

PCA in diploids

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```
library(ggplot2)
library(grid)
library(ggpubr)
```

Loading libraries

```
setwd("/Users/zebinzhang/Desktop/My_Computer/Uppsala_University/Capsella_Project")
# This file contains reads counts information of each gene in all samples
rawTMM <- read.table("InputData/Diploids_individual_TMM_FLR.txt", header=TRUE, row.names=1)
head(rawTMM)[1:5]
```

Set work directly and load files.

```
##           CR1_F      CR2_F      CR3_F      CR4_F      CG1_F
## Carubv10000018m.g 4275.497 5362.8262 4307.0971 3454.9034 7684.806
## Carubv10000019m.g  357.008  391.8685  348.1086  369.9587 1019.473
## Carubv10000020m.g 3025.966 2636.1311 2675.1350 2537.0918 3325.447
## Carubv10000021m.g 5897.512 6217.7578 6132.0495 4990.0279 6097.419
## Carubv10000022m.g 5522.177 4889.0774 5237.4588 4539.4333 5387.324
## Carubv10000023m.g 5517.350 4844.7843 5467.3307 4674.4514 5601.544
```

```
dim(rawTMM)
```

```
## [1] 17307    36
```

```
# This file contains informations of population and tissue in each accession
Accession <- read.table("InputData/DiploidsPhenotypicFile.txt", header = T)
head(Accession)
```

```
##  accession tissue species  MateType NewName
## 1   75.13_F flower      CR      Selfer  CR1_F
## 2   79.17_F flower      CR      Selfer  CR2_F
## 3    81.1_F flower      CR      Selfer  CR3_F
## 4    83.4_F flower      CR      Selfer  CR4_F
## 5    85.3_F flower      CG Outcrosser CG1_F
## 6    86.12_F flower      CG Outcrosser CG2_F
```

```

CR_F <- c(Accession$species=="CR" & Accession$tissue=="flower")
CG_F <- c(Accession$species=="CG" & Accession$tissue=="flower")
CO_F <- c(Accession$species=="CO" & Accession$tissue=="flower")

CR_L <- c(Accession$species=="CR" & Accession$tissue=="leaf")
CG_L <- c(Accession$species=="CG" & Accession$tissue=="leaf")
CO_L <- c(Accession$species=="CO" & Accession$tissue=="leaf")

CR_R <- c(Accession$species=="CR" & Accession$tissue=="root")
CG_R <- c(Accession$species=="CG" & Accession$tissue=="root")
CO_R <- c(Accession$species=="CO" & Accession$tissue=="root")

```

Define species

```
dim(rawTMM)
```

remove genes with 0 expression value in any individual

```
## [1] 17307    36
```

```

TMM <- na.omit(rawTMM)
dim(TMM)

```

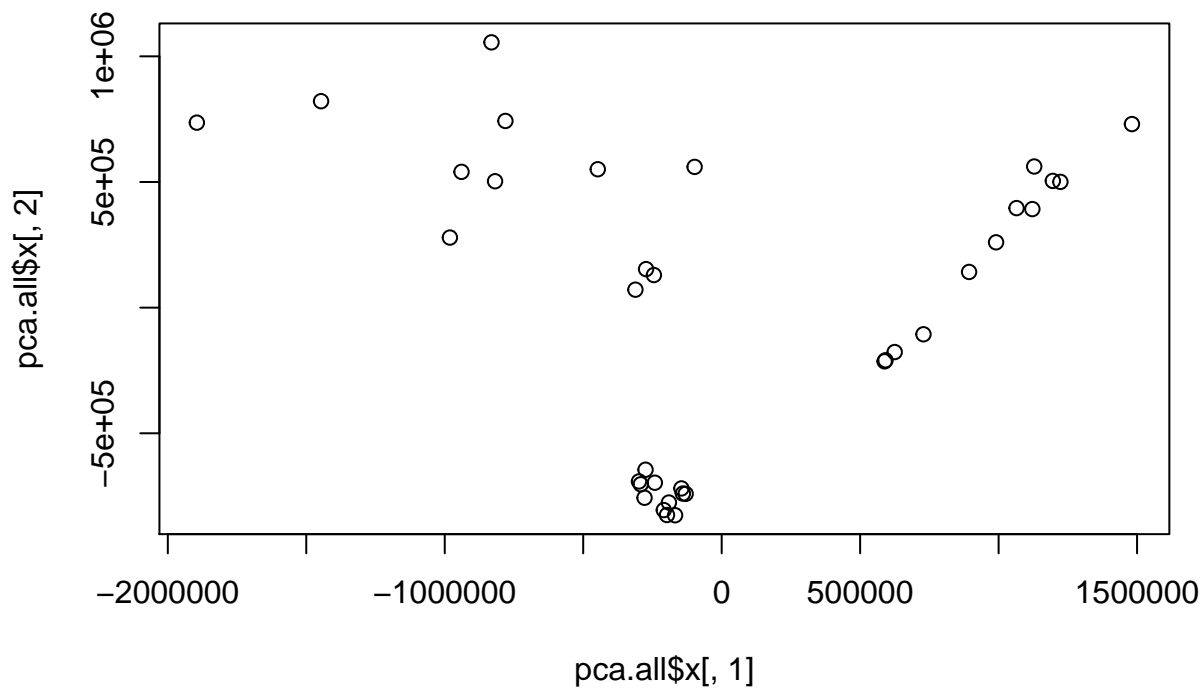
```
## [1] 17307    36
```

PCA in all tissues: Flower, Leaf, Root Perform PCA

