Mapping the Global Emergence of *Batrachochytrium dendrobatidis*, the Amphiphan Chytrid Fungus

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Abstract

The rapid worldwide emergence of the amphibian pathogen *Batrachochytrium dendrobatidis* (Bd) is having a profound negative impact on biodiversity. However, global research efforts are fragmented and an overarching synthesis of global infection data is lacking. Here, we provide results from a community tool for the compilation of worldwide Bd presence and report on the analyses of data collated over a four-year period. Using this online database, we analysed: 1) spatial and taxonomic patterns of infection, including amphibian families that appear over- and under-infected; 2) relationships between Bd occurrence and declining amphibian species, including associations among Bd occurrence, species richness, and enigmatic population declines; and 3) patterns of environmental correlates with Bd, including climate metrics for all species combined and three families (Hylidae, Bufonidae, Ranidae) separately, at both a global scale and regional (U.S.A.) scale. These associations provide new insights for downscaled hypothesis testing. The pathogen has been detected in 52 of 82 countries in which sampling was reported, and it has been detected in 516 of 1240 (42%) amphibian species. We show that detected Bd infections are related to amphibian biodiversity and locations experiencing rapid enigmatic declines, supporting the hypothesis that greater complexity of amphibian communities increases the likelihood of emergence of infection and transmission of Bd. Using a global model including all sampled species, the odds of Bd detection decreased with increasing temperature range at a site. Further consideration of temperature range, rather than maximum or minimum temperatures, may provide new insights into Bd-host ecology. Whereas caution is necessary when interpreting such a broad global dataset, the use of our pathogen database is helping to inform studies of the epidemiology of Bd, as well as enabling regional, national, and international prioritization of conservation efforts. We provide recommendations for adaptive management to enhance the database utility and relevance.

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Introduction

The specter of multi-host infectious diseases leading to the rapid devastation of entire communities is alarming but, so far, rare. However, a handful of highly virulent multi-host pathogens are known, and these have had profound impacts across populations and species (e.g., the rinderpest Morbillivirus; West Nile Virus; bat *Georgiella destructans*). Rapid aggregation, synthesis, and analysis of disease data are needed to advance both science and management of such diseases. While technological advances have aided our ability to respond to emerging disease, such as by the use of proactive modeling [1], our ability to rapidly detect, assess, and report globally-emerging pathogens has lagged for wild animal diseases that are not zoonotic or are not transmitted to livestock or companion animals. For these diseases, infection and mortality are often cryptic and research studies may take years to be completed before being published.

Herein we report on a community surveillance effort addressing the emergence of the amphibian chytrid fungus *Batrachochytrium dendrobatidis* (Bd). Approximately one-third of global amphibian species have imported status, and the emergence of Bd is known to be a proximate driver of amphibian species declines and extinctions [2–8]. Bd has an unusually wide host-range: it has been detected infecting hundreds of species. Cascading effects owing to declines of multiple host species as a result of chytridiomycosis could potentially undermine ecosystem stability and function [9,10], although some studies have not supported key functional roles of amphibians in local ecosystems [11]. Hence, although caution is necessary because amphibians’ ecological roles will vary with location, as a worst case scenario, the potential of Bd...
to act synergistically with other anthropogenic drivers may catastrophically disturb biological communities and substantially contribute to the ongoing 6th mass extinction event [8].

Although infectious diseases are a normal component of wildlife ecology, only recently arc diseases being widely recognized as critical conservation concerns, notably for amphibians [12]. Bd was described as a species in the late 1990s [13], and our basic biological knowledge of this novel pathogen is still accruing. After over a decade of research, many aspects of its ecology, epidemiology, and pathogenicity remain uncertain. For example, vectors and pathways of transmission of Bd across spatial scales are little understood, as are the factors affecting species-specific susceptibility to the clinical disease, chytridiomycosis [14,15], although reservoir amphibian hosts have been identified, including some involved in commercial trade of amphibians [16-20]. Bd genotype affects Bd virulence, and population genomics studies have shown that the worldwide emergence of chytridiomycosis has occurred as a consequence of 20th-century emergence of an aggressive lineage [16]. Concern has heightened about Bd transmission in various arms of the commercial trade in amphibians which annually involves millions of individuals [15,21]. Transmission pathways may include migratory waterbirds [22], and some regional studies have reported an association of Bd occurrence with amphibian species’ proximity to human development, suggesting uncharacterised pathways of introduction and transmission [23]. Bd has been detected in water samples [24], supporting transmission via fomites. Due to the recent recognition of Bd as an emerging infectious disease and an invasive species, Bd was listed as a notifiable disease by the World Organization of Animal Health (OIE) in 2009, resulting in international recommendations to forestall further spread via anthropogenic activities [25].

The need for effective surveillance, including rapid data accrual and interpretation in order to identify emerging patterns and processes, is integral to the progression of both Bd science and management objectives. In particular, our incomplete understanding of the global distribution of Bd is key information that is required to assess the importance of various mechanisms that may have contributed to the rapid emergence of chytridiomycosis [26] or continue to contribute to the ongoing spread of Bd.

We report on the outputs of a multi-phase project for compiling worldwide Bd data and maintaining an updated system for global-scale assessment of the pathogen, the Global Bd Mapping Project [27]. A web-based system (www.Bd-maps.net) for collation of Bd incidence and associated metadata was produced [28], with the aims of providing new insights into both Bd occurrence patterns and the development of hypotheses for the study of Bd ecology and epidemiology, and to inform species- and land-management planning efforts. Using data gleaned from the published literature or submitted as primary data to Bd-maps.net by members of the Bd Mapping Group over a four year period, we report: 1) spatial and taxonomic patterns of infection, including amphibian families that appear over- and under-infected; 2) relationships between Bd occurrence and declining amphibian species, including an analysis of Bd occurrence, species richness, and occurrence of enigmatic population declines; and 3) patterns of environmental correlates, including climate metrics for all species and for three families (Hylidae, Bufonidae, Ranidae), at both the global scale and the U.S.A. scale (which has higher spatial resolution of climate metrics allowing downscaled analyses). Our analyses contribute new insights to hypotheses of geographic and climatic relationships with Bd occurrence [29-37].

Results

Bd-maps.net

Bd-maps.net contains a web-accessible database for data storage, facilities for data uploading, basic summary statistics, and visualisation of spatial data utilising Google Maps and Google Earth. Data are submitted in a number of ways, through: 1) literature review with subsequent geo-coding and direct database submission; 2) direct online data entry by community users; and 3) a smartphone interface for the direct submission of location and metadata using the data gathering tool, EpiCollect [38]. Summaries of Bd data can be viewed globally or by country, and all data are now available for community downloading for further offline analyses such as those presented here (see Materials and Methods).

