#### RESEARCH ARTICLE



# Integrating small population effects on both detection and persistence probabilities in the design and interpretation of presence/absence surveys. [v1; ref status: not approved 1,

http://f1000r.es/21l]

Brian R. Hudgens, David K. Garcelon

Institute for Wildlife Studies, Arcata, CA 95518, USA

v1

First published: 06 Nov 2013, **2**:234 (doi: 10.12688/f1000research.2-234.v1)

Latest published: 06 Nov 2013, **2**:234 (doi: 10.12688/f1000research.2-234.v1)

#### Abstract

Determining whether or not a species has been extirpated from part of its range is one of the most important yet difficult challenges facing ecologists. Presence/absence surveys are a primary method to determine local extinction, but negative results are difficult to interpret, particularly when informal sightings appear to support local persistence of elusive species. Such conflicts are particularly likely for rare species, which are the most difficult to detect during a presence/absence survey due to small population size. However, integrating small population effects on both detection and extinction probabilities resolves such conflicts by reframing occupancy questions from present to near future terms. We developed methods for integrating the simultaneous effects of survey effort and population size on detection and extinction probabilities in the design and interpretation of presence/absence surveys. We applied these methods to the design and interpretation of a wolverine survey in Seguoia-Kings Canyon National Parks, California, USA, to address questions about the appropriateness of translocations from other populations. Ecological applications that can benefit from this type of analysis, simultaneously incorporating small population effects on both detection and population processes, include management of charismatic, yet elusive species, management of disease risk in translocations, and management of nonnative pests.

Article Stat	Article Status Summary								
Referee Responses									
Referees	Referees 1								
v1 published 06 Nov 2013	report								
1 Jonathan	1 Jonathan Rhodes, University of								
Queenslar	Queensland Australia								
Latest Comments									
Brian Hudgens, Institute for Wildlife Studies,									
USA									
07 Apr 2014 (	07 Apr 2014 (V1)								

Corresponding author: Brian R. Hudgens (hudgens@iws.org)

How to cite this article: Hudgens BR and Garcelon DK (2013) Integrating small population effects on both detection and persistence probabilities in the design and interpretation of presence/absence surveys. [v1; ref status: not approved 1, http://f1000r.es/21l] F1000Research 2013, 2:234 (doi: 10.12688/f1000research.2-234.v1)

Copyright: © 2013 Hudgens BR and Garcelon DK. This is an open access article distributed under the terms of the Creative Commons Attribution Licence, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited. Data associated with the article are available under the terms of the Creative Commons Zero "No rights reserved" data waiver (CC0 1.0 Public domain dedication).

Grant information: The author(s) declared that no grants were involved in supporting this work.

Competing interests: No competing interests were disclosed.

First published: 06 Nov 2013, 2:234 (doi: 10.12688/f1000research.2-234.v1)

#### Introduction

Detection of rare species is one of the fundamental problems in conservation biology. A species cannot be managed for unless its presence is known. Consequently, presence-absence surveys are common first steps in guiding management policies. But, while formal surveys are very good at confirming a species presence, determining if a species is absent from a management unit is more problematic. The crux of the problem lies in the omnipresent possibility that nondetection does not indicate a species absence. Clearly, incorporating nondetection probability, through its inverse, detection probability, is a critical component to both the design and interpretation of presence-absence surveys, and there are well-developed methods for doing so [e.g., 1.2].

The simplest of these methods assume that detection probabilities depend only on survey effort<sup>1</sup>. Consequently, designing a survey with a small chance of failing to detect a species present in a management unit depends only on knowing detection probabilities, which can be estimated from surveys conducted in areas of known occupancy [e.g., <sup>1,3</sup>]. In the absence of reliable information from surveys in known occupied areas, detection probability is often estimated from similar species<sup>4</sup> or simulation<sup>5,6</sup>.

Of course, detection probabilities do not depend only on survey effort. At the very least, the size of the population within the survey area influences detection probabilities; smaller populations are more difficult to detect than larger ones. Moreover, the influence of population size on detection probability is nonlinear such that the loss of an individual has a larger impact on the detection probability of a small population than of a large one<sup>7</sup>. This relationship has two underappreciated consequences for designing and interpreting presence-absence surveys. First, the effort required to detect a small population increases at an accelerating pace as population size dwindles. Second, determining the survey effort required to have an acceptably small nondetection probability requires specifying the size of the population being looked for (hereafter, target population size).

Given limited resources, it generally will not make sense to dedicate a lot of resources to ensure a high probability of detecting the last individual of a species within a management unit. In other words, it is inefficient to spend limited resources to detect a species which may technically be present within a management unit but is effectively locally extinct. It would be more efficient to design a survey to detect a population of two individuals, but such a population would still likely require a large effort to detect and would still face a high risk of local extinction in the near term. It would generally be even more efficient to design a survey based on a target population of three individuals, but because increasing population size generally has a diminishing effect on both detection probability and extinction risk (assuming no Allee effect<sup>8</sup>), the gain in efficiency would not be as large as the gain in efficiency realized from designing for a target population of two vs. one.

Ideally, a survey design avoids two sources of inefficiency: failing to detect a population that persists into the near future and detecting a population that fails to persist into the near future. As such, the target population size should account for the effects of population size on

both detection probability and extinction risk (Figure 1). This can be done thinking of survey power not in terms of the probability of failing to detect a population, but in terms of the probability of failing to detect a viable population. The latter is the equivalent of the product of both the probability of failing to detect at least one individual, which is an increasing function of population size, and the probability of a population persisting into the near future, which is a decreasing function of population size. Framing a presence-absence survey in terms of detecting a viable population also changes the interpretation of null survey results. Rather than a null result implying a species absence, a null result implies that if the species is present in a management unit, its population is so small that it is unlikely to persist into the near future.

In this paper, we develop a framework for survey design which incorporates small population effects on both detection probability and short-term persistence. We then demonstrate the application of this framework to the design and interpretation of a presence-absence survey for wolverines in Sequoia-Kings Canyon National Parks, California, USA. Finally, we discuss the implications of the wolverine survey for wolverine management in California and the circumstances under which the framework developed here is, and is not, appropriate for designing and interpreting presence-absence surveys.

#### **Methods**

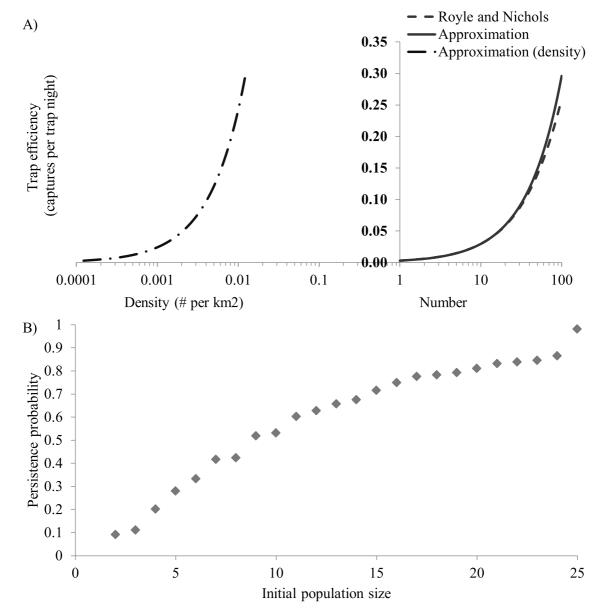
#### Analytical framework

We developed methods for optimizing survey effort to detect a viable population given that one persists in the survey region, and for a post-hoc analysis of survey power which serves as the basis for evaluating negative survey results in terms of the probability of overlooking a population that will persist into the near future. A good survey has two critical components: 1) an appropriate method to detect the species of interest, and 2) adequate sampling effort to have a low probability of overlooking the species if it is present. Here we focus on how to determine the amount of survey effort required once appropriate methods have been chosen. Although we envision a survey based on baited camera traps, with detections based on photographing an animal visiting a bait station, the methods described work with any survey protocol for which effort can be quantified in discrete units (e.g., trap-nights, kilometers traversed, etc.) and detection probability can be expressed per unit of effort.

