

```

#-----
## Created EN 20160106
## Resampling and FDM model fitting procedures using the EM algorithm on Sparus aurata
#-----
rm(list=ls())
setwd("~/Sparus_2013"); getwd()
library(Matrix); library(gtools); library(vegclust)
library(mvnormtest); library(mvtnorm); library(MVN);
library(car); library(MASS); library(Matrix)
library(doParallel); library(foreach);
# Paralell computing set up
cl<-2; # Number of available cores for parallel computing
registerDoParallel(cl); getDoParWorkers()
nrep<-cl
#-----
# Basic set up
#-----
k2test=0:4; # Number of known sources to simulate
kmin=1; # Minimum number of sources to evaluate (should be >=known)
kmax=8; # Maximum number of sources to evaluate
yr=4; # Number of years to included in the analysis
by.year=TRUE; # If false, it pools data from all years in a single analysis
bl.fix=FALSE; # If true, conditional fit; if falase, unconditional fit
p=c(0.1,0.2,0.3,0.4); # mixing proportions in the mixed-stock sample
n.y=100 # sample size per year
sel.var<-c("Sr","Y","Ba","B","Li","Rb","Mg") # Tournois et al (2013)'s ordered selection I (solution)
#-----
#1. Data manipulation
#-----
#1. Read data
excl<-c(212,223,274) # evident outliers to exclude
elements<- read.table("datos/three_years_data_elemental_ratio.csv",head=TRUE,sep=";",skip=1)[-excl,]
elements$Year=factor(elements$Year)
#-----
# Test normality and performs transformations
#-----
lcu<-powerTransform(as.matrix(elements[,sel.var])~elements$Habitat*elements$Year,family="bcPower")$lambda
elements3<-elements
elements3[,sel.var]=scale(bcPower(elements[,sel.var],lcu))
n.sel.var=length(sel.var)

# manova1<-manova(as.matrix(elements3[,sel.var])~elements3$Habitat*elements3$Year)
# r1<-resid(manova1)
# roystonTest(r1,qqplot=T)
# uniNorm(r1)
#-----
# Resampling and modeling routines
#-----
for (n.known in k2test)
{
  #Start the clock
  ptm<-proc.time()
  #define total number of origins to test so K>=KS and K=KS+KU
  if(n.known>=1) {
    if (kmin>=n.known){
      origins2test<-c(kmin:kmax)
    } else {origins2test<-c(n.known:kmax)}
  }else origins2test<-c(1:kmax)

  out2<-foreach (i=1:nrep,.combine=rbind,.errorhandling = "remove") %dopar%
  {
    #-----
    # 1. Bootstrapping
    #-----
    {
      #select year
      if(by.year==TRUE & yr<3){years<-sample(unique(elements3$Year),yr)} else {years=unique(elements3$Year)}
      years=years[order(years)]
      # split vailable observations by year and habitat
      list1<-split((1:nrow(elements3))[elements3$Year %in% years],elements3[elements3$Year %in%
years,c("Habitat","Year")],drop=T)
      # produce nursery-source (training) samples for all sources
      sel<-do.call("rbind",lapply(list1,function(x) {
        mu=colMeans(elements3[x,sel.var])
        vcov=cov(elements3[x,sel.var])
        data.frame(Individuals=1:25,mvnorm(25,mu=mu,Sigma = vcov),
          Habitat=unique(elements3$Habitat[x]),Year=unique(elements3$Year[x]))
      })))
      # Select and order known sources to be included in nursery-source dataset
      origin.sel<-sample(unique(elements3$Habitat),n.known);
      origin.sel=origin.sel[order(origin.sel)]
      # exclude unknown sources
      if(n.known>0) {
        known.data<-sel[sel$Habitat %in% origin.sel,]
      } else {known.data<-NULL}
      #---
      # produce mixed-stock data
      p.r=data.frame(origin=unique(elements3$Habitat),sample(p)) # Proportions in mixed-stock data
      new.data<-do.call("rbind",lapply(years,function(i) {
        do.call("rbind",lapply(1:4,function(j) {

