Macropodoid Modelling

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Getting the Data Together

Now read in the tree files we'll be working with

```
#min.tree <- read.nexus("/Users/Ian/Google.Drive/ANU Herp Work/Macropod_Dating/FossilUncertainty/Macro_
#mean.tree <- read.nexus("/PATH/Macro_MeanAges_CON.tre")</pre>
#max.tree <- read.nexus("/PATH/Macro_MaxAges_CON.tre")</pre>
#cp.min <- read.nexus("/PATH/Macro_CP_MinAges_CON.tre")</pre>
#cp.mean <- read.nexus("/PATH/Macro_CP_EstAges_CON.tre")</pre>
#cp.max <- read.nexus("/PATH/Macro_CP_MaxAges_CON.tre")</pre>
#sampled.trees <- read.tree("~/Google.Drive/ANU/Macropod_Dating/REAL_Run5_AllSchemes_508trees_Macropo
sampled.trees <- read.tree("Trees/Tree_Span.trees")</pre>
\textit{\#empirical.trees} \textit{ $<-$ read.tree("~/Google.Drive/ANU/Macropod_Dating/MODEL110\_Sampled_Run2\_Macropodinae.tree) and the property of the p
empirical.trees <- read.tree("Trees/Macropodinae_SampledAges.trees") # Don't know which set of trees th
                                               <- read.tree("~/Google.Drive/ANU/Macropod_Dating/REAL_Run4_Fossil_519trees_Macropodina</pre>
fossil.trees <- read.tree("Trees/Macropodinae_Fossil.trees")</pre>
#consensus.tree <- read.nexus("~/Google.Drive/ANU/Macropod_Dating/Operators/REAL_Macropodinae_ALL/Run5_
consensus.tree <-read.nexus("Trees/Macropodinae_Consensus.tre")</pre>
```

Choose the current tree we want to work with

```
tree <- consensus.tree
And the hypsodonty data
# raw data for all the macropodoid taxa
                  <- read.csv("Data/CrownHeight_Data.csv", header=T)</pre>
all.HI
```

```
# or just species means for the Macropodinae
hypsodonty.index <- read.csv("Data/Macropodinae_Hypsodonty_Data.csv", header=T)
head(hypsodonty.index)
##
                                 ΗI
                                                   HI Source Diet Guild Guild Source
## 1 Baringa_nelsonensis 1.3259480 Couzens & Prideaux 2018
                                                                 Browser Dawson 2006
## 2
        Baringa_sp_indet 1.1395682 Couzens & Prideaux 2018
                                                                 Browser Dawson 2006
## 3
          Bohra_bandharr 1.0119760 Couzens & Prideaux 2018
                                                                 Unknown
                                                                                  <NA>
## 4
        Bohra_illuminata 0.9373134 Couzens & Prideaux 2018
                                                                 Unknown
                                                                                  <NA>
## 5
        Bohra_nullarbora 1.0722913 Couzens & Prideaux 2018
                                                                 Unknown
                                                                                  <NA>
## 6
          Bohra_sp_indet 1.2234637 Couzens & Prideaux 2018
                                                                 Unknown
                                                                                  <NA>
##
     Alt_Diet_Guild
                        \min_{HI}
                                  \max_{HI}
                                           mean_HI lognormal_mean_error
## 1
               <NA> 1.1399594 1.4338624 1.3259480
                                                               0.25580763
               <NA> 0.7931034 1.3394077 1.1395682
## 2
                                                               0.05568965
               <NA> 1.0119760 1.0119760 1.0119760
## 3
                                                               0.01190490
               <NA> 0.9373134 0.9373134 0.9373134
## 4
                                                              -0.06473755
## 5
               <NA> 0.9965096 1.1480730 1.0722913
                                                               0.06730667
## 6
               <NA> 1.2234637 1.2234637 1.2234637
                                                               0.20168592
##
     lognormal_sd_error
## 1
             0.11399677
## 2
             0.24949454
## 3
                      NΑ
## 4
                      NΑ
## 5
             0.07058468
## 6
                      NA
The C4 plant reconstruction data from Andrae
enviro.data <- read.csv("Data/Andrae_S1.csv", header=T)</pre>
  grass.data <- enviro.data[1:25, c(1,3)]</pre>
  # this will also remove the last sample (from 9.5 mya) which is spurious, and keep just the mean esti
  # I broke the code somewhere, so let's just make an extra bit of data at the far end of the timescale
  grass.data <- rbind(grass.data, data.frame(Age = 15, C4_recon_mean = 5))</pre>
head(enviro.data); head(grass.data)
     Age Age_Error C4_recon_mean C4_recon_lower C4_recon_upper
##
## 1 1.0
             0.002
                             59.2
                                             36.7
                                                             81.6
## 2 2.0
             0.001
                             36.6
                                             11.8
                                                             61.3
## 3 2.5
             0.008
                             36.0
                                             11.2
                                                             60.8
## 4 2.8
                             35.4
                                             10.5
                                                             60.2
             0.005
## 5 2.8
             0.013
                             21.8
                                             0.0
                                                             48.1
## 6 3.0
             0.005
                             38.8
                                             14.3
                                                             63.3
     Age C4_recon_mean
## 1 1.0
                  59.2
## 2 2.0
                  36.6
## 3 2.5
                  36.0
## 4 2.8
                  35.4
## 5 2.8
                   21.8
## 6 3.0
                  38.8
And the dust flux data from Andrae
flux.data <- read.csv("Data/Aeolian_Flux.csv", header=T)</pre>
# I broke the code somewhere, so let's just make an extra bit of data at the far end of the timescale
flux.data <- rbind(flux.data, data.frame(Age = 15, A_Flux = 10))</pre>
```

```
head(flux.data)
                   A_Flux
            Age
## 1 0.05931646 108.3949
## 2 0.10874684 113.8622
## 3 0.20760760 118.6517
## 4 0.25703797 121.9088
## 5 0.35589873 109.4722
## 6 0.40532911 116.3859
As well as the paleotemperature data
data(InfTemp)
head(InfTemp)
       Age Temperature
## 1 0.000
              3.902176
## 2 0.000
               2.900296
## 3 0.002
              4.309984
## 4 0.002
              5.172534
## 5 0.004
              3.733446
## 6 0.004
               4.309984
Trim tree and data down to overlapping taxa
# extract the taxa that are in both the tree and
overlaps <- intersect(tree$tip.label, unique(hypsodonty.index$Taxon))</pre>
macro.tree <- drop.tip(tree, setdiff(tree$tip.label, overlaps))</pre>
#macro.tree <- lapply(tree, drop.tip, tip=tip.drops) # if you're using a set of trees (fossil, sampled)
trim.data <- dplyr::filter(hypsodonty.index, Taxon %in% overlaps)</pre>
macro.HI <- trim.data[,2]; names(macro.HI) <- trim.data[,1]; geiger::name.check(macro.tree, macro.HI)
## [1] "OK"
macro.HI
##
                                              Bohra illuminata
            Baringa_nelsonensis
##
                                                      0.9373134
                       1.3259480
##
                                          Dendrolagus_dorianus
       Dendrolagus_bennettianus
##
                       1.0115887
                                                      0.9400000
##
        Dendrolagus_goodfellowi
                                           Dendrolagus_inustus
##
                       0.8500000
                                                      0.8899909
##
          Dendrolagus_lumholtzi
                                         Dendrolagus_matschiei
##
                       0.9342720
                                                      0.9100000
##
               Dorcopsis_hageni
                                             Dorcopsis_veterum
##
                       0.880000
                                                      1.0200000
##
          Dorcopsoides_fossilis
                                         Dorcopsulus_vanheurni
##
                       0.7554314
                                                      0.9793040
##
               Kurrabi mahoneyi
                                   Lagorchestes conspicillatus
##
                       1.4331210
                                                      1.1378107
##
          Lagorchestes hirsutus
                                        Lagostrophus_fasciatus
##
                       1.2047786
                                                      1.1600000
##
                 Macropus_agilis
                                          Macropus_antilopinus
##
                       1.1692005
                                                      1.2300000
##
               Macropus eugenii
                                          Macropus fuliginosus
##
                       1.1109478
                                                      1.3630443
```

