducted in December by the NSO. The gross domestic product (GDP) for Mongolia (per capita in current USD and value added in the agricultural sector as percentage of GDP) as an indicator of economic conditions were taken from the World Bank during 1981–2014. We selected 14 indicators encompassing six groups (one group for conditions of livestock and five groups for coping-capacity; Table 1). The preceding livestock mortality (LOSSpre) and population size (POPpre) were selected as livestock factors. We measured household possession of transportation as the number of cars and tractors (trucks), possession of tools to acquire information on weather and pasture through televisions, economic conditions by aimag's GDP, herders' experience by fraction of aged herders (Hexp) and well-being or poverty conditions by proportion of households who have 101–200 livestock (L101–200). The number of livestock owned is a common measure of wealth/poverty in rural areas. For the risk assessment, we selected six factors based on the highest correlation with losses (Table 1): POPpre, Hexp, L101–200, trucks, hay/forage and GDP.

**Study area.** Most areas of Mongolia located in more than 1500 m above sea level. Elevation ranges between 800 and 4000 m, with marked differences in elevation from west to east and north to south. Climate and topography organize the terrain into pronounced eastern, central, western and southern regions. Eastern Mongolia is composed of flat or undulating plains. The central to north-western regions are made up of the Altai, Khangai, Khentii, and Khuvsgul mountain ranges (Fig. 1b), interspersed with depressions or basins. The southern region encompasses the Gobi Desert. The annual precipitation varies from over 300 mm in the north to below 100 mm in the south and is concentrated in June–August (Fig. 1c). The pasture growing season is short (May–August) and peaks at the end of August (normalized difference vegetation index, NDVI8; Fig. 1c). Pastureland is covered with snow from mid-October to the end of April, and the yearly maximum snow depth is observed in January, ranging from over 100 mm in the northern mountains to below 10 mm in the south (Fig. 1d).

**Poisson regression model and cross-validation.** We used Poisson multi-regression (PMR) to determine the factors that best explained the livestock deaths as risk indices. Death counts are typically distributed through a Poisson process, generating independent and random occurrences across time or space. If time were divided into discrete periods, death counts would be theoretically distributed as a Poisson distribution. The probability distribution of the number of occurrences, Y, of some random event in an interval of time or space is (Eq. 1):

\[
\Pr(Y = y_i) = \frac{e^{-\mu_i} \mu_i^{y_i}}{y_i!} \quad (y_i = 0, 1, 2, 3 \ldots)
\]

where \(y_i\) is the number of deaths for a particular time \(i\), and \(\mu_i\) is the expected number of deaths per period. The mean and the variance of the distribution are both \(\mu_i\). The mean and variance of this distribution is \(y_i \sim \Pr(\mu_i)\).

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Poisson multiple regression (PMR) is a basic method for analysing the relationship between observed mortality and a set of explanatory variables. In this study, the dependent variable was total livestock deaths of five species of livestock during 2000–2014 cold-season, which were converted to SU as standardize feed requirements. The factors selected as explanatory variables were multi-hazards ($P_{6–8}$, SD and $T_{11–2}$) and vulnerability (POPpre, $H_{exp}$, hay/forage, $L_{101–200}$, trucks and GDP). In this study, we assumed that observed livestock deaths occurred over a fixed time interval, and because the counts were nonnegative, PMR is defined in terms of log of expected deaths (Eq. 2):

$$\mu_i = \exp(\beta_0 + \beta_1 x_1 + \beta_2 x_2 + \ldots + \beta_k x_k)$$

where $x$ is the explanatory variable and $\beta$ is the regression coefficient. Thus, the typical PMR expresses the log outcome rate as a log linear function of a set of predictors. Initially, the PMR included only $P_{6–8}$ each hazard of SD and $T_{11–2}$. We then added vulnerability as POPpre and finally added all variables, including each coping-capacity factor. Separate models were developed for each aimag, because the contributions of each factor to deaths vary from region to region. The contribution of a given predictor is measured by the increment of variance ($R^2$; McFadden's Pseudo-$R^2$) based on log-likelihoods that are calculated from adding each predictor to the model. $R^2$ closely parallels $R^2$ in linear regressions, both conceptually and mathematically. The measure may be thought of as intuitively as a proportional reduction in error measure. To measure the relative importance of predictors, we performed a dominance analysis, which is an examination of the $R^2$ values for all possible subset models. This is performed by calculating each predictor’s added predictive ability in the presence of other predictors.