Cautions expressed in Materials and Methods for use of Bd-maps.net data should be reiterated here; in particular, results of analyses should be used to inform predictions for further examination, and are not conclusive due to several forms of potential data bias. Importantly, distribution maps derived from the Bd-maps database do not represent the distribution of chytridiomycosis, but locations where Bd sampling has occurred, where the fungus has been detected, or where it was not detected. These locations have a suite of potential biases as discussed in Materials and Methods, and we have addressed these biases to some extent by the data aggregations that we have conducted. Type I errors indicate false Bd-positive findings, and may occur in the dataset if Bd were detected at a site but the assay to determine this was incorrectly run, or was subject to cross-contamination. A Type II error is a false Bd-negative finding. This might occur if Bd occurs at a site, but was not detected. A low-level Bd infection, an insufficient sample size, or the presence of PCR inhibitors, could result in a false negative report [39]. We did not conduct quality assurance procedures to examine data for the use of acceptable swabbing, histology or PCR techniques, or for sample sizes. These types of errors have been increasingly recognized over the last decade, and are likely reduced for surveillance efforts reported in peer-reviewed publications. Nevertheless, in our analyses, by aggregating all data at a site level and by conducting analyses at the 0.5 degree latitude and longitude grid cell size, we may have reduced the chance of both false-positive and false-negative findings by the compilation of results among different studies (and their different sampling and analytical approaches), researchers, species, seasons, or years. An additional consideration is the sample sizes used in our various analyses and summary statistics; the larger sample may represent more robust datasets with less bias. Nevertheless, we offer our results as hypotheses for further testing, and temper our conclusions accordingly.

Spatial and taxonomic patterns of Bd detection

Bd has been detected in 36 of 82 (43%) countries (Figure 1), and in 516 of 1,240 (42%) species, determined using a data set of more than 36,000 individuals (Bd-maps.net data snapshot August 2012). Bd was detected at 1,814 of 4,281 (43%) sites tallied for our analyses (data snapshot July 2010). Broad-scale distribution patterns are evident, with Bd widely detected in the Americas, and detected only patchily in Africa, Asia, and Europe. Spatial patterns and gaps in sampling across the globe contribute to the clusters of Bd locations on the world map. In particular, Asia had been largely unsampled at the time of our data analyses; however, recently, Swei et al. [26] sampled in 15 Asian countries and found only 2.35% Bd prevalence.

Bd was detected in 41 of 50 (82%) sampled families of amphibians (Table 1; data snapshot January 2011). Anurans and caudates were comparatively sampled: 955 of 6,177 (15%) extant
Figure 1. Global distribution of the amphibian chytrid fungus, Batrachochytrium dendrobatidis (Bd). Maps downloaded from www.Bd-maps.net (15 August 2012); Bd-positive (red) and Bd-negative (white, blue) sites are shown.
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and recognized anuran species were included in our study, while 114 of 629 (18%) caecilians were also included (Table 1). Caecilians remained relatively unsampled, with only 6 of 190 (3%) species registered in the database (Table 1, Table 2) (total species counts from AmphibiaWeb.org, accessed 25 July 2012). Three speciose anuran families, Hylidae, Ranidae, and Bufonidae, accounted for 30% of the species in which Bd was detected (Table 2, Table S1).

We examined which families deviated from the expected level of infection given the overall background prevalence in our data set (508/1,055; 48%), and found broad taxonomic patterns (Table 2). Microhylidae, Anostomidae, and Mantellidae were under-
Table 1. Numbers of amphibian species and families with Batrachochytrium dendrobatidis detections as of January 2011.

| Order       | Species | B. d. Detected | B. d. Not Detected | Total | Percent Detected |
|-------------|---------|----------------|--------------------|-------|------------------|
| Anura       | 449     | 486            | 935                | 48.0  | 35               |
| Caudata     | 59      | 55             | 114                | 51.8  | 6                |
| Gymnophiona | 0       | 6              | 6                  | 0     | 0                |
| Total       | 508     | 547            | 1055               | 48.2  | 41               |

For species in which Bd infection was detected, 106 of 449 (24%) anurans and 12 of 60 (20%) caudates were listed in the three Threatened categories of the IUCN Red List (Critically Endangered, Endangered, Vulnerable; Table S1). It is important to note that these are counts of species that are recognized as being threatened without reference to what threats, according to the IUCN Threats Classification Scheme, are responsible for their classification. Animals under threat from Bd would be classified as 8.1 ‘Invasive non-native/ alien species/ diseases’, 8.2 ‘Problematic native species/ diseases’ or 8.4 ‘Problematic species/ diseases of unknown origin’. Although Bd infection does not equate to being a threat in every instance, upon inspection of the Red List, Bd may be under-reported as a concern, given the frequency of its occurrence as we report it here. For example, of the 12 caudates known to be infected with Bd that are currently ranked as threatened, Bd infection was listed as a threat for only one species, and disease, generally, was listed for a second species. To assess the relative impact of threats globally, an appropriate analysis would involve comparing the number of threatened species that are Red Listed because of a given process, versus the number of species that are exposed to that process. Currently, we are unable make such a comparison in the context of Bd because the only data we have are the infection status of a relatively small subset of the total number of amphibian species (1,240/6,996 species, 18%; species count from AmphibiaWeb.org, accessed 25 July 2012) and we lack conclusive data on symptoms of the disease chytridiomycosis for most of these species.