Macropus_irma

##

Macropus_giganteus

```
##
                       1.3300000
                                                      1.1372544
##
                 Macropus_parma
                                                Macropus_parryi
##
                       1.3176042
                                                      1.3475744
##
                 Macropus_pavana
                                             Macropus_robustus
##
                       1.1823511
                                                      1.2418003
##
           Macropus rufogriseus
                                                 Macropus rufus
##
                       1.3500000
                                                      1.3978825
           Onychogalea_fraenata
##
                                         Onychogalea_unguifera
##
                       1.5000000
                                                      1.2694731
##
            Peradorcas_concinna
                                           Petrogale_assimilis
##
                       1.222222
                                                      1.1667479
##
           Petrogale_brachyotis
                                            Petrogale_inornata
##
                       1.0074257
                                                      1.6500000
##
            Petrogale_lateralis
                                         Petrogale_penicillata
##
                       1.2076257
                                                      1.2858835
##
       Petrogale_purpureicollis
                                         Petrogale_rothschildi
##
                       1.8200000
                                                      1.4000000
##
            Petrogale_xanthopus Prionotemnus_palankarinnicus
##
                       1.2100000
                                                      1.0221543
##
               Protemnodon anak
                                            Setonix brachyurus
##
                       1.0869565
                                                      1.0196716
##
         Thylogale_billardierii
                                               Thylogale_brunii
##
                       1.1231331
                                                      1.0000000
##
           Thylogale_stigmatica
                                               Thylogale_thetis
##
                       1.1895715
                                                      1.1652379
##
               Wallabia bicolor
##
                       1.1218826
```

If you're working with the raw data, trim tree and data down to just Macropodinae

```
macros <- dplyr::filter(all.HI, Higher_tax == "Macropodinae")
overlaps <- intersect(tree$tip.label, unique(all.HI$Taxon))
trim.tree <- drop.tip(tree, setdiff(tree$tip.label, overlaps))
trim.raw <- filter(macros, Taxon %in% overlaps)</pre>
```

create a tibble to get the species means (if you haven't done this already)

```
library(dplyr)
sp.means <- trim.raw %>%
group_by(Taxon) %>%
summarise_at(vars(H_HYPCD/PW), mean)
#write.csv(sp.means, row.names=FALSE, file="/PATH/CrownHeight_Macropodinae_spMEANS.csv") # uncomment
```

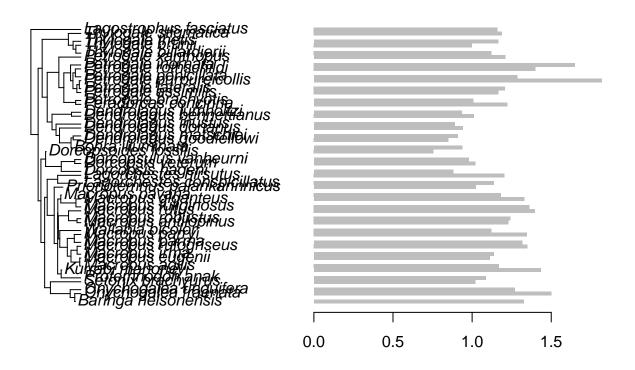
Visualizing Our Data

Let's quickly visualize the data in a few different ways to get an idea of what's going on

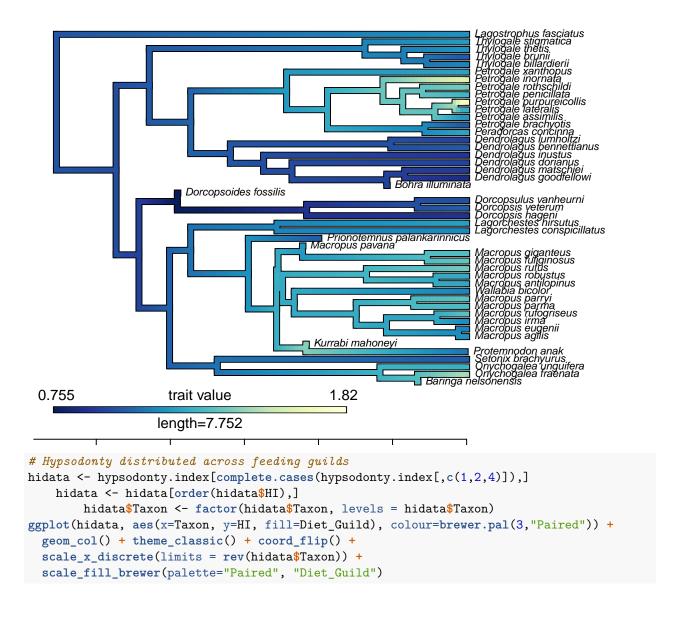
```
# Barplot of trait value
plotTree.barplot(macro.tree, macro.HI, args.barplot=list(beside=TRUE, border=F))
## Warning in space + width: longer object length is not a multiple of shorter
## object length
## Warning in space + width: longer object length is not a multiple of shorter
## object length
```

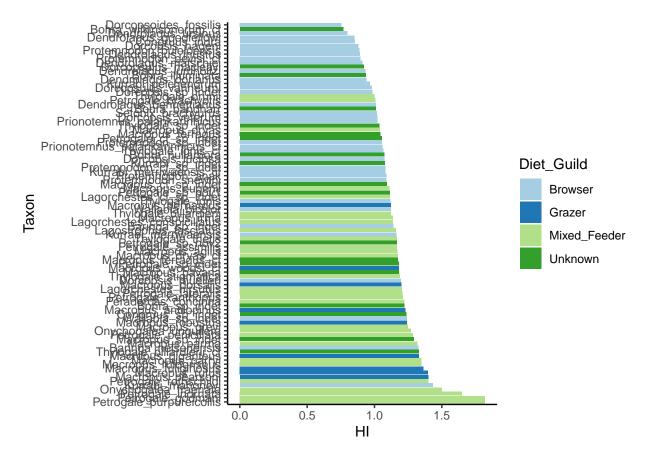
```
## Warning in space + width: longer object length is not a multiple of shorter
## object length
## Warning in space + width: longer object length is not a multiple of shorter
```

object length



```
# Continuous trait map
obj1 <- contMap(macro.tree, macro.HI, plot=FALSE, outline=F);
   n<-length(obj1$cols);
   obj1$cols[1:n] <- rev(colorRampPalette(brewer.pal(9, "YlGnBu"))(n));
plot(obj1,legend=0.7*max(nodeHeights(obj1$tree)),
      fsize=c(0.7,0.9), lwd=5, border=F); axisPhylo(1, backward=T)</pre>
```





