

# Approximate Bayesian model selection with the deviance statistic

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Bayesian model selection poses two main challenges: the specification of parameter priors for all models, and the computation of the resulting Bayes factors between models. There is now a large literature on automatic and objective parameter priors, which unburden the statistician from eliciting them manually in the absence of substantive prior information. One important class are g-priors, which were recently extended from linear to generalized linear models. To solve the computational challenge, we show that the resulting Bayes factors can conveniently and accurately be approximated by test-based Bayes factors (Johnson, 2008) using the deviance statistics of the models. For the estimation of the hyperparameter  $g$ , we show how empirical Bayes estimates correspond to shrinkage estimates from the literature, and propose a conjugate prior as a fully Bayes alternative. Connections to minimum Bayes factors are also discussed. We illustrate the methods with the development of a clinical prediction model for 30-day survival in the GUSTO-I trial, and with variable and function selection in Cox regression for the survival times of primary biliary cirrhosis patients.

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