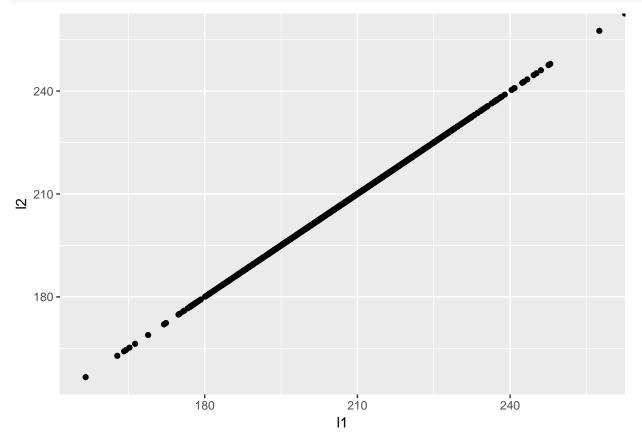
Tests

In order to search for possible bugs and test the package we have create alternative functions to the original ones which should be equivalent. There are a series of possible test to check that the package is doing the proper things to do

log-likelihood function

```
time = proc.time()
#dendroica
btdd = c(4.9999999998,4.806886544,4.70731246478,4.50735197578,4.37856240588,4.29594855558,4.19207515688
pars = c(3.2,0.3,40)
m = 1000
S = sim.sct(btdd,pars,m,print=FALSE)
l1 = vector(mode="numeric",length=m)
l2 = vector(mode="numeric",length=m)
for(i in 1:m){
    s = S$rec[[i]]
    l1[i] = nllik.tree(pars,s)
    l2[i] = nllik.tree2(pars,s)
}
qplot(l1,12)
```



```
all.equal(11,12)

## [1] TRUE

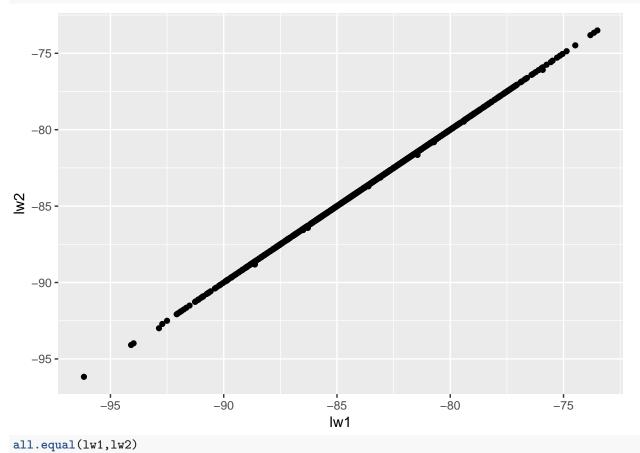
get.time(time)

## [1] 5.298
```

Importance sampling weights

To check if the importance sampling weights are well calculated we use the previous and the current method using the sim.extinct_old function.

```
lw1 = vector(mode="numeric",length=m)
lw2 = vector(mode="numeric",length=m)
w1 = vector(mode="numeric",length=m)
w2 = vector(mode="numeric",length=m)
for(i in 1:1000){
    si = sim.extinct_old(btdd,pars)
    w1[i] = si$weight
    w2[i] = si$weight2
lw1[i] = si$logweight
lw2[i] = si$logweight2
}
qplot(lw1,lw2)
```



```
## [1] "Mean relative difference: 2.491663e-05"
all.equal(w1,w2)
## [1] TRUE
now we check other ..
test.g <- function(df,pars){</pre>
  df$xi = 0
  df$xi[df$bte<df$bt[dim]] = 1</pre>
 dim = dim(df)[1]
 wtT = diff(c(0,df\$bt))
 n.tree = list(wt=wtT,E=df$to[-dim])
 n.tree$E[n.tree$E==2] = 1
 E = n.tree$E
 n = c(2,2+cumsum(E)+cumsum(E-1))
  lambda = (pars[1]-(pars[1]-pars[2])*(n/pars[3]))
  mu = pars[2]
  s = lambda*n
  lsprob2 = da.prob2(wt=wtT,t_ext=df$t.ext,s=s,mu=pars[2],r=df$bt[dim]-c(0,df$bt[-dim]),n=n)
  ls2 = sum(lsprob2)
  lsprob = da.prob(xi=df$xi,wt=wtT,t_ext=df$t.ext,s=s,mu=mu0,r=df$bt[dim]-c(0,df$bt[-dim]),n=n)
 ls = sum(lsprob)
 return(list(ls=ls,ls2=ls2))
df = sim.extinct(btdd,pars)
test.g(df,pars)
```