Now we can look at the time-sampled variables

```
# make a plot of the grass data

pp <- ggplot(enviro.data[1:25,], aes(Age)) +
    geom_ribbon(aes(ymin = C4_recon_lower, ymax = C4_recon_upper), fill = "light green") +
    geom_line(aes(y = C4_recon_mean), color="DarkGreen") + scale_x_reverse() + theme_classic() +
    coord_cartesian(xlim = c(0, 20), ylim = c(0,80), expand = FALSE)

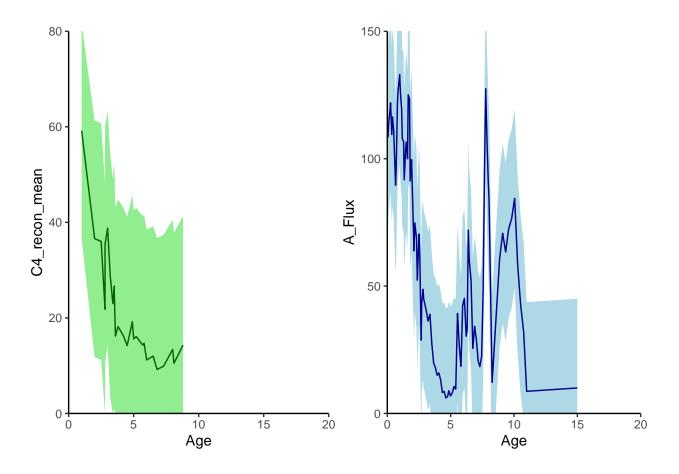
#qq <- gggeo_scale(pp, dat="epochs") # if you want to plot a geological timescale

# make a plot of the flux data

rr <- ggplot(flux.data, aes(Age)) +
    geom_ribbon(aes(ymin = A_Flux-35, ymax = A_Flux+35), fill = "light blue") +
    geom_line(aes(y = A_Flux), color="DarkBlue") + scale_x_reverse() + theme_classic() +
    coord_cartesian(xlim = c(0, 20), ylim = c(0,150), expand = FALSE)

#ss <- gggeo_scale(rr, dat="epochs") # if you want to plot a geological timescale

grid.arrange(pp, rr, nrow=1)</pre>
```



Fitting Models of Trait Evolution

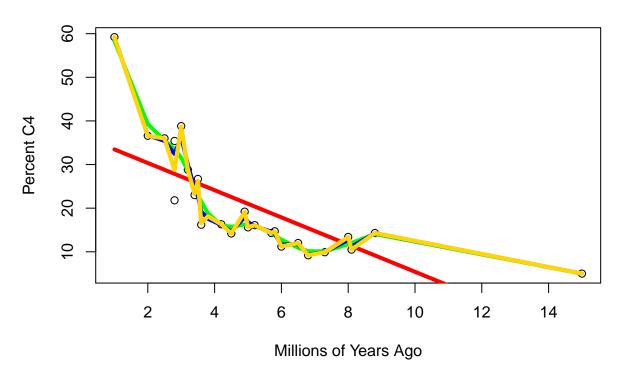
Correlative models like the environmental models in RPANDA will be sensitive to the amount of smoothing to the trend line of the input data (see Clavel & Morlon, PNAS). To address this, we'll create a function that searches for the optimum smoothness of the trend by fitting a set of values.

We can have a look at what this smoothing actually does to our data. We can come back to look at these once we get the optimum fits for our models.

```
grass.spline0 <- sm.spline(x=grass.data$Age, y=grass.data$C4_recon_mean, df=0)
grass.spline10 <- sm.spline(x=grass.data$Age, y=grass.data$C4_recon_mean, df=10)
grass.spline20 <- sm.spline(x=grass.data$Age, y=grass.data$C4_recon_mean, df=20)
grass.spline30 <- sm.spline(x=grass.data$Age, y=grass.data$C4_recon_mean, df=30)
grass.spline40 <- sm.spline(x=grass.data$Age, y=grass.data$C4_recon_mean, df=40)
grass.spline50 <- sm.spline(x=grass.data$Age, y=grass.data$C4_recon_mean, df=50)

plot(grass.data, main="C4 Grass Reconstruction Through Time", xlab="Millions of Years Ago", ylab="Percentions(grass.spline0, col="red", lwd=4)
lines(grass.spline10, col="green", lwd=4)
lines(grass.spline20, col="blue", lwd=4)
lines(grass.spline30, col="yellow", lwd=4)
lines(grass.spline40, col="violet", lwd=4)
lines(grass.spline50, col="gold", lwd=4)
lines(grass.spline50, col="gold", lwd=4)</pre>
```

C4 Grass Reconstruction Through Time

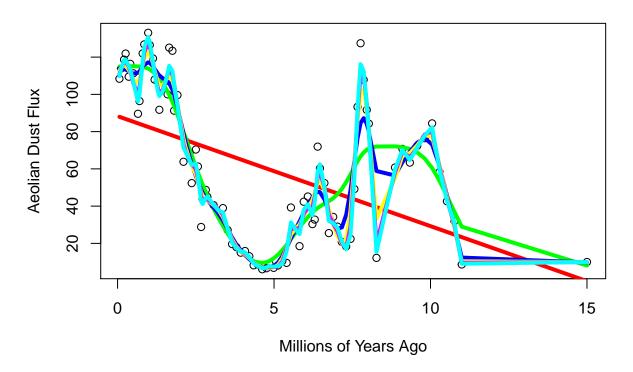


```
flux.spline0 <- sm.spline(x=flux.data$Age, y=flux.data$A_Flux, df=0)
flux.spline10 <- sm.spline(x=flux.data$Age, y=flux.data$A_Flux, df=10)
flux.spline20 <- sm.spline(x=flux.data$Age, y=flux.data$A_Flux, df=20)
flux.spline30 <- sm.spline(x=flux.data$Age, y=flux.data$A_Flux, df=30)
flux.spline40 <- sm.spline(x=flux.data$Age, y=flux.data$A_Flux, df=40)
flux.spline50 <- sm.spline(x=flux.data$Age, y=flux.data$A_Flux, df=50)

plot(flux.data, main="Aeolian Dust Flux Through Time", xlab="Millions of Years Ago", ylab="Aeolian Dust lines(flux.spline0, col="red", lwd=4)
lines(flux.spline10, col="green", lwd=4)
lines(flux.spline20, col="blue", lwd=4)</pre>
```

```
lines(flux.spline30, col="yellow", lwd=4)
lines(flux.spline40, col="magenta", lwd=4)
lines(flux.spline50, col="cyan", lwd=4)
```

