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MYCOBACTERIOSIS-ASSOCIATED MORTALITY IN WILD STRIPED BASS (MORONE SAXATILIS) FROM CHESAPEAKE BAY, USA

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Abstract. The striped bass (Morone saxatilis) is an economically and ecologically important finfish species along the Atlantic seaboard of the United States. Recent stock assessments in Chesapeake Bay (USA) indicate that non-fishing mortality in striped bass has increased since 1999, concomitant with very high (>50%) prevalence of visceral and dermal disease caused by Mycobacterium spp. Current fishery assessment models do not differentiate between disease and other components of non-fishing mortality (e.g., senescence, predation); therefore, disease impact on the striped bass population has not been established. Specific measurement of mortality associated with mycobacteriosis in wild striped bass is complicated because the disease is chronic and mortality is cryptic. Epidemiological models have been developed to estimate disease-associated mortality from cross-sectional prevalence data and have recently been generalized to represent disease processes more realistically. Here, we used this generalized approach to demonstrate disease-associated mortality in striped bass from Chesapeake Bay. To our knowledge this is the first demonstration of cryptic mortality associated with a chronic infectious disease in a wild finfish. This finding has direct implications for management and stock assessment of striped bass, as it demonstrates population-level negative impacts of a chronic disease. Additionally, this research provides a framework by which disease-associated mortality may be specifically addressed within fisheries models for resource management.

Key words: disease-associated mortality; force-of-infection; Morone saxatilis; mycobacteriosis; striped bass.

INTRODUCTION

The anadromous striped bass, Morone saxatilis, is one of four dominant piscivores in the Chesapeake Bay (USA) and fills a critical ecological niche in estuarine food webs (Hartman and Brandt 1995, Latour et al. 2008). Striped bass are also of high economic importance, forming the basis for large recreational and commercial fisheries within Chesapeake Bay (Kirkley et al. 2000). Disease outbreaks have been reported in Chesapeake Bay striped bass since the late 1980s. Mortality in fish from Maryland waters of the Bay in 1988 was associated with Streptococcus spp. (Baya et al. 1990), and visceral and dermal lesions observed in 1994 were attributed to Edwardsiella spp. (Baya et al. 1997). Beginning in 1997, striped bass exhibiting poor body condition and ulcerative skin lesions were observed in Virginia and Maryland waters of Chesapeake Bay. Histopathology revealed granulomatous inflammation associated with acid-fast bacteria, consistent with infection by Mycobacterium spp. (Vogelbein et al. 1999). Subsequent surveys have demonstrated high (>50%) prevalence of visceral and dermal mycobacteriosis in striped bass from Chesapeake Bay (Vogelbein et al. 1999, Cardinal 2001, Overton et al. 2003). Coincident with reports of high disease prevalence, recent fishery stock assessments have demonstrated an increase in natural, non-fishing mortality among striped bass in Maryland waters since 1999 (Jiang et al. 2007). These observations suggest that mycobacteriosis may be responsible for increased natural mortality of striped bass. As with most fishery assessment models, however, that of Jiang et al. (2007) does not differentiate between disease and other components of non-fishing mortality, such as predation and senescence. Therefore, the relationship between disease and increased natural mortality has not yet been established.

Mycobacteriosis is a subacute to chronic disease common in wild and captive fishes worldwide. Mortality resulting from mycobacteriosis is not generally reported in wild finfish populations, although high mortality is commonly observed in aquaculture (Nigrelli and Vogel...
Mycobacterium marinum, M. fortuitum, and M. chelonae are the most frequently cultured isolates from diseased fishes, although several other species have been reported (Lansdell et al. 1993, Tortoli et al. 1996, Herbst et al. 2001, Whippes et al. 2003, Rhodes et al. 2004). In most cases, disease is visceral, with spleen, liver, and kidney being the primary target organs. Granulomatous inflammation, often with extensive tissue destruction, is characteristic, although more poorly organized inflammatory responses are observed, typically in association with high bacillary loads (Kent et al. 2004). External clinical signs include scale loss, skin ulceration, emaciation, exophthalmia, pigmentation changes, and spinal defects (Nigrelli and Vogel 1963, Snieszko 1978, Wolke and Stroud 1978, Bruno et al. 1998). Mycobacteriosis has previously been described in wild and cultured striped bass from the U.S. Pacific Coast, with prevalence values as high as 68% and 80%, respectively (Sakanari et al. 1983, Hedrick et al. 1987).

