## mantelNSGA vignette

```
library(mantelNSGA2)
library(vegan)
```

#### 1 Introduction

bla (Clarke and Warwick 1994). Clarke and Warwick (1994) total combinations tested by sinkr::bioEnv() 2^ncol(var.mat)-1 = {r 2^ncol(var.mat)-1}

The following tutorial outlines the use of the ELEFAN functions available in TropFishR (ELEFAN, ELEFAN\_GA, and ELEFAN\_SA). These functions can be used to estimate growth model parameters from length-frequency (lfq) data. The ability to convert length to relative age is the initial step in length-based stock assessment, and underpins subsequent analyses (see the sinkr package for a broader overview of length-based stock assessment).

The growth model presently implemented within TropFishR is the von Bertalanffy growth function (VBGF),

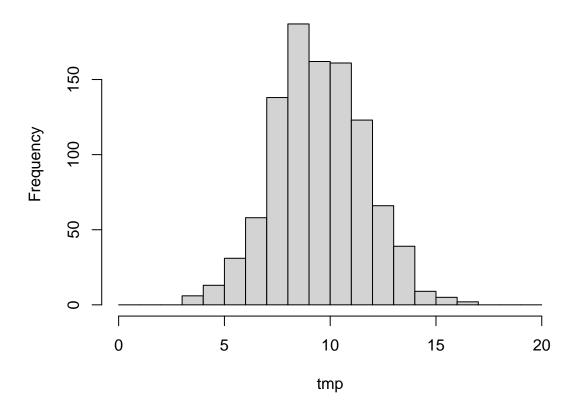
$$L_t = L_{\infty}(1 - \exp(-K(t - t_0))),$$

where  $L_t$  is length-at-age t,  $L_{\infty}$  is asymptotic length, K is the von Bertalanffy growth constant, and  $t_0$  is the theoretical age when length equals zero.  $t_0$  is usually negative, resulting in a positive length at the time of recruitment (t = 0). In TropFishR, these parameters are referred to as Linf, K, and t0.

The following plot shows the basic form of the VBGF under varying K settings:

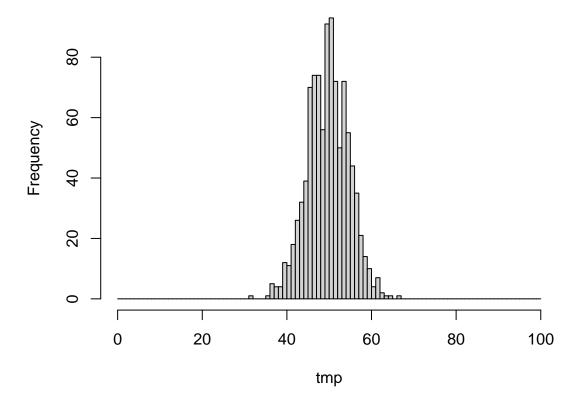
```
# chance of having few or many "genes" turned on is low, so one ends up
# mainly exploring the middle
nvar <- 20
tmp <- replicate(1000, expr = sum(sample(x = c(0,1), nvar, replace = TRUE)))
hist(tmp, breaks = seq(0, nvar))</pre>
```

# Histogram of tmp



```
# more extreme when number of genes is high
nvar <- 100
tmp <- replicate(1000, expr = sum(sample(x = c(0,1), nvar, replace = TRUE)))
hist(tmp, breaks = seq(0, nvar))</pre>
```

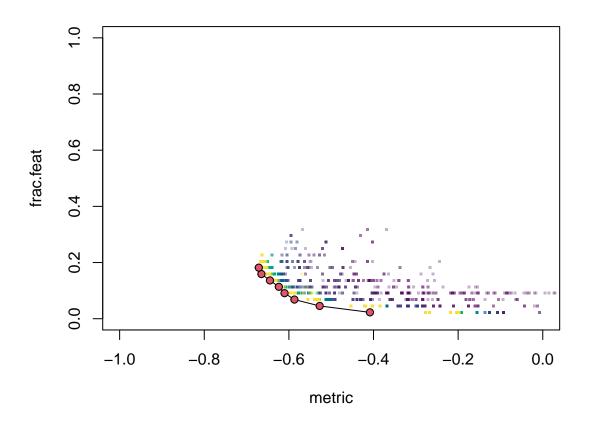
### Histogram of tmp



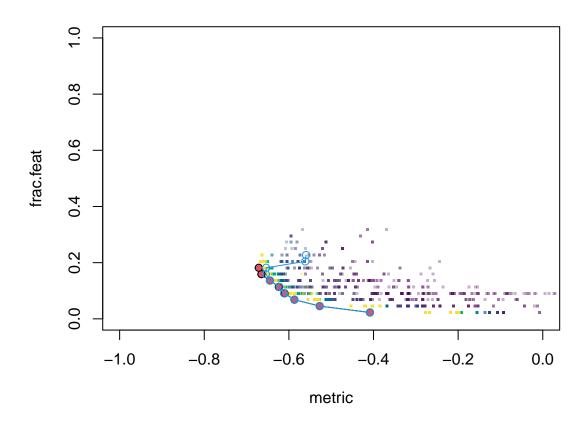
#### 1.1 EnvBio example

```
data("varechem")
data("varespec")
# biological community variables that best correlate with environmental data
fix.mat = varechem
var.mat = wisconsin(varespec)
fix.dist.method = "euclidean"
var.dist.method = "bray"
scale.fix = TRUE
scale.var = FALSE
p.feat <- 0.1
# mantelNSGA2
set.seed(1111)
fitGA <- mantelNSGA2(</pre>
 fix.mat = fix.mat, var.mat = var.mat,
 fix.dist.method = fix.dist.method, var.dist.method = var.dist.method,
  scale.fix = scale.fix, scale.var = scale.var, p.feat = p.feat,
 pop.size = 50, max.iter = 100, stop.criterion = 50,
```

```
mutation.rate = 0.2, crossover.rate = 0.8, verbose = FALSE
)
# plot(fitGA$generation.fitness)
# fitGA$pareto.solution
plot(fitGA, parFrontT = "o")
```