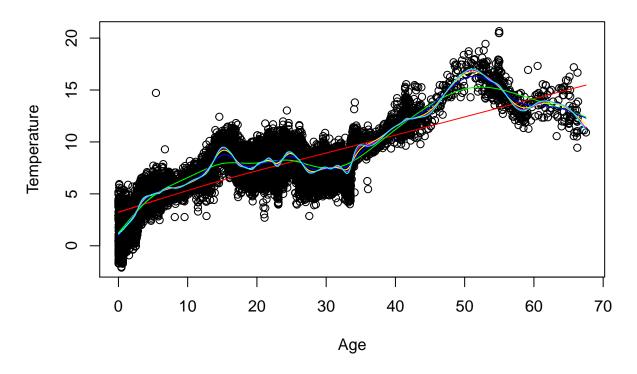
Aeolian Dust Flux Through Time



```
data(InfTemp)
env.spline0 <- sm.spline(x=InfTemp$Age, y=InfTemp$Temperature, df=0)
env.spline10 <- sm.spline(x=InfTemp$Age, y=InfTemp$Temperature, df=10)
env.spline20 <- sm.spline(x=InfTemp$Age, y=InfTemp$Temperature, df=20)
env.spline30 <- sm.spline(x=InfTemp$Age, y=InfTemp$Temperature, df=30)
env.spline40 <- sm.spline(x=InfTemp$Age, y=InfTemp$Temperature, df=40)
env.spline50 <- sm.spline(x=InfTemp$Age, y=InfTemp$Temperature, df=50)

plot(InfTemp, main="Paleotemperature Through Time")
lines(env.spline0, col="red")
lines(env.spline10, col="green")
lines(env.spline20, col="blue")
lines(env.spline30, col="yellow")
lines(env.spline40, col="magenta")
lines(env.spline50, col="cyan")</pre>
```

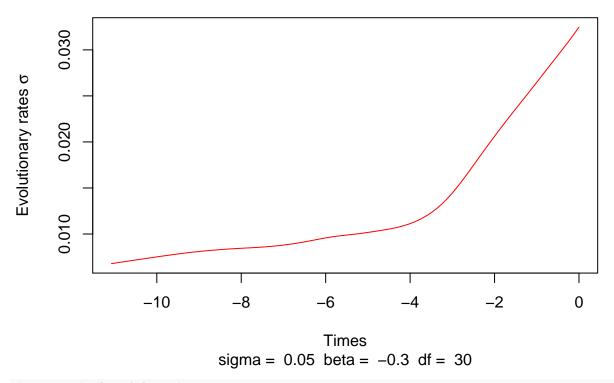
Paleotemperature Through Time



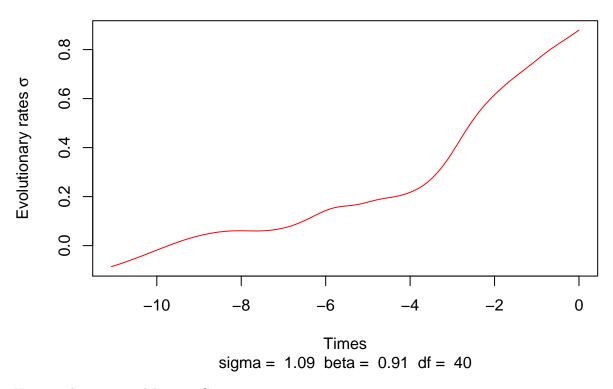
Next we'll fit a number of models to our tree and data

Start with the environmental model of paleotemperature. You can designate the number of cores and the amount of smoothing

EnvExp; **AICc** = 14.21

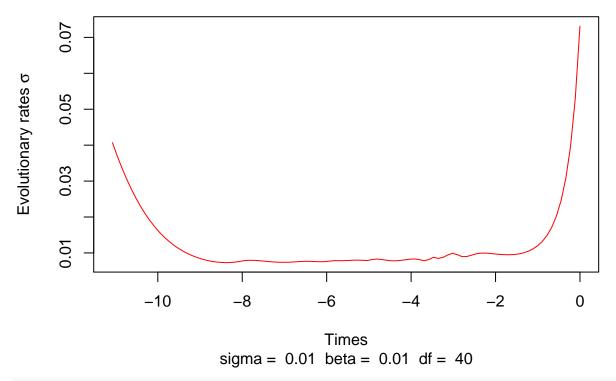


EnvLin; AICc = 82.08

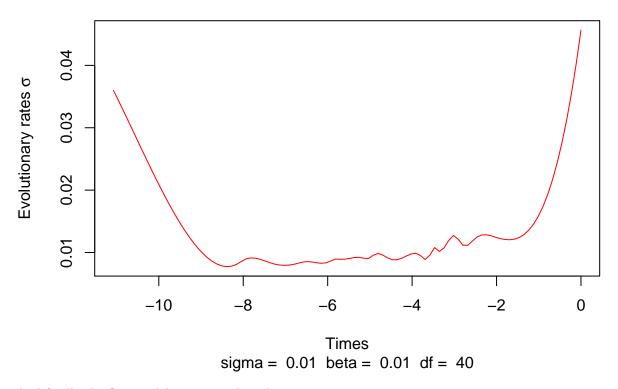


Next up the grass model, using C4 reconstructions.

EnvExp; AICc = 3.28



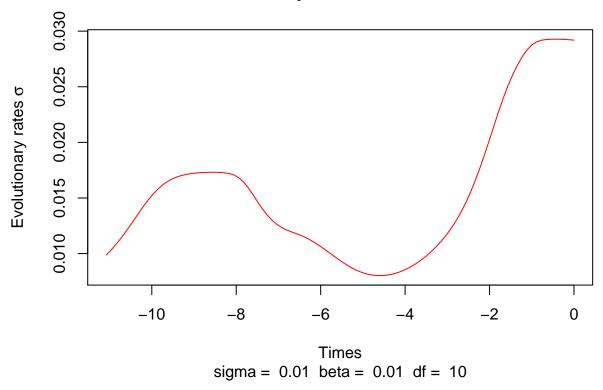
EnvLin; AICc = 6.87



And finally the flux models, using aeolian dust measurements.