Characterization and identification of Mycobacterium spp. associated with the epizootic of mycobacteriosis in Chesapeake Bay striped bass are ongoing. A number of isolates have been cultured, including the fish pathogen M. marinum, the newly described species M. shottsii and M. pseudoshottsii, and species typically considered to be saprophytes (e.g., M. gordonae; Rhodes et al. 2001, 2003, 2004). Other isolates do not exactly match known phenotypic profiles, and potentially represent additional new species (Rhodes et al. 2004). Mycobacterium shottsii and M. pseudoshottsii are the isolates most commonly recovered from diseased striped bass, and co-infections with multiple mycobacterial species occur (Rhodes et al. 2004). Both M. shottsii and M. pseudoshottsii are closely related to M. marinum and M. ulcerans, important pathogens of fishes and humans, respectively. IS2404, an insertion sequence associated with mycolactone, a plasmid-encoded polyketide toxin, has been detected in M. pseudoshottsii, and production of variant mycolactones has been demonstrated by this species, as well as other related fish-pathogenic mycobacteria (Rhodes et al. 2005, Ranger et al. 2006, Yip et al. 2007).

Observation of mortality among wild finfishes is typically limited to acute kill events of schooling species (e.g., Patterson 1996, Whittington et al. 1997, Blazer et al. 2002). Due to the chronic nature of mycobacteriosis in striped bass, however, mortality is cryptic, and has not been measurable previously. Several modeling approaches have been developed to estimate disease-associated mortality from cross-sectional prevalence data (Cohen 1973, Caley and Hone 2002), and have recently been generalized to allow use of complex age-varying hazards that more realistically represent disease processes (Heisey et al. 2006). In this study, we apply the approach of Heisey et al. (2006) to a comprehensive three-year survey of mycobacteriosis in Chesapeake Bay striped bass. This work was conducted using the Chesapeake Bay Multispecies Monitoring and Assessment Program (ChesMMAP), which is a fisheries-independent survey in support of ecosystem-based approaches to fisheries management. We present apparent prevalence data with respect to age, sex, and cruise (season), and examine age dependency of force-of-infection, influence of covariates on force-of-infection, and existence of disease-associated mortality.

**Materials and Methods**

**Field collections**

Striped bass were collected by the ChesMMAP survey in the main stem of Chesapeake Bay. During 2003–2005, a total of 15 cruises were conducted (annually during March, May, July, September, and November) and 80–90 sites were sampled per cruise (Fig. 1). Sampling locations were chosen according to a stratified random design, with strata based on water depth (3–9 m, 9–15 m, and >15 m) within five latitudinal 30-minute regions of the bay. The number of locations sampled in each regional stratum was randomly selected in proportion to its area. At each sampling location, a 13.7 m, four-seam balloon otter trawl (15.2 cm stretch mesh in the wings and body and 7.6 cm stretch mesh in the cod end) was towed for 20 min at ~6.5 km/h with the current. If distinct size classes (based on fork length) of striped bass were evident in the catch, length measurements were coded according to a coarse size categorization (e.g., small, medium, large). A random subsample of each size...
class (if multiple classes were present) was further processed for visual sex determination, otolith-based ageing, and histological detection of splenic mycobacteriosis.

**Laboratory procedures**

A transverse section of the right sagittal otolith was used to determine fish age following the methods of Secor et al. (1995). Annuli were counted by viewing the slide under a dissecting microscope using transmitted light (25× magnification). Three readers aged all specimens once and a randomly selected subsample of 200 specimens a second time. Year class ages were assigned based on the conventional 1 January birth date.

Spleens were fixed in Z-fix (Anatech, Battle Creek, Michigan, USA) for at least 72 hours, then divided transversely into six approximately equal portions, while removing an additional 1–2 mm section between portions to avoid examination of contiguous sections. Portions were then processed for routine paraffin histology (Prophet et al. 1992). If all six portions for an individual spleen would not fit in one cassette, portions were randomized and distributed to multiple cassettes. Sections were cut at 5 μm and stained with hematoxylin and eosin (HE). One section of up to six spleen portions and a 50-μm step section were included on each slide. Tissue sections were examined on an Olympus AX-70 light microscope (Olympus, Center Valley, Pennsylvania, USA) for the presence of granulomas. A granuloma was defined as any lesion containing epithelioid macrophages (Cotran et al. 1999). Granulomas containing helminth parasites (typically metacercariae of digenean trematodes) were not counted. HE-stained sections were examined for granulomas until (1) 12 sections (six sections + six 50-μm step sections) per fish were determined to be negative, or (2) granulomas were found.