Bd and environmental correlates: World and U.S.A. scales

Our full model of odds of Bd occurrence by environmental parameters at sites using all world data (data snapshot from July 2010) included biomes, latitude, occurrence of enigmatic declines, temperature range, and average annual precipitation as significant explanatory variables (Table S2). Removing latitude from our analysis, which we considered warranted due to the overwhelming focus of Bd-surveillance in the northern hemisphere (Figure S1, Figure S2), resulted in only minor changes in the other coefficients, and slight changes to the order of biome entry into the model (Table S3). N = 3,733 sites used in analysis. After accounting for differences in odds of Bd detection between biomes, we found that the odds of Bd detection at discrete sites; 1) increased by 14% with each known increase in occurrence of enigmatic declines at a site; 2) decreased by 8.8% with each degree increase in temperature range at a site; and 3) decreased by 0.05% with each mm increase in average annual precipitation at a site (Table 3, Figure 3, Figure 4). Among biomes, sites in Montana Grasslands and Shrubslands (N = 56 sites) had the highest odds of Bd occurrence. To better understand the inverse relationship of Bd presence with world temperature range values at sites, it is important to
Table 2. Numbers of species with *Batrachochytrium dendrobatidis* (*Bd*) detections by amphibian family, and results of randomisation tests (see Materials and Methods section) to determine whether each sampled amphibian family was over- or under-infected, compared to what we would expect by chance given an overall prevalence of 508/1055 species.

| Family            | Species | Bd Detected | Bd Not Detected | Total Sampled | Prev. | p-value | Prev. pattern |
|-------------------|---------|-------------|-----------------|---------------|-------|---------|---------------|
| **Amphibia**       |         |             |                 |               |       |         |               |
| Anura             |         |             |                 |               |       |         |               |
| Heleophrynidae    | 3       | 0           | 3               | 1             | 0.9652| NA      |               |
| Pelobatidae       | 2       | 0           | 2               | 1             | 0.9278| NA      |               |
| Calyptocephalidae | 1       | 0           | 1               | 1             | 0.8494| NA      |               |
| Ceratophryidae    | 11      | 1           | 12              | 0.92          | 0.9998*| Over    |               |
| Hylidae           | 8       | 1           | 9               | 0.89          | 0.9936*| Over    |               |
| Hemiphractidae    | 7       | 2           | 9               | 0.78          | 0.9641| NSD     |               |
| Leiperidae        | 6       | 2           | 8               | 0.75          | 0.9386| NSD     |               |
| Amaryllididae     | 11      | 5           | 16              | 0.69          | 0.9510| NSD     |               |
| Craugastoridae    | 17      | 8           | 25              | 0.68          | 0.9782*| Over    |               |
| Dendrobatidae     | 14      | 7           | 21              | 0.67          | 0.9542| NSD     |               |
| Limnodynastidae   | 12      | 6           | 18              | 0.67          | 0.9442| NSD     |               |
| Cycloramphidae    | 4       | 2           | 6               | 0.67          | 0.8173| NSD     |               |
| Pipidae           | 10      | 6           | 16              | 0.63          | 0.8761| NSD     |               |
| Hylidae           | 89      | 59          | 148             | 0.6           | 0.5992*| Over    |               |
| Myobatrachidae    | 20      | 14          | 34              | 0.59          | 0.8980| NSD     |               |
| Ranidae           | 55      | 43          | 98              | 0.56          | 0.9513| NSD     |               |
| Eleutherodactylida| 15      | 12          | 27              | 0.56          | 0.7858| NSD     |               |
| Centrolenidae     | 9       | 7           | 16              | 0.56          | 0.7361| NSD     |               |
| Hyperoliidae      | 27      | 27          | 54              | 0.50          | 0.6162| NSD     |               |
| Bombinatoridae    | 2       | 2           | 4               | 0.50          | 0.5109| NSD     |               |
| Strabomantidae    | 30      | 33          | 63              | 0.48          | 0.4610| NSD     |               |
| Leptodactylidae   | 8       | 10          | 18              | 0.44          | 0.3776| NSD     |               |
| Leiopelmatidae    | 2       | 3           | 6               | 0.40          | 0.3572| NSD     |               |
| Bufonidae         | 43      | 68          | 111             | 0.39          | 0.0178*| Under   |               |
| Ptychadenidae     | 5       | 8           | 13              | 0.38          | 0.2439| NSD     |               |
| Phrynobatrachidae | 5       | 8           | 13              | 0.38          | 0.2418| NSD     |               |
| Notophthalmidae   | 6       | 15          | 21              | 0.29          | 0.0133| NSD     |               |
| Rhacophoridae     | 5       | 14          | 19              | 0.26          | 0.0286| NSD     |               |
| Petropedilidae    | 1       | 5           | 6               | 0.17          | 0.0643| NSD     |               |
| Scaphiopodidae    | 1       | 5           | 6               | 0.17          | 0.0609| NSD     |               |
| Microhylidae      | 4       | 21          | 25              | 0.16          | 0.0006*| Under   |               |
| Megophyidae       | 1       | 6           | 7               | 0.14          | 0.0244| NSD     |               |
| Ambystomatidae    | 4       | 26          | 30              | 0.13          | 6.19E-05*| Under   |               |
| Ranellidae        | 0       | 48          | 48              | 0             | 1.05E-11*| Under   |               |
| Ceratobatrachidae | 0       | 3           | 3               | 0             | 0.04688| NA      |               |
| Brevipaludimidae  | 0       | 3           | 3               | 0             | 0.0456 | NA      |               |
| Pelodytidae       | 0       | 1           | 1               | 0             | 0.1676| NA      |               |
| Hemisomatidae     | 0       | 1           | 1               | 0             | 0.1652| NA      |               |
| **Total Anura**   | 449     | 486         | 935             |               |       |         |               |
| **Caudata**       |         |             |                 |               |       |         |               |
| Cryptobranchidae  | 3       | 0           | 3               | 1             | 0.9646| NA      |               |
| Salamandridae     | 14      | 4           | 18              | 0.78          | 0.9947*| Over    |               |
| Ambystomatidae    | 13      | 5           | 18              | 0.72          | 0.9809*| Over    |               |
recognize that these temperature ranges represented the differences in average monthly values, and spanned 4.8 to 20.3 °C. Large temperature ranges were found in: 1) Deserts and Xeric Shrublands; 2) Tropical and Subtropical Coniferous Forests; and 3) Temperate Coniferous Forests. Sites in montane areas of Mexico and Chile had moderately large temperature ranges. Sites with small temperature ranges in our analysis occurred in Tasmania, Italy, Denmark, Alaska, and South Africa.

Family-scale analyses were first conducted using all data to explore potential Bd-associations between higher-order taxonomy (family) and geographic patterns. Our analyses included the three main amphibian families in our world sample (Ranidae, Bufonidae, Hylidae; 71% of sites; Table S3). In our Bufonidae model (N = 888 sites; Figure 4), the odds of Bd detection increased by 14.7% with each 1,000-m increase in elevation, and were highest in Boreal Forest and Taiga, Temperate Coniferous Forests, Montane Grasslands and Shrublands, and Tropical and Subtropical Dry Broadleaf Forests. In Hylidae (N = 871 sites; Figure 4), the odds of detecting Bd 1) increased by 10% with each incremental increase in occurrence of enigmatic declines; 2) increased by 1.9% with each degree increase in average annual temperature; and 3) decreased by 14.9% with each degree increase in average temperature range. The highest odds of Bd detection for Hylidae were in Montane Grasslands and Shrublands, Tropical and Subtropical Dry Broadleaf Forests, and Temperate Grasslands, Savannas, and Shrublands. In Ranidae (N = 1,184 sites; Figure 4), the odds of Bd detection increased by 6.4% with each degree increase in average temperature range, decreased by 4% with each incremental increase in species richness, and decreased by 20% with each 1,000-m increase in elevation. The highest odds of Bd detection in ranids occurred in two biomes, Tropical and Subtropical Grasslands, and Deserts and Xeric Shrubland. Several biomes were dropped from the ranid analysis due to data gaps.

