

APPENDIX 2. WinBUGS (Bayesian inference Using Giggs Sampling) language descriptions of the hierarchical Bayesian models .

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##### Appendix 2 A. #####
##### BUGS language description of the hierarchical Bayes (HB3) model for estimating
##### trends and annual indices from the North American Breeding Bird Survey in Canada

{ # start model

##### counts, overdispersion effects, and goodness of fit stats ######

for( k in 1 : ncounts ) {
    log(lambda[k]) <- beta[strat[k]] * (year[k] - fixedyear) + obs[strat[k],obser[k]] + eta*firstyr[k] +
strata[strat[k]] + noise[k] + yeareffect[year[k],strat[k]]
    noise[k] ~ dnorm(0.0, taunoise)
    count[k] ~ dpois(lambda[k])
    fcount[k] ~ dpois(lambda[k])
    err[k] <- pow(count[k]-lambda[k],2)/lambda[k]
    ferr[k] <- pow(fcount[k]-lambda[k],2)/lambda[k]
    fzero[k] <- equals(fcount[k],0)
}
} # end of k-counts loop

nfzero <- sum(fzero[1:ncounts])
gof <- sum(err[1:ncounts])
fgof <- sum(ferr[1:ncounts])
diffgof <- gof-fgof
posdiff <- step(diffgof)

maxf <- ranked(fcount[1:ncounts],ncounts)
meanf <- mean(fcount[1:ncounts])

taunoise ~ dgamma(0.001,0.001)
sdnoise <- 1 / pow(taunoise, 0.5)

##### observer effect hyperparameters ######

mulogtauobs ~ dnorm(0.0,1.0E-6)
taulogtauobs ~ dgamma(0.001,0.001)

##### first-year observer effect ######

eta ~ dnorm( 0.0,1.0E-6)

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##### stratum-level effects #####
for( s in 1 : nstrata ) {

    beta[s] ~ dnorm( 0.0,1.0E-6) # long-term trend in stratum
    expbeta[s] <- exp(beta[s])
    strata[s] ~ dnorm( 0.0,1.0E-6) # stratum average abundance
    expstrata[s] <- exp(strata[s])
    overdisp[s] <- 1 + 1/(expstrata[s]*taunoise)
    tauyear[s] ~ dgamma(0.001,0.001) # precision of year-effects in stratum
    sdyear[s] <- 1 / pow(tauyear[s],0.5)

##### observer effects #####
for( i in 1 : nobservers[s] ) {
    obs[s,i] ~ dnorm( 0.0,tauobs[s])
} # end i-observer loop

log(tauobs[s]) <- logtauobs[s]
logtauobs[s] ~ dnorm(mulogtauobs,taulogtauobs)
sdobs[s] <- 1 / pow(tauobs[s], 0.5)

##### year effects #####
for( y in ymin : ymax ) {
    yeareffect[y,s] ~ dnorm( 0.0, tauyear[s])
} # end y-year loop

} # end of s-strata loop

##### summary statistics - stratum level annual indices = n #####
##### n[,] values are combined across strata, weighted by stratum-area outside of BUGS #####
##### after saving the full posterior distributions of n[,] #####
for( i in 1 : nstrata ) {
    for( t in ymin : ymax ) {
        n[i,t] <- nonzeroweight[i]*exp(strata[i]+beta[i]*(t-fixedyear)+yeareffect[t,i]+ 0.5*sdnoise*sdnoise+
        0.5*sdobs[i]*sdobs[i])
    } # end t-year loop
} # end i-strata loop

} # end model

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##### Appendix 2 B. #####
##### BUGS language description of the trend comparison model #####
##### model for comparing trend estimates from two (indexed by e) different analyses, for species
##### (indexed by s)
##### for the HB3-ML comparison, e[1] are estimates from the HB3 model and e[2] from the ML model
##### the trend comparison model is modified from the grassland bird model from Link and Barker 2010,
##### pg 90, also described in Sauer and Link 2002, Ecology 83:1743-1751
##### in essence, this is a paired t-test style comparison, that accounts for the imprecision in each s*e
##### trend estimate.

##### input data compiled in R, consist of nspecies (the number of species), as well as 3 matrices: varhat,
##### betahat, and n, each of which has nspecies rows and 2 columns
## varhat = estimates of the s*e variances of trends
## betahat = estimates of the s*e trends
## n = sample size of the s*e trends (for the HB3-ML comparison, this is the number of routes used to
##### generate the trend)

model{

for (e in 1:2) {

for(s in 1:nSpecies) {

varhat[s,e] ~ dgamma(p[s,e],lam[s,e])
p[s,e] <- n[s,e] / 2
lam[s,e] <- p[s,e] * tau.betahat[s,e]
tau.betahat[s,e] ~ dgamma(0.001,0.001)
sd.betahat[s,e] <- 1/sqrt(tau.betahat[s,e])
betahat[s,e] ~ dnorm(beta[s,e],tau.betahat[s,e])
beta[s,e] ~ dnorm(0.0,1.0E-6)
pos[s,e] <- step(beta[s,e])
neg[s,e] <- 1-pos[s,e]

} # end of first s-species loop
numpos[e] <- sum(pos[,e])
numneg[e] <- sum(neg[,e])

} #end of e loop (indexing two models being compared)

for(s in 1:nSpecies) {
dif[s] <- beta[s,1]-beta[s,2] # dif is a vector of the species-specific trend differences after accounting for
# the imprecision of each estimate's trend and the group (survey/monitoring program) structure
difvar[s] <- sd.betahat[s,1]-sd.betahat[s,2] # difvar is a vector of the species-specific differences in sd of
# the trends
posdifvar[s] <- step(difvar[s])
}
}
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} # end of second s-species loop

p.HBimp <- (nspecies-sum(posdifvar[]))/nspecies

m.dif <- mean(dif[])
m.difvar <- mean(difvar[])

dif.numpos <- numpos[1]-numpos[2]

} # end of model
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