```

pca.all <- prcomp(t(TMM))
# NOTE: By default, prcomp() expects the samples to be rows and the genes to be columns
# So the data have to be transpose the matrix using the t() function
# If don't transpose the matrix, you will ultimately get a graph that shows how the genes are related to each other
# prcomp() returns three things:
# 1) x -- x contains the principal components (PCs) for drawing a graph.
## plot PC1 and PC2
plot(pca.all$x[,1],pca.all$x[,2])

```

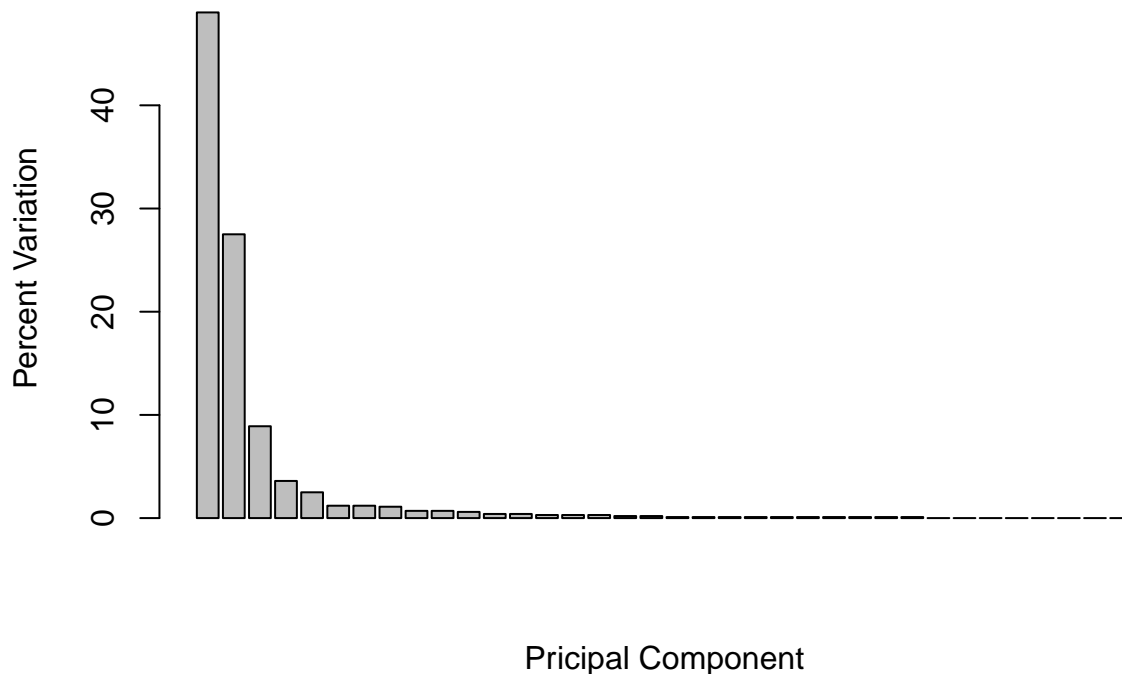


```
# 2) sdev -- "standard deviation", to calculate how much variation in the original data each PC accounts for.
## how much variation in the original data PC1 accounts for.
pca.all.var <- pca.all$sdev^2
head(pca.all.var)
```

```
## [1] 649042747103 363638260151 118449733159 47121684231 32665486503
## [6] 16427751269
```

```
## Since the percentage of variation that each PC accounts for is way more interesting than the actual values.
## we calculate the percentage..
pca.all.var.per <- round(pca.all.var/sum(pca.all.var)*100,1)
## Plotting the percentage is easy with barplot()
barplot(pca.all.var.per, main = "Contribution of PCs", xlab = "Principal Component", ylab = "Percent Variance Explained")
```

Contribution of PCs



```
# 3) rotation -- the loading scores rotation
```

```
dim(pca.all$x)[1] # check how many PCs in total
```

Format the data the way of ggplot2 needs

```
## [1] 36
```

```
pca.all.data <- data.frame(ID = 1:dim(pca.all$x)[1]) # create a new dataframe
```

```
for (i in 1:dim(pca.all$x)[1]){ # name each PCs
  pcs <- paste0("PC", i)
  pca.all.data[[pcs]] <- pca.all$x[,i]
}
```

```
# Set ID, Tissue, species, and matingSystem for ggplot
```

```
pca.all.data$ID <- row.names(pca.all.data)
pca.all.data$Tissue <- c(rep("Flower",12), rep("Leaf",12), rep("Root",12))
pca.all.data$Species <- rep(c(rep("CR",4), rep("CG",4), rep("CO", 4)), 3)
pca.all.data$Species <- factor(pca.all.data$Species, levels = c("CG","CR","CO"))
pca.all.data$MatingSystem <- rep(c(rep("Selfer",4), rep("Outcrosser",4), rep("Selfer", 4)), 3)
```