**Statistical analyses**

**Force-of-infection.**—Various measures exist for establishing the occurrence of disease in an animal population. One such measure, cumulative incidence, is useful in that it allows calculation of the force-of-infection rate constant λ, according to the formula \( \pi(t) = 1 - e^{-\lambda t} \), where \( \pi(t) \) is cumulative incidence, or proportion of individuals that have become positive within the cohort, at time \( t \). Measurement of cumulative incidence requires constant monitoring of a cohort over time (i.e., longitudinal monitoring), so that diseased individuals that die before time \( t \) are still counted within the disease-positive proportion. Thus cumulative incidence is not influenced by population age structure or disease-associated mortality, and may be thought of roughly as the “true” measure of disease occurrence in a study group. The longitudinal monitoring requirement, however, is seldom possible in studies of wild animals, especially finfishes.

Due to this logistical constraint, apparent prevalence, or the proportion of individuals in a sample that test positive for a disease of interest, is a more commonly used measure of disease in wild animal populations. Apparent prevalence is conveniently measured in a cross-sectional (i.e., single point-in-time) study design, but may be strongly influenced by population age structure and potentially by disease-associated mortality. Assuming that animals test negative for disease at birth, time of exposure to disease agents and risk factors increases with age. Thus the apparent prevalence is directly related to the age structure of the population. Disease-associated mortality preferentially removes diseased individuals from the study population, thus biasing cross-sectional samples toward negative individuals. Many studies estimating incidence from age-prevalence data avoid this complication, known as left truncation, by assuming the disease in question to be benign. Various models have, however, been designed for human (Podgor and Leske 1986, Keiding 1991) and wildlife (Cohen 1973, Woolhouse and Chandiwana 1992, Caley and Hone 2002) epidemiology to estimate force-of-infection from age-prevalence data while taking into account disease-associated mortality. To date, these models have been able to accommodate only age-invariant or coarsely defined age-varying force-of-infection hazards. A piecewise approximation method has recently been developed, however, which allows incorporation of disease-associated mortality into apparent prevalence-based models, including complex age-varying infection hazards functions (Heisey et al. 2006). Here, we apply this method to explore disease-associated mortality of striped bass in Chesapeake Bay, USA.

In this work, we refer to “force-of-infection,” but the event we actually analyze is the onset of detectable disease (granulomatous inflammation). Indeed, in many epidemiological studies, some advanced disease state is used as a proxy for the infection event because the event itself is too subtle to be detected. Thus, in our case, the “force-of-infection” would be more accurately referred to as the “hazard of detectable disease.” For the sake of readability and consistency with previous nomenclature, however, we use the former term.

The three-state irreversible disease model (Fig. 2) characterizes the probabilities of an individual being non-diseased, \( N(t) \), diseased and alive, \( P(t) \), or dead, \( D(t) \) at age \( t \) (Heisey et al. 2006). All animals are assumed to be born disease negative so that \( N(0) = 1 \), \( P(0) = 0 \), \( D(0) = 0 \), and self cure is assumed not to occur. Transitions between states are controlled by three age-specific hazards functions: \( \lambda(t) \) is the disease hazard, or the rate at which disease-negative animals become disease-positive; \( \delta(t) \) is the rate at which disease-negative animals die; and \( \delta(t) + \mu(t) \) is the disease-positive mortality hazard, which represents the additive disease-negative and disease-positive mortality rates. Note that the disease-associated hazard, \( \mu(t) \) is assumed to be independent from background (disease-negative) mor-
tality. Disease-associated hazard $\mu(t)$ thus potentially includes mortality directly resulting from disease, as well as indirect disease-related mortality mechanisms (e.g., higher susceptibility of diseased fish to capture or predation). Cause-specific mortality is not implied by $\mu(t)$.

