

Lottia.digitalis

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Setup

```
library(rentrez)
library(ape)
library(pegas)
library(strataG)
library(stringr)
library(knitr)
library(ggplot2)
library(reshape2)

#set the working directory
#opts_knit$set(root.dir = "~/Users/crandall_lab/Desktop/Lilli_Krier/CapeMendocino")

# read in the data to a DNAbin object

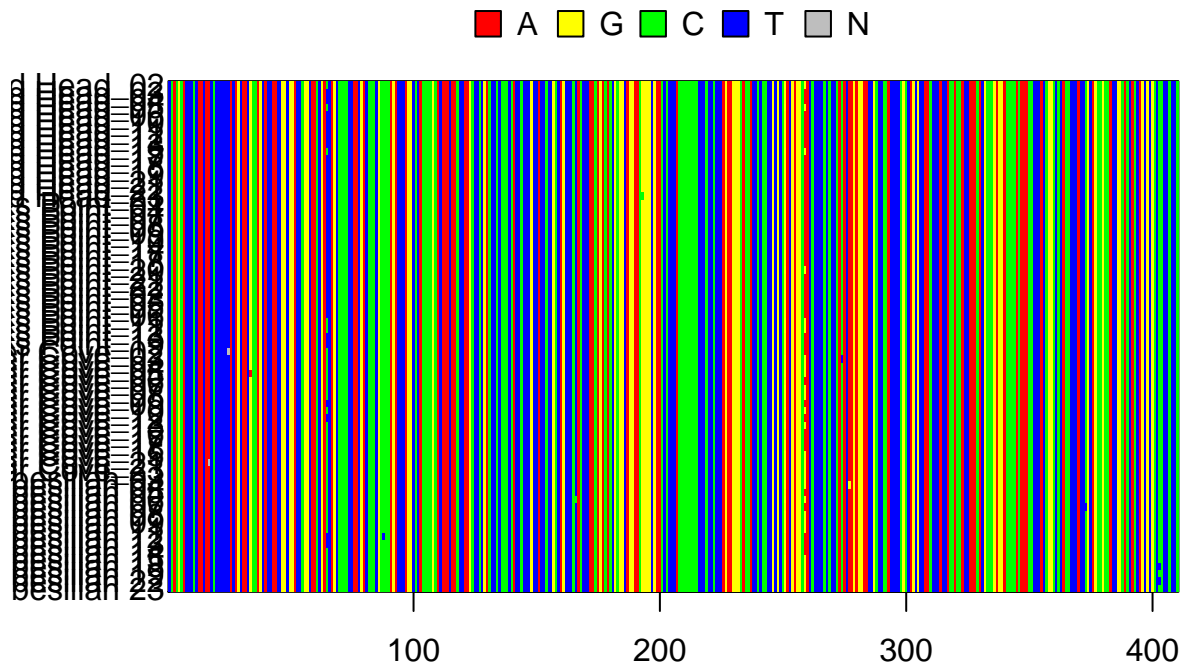
Ldigitalis <-read.FASTA("Final_Lottia_Alignments/Lottia_digitalis.fasta")
```

Initial look at data

Align the data and look at a quick image of it

```
Ldigitalis <- muscle(Ldigitalis)

image.DNAbin(Ldigitalis)
```



Haplotype network

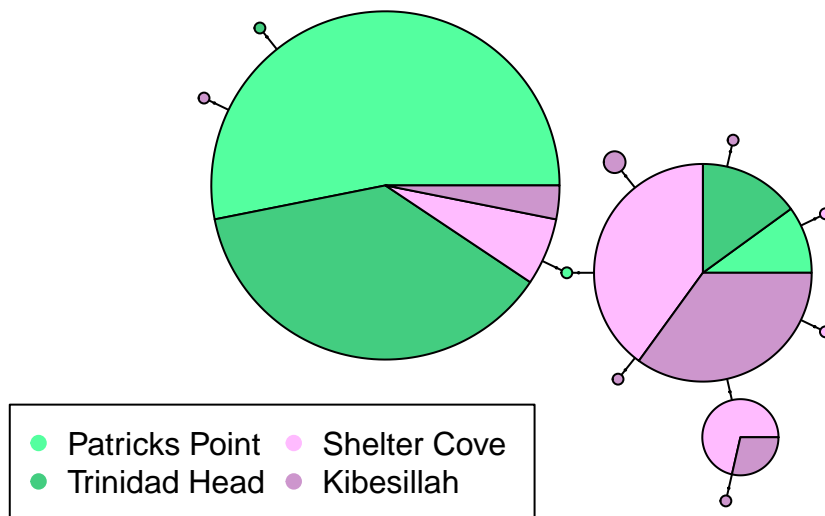
First make a network that colors each population separately

```
#make a population map by stripping the population name off of the DNAbin labels
pop <- str_extract(rownames(Ldigitalis), "[A-Za-z]+\\s*[A-Za-z]+")
```

```
#Label the haplotypes and sort
h <- pegas::haplotype(Ldigitalis)
h <- sort(h, what = "label")
#Create the network
net <- pegas::haploNet(h)
#Funky code to make a table of which haplotypes occur in which populations
i<-stack(setNames(attr(h, "index"), rownames(h)))
i<-i[order(i$values),]
ind.hap<-table(hap=i$ind, pop=pop)
# put columns of ind.hap into North to South order
ind.hap<-ind.hap[,c("Patricks Point", "Trinidad Head", "Shelter Cove", "Kibesillah")]
```

```
#Pick some funky colors
mycolors <- c("seagreen1", "seagreen3", "plum1", "plum3")
```

```
plot(net, size=attr(net, "freq"), scale.ratio=2, pie = ind.hap, legend = F, labels = F, threshold = 0, sl
legend("bottomleft", colnames(ind.hap), col=mycolors, pch=19, ncol = 2)
```