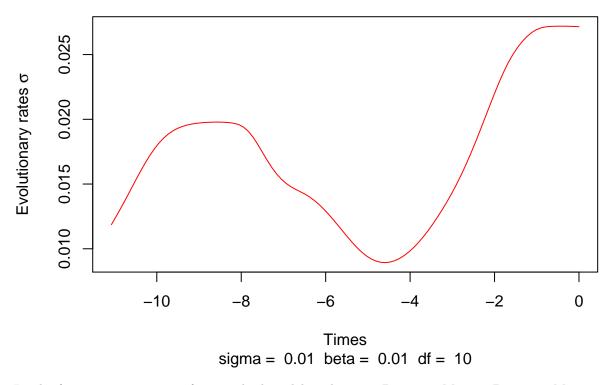
```
FLUXexp <- best.smoothing(macro.tree, macro.HI, time.data=flux.data, degrees=c(10,20,30,40,50), model="EnvExp", cores=5)
```

EnvExp; AICc = 15.03



FLUXlin <- best.smoothing(macro.tree, macro.HI, time.data=flux.data, degrees=c(10,20,30,40,50), model="EnvLin", cores=5)

EnvLin; AICc = 16.2



Lastly, for comparison, run a few standard models. These are Brownian Motion, Brownian Motion with a Trend, and Early Burst.

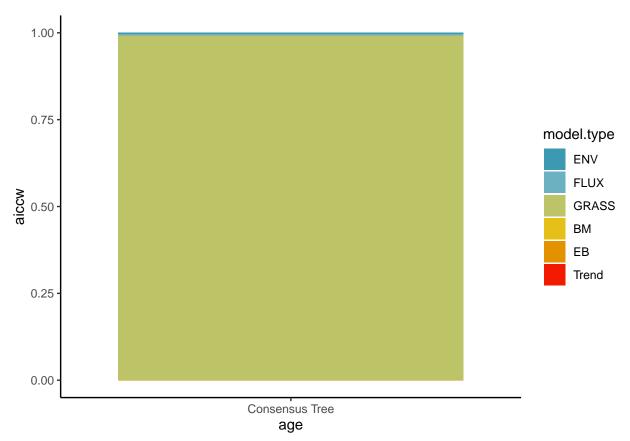
```
BM_res <- fitContinuous(macro.tree, macro.HI, model="BM")
trend_res <- fitContinuous(macro.tree, macro.HI, model="drift")
EB_res <- fitContinuous(macro.tree, macro.HI, model="EB")</pre>
```

Compare the models with AICc, and check differences across the trees

```
model_FIT <- c(ENVexp$best.result$aicc, ENVlin$best.result$aicc,</pre>
                GRASSlin$best.result$aicc, GRASSexp$best.result$aicc,
               FLUXexp$best.result$aicc, FLUXlin$best.result$aicc,
               BM_res$opt$aicc, trend_res$opt$aicc, EB_res$opt$aicc);
names(model_FIT) <- c("ENVexp", "ENVlin", "GRASSlin", "GRASSexp", "FLUXexp", "FLUXlin", "BM", "Trend",</pre>
aic.w(model_FIT)
       ENVexp
                   ENVlin
                            GRASSlin
                                        GRASSexp
                                                                 FLUXlin
                                                     FLUXexp
## 0.00360654 0.00000000 0.14169000 0.85030512 0.00239677 0.00133317 0.00035646
## 0.00019825 0.00011369
fit.aic <- as.data.frame(as.vector(aic.w(model_FIT)));</pre>
    fit.aic$model <- names(model_FIT);</pre>
        colnames(fit.aic) <- c("aiccw", "model");</pre>
            fit.aic$age <- "Consensus Tree"</pre>
```

Quickly collapse models from the same data

Then plot the model fits as AICc weights



Fitting Models to Our Data as a Function of Time

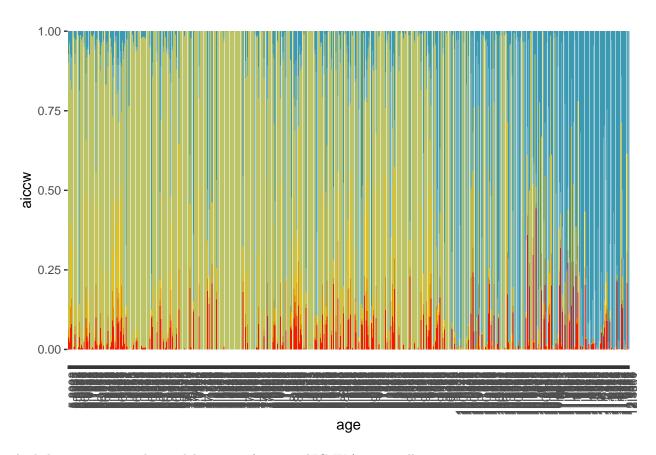
Ok, now that we've fit the models to a given tree, we want to fit the models to lots of trees of different ages, shapes, etc. Now we'll run a loop across all of these trees, to fit the models to each one. It may take a little while.

```
# MAKE THIS EVAL=TRUE IF YOU WANT TO DO THIS BIT FOR REAL
tree.span <- fossil.trees
mean.data <- fossil.HI
# Fit all the models to a series of trees!
all.aics <- NULL; all.results <- NULL; timer <- progress_estimated(length(tree.span))
for (k in 1:length(tree.span)){</pre>
```