Table 2. Cont.

| Family | Species | Bd Detected | Bd Not Detected | Total Sampled | Prev. | p-value | Prev. pattern |
|--------|---------|-------------|-----------------|---------------|-------|---------|---------------|
| Proteidae | 1 | 1 | 2 | 0.50 | 0.5264 | NA |
| Gymnophiona | 0 | 0 | 0 | 0.0018* | Under |
| Lichthyophila | 0 | 1 | 1 | 0.1663 | NA |
| Gymnophiona | 0 | 5 | 5 | 0 | 0.1665* | Under |
| Amphiumidae | 27 | 34 | 61 | 0.44 | 0.2647 | NSD |
| Plethodontidae | 9 | 8 | 17 | 0 | 0.0165* | Under |

Prev. = Prevalence; Bd detected/total sampled; * denotes a significant deviation from the expected level of infection in that family. NSD = not significant. NA denotes that insufficient species had been sampled in that family to detect a significant deviation from a random level of infection. Power-analyses using binomial tests indicated that with six species sampled in a Family, there was sufficient power to detect a deviation in either direction. 'Bd not detected' data have limitations, as indicated in Supplemental Information. Notes on taxonomic names used are provided in Supplemental Information. Detections in table updated as of January 2011. doi:10.1371/journal.pone.0056802.t002

Figure 2. Global locations with enigmatic amphibian declines were positively associated with amphibian species richness data from GAA [7], and this relationship increased with the occurrence of the amphibian chytrid fungus, Batrachochytrium dendrobatidis, as modeled with logistic regression. Dashed lines indicate 95% confidence interval. doi:10.1371/journal.pone.0056802.g002
with average annual minimum temperatures<5°C. Sites with richness<10 species had roughly constant odds of Bd detection. For sites with average minimum temperature above 5°C, there was little change in odds of detection. Bd detection likelihood was greatest in Temperate Broadleaf and Mixed Forests. The U.S.A. Bufonidae model showed that the odds of Bd detection increased by 6.7% with each additional species at a site, and decreased 12.5% with each degree of average annual temperature. Two biomes were positively associated with bufonid Bd odds of detection: Deserts and Xeric Shrublands, and Temperate Coniferous Forests. The U.S.A. Hylidae model included only species richness and occurrence of enigmatic declines when enigmatic declines = 1, there was an estimated 29% increase in odds of Bd detection when species richness increased by one. The map of this model was not informative due to the few enigmatic declines known for U.S.A. Hylids. The U.S.A. Ranidae model included only biome. The odds of detecting Bd in U.S.A. ranids were highest in Desert and Xeric Shrublands, Temperate Broadleaf and Mixed Forests, and Temperate Coniferous Forests.

**Discussion**

Preliminary analyses of data gathered over a four-year period support broad-scale differences in patterns of infection between amphibian families. These results provide insights to family-specific environmental traits associated with Bd infection. Downscaled environmental hypotheses and assessments are warranted to gauge relationships that may help explain infection patterns at local to regional scales.

Largely due to efforts of the Bd Mapping community, we show that the spatial distribution of Bd is highly heterogeneous. It appears that a proximal driver of this heterogeneity is the rapid, continuing spread of Bd into naïve regions, and it appears that Bd has not yet reached a global equilibrium. Within this increasing pattern of spread, we identified nested variables that may influence the ecology and epidemiology of Bd following potential introduction into a region; these variables contain taxonomic (see under- and over-infected families), environmental (e.g., temperature range and precipitation at a site), and community-level (e.g., species richness) determinants. While models that combine spatial-spread dynamics with stationary site-specific equilibria are required for understanding and predicting the current and further impact of Bd, model clarity will only be achieved when we have a better understanding of the fine-scale epidemiological processes that reflect amphibian diversity, habitat, and community-level interactions.

Bd relationships with temperature are supported by experimental work showing that Bd does not grow well at temperatures<20°C [13,41], field work describing increased infection prevalence and mortality under cooler conditions [39,31,32,43], and other Bd-occurrence models, such as the downscaled model using the software Maxent which was applied to data from Australia [44]. Although altitude was not a significant predictor in the current analysis (Figure S3), the highest odds of detecting Bd were in the biome Montane Grasslands and Shrublands (Figure S4). This suggests that altitude per se is not a globally-uniform predictor of Bd presence, and that the risk of chytridiomycosis to montane-associated species may need to be assessed using regional, spatially restricted data sets. Species-specific and site-specific approaches may yield a better understanding of the dynamics of Bd-occurrence among such montane amphibians.

Our analyses support the associations of Bd with global-level temperature metrics, supporting hypotheses that the future occurrence of chytridiomycosis will respond to changes in global climate conditions [32,45]. However, there appears to be little support at this time that climate change is associated with chytridiomycosis, within at least parts of its range where this has been studied [35]. Our global data and the model created for Australia [44] suggest that both temperature range and precipitation may be of particular importance for future investigations. Our model predicts an almost 9% decrease in the odds of Bd detection with each degree-C increase in temperature range at a site, with Hylidae contributing significantly to this result and Ranidae having an opposing tendency. This aspect of our study will be particularly informative for predicting the effects of warmer or more variable weather patterns on the worldwide incidence of
Figure 4. Extrapolated global maps of the odds of *Batrachochytrium dendrobatidis* (*Bd*) detection in amphibians derived from regression models. The global model using data from all species combined is shown (A), as well as subsets of the world data for species within the amphibian families Bufonidae (B), Hylidae (C), and Ranidae (D). Although these map projections depict the odds of *Bd* detection at the world scale, it should be noted that amphibians do not occur everywhere in the world, and in particular, the three amphibian families (B,C,D) are not native to all regions world-wide. Significant model parameters differed among models and included landscape-scale site attributes including climate metrics and biotic factors, such as biome or amphibian species richness (see text). Gaps in the mapped models are due to a lack of species richness data, or amphibian absence, for certain regions.