```
ZscoreAll <- pca.all.data
head(ZscoreAll)[1:7]
```

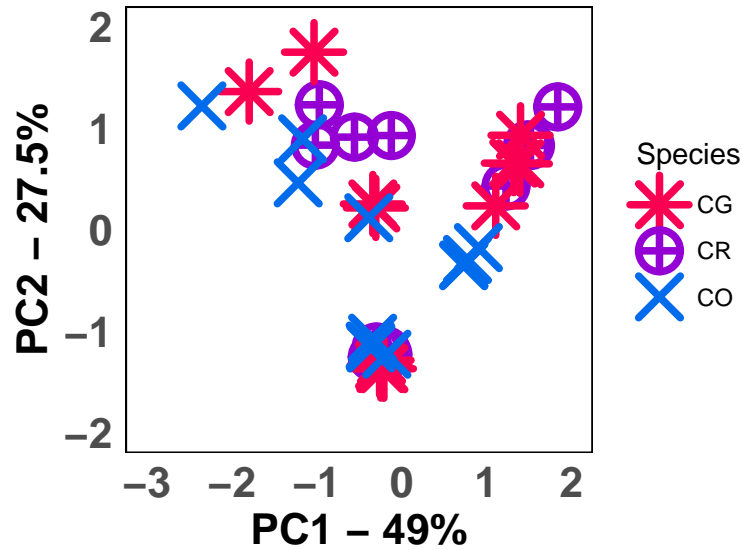
```
##   ID      PC1      PC2      PC3      PC4      PC5      PC6
## 1  1 -278544.1 -756548.9  94069.71 117903.16 -81764.805  2940.939
## 2  2 -145390.7 -719431.4 -37627.00  39356.61  20915.336  2026.887
## 3  3 -130108.3 -740925.1 -19649.46  51476.50  -3602.299 -31941.098
## 4  4 -241297.2 -696687.0  44972.37  42775.38   6835.263 -40415.846
## 5  5 -190536.9 -775558.3  30242.91 162820.18 -161845.001 220102.247
## 6  6 -197479.0 -825155.7  33887.92 184174.11 -221366.092 203952.704
```

```
for(i in 2:13){
  ZscoreAll[,i] <- scale(ZscoreAll[,i])
}
head(ZscoreAll)[1:7]
```

```
##   ID      PC1      PC2      PC3      PC4      PC5      PC6
## 1  1 -0.3457461 -1.254591  0.27332715 0.5431440 -0.45239921 0.02294547
## 2  2 -0.1804679 -1.193039 -0.10932830 0.1813039  0.11572316 0.01581395
## 3  3 -0.1614984 -1.228682 -0.05709310 0.2371366 -0.01993128 -0.24920732
## 4  4 -0.2995130 -1.155322  0.13067086 0.1970532  0.03781906 -0.31532805
## 5  5 -0.2365061 -1.286115  0.08787323 0.7500631 -0.89547760  1.71725748
## 6  6 -0.2451231 -1.368362  0.09846410 0.8484342 -1.22480382  1.59125730
```

```
ggplot(ZscoreAll, aes(PC1, PC2, col = Species, fill = Species)) +
  geom_point(size = 6, stroke = 2, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  scale_fill_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  xlab(paste("PC1 - ", pca.all.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.all.var.per[2], "%", sep = "")) +
  scale_x_continuous(breaks=seq(-3, 2, 1), limits=c(-3, 2)) +
  scale_y_continuous(breaks=seq(-2, 2, 1), limits=c(-2, 2)) +
  theme_minimal() +
  theme(panel.grid = element_blank(),
        panel.border = element_rect(fill= "transparent")) +
  theme(axis.text=element_text(size=16,face="bold"),
        axis.title=element_text(size=16,face="bold")) +
  theme(plot.title = element_text(size=16,face="bold"))
```

Output raw PCA plot



```
# Tissue
p1 <- ggplot(ZscoreAll, aes(PC1, PC2, col = Tissue, fill = Species)) +
  geom_point(size = 6, stroke = 2, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#fa990e", "#155800", "#089400")) +
  scale_fill_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  # stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
  xlab(paste("PC1 - ", pca.all.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.all.var.per[2], "%", sep = "")) +
  # stat_ellipse(geom="polygon", aes(fill = tissue), # add frame
  #             alpha = 0.05,
  #             show.legend = FALSE,
  #             level = 0.90) +
  scale_x_continuous(breaks=seq(-3, 2, 1), limits=c(-3, 2)) +
  scale_y_continuous(breaks=seq(-2, 2, 1), limits=c(-2, 2)) +
  theme_minimal() +
  theme(panel.grid = element_blank(),
        panel.border = element_rect(fill= "transparent")) +
  theme(axis.text=element_text(size=16,face="bold"),
        axis.title=element_text(size=16,face="bold")) +
  #theme(legend.position = "none") +
  ggtitle("by Tissue") +
  theme(plot.title = element_text(size=16,face="bold"))

# Species
p2 <- ggplot(ZscoreAll, aes(PC1, PC2, col = Species, fill = Species)) +
  geom_point(size = 6, stroke = 2, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  scale_fill_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  # stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
  xlab(paste("PC1 - ", pca.all.var.per[1], "%", sep = "")) +
```

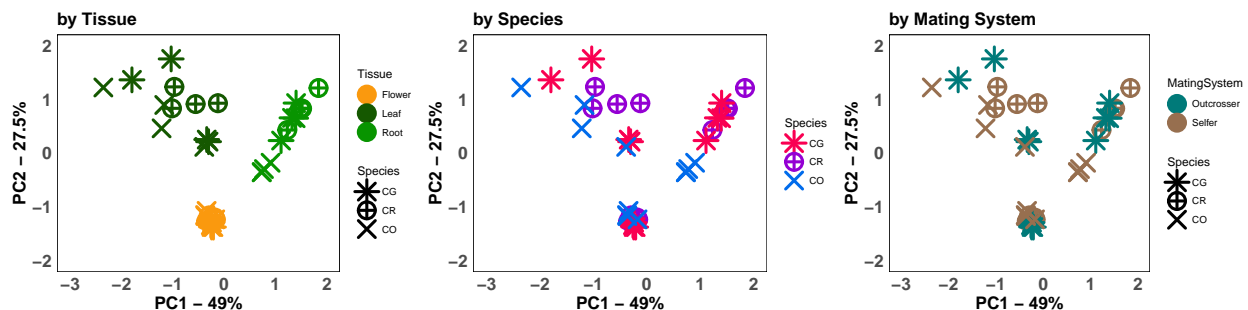
```