To evaluate the full PRM model’s predictive ability, the leave-one-out cross-validation (LOO-CV) analysis was applied for each aimag. The LOO-CV is a commonly used error estimation technique, and it consists of excluding each sample once, constructing a new model without this sample, and predicting the value of its dependent variable, $y$. In the LOO-CV, each observation in the sample dataset of size $n$ ($n = 15$ years in this study) is successively taken out and the remaining $n-1$ observations of the set are used to train the prediction model to estimate the livestock mortality. This process is done $n$ times. The predictive ability of the model is assessed by prediction error measure, $Q^2$ (Eq. 3):

$$Q^2 = 1 - \frac{\sum_{i=1}^{n}(\hat{y}_i - y_i)^2}{\sum_{i=1}^{n}(y_i - \bar{y})^2}$$

where $y_i$ and $\hat{y}_i$ are the observed and predicted livestock mortalities for an individual compound in the training set. The value $\bar{y}$ is the mean value of $y$ for all samples. The $Q^2$ statistic provides cross-validation, as well as a qualitative measure of consistency between the predicted and original data. An acceptable value for $Q^2$ in biological models is $\geq 0.4$.

References

1. IPCC A Special Report of Working Groups I and II: Managing the Risks of Extreme Events and Disasters to Advance Climate Change Adaptation (eds Field, C. B. et al.) (Cambridge Univ. Press, 2012).
2. CRED (Centre for Research on the Epidemiology of Disasters) and UNISDR (United Nations Office for Disaster Risk Reduction): The Human Cost of Weather-Related Disasters 1995–2015 (Brussels, BELGIUM, 2015).
3. Nandintsetseg, B., Shinoda, M. & Erdenetsetseg, B. Contributions of multiple climate hazards and overgrazing to the 2009/2010 winter disaster in Mongolia. Nat Hazards. https://doi.org/10.1007/s11069-014-1795-9 (2015).
4. Shinoda, M. Evolving a multi-hazard focused approach for arid Eurasia (ed. Sternberg, T.) Climate hazard crises in Asian societies and environments (Routledge, Oxon, 2017).
5. Birkmann, J. Measuring vulnerability to promote disaster-resilient societies: Conceptual frameworks and definitions. (ed. Birkmann, J.) Measuring vulnerability to natural hazards: towards disaster resilient societies (United Nations Univ. Press, 2006).
6. Sneath, D. State policy and pasture degradation in Inner Asia. World Develop. 27, 141–1425, https://doi.org/10.1016/j.worlddev.2008.06.015 (2008).
7. Wang, W. et al. Early warning of snow-caused disasters in pastoral areas on the Tibetan Plateau. Nat. Hazards Earth Syst. Sci. 13, 1411–1425, https://doi.org/10.5194/nhess-13-1411-2013 (2013).
8. Nandintsetseg, B., Greene, J. S. & Goulden, C. E. Trends in extreme daily precipitation and temperature near Lake Hovsgol, Mongolia. Int. J. Climatol. 72, 341–347, https://doi.org/10.1002/joc.1404 (2007).
9. Nandintsetseg, B. & Shinoda, M. Assessment of drought frequency, duration, and severity and its impact on pasture production in Mongolia. Nat. Hazards 66, 993–1008, https://doi.org/10.1007/s11069-012-0537-4 (2013).
10. UNDP & NEMA. Dzud national report 2009–2010, Project ID: 90074253 (UNDP-NEMA, Ulaanbaatar 2010).
11. Rao, M. P. et al. Dzuds, droughts, and livestock mortality in Mongolia. Environ. Res. Let. 10(7), 074012, https://doi.org/10.7916/DQ8Q1D75 (2015).
12. Tachiiri, K., Shinoda, M., Klinkenberg, B. & Morinaga, Y. Assessing Mongolian snow disaster risk using livestock and satellite data. J. Arid Environ. 72, 2251–2263, https://doi.org/10.1016/j.jaridenv.2008.06.015 (2008).
13. Erdenetsetseg, B., Doljinsuren, M. & Nandintsetseg, B. Drought and 2009–2010 Dzud. (eds. Shinoda, M. & Nandintsetseg, B.) Scientific report on climate change and hazards in Mongolia (Nagoya Univ. Nagoya, 2015).
14. Joly, F., Sabatier, R. & Hubert, B. Modelling interacting plant and livestock renewal dynamics helps disentangle equilibrium and nonequilibrium aspects in a Mongolian pastoral system. Sci. Total Environ. 625, 1390–1404, https://doi.org/10.1016/j.scitotenv.2017.12.215 (2017).
15. Murray, V. et al. Case studies. In: Managing the Risks of Extreme Events and Disasters to Advance Climate Change Adaptation (eds Field, C. B. et al.) A Special Report of Working Groups I and II of IPCC (Cambridge Univ. Press, Cambridge, UK, & New York, NY, USA) 487–542 (2012).
16. Du, C. et al. Mongolian herders’ vulnerability to dzud: A study of record livestock mortality levels during the severe 2009/2010 winter. Nat. Hazards. https://doi.org/10.1007/s11069-017-2893-4 (2017).
