IDM_project

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The following are used for the simulation: We assume n.genus q = 1, 2, ..., n.genus. * For each g genera, we assume n.species in that geunus * We assume the data was collected at n.sites * We also assume the total number of visits to each site is n.visit. * We also assume that some of the data are collected with species ID * Most of them are collected with genus ID * shan.index is the shannon's index * p.tag is the detection probability for the occurrence model (species data) * includecov (T/F) indicates whether it is a model with a covariate (TRUE) or no covariate (FALSE)

Simulating the data

```
set.seed(2020)
includecov= FALSE
#includecov=TRUE means that there is covariate effect and false otherwise
n.sites=20
n.species = 5
n.visit=3
covte = covariates(n.sites)
data = sim(n.sites, n.species, n.visit, n.sites, n.sites, cov=covte, includecov) #covte is the covariates
summary(data)
##
               Length Class Mode
## mat.species 300
                      -none- numeric
## mat.genus
                60
```

Original Parameters

60

20

covariates

shan.index

-none- numeric

-none- numeric

-none- numeric

$$\sigma_{\alpha}^{2} = 0.4^{2}$$

$$\sigma_{species}^{2} = 2$$

$$\sigma_{site}^{2} = 1$$

$$p.tag = 0.7$$

Running the model

```
mcmc.out <- est(data, code)</pre>
## defining model...
## building model...
```

```
## setting data and initial values...
## Warning in model$setData(data): data not used in model: cov
## running calculate on model (any error reports that follow may simply reflect
## missing values in model variables) ...
## checking model sizes and dimensions... This model is not fully initialized. This is not an error. To
## model building finished.
## compiling... this may take a minute. Use 'showCompilerOutput = TRUE' to see C++ compilation details.
## compilation finished.
## ===== Monitors =====
## thin = 1: p.tag, shan.index, sigma.alpha, sigma.site, sigma.species
## ===== Samplers =====
## RW sampler (30)
##
    - sigma.site
##
   - sigma.species
##
    - sigma.alpha
##
    - p.tag
##
    - beta.site[] (20 elements)
##
    - beta.species[] (5 elements)
##
    - alpha
## binary sampler (100)
    - z[] (100 elements)
## Monitored nodes are valid for WAIC
## compiling... this may take a minute. Use 'showCompilerOutput = TRUE' to see C++ compilation details.
## compilation finished.
## runMCMC's handling of nburnin changed in nimble version 0.6-11. Previously, nburnin samples were dis
## running chain 1...
## |-----|-----|
## running chain 2...
## |-----|
## |-----|
## WAIC was calculated as NaN. You may need to add monitors to model latent states, in order for a val
#Diagnostics of the model
coda_samples <- mcmc(mcmc.out$mcmc.out$samples)</pre>
#mcmcplot(coda_samples)
```

Parameter Estimation

- Mean bias of shannon index
- MSE of shannon index
- Estimate of σ_{α} , σ_{β} , σ_{γ} and p.tag

```
est_shan = shanest(est_beta,data$covariates, TRUE) #estimation of shannon index
  mse_shan = sum(((true_shan-est_shan)^2)/n.sites) #MSE of shannon index
  mbias_shan = mean(true_shan-est_shan) #mean bias of shannon index
  p_est <- mcmc.out$mode.out[(n.species*3)+1] #detection probability</pre>
  sig_alpha <- NA
  sig_beta <- NA
  sig_gamma <- NA
  }else{
   est_shan <- mcmc.out$mode.out[2:(n.sites+1)] #estimates of shannon index
   mse_shan = sum(((true_shan-est_shan)^2)/n.sites) #MSE of shannon index
   mbias_shan = mean(true_shan-est_shan)#mean bias of shannon index
  p_est <- mcmc.out$mode.out[1] #detection probability</pre>
  sig_alpha <- tail(mcmc.out$mode.out,3)[1] #variance of intercept</pre>
  sig_beta <- tail(mcmc.out$mode.out,3)[2] #variance of site effect</pre>
  sig_gamma <- tail(mcmc.out$mode.out,3)[3] #variance of species effect
  }
data = data.frame(p_est, mse_shan, mbias_shan, sig_alpha, sig_beta, sig_gamma)
print(data)
```

p_est mse_shan mbias_shan sig_alpha sig_beta sig_gamma ## p.tag 0.7098417 0.008709304 -0.09332365 0.1455323 0.6132096 0.3414869