ylab(paste("PC2 - ", pca.all.var.per[2], "%", sep = "")) +
# stat_ellipse(geom="polygon", aes(fill = tissue), # add frame
#
#       alpha = 0.05,
#       show.legend = FALSE,
#       level = 0.90) +
scale_x_continuous(breaks=seq(-3, 2, 1), limits=c(-3, 2)) +
scale_y_continuous(breaks=seq(-2, 2, 1), limits=c(-2, 2)) +
theme_minimal() +
theme(panel.grid = element_blank(),
      panel.border = element_rect(fill= "transparent")) +
theme(axis.text=element_text(size=16,face="bold"),
      axis.title=element_text(size=16,face="bold")) +
#theme(legend.position = "none") +
ggtitle("by Species") +
theme(plot.title = element_text(size=16,face="bold"))
# Mating System
p3 <- ggplot(ZscoreAll, aes(PC1, PC2, col = MatingSystem, fill = Species)) +
geom_point(size = 6, stroke = 2, aes(shape = Species)) +
scale_shape_manual(values=c(8,10,4))+
scale_color_manual(values = c("#007a79", "#976f4f")) +
#scale_fill_manual(values = c("#f90052", "darkviolet", "#006beb")) +
# stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
xlab(paste("PC1 - ", pca.all.var.per[1], "%", sep = "")) +
ylab(paste("PC2 - ", pca.all.var.per[2], "%", sep = "")) +
scale_x_continuous(breaks=seq(-3, 2, 1), limits=c(-3, 2)) +
scale_y_continuous(breaks=seq(-2, 2, 1), limits=c(-2, 2)) +
theme_minimal() +
theme(panel.grid = element_blank(),
      panel.border = element_rect(fill= "transparent")) +
theme(axis.text=element_text(size=16,face="bold"),
      axis.title=element_text(size=16,face="bold")) +
#theme(legend.position = "none") +
ggtitle("by Mating System") +
theme(plot.title = element_text(size=16,face="bold"))

grid.newpage()
ggarrange(p1, p2, p3, ncol=3, nrow=1) # set the common legend

```

We also want to know how this plot looks like when color by tissue, by species, and by mating system



Causing all dots of flowers crowded together, I have to add zoom in plot for flower The PCA analysis only for flower is need to run, where all steps are typically same as above mentioned but in different input data

```
## PCA in Flowers
### Define flower
setwd("/Users/zebinzhang/Desktop/My_Computer/Uppsala_University/Capsella_Project")
TMM.Flower <- read.table("InputData/TMM_Diploids_F.txt", header = T, sep = "\t")
head(TMM.Flower)
```

```
##           CG1_F    CG2_F    CG3_F    CG4_F    CR1_F    CR2_F
## Carubv10000018m.g 7684.806 6298.346 5793.3873 6236.9802 4275.497 5362.8262
## Carubv10000019m.g 1019.473 1690.816 401.1901 991.6462 357.008 391.8685
## Carubv10000020m.g 3325.447 2731.951 2855.3056 3079.5167 3025.966 2636.1311
## Carubv10000021m.g 6097.419 5552.219 5613.8344 6340.5385 5897.512 6217.7578
## Carubv10000022m.g 5387.324 5080.437 5374.0826 5583.7381 5522.177 4889.0774
## Carubv10000023m.g 5601.544 5348.137 5554.6822 5333.5295 5517.350 4844.7843
##           CR3_F    CR4_F    C01_F    C02_F    C03_F    C04_F
## Carubv10000018m.g 4307.0971 3454.9034 2968.84893 3585.6948 3102.7761 3045.1453
## Carubv10000019m.g 348.1086 369.9587 89.85919 289.2978 256.7454 268.3176
## Carubv10000020m.g 2675.1350 2537.0918 2212.86192 2552.3922 2331.6557 2050.6528
## Carubv10000021m.g 6132.0495 4990.0279 7223.50950 7483.9381 6995.9217 6537.8976
## Carubv10000022m.g 5237.4588 4539.4333 4020.72119 4467.3627 3717.1696 3824.5072
## Carubv10000023m.g 5467.3307 4674.4514 4440.56247 4828.6397 4427.7404 3767.8463
```

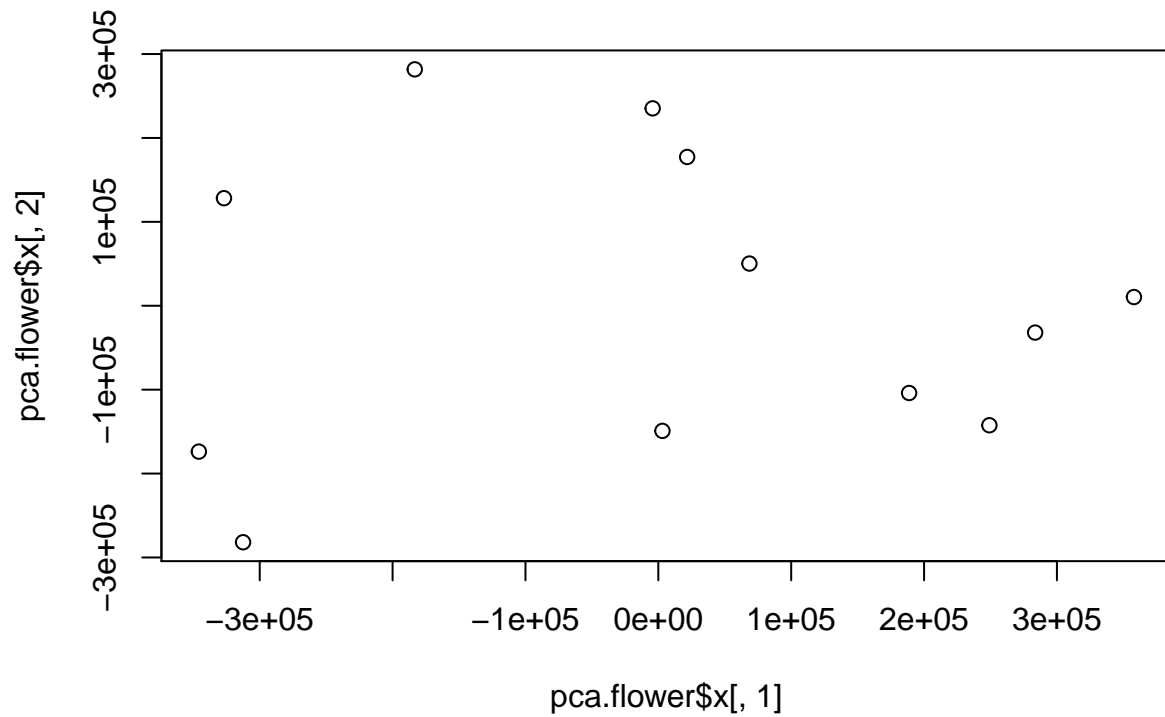
```
dim(TMM.Flower)
```

