# Sperm size figures

jeremy

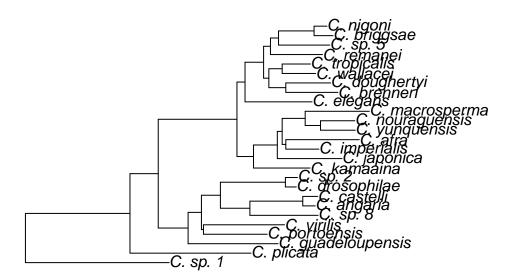
Wednesday, January 14, 2015

#### Sperm size figures

This is the output of an R markdown script - a fancy way of embedding r in a page so we can rerun/tune it as needed and then have the commands

I'm embedding, but hiding a bynch of stuff for readability - the actual code is in the .Rmd file.

Here's the tree from Kiontke et al 2011, figure 3 intial values I recieved had angaria and sp12 flipped - I've corrected it here Here's the tree:



Now read in the compiled data - we have oocytesize, spermsize, primary spermatocytesize, and oocytearea. The species names in the files are inconsistent - they are fixed now. For now, the oocyte area is using the area formulation as made by the initial sperm size formula. All response variables are logged unless otherwise noted.

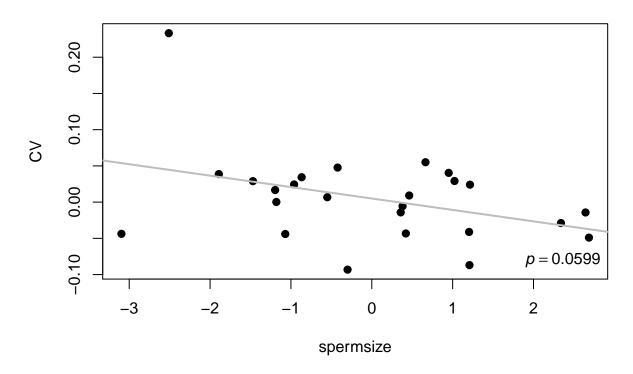
I have filtered out some of the data: Any oocytesize that is not stage -1, and any primaryspermatocyte data that is from hermaphrodites rather than males.

We now have clean data, clustered into species, strains or individuals. Variables ending in 1 are in species, 2 are strains and 3 are individuals. The non-clustered data is as is.

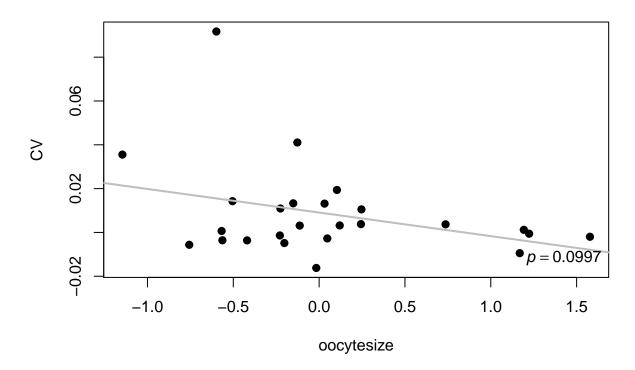
#### PIC plots - mean vs CV

Plotting the phylogenetic independent contrasts (PIC) of means of traits vs CV of traits. All means and CVs are grouped at strain level.