Diversity Statistics

```
Ldigitalis_g <- sequence2gtypes(Ldigitalis, strata = pop)

Ldigitalis_g <- labelHaplotypes(Ldigitalis_g)

stratastat<-function(x,pop=pop,fun=nuc.div){
  #this function will calculate stats for a DNABin object (x), stratified across populations given in pop
  # Some functions this will work with: nuc.div(), theta.s(), tajima.test() from pegas, FusFs(), exptdH
  stats<-NULL
  for(p in unique(pop)){
    stat<-fun(x[grepl(p,rownames(x)),])
    stats<-c(stats,stat)
  }
  names(stats)<-unique(pop)
  return(stats)
}

nucd<-stratastat(Ldigitalis,pop=pop,fun=nuc.div)
Fs<-stratastat(Ldigitalis,pop=pop,fun=fusFs)

kable(data.frame("Nucleotide Diversity"=nucd, "Fu's Fs" = Fs ))
```

| | Nucleotide.Diversity | Fu.s.Fs |
|----------------|----------------------|------------|
| Trinidad Head | 0.0018902 | 0.4582495 |
| Patricks Point | 0.0011168 | -0.3774818 |
| Shelter Cove | 0.0023789 | -1.4136598 |
| Kibesillah | 0.0034756 | -4.1917041 |

```
Ldigitalis_g
```

```
## $gtypes
##
```

```
## <<< gtypes created on 2019-10-25 11:28:40 >>>
##
## Contents: 69 samples, 1 locus, 4 strata
##
## Strata summary:
##          num.samples num.missing num.alleles prop.unique.alleles
## Kibesillah           16           0           8           0.6250000
## Patricks Point       20           0           3           0.3333333
## Shelter Cove         17           0           5           0.4000000
## Trinidad Head        16           0           3           0.3333333
##          heterozygosity
## Kibesillah           0.8083333
## Patricks Point       0.2789474
## Shelter Cove         0.7132353
## Trinidad Head        0.4250000
##
## $unassigned
## $unassigned$gene.1
## NULL
```

F-statistics

Took some hints from [here](#), and [here](#)