17. Fernandez-Gimenez, M. E., Batkhishig, B., Babuyan, B. & Ulambayar, T. Lessons from the dzud: Community-based rangeland management increases adaptive capacity of Mongolian herders to winter disasters. World Develop. 68, 48–65, https://doi.org/10.1016/j.worlddev.2014.11.015 (2015).
19. Johnson, D. A., Sheehy, D. P., Miller, D. & Damiran, D. Mongolian rangelands in transition. *Sci. et Changements Planétaires/ Sécheresse* **17**, 133–41 (2006).
20. Mearns, R. Sustaining livelihoods on Mongolia’s pastoral commons: insights from a participatory poverty assessment. *Dev. Change* **35**(1), 107–139, https://doi.org/10.1111/j.1467-7660.2004.00345.x (2004).
21. Sternberg, T. Unravelling Mongolia’s extreme winter disaster of 2010. *Nomadic Peoples* **14**, 72–86, https://www.jstor.org/stable/43123863 (2010)
22. Templer, G., Swift, J. & Payne, P. The changing significance of risk in the Mongolian pastoral economy. *Nomadic Peoples* **33**, 105–122, https://www.jstor.org/stable/43124055 (1993).
23. Shinoda, B. & Nandintsetseg, B. Soil moisture and vegetation memories in a cold, arid climate. *Glob. Planet. Change* **79**, 110–117, https://doi.org/10.1016/j.gloplacha.2011.08.005 (2011).
24. Ilkhagvadorj, D., Hauck, M., Dulamsuren, C. & Tsoigbaatar, J. Pastoral nomadism in the forest-steppe of the Mongolian Altai under a changing economy and a warming climate. *J. Arid. Environ.* **88**, 82–89, https://doi.org/10.1016/j.jaridenv.2012.07.019 (2013).
25. Nandintsetseg, B. & Shinoda, M. Land surface memory effects on dust emission in a Mongolian temperate grassland. *J. Geophys. Res. Biogeosci.* **120**, 414–427, https://doi.org/10.1002/2014BG002708 (2015).
26. Wood, N. Understanding risk and resilience to natural hazards. *U.S. Geol. Survey Fact Sheet* **2011–3008**, https://pubs.usgs.gov/fs/2011/3008/ (2011).
27. Rhodes, T. E. & Freitas, S. A. *Advanced Statistical Analysis of Mortality* (Boston, USA) (2004).
28. Hessl, A. E. *Veterinary Epidemiologic Research* (2nd ed.), (VER Inc., Charlottetown, Canada, 2009), 445–466 (2002–2014).
29. Iijima, Y. & Hori, M. E. Cold air formation and advection over Eurasia during “Dzud” cold disaster winters in Mongolia. *Nat. Hazards* https://doi.org/10.1007/s11069-017-2893-4 (2016).
30. Liu, J. P. et al. Impact of declining Arctic sea ice on winter snowfall. *Proc. Natl. Acad. Sci. USA* **109**, 4074–4079, https://doi.org/10.1073/pnas.1114910109 (2012).
31. Arzel, O., Fichefet, T. & Goosse, H. Sea ice evolution over the 20th and 21st centuries as simulated by current AOGCMs. *Ocean Model* **12**, 401–415, https://doi.org/10.1016/j.ocemod.2005.08.002 (2006).
32. NSO (National Statistical Office). *Mongolian statistical yearbook* 1999–2013 (Ulaanbaatar, in Mongolian) (2000–2014).
33. NSO (National Statistical Office). *Monthly bulletin statistics* 2000–2014 (Ulaanbaatar, in Mongolian) (2000–2014).
34. NSO (National Statistical Office). *Agricultural sector* 2001–2013 (Ulaanbaatar, in Mongolian) (2001–2014).
35. World bank, World Bank national accounts data, and OECD National Accounts data files, https://data.worldbank.org/indicator/NV.AGR.TOTL.ZS?locations=MN&view=chart (2018).
36. Tucker, C. J. et al. An extended AVHRR 8-km NDVI data set compatible with MODIS and SPOT vegetation NDVI data. *Int. J. Remote Sens.* **26**(20), 4483–4498, https://doi.org/10.1080/01431160500168686 (2005).
37. Dohoo, W. & Martin, H. S. *Modelling count and rate data*. Veterinary Epidemiologic Research (2nd ed.), (VER Inc., Charlottetown, Canada, 2009), 445–466 (2002–2014).
38. Azen, R. & Traxel, N. Using Dominance Analysis to Determine Predictor Importance in Logistic Regression. *J. Educ. Behav. Stat.* **34**(3), 319–347, https://www.jstor.org/stable/40263507 (2009).
39. Johan, A. W. et al. Assessment of PLSDA cross validation. *Metabolomics* **4**, 81–89, https://doi.org/10.1007/s11306-007-0099-6 (2008).

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**Author Contributions**

B.N. and M.S. designed the study. B.N. analysed the data and wrote the manuscript and prepared all Figures with comments and edits from M.S. Ch.D. and E.M. prepared meteorological and socioeconomic datasets.

**Additional Information**

**Competing Interests:** The authors declare no competing interests.

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