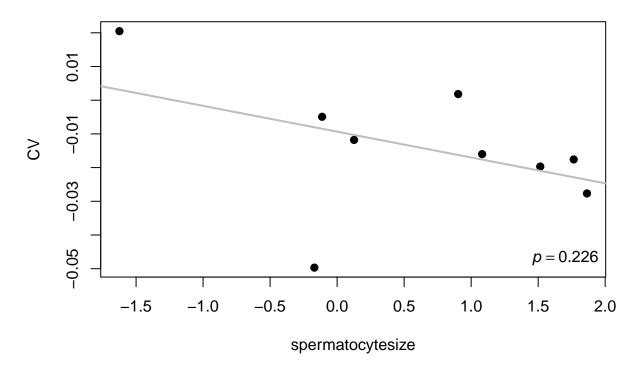
## PIC of spermsize means vs CV



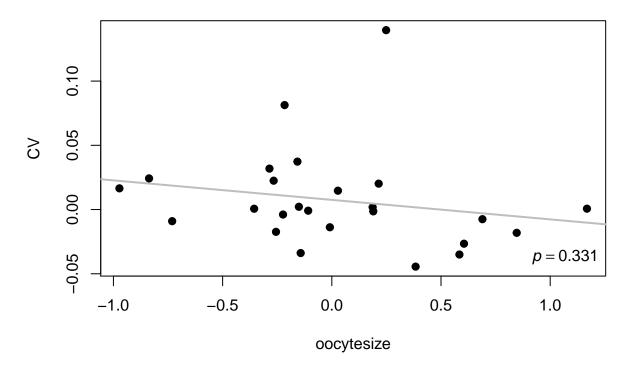
## PIC of oocytesize means vs CV



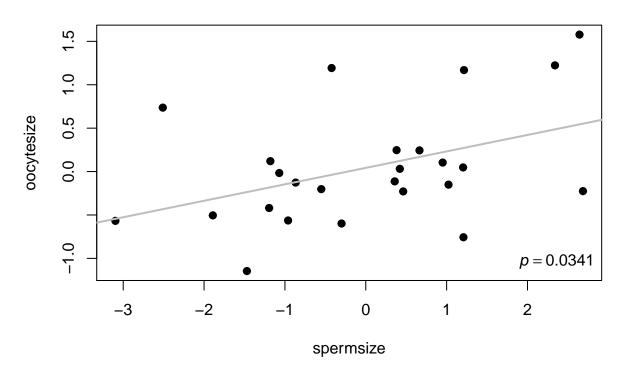
PIC of spermatocytesize means vs CV



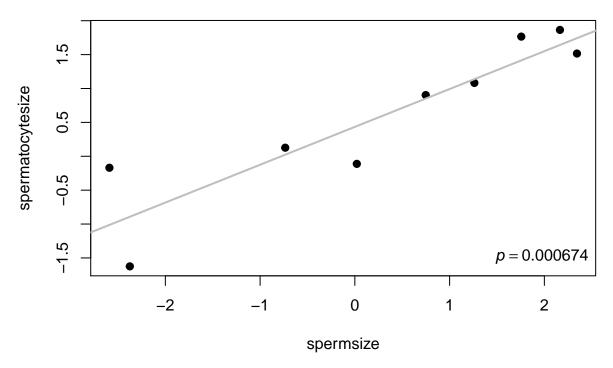
# PIC of oocytesizearea means vs CV



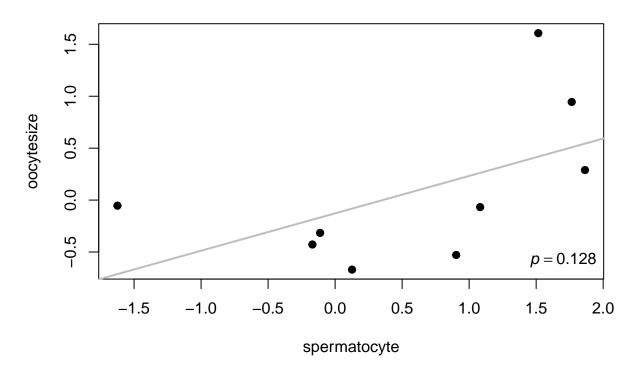
PIC of spermsize vs oocytesize



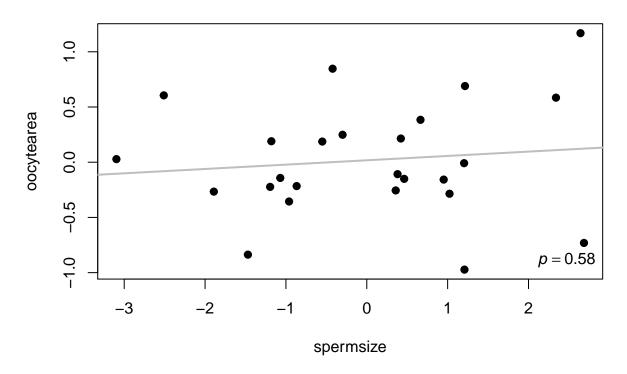
PIC of spermsize vs spermatocytesize



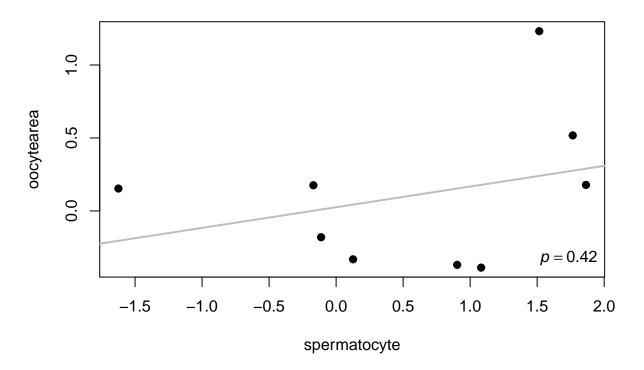
## PIC of spermatocytesize vs oocytesize