From the three-state model, a system of differential equations can be constructed and solved for the state probabilities $N(t)$, $P(t)$, and in turn $v(t)$, which is defined as the probability that an animal is disease positive conditional on being alive at age $t$; i.e., $Pr(P = 1 | t) = v(t)$. If $\lambda(t)$ and $\mu(t)$ are assumed to not vary with age such that $\lambda(t) = \lambda$ and $\mu(t) = \mu$, then the link function for age-specific apparent prevalence is

$$v_c(t) = \frac{1 - e^{-\lambda - \mu}t}{1 - \frac{1}{\lambda}e^{-(\lambda - \mu)t}}$$

(1)

where the subscript $c$ indicates that the hazard functions are constant across age. Eq.1 specifies the age-specific prevalence while adjusting for the potentially higher mortality rate of disease-positive animals. Note that the disease-negative mortality hazard $\delta(t)$ has no influence on $v_c(t)$ (Cohen 1972, Heisey et al. 2006).

The assumption that the infection hazard is invariant with respect to age is often unrealistic. However, relaxation of this assumption generally requires specification of a functional form for the age-specific infection hazard. Following the lead of Heisey et al. (2006), we specify the following models.

- **Weibull:** $\lambda(t) = \alpha \beta (\alpha t)^{\beta - 1}$
- **Pareto:** $\lambda(t) = \frac{\alpha \beta}{1 + \beta t}$
- **Gompertz:** $\lambda(t) = e^{\alpha + \beta t}$
- **Log-logistic:** $\lambda(t) = \frac{\alpha \beta (\alpha t)^{\beta - 1}}{[1 + (\alpha t)^\beta]}$

where for all models $\alpha$ and $\beta$ are scale and shape parameters, respectively, that control age dependency. Note that the Pareto is technically not an “age-varying” infection hazard model, but it is included with this group because its assumption of heterogeneous hazard offers flexibility to the baseline hazard function. Unlike the first three models, the log-logistic model allows the hazard to achieve a peak (i.e., it is not monotonic).

Investigation of the influence of covariates on the infection hazard is accomplished in the following manner:

$$\lambda(t) = \lambda_0(t) e^{Xb}.$$  

(6)

Here $\lambda_0$ is the baseline force-of-infection, $X$ is the design matrix, and $b$ is the vector of parameters associated with the covariates. The unit hazards ratio $e^b$ measures how much of a unit change the $i$th covariate, modeled as $e^Xb$, shifts $\lambda(t)$ up or down (Collett 2003, Heisey et al. 2006). In the case of age-invariant infection hazard, $\lambda$ is not dependent on $t$, and Eq. 6 simplifies to $\lambda = \lambda_0 e^{Xb}$.

Application of the above age-variant hazard models within the three-state irreversible disease model framework does not allow for a closed-form solution of $v(t)$. To overcome this difficulty, we use a recently developed piecewise approximation method that facilitates computation of $v(t)$ for any age-varying hazard function. Details of this method are given in Heisey et al. (2006).

The confidence intervals around $\beta$ and $\mu$ were generated by a profile likelihood approach. Models were re-fitted using fixed individual parameters $\theta$ surrounding the original estimated parameter $\hat{\theta}$. The 95% confidence interval around $\theta$ was defined as the set of $\theta$ for which the likelihood ratio statistic $G^2 = 2(L_0 - L_\theta)$ was <3.84($\chi^2_{0.05,1}$) (Venzon and Moolgavkar 1988, Heisey et al. 2006).

Application to striped bass.—We fitted eight model structures with varying parameterizations for each specified force-of-infection, either $\lambda$ or $\lambda(t)$ depending on the assumptions of age dependency. All models were fitted using the R function (R Development Core Team 2005) given in the supplement of Heisey et al. (2006). Age-1 through Age-10 fish were treated as individual age classes, while Age-11+ fish were binned. Four hundred
intervals were used to estimate $v(t)$ with piecewise methods (Heisey et al. 2006). Age-invariant hazard models included parameterization with or without the mortality term $\mu$ and age, sex, cruise (season), and/or year as covariates. The null model included only parameter(s) associated with the form of $\lambda$. Models including age as a covariate were intended to examine age dependency of force-of-infection. These models allowed $\ln \lambda$ to vary linearly ($\beta_0 + \beta_1 t$) or quadratically ($\beta_0 + \beta_1 t^2$) with age. Since $m_c(t)$ already incorporated age, including age as a covariate led to a model that did not correspond to a solution of the differential equations from the three-compartment model and thus did not have a hazards-based interpretation. However, inclusion of age as a covariate gave the model some flexibility and allowed examination of the assumption of age invariance of $k$. If the inclusion of age as a covariate significantly improved model fit, then it was likely that assumption of age invariance of the infection hazard was incorrect and models other than $m_c(t)$ needed to be considered (Heisey et al. 2006). The final age-invariant hazard model included year as a covariate in order to examine the possibility that apparent prevalence varied annually.

