## Bakeoff Pipeline

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Bakeoff is a pipeline for fitting various single and joint species distribution models to community data. The main script used is 'bakeoff.pipeline.r', which guides through the importing and formatting the data, fitting the models, producing predictions based on validation data.

All the scripts sourced, the data used and the folder structure are available in Github: AnnaNorb/bakeoff. For the pipeline to run smoothly, the required folder structure is created as a part of the pipeline to the location indicated by the user. Please make sure, that the location of the pipeline (where it is downloaded), is also a location where additional results folders can be created and results saved in the end.

The pipeline works in R environment, except for the method HMSC, for which all models are fitted and predictions made in Matlab (see the document 'bakeoffHMSCvignette'). The workflow is the following: 1) first we construct the folder structure needed, 2) then fit the HMSC models in Matlab, 3) after which we fit all models in R, 4) and finally produce all predictions and performance measures.

## 1 Preparations

OS<-get\_os()

```
First clean the workspace.
rm(list = ls(all=TRUE)); gc()
Then identify and indicate your operating system.
get_os <- function() {
   if (.Platform$OS.type == "windows") {
      "win"
   } else if (Sys.info()["sysname"] == "Darwin") {
      "osx"
   } else if (.Platform$OS.type == "unix") {
      "unix"
   } else {
      stop("Unknown OS")
   }
}</pre>
```

Define the path for the location of the 'bakeoff' folder, which defines the paths to all files of the pipeline, including the script 'settings.r', which and runs a series of setting.

```
pth<-"~/OneDrive - University of Helsinki/"
SETT<-paste(pth,"bakeoff/pipeline/SCRIPTS/settings.r",sep="")
source(SETT)
setwd(WD)</pre>
```

Next we will create the results folders unless they already exist, in which case the following lines can be ignored.

```
dir.create(PD2)
dir.create(RD)
for (set in Sets) {
    dir.create(paste(RD,set,sep=""))
    }
```

```
dir.create(RDfinal)
dir.create(paste(RDfinal,"150/",sep=""))
dir.create(paste(RDfinal,"300/",sep=""))
```

At this point, the user should switch to fitting the HMSC models: open the 'bakeoffHMSC vignette' and follow the workflow until you have successfully produced the fits and predictions.

After returning to the R pipeline, the user should check whether she has the required packages installed already and install the ones missing. With the command below, all the required packages are downloaded, along with their dependencies.

The 'mvpart' package is available from the CRAN archive (https://cran.r-project.org/src/contrib/Archive/mvpart/), and it can be downloaded from there, but it is included in the pipeline, and can hence be installed using the lines below.

Species archetypes models are provided as a source code along with the pipeline (".../MODELS/coordinSAMsv2.R"). The code is sourced within the model fitting proedure.

Then select the number of cores ('crs') you wish to use in parallel computing and set the clusters. The default is two cores.

```
crs<-2
if (OS=="osx") {
    require(doParallel)
    registerDoParallel(cl=makeCluster(crs))
}
if (OS=="win") {
    require(doMC)
    registerDoMC(cores=crs)
}</pre>
```

## 2 Model fitting and predictions

The provided data sets have been divided into training and validation beforehand (see the main text of Norberg et al. for details). The set of 300 sampling units is randomly sampled from the full set and the set of 150 is randomly sampled from this set.

Then the user has to choose whether to fit the Bayesian methods with shorter or longer chains. The default ('FALSE') means shorter chains (max. MCMC 50 000 iterations and/or 24h fitting time), and 'TRUE' mean double the default.