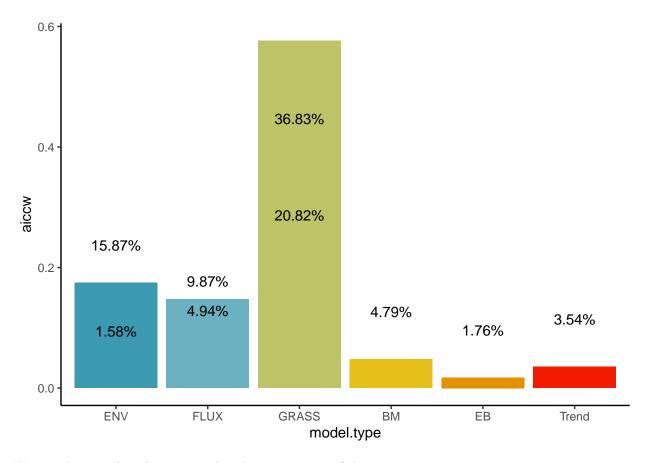
```
int.results <- NULL</pre>
# Fit a number of models to the data (ENV, GRASS, BM, EB, Trend, Drift)
ENVexp <- best.smoothing(tree.span[[k]], mean.data, time.data=InfTemp,</pre>
                          degrees=c(10,20,30,40,50), model="EnvExp", cores=5);
                          int.results[["ENVexp"]] <- ENVexp$best.result;</pre>
ENVlin <- best.smoothing(tree.span[[k]], mean.data, time.data=InfTemp,</pre>
                          degrees=c(10,20,30,40,50), model="EnvLin", cores=5);
                          int.results[["ENVlin"]] <- ENVlin$best.result;</pre>
GRASSexp <- best.smoothing(tree.span[[k]], mean.data, time.data=grass.data,</pre>
                            degrees=c(30,40,50), model="EnvExp", cores=3);
                            int.results[["GRASSexp"]] <- GRASSexp$best.result;</pre>
GRASSlin <- best.smoothing(tree.span[[k]], mean.data, time.data=grass.data,</pre>
                            degrees=c(30,40,50), model="EnvLin", cores=3);
                            int.results[["GRASSlin"]] <- GRASSlin$best.result;</pre>
FLUXexp <- best.smoothing(tree.span[[k]], mean.data, time.data=flux.data,
                           degrees=c(10,20,30,40,50), model="EnvExp", cores=5);
                           int.results[["FLUXexp"]] <- FLUXexp$best.result;</pre>
FLUXlin <- best.smoothing(tree.span[[k]], mean.data, time.data=flux.data,</pre>
                           degrees=c(10,20,30,40,50), model="EnvLin", cores=5);
                           int.results[["FLUXlin"]] <- FLUXlin$best.result;</pre>
          <- fitContinuous(tree.span[[k]], mean.data, model="BM");</pre>
BM_res
             int.results[["BM"]] <- BM res</pre>
trend_res <- fitContinuous(tree.span[[k]], mean.data, model="drift");</pre>
             int.results[["Trend"]] <- trend_res</pre>
EB_res
          <- fitContinuous(tree.span[[k]], mean.data, model="EB");</pre>
             int.results[["EB"]] <- EB_res</pre>
curr_tree_FIT <- c(ENVexp$best.result$aicc, ENVlin$best.result$aicc,</pre>
                   GRASSexp$best.result$aicc, GRASSlin$best.result$aicc,
                   FLUXexp$best.result$aicc, FLUXlin$best.result$aicc,
                   BM res$opt$aicc, trend res$opt$aicc, EB res$opt$aicc);
names(curr_tree_FIT) <- c("ENVexp", "ENVlin", "GRASSexp", "GRASSlin",</pre>
                            "FLUXexp", "FLUXlin", "BM", "Trend", "EB")
curr_tree_SIG <- c(ENVexp$best.result$param[[1]], ENVlin$best.result$param[[1]],</pre>
                    GRASSexp$best.result$param[[1]], GRASSlin$best.result$param[[1]],
                    FLUXexp$best.result$param[[1]], FLUXlin$best.result$param[[1]],
                    BM_res$opt$sigsq, trend_res$opt$sigsq, EB_res$opt$sigsq);
names(curr_tree_SIG) <- c("ENVexp", "ENVlin", "GRASSexp", "GRASSlin",</pre>
                            "FLUXexp", "FLUXlin", "BM", "Trend", "EB")
curr_tree_PAR <- c(ENVexp$best.result$param[[2]], ENVlin$best.result$param[[2]],</pre>
                    GRASSexp$best.result$param[[2]], GRASSlin$best.result$param[[2]],
                    FLUXexp$best.result$param[[2]], FLUXlin$best.result$param[[2]],
                    NA, trend_res$opt$slope, EB_res$opt$a);
names(curr_tree_PAR) <- c("ENVexp", "ENVlin", "GRASSexp", "GRASSlin",</pre>
                           "FLUXexp", "FLUXlin", "BM", "Trend", "EB")
curr.aic <- as.data.frame(as.vector(aic.w(curr_tree_FIT)));</pre>
```

```
curr.aic$model <- names(curr_tree_FIT); colnames(curr.aic) <- c("aiccw", "model");</pre>
            curr.aic$age <- round(max(nodeHeights(tree.span[[k]])), 4)</pre>
                 curr.aic$tree <- k</pre>
                     curr.aic$sigsq <- curr_tree_SIG</pre>
                         curr.aic$par <- curr_tree_PAR</pre>
  all.results[[k]] <- int.results
  curr.aic$model.type <- c("ENV", "ENV", "GRASS", "GRASS",</pre>
                            "FLUX", "FLUX", "BM", "Trend", "EB")
  all.aics <- rbind.data.frame(all.aics, curr.aic);</pre>
  print(timer$tick())
Save the file externally:
saveRDS(all.aics.
                     file="/PATH/Fossil Trees Model Fitting AICCs.RDS")
saveRDS(all.results, file="/PATH/Fossil_Trees_Model_Fitting_Results.RDS")
#saveRDS(all.aics, file="/PATH/Model_Fitting_AICCs.RDS")
Or skip the work and read in the file instead:
all.aics <- readRDS("Data/FINAL SAMPLED Model Fitting AICCs.RDS")
all.aics$model.type <- factor(all.aics$model.type,</pre>
                        levels=c("ENV", "FLUX", "GRASS", "BM", "EB", "Trend"))
all.aics$age <- as.factor(all.aics$age)</pre>
sampled.res <- (ggplot(all.aics)</pre>
 + geom_bar(aes(y=aiccw, x=age, fill=model.type), stat="identity")
  + theme(axis.text.x=element_text(angle=90, hjust=1),
          panel.background=element_blank(), legend.position="none")
  + scale_fill_manual( values=wes_palette("Zissou1", 6, "continuous")))
```

sampled.res



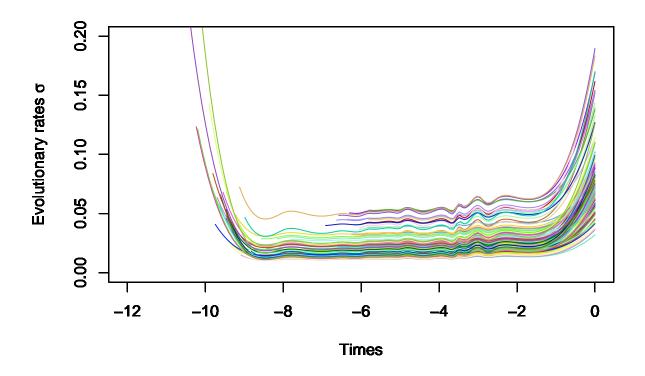
And then summarize the model support (average AICcWt) across all trees



We can also visualize the estimated evolutionary rates of the trait. First create an adjusted version of the plotting function from RPANDA:

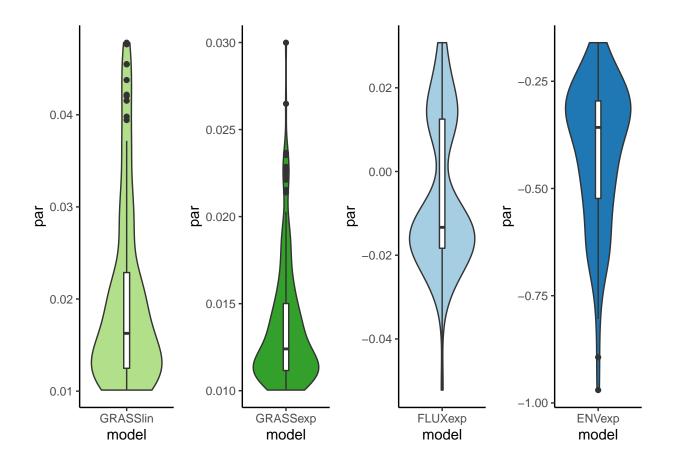
```
# Adjust the RPANDA plotting function so we can fix the axes, and do a bunch of plots
plot.fixed_t_env <- function (x, steps = 100,</pre>
                                 xlim=c(-10,0), ylim=c(0,1), linecol="red", ...)
{
  if (is.function(x$model)) {
    fun_temp <- function(x, temp, model, param) {</pre>
      rate_fun <- function(x) {</pre>
        model(x, temp, param)
      rate <- rate_fun(x)
      return(rate)
    }
  else if (x$model == "EnvExp") {
    fun_temp <- function(x, temp, model, param) {</pre>
      sig <- param[1]</pre>
      beta <- param[2]</pre>
      rate <- (sig * exp(beta * temp(x)))</pre>
      return(rate)
    }
  }
  else if (x$model == "EnvLin") {
    fun_temp <- function(x, temp, model, param) {</pre>
      sig <- param[1]</pre>
```

I'll plot just the results from the GRASS-linear model, but you could do this for all.