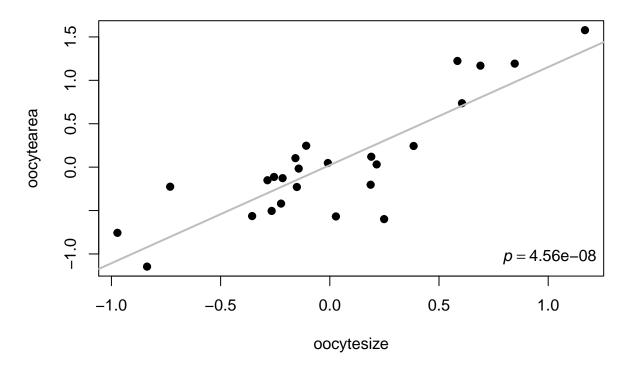
# PIC of spermsize vs oocytearea



## PIC of spermatocytesize vs oocytearea



#### PIC of oocytesize vs oocytearea



#### PGLS - testing for correlation with evolutionary models

Output are summary tables from fits. The OU model won't fit the spermatocyte fits due to overparameterisation.

```
##
## pgls(formula = meansperm ~ meanooarea, data = compsperm, lambda = "ML")
##
## Residuals:
##
                1Q Median
                                3Q
                                       Max
  -0.9643 -0.5988 0.4323
                           0.6861
                                    1.8263
##
##
## Branch length transformations:
##
## kappa [Fix]
                : 1.000
  lambda [ ML]
                : 0.000
      lower bound : 0.000, p = 1
##
##
      upper bound : 1.000, p = 0.11287
      95.0% CI
##
                 : (NA, NA)
## delta [Fix]
                : 1.000
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) -6.09037
                           6.57159 -0.9268
                                             0.3812
```

```
## meanooarea 1.46171
                          0.94975 1.5390 0.1624
##
## Residual standard error: 1.057 on 8 degrees of freedom
## Multiple R-squared: 0.2284, Adjusted R-squared: 0.132
## F-statistic: 2.369 on 1 and 8 DF, p-value: 0.1624
##
## Call:
## pgls(formula = meansperm ~ meanoo, data = compsperm, lambda = "ML")
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -1.4586 -1.1960 0.6374 1.0989 1.4686
##
## Branch length transformations:
##
## kappa [Fix] : 1.000
## lambda [ ML] : 0.895
##
     lower bound : 0.000, p = 1
     upper bound : 1.000, p = 0.79696
##
     95.0% CI : (NA, NA)
## delta [Fix] : 1.000
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) -11.31303
                           6.94067 -1.6300 0.14176
## meanoo
                1.50685
                           0.65983 2.2837 0.05177 .
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.315 on 8 degrees of freedom
## Multiple R-squared: 0.3946, Adjusted R-squared: 0.319
## F-statistic: 5.215 on 1 and 8 DF, p-value: 0.05177
##
## Call:
## pgls(formula = meansperm ~ meanprimsp, data = compsperm, lambda = "ML")
## Residuals:
      Min
               1Q Median
                               30
                                      Max
## -0.8244 -0.3185 -0.1789 0.1091 0.5614
## Branch length transformations:
##
## kappa [Fix] : 1.000
## lambda [ ML] : 0.000
##
     lower bound : 0.000, p = 1
     upper bound : 1.000, p = 0.0066141
##
     95.0% CI
               : (NA, 0.708)
## delta [Fix] : 1.000
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.63590
                         0.94589 -2.7867 0.0236793 *
```

```
## meanprimsp
              1.21417
                          0.17131 7.0873 0.0001033 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.446 on 8 degrees of freedom
## Multiple R-squared: 0.8626, Adjusted R-squared: 0.8454
## F-statistic: 50.23 on 1 and 8 DF, p-value: 0.0001033
## Generalized least squares fit by REML
     Model: meansperm ~ meanooarea
##
     Data: fullpglsframe
##
         AIC
                  BIC
                         logLik
##
     54.95884 59.67106 -23.47942
##
## Correlation Structure: corMartins
## Formula: ~1
## Parameter estimate(s):
##
     alpha
## 6.841132
##
## Coefficients:
##
                  Value Std.Error
                                     t-value p-value
## (Intercept) -3.044212 3.398811 -0.8956698 0.3793
## meanooarea
              1.005070 0.498035 2.0180725 0.0549
##
## Correlation:
##
              (Intr)
## meanooarea -0.999
##
## Standardized residuals:
##
         Min
                     Q1
                                           QЗ
                               Med
                                                      Max
## -1.7775727 -0.6837636 -0.2886598 0.4237510
                                              2.2862305
## Residual standard error: 0.6334179
## Degrees of freedom: 26 total; 24 residual
## Generalized least squares fit by REML
    Model: meansperm ~ meanoo
##
     Data: fullpglsframe
         AIC
##
                  BIC
                         logLik
    50.30285 55.01506 -21.15142
##
## Correlation Structure: corMartins
## Formula: ~1
## Parameter estimate(s):
##
      alpha
## 5.469326
##
## Coefficients:
                  Value Std.Error t-value p-value
## (Intercept) -7.068653 3.346328 -2.112361 0.0452
               1.074887 0.329803 3.259179 0.0033
## meanoo
##
## Correlation:
```

```
(Intr)
## meanoo -0.999
##
## Standardized residuals:
          Min
                                Med
                                             QЗ
## -1.9084685 -0.7013875 -0.3474741 0.5038124 2.4034645
## Residual standard error: 0.5843052
## Degrees of freedom: 26 total; 24 residual
Some phylogenetic signal methods - lambda
## Loading required package: parallel
       Length Class
                         Mode
## lik
         1
              bm
                         function
## bnd
              data.frame list
## res 400
              -none-
                         numeric
## opt
              -none-
                         list
##
       Length Class
                         Mode
## lik
         1
              bm
                         function
## bnd
         2
              data.frame list
## res 400
              -none-
                         numeric
## opt
         8
              -none-
                         list
       Length Class
                         Mode
## lik
         1
              bm
                         function
## bnd
         2
              data.frame list
## res 400
              -none-
                         numeric
## opt
         8
              -none-
                         list
##
       Length Class
                         Mode
## lik
         1
              bm
                         function
## bnd
         2
              data.frame list
## res 400
              -none-
                         numeric
## opt
         8
              -none-
                         list
Blombergs K
             K PIC.variance.obs PIC.variance.rnd.mean PIC.variance.P
## 1 0.3738911
                       2.289904
                                              3.954053
                                                                  0.03
   PIC.variance.Z
## 1
          -1.472221
            K PIC.variance.obs PIC.variance.rnd.mean PIC.variance.P
## 1 0.670685
                                             1.230789
                     0.4572932
                                                                0.001
   PIC.variance.Z
```