Models incorporating age-variant force-of-infection functions were fitted according to the piecewise approximation method of Heisey et al. (2006). A total of 32 models including combinations of covariates sex and cruise and the mortality parameter $\mu$ were fitted. Individual hazard ratios were calculated for cruises 2–5, with cruise 1 set as the reference level, and for sex, with males as the reference category. All model fits were assessed with Akaike’s Information Criterion (AIC; Burnham and Anderson 2002). Likelihood ratio testing was used to examine differences between parameterizations including $\mu$ and corresponding nested parameterizations not including $\mu$ (Woodward 2005).

RESULTS

Apparent prevalence

The analyses presented here include 1420 striped bass collected during the 2003–2005 sampling years. Of these, 519 were female and 901 were male. Ages 1–16 were represented in the sample, with the majority of fish between Ages 1 and 4 ($n = 1152$ fish).

Apparent prevalence increased by age category through Age 5 for both male and female fish, reaching 80.0% and 88.9%, respectively (Fig. 3). Prevalence in Age-6 males and females was 47.8% and 100%, respectively, although the latter value is based on a sample size of four fish. Apparent prevalence was >85% for Age-7+ males, whereas a marked reduction in apparent prevalence was observed in Age-7+ females.

Age-invariant $\lambda$ models.—The linear age covariate model did not demonstrate appreciably improved fit to the observed data relative to the constant $\lambda$ model without age covariate. Fit was markedly improved, however, in all parameterizations of the quadratic age covariate model (Table 1). The improved fit of the quadratic age covariate model indicated that assumption of age invariance of $\lambda$ within $v_c(t)$ was incorrect, and incorporation of hazard functions describing age variance of $\lambda$ was warranted. Treatment of year as a covariate did not improve fit significantly for any parameterization. This indicated that prevalence did not vary on an annual basis over the period of the study, and inclusion of year as a model covariate was contraindicated.

Age-variant models: $\lambda(t)$.—Among age-variant force-of-infection models, the log-logistic yielded consistently better model fits for all combinations of covariates and $\mu$ when compared to Weibull, Pareto, and Gompertz
Table 1. Assessment of model fits based on Akaike’s information criteria.

<table>
<thead>
<tr>
<th>Models</th>
<th>Null</th>
<th>μ</th>
<th>Sex</th>
<th>Sex, μ</th>
<th>Cruise</th>
<th>Cruise, μ</th>
<th>Cruise, Sex</th>
<th>Cruise, Sex, μ</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age-invariant λ</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Constant λ</td>
<td>97.3</td>
<td>92.8</td>
<td>78.3</td>
<td>69.8</td>
<td>43.1</td>
<td>44.4</td>
<td>30.3</td>
<td>30.5</td>
</tr>
<tr>
<td>Linear age</td>
<td>87.7</td>
<td>88.7</td>
<td>69.8</td>
<td>69.9</td>
<td>42.7</td>
<td>43.8</td>
<td>29.5</td>
<td>31.4</td>
</tr>
<tr>
<td>Quadratic age</td>
<td>80.1</td>
<td>69.8</td>
<td>61.9</td>
<td>60.6</td>
<td>36.4</td>
<td>17.1</td>
<td>23.9</td>
<td>13.5</td>
</tr>
<tr>
<td>Year</td>
<td>98.5</td>
<td>94.7</td>
<td>80.9</td>
<td>72.8</td>
<td>44.8</td>
<td>46.5</td>
<td>32.7</td>
<td>33.2</td>
</tr>
<tr>
<td>Age-variant λ</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Weibull</td>
<td>95.9</td>
<td>93.9</td>
<td>75.5</td>
<td>69.7</td>
<td>41.0</td>
<td>41.0</td>
<td>32.3</td>
<td>26.2</td>
</tr>
<tr>
<td>Gompertz</td>
<td>88.6</td>
<td>90.6</td>
<td>68.8</td>
<td>70.1</td>
<td>42.9</td>
<td>44.8</td>
<td>29.7</td>
<td>31.7</td>
</tr>
<tr>
<td>Pareto</td>
<td>90.6</td>
<td>92.2</td>
<td>70.6</td>
<td>70.7</td>
<td>43.5</td>
<td>45.5</td>
<td>30.3</td>
<td>32.0</td>
</tr>
<tr>
<td>Log-logistic</td>
<td>83.3</td>
<td>73.0</td>
<td>65.0</td>
<td>56.2‖</td>
<td>26.3</td>
<td>7.6‖</td>
<td>15.5</td>
<td>0†</td>
</tr>
</tbody>
</table>