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A Global, all species
Odds of *Bd*
High
Low
N = 3733 sampled sites

B Global, Bufonidae
N = 888 sampled sites

C Global, Hylidae
N = 871 sampled sites

D Global, Ranidae
N = 1184 sampled sites

Kilometers

A caveat here is that such broad-brush analyses inevitably miss fine-scale relationships between amphibians, their microhabitats, and the little-understood relationship between temperature and the ectothermic immune response; these factors are all likely key to determining the local outcome of infection [46].

Our finding of increased occurrence of *Bd* with species richness supports the hypothesis that greater complexity of amphibian
Table 3. Significant parameters in the regression model using *Btrachyctrymum dendrobatidis* (Bd) occurrence and data from wild-occurring amphibians having precise locations in the global Bd database (N = 3,733 locations). The impact of the Bd mapping project is still unfolding. First, this project demonstrates the value of social networking among scientists, with data compiled from hundreds of contributors who have assessed the occurrence of Bd at over 4,000 sites worldwide. Bd-maps.net allows an ongoing synergy among researchers and managers as global data are communicated or exported for further uses. Second, the project has led to a number of new studies and programs. As a consequence of the project, hypotheses have taken the initiative to: 1) initiate Bd monitoring and assessment in their region and to leverage funding for projects with expanded scope (e.g., the pan-European chytridiomycosis risk assessment RACE; www.bd-maps.eu); 2) use Bd-maps.net to direct work toward spatial and taxonomic gaps (e.g., U.S. states, Southeast Asia; e.g., Bd surveillance papers published in the Amphibian Diseases section of *Herpetological Review*); 3) to instigate disease mitigation programs in local systems (e.g., Mallorca and *Alytes muelleri* [54]); 4) include Bd-occurrence data in aquatic invasive: species alerts for land management (e.g., for water-draw decisions for western US wildlife management; Figure S5); and 5) to facilitate novel investigations into the complex ecological patterns associated with Bd infection [55].

To enhance the utility of Bd-maps.net for the global science and management community, dedicated curation and community deposition of data are needed. Currently, part-time efforts of individuals at two institutions, the U.S. Forest Service, Pacific Northwest Research Station, Oregon, U.S.A., and Imperial College, London, U.K., allow limited technical assistance for data contributors. Broader institutional support from the global community could tie data reporting to permit processes and publications for broader community participation and uses. For example, infection trend monitoring could easily be conducted for local jurisdictions, would be more effective with surveillance reporting tied to existing permitting processes, and could lead to insights for understanding transmission vectors or infectivity correlates. Molecular typing of Bd isolates [16] is continuing and is planned to be integrated with mapping data to understand the global spread of infection within a population genomics framework [56]. The addition of molecular data to Bd-maps.net, alongside rapid next-generation genotyping techniques, may aid in identifying the major sources and sinks of infection, and the dominant global vectors of Bd. We are applying our lessons-learned in the development of Bd-maps to the current development of a global Ranavirus Reporting System [57,58]. This developing system includes data fields to assess data certainty and potential bias. Understanding of the vectors for both Bd and Ranaviruses will be key for regional biosecurity, with subsequent emphasis on management of the human-mediated transmission pathways. An expected outcome is the need for increased transparency of disease transmission pathways in amphibian trade within the context of the OIE legislation, as well as a call for all countries contributing to the OIE infrastructure to contribute to the detection and containment of novel wildlife diseases. These approaches are synergistic with those proposed to contain the emergence of newly emerging human diseases, and while the challenges are great, the

| Parameter                  | Coefficient | SE     | p value |
|----------------------------|-------------|--------|---------|
| Intercept                  | 1.186       | 0.262  |         |
| Enigmatic Declines         | 0.141       | 0.028  | <0.001  |
| Temperature Range          | -0.089      | 0.017  | <0.001  |
| Average Annual Precipitation | -0.0005   | 0.003  | <0.001  |
| Biome                      | --          | --     |         |
| Tundra (included in intercept) | --      | --     |         |
| Montane Grasslands & Shrubs | 1.092     | 0.366  |         |
| Mediterranean Forest, Woodlands & Scrub | -0.598   | 0.126  |         |
| Tropical & Subtropical Dry Broadleaf Forests | -0.346  | 0.406  |         |
| Boreal Forests/Taiga        | -0.649      | 0.340  |         |
| Mangroves                  | 0.251       | 0.773  |         |
| Deserts & Xeric Shrublands | 0.471       | 0.186  |         |
| Flooded Grasslands & Savannas | 0.323    | 0.576  |         |
| Temperate Broadleaf & Mixed Forests | 0.397   | 0.122  |         |
| Temperate Grasslands, Savannas & Shrubs | -0.346  | 0.172  |         |
| Temperate Coniferous Forests | 0.229    | 0.124  |         |
| Tropical & Subtropical Grasslands | -0.582  | 0.198  |         |
| Tropical & Subtropical Moist Broadleaf Forests | 0.224  | 0.172  |         |
| Tropical & Subtropical Coniferous Forests | 0.503  | 0.278  |         |

| Parameter                  | Coefficient | SE     | p value |
|----------------------------|-------------|--------|---------|
| Intercept                  | 1.186       | 0.262  |         |
| Enigmatic Declines         | 0.141       | 0.028  | <0.001  |
| Temperature Range          | -0.089      | 0.017  | <0.001  |
| Average Annual Precipitation | -0.0005   | 0.003  | <0.001  |
| Biome                      | --          | --     |         |
| Tundra (included in intercept) | --      | --     |         |
| Montane Grasslands & Shrubs | 1.092     | 0.366  |         |
| Mediterranean Forest, Woodlands & Scrub | -0.598   | 0.126  |         |
| Tropical & Subtropical Dry Broadleaf Forests | -0.346  | 0.406  |         |
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| Tropical & Subtropical Moist Broadleaf Forests | 0.224  | 0.172  |         |
| Tropical & Subtropical Coniferous Forests | 0.503  | 0.278  |         |

Individual biome coefficients represent additive shifts in the Bd-occurrence odds ratio, whereas other coefficients represent multiplicative changes in odds of detecting Bd. Logistic regression with likelihood ratio test statistic (Chi²) was used.

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Global Bd Mapping Project
imperative to curb the spread of these globally-devastating infections has never been greater.