## 1

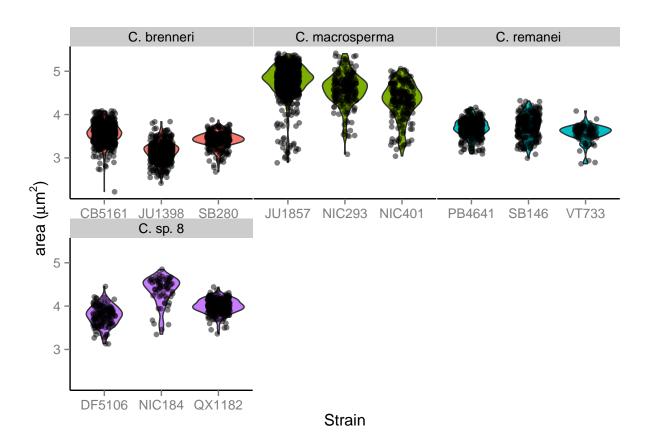
-1.935959

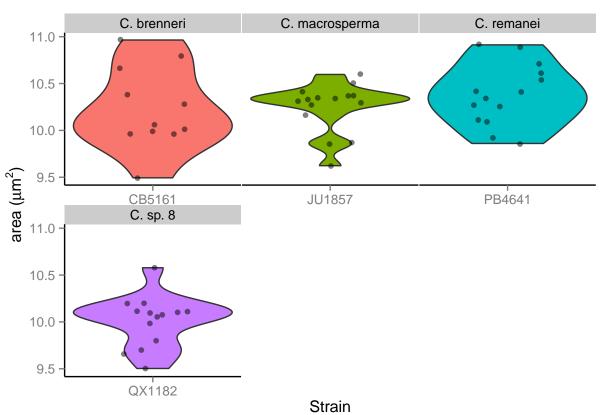
```
## [1] "Dropping tips from the tree because they are not present in the data:"
## [1] "C. sp. 1"
                         "C. portoensis"
                                           "C. virilis"
                         "C. angaria"
## [4] "C. sp. 8"
                                           "C. castelli"
## [7] "C. sp. 2"
                          "C. kamaaina"
                                           "C. imperialis"
## [10] "C. afra"
                          "C. nouraguensis" "C. doughertyi"
## [13] "C. wallacei"
                         "C. tropicalis"
                                           "C. sp. 5"
## [16] "C. nigoni"
           K PIC.variance.obs PIC.variance.rnd.mean PIC.variance.P
##
## 1 1.218587
                     1.696502
                                            2.58208
## PIC.variance.Z
## 1
         -1.943915
            K PIC.variance.obs PIC.variance.rnd.mean PIC.variance.P
## 1 0.5877787
                     0.2622437
                                           0.6191475
                                                              0.002
## PIC.variance.Z
## 1
         -1.763413
Morans I
## class: krandtest
## Monte-Carlo tests
## Call: as.krandtest(sim = matrix(res$result, ncol = nvar, byrow = TRUE),
      obs = res$obs, alter = alter, names = test.names)
##
## Number of tests: 1
## Adjustment method for multiple comparisons:
                                                none
## Permutation number:
                        999
   Test
                 Obs
                      Std.Obs Alter Pvalue
## 1 x -0.09396389 -1.515518 greater 0.954
## other elements: adj.method call
## class: krandtest
## Monte-Carlo tests
## Call: as.krandtest(sim = matrix(res$result, ncol = nvar, byrow = TRUE),
      obs = res$obs, alter = alter, names = test.names)
##
## Number of tests: 1
##
## Adjustment method for multiple comparisons:
                                                none
## Permutation number:
                        999
## Test
                 Obs
                       Std.Obs Alter Pvalue
## 1
       x -0.01730539 0.5842424 greater 0.206
## other elements: adj.method call
## class: krandtest
## Monte-Carlo tests
## Call: as.krandtest(sim = matrix(res$result, ncol = nvar, byrow = TRUE),
      obs = res$obs, alter = alter, names = test.names)
```