Assuming that additional traps do not interfere with one another, the probability of detecting a species (D) for a given amount of effort is:

$$D=1-(1-E)^{T} (equation 1)$$

where E is the trap efficiency (i.e., the probability of a detection during a single unit of survey effort) and T is the number of trap-days (i.e., total survey effort) in the study. If all individuals within the surveyed population have the same chance of being detected, trap efficiency will be related to population size as  $E=1-(1-e)^{N7}$ , where e is the probability that a particular individual is detected and N is the number of individuals in the survey area. After substitution and simplification, Equation 1 can then be rewritten to account for population size as:



**Figure 1. Small population effects on detection and persistence probabilities for wolverines. A**) Lines to the right of the Y-axis depict how trap efficiency increases with population size calculated using the methods of Royle and Nichols<sup>7</sup> (dashed line) and the approximation presented in this paper (solid line). The line to the left of the Y-axis shows how trap efficiency increases with increasing density. Density values correspond to the population sizes used in the right-hand side of the graph interacting in an 8171 km² area. Trap efficiencies were calculated assuming an individual capture probability of 0.003 or reference trap efficiency of 0.024 for a reference population size of 8 individuals at a corresponding reference density of 0.0024 animals per square km. The X-axis is plotted on a log<sub>10</sub> scale to emphasize small population numbers. For raw data, please see Dataset 1. **B**) Points indicate the probability that a population starting at the population size on the X-axis persists for 25 years predicted by PVA (see text for details). For raw data, please see Dataset 2.

$$D(N)=1-((1-e)^{N})^{T}$$
 (equation 2)

To calculate the survey effort required assuming a population size N, to achieve the desired detection probability (D(N)), we solve for T(N):

$$T(N)=\ln(1-D(N))/\ln((1-e)^{N})$$
 (equation 3)

Note that as the population size of the target species in the survey region approaches zero, the required effort approaches infinity.

The nonlinear relationship between T and N as N approaches zero means that surveys designed to detect the lowest possible size (i.e., a single individual within the survey area) are likely to be prohibitively large.

The next step to optimize survey effort is to consider the effects of population size not only on detection probability, but also on near-term population viability (i.e., the probability that a population persists beyond some time, t, in the near future). There are many

ways to determine the relationship between density and population viability using population viability analysis (PVA). Choosing the most appropriate type of PVA depends on what data are available for the species. Commonly used count-based PVA, which treats changes in the (log) population size as a diffusion process, explicitly assumes a relationship between population size and extinction risk. For more complex population dynamics (e.g., density dependent or age structured populations), the relationship can be determined by simulation. Here, we assume that there is some relationship between near-term population viability and the current population size (Z(N)) and that this relationship has been determined using PVA (e.g., Figure 1B).

When taking population viability into account, the relevant statistic is probability of failing to detect a viable population given that the species inhabits the survey area at a given population size:

$$P(N)=(1-Z(N))\times(1-D(N))$$

Substituting for D(N) and simplification yields

$$P(N)=(1-Z(N))\times(((1-e)^{N})^{T})$$
 (equation 4)

The maximum value of P(N), as N varies in equation 4, approximates the probability that a survey failed to detect a population that will persist into the near future. The exact probability is equal to the integral of the product of P(N) and the probability that the population being surveyed has N individuals over all possible values of N. However, the latter probability distribution is extremely unlikely to be known or even estimable for populations which are the focus of presence/absence surveys. Approximating this interval at the maximum value of P(N) results in an estimate of P(N) that is biased high, which for most applications of this analysis will be a conservative bias.

Because most power analyses calculate the probability of success (e.g., that a study will show a significant difference given that observed groups do differ), it is tempting to estimate survey power as 1-P(N). We caution against doing so because this quantity is difficult to interpret; it is the union of the probabilities that a population is detected or goes extinct in the near future, not the probability that a population is detected given that a viable population persists in the survey area, nor the probability that an undetected population goes extinct.

To determine the **optimal** trap effort, one must first determine the desired level of risk for overlooking a viable population independent of population size,  $P(\bullet)$ . Choosing a critical value for  $P(\bullet)=(P_{crit})$  and rearranging equation 4 to solve for T yields

$$T=(ln(P_{crit})-ln(1-Z(N)))/ln((1-e)^{N})$$
 (equation 5)

At N=0, equation 5 evaluates to infinity, reflecting that no amount of effort can detect an extinct viable population (itself a logical impossibility). As N increases, equation 5 has a maximum value which indicates the optimal survey effort. If Z(N) is a differentiable function,  $T_{max}$  can be calculated analytically. If Z(N) was determined by simulation then  $T_{max}$  must be determined by iterating equation 5, but since extinction risk generally decreases sharply with population size, iterations will typically only have to span a small range of N. Note that if population viability and reference detection probabilities are provided in terms of population density rather than population numbers, equation 5 may still be used by defining N as population density.

#### Application to Sierra Nevada wolverines

Pre-survey power analysis. We applied the techniques outlined above to design and interpret a presence-absence survey for wolverines in Sequoia-Kings Canyon National Parks (SKCNP) to address concerns related to the impact that a proposed translocation of wolverines into the southern Sierra Nevada would have on a possible remnant population. The concerns arise from a contradiction between management agencies' - in particular, California Department of Fish and Wildlife and SKCNP — stance that wolverines persist and should be managed for in the Sierra Nevada<sup>10,11</sup> and published conclusions that the species has been extirpated from the state 12,13. The conflicting stances largely stem from differences in opinion about how to treat unverified reports by citizen observers and a failure of formal surveys to address survey power. Because most carnivore surveys in the Sierra Nevada take place in the summer<sup>2,14</sup>, when wolverines are not likely to visit bait stations (there are no published records of summertime efforts which have trapped wolverines), it is not unreasonable to suppose that formal surveys may have overlooked a small remnant population in the Sierra Nevada given numerous reported sightings of the species. The recent discovery of an immigrant male wolverine more than 250 km north of SKCNP<sup>15</sup> has done little to quell the debate.

In order to resolve this conflict, we carried out a winter survey for wolverines using baited camera stations in SKCNP designed to have a high power to detect a wolverine population large enough to have a reasonable probability of persisting at least 25 years. Given a 75-year absence of physical evidence of wolverine presence in the area<sup>13</sup>, we assumed that if wolverines did remain in the parks, they persisted in low densities compared to areas where detection probabilities were available from previous winter surveys. Consequently, it was clear that the effect of population size on detection probability needed to be accounted for.