Also plot the beta value of the correlation between rate and the variable

```
# And plot the beta values for preferred models
col.pal <- brewer.pal(8, "Paired")
beta.values <- filter(all.aics, model %in% c("GRASSexp", "FLUXexp", "GRASSlin", "ENVexp") & aiccw >= 0..
ge_beta <- filter(beta.values, model=="GRASSexp");
ge <- ggplot(ge_beta, aes(x=model, y=par)) + geom_violin(fill=col.pal[[4]]) + theme_classic() + geom_body
gl_beta <- filter(beta.values, model=="GRASSlin");
gl <- ggplot(gl_beta, aes(x=model, y=par)) + geom_violin(fill=col.pal[[3]]) + theme_classic() + geom_body
fe_beta <- filter(beta.values, model=="FLUXexp");
fe <- ggplot(fe_beta, aes(x=model, y=par)) + geom_violin(fill=col.pal[[1]]) + theme_classic() + geom_body
ee_beta <- filter(beta.values, model=="ENVexp");
ee <- ggplot(ee_beta, aes(x=model, y=par)) + geom_violin(fill=col.pal[[2]]) + theme_classic() + geom_body
gridExtra::grid.arrange(gl, ge, fe, ee, nrow=1)</pre>
```



Comparing Estimated Node Ages Across Dating Schemes

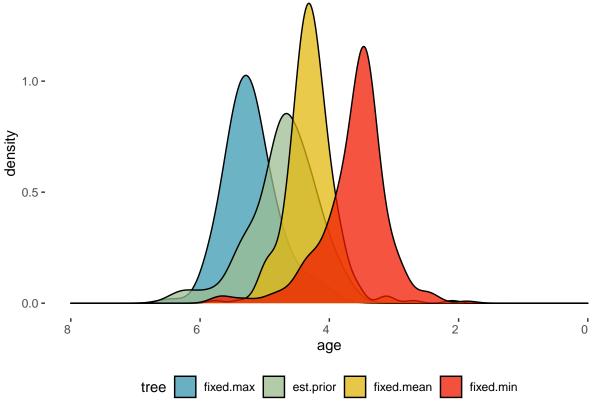
Make a quick function to get the DEPTH of a node (from present), instead of the HEIGHT (from root)

MRCA.depth <- function(phy){max(nodeHeights(phy)) - findMRCA(phy, tips=c("Macropus_irma", "Wallabia_bic")

Prepare trees for comparison of ages

```
# Prepare trees for comparison of ages
#max.age.trees <- read.nexus("~/Google.Drive/ANU/Macropod_Dating/Operators/REAL_Macropodinae_ALL/Altern
max.age.trees <- read.nexus("Trees/Macropodinae_MaxAges.trees")</pre>
max.ages <- max.age.trees[sample(1:length(max.age.trees),300)]</pre>
\# max. ages <- lapply (max. age. trees, drop. tip, tip = set diff (max. age. trees [[1]] \$ tip. label, overlaps)); \ classians the set of the
age.max <- as.data.frame(unlist(lapply(max.ages, MRCA.depth)));</pre>
age.max$tree <- "fixed.max"; colnames(age.max) <- c("age", "tree")
#mean.age.trees <- read.nexus("~/Google.Drive/ANU/Macropod_Dating/Operators/REAL_Macropodinae_ALL/Alter
mean.age.trees <- read.nexus("Trees/Macropodinae_MeanAges.trees")</pre>
mean.ages <- mean.age.trees[sample(1:length(mean.age.trees),300)]</pre>
#mean.ages <- lapply(mean.age.trees, drop.tip, tip=setdiff(mean.age.trees[[1]]$tip.label, overlaps)); c</pre>
age.mean <- as.data.frame(unlist(lapply(mean.ages, MRCA.depth)));</pre>
age.mean$tree <- "fixed.mean"; colnames(age.mean) <- c("age", "tree")
\#min. age. trees <- \ read. nexus ("~/Google.Drive/ANU/Macropod_Dating/Operators/REAL\_Macropodinae\_ALL/Alternative for the property of the p
min.age.trees <- read.nexus("Trees/Macropodinae_MinAges.trees")</pre>
min.ages <- min.age.trees[sample(1:length(min.age.trees),300)]</pre>
```

```
\#min.ages \leftarrow lapply(min.age.trees, drop.tip, tip=setdiff(min.age.trees[[1]]$tip.label, overlaps)); clastically contains the set of 
age.min <- as.data.frame(unlist(lapply(min.ages, MRCA.depth)));</pre>
age.min$tree <- "fixed.min"; colnames(age.min) <- c("age", "tree")
\#sampled500.trees <- read.nexus("~/Google.Drive/ANU/Macropod_Dating/Operators/REAL\_Macropodinae\_ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-ALL/Runces-A
#sampled500.trees <- read.nexus("~/Google.Drive/ANU/Macropod_Dating/Operators/REAL_Macropodinae_ALL/Run
\#sampled.ages \leftarrow sampled500.trees[(length(sampled500.trees)-200):length(sampled500.trees)]
sampled.ages <- read.tree("Trees/Macropodinae_SampledAges.trees")</pre>
#min.ages <- lapply(min.age.trees, drop.tip, tip=setdiff(min.age.trees[[1]]$tip.label, overlaps)); clas
age.sampled <- as.data.frame(unlist(lapply(sampled.ages, MRCA.depth))); age.sampled$tree <- "est.prior"
Combine the ages from different dating schemes, and plot them
require(wesanderson)
age.all <- rbind(age.min, age.sampled, age.mean, age.max)
age.all$tree <- factor(age.all$tree, levels=c("fixed.max","est.prior", "fixed.mean", "fixed.min"))
(ggplot(age.all, aes(x=age, fill=tree))
        + geom_density(alpha=0.75, adjust=1.5)
        + theme(axis.text.x=element_text(angle=0, hjust=1), panel.background=element_blank(), legend.position
       + scale_fill_manual(values=wes_palette("Zissou1", type="continuous", 4))
       + scale_x_reverse(lim=c(8,0)))
```