```
## [1] 17307    12
```

```
dim(na.omit(TMM.Flower))
```

```
## [1] 17307    12
```

```
### perform PCA in Flower
pca.flower <- prcomp(t(na.omit(TMM.Flower)))
# 1) X
## plot PC1 and PC2
plot(pca.flower$x[,1],pca.flower$x[,2])
```

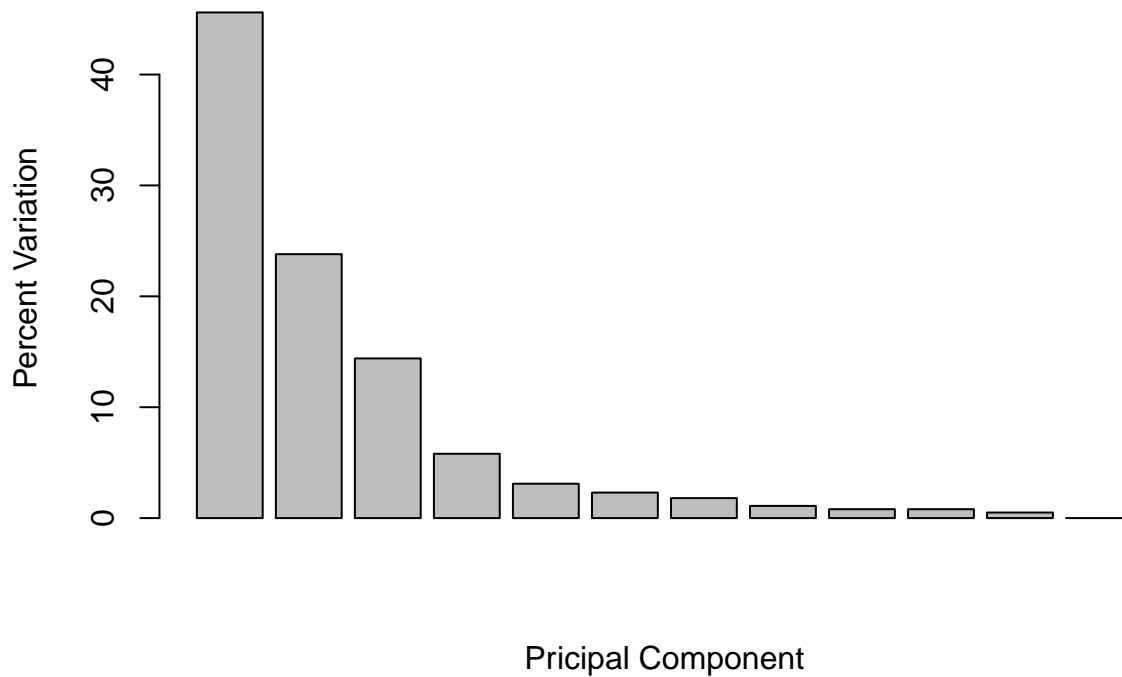



```
# 2) sdev
pca.flower.var <- pca.flower$sdev^2
head(pca.flower.var)
```

```
## [1] 60837790941 31752477312 19214354990 7725694762 4081414720 3116208228
```

```
## calculate the percentage..
pca.flower.var.per <- round(pca.flower.var/sum(pca.flower.var)*100,1)
## Plotting the percentage is easy with barplot()
barplot(pca.flower.var.per, main = "Contribution of PCs in Flowers", xlab = "Principal Component", ylab = "Percentage of Variance Explained")
```

Contribution of PCs in Flowers



```
# 3) rotation -- the loading scores rotation
# Format the data the way ggplot2 likes it
dim(pca.flower$x)[1]
```

```
## [1] 12
```

```
pca.flower.data <- data.frame(ID = 1:dim(pca.flower$x)[1])
for (i in 1:dim(pca.flower$x)[1]){
  pcs <- paste0("PC", i)
  pca.flower.data[[pcs]] <- pca.flower$x[,i]
}

pca.flower.data$ID <- row.names(pca.flower$x)

pca.flower.data$ID <- row.names(pca.flower.data)
pca.flower.data$tissue <- "Flower"
pca.flower.data$Species <- c(rep("CG",4), rep("CR",4), rep("CO", 4))
pca.flower.data$Species <- factor(pca.flower.data$Species, levels = c("CG", "CR", "CO"))
#pca.flower.data$species <- c(c("CR", "CG", "CO"))