Only two published winter carnivore surveys provide estimates of both detection rates from baited camera stations and wolverine density <sup>16,17</sup>, and none report both the trap efficiency and population estimate within the effective trap area. Consequently, it is not possible to estimate the per capita detection probability (e) of a wolverine. However, when N and e are small, E is closely approximated by (N/N\*)E\* where E\* is the trap efficiency at a reference site with population size N\* (Figure 1A). Application of this approximation for E into equation 5 yields:

$$T = (ln(P_{crit}) - ln(1 - Z(N)))/ln(1 - (N/N^*)E^*)$$
 (equation 6)

Assuming a constant area, changes in density (d) are equivalent to changes in population size, such that N/N\* may be replaced by  $d/d^*$ . The average trap efficiency and density estimates from these studies were E\*=0.024 detections per trap-night and  $d^*$ =0.0098 wolverines per km². Corresponding trap efficiencies over a range of wolverine densities are shown in Figure 1A.

We determined how wolverine population size affects extinction risk using the PVA modeling program VORTEX (v.9.7.2, Chicago Zoological Society 2007) parameterized with demographic rates taken from the published literature (see Appendix A). VORTEX is a widely used, flexible, individual based simulation<sup>18</sup> that has been found to perform relatively well compared to other PVA packages<sup>19</sup>. We set initial population size between 2 and 25 animals and

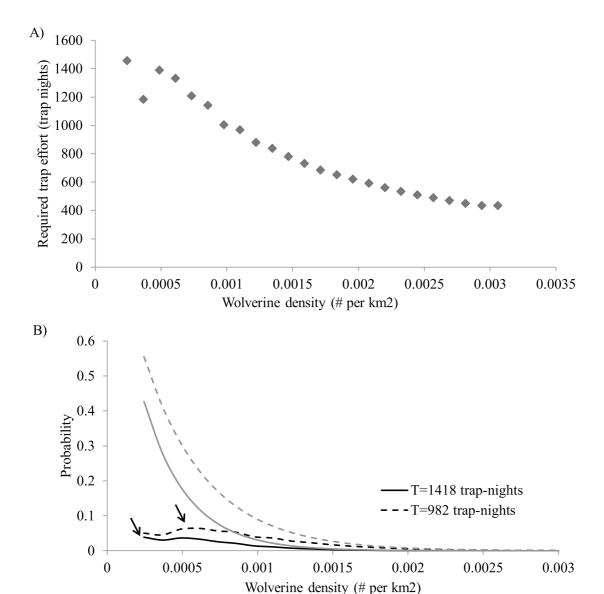


Figure 2. Preliminary and post-hoc power analysis for wolverine survey. Top panel A) shows the trap effort required to have 95% probability of detecting a viable wolverine population given that wolverines inhabit the survey area at the density indicated on the X-axis. The apparent dip of the second point from the left reflects a dramatic decrease in population viability but not detection probability for a population of three compared to one of four individuals. The bottom panel (B) shows the realized power of the survey. The grey lines indicate the probability of failing to detect at least one wolverine and the dark lines indicate the probability of failing to detect a member of a viable wolverine population given that wolverines are present in the study region at the density indicated by the X-axis. Probabilities were calculated from equation 2 and equation 4, respectively, assuming a trap effort of 1418 (solid line) or 982 (dashed line) trap-nights, and reference trap efficiency of 0.024 captures per trap-night at a reference density of 0.0098 wolverines per square km. Black arrows indicate maximum values reported in text. For raw data, please see Dataset 3.

simulated population dynamics for 25 years, corresponding to 5 to 8 wolverine generations. We estimated short-term extinction risk at each initial population size as the fraction of 10,000 replicate simulated populations that went extinct before the end of the simulation (Figure 1B).

Because published detection probabilities are tied to population densities and extinction risk is tied to population size, we had to translate population size to density in the survey area. Our PVA assumed a well-mixed population without spatial structuring. We

therefore assumed that the corresponding area allowed a juvenile at the center of the population to interact with all other animals in the population if it covered the average dispersal distance. The average distance covered by a dispersing wolverine is 51 km<sup>20</sup>, so we divided population size by 8171 km<sup>2</sup> to calculate the corresponding density.

Maximizing equation 5 using the density dependent detection and extinction risks shown in Figure 1 and Figure 2, we calculated that 1160 trap-nights would be sufficient to have a less than 5% probability of failing to detect a viable population of wolverines inhabiting

SKCNP (Figure 2A). Assuming that baited camera stations would be available for 2.5 months (75 days), this effort would require a minimum of 16 stations.

Survey methods. We placed 18 baited camera stations throughout SKCNP between January 26 and 28, 2006. Baited camera stations have been successfully used to detect wolverines outside our survey region<sup>15–17,21</sup>, and to detect other mesocarnivores in the southern Sierra Nevada<sup>12,14</sup>. We exceeded the recommended 16 stations so that camera malfunctions or other unforeseen circumstances affecting up to two stations would not reduce our survey power below 95%.

Each station consisted of a  $1.2~\mathrm{m}\times10~\mathrm{cm}\times10~\mathrm{cm}$  wooden post wrapped in barbed wire held  $0.46~\mathrm{m}$  from a tree by an aluminum frame. On top of each post was a ~0.5 m × 0.5 cm steel pole holding 1 to 5 kg of meat, bone, and hair cut from a pig carcass and wrapped in hardware cloth. We also attached a perforated can of wet cat food injected with Gusto (a combination of skunk glands and beaver castor oil; Caven's Quality Animal Lures) to each post to provide a scent lure. The bottom of each station was placed approximately 1 m above snow level so that the bait was approximately 3 m above snow level (approximately 5–6 m above ground). We mounted a heat/motion triggered camera to a second tree 1.5–5 m away from and 0.5–1 m above the bait station. We constructed camera units using PixControl<sup>TM</sup> control boards equipped with passive infrared (PIR) sensors and Sony P32 digital cameras with 256 MB memory cards. All components were mounted in waterproof plastic boxes.

We established survey stations by helicopter on 26–28 January 2006 (Table 1). To ensure adequate coverage across the parks, a  $20 \text{ km} \times 20 \text{ km}$  grid was overlaid on the park and at least one station was placed in suitable wolverine habitat in each cell (Figure 3). This represents one station per average female home range area ( $400 \text{ km}^2$ ) or two stations per male home range area ( $800 \text{ km}^2$ ) at low wolverine density. We placed additional stations in cells that contained a high number of alleged wolverine sightings reported in the 25 years prior to the survey.

The primary determinants of station locations within each grid were elevations within approximately 200 m elevation of treeline (2744–3506 m), the availability of trees for anchoring bait stations and cameras, and proximity to safe helicopter landing sites. We also considered the accessibility of locations to monitoring teams that were to check the stations and the location of reported wolverine sightings within the past 25 years. The two sites located below 2500 m were placed within 1 km of where purported wolverine tracks were photographed during snow surveys in 1979 and 1980 (T. Andrews, unpublished data) and in the Mineral King area where many unsubstantiated sightings have been reported since 1980.

