

# 1 Introduction

2 Infectious agents can have major impacts on animal populations through changing  
3 population dynamics and stability (Dobson 1992), altering predator prey interac-  
4 tions (Joly 2004), and even causing species decline and extinction (Viana 2015,  
5 Daszak 1999). Accurate estimates of parasite-induced host mortality (PIHM)  
6 in wild animals are important for understanding what regulates both host and  
7 parasite populations and to make predictions about disease transmission in natural  
8 systems. Although a negative impact on host fitness is a fundamental component  
9 of parasitism (Lafferty and Kuris 2002, Poulin book, other definition sources), it is  
10 notoriously difficult to quantify PIHM in wild animal populations (cite, McCallum  
11 2008?, lester 1984).

12 To conclusively identify PIHM in wild animal populations, it is necessary to  
13 experimentally infect and track host populations in ways that are rarely possible in  
14 most host-parasite systems (citations). Instead, parasitologists are often only able  
15 to collect a certain number of hosts and determine each host's parasite intensity.  
16 This snapshot, distributional data is far from the ideal type of data for addressing  
17 questions regarding PIHM, but the reality is that this is the type of data on  
18 which most questions regarding PIHM are asked (citations). The two primary  
19 questions that one may wish to ask given a snapshot host- parasite dataset are:  
20 1) Is PIHM occurring in this system? and 2) How does host survival change as  
21 parasite intensity increases? MORE

22 The first of these two questions was addressed by Crofton (1971) developed  
23 a method to test for PIHM using the truncation of the negative binomial distribu-  
24 tion. In short, the Crofton Method assumes that the distribution of parasites across  
25 hosts before mortality occurs follows a negative binomial distribution (Anderson  
26 & May 1978; Shaw *et al.* 1998). As heavily infected hosts begin to die, the negative  
27 binomial distribution gets truncated because heavily infected hosts dies and are no

longer observed in a sample. In other words, the observed host-parasite distribution and the pre-mortality host-parasite distribution will predict substantially different numbers of highly infected hosts (because those have died due to infection) but similar number of hosts with low infection loads (because those have survived). (Crofton 1971) noted that by starting with all the observed data and iteratively fitting a negative binomial distribution to hosts with lower and lower parasite loads, one could determine whether or not PIHM was occurring in the system (Figure XA). This was done graphically by determining whether parameters of the truncated negative binomial distributions showed a substantial change as hosts with smaller and smaller parasite intensities were fitted (Figure XB). The Crofton Method and this graphical technique for determining whether or not PIHM is occurring are both still used (Ferguson *et al.* 2011). A more thorough description and implementation of the Crofton Method is given in Appendix X.

The second question regarding PIHM is how host survival changes as parasite intensity increases. The Crofton Method on its own cannot answer this question and Adjei *et al.* (1986) proposed a method to determine how host survival probability changes with increasing parasite-load. The Adjei Method proceeds by first using the Crofton Method to estimate the pre-mortality parameters for a host-parasite distribution (describe what these are) and then, given these parameters, estimates a host-survival function that describes how the probability of host-survival changes with increasing parasite load (see Appendix X for a full description). From this host-survival function, the Adjei Method can estimate important host-parasite quantities such as the parasite intensity at which 50% of hosts succumb to PIHM ( $LD_{50}$ ), as well as the percent of hosts in a population succumbing to PIHM (Adjei *et al.* 1986).

Asking questions regardin

## 54 Methods

### 55 The likelihood method for estimating PIHM

56 Given the potential deficiencies of the Adjei Method, we provide an alternative  
57 approach for estimating parasite-induced host mortality (PIHM) that makes less  
58 assumption than the previously proposed Adjei Method. The likelihood method  
59 does not require any binning or alteration of the data, potentially reduces the  
60 number of parameters that need to be estimated, and allows for standard statistical  
61 techniques to be used to assess the significance of PIHM in a system.

62 The likelihood method makes the following assumption about the host-  
63 parasite system. First, as with all previously proposed methods for estimating  
64 PIHM, the likelihood method assumes that the pre-mortality distribution follows  
65 a negative binomial distribution ( $g(x)$ ) with parameter  $\mu_p$  and  $k_p$ . The validity of  
66 the assumption is an inherent problem with all the PIHM methods proposed to  
67 date and we address this thoroughly in the Discussion. The second assumption  
68 that the likelihood method makes is that the host survival function takes the form  
69 of a logistic curve given by

$$h(x|a, b) = h_x = \frac{e^{a+b \log(x)}}{1 + e^{a+b \log(x)}} \quad (1)$$

70 where  $x$  is the parasite intensity in a given host and  $a$  and  $b$  are the two  
71 parameters of the function. Generally, a larger  $a$  allows for hosts to tolerate larger  
72 parasite loads before experiencing parasite-induced mortality and a more negative  
73  $b$  leads to a more rapid decline in the probability of host survival as parasite  
74 intensity increases. The value  $\exp(a/|b|)$  (notation) is typically referred to as the  
75  $LD_{50}$ , which gives the parasite intensity at which 50% of host experience mortality.  
76 With these two explicit assumptions, the likelihood method tries to estimate the  
77 4 parameters  $\mu_p$ ,  $k_p$ ,  $a$ , and  $b$ .