Materials and Methods

Data compilation

*Bd* location data were compiled from published literature and unpublished studies (Table S1). We defined 'site' as a geographically distinct location having specific coordinates and/or locality description. A 'record' was a database entry for a single species for a single year, at a particular site. A given site may have a number of records for different species or years, some for which *Bd* was found, others for which it was not found. A 'country centroid' is a point designation for a type of site that placed sampled species in the country of origin, where finer-scale location data were not available. Country centroid data were compiled generally from early years of *Bd* surveillance before Global Positioning Systems were frequently used to determine site coordinates.

Site-level data were used in analyses, and if both positive and negative *Bd* records were available for a site, the site was designated *Bd*-positive. Analyses did not include data from captive animals, vague locations, or country centroids. For analyses, a snapshot of site-level data was taken in July 2010, including all data compiled in our database up to that time. Analyses were conducted using a global grid spatial resolution of 0.5 degrees latitude and longitude, which was the finest scale available for the global climate data. Finer-scale habitat patterns will not be apparent from our analyses, such as those with more restricted temperature regimes. Parameters used in our models are listed in Table S3.

To provide a more updated summary of *Bd* occurrence (Table 1, Table 2, Table S1), we report tallies of *Bd*-positive and *Bd*-negative data compiled up through a data snapshot taken on 30 January 2011. These summaries were used for the over/under-infection randomization analysis. Our data are available via a world *Bd* mapping database and interactive web application that we developed for the *Bd* mapping project, which has merged and extended pre-existing data compilation efforts (http://www.Bdmaps.net). The Bdmaps.net website facilitates automatic surveillance by country or species at several spatial scales which can be user-determined by an interactive mapping tool using Google Maps technology. Hence, real-time counts by country and species are available online, and recent data snapshots of these tallies were taken on 23 July 2012 for this paper. Some sensitive data were not released for precise mapping, and hence our hierarchical mapping method enables their mapping at world scale but not at finer scales. However, sensitive data were included in our data snapshots for both analyses and more recent summaries. For the purposes of this paper, a data snapshot was undertaken for our analyses (June 2010), and we have provided additional time-sensitive snapshots for more updated information in Figure 1 (August 2012) and Table S1 (January 2011). Data updates and analyses are time-intensive procedures to ensure duplicate records are screened from our counts, accounting for different snapshot dates reported here. The most recent (July 2012) counts of *Bd*-infected countries and species is the only realtime data summary available at this time.

Caution is needed due to several potential data constraints when interpreting our maps and analyses. Limited quality assurance was conducted, such that if inaccurate methods were used or results reported, those data may not have been screened out of this compilation (e.g., false positives and negatives may be included in the data). *Bd* prevalence in a population may be low, such that a small sample size of individuals screened for *Bd* may not detect it, and consequently could yield a false negative report (‘no detection’ does not indicate absence of *Bd*; see [39]). This could be a particular concern for data collected before 2008, when the recommendation for sample sizes>50 be collected to assess *Bd* occurrence when prevalence is low [39]. In addition, it is possible that negative data are not consistently reported, especially from early exploratory studies or due to the challenge of publishing negative data, and hence negative data may be under-represented in our compilation. To conduct adequate quality assurance of the global data set to reduce potential biases, individual studies would need to be screened, which would require each data source to be re-visited and assessed. Our aggregation of data at the 0.5 degree latitude and longitude scale likely reduces these biases by compiling data among different sampling events, including separate studies (and their methods), researchers, species, seasons, and years. Finally, our world map of *Bd* detections (Figure 1) includes samples taken from living or dead animals—appearing healthy or symptomatic, native and introduced animals, and wild and captive animals. We do not map the disease chytridiomycosis wherein animals showed symptoms of the disease. This last point should be emphasized; we report *Bd* occurrence and hence the threat of disease and potential taxonomic or geographic sources of disease transmission, but we have not distinguished locations with mortality attributed to the disease and our maps do not represent declining amphibian populations. In addition, our logistic regression models should not be used for direct prediction, and coefficients from the fitted models should only be interpreted in terms of relative (not absolute) contribution to the odds of *Bd* detection. Because sites were not randomly selected, the inference of each model is restricted to those sites used in the analysis. However, with 3,733 sites used in our full model, we feel there is useful predictive capacity of those data to be used as hypotheses for future investigation and later downscaled analyses, especially for the better-sampled regions of the world and more heavily-sampled amphibian families.

During modeling, all of our 3,961 *Bd* sites during that data snapshot did not fall cleanly within available environmental coverages, and hence some were not included in regression analyses. For example, the biome coverage did not overlay cleanly with the *Bd* site data. For 227 sites, site location did not correspond to an environmental coverage after the overlay was conducted, and these were eliminated from the model. These sites were distributed around the world and often were near coastlines, suggesting the coastal accuracy of the landscape-scale environmental boundaries was poor; hence, these sites had very low elevations. However, there were often neighboring locations that were sorted into a biome, so locations with clusters of samples were not generally lost from analyses. We also deleted the 'lake' biome from analyses (1 site) because lentic habitats within other biome types were not distinguished.

Statistical Analysis

Using logistic regression (programmed in R software [59]), we examined associations between the *Bd* occurrence at a site (world scale: all data and three well-sampled families separately – Bufonidae, Hylidae, Ranidae) and: 1) latitude; 2) elevation [60]; 3) biome [61]; 4) amphibian species richness [27]; and 5) global temperature and precipitation metrics [62]; Climatic Research Unit, University of East Anglia, Norwich, UK; www.cru.uea.ac.uk) (Table S2). These parameters were chosen to represent world-scale ecological factors that may be correlated with *Bd*. Our intent was to use our robust data set to address existing hypotheses of *Bd* occurrence correlations, and to contribute to new ecological research and surveillance directions. We did not include other
factors that may better represent a human footprint signal from modern-day disturbances (e.g., proximity to roads, cities, agricultural areas) largely because these data layers are inconsistent at the world scale.

Additionally, because the United States was a relatively well-sampled large-scale region, we conducted similar analyses for all U.S.A. data and well-sampled families in the U.S.A. (Bufonidae, Hylidae, Ranidae). More precise temperature and precipitation data were used in the U.S.A. analyses (Parameter-elevation Regressions on Independent Slopes Model [PRISM], C. Dole. PRISM Climate Group, http://prism.oregonstate.edu/). Altogether, eight Bd-occurrence models were generated: world scale, all data and data from three families; U.S.A. scale, all data and data from three families.

Logistic regression also was used to examine the relationship between locations with enigmatic population declines (7) and both Bd detection and species richness (decline and richness data sources: (7)) detailed in Table S3. Again, we conducted eight analyses: world scale with all world data, and world data from three families; U.S.A. scale with all U.S.A. data, and U.S.A. data from three families.