Notes: Age-invariant models use the link function \( v(t) \). The null model includes only parameter(s) associated with the form of \( \lambda \). Age is included as a covariate in “linear” (\( \ln \lambda = b_0 + b_1t \)) and “quadratic” (\( \ln \lambda = b_0 + b_1t + b_2t^2 \)) models. Year is included as a covariate in the “year” model. Age-variant force-of-infection models use different hazard functions to describe \( \lambda(t) \) as defined in the text. Covariate parameterizations are in columns, and all fits for all models are calculated relative to the best-fit model (log-logistic, sex, cruise, μ). In all cases, significantly better fits were observed in models containing \( μ (P \leq 0.001) \). Where models have more than one parameter, they are not interactions.

† Models containing \( μ \) were compared to their respective restricted model in which \( μ = 0 \) with the likelihood ratio test.
irreversibility have been made explicitly or implicitly in previous studies of mycobacteriosis in wildlife (Anderson and Trewhella 1985, Caley and Hone 2002, Heisey et al. 2006), and it is generally assumed that mycobacteriosis in fishes is chronic, progressive, and ultimately fatal (Van Duijn 1981, Frerichs 1993, Decostere et al. 2004). Regression or resolution of mycobacteriosis in naturally or experimentally infected fishes has not, to our knowledge, been described. Long-term (45 weeks) experimentally induced mycobacterial infections in striped bass resulted in progressive granuloma formation and, in *M. marinum* infections, secondary recrudescent disease (Gauthier et al. 2003). Death of diseased female wild striped bass in Chesapeake Bay, perhaps due to energetic demands of spawning and migration, is therefore plausible and not contraindicated by available data. The ability of mycobacteria of human and veterinary importance to enter latency, however, is well known (Flynn and Chan 2001), and the possibility that fishes (striped bass in particular) may in some circumstances resolve mycobacterial granulomas remains an important consideration to the model used in this study.

Simple models of disease dynamics often assume that force-of-infection is constant through time. This assumption, however, is seldom realistic in natural systems, as force-of-infection can vary considerably with several factors, including host/pathogen density and life history traits (Scott and Smith 1994). The assumption of age-invariant $\lambda$ in modeling these data was tested with “diagnostic” models incorporating age as a covariate (Heisey et al. 2006). The improved fit of these models suggested age dependency of $\lambda(t)$, and indicated the use of more complex hazard functions describing force-of-infection in the context of $v(t)$. It is worth noting that we assumed age-invariant disease-associated mortality ($\mu$). Age-variant disease-associated mortality could have significant implications for this model, and future efforts will explore this possibility.

Model parameterizations including year as a covariate were applied assuming age-invariant $\lambda$, in order to test possible effects of annual variation in disease prevalence. The apparent constancy of annual prevalence demonstrated by these parameterizations indicates that annual prevalence of mycobacteriosis in Chesapeake Bay striped bass was stable from 2003 to 2005. Diseases with low pathogenicity and long persistence times generally tend toward endemicity (Anderson and May 1979), and it may be that striped bass mycobacteriosis would best be described as an endemic, rather than epidemic disease. An additional consideration is the recent, large fluctuation in striped bass population size. Striped bass populations along the eastern seaboard of the United States experienced a collapse in the late 1970s and 1980s, followed by a robust recovery thought to be due primarily to fishing moratoria and favorable environmental conditions leading to strong recruitment (Field 1997, Richards and Rago 1999). The level of disease prevalence currently observed may be related in part to increased host density, and may therefore have a different “stable” level when striped bass populations are lower. Unfortunately, comparable historical information about bay-wide disease prevalence is not available, so meaningful inferences about long-term disease trends are not possible. Accommodation of calendar time as well as age influences in the force-of-infection models used in this work is theoretically feasible, but will require long-term data sets to avoid confounding effects (Heisey et al. 2006).

