# fishStan: Hierarchical Bayesian models for fisheries

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Summary

Fisheries managers and ecologists use statistical models to estimate population-level relations and demographic rates (e.g., length-maturity curves, growth curves, and mortality rates). These relations and rates provide insight into populations and inputs for other models. For example, growth curves may vary across lakes showing fish populations differ due to management actions or underlying environmental conditions. A fisheries manager could use this information to set lake-specific harvest limits or an ecologist could use this information to test scientific hypotheses about fish populations. The above example also demonstrates how populations exist within hierarchical structures where sub-populations may be nested within a meta-population. More generally, these hierarchical structures may be both biological (e.g., different lakes or river pools) and statistical (e.g., correlated error structures). Currently, limited options exist for fitting these hierarchical models and people seeking to use them often must program their own implementations. Furthermore, many fisheries managers and researchers may not have Bayesian programming skills, but many can use interactive languages such as R. Additionally, programs such as JAGS often require long run times (e.g., hours if not days) to fit hierarchical models and programs such as Stan can be more difficult to program because it is a compiled language. We created `fishStan` to share hierarchical models for fisheries and ecology in an easy-to-use R package.

Statement of need

We designed `fishStan` to be used by researchers, managers, and students wanting to apply hierarchical models to fisheries and ecological data (Erickson et al., 2020). An archived version of the package exists on Zenodo (10.5281/zenodo.6373008). The package enables users to take advantage of compiled languages (e.g., C++, Stan) for speed without losing the flexibility or ease-of-use of R. Specifically, `fishStan` uses RStan (Stan Development Team, 2020) to call Stan (Carpenter et al., 2017) for fisheries models including hierarchical growth models, hierarchical linear regression, hierarchical logistic regression, and a catch-curve model.

The initial model incorporated in `fishStan` was a hierarchical von Bertalanffy model presented in Midway et al. (2015) who used JAGS (Plummer, 2003) as their programming language. We included this model and other common fisheries models developed as part of ongoing research and management. These include hierarchical logistic models (e.g., maturity curves), hierarchical linear models (e.g., log-log length weight relations), and other growth curves.

We created the original package for Erickson et al. (2021b) that estimated bighead carp (*Hypophthalmichthys nobilis*) and silver carp (*H. molitrix)* demographic rates in the Illinois, Mississippi, and Ohio Rivers. This paper also includes a public, peer-reviewed dataset (Erickson et al., 2021a) and a public, peer-reviewed workflow (Erickson & Kallis, 2021). A subset of Erickson et al. (2021a) and Erickson & Kallis (2021) is included in Erickson et al. (2020) as a vignette to demonstrate an application of the package. Other applications of our Stan-based

models include the growth models in ASMFC -Atlantic States Marine Fisheries Commission (ASMFC) (2020) and Caves et al. (2020). Books such as Ogle (2016) provide an introduction and overview to the fisheries models included within the package. Books such as Gelman & Hill (2007) provide an introduction to and overview of hierarchical Bayesian modeling.

Acknowledgments

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