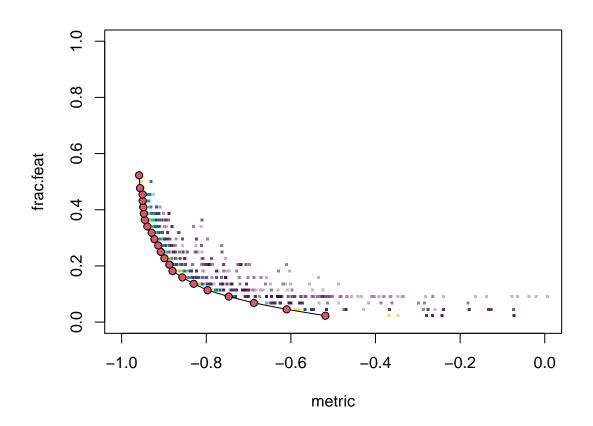
```
# Compare to buStep
set.seed(1111)
fitBV <- bvStep(
    fix.mat = fix.mat, var.mat = var.mat,
    fix.dist.method = fix.dist.method, var.dist.method = var.dist.method,
    scale.fix = scale.fix, scale.var = scale.var,
    num.restarts = 50,
    random.selection = TRUE,
    prop.selected.var = 0.4,
    verbose = FALSE
)
# fitBV$order.by.best # mantelGA not looking in simpler solutions enough
# fitBV$order.by.i.comb
plot(fitGA, parFrontT = "o")
points(-fitBV$order.by.i.comb$rho, fitBV$order.by.i.comb$n.var/ncol(var.mat), pch = 1, col = 4, t = "o"</pre>
```



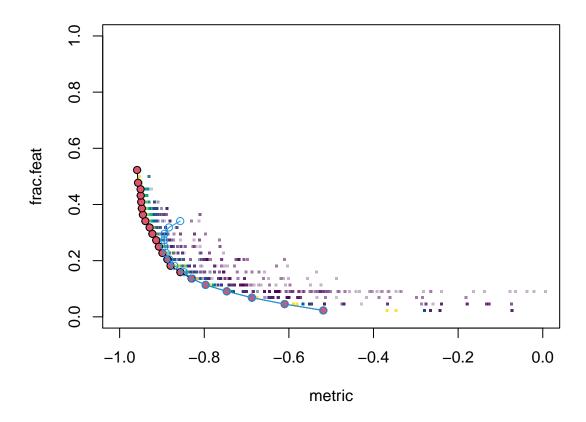
### 1.2 BioBio example

```
# biological community variables that best correlate with the
# overall biological community
fix.mat = wisconsin(varespec)
var.mat = wisconsin(varespec)
fix.dist.method = "bray"
var.dist.method = "bray"
scale.fix = FALSE
scale.var = FALSE
p.feat <- 0.1
# mantelNSGA2
set.seed(1111)
fitGA2 <- mantelNSGA2(</pre>
  fix.mat = fix.mat, var.mat = var.mat,
  fix.dist.method = fix.dist.method, var.dist.method = var.dist.method,
  scale.fix = scale.fix, scale.var = scale.var, p.feat = p.feat,
  pop.size = 50, max.iter = 100, stop.criterion = 50,
  mutation.rate = 0.2, crossover.rate = 0.8, verbose = FALSE
```

```
# plot(fitGA2$generation.fitness)
# fitGA2$pareto.solution
plot(fitGA2, parFrontT = "o")
```



```
# Compare to buStep
set.seed(1111)
fitBV2 <- bvStep(</pre>
  fix.mat = fix.mat, var.mat = var.mat,
  fix.dist.method = fix.dist.method, var.dist.method = var.dist.method,
  scale.fix = scale.fix, scale.var = scale.var,
  num.restarts = 50,
  random.selection = TRUE,
  prop.selected.var = 0.3,
  var.always.include = c(15,23,26),
  verbose = FALSE
)
# fitBV2$order.by.best # mantelGA not looking in simpler solutions enough
# fitBV2$order.by.i.comb
plot(fitGA2, parFrontT = "o")
points(-fitBV2$order.by.i.comb$rho, fitBV2$order.by.i.comb$n.var/ncol(var.mat), pch = 1, col = 4, t = "
```



# References

Clarke, KR, and RM Warwick. 1994. "An Approach to Statistical Analysis and Interpretation." Change in Marine Communities 2: 117-43.