Fourteen of the camera stations were checked at least once between 18 February and 1 May 2006 by Institute for Wildlife Studies personnel, California snow survey teams, or park staff. The person checking each station replaced the camera batteries and memory cards. Frequent snow storms and high avalanche risk prevented surveyors from checking four of the most remote stations until they were taken down by helicopter in early May. Stations remained active for up to 105 days. We reviewed all pictures taken from

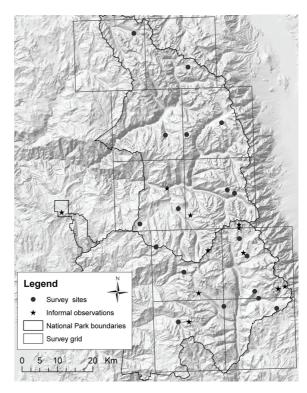


Figure 3. Map of baited survey sites with overlaying 20 km x 20 km grid. Black dots indicate locations of baited survey stations; stars indicate locations of informal wolverine sightings reported to Sequoia-Kings Canyon National Parks (courtesy NPS) since 1980. Map generated by IWS staff using ArcGIS and World Shaded Relief (ESRI 2009).

each camera after memory cards were removed from the cameras during mid-survey checks, and at the end of the survey. We recorded all animals photographed by each camera, days during which photos indicated a camera was obscured by snowfall, how many days the station was available (see results), and the date the last photo was taken.

#### Results

Camera performance. Overall, the PIR-triggered cameras worked very well under a wide range of conditions. We collected 2939 pictures during our survey excluding pictures taken while surveyors set up, maintained, and took down the stations. Pictures were taken during all hours of the day and night throughout the entire survey period. Animals were visible in 602 pictures (Figure 4); most remaining photographs were probably triggered by rapid changes in ambient temperature or by wind-blown branches. An additional 400 pictures were used to identify periods when cameras were operable. These included pictures taken during and immediately following snow events which obscured the camera lens, subsequent pictures indicating that the lens had been cleared, and pictures of surveyors approaching the camera before switching out memory cards or taking down survey stations.

In total, cameras were considered operable for 1418 survey-days from 27 January through 11 May. Individual stations were available for between 20 and 105 days (Table 1). A station was judged to be operable

Table 1. Availability of baited survey stations within Sequoia-Kings Canyon National Parks.