Randomisation tests were used to determine whether each sampled amphibian family was over- or under-infected, compared to what we would expect by chance given an overall prevalence of 50±1055 species (using the January 2011 data snapshot reflected in Table 1, Table 2, and Table S1). Ten thousand unconstrained randomisations were performed using a computer program written in the statistical language R (59); infection data was randomised across all species in the data set, and the number of species in the focal family that were randomly assigned as infected was counted. The null distribution generated was then compared to the observed number of infected species within the focal family, and the null hypothesis (that infection was random) was rejected if the observed value lay in the 2.5% for either tail of the null distribution, and statistical power was sufficient (n. species≥6).

Power-analyses using binomial tests indicated that with six species sampled in a family, there was sufficient power to detect a elevation in either direction.

Supporting Information

Figure S1 Occurrence of Batrachochytrium dendrobatidis (Bd) at sites by latitude. The high number of sample sites in the Northern Hemisphere, particularly the United States and Spain, gives a pronounced skew to this distribution. (TIF)

Figure S2 Taxonomic patterns of Batrachochytrium dendrobatidis (Bd) occurrence by latitude. The extraordinary peak in species richness in the 8°N range highlights the overlap of many families of amphibians in Central America and also the intensive sampling represented by Lips et al. (2003, 2006).

Table S1 Families considered Threatened (the previous three categories are considered Threatened by the IUCN; NT = Near Threatened; LC = Least Concern; DD = Data Deficient; NE = Not Evaluated (newly discovered or newly recognized species). Last updated from published literature in March 2011; references are given in Supplemental Information text, File S1. (DOCX)

Table S2 Parameters used in regression analyses. Numbers in parentheses after biomes indicate number of sites in the full analysis (N = 3733 total sites). Sites with no biome assigned (n = 227) were dropped from the analysis. Lake biome (N = 1 site) was also dropped due to inconsistency of lentic habitat designation. (DOCX)

Table S3 Significant parameters in three family-scale logistic regression models using Batrachochytrium dendrobatidis (Bd) occurrence data from wild-occurring amphibians having exact and approximate locations in

Global Bd Mapping Project

PLOS ONE | www.plosone.org 11

February 2013 | Volume 8 | Issue 2 | e56802
the global Bd database. Individual bioclimatic coefficients represent additive shifts in the *Bd*-occurrence odds ratio, whereas other coefficients represent multiplicative changes in odds of detecting *Bd*. Logistic regression with likelihood ratio test statistic (Chi) was used.

(BOC)

**File S1** Supplemental Information: Taxonomic Notes; References for Supplemental Material.

(BOC)

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**Author Contributions**

Conceived Global Bd Mapping Project: DHO. Conceptually designed interactive website: DMA MCF D0027. Led website development: DMA. Assisted website development: SWF. Conceived and designed the experiments: CH. Led compilation of world data and summarisation: KLR. Analyzed the data: GW. BJH DHO MCF KLR SWF. Contributed reagents/materials/analysis tools: DHO DMA KLR. Led compilation of world data and summarisation: KLR. Analyzed the data: GW. BJH DHO MCF KLR SWF. Contributed reagents/materials/analysis tools: DHO DMA KLR. Contributed to the writing of the paper: DHO DMA KLR BJH TWVG SWF GW CIP.

**References**

1. Ferguson NM, Donnelly CA, Anderson RM (2001) Transmission intensity and impact of control policies on the first and second epidemic in Great Britain. Nature 413: 542-548.
2. Berger L, Speare R, Da Fonseca DE, Cunningham AA, et al. (1998) Chytridiomycosis causes amphibian mortality associated with population declines in the rain forests of Australia and Central America. Proc Natl Acad Sci U S A 95: 9031-9036.
3. Johnson PJ (2006) Amphibian diversity: Decimation by disease. Proc Natl Acad Sci U S A 103: 3011-3012.
4. Guisan A (2001) Amphibian Conservation Action Plan, IUCN/SSC Amphibian Specialist Group, Gland, Switzerland and Cambridge, UK.
5. McCallum H, Dobson A (2007) Amphibian decline or extinction? Current declines cross eco-region boundaries: J Herpetol 41: 483-483.
6. Stewart IJF, Berger L, Speare R, Cash, S, McDonald KR, et al. (2007) Spread of chytridiomycosis has caused the rapid global decline and extinction of frogs. Trends Ecol Evol 4: 124-134.
7. Stuart SN, Chanson JS, Cox NA, Rodrigues L, et al. (2004) Status and trends of amphibian declines and extinctions worldwide. Science 306: 1783-1786.
8. Wall DH, Vredenburg VT (2008) Are we in the midst of the mass extinction? A view from the world of amphibians. Proc Natl Acad Sci U S A 105: 11466-11473.
9. Colón-Gaud C, White MR, Kilham SS, Lips KR, Pringle CM, et al. (2009) Assessing ecological responses to catastrophic amphibian declines: patterns of macroinvertebrate production and food web structure in upland Pennsylvania streams. Nord Oecol 54: 331-343.
10. Colón-Gaud C, White MR, Kilham SS, Lips KR, Pringle CM, et al. (2010) Macroinvertebrate community responses to catastrophic amphibian declines in upland streams: J North Am Benthol Soc 29: 1105-1180.
11. Connolly S, Pringle CM, Kilham SS, Lips KR, Pringle CM, et al. (2011) Do tadpoles affect leaf decomposition in upland streams? Freshwater Biol 56: 1863-1873.
12. Fisher MG, Hoss DA, Beiges C, Brownstein JS, Mayolf L, et al. (2011) Emerging fungal threats to animal, plant and ecosystem health. Nature 484: 109-114.
13. Legg N, Kile DM, Nicholls DK (1999) *Batrachochytrium dendrobatidis*. gyn sp nov, a chytrid pathogenic to amphibians. Mycologia 91: 219-227.
14. Fisher MG, Garner T, Walter J (2009) Global emergence of *Batrachochytrium dendrobatidis* and amphibian chytridiomycosis in space, time and host. Annu Rev Microbiol 63: 291-310.
15. Garner TW, Walker S, Boettcher J, Leitch A, Rayfield E, et al. (2008) Life history trade-offs influence mortality associated with the amphibian pathogen *Batrachochytrium dendrobatidis*. Oikos 118:783-791.
16. Fallows RA, Weinsten LA, Bielby J, Garner TW, Wallis F, et al. (2011) Multiple emergences of genetically diverse amphibian-infecting chytrids include a
Global Bd Mapping Project

32. Bond J, Carracluel LM, Durrin L, Walker S, Fisher MC (2007) Climate change and outbreaks of amphibian chytridiomycosis in a montane area of Central Spain: is there a link? Proc R Soc Lond [Biol] 274: 237-240.