ZscoreFlower <- pca.flower.data
head(ZscoreFlower)
```

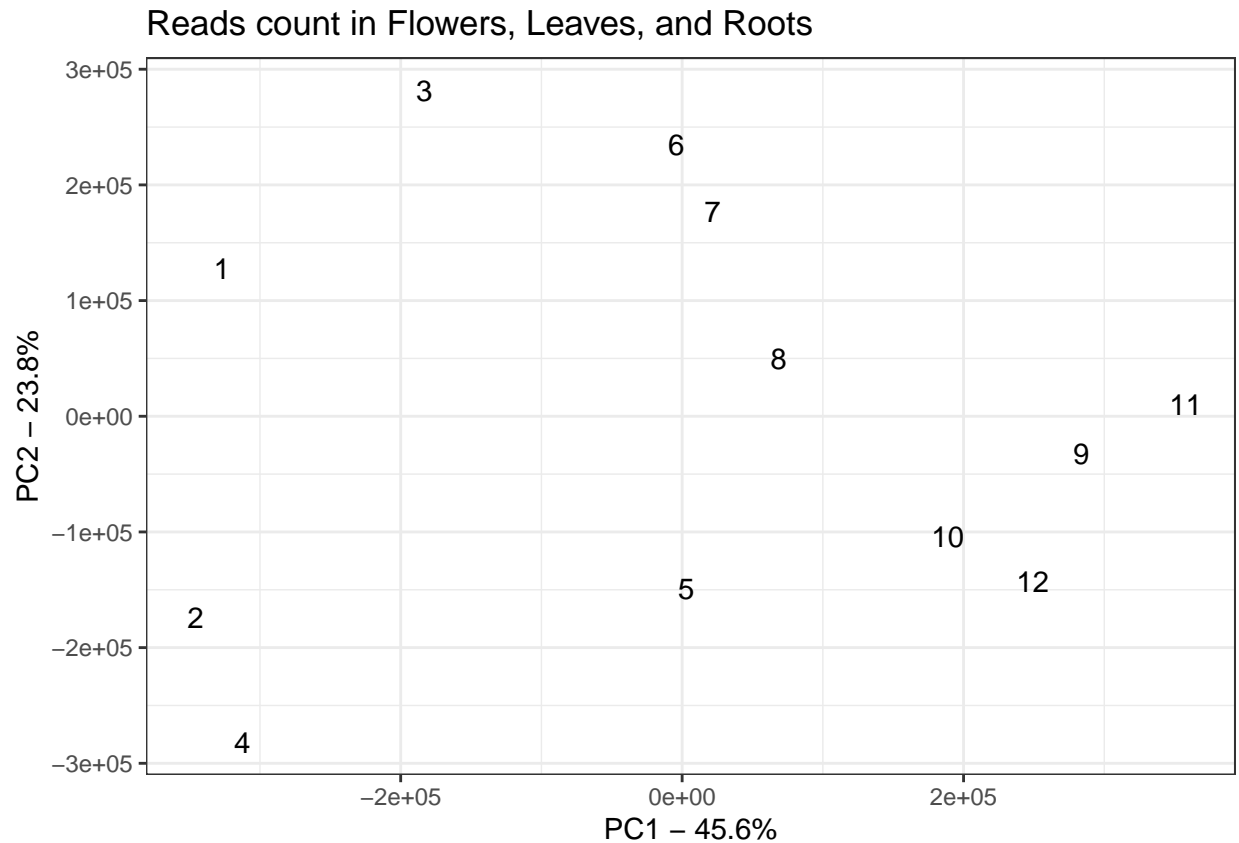
```
##   ID      PC1      PC2      PC3      PC4      PC5      PC6
## 1  1 -326902.058 128200.6 155351.0038  8347.959 -33883.23 -65092.43
```