MK         36         26         11         118         35         21         2400         24-Apr         13         74         9-Feb           SP         36         28         32         118         16         2         3338         25-Mar         0         57         NA           RC         36         29         43         118         19         33         2955         11-May         0         104         29-Jan           CP         36         29         49         118         26         34         3161         9-May         16         86         3-Feb           GF         36         31         30         118         21         13         3234         10-May         0         103         26-Mar           TL         36         34         54         118         34         7         2839         21-Feb         0         25         7-Feb           WC         36         36         19         118         32         30         3311         10-May         0         103         28-Mar           TM         36         37         45         118         23         36         3165	Site		Latitud	le		Longitu	de	Elevation (m)	Last picture <sup>a</sup>	Days closed	Days available <sup>b</sup>	1st carnivore <sup>c</sup>
RC         36         29         43         118         19         33         2955         11-May         0         104         29-Jan           CP         36         29         49         118         26         34         3161         9-May         16         86         3-Feb           GF         36         31         30         118         21         13         3234         10-May         0         103         26-Mar           TL         36         34         54         118         34         7         2839         21-Feb         0         25         7-Feb           WC         36         36         19         118         22         30         3311         10-May         0         103         28-Mar           TM         36         37         10         118         38         4         3149         11-May         0         84         27-Mar           TC         36         37         45         118         23         36         3165         11-May         0         104         30-Jan           BD         36         40         56         118         43         43 <td< td=""><td>MK</td><td>36</td><td>26</td><td>11</td><td>118</td><td>35</td><td>21</td><td>2400</td><td>24-Apr</td><td>13</td><td>74</td><td>9-Feb</td></td<>	MK	36	26	11	118	35	21	2400	24-Apr	13	74	9-Feb
CP 36 29 49 118 26 34 3161 9-May 16 86 3-Feb  GF 36 31 30 118 21 13 3234 10-May 0 103 26-Mar  TL 36 34 54 118 34 7 2839 21-Feb 0 25 7-Feb  WC 36 36 19 118 22 30 3311 10-May 0 103 28-Mar  TM 36 37 10 118 38 4 3149 11-May 20 84 27-Mar  TC 36 37 45 118 23 36 3165 11-May 0 104 30-Jan  BD 36 40 56 118 43 43 2773 26-Apr 4 85 31-Jan  MC 36 44 42 118 37 37 2154 7-Apr 4 66 1-Feb  CL 36 46 13 118 25 4 3268 10-May 6 97 5-Mar  BL 36 46 21 118 23 52 3260 4-May 0 97 7-Mar  GB 36 56 57 118 37 18 2505 16-Feb 0 20 7-Feb  HL 36 56 58 118 35 2 3092 24-Mar 4 52 5-Feb  MI 36 57 0 118 27 14 3290 10-May 0 103 29-Mar	SP	36	28	32	118	16	2	3338	25-Mar	0	57	NA
GF 36 31 30 118 21 13 3234 10-May 0 103 26-Mar  TL 36 34 54 118 34 7 2839 21-Feb 0 25 7-Feb  WC 36 36 19 118 22 30 3311 10-May 0 103 28-Mar  TM 36 37 10 118 38 4 3149 11-May 20 84 27-Mar  TC 36 37 45 118 23 36 3165 11-May 0 104 30-Jan  BD 36 40 56 118 43 43 2773 26-Apr 4 85 31-Jan  MC 36 44 42 118 37 37 2154 7-Apr 4 66 1-Feb  CL 36 46 13 118 23 52 3260 4-May 0 97 7-Mar  GB 36 56 57 118 37 18 2505 16-Feb 0 20 7-Feb  HL 36 56 58 118 35 2 3092 24-Mar 4 52 5-Feb  MI 36 57 0 118 27 14 3290 10-May 0 103 29-Mar	RC	36	29	43	118	19	33	2955	11-May	0	104	29-Jan
TL 36 34 54 118 34 7 2839 21-Feb 0 25 7-Feb  WC 36 36 19 118 22 30 3311 10-May 0 103 28-Mar  TM 36 37 10 118 38 4 3149 11-May 20 84 27-Mar  TC 36 37 45 118 23 36 3165 11-May 0 104 30-Jan  BD 36 40 56 118 43 43 2773 26-Apr 4 85 31-Jan  MC 36 44 42 118 37 37 2154 7-Apr 4 66 1-Feb  CL 36 46 13 118 25 4 3268 10-May 6 97 5-Mar  BL 36 46 21 118 23 52 3260 4-May 0 97 7-Mar  GB 36 56 57 118 37 18 2505 16-Feb 0 20 7-Feb  HL 36 56 58 118 35 2 3092 24-Mar 4 52 5-Feb  MI 36 57 0 118 27 14 3290 10-May 0 103 29-Mar	СР	36	29	49	118	26	34	3161	9-May	16	86	3-Feb
WC       36       36       19       118       22       30       3311       10-May       0       103       28-Mar         TM       36       37       10       118       38       4       3149       11-May       20       84       27-Mar         TC       36       37       45       118       23       36       3165       11-May       0       104       30-Jan         BD       36       40       56       118       43       43       2773       26-Apr       4       85       31-Jan         MC       36       44       42       118       37       37       2154       7-Apr       4       66       1-Feb         CL       36       46       13       118       25       4       3268       10-May       6       97       5-Mar         BL       36       46       21       118       23       52       3260       4-May       0       97       7-Mar         GB       36       56       57       118       37       18       2505       16-Feb       0       20       7-Feb         HL       36       56       58 <td>GF</td> <td>36</td> <td>31</td> <td>30</td> <td>118</td> <td>21</td> <td>13</td> <td>3234</td> <td>10-May</td> <td>0</td> <td>103</td> <td>26-Mar</td>	GF	36	31	30	118	21	13	3234	10-May	0	103	26-Mar
TM 36 37 10 118 38 4 3149 11-May 20 84 27-Mar  TC 36 37 45 118 23 36 3165 11-May 0 104 30-Jan  BD 36 40 56 118 43 43 2773 26-Apr 4 85 31-Jan  MC 36 44 42 118 37 37 2154 7-Apr 4 66 1-Feb  CL 36 46 13 118 25 4 3268 10-May 6 97 5-Mar  BL 36 46 21 118 23 52 3260 4-May 0 97 7-Mar  GB 36 56 57 118 37 18 2505 16-Feb 0 20 7-Feb  HL 36 56 58 118 35 2 3092 24-Mar 4 52 5-Feb  MI 36 57 0 118 27 14 3290 10-May 0 103 29-Mar	TL	36	34	54	118	34	7	2839	21-Feb	0	25	7-Feb
TC       36       37       45       118       23       36       3165       11-May       0       104       30-Jan         BD       36       40       56       118       43       43       2773       26-Apr       4       85       31-Jan         MC       36       44       42       118       37       37       2154       7-Apr       4       66       1-Feb         CL       36       46       13       118       25       4       3268       10-May       6       97       5-Mar         BL       36       46       21       118       23       52       3260       4-May       0       97       7-Mar         GB       36       56       57       118       37       18       2505       16-Feb       0       20       7-Feb         HL       36       56       58       118       35       2       3092       24-Mar       4       52       5-Feb         MI       36       57       0       118       27       14       3290       10-May       0       103       29-Mar	WC	36	36	19	118	22	30	3311	10-May	0	103	28-Mar
BD 36 40 56 118 43 43 2773 26-Apr 4 85 31-Jan  MC 36 44 42 118 37 37 2154 7-Apr 4 66 1-Feb  CL 36 46 13 118 25 4 3268 10-May 6 97 5-Mar  BL 36 46 21 118 23 52 3260 4-May 0 97 7-Mar  GB 36 56 57 118 37 18 2505 16-Feb 0 20 7-Feb  HL 36 56 58 118 35 2 3092 24-Mar 4 52 5-Feb  MI 36 57 0 118 27 14 3290 10-May 0 103 29-Mar	TM	36	37	10	118	38	4	3149	11-May	20	84	27-Mar
MC       36       44       42       118       37       37       2154       7-Apr       4       66       1-Feb         CL       36       46       13       118       25       4       3268       10-May       6       97       5-Mar         BL       36       46       21       118       23       52       3260       4-May       0       97       7-Mar         GB       36       56       57       118       37       18       2505       16-Feb       0       20       7-Feb         HL       36       56       58       118       35       2       3092       24-Mar       4       52       5-Feb         MI       36       57       0       118       27       14       3290       10-May       0       103       29-Mar	TC	36	37	45	118	23	36	3165	11-May	0	104	30-Jan
CL       36       46       13       118       25       4       3268       10-May       6       97       5-Mar         BL       36       46       21       118       23       52       3260       4-May       0       97       7-Mar         GB       36       56       57       118       37       18       2505       16-Feb       0       20       7-Feb         HL       36       56       58       118       35       2       3092       24-Mar       4       52       5-Feb         MI       36       57       0       118       27       14       3290       10-May       0       103       29-Mar	BD	36	40	56	118	43	43	2773	26-Apr	4	85	31-Jan
BL 36 46 21 118 23 52 3260 4-May 0 97 7-Mar GB 36 56 57 118 37 18 2505 16-Feb 0 20 7-Feb HL 36 56 58 118 35 2 3092 24-Mar 4 52 5-Feb MI 36 57 0 118 27 14 3290 10-May 0 103 29-Mar	MC	36	44	42	118	37	37	2154	7-Apr	4	66	1-Feb
GB 36 56 57 118 37 18 2505 16-Feb 0 20 7-Feb HL 36 56 58 118 35 2 3092 24-Mar 4 52 5-Feb MI 36 57 0 118 27 14 3290 10-May 0 103 29-Mar	CL	36	46	13	118	25	4	3268	10-May	6	97	5-Mar
HL 36 56 58 118 35 2 3092 24-Mar 4 52 5-Feb MI 36 57 0 118 27 14 3290 10-May 0 103 29-Mar	BL	36	46	21	118	23	52	3260	4-May	0	97	7-Mar
MI 36 57 0 118 27 14 3290 10-May 0 103 29-Mar	GB	36	56	57	118	37	18	2505	16-Feb	0	20	7-Feb
	HL	36	56	58	118	35	2	3092	24-Mar	4	52	5-Feb
DB 37 5 42 118 33 9 3450 13-Apr 0 76 29-Jan	MI	36	57	0	118	27	14	3290	10-May	0	103	29-Mar
	DB	37	5	42	118	33	9	3450	13-Apr	0	76	29-Jan
EV 37 10 24 118 42 57 3012 20-Apr 1 82 23-Mar	EV	37	10	24	118	42	57	3012	20-Apr	1	82	23-Mar

<sup>&</sup>lt;sup>a</sup> Date last picture was taken.

from the day it was set up to the day the last photograph of the bait station was taken, excluding days when snow accumulation either obscured the camera lens or buried the bait station. The camera lenses at five stations were temporarily obscured by snowfall accumulating on the camera case and were judged unavailable until the time of the next unobscured photo. This period ranged from 10 minutes to four days. There were no photos taken more than six hours after the first obscured photo that remained obstructed by snow accumulation. Any station known to have been obscured for more than one hour was deemed unavailable for at least one day. Stations were also judged to be unavailable when the bait was buried by snow or was missing.

We detected no wolverines. Species detected included nine mammals and four birds. Martens (Martes americana Turton) were recorded in 419 pictures from 14 stations (Dataset 4). Other mesocarnivore species detected include coyote (Canis latran, Say), fisher (Martes pennanti Erxleben), and black bear (Ursus americanus Pallas). We also recorded northern flying squirrels (Glaucomys sabrinus Shaw), Douglas' squirrels (Tamiasciurus douglasi Bachman), golden-mantle ground squirrel (Spermophilus lateralis Say), deer mouse (Peromyscus maniculatus Wagner), white-tailed jackrabbit (Lepus townsendii Bachman), Clark's nutcrackers (Nucifraga columbiana Wilson), dark-eyed juncos

(*Junco hyemalis* Linnaeus), green-tailed towhee (*Pipilo chlorurus* Audubon), and white-crowned sparrow (*Zonotrichia leucophrys* Forster).