Data augmentation algorithm

We have two options on ..

```
no cores <- detectCores()
cl <- makeCluster(no_cores)</pre>
registerDoParallel(cl)
m=1000000
time = proc.time()
trees <- foreach(i = 1:m, combine = list) %dopar% {</pre>
  df = emphasis::sim.extinct(brts = btdd,pars = pars)
  11 = length(df$bt)
  return(list(l1=l1))
}
t1 = get.time(time)
11 = sapply(trees,function(list) list$11)
time = proc.time()
trees <- foreach(i = 1:m, combine = list) %dopar% {</pre>
  df = emphasis::sim.extinct2(brts = btdd,pars = pars)
  12 = length(df$bt)
 return(list(12=12))
12 = sapply(trees,function(list) list$12)
t2 = get.time(time)
stopCluster(cl)
```

```
dfh = data.frame(l=c(l1,l2),alg = c(rep("a1",length(l1)),rep("a2",length(l2))))
ggplot(dfh,aes(x=1,fill=alg)) + geom_histogram(alpha=0.5,position='identity',binwidth = 1)
  80000 -
  60000 -
                                                                                    alg
40000 -
                                                                                         а1
                                                                                         a2
  20000 -
      0 -
                      100
                                      125
                                                      150
                                                                      175
summary(11)
                               Mean 3rd Qu.
##
      Min. 1st Qu.
                     Median
                                                Max.
                                               184.0
##
      82.0
             122.0
                      130.0
                              129.3
                                       136.0
summary(12)
##
      Min. 1st Qu.
                    Median
                               Mean 3rd Qu.
                                                Max.
      82.0
##
             122.0
                      130.0
                              129.4
                                       136.0
                                               180.0
t1
## [1] 1458.284
t2
## [1] 14992.69
two sampling algorithms
sim.extinct <- function(brts,pars,model='dd',seed=0){</pre>
  if(seed>0) set.seed(seed)
  wt = -diff(c(brts, 0))
  ct = sum(wt)
  lambda0 = pars[1]
```

```
mu0 = pars[2]
K = pars[3]
dim = length(wt)
ms = NULL # missing speciations, for now we just add time. When we consider topology we do it with sp
me = NULL # missing extinctions (in the uniform plane)
bt = NULL
bte = NULL
to = NULL
cbt = 0
N = 2
for(i in 1:dim){
  cwt = wt[i]
  cbt = sum(wt[0:(i-1)])
 key = 0
  while (\text{key} == 0) {
    if(model == "dd"){  # diversity-dependence model
      lambda = max(1e-99, lambda0 - (lambda0-mu0)*N/K)
      mu = mu0
      s = N*lambda
    }else{print('Model not implemented yet, try dd')}
    t.spe = rexp(1,s)
    t.ext = extinction.processes(u=me,inits=ms,mu0=mu0)
    t_ext = ifelse(length(t.ext)>0,min(t.ext),Inf)-cbt # if is not empty gives the waiting time for
    mint = min(t.spe, t_ext)
    if(mint < cwt){</pre>
      if(mint == t.spe){#speciation
        u = runif(1)
        if(u < pexp(ct-(cbt+t.spe),mu)){</pre>
          ms = c(ms, cbt+t.spe)
          me = c(me, u)
          bt = c(bt, cbt+t.spe)
          text = extinction.processes(u=u,inits=cbt+t.spe,mu0=mu0)
          bte = c(bte,text)
          to = c(to,1)
          N = N + 1
        cwt = cwt - t.spe
        cbt = cbt + t.spe
      else{#extinction
        extinctone = which(t.ext == min(t.ext))
        text = t.ext[extinctone]
        bt = c(bt, text)
        bte = c(bte, Inf)
        to = c(to,0)
        ms = ms[-extinctone]
        me = me[-extinctone]
        cwt = cwt - t_ext
        cbt = cbt + t_ext
        N = N-1
    }
    else{
```

```
key = 1
      }
    }
    N = N+1
  df = data.frame(bt = c(bt,ct-brts),bte = c(bte, rep(Inf,length(wt))),to = c(to,rep(2,length(wt))))
  df = df[order(df$bt),]
  df$t.ext = df$bte-df$bt
  df = df[-1,]
  df = rbind(df,data.frame(bt=ct,bte=Inf,to=0,t.ext=Inf))
  return(df)
}
sim.extinct2 <- function(brts,pars,model='dd',seed=0){</pre>
  if(seed>0) set.seed(seed)
  wt = -diff(c(brts, 0))
  ct = sum(wt)
  lambda0 = pars[1]
  mu0 = pars[2]
  K = pars[3]
  dim = length(wt)
  bt = NULL
  bte = NULL
  to = NULL
  N = 2
  for(i in 1:dim){
    cwt = wt[i]
    cbt = sum(wt[0:(i-1)])
    key = 0
    while(key == 0){
      if(model == "dd"){  # diversity-dependence model
        lambda = max(1e-99, lambda0 - (lambda0-mu0)*N/K)
        mu = mu0
        s = N*lambda
      }else{print('Model not implemented yet, try dd')}
      rns = rnhe(lambda=s,mu=mu,Ti=ct-cbt)
      t.spe = rns$rv
      ex = rns\$ex
      sbte = bte[bte>cbt]
      t_ext = ifelse(length(sbte)>0,min(sbte),Inf) - cbt
      mint = min(t.spe,t_ext)
      if(mint < cwt){</pre>
        if(mint == t.spe){#speciation
          bt = c(bt, cbt+t.spe)
          text = cbt+t.spe+ex/mu0
          bte = c(bte,text)
          to = c(to,1)
          N = N + 1
          cwt = cwt - t.spe
          cbt = cbt + t.spe
        }
        else{#extinction
          bt = c(bt,cbt+t_ext)
```

```
bte = c(bte, Inf)
          to = c(to,0)
         cwt = cwt - t_ext
         cbt = cbt + t_ext
         N = N-1
     }
     else{
       key = 1
   }
   N = N+1
 df = data.frame(bt = c(bt,ct-brts),bte = c(bte, rep(Inf,length(wt))),to = c(to,rep(2,length(wt))))
  df = df[order(df$bt),]
 df$t.ext = df$bte-df$bt
 df = df[-1,]
 df = rbind(df,data.frame(bt=ct,bte=Inf,to=0,t.ext=Inf))
 return(df)
}
```