```
## 2 2 -345819.621 -173820.0 62148.3607 6351.159 137084.95 49138.03
## 3 3 -183308.300 281754.4 108521.1338 -113651.996 -69034.07 37027.22
## 4 4 -312534.713 -281851.5 -878.2356 70210.075 -64291.80 -28275.65
## 5 5 3081.972 -149151.9 -224566.3383 -174966.191 -16428.40 31686.95
## 6 6 -4257.591 235252.3 -100864.0258 97625.278 61320.68 11853.72
##      PC7      PC8      PC9      PC10      PC11      PC12 tissue
## 1 -8207.9039 82616.243 -43.30106 13936.050 -5044.585 -1.625375e-09 Flower
## 2 -45283.8653 -14082.197 -17223.44109 17818.393 2786.605 -4.885407e-10 Flower
## 3 -23164.8764 -61672.001 -9213.33212 -9187.486 10439.164 -9.453557e-10 Flower
## 4 77513.4889 -36549.057 11864.09877 -22810.445 2737.705 -8.174623e-10 Flower
## 5 -935.6943 46954.285 4967.81207 -20801.146 17560.981 2.672199e-10 Flower
## 6 -10080.8005 4860.015 48589.98730 -58728.041 -7841.190 5.280437e-10 Flower
## Species
## 1 CG
## 2 CG
## 3 CG
## 4 CG
## 5 CR
## 6 CR
```

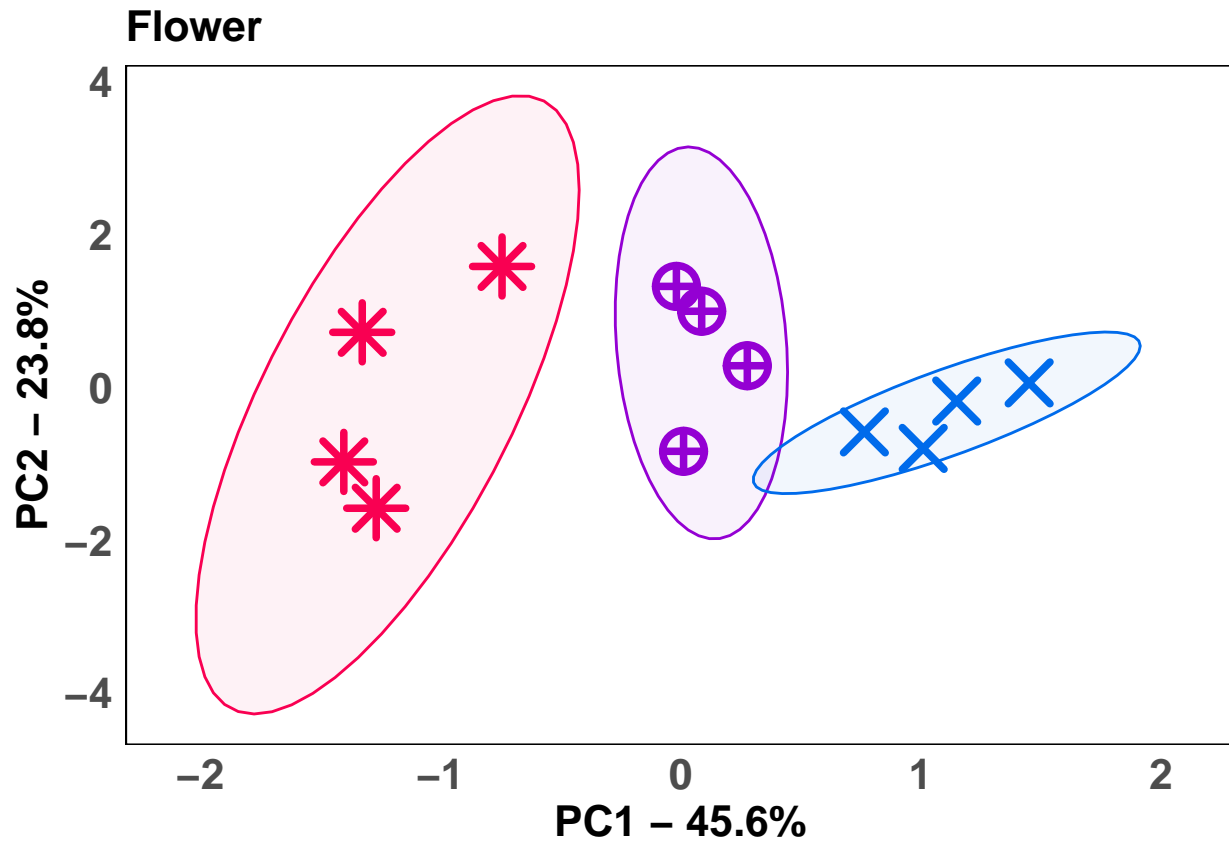
```
for(i in 2:13){
  ZscoreFlower[,i] <- scale(ZscoreFlower[,i])
}
head(ZscoreFlower)
```

```
## ID      PC1      PC2      PC3      PC4      PC5      PC6
## 1 1 -1.32535108 0.7194510 1.120730417 0.09497549 -0.5303706 -1.1660503
## 2 2 -1.40204810 -0.9754631 0.448349586 0.07225771 2.1457762 0.8802470
## 3 3 -0.74318239 1.5811823 0.782891211 -1.29302909 -1.0805829 0.6632968
## 4 4 -1.26710190 -1.5817271 -0.006335752 0.79878641 -1.0063528 -0.5065232
## 5 5 0.01249516 -0.8370280 -1.620062437 -1.99060626 -0.2571520 0.5676324
## 6 6 -0.01726145 1.3202164 -0.727651440 1.11069165 0.9598461 0.2123446
##      PC7      PC8      PC9      PC10      PC11      PC12 tissue
## 1 -0.16804663 2.1562284 -0.001302233 0.4326155 -0.1960269 -1.8059882 Flower
## 2 -0.92713085 -0.3675359 -0.517976651 0.5531347 0.1082844 -0.5477301 Flower
## 3 -0.47427205 -1.6095977 -0.277081153 -0.2852062 0.4056543 -1.0533369 Flower
## 4 1.58699232 -0.9539058 0.356800138 -0.7081024 0.1063842 -0.9117834 Flower
## 5 -0.01915718 1.2254753 0.149401658 -0.6457280 0.6824002 0.2887523 Flower
## 6 -0.20639186 0.1268431 1.461292130 -1.8230889 -0.3046999 0.5774343 Flower
## Species
## 1 CG
## 2 CG
## 3 CG
## 4 CG
## 5 CR
## 6 CR
```

```
# graph parttern 1
ggplot(data = pca.flower.data, aes(x=PC1, y=PC2, label=ID)) +
  geom_text() +
  xlab(paste("PC1 - ", pca.flower.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.flower.var.per[2], "%", sep = "")) +
  theme_bw() +
  ggtitle("Reads count in Flowers, Leaves, and Roots")
```



```
# graph parttern 2
# PC1 vs PC2
ggplot(ZscoreFlower, aes(PC1, PC2, col = Species, fill = Species)) +
  geom_point(size = 6, stroke = 2, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#f90052","darkviolet","#006beb")) +
  scale_fill_manual(values = c("#f90052","darkviolet","#006beb")) +
  # stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
  xlab(paste("PC1 - ", pca.flower.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.flower.var.per[2], "%", sep = "")) +
  stat_ellipse(geom="polygon", aes(fill = Species), # add frame
    alpha = 0.05,
    show.legend = FALSE,
    level = 0.90) +
  scale_x_continuous(breaks=seq(-2, 2, 1), limits=c(-2.1, 2.1)) +
  theme_minimal() +
  theme(panel.grid = element_blank(),
    panel.border = element_rect(fill= "transparent")) +
  theme(axis.text=element_text(size=16,face="bold"),
    axis.title=element_text(size=16,face="bold")) +
  theme(legend.position = "none") +
  ggtitle("Flower") +
  theme(plot.title = element_text(size=16,face="bold"))
```