Post-hoc power analysis. Assuming 1418 trap-nights and the density dependent detection and persistence probabilities from Figure 1 and Figure 2, maximizing equation 4 yields an estimated probability for failing to detect a viable wolverine population of P=0.039 (Figure 2B). Because most wolverine surveys are conducted following a pre-bait period, we also calculated the power of our survey assuming that survey stations were not effective until the average date of first visitation by mesocarnivores to the 17 stations that were visited by at least one mesocarnivore (February 21). This reduced the effective survey time to 982 trap-days, increasing the estimated probability of failing to detect a viable wolverine population to P=0.064 (Figure 2B).

Viable Population Survey Data for wolverines in the Sierra Nevada, California, USA

4 Data Files

http://dx.doi.org/10.6084/m9.figshare.831467

<sup>&</sup>lt;sup>b</sup> Number of days between 27 January and date of last picture minus the number of days closed.

<sup>&</sup>lt;sup>c</sup> Date first picture of any mesocarnivore taken at the station.



Figure 4. Examples of animals photographed at baited camera stations. Panels show four carnivores identified at four stations: a) black bear (*Ursus americanus*), b) coyote (*Canis latrans*), c) fisher (*Martes pennanti*), and d) marten (*Martes americana*).

#### **Discussion**

A growing literature is devoted to inferring true occupancy from spatially replicated surveys1, and incorporating density effects on detection rates to infer extinction probabilities from temporally replicated surveys<sup>2,7</sup> and population size from detection patterns<sup>22</sup>. The framework presented here extends the application of these methods to the special but important case where replicated surveys fail to detect a species on any occasion. We accomplish this by extending the analyses to simultaneously consider the effects of population size on detection and population processes (i.e., trap efficiency and persistence probability). This extension makes two important advances to the design and interpretation of presenceabsence surveys. First, explicitly acknowledging that extremely small populations are very difficult to detect but unlikely to persist facilitates efficient survey designs with enough power to detect species which need to be managed without wasting effort on effectively extinct populations. Second, the framework presented here shifts the interpretation of a failure to detect a species from the conclusion that zero individuals remain within the survey area to the conclusion that no more than a very small number could be within the survey area.

Choosing a target population density based on the effects of small populations on both detection and persistence probabilities was critical to design of an efficient survey. If only a single wolverine inhabited the park, the survey would require 3730 trap days to have 5% failure probability. In contrast, consideration of small population effects on persistence led to the choice of target density that required less than one-third the effort, and expense, to have the same power to detect a viable population.

A consequence of choosing a target population size greater than one is that while the lack of detections during our survey does provide strong evidence that there is not a viable wolverine population within SKCNP, it does not necessarily mean that recent informal sightings are false (somebody could have seen the last remaining individual). We see this as a useful uncertainty when making management decisions. Evaluating null results in terms of a viable population rather than simple occupancy provides a basis for making management decisions that are scientifically justified regardless of the credibility of informally reported observations. When survey results indicate no viable population is present based on density dependent detection and persistence probabilities, informal sightings do not bear on estimates of future occupancy because 1) the possibility that one or a few individuals inhabit the region, and their potential to contribute to future occupancy, has been accounted for, and 2) it is generally not possible to quantify accurately the effort that could lead to an informal observation. Informal sightings often suffer the additional weakness that the probability of misidentification cannot be quantified.

Wolverines are but one of many charismatic, yet elusive species, for which informal sightings contradict current published range limits distributions inferred from formal surveys. Other notable examples include several heavily managed species such as pumas (*Puma concolor*)<sup>23</sup>, fisher (*Martes pennant*)<sup>24</sup>, Iberian lynx (*Lynx pardinus*)<sup>25</sup>, or koalas (*Phascolarctos cinereus*)<sup>26</sup>. At the core of this conflict are the different types of errors possibly committed when interpreting informal

sightings and unsuccessful formal surveys. The interpretation of a lack of verifiable evidence from formal surveys as evidence for extinction may lead to the error of accepting a false negative (i.e., the species is present but undetected). Alternatively, interpretation of informal sightings, particularly sightings unaccompanied by verifiable physical evidence, as confirmation of an extant population despite negative results from formal surveys may lead to the error of rejecting a true negative (i.e., the species was undetected by surveys because it is truly absent). The variety of classification schemes and the evidentiary value of different types of observations in the published literature highlight the range of willingness among biologists to accept one type of error over the other<sup>13,23,24</sup>. Differences in opinion about which of these errors is more egregious are not merely academic, as the allocation of limited resources often depends on the presence or absence of charismatic species<sup>24</sup>.

Appropriate application of the methods presented in this paper depends on choosing the proper time horizon to evaluate population viability. Density dependent persistence for time horizons near or below the species' generation time or for thousands of generations will rarely vary (from near 100% and 0%, respectively) and therefore contribute little to enhancing survey design or interpreting negative results. Worse, using persistence probabilities estimated for time periods beyond the management horizon may yield misleading results. Incorporating small population effects on both detection and population processes to shift analyses from present to future occupancy should only be done if the latter is pertinent to management goals. For example, when the population in question comprises the entirety of a species, even the last pair has high conservation value.

A management challenge particularly suited to the methods developed in this paper is evaluating the appropriateness of translocations as a management strategy. Translocations are powerful tools to reintroduce extirpated species or to prevent local extinction when populations decline to levels highly susceptible to demographic, environmental, and genetic stochastic events or suffer from Allee effects. However, translocations may have the unintended effects of eroding genetic diversity when relatively large numbers are introduced into small but viable populations. For example, the historic Sierra Nevada wolverine population was highly differentiated from other North American populations evidenced by alleles not known from any other population<sup>27</sup>. Uncertainty about the status of wolverines in the Sierra Nevada caused by numerous informal sightings in SKCNP weighs against translocating animals into the region as a strategy to restore healthy populations to the area, while translocations would be the only recourse if the species truly is extirpated from the region. By evaluating formal survey efforts in terms of near future occupancy, we can conclude from our negative result that even if a small number of wolverines inhabit the parks, their persistence is so tenuous that any contribution they may make to species-wide genetic diversity will most likely be lost to extinction in the absence of translocations. Consequently, translocations may be viewed as the most appropriate strategy to restore a healthy wolverine population to the southern Sierra Nevada regardless of the validity of informal sightings.

Incorporating the concurrent effects of small populations on multiple processes, such as detection and persistence probabilities, has

utility for many ecological applications. For example, fish stocking programs face the problem of preventing disease transfer to stocked populations. Disease monitoring of source populations must balance concerns about overlooking a pathogen against logistical constraints, especially because destructive sampling is often necessary for pathogen detection. The detection effort required to achieve some certainty of pathogen detection depends on pathogen prevalence<sup>28</sup>, which also determines the risk of disease transfer if pathogens are present but not detected. Disease prevalence is often unknown, and assumed at an arbitrary level<sup>29</sup>. A better strategy would be to determine sampling effort based on the product of prevalence dependent detection and transfer probabilities. Similar problems are faced in determining the success of nonnative eradication efforts. For example, optimal efforts to remove golden eagles (Aquila chrysaetos) from the California Channel Islands because they threaten endemic fox populations would account for increasing costs to find eagles and reduced impact of eagles on foxes as eagle density on the islands approaches zero<sup>30</sup>. Perhaps the most important benefit of considering the simultaneous effects of population size on detection and population processes is that doing so frames detection studies and their interpretation in terms directly related to their management application and away from the onerous task of proving something does not exist.