MCMC2<-FALSE

Then we can fit the models and make predictions. Within the model fitting scripts, also computation times are calculated and they are saved in the results directory.

```
for (sz in 1:2) {
    for (d in 1:length(Sets)) {
        set_no <- Sets[d]</pre>
        source(readdata)
        # fit
        #############
        source(paste(MD, "fit.glm.r", sep=""))
                                                      # ssGLM 1
        source(paste(MD, "fit.glmmPQL.r", sep=""))
                                                      # ssGLM 2
        source(paste(MD, "fit.gam.r", sep=""))
                                                      # GAM
        source(paste(MD, "fit.mrt.r", sep=""))
                                                      # MRTs
        source(paste(MD, "fit.rf.r", sep=""))
                                                      # RFs
        source(paste(MD, "fit.brt.r", sep=""))
                                                      # BRT
        source(paste(MD, "fit.svm.r", sep=""))
                                                      # SVM
        source(paste(MD, "fit.gnn.r", sep=""))
                                                      # GNNs
        source(paste(MD, "fit.mars.r", sep=""))
                                                      # MARS
        source(paste(MD, "fit.gjam.r", sep=""))
                                                      # GJAMS
        source(paste(MD, "fit.mistnet.r", sep=""))
                                                      # mistnet
        source(paste(MD, "fit.manyglm.r", sep=""))
                                                      # MVABUND
        source(paste(MD, "fit.bc.r", sep=""))
                                                      # BayesComm
        source(paste(MD, "fit.sam.r", sep=""))
                                                      # SAMs
        source(paste(MD, "fit.boral.r", sep=""))
                                                      # BORAL
        # predict
        ####################
        # ss GLM 1
        rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
        source(paste(PD, "predict.glm.r", sep=""))
        # ss GLM 2
        rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
        source(paste(PD, "predict.glmmPQL.r", sep=""))
        rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
        source(paste(PD, "predict.brt.r", sep=""))
        rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
        source(paste(PD, "predict.svm.r", sep=""))
        rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
        source(paste(PD, "predict.gnn.r", sep=""))
        rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
        source(paste(PD, "predict.mars.r", sep=""))
        rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
        source(paste(PD, "predict.manyglm.r", sep=""))
        # GAM
```

```
rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
    source(paste(PD, "predict.gam.r", sep=""))
    # RFs
    rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
    source(paste(PD, "predict.rf.r", sep=""))
    rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
    source(paste(PD, "predict.mrt.r", sep=""))
    rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
    source(paste(PD, "predict.gjam.r", sep=""))
    # SAMs
    rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
    source(paste(PD, "predict.sam.r", sep=""))
    # mistnet
    rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
    source(paste(PD, "predict.mistnet.r", sep=""))
    # BayesComm
    rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
    source(paste(PD, "predict.bc.r", sep=""))
    rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
    source(paste(PD, "predict.boral.r", sep=""))
    rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
    source(paste(PD, "predict.hmsc.ss.r", sep=""))
    # HMSC
    rm(list=ls()[!(ls() %in% saveobjs)]); gc(); source(SETT); source(readdata)
    source(paste(PD, "predict.hmsc.all.r", sep=""))
}
```