```

```

e.ij=elements3[elements3$Year==i & elements3$Habitat==unique(elements3$Habitat)[j],sel.var]
mu=colMeans(e.ij);vcov=cov(e.ij)
n.ij=n.y*p.r[j,2]
data.frame(Individuals=1:n.ij,mvrnorm(n.ij,mu=mu,Sigma = vcov),
           Habitat=unique(elements3$Habitat)[j],Year=i)
}))
}))
if(by.year==FALSE) {new.data$Year="all";known.data$Year="all"}
known.origin<-known.data$Habitat
}
#-----
# 2. Starting mean and covariance parameters for known sources
#-----
if (n.known>0)
{
  if (length(unique(known.data$Habitat))>0){
    if (by.year==TRUE){
      sp1<-split(known.data,known.data$Year)}else{
        sp1<-list(known.data)}
    mu.known=lapply(sp1,function(x){
      sp2<-split(x,factor(x$Habitat))
      lapply(sp2,function(y) colMeans(y[,sel.var]))
    })
    vcov.known=lapply(sp1,function(x){
      sp2<-split(x,factor(x$Habitat))
      lapply(sp2,function(y) cov(y[,sel.var]))
    })
  }
}
#-----
# 3. Semi-supervised clustering
#-----
{# Fits a semi-supervised cluster model
  if (by.year==TRUE) years.hs=years else years.hs="all"
  hs<-lapply(origins2test[origins2test!=n.known], function(k) {
    hs03<-lapply(years.hs, function(j) {
      hs02=new.data[new.data$Year==j,]
      if (n.known>0){
        vc=vegclust(hs02[,sel.var], mobileCenters=(k-n.known),nstart=100,
                    fixedCenters = do.call("rbind",mu.known[[paste(j)]]),method="KM")
        origin0=names(vc$memb)[apply(vc$memb,1,which.max)]
      }else{
        vc=vegclust(hs02[,sel.var], mobileCenters=k,nstart=100,method="KM")
        origin0=names(vc$memb)[apply(vc$memb,1,which.max)]
      }
      if(!is.null(origin0)){
        origin01<-c(as.numeric(substr(origin0,2,2))-(k-n.known))
        hs02$origin=NA
        hs02$origin[substr(origin0,1,1)=="M"]=origin0[substr(origin0,1,1)=="M"]
        if (n.known>0){hs02$origin[is.na(hs02$origin)]=names(mu.known[[1]])[origin01[substr(origin0,1,1)!="M"]]}
      }
      hs02})
    names(hs03)<-as.character(years.hs)
    hs04<-do.call(rbind,hs03)
    hs04
  })
  if (length(hs)>0){
    names(hs)<-origins2test[origins2test!=n.known]
    datamix<-hs[[1]]
  }else{datamix=new.data}
}
#-----
# 4. Produce list of starting parameters
#-----
{
  ipar.lst<-lapply(origins2test,function(s){
    b.list<-fpar<-ipar<-list()
    origin.list=c(1:s)
    #-----
    # 4.1. Starting mean and covariance parameters
    #-----
    # Update mu and cov for a mix known and unknown
    if (s>n.known)
    {
      test.data2=hs[[paste(s)]]
      #Means and covariances
      if (length(unique(test.data2$origin))>1){
        if (by.year==TRUE){
          sp1<-split(test.data2,test.data2$Year)}else{
            sp1<-list(test.data2)}
          mu.new=lapply(sp1,function(x){
            sp2<-split(x,x$origin)
            lapply(sp2,function(y) colMeans(y[,sel.var]))
          })
          vcov.new=lapply(sp1,function(x){
            sp2<-split(x,x$origin)
            lapply(sp2,function(y) cov(y[,sel.var]))
          })
        }else{
          if (by.year==TRUE){
            sp1<-split(test.data2,test.data2$Year)}else{

```

```

    sp1<-list(test.data2)}
    mu.new=lapply(sp1,function(x) colMeans(x[,sel.var]))
    vcov.new=lapply(sp1,function(x){
      cov(x[,sel.var])})
  }
}
#-----
# 4.2. Starting mean and covariance parameters
#-----
if(s==n.known){
  mu.all=mu.known;vcov.all=vcov.known;
}else{
  if(n.known==0){mu.all=mu.new;vcov.all=vcov.new}else{
    vcov.all<-lapply(as.character(years),function(y){
      l1=c(vcov.known[[y]],vcov.new[[y]][!(names(vcov.new[[y])) %in% origin.sel])])
    });
    names(vcov.all)=years
    mu.all<-lapply(as.character(years),function(y){
      l1=c(mu.known[[y]],mu.new[[y]][!(names(mu.new[[y])) %in% origin.sel])])
    }); names(mu.all)=years
  }
}

```

```

#Ensure all covariance matrices are positive definite
vcov.all<-lapply(names(vcov.all),function(i)