#### **Author contributions**

BH and DG conceived the study and collected the field data. BH conducted the analysis and prepared the manuscript. Both authors were involved in the revision of the draft manuscript and have agreed to the final content.

#### Competing interests

No competing interests were disclosed.

#### Grant information

The author(s) declared that no grants were involved in supporting this work.

#### Acknowledgments

We thank Sequoia-Kings Canyon National Parks staff, particularly D. Graber, H. Werner, and R. Mazur. We are grateful to C. Rall, who headed IWS monitoring efforts, with C. Miles and B. Czibesz, J. Nelson, and K. Thompson, and to J. King and the California Snow Surveyors who checked stations for us and D. Johnson, M. Potter, R. Rall, and G. Schmidt. Comments from K. Aubry, R. Powell, N. Haddad, and K. Pollock substantially improved this paper.

#### References

- MacKenzie DI, Royle JA, Brown JA, et al.: Occupancy estimation and modeling for rare and elusive populations. In: Thompson WL, editor. Sampling rare or elusive species: Concepts, designs, and techniques for estimating population parameters. Washington, D.C.: Island Press. 2004; pp. 149–172.
   Reference Source
- Dorazio RM: On the choice of statistical models for estimating occurrence and extinction from animal surveys. Ecology. 2007; 88(11): 2773–2782.
   PubMed Abstract | Publisher Full Text
- Peterson JT, Bayley PB: A Bayesian approach to estimating presence when a species is undetected. In: Thompson WL editor. Sampling rare or elusive species: Concepts, designs, and techniques for estimating population parameters. Washington, D.C.: Island Press. 2004; pp. 173–188.
   Reference Source
- Reed MJ: Using statistical probability to increase confidence of inferring species extinction. Conserv Biol. 1996; 10(4): 1283–1285.
   Publisher Full Text
- Choquenot D, Ruscoe WA, Murphy E: Colonisation of new areas by stoats: time to establishment and requirements for detection. N Z J Ecol. 2001; 25: 83–88.
   Reference Source
- Conn PB, Bailey LL, Sauer JR: Indexes as surrogates to abundance for low abundance species. In: Thompson WL editor. Sampling rare or elusive species: Concepts, designs, and techniques for estimating population parameters. Washington, D.C.: Island Press. 2004; pp. 59–74.

  Reference Source
- Royle JA, Nichols JD: Estimating abundance from repeated presence-absence data or point counts. Ecology. 2003; 84: 777–790.
   Publisher Full Text
- Allee WC: Animal aggregations: a study in general sociology. AMS Press. 1978.

  Reference Source
- Morris WF, Doak DF: Quantitative conservation biology: Theory and practice of population viability analysis. Sunderland: Sinauer and Associates. 2002; 480.
   Reference Source
- Graber DM: Status of terrestrial vertebrates. In: Sierra Nevada Ecosystem Project: Finalreport to Congress, vol. II, Assessments and scientific basis for management options. Davis: University of California, Centers for Walter and Wildland Resources. 1996.
   Reference Source
- 1. California Natural Diversity Database. 2012
- Reference Source

  12. Zielinski WJ, Truex RL, Schlexer FV, et al.: Historical and contemporary distributions of carnivores in forests of the Sierra Nevada, California, USA. J Biogeogr. 2005; 32(8): 1385–1407.

  Publisher Full Text

- Aubry KB, McKelvey KS, Copeland JP: Distribution and broadscale habitat relations of the wolverine in the contiguous United States. J Wildl Manage. 2007; 71(7): 2147–2158.
   Publisher Full Text
- Green R: Distribution of American marten, fisher, and other carnivores in Sequoia and Kings Canyon National Parks. Master's thesis, Humboldt State University. 2007.
- Moriarty KM, Zielinski WJ, Gonzales AG, et al.: Wolverine confirmation in California after nearly a century: Native or long-distance immigrant? Northwest Sci. 2009; 83(2): 154–162.
   Publisher Full Text
- Lofroth EC, Krebs J: The abundance and distribution of wolverines in British Columbia, Canada. J Wildl Manage. 2007; 71(7): 2159–2169.
   Publisher Full Text
- Mulders R, Boulanger J, Paetkau D: Estimation of population size for wolverines Gulo gulo at Daring Lake, Northwest Territories, using DNA based mark-recapture methods. Wildl Biol. 2007; 13(sp2): 38–51.
   Publisher Full Text
- Lacy RC: Structure of the VORTEX simulation model for population viability analysis. Ecol Bull. 2000; 48: 191–203.
   Reference Source
- Lindenmayer DB, Possingham HP, Lacy RC, et al.: How accurate are population models? Lessons from landscape-scale tests in a fragmented system. Ecol Lett. 2003; 6(1): 41–47.
   Publisher Full Text
- Vangen KM, Persson J, Landa A, et al.: Characteristics of dispersal in wolverines. Can J Zool. 2001; 79(9): 1641–1649.
   Publisher Full Text
- Banci V: Wolverine. In: Ruggiero LF, Aubry KB, Buskirk SW, Lyon LJ, Zielinski WJ, editors. The scientific basis for conserving forest carnivores: American marten, fisher, lynx and wolverine in the western United States. Fort Collins: Rocky Mountain Research Station. 1994; pp. 99–127. Reference Source
- Zhou S, Griffiths SP: Estimating abundance from detection-nondetection data for randomly distributed or aggregated elusive populations. *Ecography*. 2007; 30(4): 537–549.
   Publisher Full Text
- Clark DW, White SC, Bowers AK, et al.: A survey of recent accounts of the mountain lion (Puma concolor) in Arkansas. Southeast Nat. 2002; 1(3): 269–278. Publisher Full Taxt
- McKelvey KS, Aubry KB, Schwartz MK: Using anecdotal occurrence data for rare or elusive species: The illusion of reality and a call for evidentiary standards. *BioScience*. 2008; 58(6): 549–555.
   Reference Source

- Sarmento P, Cruz J, Monterroso P, et al.: Status survey of the critically endangered Iberian Iynx Lynx pardinus in Portugal. Eur J Wildl Res. 2009; 55(3): 247–253.
   Publisher Full Text
- Curtin A, Lunney D, Matthews A: A survey of a low-density koala population in a major reserve system, near Sydney, New South Wales. Aust Mammal. 2001; 23(2): 135–144.
   Publisher Full Text
- Schwartz MK, Aubry KB, McKelvey KS, et al.: Inferring geographic isolation of wolverines in California Using Historical DNA. J Wildl Manage. 2007; 71(7): 2170–2179.
   Publisher Full Text
- Dell'Omodarme M, Prati MC: The probability of failing in detecting an infectious disease at entry points into a country. Stat Med. 2005; 24(17): 2669–2679.
   PubMed Abstract | Publisher Full Text
- Fenichel EP, Tsao JL, Jones M, et al.: Fish pathogen screening and its influence on the likelihood of accidental pathogen introduction during fish