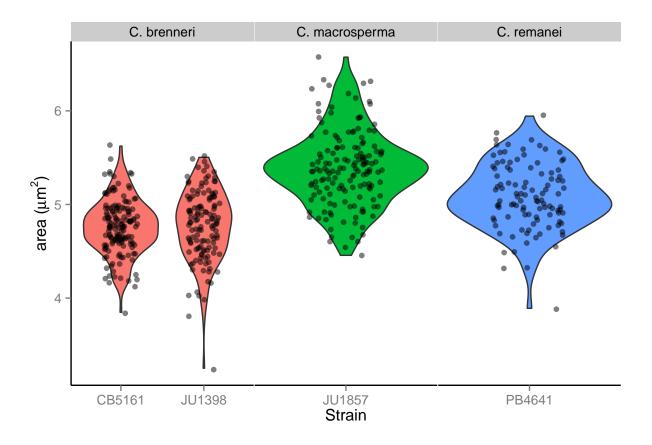
```
##
## Number of tests: 1
##
## Adjustment method for multiple comparisons:
                                                none
## Permutation number:
                        999
    Test
                Obs
                       Std.Obs
                                Alter Pvalue
       x -0.1482137 -0.9332742 greater 0.812
##
## other elements: adj.method call
## class: krandtest
## Monte-Carlo tests
## Call: as.krandtest(sim = matrix(res$result, ncol = nvar, byrow = TRUE),
##
       obs = res$obs, alter = alter, names = test.names)
##
## Number of tests: 1
## Adjustment method for multiple comparisons:
                                                none
## Permutation number:
                        999
    Test
                 Obs
                       Std.Obs Alter Pvalue
       x -0.03082416 0.2119096 greater 0.37
## other elements: adj.method call
abouheifs cmean
## class: krandtest
## Monte-Carlo tests
## Call: as.krandtest(sim = matrix(res$result, ncol = nvar, byrow = TRUE),
       obs = res$obs, alter = alter, names = test.names)
##
## Number of tests: 1
## Adjustment method for multiple comparisons:
## Permutation number:
                        999
## Test
                 Obs Std.Obs Alter Pvalue
## 1
       x -0.09396389 -1.529576 greater 0.953
## other elements: adj.method call
## class: krandtest
## Monte-Carlo tests
## Call: as.krandtest(sim = matrix(res$result, ncol = nvar, byrow = TRUE),
##
       obs = res$obs, alter = alter, names = test.names)
## Number of tests: 1
## Adjustment method for multiple comparisons:
## Permutation number:
                        999
## Test
                 Obs
                       Std.Obs Alter Pvalue
## 1
       x -0.01730539 0.5925913 greater 0.205
## other elements: adj.method call
```

```
## class: krandtest
## Monte-Carlo tests
## Call: as.krandtest(sim = matrix(res$result, ncol = nvar, byrow = TRUE),
##
       obs = res$obs, alter = alter, names = test.names)
## Number of tests:
## Adjustment method for multiple comparisons:
                                                none
## Permutation number:
                        999
    Test
                 Obs
                        Std.Obs Alter Pvalue
## 1
       x -0.1482137 -0.9384371 greater 0.814
## other elements: adj.method call
## class: krandtest
## Monte-Carlo tests
## Call: as.krandtest(sim = matrix(res$result, ncol = nvar, byrow = TRUE),
       obs = res$obs, alter = alter, names = test.names)
##
## Number of tests:
                    1
## Adjustment method for multiple comparisons:
## Permutation number:
                        999
    Test
                 Obs
                       Std.Obs Alter Pvalue
## 1
       x -0.03082416 0.2241026 greater 0.365
##
## other elements: adj.method call
```