## 3 Measures for evaluating the predictive performance of the models

}

First we calculate species-specific occurrence probabilities, species richnesses and beta index values.

```
}
             for (d in 1:length(Sets)) {
                 set no <- Sets[d]; source(readdata)</pre>
                 source(paste(modpredsFolder,"beta_inds.r",sep="/"))
             }
        }
    }
}
Then we produce the measures for evaluating the predictive performance of the modelling frameworks and their variants.
rm(list=ls()[!(ls() %in% saveobjs2)]); gc(); source(SETT); setwd(WD)
ENS<-list(NULL,</pre>
        unlist(mod_names),
         c("HMSC3", "MARS1", "BRT1", "MISTN1", "GNN1"))
PRV<-list(NA, 0.1)
for (e in 1:length(ENS)) {
    for (p in 1:length(PRV)) {
        opts<-list(modelEnsemble=ENS[[e]],prevaleceThreshold=PRV[[p]])</pre>
        for (sz in 1:2) {
             PMS<-NA
             PMs<-list()
             for (d in 1:length(Sets)) {
                 source(SETT); set no <- Sets[d]; source(readdata)</pre>
                 source(paste(RD,"expct.r",sep=""))
                 source(paste(RD, "aucs.r", sep=""))
                 source(paste(RD, "sqrtPrs.r", sep=""))
                 source(paste(RD, "probBinRMSE.r", sep=""))
                 source(paste(RD, "mse.r", sep=""))
                 source(paste(RD, "spearm.r", sep=""))
                 source(paste(RD, "sds.r", sep=""))
                 source(paste(RD, "p50.r", sep=""))
                 pms<-simplify2array(PMs)
                 PMS<-rbind(PMS,pms)
             }
             PMS \leftarrow PMS[-1,]
             filebody <- paste (RDfinal, dataN[sz], "/meta_analysis/PMS", sep="")
             if (is.numeric(prevThrs)) {
                 filebody<-paste(filebody,"_spThr",prevThrs*100,sep="")</pre>
             }
             if (is.null(ensmblModels)!=TRUE){
                 if(length(ensmblModels) == nmodels){
                      ensmbl<-"all"
                 }
                 if(length(ensmblModels)!=nmodels){
                      ensmbl<-paste(ensmblModels,collapse="_")</pre>
                 filebody<-paste(filebody,"_ensmbl_",ensmbl,sep="")</pre>
             }
             save(PMS, file=paste(filebody,".RData",sep=""))
```

```
}
    }
}
And then we compile all the results into one table.
rm(list=ls()[!(ls() %in% saveobjs2)]); gc(); source(SETT); setwd(WD)
TBL_PMS_ALL<-NA
for (e in 1:length(ENS)) {
    for (p in 1:length(PRV)) {
        opts<-list(modelEnsemble=ENS[[e]],prevaleceThreshold=PRV[[p]])
        TBL PMS all<-NA
        for (sz in 1:2) {
            filebody<-paste(RDfinal,dataN[sz],"/meta_analysis/PMS",sep="")</pre>
            if (is.numeric(opts$prevaleceThreshold)) {
                 filebody<-paste(filebody,"_spThr",opts$prevaleceThreshold*100,sep="")
            }
            if (is.null(opts$modelEnsemble)!=TRUE) {
                 if(length(opts$modelEnsemble) == nmodels){
                     ensmbl<-"all"
                 if(length(opts$modelEnsemble)!=nmodels){
                     ensmbl<-paste(opts$modelEnsemble,collapse="_")</pre>
                 filebody<-paste(filebody,"_ensmbl_",ensmbl,sep="")</pre>
            load(file=paste(filebody, ".RData", sep=""))
            noMod<-dim(PMS)[1]/(3*length(Sets))
            if (is.null(opts$modelEnsemble)!=TRUE) {
                 tblrep<-matrix(rep(c(paste("ENS",ensmbl,sep="_"),"ensemble",
                         1*(colSums(feats[ENS[[e]],3:ncol(feats)])>0)),
                         each=(length(Sets)*3)),nrow=length(Sets)*3)
                 colnames(tblrep)<-colnames(feats)</pre>
                 TBL<-cbind(tblrep,dataN[sz],rep(rep(c("i","e1","e2"),
                     each=noMod),length(Sets)),rep(Sets,each=(noMod*3)))
            } else {