```

```

{
  if(s>1){
    aux2<-lapply(vcov.all[[i]], function(j) {
      if(!is.na(det(j))){if(det(j)<=1E-9){
        aux1<-cov(new.data[new.data$Year==i,sel.var])} else {
          aux1=j}} else {aux1<-cov(new.data[,sel.var])}
      names(aux1)=names(j)
      aux1
    })
  }else
  {
    if(n.known==0) j=vcov.all[[i]] else j=vcov.all[[i]][[1]]
    if(!is.na(det(j))){if(det(j)<=1E-9){
      aux1<-cov(test.data2[test.data2$Year==i,sel.var])} else {
        aux1=j}} else {aux1<-cov(test.data2[,sel.var])}
    names(aux1)=names(j)
    aux2=aux1
  }
}

```

```

names(vcov.all)=names(mu.all)
ipar<-list(mu.all=mu.all,vcov.all=vcov.all)
#-----

```

```

# 4.3. Starting values for mixing proportions
#-----

```

```

# values are re-scaled logit (needed for other maximization algorithms although not for EM)

```

```

if(s>1){
  if (by.year==TRUE)
  {
    pred<-data.frame()
    for (j in years){
      new.data.y=new.data[new.data$Year==j,]
      i.names=names(vcov.all[[paste(j)]])
      p.i=sapply(i.names,function(i){
        dmvnorm(as.matrix(new.data.y[,sel.var]),mu.all[[paste(j)]][[i]],vcov.all[[paste(j)]][[i]])
      })
      pred0<-data.frame(Year=j,pred=colnames(p.i)[apply(p.i,1,which.max)])
      pred<-rbind(pred,pred0)
    }
    if(s>n.known) {pred=hs[[paste(s)]][,c("Year","origin")];names(pred)[2]="pred"}
    for (y in 1:length(years))
    {
      datamix$mixture=1
      (p.mat<-table(datamix$mixture[datamix$Year==years[y]],pred$pred[pred$Year==years[y]],useNA = "ifany"))
      p.mat2<-(as.matrix(rbind(p.mat/rowSums(p.mat))))
      p.mat2<-ifelse(is.na(p.mat2),0.001,ifelse(p.mat2==1,1-(s-1)/1000,ifelse(p.mat2==0,0.001,p.mat2)))
#avoid log(0)
      p.mat2<-p.mat2/rowSums(p.mat2)
      b.list[[y]]<-log((p.mat2)/(1-p.mat2))[1:(s)]
      names(b.list)[[y]]=paste(years[y])
    }
  }else{
    pred<-data.frame()
    p.i<-matrix(NA,nc=s,nr=nrow(new.data))
    for (i in 1:s){
      i.names=names(vcov.all[[1]])
      p.i[,i]<-dmvnorm(as.matrix(new.data[,sel.var]),mu.all[[1]][[i.names[[i]]]],vcov.all[[1]]
[[i.names[[i]]]])
    }
    pred0<-data.frame(Year="all",pred=names(mu.all[[1]])[apply(p.i,1,which.max)])
    pred<-rbind(pred,pred0)
    if(s>n.known) {pred=hs[[paste(s)]][,c("Year","origin")];names(pred)[2]="pred"}
    datamix$mixture=1
    (p.mat<-table(datamix$mixture,pred$pred,useNA = "ifany"))
    p.mat2<-(as.matrix(rbind(p.mat/rowSums(p.mat))))
    p.mat2<-ifelse(is.na(p.mat2),0.001,ifelse(p.mat2==1,1-(s-1)/1000,ifelse(p.mat2==0,0.001,p.mat2))) #avoid
log(0)
    p.mat2<-p.mat2/rowSums(p.mat2)
  }
}