Pairwise PhiST

Let's plot the pairwise PhiST statistics as a heatmap

```
pairwise_phi <- pairwiseTest(Ldigitalis_g$gtypes, stats = "phist", nrep = 10000, model = "raw")
```

```
##
## <<< gtypes created on 2019-10-25 11:28:40 >>>
## 2019-10-25 11:28:40 : Pairwise tests : 10000 permutations
## 2019-10-25 11:28:40 : Kibesillah v. Patricks Point
## 2019-10-25 11:28:41 : Kibesillah v. Shelter Cove
## 2019-10-25 11:28:42 : Kibesillah v. Trinidad Head
## 2019-10-25 11:28:42 : Patricks Point v. Shelter Cove
## 2019-10-25 11:28:43 : Patricks Point v. Trinidad Head
## 2019-10-25 11:28:43 : Shelter Cove v. Trinidad Head
##
## Population structure results:
##          pair.label          PHist PHist.p.val
## 1    Kibesillah (16) v. Patricks Point (20)  0.36982491  0.00019998
## 2    Kibesillah (16) v. Shelter Cove (17) -0.01538677  0.56004400
## 3    Kibesillah (16) v. Trinidad Head (16)  0.33153153  0.00079992
## 4    Patricks Point (20) v. Shelter Cove (17)  0.41626979  0.00009999
## 5    Patricks Point (20) v. Trinidad Head (16) -0.02885108  0.74872513
## 6    Shelter Cove (17) v. Trinidad Head (16)  0.44285870  0.00019998

phiST <- pairwise_phi$pair.mat$PHist

#symmetrize the matrix
phiST[upper.tri(phiST)] = t(phiST)[upper.tri(phiST)]
```

```

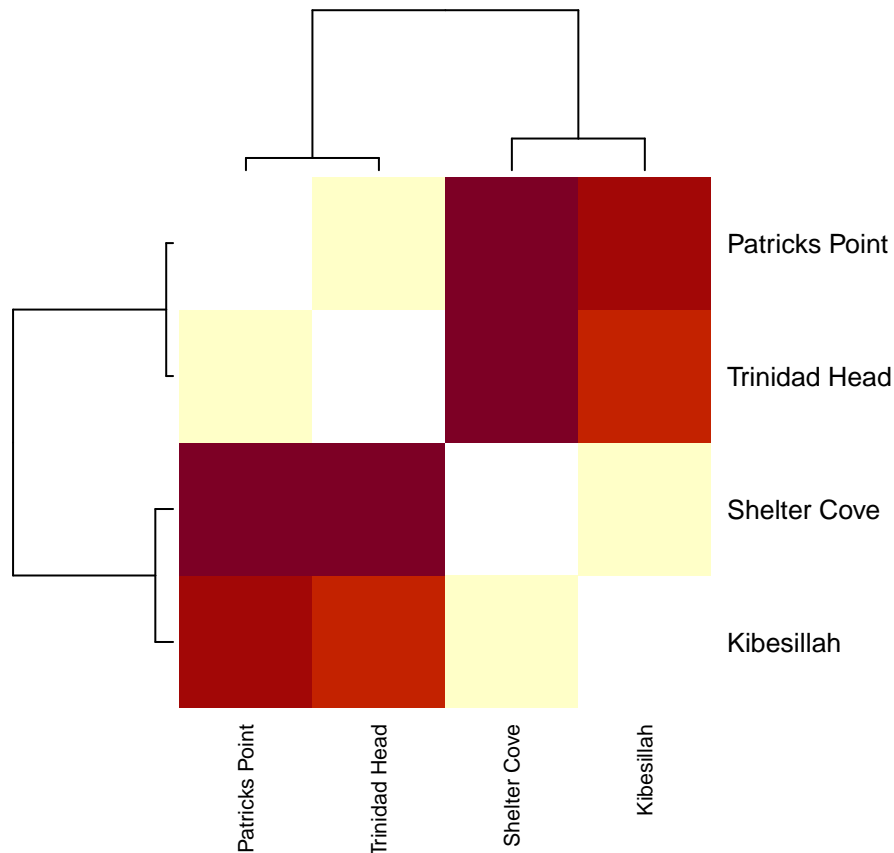
#put it in North to South Order
phiST<-phiST[c(2,4,3,1),c(2,4,3,1)]

distance <- dist(phiST)
cluster <- hclust(distance, method="ward.D")
dendrogram <- as.dendrogram(cluster)
Rowv <- c(4,3,2,1)
dendrogram <- reorder(dendrogram, Rowv)
reorderfun <- function(d,w) { d }

heatmap(phiST,scale="none",Rowv=dendrogram, reorderfun=reorderfun, cexRow = 1, cexCol=0.8, legend = "col")

## Warning in plot.window(...): "legend" is not a graphical parameter
## Warning in plot.xy(xy, type, ...): "legend" is not a graphical parameter
## Warning in title(...): "legend" is not a graphical parameter

```



AMOVA

```

#write.csv(pop,"amova_hypotheses_lottia.csv",row.names = F, quote=F)
#Open this file up in a spreadsheet program and fill out your hypotheses for higher level hierarchies, i

#read them back into R
amovahyps<-read.csv("amova_hypotheses_lottia.csv")
CapeMendocino<-amovahyps$CapeMendocino

```

```
pop1<-as.factor(pop)
```

```
#Calculate the p-distance among all sequences  
dists<-dist.dna(Ldigitalis,model="raw")
```

Test the Cape Mendocino hypothesis:

```
amova_out<-pegas::amova(formula=dists~CapeMendocino/pop1,nperm=1000)
```

```
amova_out
```

```
##  
## Analysis of Molecular Variance  
##  
## Call: pegas::amova(formula = dists ~ CapeMendocino/pop1, nperm = 1000)  
##  
##              SSD              MSD df  
## CapeMendocino 1.979667e-04 1.979667e-04 1  
## pop1          1.103392e-06 5.516958e-07 2  
## Error         3.302693e-04 5.081065e-06 65  
## Total         5.293393e-04 7.784402e-06 68  
##  
## Variance components:  
##              sigma2 P.value  
## CapeMendocino 5.7346e-06 0.0000  
## pop1          -2.6439e-07 0.9411  
## Error         5.0811e-06  
##  
## Phi-statistics:  
## CapeMendocino.in.GLOBAL (Phi_CT)          pop1.in.GLOBAL (Phi_ST)  
##              0.54349612              0.51843824  
##   pop1.in.CapeMendocino (Phi_SC)  
##              -0.05489084  
##  
## Variance coefficients:  
##              a              b              c  
## 17.13131 17.33158 34.43478
```

```
# You can see the output above, but we can re-calculate it just to be sure.
```

```
FCT<-amova_out$varcomp[1,1]/sum(amova_out$varcomp[,1])
```

```
FCTp<-amova_out$varcomp[1,2]
```

```
FSC<-amova_out$varcomp[2,1]/(amova_out$varcomp[2,1]+amova_out$varcomp[3,1])
```

```
FSCp<-amova_out$varcomp[2,2]
```

```
FST<-(amova_out$varcomp[1,1]+amova_out$varcomp[2,1])/sum(amova_out$varcomp[,1])
```

```
FSTp<-NA
```

```
result<-c(FCT,FSC,FST)
```