figure 3A with violin plots







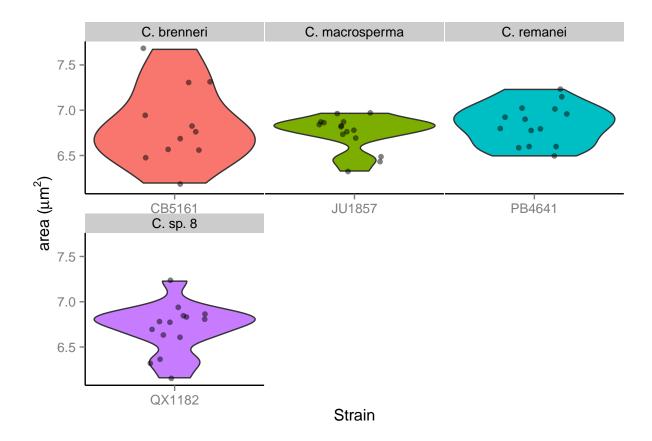
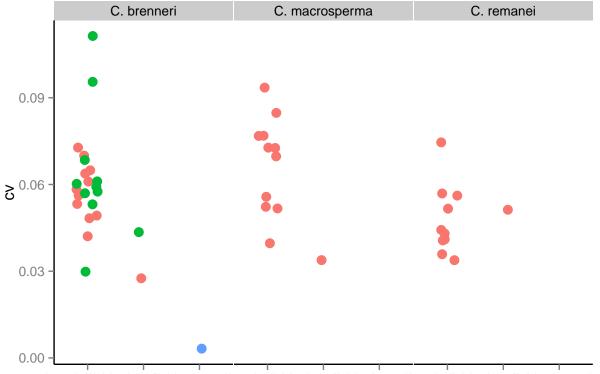
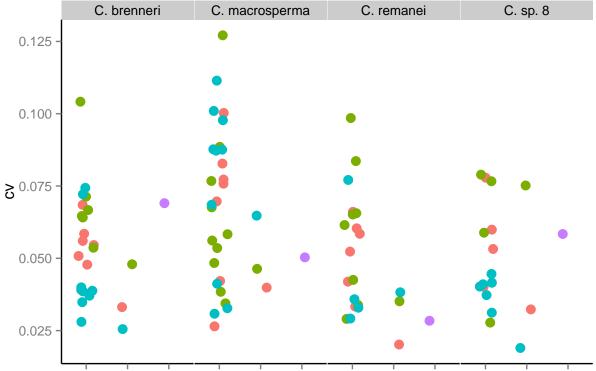


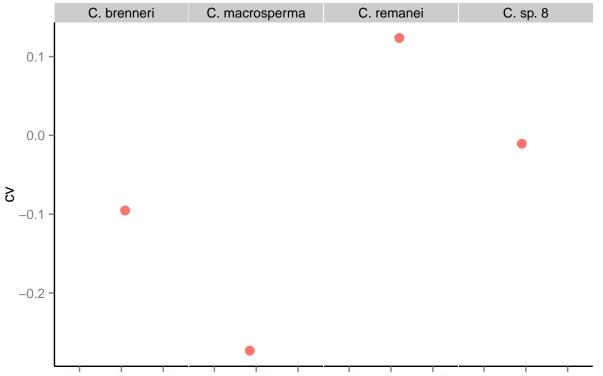
figure 3B



intra individuean individuean straintsa individuean individuean straintsa individuean individuean straints



intrabedixeenabedixee



intra biediweenalbetiweena listtearbiediweena listt

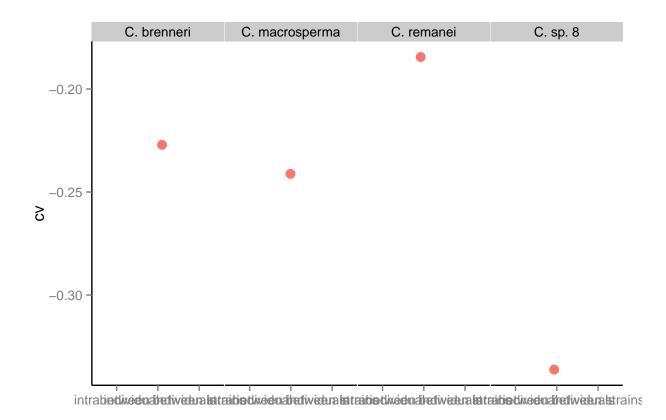


figure mockup a we are starting agin here, as I've modified some of the variables etc above Easier to start with a clean slate Clean and read in data:

mean and sd

```
groupedmeans=ff %>% group_by(Species,sex) %>% summarise(mean=mean(area,na.rm=T),n=n(),sd=sd(area,na.rm=T))
limits <- aes(ymax = mean + 1.96*sd/sqrt(n), ymin=mean - 1.96*sd/sqrt(n))
dodge <- position_dodge(width=0.9)
ggplot(data=groupedmeans, aes(x=Species, y=mean, fill=sex)) + geom_bar(stat="identity", position=positi
    scale_fill_manual(values=c("gray","white"))+ylab( expression(paste("area (", mu, m^{2},")")))+xlab(
    theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
    panel.background = element_blank(),axis.line = element_line(colour = "black"),legend.position = "n</pre>
```