78 To estimate these parameters, we need to define a probability distribution  
 79 that gives the probability of having a parasite load of  $x$  parasites conditional on  
 80 host survival. Using the standard rules of conditional probability This distribution  
 81 can be written as

$$P(x|\text{survival}) = \frac{P(\text{survival}|x) * P(x)}{P(\text{survival})} \quad (2)$$

82 One can now see that  $P(\text{survival}|x)$  is the survival function  $h(x; a, b)$ ,  
 83  $P(x)$  is the pre-mortality parasite distribution  $g(x; \mu_p, k_p)$  and  $P(\text{survival}) =$   
 84  $\sum_{x=0}^{\infty} P(\text{survival}|x) * P(x) = \sum_{x=0}^{\infty} h(x; a, b) * g(x; \mu_p, k_p)$ . Therefore equation 2 can  
 85 be written as

$$P(x|\text{survival}) = \frac{h(x; a, b) * g(x; \mu_p, k_p)}{\sum_{x=0}^{\infty} h(x; a, b) * g(x; \mu_p, k_p)} \quad (3)$$

86 Using this probability distribution, one can then find the parameters  $\mu_p$ ,  
 87  $k_p$ ,  $a$ ,  $b$  that maximize the likelihood of an observed host-parasite dataset  $\mathbf{x}$ .

88 Alternatively, one could apply the Crofton Method to estimate  $\mu_p$  and  $k_p$   
 89 and then find the maximum likelihood estimates of  $a$  and  $b$  and the corresponding  
 90  $LD_{50}$ . A final option would be to follow the example of (Ferguson *et al.* 2011) and  
 91 assume  $k_p = 1$  and only estimate  $a$ ,  $b$  and  $\mu_p$ .

## 92 0.1 Testing the ability of approaches to identify PIHM

93 To test the ability of the Adjei Method and the Likelihood Method to identify  
 94 whether or not PIHM was occurring in a system, we randomly generated data  
 95 using the following procedure. First, we drew  $N_p$  randomly infected hosts from  
 96 a negative binomial distribution with parameters  $\mu_p$  and  $k_p$ . This represented  
 97 the dataset observed before mortality. Second, we chose values of  $a$  and  $b$  and  
 98 calculated the probability of survival for all  $N_p$  hosts using equation 1. Third, we

drew  $N_p$  random numbers from a uniform distribution between 0 and 1 and if host survival probability was less than this random number, the host experienced parasite-induced mortality. The surviving hosts comprised the dataset ( $\mathbf{x}$ ) that would be obtained in the field, after PIHM.

Using both the pre-mortality and post-mortality simulated datasets, we assumed that the values of  $N_p$ ,  $\mu_p$ , and  $k_p$  were known and tested the ability of both methods to correctly determine whether or not PIHM was occurring. While this scenario is unrealistic because the parameters  $N_p$ ,  $\mu_p$ , and  $k_p$  are always unknown, we implemented this scenario as a baseline to establish the efficacy of the methods independent of the estimates of  $N_p$ ,  $\mu_p$  and  $k_p$ . For the Adjei Method,  $N_p$ ,  $\mu_p$ , and  $k_p$  are estimated using the Crofton Method, while the  $\mu_p$  and  $k_p$  in the likelihood method can be estimated jointly with  $a$  and  $b$  or via the Crofton Method. If a method could not correctly predict whether or not PIHM was occurring under these idealistic conditions, we considered this strong evidence of the unreliability of this method.

We used three different values of  $\mu_p$  (10, 100, 500) and for each  $\mu_p$  we examined three different survival functions that had graduate, moderate, and sharp decreases in host survival with increasing parasite intensity. For a given  $\mu_p$ , each survival function had the same  $LD_{50}$ , but different values of  $a$  and  $b$ . We examined each  $\mu_p$ -survival function pair at three levels of parasite aggregation,  $k_p = 0.1, 0.5, 1$ , which are realistic values of parasite aggregation in natural populations (Shaw *et al.* 1998). For each of these parameter combinations we simulated 150 datasets and tested the probability of each method incorrectly identifying PIHM in the pre-mortality dataset (Type I error) and incorrectly failing to identify PIHM in the post-mortality dataset (Type II error). For each method, we used likelihood ratio test to determine whether the full model with PIHM provided a significantly better fit than the reduced model without PIHM at significance level 0.05 (Appendix X). We tested all each parameter combinations

127 for pre-mortality population sizes of  $N_p = [300, 500, 1000, 2000, 5000, 7500, 10000]$ .