                 tblrep<-apply(feats,2,rep,times=3*length(Sets))
                 TBL<-cbind(tblrep,dataN[sz],rep(rep(c("i","e1","e2"),each=noMod),
                     length(Sets)),rep(Sets,each=(noMod*3)))
            }
            colnames(TBL)[(ncol(TBL)-2):ncol(TBL)]<-c("dataSize", "predType", "dataSet")
            TBL_PMS<-cbind(TBL,PMS,PRV[[p]])
            colnames(TBL_PMS)[ncol(TBL_PMS)]<-"prevThr"</pre>
            #print(paste("Number of NAs",sum(is.na(TBL_PMS)),"for data size",dataN[sz]))
            TBL PMS all <- rbind (TBL PMS all, TBL PMS)
        TBL_PMS_all<-TBL_PMS_all[-1,]</pre>
        filebody<-paste(RDfinal,dataN[sz],"/meta_analysis/finalTBLall",sep="")</pre>
        if (is.numeric(opts$prevaleceThreshold)) {
            filebody<-paste(filebody,"_spThr",opts$prevaleceThreshold*100,sep="")
```

```
}
        if (is.null(opts$modelEnsemble)!=TRUE) {
             if(length(opts$modelEnsemble) == nmodels){
                 ensmbl<-"all"
             }
             if(length(opts$modelEnsemble)!=nmodels){
                 ensmbl<-paste(opts$modelEnsemble,collapse=" ")</pre>
             }
            filebody<-paste(filebody, "_ensmbl ",ensmbl,sep="")</pre>
        }
        save(TBL_PMS_all, file=paste(filebody,".RData",sep=""))
        TBL_PMS_ALL<-rbind(TBL_PMS_ALL,TBL_PMS_all)</pre>
    }
}
TBL_PMS_ALL<-TBL_PMS_ALL[-1,]
rownames(TBL PMS ALL) <- NULL
save(TBL_PMS_ALL, file=paste(RDfinal, "TBL_PMS_ALL.RData", sep=""))
write.table(TBL PMS ALL,file=paste(RDfinal,"TBL PMS ALL.csv",sep=""),
             sep=",",row.names=F,col.names=T)
Let's plot the results. The results are otherwise raw, except that we'll take the reciprocal of some of the measures, so
that a bigger number is always better for all the measures.
rm(list=ls()[!(ls() %in% c("OS","pth","SETT"))]); gc()
source(SETT)
setwd(WD)
load(file=paste(RDfinal, "TBL_PMS_ALL.RData", sep=""))
resTBL<-data.frame(TBL PMS ALL)</pre>
resTBL[,c(3:8,11:(ncol(resTBL)-1))]<-apply(resTBL[,c(3:8,11:(ncol(resTBL)-1))],2,as.numeric)
pms<-colnames(resTBL[,11:(ncol(resTBL)-1)])</pre>
pms < -pms[c(1,2,4,3,
        5,9,17,13,
        6,10,18,14,
        7,11,19,15,
        8,12,20,16)]
resTBL[,minTomaxBest] <- -resTBL[,minTomaxBest]</pre>
pdf(file=paste(RDfinal,"raw res fig.pdf",sep=""),bg="transparent",width=15,height=15)
    par(family="serif", mfrow=c(5,4), mar=c(7,3,2,1))
    for (p in 1:length(pms)) {
        plot(0,0,xlim=c(0,length(mod_names2)),ylim=c(min(resTBL[,pms[p]]),max(resTBL[,pms[p]])),
             type='n',xaxt='n',xlab="",ylab="",main=pms[p])
        for (m in 1:length(mod names)) {
             lines(resTBL[which(resTBL$Abbreviation==mod_names[[m]]),pms[p]],
                 x=rep(m,times=length(resTBL[which(resTBL$Abbreviation==mod names[[m]]),pms[p]])),
                 lwd=2)
            points(mean(resTBL[which(resTBL$Abbreviation==mod names[[m]]),pms[p]]),
                 x=m,pch=21,col="black",bg="red3",cex=2)
        }
        axis(1,1:length(mod_names),unlist(mod_names),las=2)
dev.off()
```

```
Finally, let's have a look at the computation times. First compile the results.
rm(list=ls()[!(ls() %in% c("OS","pth","SETT"))]); gc(); source(SETT); setwd(WD)
require(rmatio)
require(R.matlab)
for (sz in 1:2) {
    for (d in 1:length(Sets)) {
        set_no <- Sets[d]
        comtimes<-list()</pre>
        for (m in 1:length(mod_names)) {
             if (m \le 20) {
                 load(file=paste(FD,set no,"/comTimes ",mod names[m]," ",dataN[sz],
                          ".RData", sep=""))
                 comtimes[[m]]<-as.numeric(comTimes, units = "mins")</pre>
             }
             if (m > 20 \& m \le 22) {
                 tmp<-read.mat(filename=paste(FD,set_no,"/ssHMSC/compTime_",Sets[d],"_",</pre>