```

```

        b.list[[1]]<-log((p.mat2)/(1-p.mat2))
        names(b.list)[[1]]="all"
    }
    {ipar[["b.list"]]=b.list}
}
if (s>1) {ipar[["b.list"]]=b.list}
#-----
# 4.4. Compilation of all starting parameter, data and other useful values
#-----
list(ipar=ipar,datamix=datamix[,c(sel.var,"Year")],origin.sel=origin.sel,year.sel=years)
})
names(ipar.lst)=origins2test
}
#-----
# Likelihood maximization through the EM algorithm
#-----
out1=lapply(ipar.lst,function(x)
{
  if(by.year==TRUE) years2=x$year.sel else years2="all"
  em.year=lapply(years2,function(y)
  {
    # Data matrix
    X=datamix;X=X[X$Year==y,sel.var]
    # Reconstruct parameter vectors and matrices (heritage from optim)
    known.origins=x$origin.sel
    pi=if(is.null(x$ipar$b.list)) 1 else 1/(1+exp(-x$ipar$b.list[[paste(y)]]))
    mu=x$ipar$mu.all[[paste(y)]];mu=mu[order(names(mu))]
    sigma=x$ipar$vcov.all[[paste(y)]]
    if(length(pi)>1){sigma<-sigma[order(names(sigma))]}
    # Working starting values
    mu.em=mu;pi.em=pi;sigma.em=sigma;nsources=length(pi)
    # Produces joint dataset
    XK=known.data[known.data$Year==y,]
    if(n.known>0){X2=rbind(X,XK[,sel.var])}else{X2=X}
    # Likelihood maximization for K>1
    if(length(pi)>1)
    {
      #Compute starting likelihood
      loglik<- vector()
      loglik[1]<-0
      ld0=sapply(names(mu),function(i){
        mvtnorm::dmvnorm(as.matrix(X),mu.em[[i]],sigma.em[[i]])
      })
      loglik[2]<-sum(log(rowSums(t(apply(ld0,1,function(j) pi.em*j))))))
      k=2
      # EM algorithm
      while(abs(loglik[k]-loglik[k-1])>=1E-7 & k<=300) {
        # E step
        tau.source=sapply(names(mu),function(i){
          (XK$Habitat==i)*1
        })
        dd=sapply(names(sigma),function(i){
          dmvnorm(as.matrix(X),mu.em[[i]],sigma.em[[i]])
        })
        tau.mixed<-t(apply(dd,1,function(j) j*pi.em/sum(j*pi.em)))
        # Add source data
        if(n.known>0){tau=rbind(tau.mixed,tau.source)}else{tau=tau.mixed}
        if(ncol(dd)==1) tau=t(tau)
        colnames(tau)=names(mu)
        tau[is.na(tau)]=1/nsources
        # M step
        pi.em<-apply(tau.mixed,2,function(i) sum(i)/length(i))
        if(bl.fix==FALSE){# Performs Unconditional estimation of mu and theta for all nursery-sources
          # Compute updated mu values for all sources
          mu.em<-split(t(apply(tau,2,function(i) colSums(i*X2)/sum(i))),names(mu.em))
          # rename
          mu.em<-lapply(mu.em,function(i) {j=matrix(i,nr=1);colnames(j)=sel.var;j})
          # Compute updated sigma values for all sources
          sigma.em<-lapply(names(sigma.em),function(i) {
            res.i<-t(apply(X2,1,function(x) x-mu.em[[i]]))
            cov.i=cov.wt(res.i, wt=tau[,i])$cov
            if(is.finite(det(cov.i)) & det(cov.i)>1e-9) cov.i else sigma.em[[i]]
          })
          # rename
          names(sigma.em)=names(sigma)
        }else{# Limits unconditional estimation of mu and theta to unknown nursery-sources
          # Compute updated mu values for all sources
          mu.em0<-split(t(apply(tau,2,function(i) colSums(i*X)/sum(i))),names(mu.em))
          # rename
          mu.em0<-lapply(mu.em0,function(i) {j=matrix(i,nr=1);colnames(j)=sel.var;j})
          # Pass along updated mu values only for unknown origins
          mu.em=c(mu[names(mu) %in% known.origins],mu.em0[!(names(mu.em0) %in% known.origins)])
          mu.em=mu.em[order(names(mu.em))]
          # Compute updated sigma values for all sources
          sigma.em0<-lapply(names(sigma),function(i) {
            res.i<-t(apply(X,1,function(x) x-mu.em[[i]])
            cov.i=cov.wt(res.i, wt=tau[,i])$cov
            if(is.finite(det(cov.i)) & det(cov.i)>1e-9) cov.i else sigma.em[[i]]
          })
          names(sigma.em0)=names(sigma)
          # Pass along updated sigma values only for unknown origins