Getting Bayes Factors for Fossil Taxa

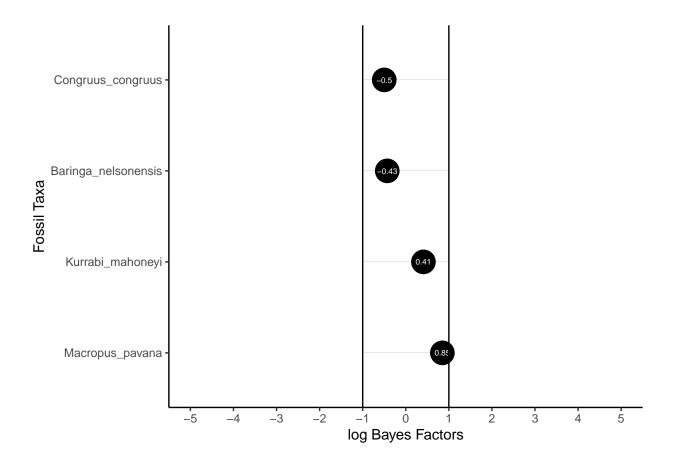
Investigating Fossil Taxa as putative Sampled Ancestors or Terminals

```
prior.trees <- read.tree("Trees/Macropodinae SampledAges PriorOnly.trees");</pre>
prior.trees <- prior.trees[(length(prior.trees)-1000):length(prior.trees)]</pre>
post.trees <- read.nexus("Trees/Macropodinae_MeanAges.trees");</pre>
post.trees <- post.trees[(length(post.trees)-1000):length(post.trees)]</pre>
Choose which tips you want information for:
fossil_taxa <- c("Baringa_nelsonensis", "Congruus_congruus",</pre>
                  "Kurrabi mahoneyi", "Macropus pavana")
Get the node numbers of the tips
nodes <- sapply(fossil taxa,function(x,y) which(y==x),y=tree$tip.label)
Then get the edge lengths for those nodes
edge.lengths <- setNames(tree$edge.length[sapply(nodes,
                function(x,y) which(y==x),y=tree$edge[,2])],names(nodes))
The faster way is to make a function to do this:
get_terminal_branchlengths <- function(phy, tipnames){</pre>
  ## Get the node numbers of the tips
  nodes <- sapply(tipnames,function(x,y) which(y==x),y=phy$tip.label)</pre>
  ## Then get the edge lengths for those nodes
  edge.lengths <- setNames(phy$edge.length[sapply(nodes,
                  function(x,y) which(y==x),y=phy$edge[,2])],names(nodes))
 return(edge.lengths)
}
Now that we've got the tips and branch lengths, we can compare the posterior to the prior
BFSA <- function(prior.phy, posterior.phy, tips){</pre>
  post <- lapply(posterior.phy, get_terminal_branchlengths, tipnames=tips);</pre>
      names(post) <- NULL; post <- unlist(post)</pre>
  names(prior)<- NULL; prior <- unlist(prior)</pre>
  BFs <- NULL
  for (j in 1:length(tips)){
    curr.tip <- subset(post, names(post)==tips[j]);</pre>
    probSA <- sum(curr.tip<=0); probTIP <- length(curr.tip)-probSA;</pre>
    curr.tip <- subset(prior, names(prior)==tips[j]);</pre>
    priorSA <- sum(curr.tip<=0); priorTIP <- length(curr.tip)-priorSA;</pre>
    curr.BF <- log((probSA * priorTIP) / (probTIP * priorSA))</pre>
    if(is.na(curr.BF)){curr.BF <- 0}</pre>
    #curr.BF <- log(probSA/(length(curr.tip)-probSA))</pre>
    names(curr.BF) <- tips[j]; curr.BF <- round(curr.BF, 2)</pre>
    BFs <- append(BFs, curr.BF)</pre>
  }
 return(BFs)
```

Orient the data appropriately

macro_BFs <- BFSA(prior.trees, post.trees, tips=fossil_taxa)</pre>

```
macro_BFs <- as.data.frame(macro_BFs) # make the vector a data frame
macro_BFs[which(macro_BFs$macro_BFs > 5),] <- 5 # change any really big (INF) numbers to 5
macro_BFs[which(macro_BFs$macro_BFs < -5),] <- -5 # change any really small (-INF) numbers to -5
macro BFs$taxa <- rownames(macro BFs); # create column with the taxon names
macro_BFs <- macro_BFs[order(macro_BFs$macro_BFs),] # reorder by BF values
# set colors for plotting
macro_BFs$color <- "black";</pre>
if(any(macro_BFs$macro_BFs > 1)){macro_BFs[which(macro_BFs$macro_BFs > 1),]$color <- "#3d98d3"}</pre>
if(any(macro_BFs$macro_BFs < -1)){macro_BFs[which(macro_BFs$macro_BFs < -1),]$color <- "#FF7175"}
macro_BFs$taxa <- factor(macro_BFs$taxa, levels=c(macro_BFs$taxa)) # set factors for plotting (NOT NECE
Then plot the data
ggplot(macro_BFs, aes(x=taxa, y=macro_BFs, label=macro_BFs)) +
  geom_ribbon(aes(ymin=-1, ymax=+1)) +
  geom_point(stat='identity', size=8, color=macro_BFs$color) +
  \# geom\_segment(aes(y = 0,
                     x = taxa,
  #
                     yend = macro_BFs,
  #
                     xend = taxa),
                 color = macro_BFs$color) +
  geom_text(color="white", size=2) +
  #labs(title="Bayes Factor Support", subtitle="for Fossil Taxa as Sampled Ancestors") +
  #ylim(-5, 5) +
  scale_y_continuous(name="log Bayes Factors", limits=c(-5,5), breaks=c(-5:5)) +
  scale_x_discrete(limits = rev(unique(sort(macro_BFs$taxa)))) + # drop this if you want to order it di
  #theme(panel.background=element_blank()) +
  geom_hline(yintercept=-1) +
  geom_hline(yintercept=1) +
  xlab("Fossil Taxa") +
  #ylab("log Bayes Factors") +
  theme_classic() +
  #theme(axis.text.y=element_blank(), axis.title.y=element_blank()) +
  coord_flip()
```



Investigating estimated ages of fossils

Create a function to pull the ages of each fossil taxon estimated

```
get.fossil.ages <- function(fossil.tips, trees){
   tree.tables <- lapply(trees, print.tree)
   fossil.tables <- lapply(1:length(tree.tables), function(x) {
      subset(tree.tables[[x]], tree.tables[[x]]$label %in% fossil.tips)
   })
   fossil.ages <- lapply(1:length(fossil.tables), function(x) {
      select(fossil.tables[[x]], label, time_bp)
   })
   final <- bind_rows(fossil.ages)
}</pre>
```

Need to create a function called "print.tree" that I borrowed some code from biogeoBEARS

Now pull out the info on those fossils

```
my.test <- get.fossil.ages(fossil.tips = fossil_taxa, trees = post.trees)

(ggplot(my.test, aes(x=time_bp, y=label, fill=..x..))
    + scale_fill_gradientn(colours=wes_palette("Zissou1"))
    + geom_density_ridges_gradient(scale=1.5)
    + scale_x_reverse()
    + theme_classic())</pre>
```

Picking joint bandwidth of 0.6