                              mod_names[m],"_",dataN[sz],".mat",sep=""))
                 comtimes[[m]]<-as.numeric(tmp)/60</pre>
             }
             if (m > 22) {
                 tmp<-read.mat(filename=paste(FD,set_no,"/compTime_hmsc_",Sets[d],"_",</pre>
                          m-22, "_", dataN[sz], ".mat", sep=""))
                 comtimes[[m]]<-as.numeric(tmp)/60</pre>
             }
        }
        save(comtimes, file=paste(RD2,set no,"/comtimes ",dataN[sz],".RData",sep=""))
    }
}
And the plot them.
rm(list=ls()[!(ls() %in% c("OS","pth","SETT"))]); gc(); source(SETT); setwd(WD)
# define the ensemble
ens<-c("HMSC3", "BC2", "BRT1", "MISTN1", "GNN1")
ComTimesAll<-NA
for (sz in 1:2) {
    set_no <- Sets[1]</pre>
    load(file=paste(RD2,set_no,"/comtimes_",dataN[sz],".RData",sep=""))
    ComTimes<-unlist(comtimes)</pre>
    for (d in 2:length(Sets)) {
        set no <- Sets[d]</pre>
        load(file=paste(RD2,set no,"/comtimes ",dataN[sz],".RData",sep=""))
        ComTimes<-cbind(ComTimes,unlist(comtimes))</pre>
    }
    ComTimes<-cbind(mod names,ComTimes,rep(dataN[sz],times=nrow(ComTimes)))</pre>
    colnames(ComTimes)<-c("Variant",Sets,"dataSize")</pre>
    ComTimesAll<-rbind(ComTimesAll,ComTimes)</pre>
}
ComTimesAll<-ComTimesAll[-1,]</pre>
```

```
comtimes<-ComTimesAll[,2:7]
comtimes<-matrix(unlist(comtimes),ncol=6)</pre>
rownames(comtimes) <- ComTimesAll[,1]
colnames(comtimes)<-c(Sets, "size")</pre>
modNams<-rownames(comtimes[comtimes[,ncol(comtimes)]=="150",])</pre>
comtimes150<-apply(comtimes[comtimes[,ncol(comtimes)] == "150",],2,as.numeric)</pre>
comtimes300<-apply(comtimes[comtimes[,ncol(comtimes)] == "300",],2,as.numeric)</pre>
# ensembles
comtimes150 <- rbind(comtimes150, c(colSums(comtimes150[,-ncol(comtimes150)]),150))</pre>
rownames(comtimes150)<-c(modNams, "ENS_all")</pre>
comtimes300 <- rbind(comtimes300, c(colSums(comtimes300[,-ncol(comtimes300)]),300))</pre>
rownames(comtimes300)<-c(modNams, "ENS_all")</pre>
comtimes150 <- rbind(comtimes150,c(colSums(comtimes150[ens,-ncol(comtimes150)]),150))</pre>
rownames(comtimes150)[nrow(comtimes150)] <- paste(c("ENS", ens), collapse=" ")
comtimes300 <- rbind(comtimes300,c(colSums(comtimes300[ens,-ncol(comtimes300)]),300))</pre>
rownames(comtimes300)[nrow(comtimes300)] <- paste(c("ENS",ens),collapse=" ")
comtimes<-rbind(comtimes150,comtimes300)</pre>
comtimes<-cbind(rownames(comtimes),comtimes)</pre>
write.table(comtimes,file=paste(RDfinal,"comtimes.csv",sep=""),
            row.names=FALSE,col.names=TRUE,sep=",", quote=FALSE)
comTimes<-apply(comtimes[,-1],2,as.numeric)</pre>
rownames(comTimes)<-rownames(comtimes)</pre>
pdf(file=paste(RDfinal,"/comptimes.pdf",sep=""),bg="transparent",width=10,height=7)
    par(family="serif",mar=c(18,4,4,1))
    plot(log10(apply(comTimes[which(comTimes[,ncol(comTimes)] == 300), -ncol(comTimes)],1,max)),
        col="black",pch=17, xlab="",ylab="",yaxt="n",xaxt="n",
        ylim=c(log10(0.0004),log10(4000)), main="Computation times")
    points(log10(apply(comTimes[which(comTimes[,ncol(comTimes)]==150),-ncol(comTimes)],1,min)))
    axis(side=2,at=log10(c(0.0005,0.0167,1,10,60,1440,3900)),
        labels=c("0.03 sec","1 sec","1 min","10 min","1 h","24 h","65 h"),las=2)
    axis(side=1,at=1:nrow(comTimes[which(comTimes[,ncol(comTimes)]==150),]),
        labels=rownames(comTimes[which(comTimes[,ncol(comTimes)]==150),]),las=2)
    for (i in 1:nrow(comTimes[which(comTimes[,ncol(comTimes)]==300),-ncol(comTimes)])) {
        lines(
        y=c(log10(apply(comTimes[which(comTimes[,ncol(comTimes)]==150),-ncol(comTimes)],1,min))[i],
        log10(apply(comTimes[which(comTimes[,ncol(comTimes)] == 300), -ncol(comTimes)],1,max))[i]),
        x=c(i,i)
    abline(h=log10(1440))
dev.off()
```