![Force-of-infection, $\lambda(t)$, calculated from the best-fit model (log-logistic, sex, cruise, $\mu$) for (a) male and (b) female striped bass. Cruises: 1 (solid diamond), 2 (open square), 3 (solid triangle), 4 (X), and 5 (open triangle) represent March, May, July, September, and November, respectively, from 2003 to 2005.](image-url)

Age-variant models using the log-logistic hazard function demonstrated markedly improved fit over models using Weibull, Gompertz, or Pareto hazard functions. The improved fit of the log-logistic model is perhaps not surprising, as the freedom of the log-logistic hazard function to peak lends it flexibility not found in monotonic models (Collett 2003). Lower initial force-of-infection values may reflect a lag period between infection and disease expression in younger fish. Interpretation of the post-modal decline in force-of-infection with increasing age is dependent in large part on the mechanism(s) of mycobacterial transmission to...
Based on ΔAIC estimation of model fit, both cruise and sex covariates were found to influence infection hazard in the log-logistic model (Fig. 4). Female fish were found to have a lower infection hazard than male fish, and hazard by cruise increased as the year progressed. Sex-specific differences in force-of-infection are likely related to life history differences (e.g., migration), and may also be due to physiological differences such as spawning stress. Seasonality of hazard may be due to environmental stressors, such as increased summer temperatures and seasonal hypoxia found in Chesapeake Bay (Kemp et al. 2005). Region was not included as a covariate in these analyses, as the highly migratory nature of striped bass would make interpretation of regional effects on the model difficult.

Both ΔAIC model comparisons and likelihood ratio testing supported the inclusion of disease-associated mortality μ as a model parameter. Additionally, the 95% confidence interval around the estimated value of disease-associated mortality $e^{\mu}$ did not include unity. Taken together, these analyses indicated that, under the assumptions of this modeling approach, significant disease-associated mortality occurred in Chesapeake Bay striped bass. The manner or degree in which mycobacteriosis may be regulating the striped bass population, however, remains unknown. Development of predictive dynamic models of disease effects on the striped bass population is beyond the scope of this work; however, it seems plausible that the direction of bias in this case would be towards collection of diseased individuals. This would again be expected to make the analysis more conservative with respect to estimation of $\mu$; however we acknowledge that effects of disease on relative distribution and catchability of striped bass are unknown.

Disease-associated mortality was demonstrated at the level of the entire data set, which included both male and female fish. Visual inspection of the apparent prevalence data, however, gives the impression that sex-specific disease processes may be present. Moreover, the depression of apparent prevalence in older female fish begs the question of whether this group may be disproportionately driving the model-estimated mortality. Refitting the full log-logistic model to male fish alone narrowly failed to reject the null hypothesis of no mortality (likelihood ratio test, $P = 0.057$). Sensitivity analysis, however, indicated that recategorizing 5% of disease-positive male fish to disease negative was sufficient to reject the null hypothesis and support the model in which $\mu \neq 0$. These results may best be interpreted as indicating that disease-associated mortality in male fish, if it occurs, is modest.

The striped bass is an economically and ecologically important finfish along the eastern seaboard of the United States. As such, high prevalence of a potentially lethal infectious disease in this species is of considerable concern to fisheries managers and user groups. In this work, we provide the first evidence for mycobacteriosis-associated mortality in striped bass of Chesapeake Bay, the dominant breeding and nursery ground for the striped bass along the Atlantic coast. This evidence is based on a novel modeling approach to estimating force-of-infection from age-prevalence data (Heisey et al. 2006). The method we describe here depends on several assumptions, including irreversibility of disease and age-invariant disease-associated mortality. These assumptions are reasonable given the current understanding of mycobacterial pathobiology in fishes, but require testing, especially in the context of wild populations. We intend this work to drive formulation of hypotheses...
to test these assumptions, and thus to aid in development of more sophisticated models of disease dynamics based on directed field and experimental data.

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