- translocation. J Aquat Anim Health. 2008; 20(1): 19–28. PubMed Abstract | Publisher Full Text
- Bakker VJ, Doak DF, Roemer GW, et al.: Incorporating ecological drivers and uncertainty into a demographic population viability analysis for the island fox. Ecol Monogr. 2009; 79: 77–108.
   Publisher Full Text
- Krebs J, Lofroth E, Copeland J, et al.: Synthesis of survival rates and causes of mortality in North American wolverines. J Wildlife Man. 2004; 68(3): 493–502.
   Bublisher Full Toxt
- Persson J, Landa A, Andersen R, et al.: Reproductive characteristics of female wolverines (Gulo gulo) in Scandinavia. J Mammal. 2006; 87(1): 75–79.
   Reference Source
- Carroll C, Noss RF, Schumaker NH, et al.: Is the return of the wolf, wolverine, and grizzly bear to Oregon and California biologically feasible. In: Maehr, DS, Noss RF, Larkin JL. editors. Large Mammal Restoration. Island Press, Washington. 2001; pp. 25–46.
   Reference Source

## **APPENDIX A**

#### Demographic rates input into VORTEX for PVA simulations.

Parameter	Value	Notes (parameter origin)
Age of first reproduction	3 yrs	31, 32
Maximum age of reproduction	13 yrs	31
Maximum progeny/year	4	32
Sex ratio at birth (M:F)	45:55	32ª
% Adult females breeding	50	31, 32 <sup>b</sup>
Mean offspring/female/yr	1.5	30, 31 <sup>b</sup>
Std	0.1	32°
Mortality		
juv male	0.135	31 <sup>d</sup>
juv female	0.175	31 <sup>d</sup>
subadult male	0.1775	31 <sup>d</sup>
subadult female	0.175	30 <sup>d</sup>
adult male	0.1775	31 <sup>d</sup>
adult female	0.175	31 <sup>d</sup>
% males breeding	33	Corresponds to 4 female home ranges/male home range
Initial population size	2–25	
Carrying capacity	100	33

<sup>&</sup>lt;sup>a</sup> Average of two populations (50% male in population 1, 40% male in population 2).

<sup>&</sup>lt;sup>b</sup> Persson *et al.*<sup>32</sup> report 53% of females ≥ 3 reproduce with a mean annual birth rate of 0.74 kits/female. We assumed 0.5 females breeding \* 1.5 kit/breeding female =0.75 kits/female. This assumption approximates the 0.375 female kits/female adult/year reproductive rate assumed by Krebs *et al.*<sup>31</sup>.

 $<sup>^{\</sup>circ}$  Persson et al.  $^{32}$  report litter sizes between 1–4, averaging 1.88 kits/litter and a 95% CI 1.68–2.07, corresponding to a std of 0.1.

<sup>&</sup>lt;sup>d</sup> Mortality rates were calculated as an average of mortality from natural causes of radio-tracked wolverines in montane habitats from both trapped and untrapped sites, weighted by effort in each site type<sup>31</sup>.

Simulations assuming 100% male breeding yielded same results.

# Current Referee Status:



### Referee Responses for Version 1



#### **Jonathan Rhodes**

Landscape Ecology and Conversation Group, University of Queensland, Brisbane, Qld, Australia

Not Approved: 11 March 2014

Referee Report: 11 March 2014 doi:10.5256/f1000research.2649.r3623

This paper presents an interesting idea, but I believe there are potentially some technical problems with it.

My main concern is with Equation (4). You state that P(N) in Equation (4) is the probability of failing to detect a viable population, given that the species inhabits the survey area at a given population size, N. However, since (1 - Z(N)) is the probability that the population is not viable and (1 - D(N)) is the probability that the species is not detected, then  $(1 - Z(N)) \times (1 - D(N))$  is the probability of that the population is not viable and is not detected. I suspect what you want is the probability that the population is viable and missed, which would be Z(N) x (1 – D(N)). It is also not clear to me why you don't integrate over N. If you assume that N has a Poisson distribution, then this integral should not be too hard, or at least it will be possible to approximate it (see Royle, 2004).

I think some broader discussion of this approach in relation to statistical methods for estimating detection error and estimating population parameters in the presence of detection errors would be useful. One thing that I think needs discussion is why here you focus on estimating the detection error and use that to estimating a suitable sample size. This is quite different to many recently developed approaches where the focus is on inferring the underlying population parameters from imperfect observation data. Here your focus is on the detection error itself and it is not clear why. I think some discussion of this in relation to how the data is intended to be used is important.

Finally, you talk about maximising Equation (5). As far as I understand, you are not maximising this equation, you are solving it for a value of Pcrit.

I have read this submission. I believe that I have an appropriate level of expertise to state that I do not consider it to be of an acceptable scientific standard, for reasons outlined above.

Competing Interests: No competing interests were disclosed.

# **Article Comments**

#### Comments for Version 1

**Author Response** 



#### Brian Hudgens, Institute for Wildlife Studies, USA

Posted: 07 Apr 2014

I'll start by addressing the focus of the paper, which is to determine if a species is present or absent at a regional scale. The primary application is for species that, if present, are regionally rare and difficult to detect. The critical problem facets of the problem are:

- 1. It is impractical to search for the very last individual in a region, and also not interesting to find the very last individual because if there is only one the population is already effectively extinct.
- 2. If the species is present, there is no prior information about what population size may be.
- 3. Larger populations are both more likely to be detected for a given effort and of greater interest (i.e. less likely to go extinct in the near future, and thus more likely to need management). This is an issue for ecologists when: trying to determine regional extinctions, trying to identify major range reductions, trying to determine removal of invasive species, or trying to determine pathogen presence. In the case of wolverines, public support for a reintroduction of wolverines into the Sierra Nevada depends on the historic native population being extinct. In cases of invasive species removal, it can be applied to post eradication attempt monitoring when encounter rates are not known from pre-attempt monitoring. Recent approaches using occupancy models to infer population parameters assume relatively small, replicated units being monitored with some units being occupied by the target species. Those methods do not apply to region-wide surveys without any occupied subunits, and cannot be used to infer regional absence as, without any occupied subunits, there is no information available on detection probabilities.

You are correct that Equation (4) should be  $P(N)=(1-\text{probability of extinction given N})^*(1-\text{probability of detection given N})$ . We incorrectly implied that Z(N) is the probability of persistence in the preceding paragraph, it should be the probability of extinction, and the application of equation 6 (derived from equation 4) assumes Z(N) is extinction probability.

I agree that integrating over N would preferable, but only if there is a justified probability distribution for N. Again, the Royle (2004) method is aimed for fitting data to replicated transects with nonzero count data at some sites. The methods appear to do poorly when most observed counts are zero (based on their simulation study), and do not apply when all observed counts are zero. However, if survey data from other regions where the species of interest is present in known numbers can be used to generate an expected population, or to provide a maximum population size, in the target region, one could calculate the probability of possible population sizes p[N] from a Poisson or uniform distribution respectively. Integrating p[N]\*P(N) over N=1 to (max N) would then yield a better estimate of the true probability of overlooking a viable population. While I agree this is worth further development, I expect the calculation will be sensitive to the choice of probability distribution for N, and since we did not have information for the motivating case of wolverines in the Sierra Nevada, we relied on the simpler approximation, which effectively represents the case where N is assumed to be equal to the value yielding the greatest probability of overlooking a viable population, and is thus a conservative estimate.

Lastly, while a single value of  $P_{crit}$  is assumed, because N is unknown, we do indeed maximize Equation (5) over a range of N=[1,25].

Competing Interests: No competing interests were disclosed.



\_\_\_\_\_