```

```

        sigma.em=c(sigma[names(sigma) %in% known.origins],sigma.em0[!(names(sigma.em0) %in% known.origins)])
        sigma.em=sigma.em[order(names(sigma.em))]
    }
    #probability density given updated parameters
    ld.test=sapply(names(mu),function(i){
        mvtnorm::dmvnorm(as.matrix(X2),mu.em[[i]],sigma.em[[i]],log=T)
    })
    #log-likelihood given updated parameters
    if(ncol(ld.test)>1){
        loglik[k+1]<-sum(colSums(tau*(t(apply(ld.test,1,function(x) log(pi.em)+x))))))
    }else{
        loglik[k+1]<-sum(colSums(tau*(apply(ld.test,1,function(x) log(pi.em)+x))))
    }
    k<-k+1
}
if(k>=300){loglik<-9999}

#Final probability densities
ld.fin=sapply(names(mu.em),function(i){
    mvtnorm::dmvnorm(as.matrix(X),mu.em[[i]],sigma.em[[i]])
})
tau.fin<-t(apply(ld.fin,1,function(j) j*pi.em/sum(j*pi.em)))
}
#Final Likelihood and mixing proportions
if(length(pi)>1) {
    pi.fin<-pi.em
    loglik.fin<-sum(log(rowSums(t(apply(ld.fin,1,function(x) pi.fin*x))))))
} else {
    pi.fin=1
    if(n.known==1) mu=mu[[1]]
    loglik.fin<-sum(log(mvtnorm::dmvnorm(as.matrix(X),mu,sigma)))
}
# Rename unknown groups by maximizing correct classification (only for 4-source models)
if(length(mu.em)==4){
    cc0=data.frame(Habitat=new.data$Habitat[new.data$Year==y]
        ,ph=colnames(ld.fin)[apply(ld.fin,1, which.max)])
    cc1=as.data.frame(cbind(table(cc0$ph,cc0$Habitat)))
    #select combination that maximize correct assignments
    perm=permutations(4,4)
    perm.sum=sapply(1:nrow(perm),function(j){
        cc1[1,perm[j,1]]+cc1[2,perm[j,2]]+cc1[3,perm[j,3]]+cc1[4,perm[j,4]]
    })
    names(mu.em)=unique(new.data$Habitat)[perm[which.max(perm.sum),]]
}
#Output list
list(mu=mu.em,pi=pi.fin,sigma=sigma.em,loglik=loglik.fin,nx=nrow(X),nsources=nsources)
})
names(em.year)=years2
em.year
})
#Prepare output 2
if (!is.null(out1)){
    names(out1)=origins2test
    #AIC
    AICs<-matrix(sapply(out1,function(i) {sapply(i, function(j){
        ifelse(bl.fix==TRUE,-2*j$loglik+2*(j$nsources+(j$nsources-n.known)*(7*2+(49-7)/2)-1),
        -2*j$loglik+2*(j$nsources*(1+7*2+(49-7)/2)-1)
    })),nc=length(ipar.lst))
    #BIC
    BICs<-matrix(sapply(out1,function(i) {sapply(i, function(j){
        ifelse(bl.fix==TRUE,-2*j$loglik+log(j$nx)*(j$nsources+(j$nsources-n.known)*(7*2+(49-7)/2)-1),
        -2*j$loglik+log(j$nx)*(j$nsources*(1+7*2+(49-7)/2)-1)
    })),nc=length(ipar.lst))
    #Several values of interest
    loglik<-matrix(sapply(out1,function(i) sapply(i, function(j) j$loglik)),nc=length(ipar.lst))
    best.model.aic=names(ipar.lst)[apply(AICs,1,which.min)]
    best.aic=apply(AICs,1,function(i) i[which.min(i)])
    best.model.bic=names(ipar.lst)[apply(BICs,1,which.min)]
    best.bic=apply(BICs,1,function(i) i[which.min(i)])
    bic.dist=apply(BICs,1,function(i) (min(i)-mean(i))/sd(i))
    #Defines output data-frame
    out1.2=data.frame(by.year=by.year,bl.fix=bl.fix,n.known=n.known,run=i,
        Year=rep(names(out1[[1]]),each=4),
        origin=unlist(lapply(out1[["4"]],function(j) names(j[["mu"]]))),
        source.data=unlist(lapply(out1[["4"]],function(k) names(k[["sigma"]]))),
        do.call("rbind",lapply(out1[["4"]],function(l) {do.call("rbind",l[["mu"]])})),
        known.origins=paste(ipar.lst[["4"]]$origin.sel,collapse="-"),
        known.years=paste(ipar.lst[["4"]]$year.sel,collapse="-"),
        best.model.aic=rep(best.model.aic,each=4),
        best.aic=rep(best.aic,each=4),
        best.model.bic=rep(best.model.bic,each=4),
        best.bic=rep(best.bic,each=4),
        bic.dist=rep(bic.dist,each=4),
        pred.mix=unlist(lapply(out1[["4"]],function(l) l[["pi"]]))
    # Add true mixing proportion in mixed-stock data
    out1.2=merge(out1.2,p.r,by="origin")
    out1.2=out1.2[order(out1.2$Year,out1.2$origin),]
} else {out1.2=NULL}
out1.2
}
out2 # Stop the clock
elapsed=round((proc.time()-ptm)[3]/3600,2)

```

```
# Write the output
write.csv(out2,paste("EM known
",n.known,".by.year_",by.year,".bl.fix_",bl.fix, ".",Sys.time(),elapsed,".", ".csv",sep=""),row.names=F)
}
```