33. Kruger KM, Horo JM (2007) Large-scale seasonal variation in the prevalence and severity of chytridiomycosis. J Zool 271: 352-358.

34. Eise KS, Beeman F, Byrnes R, Reeve JD, Alford KA, et al. (2006) Emerging infectious disease and the loss of biodiversity in a Neotropical amphibian community. Proc Natl Acad Sci U S A 103: 3163-3167.

35. Lips KR, DuFresnay JF, McNeely HJ, Brooks MW (2008) Flying waves: Reconceiving the roles of distance and climate change in amphibian declines. PLoS Biol 6: e441-e454.

36. Muller EF, Pödith DB, Livo LJ (2008) Distribution and environmental limitations of an amphibian pathogen in the Rocky Mountains, USA. Biol Cons 141: 1484-1492.

37. Murray KA, Skerratt LF, Speare R, McCallum H (2009) Impact and dynamics of disease in species threatened by the chytrid fungal fungus, Batrachochytrium dendrobatidis. Conserv Biol 23: 1242-1252.

38. Ankevort DM, Humby DM, Fell EJ, de-Owen F, Syran BG (2009) EpilCoRe: Linking amphibians to web applications for epidemiology, ecology and community data collection. PLoS ONE 4(9): e7058. doi:10.1371/journal.pone.0007058.

39. Skerratt LF, Berger L, Hines HR, McDonald KR, Munder D, et al. (2010) Survey protocol for detecting chytridiomycosis in all Australian frog populations. Dis Aquat Org 86: 83-94.

40. Bulley J, Cooper N, Cunningham AA, Garner TW, Purvis A (2008) Predicting susceptibility to future declines in the world’s frogs. Conserv Lett 1: 82-90.

41. Poinar J, NCalc, Longcore JF (2009) Physiology of Batrachochytrium dendrobatidis, a chytrid pathogen of amphibians. Mycologia 91: 9-15.

42. Retaliuk RWR, McCallum H, Speare R (2004) Epidemic infection of the amphibian chytrid fungus in a frog community post-decline. Philos Biol 2: 1963-1971.

43. Forrest MJ, Schleeplema MA (2011) Nothing’s a hot bath won’t cure: infection rates of amphibian chytrid fungus correlate negatively with water temperature under natural field settings. PLoS ONE 6(4): e18444. doi:10.1371/journal.pone.0018444.

44. Murray KA, Retaliuk RWR, Porschendorf R, Skerratt LF, Reisser D, et al. (2011) Assessing spatial patterns of disease risk to biodiversity: implications for the management of the amphibian pathogen, Batrachochytrium dendrobatidis. J Anim Ecol 80:163-173.

45. Parnell AJ, Bauntonne MR, Cordera IA, Consuegra JA, Fedenr MIP, et al. (2006) Widespread amphibian extinctions from epidemic disease driven by global warming. Nature 439: 161-167.

46. Leitch R, Ryan MJ, Poiier A, Choppaib B (2008) Infection with the fungus Batrachochytrium dendrobatidis in a non-native Edithvale devilfish below sea level in the Coachella Valley, California, USA. Herpetol Rev 39.

47. Fenton A, Piotersen A (2005) Community epidemiology framework for classifying threats. Emerging Infect Dis 11: 1013-1021.

48. de Castro F, Rolle B (2010) Mechanisms of disease-induced extinction. Ecol Lett 11: 117-126.

49. Vredenburg VT, Krapf R, Fauszall T, Briggs CJ (2010) Dynamics of an emerging disease: large-scale amphibian population extinctions. Proc Natl Acad Sci U S A 107: 5680-5684.

50. Kinney VC, Hermmeyer JJ, Poiier AP, Lamou MJ (2011) Seasonal pattern of Batrachochytrium dendrobatidis infection and mortality in Edithvale devilfish: Affirmation of Vredenburg’s ‘30,000 Zoopool Rule’. PLoS ONE 6: e16781.

51. Woodhams D, Ardeshir R, Allford RA, Maranelli G, Reisser EK, et al. (2007) Resistance to chytridiomycosis varies among amphibian species and is correlated with skin psychrotroph abundance. Anim Conserv 10: 409-417.

52. Briggs CJ, Vredenburg VT, Krapf R, Rachevitz LJ (2005) Investigating the population-level effects of chytridiomycosis: An emerging infectious disease of amphibians. Ecology 86: 3149-3155.

53. Mitchell KM, Churcher TS, Garnett TWG, Fisher MC (2009) Persistence of the emerging pathogen Batrachochytrium dendrobatidis outside the amphibian host greatly increases the probability of host extinction. Proc R Soc Lond [Biol] 276: 329-334.

54. Lubick N (2010) Emergency medicine for frogs. Nature 465: 680-681.

55. Boczar R, Han B, Scoble C, Michael I, Olmer DL, et al. (2011) Environmental and biological correlates of species susceptibility to Batrachochytrium dendrobatidis in amphibians. Biol Conserv 144:1911-1920.

56. James TY, Levinton A, Vilkaski, Moogen JA, Taylor JW, et al. (2009) Rapid global expansion of the fungal disease chytridiomycosis into declining and healthy amphibian populations. PLoS Pathog 5: e1000418. doi:10.1371/journal.ppat.1000418.

57. Duffin A, Olmer D (2011) The establishment of a global Rannius reporting system. Frogsblog 96: 37.

58. Marchong RE, Miller D (2011) 2011 International Rannius symposium. J Herpetol Med Surg 21: 1-2.

59. Development Core Team (2009) R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria, ISBN 3-900051-07-0. URL: http://www.R-project.org.

60. ESRI [United States Geological Survey] (2009) Global digital elevation model GTOP030. [US Geological Survey] global digital elevation model GTOP-030<http://edc-usgs-gov/products/daspberry/geopro30/geopro30.html>accessed 28 October 2009.

61. Olson DM (2001) BioScience 51: 933-938. <http://www.worldwildlife.org/ science/data/item/075.html>accessed 4 October 2009.

62. Mitchell TD (2009) Tsaldar Centre Working Paper No. 35. <http://www. tsaldar.ac.uk/publications/working_papers/wp35_summary.html>accessed 4 November 2009.