## 128 **Testing ability of PIHM approaches to recover survival function**

129 To compare the ability of the Adjei Method and the likelihood method to  
130 recover  $LD_{50}$  and the parameters  $a$  and  $b$  or the survival function, we used the  
131 same simulation procedure and parameter combinations described above. For each  
132 parameter combination we simulated 150 datasets, estimated  $a$ ,  $b$ , and  $LD_{50}$  and  
133 calculated the standardized bias and precision (Walther & Moore 2005) for these  
134 estimates over varying pre-mortality host population sizes  $N_p = [300, 500, 1000,$   
135  $2000, 5000, 7500, 10000]$ .  $N_p$  is not technically the sample size on which the  
136 methods are being tested because parasite-induced mortality reduces  $N_p$  for each  
137 simulated dataset. We therefore used the average number of surviving hosts over  
138 all 150 simulations for a given parameter combination as our measure of sample  
139 size.

## 140 **Application to data**

141 We tested the likelihood method on the datasets given in (Crofton 1971) and  
142 (Adjei *et al.* 1986) which both papers reported seeing PIHM in the respective  
143 populations.

## Results

### Detecting PIHM

### Recovery of the survival function

### Application to data

## Discussion

Parasite-induced host mortality is of substantial interest in many systems, but determining whether it is occurring given only observational data is notoriously difficult. Many of the previous methods derived to determine the effect of PIHM on a host-parasite system limit their inference to answering the yes or no question of whether or not PIHM is occurring. While a relevant question, it is often of interest to know something about the host survival function which can provide information regarding important properties of the host-parasite system, such as  $LD_{50}$  and percent of the population suffering PIHM.

We show that the Adjei Method, the only currently proposed method to estimate the host survival function and the  $LD_{50}$  from observational PIHM, has some serious methodological problems that result in biased estimates of the host survival function even under the most idealistic conditions. Interestingly, despite these flaws, the Adjei Method can still produce unbiased and precise estimates of the  $LD_{50}$  when host-parasite systems show aggregation close to  $k = 1$

To attempt to ameliorate the flaws in the Adjei Method, we proposed a more general method to determine both whether or not PIHM is occurring in a system and the shape of the survival function. We show that this method is asymptotically unbiased when estimating the host-survival function for all of the parameter space that we explored, but can produce seriously biased estimates of the host survival function for sample sizes typically observed in many host-parasite

169 studies. However, we found that the likelihood method produces unbiased and  
170 precise estimates of the  $LD_{50}$  for small, realistic sample sizes.

171 Moreover, the likelihood method...

172 What do these findings tell us about our ability to go beyond saying whether  
173 or not PIHM is occurring in a system? While we have generalized and improved  
174 upon the previously existing methods for estimating PIHM, we cannot belie the  
175 fact that estimating the host survival function from observational data alone is  
176 ladened with assumptions and difficulties.

177 The most fundamental assumption of all methods for estimating PIHM  
178 is that the shape of the pre- mortality host-parasite distribution is known. In  
179 the discrete case, this distribution is negative binomial, while in the continuous  
180 case the distribution can be gamma or exponential ((Ferguson *et al.* 2011)).  
181 While there is substantial empirical and theoretical evidence to justify the  
182 use of the negative binomial distribution as the pre-mortality distribution for  
183 macroparasites across hosts (Crofton, Shaw, Anderson and Gordon, etc), it is  
184 widely recognized that different processes can lead to a variety of distributions  
185 of parasites across hosts (Wilber, Duerr etc). However, the critical assumption of  
186 the pre-mortality distribution is not that the processes leading the pre-mortality  
187 distribution generate a negative binomial distribution, but rather that the pre-  
188 mortality distribution is well-fit by a negative binomial. The extreme flexibility of  
189 the negative binomial distribution in the discrete case or the gamma distribution  
190 in the continuous case make them reasonable candidate distributions for the  
191 pre-mortality distributions. Therefore, we do not see this assumption as central  
192 problem in any of the proposed methods.

193 However, to use the pre-mortality distribution to infer whether or not the  
194 PIHM is occurring in a system requires an explicit assumption about the host  
195 survival function and the shape of the post-mortality distribution. Regarding the  
196 host-survival function, all methods of PIHM assume that the host-survival function



is such that uninfected individuals and individuals with low parasite intensity experience essentially no PIHM. Lanciani & Boyett (1989) illustrated this by showing that when hosts experienced a linear decrease in survival probability the Crofton Method could not detect PIHM. This result, relates to the shape of the post-mortality distribution. Given that the pre-mortality distribution is well-fit by a negative binomial, a linear host-survival function will result in a post-mortality distribution that is also well-fit by a negative binomial distribution (why?). In this case, one would be unable to identify PIHM because the pre-mortality distribution

As parasitologist, we often want to get the most out of our

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