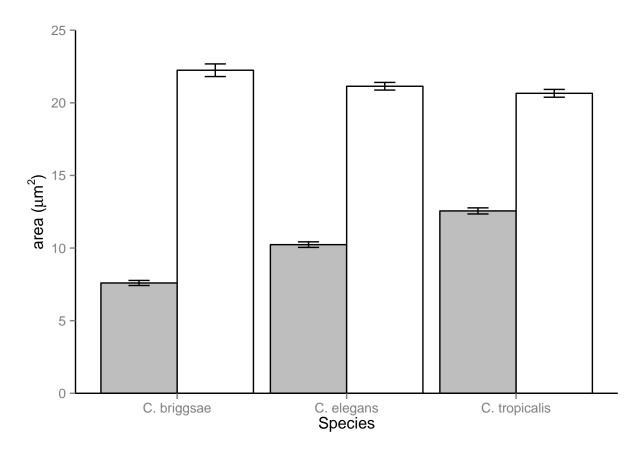


Figure mockup b group to get the data I want it's ugly, but it works

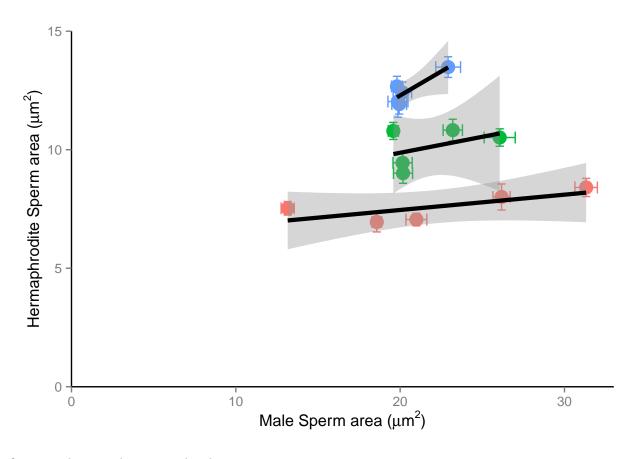
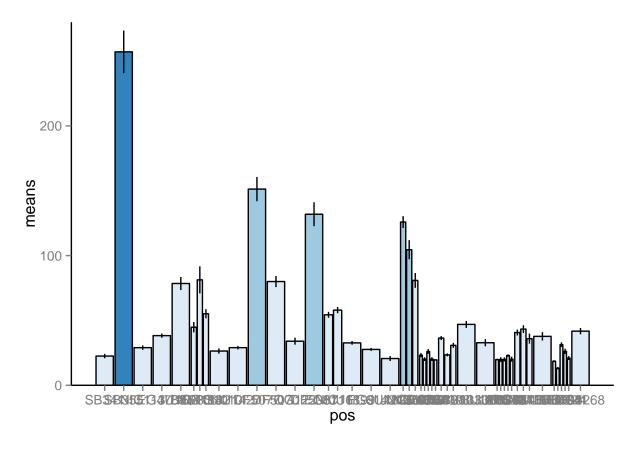


figure1a - bars need to rerun the above

```
ff<-spermsizes<-read.table("C:/Users/jeremy/Desktop/geiger/compileddata.csv",header=T,sep=",")
ff<-renameworms(ff)</pre>
spermsize2<-ff %>% group_by(Species,Strain) %>% summarise(means=mean(area,na.rm=T),cv=sd(area,na.rm=T)/
countspec=function(dataframe){
    holding=c()
    for(i in 1:nrow(dataframe)){
        holding=c(holding,sum(dataframe$Species==dataframe[i,]$Species))
    return(holding)
spermsize2$Species<-ordered(spermsize2$Species,levels=c(tr$tip.label))</pre>
spermsize2<-spermsize2[order(spermsize2$Species),]</pre>
spermsize2$Strain<-factor(spermsize2$Strain, levels = spermsize2$Strain)</pre>
spermsize2$specnum=countspec(spermsize2)
spermsize2$cols=cut(spermsize2$means,3)
findchange<-function(df){</pre>
    holding=c()
    for(i in 1:length(df$Species)){
        if(i==1){
            holding=c(holding,0)
        } else if(df$Species[i]==df$Species[i-1]){
            holding=c(holding,0)
        } else {
            holding=c(holding,1)
```

```
return(holding)
spermsize2$change<-findchange(spermsize2)</pre>
w<-(1-(spermsize2$specnum-1)*0.05)/spermsize2$specnum
pos \leftarrow 0.5 * (cumsum(w) + cumsum(c(0, w[-length(w)])))
gapsizes<-function(df,pos){</pre>
    pos1<-pos
    for(i in 1:length(df$specnum)){
        if(df$specnum[i]==1){
            pos1[i:length(pos1)]<-pos1[i:length(pos1)]+0.1
        } else if(df$change[i]==1){
            pos1[i:length(pos1)]<-pos1[i:length(pos1)]+0.1</pre>
        } else {
            pos1[i:length(pos1)]<-pos1[i:length(pos1)]+0.05
    }
    return(pos1)
}
pos<-gapsizes(spermsize2,pos)</pre>
ggplot(data=spermsize2,aes(x = pos, width = w, y = means, fill=cols,ymin=means-sem,ymax=means+sem)) + s
    geom_bar(stat = "identity",colour="black") + scale_x_continuous(labels = spermsize2$Strain, breaks
    theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
     panel.background = element_blank(),axis.line = element_line(colour = "black"),legend.position = "n
```

## Warning: position\_stack requires constant width: output may be incorrect



large tree first group all the data

```
spermsize1<-ff %>% group_by(Species) %>% summarise(means=mean(area,na.rm=T))
spermsize2<-ff %>% group_by(Species,Strain) %>% summarise(means=mean(area,na.rm=T))
spermsize3<-ff %>% group_by(Species,Strain,individual) %>% summarise(means=mean(area,na.rm=T))
```

function to fix the format of spermsizes, and fit tree

```
fitace <- function(spermsizes,tree){
  holding<-spermsizes$means
  names(holding)<-spermsizes$Species
  return(fastAnc(tree,holding,CI=T))
}</pre>
```

function to choose a subset - works on any level, just chooses one from each species

```
subsetter<-function(spermsizes){
   x=lapply(split(spermsizes, spermsizes$Species),function(subdf) subdf[sample(1:nrow(subdf), 1),])
   do.call("rbind",x)
}</pre>
```

example usage fitace(subsetter(spermsizes2),tr) now need to loop and keep the values

```
looper<-function(spermsizes,reps,tree){
  outputdata=data.frame()
  outputCI=data.frame()
  for(i in 1:reps){
     outputdata=rbind(outputdata,fitace(subsetter(spermsizes),tree)$ace)
     outputCI=rbind(outputCI,fitace(subsetter(spermsizes),tree)$CI95)
  }
  names(outputdata)=seq(from=27,to=51)
  return(c(outputdata,outputCI))
}</pre>
```

ok we can call eg x=looper(spermsize3,10,tr) then mean and sd, etc. etc.

```
x<-fitace(spermsize1,tr)
AncSperm=x$ace
#or AncSperm=apply(x,2,mean) if it's bootstrapped
spermsize1$Species <- factor(spermsize1$Species, levels = c("C. sp. 1", "C. plicata", "C. guadeloupensis"</pre>
spermsize=spermsize1[order(spermsize1$Species),]$means
names(spermsize)<-spermsize1[order(spermsize1$Species),]$Species</pre>
plot(tr,label.offset=0,no.margin = TRUE, show.tip.label = F,x.lim=c(0,1),edge.width=1.5,edge.color=c(1))
#loads the graph into memory, then plots the anc. states on the nodes
#this is taken more or less from Nicos function
lastPP <- get("last_plot.phylo", envir = .PlotPhyloEnv)</pre>
  node <- (lastPP$Ntip + 1):length(lastPP$xx)</pre>
  XX <- lastPP$xx[node]</pre>
 YY <- lastPP$yy[node]
  for (i in 1:length(XX))
    if (round(AncSperm[i], digits=1)==59.6) #this is a horrible hack to fix the labels if they are too b
      points(XX[i],YY[i],col="red",lwd=round(AncSperm[i], digits=1)/max(round(spermsize,digit=1))*50,pci
      text(XX[i]+0.02,YY[i],labels=round(AncSperm[i], digits=1),col="black",cex=1.2,adj=c(1,0))
    }
    points(XX[i],YY[i],col="red",lwd=round(AncSperm[i], digits=1)/max(round(spermsize,digit=1))*50,pch=
    text(XX[i],YY[i],labels=round(AncSperm[i], digits=1),col="black",cex=1.4,adj=c(1,0))
  }
#now we label the tips
tip <- 1:lastPP$Ntip</pre>
XX <- lastPP$xx[tip]</pre>
YY <- lastPP$yy[tip]
points(XX,YY,col="blue",lwd=(round(spermsize,digit=1)/max(round(spermsize,digit=1))*50),pch=19)
text(XX+0.15,YY,labels=round(spermsize,digit=1),col="blue",cex=1.4)
text(XX+0.07,YY,labels=c("C. sp. 1","C. plicata", "C. guadeloupensis", "C. portoensis", "C. virilis", "C. s
```