```
# by Tissue
p1 <- ggplot(ZscoreAll, aes(PC1, PC2, col = Tissue, fill = Species)) +
  geom_point(size = 6, stroke = 2, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#fa990e", "#155800", "#089400")) +
  scale_fill_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  xlab(paste("PC1 - ", pca.all.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.all.var.per[2], "%", sep = "")) +
  scale_x_continuous(breaks=seq(-3, 2, 1), limits=c(-3, 2)) +
  scale_y_continuous(breaks=seq(-2, 2, 1), limits=c(-2, 2)) +
  theme_minimal() +
  theme(panel.grid = element_blank(),
        panel.border = element_rect(fill= "transparent")) +
  theme(axis.text=element_text(size=16,face="bold"),
        axis.title=element_text(size=16,face="bold")) +
  theme(legend.position = "none") +
  ggtitle("by Tissue") +
  theme(plot.title = element_text(size=16,face="bold"))

p1_i <- ggplot(ZscoreFlower, aes(PC1, PC2, col = Species, fill = Species)) +
  geom_point(size = 5, stroke = 1, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#fa990e", "#fa990e", "#fa990e")) +
```

```

scale_fill_manual(values = c("#fa990e", "#fa990e", "#fa990e")) +
xlim(-2,2) + ylim(-2,2)+
#xlab("PC1") +
#ylab("PC2") +
theme_minimal() +
theme(panel.grid = element_blank(),
      panel.border = element_rect(fill= "transparent")) +
theme(legend.position = "none") +
theme(axis.title.x=element_blank(),
      axis.text.x=element_blank(),
      axis.ticks.x=element_blank(),
      axis.title.y=element_blank(),
      axis.text.y=element_blank(),
      axis.ticks.y=element_blank())

p_tissue <- p1 + annotation_custom(ggplotGrob(p1_i), xmin = -3.41, xmax = -1,
                                  ymin = -2.35, ymax = 0)

# Species
p2 <- ggplot(ZscoreAll, aes(PC1, PC2, col = Species, fill = Species)) +
  geom_point(size = 6, stroke = 2, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  scale_fill_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  # stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
  xlab(paste("PC1 - ", pca.all.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.all.var.per[2], "%", sep = "")) +
  # stat_ellipse(geom="polygon", aes(fill = tissue), # add frame
  #             alpha = 0.05,
  #             show.legend = FALSE,
  #             level = 0.90) +
  scale_x_continuous(breaks=seq(-3, 2, 1), limits=c(-3, 2)) +
  scale_y_continuous(breaks=seq(-2, 2, 1), limits=c(-2, 2)) +
  theme_minimal() +
  theme(panel.grid = element_blank(),
        panel.border = element_rect(fill= "transparent")) +
  theme(axis.text=element_text(size=16,face="bold"),
        axis.title=element_text(size=16,face="bold")) +
  theme(legend.position = "none") +
  ggtitle("by Species") +
  theme(plot.title = element_text(size=16,face="bold"))

p2_i <- ggplot(ZscoreFlower, aes(PC1, PC2, col = Species, fill = Species)) +
  geom_point(size = 5, stroke = 1, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  scale_fill_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  xlim(-2,2) + ylim(-2,2)+
  # stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
  #xlab("PC1") +
  #ylab("PC2") +
  theme_minimal() +
  theme(panel.grid = element_blank(),

```

```

    panel.border = element_rect(fill= "transparent")) +
  theme(legend.position = "none") +
  theme(axis.title.x=element_blank(),
        axis.text.x=element_blank(),
        axis.ticks.x=element_blank(),
        axis.title.y=element_blank(),
        axis.text.y=element_blank(),
        axis.ticks.y=element_blank())

p_species <- p2 + annotation_custom(ggplotGrob(p2_i), xmin = -3.41, xmax = -1,
                                   ymin = -2.35, ymax = 0)

# Mating System
p3 <- ggplot(ZscoreAll, aes(PC1, PC2, col = MatingSystem, fill = Species)) +
  geom_point(size = 6, stroke = 2, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#007a79", "#976f4f")) +
  #scale_fill_manual(values = c("#f90052", "darkviolet", "#006beb")) +
  # stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
  xlab(paste("PC1 - ", pca.all.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.all.var.per[2], "%", sep = "")) +
  scale_x_continuous(breaks=seq(-3, 2, 1), limits=c(-3, 2)) +
  scale_y_continuous(breaks=seq(-2, 2, 1), limits=c(-2, 2)) +
  theme_minimal() +
  theme(panel.grid = element_blank(),
        panel.border = element_rect(fill= "transparent")) +
  theme(axis.text=element_text(size=16,face="bold"),
        axis.title=element_text(size=16,face="bold")) +
  theme(legend.position = "none") +
  ggtitle("by Mating System") +
  theme(plot.title = element_text(size=16,face="bold"))

p3_i <- ggplot(ZscoreFlower, aes(PC1, PC2, col = Species, fill = Species)) +
  geom_point(size = 5, stroke = 1, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#007a79", "#976f4f", "#976f4f")) +
  scale_fill_manual(values = c("#007a79", "#976f4f", "#976f4f")) +
  xlim(-2,2) + ylim(-2,2)+
  # stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
  #xlab("PC1") +
  #ylab("PC2") +
  theme_minimal() +
  theme(panel.grid = element_blank(),
        panel.border = element_rect(fill= "transparent")) +
  theme(legend.position = "none") +
  theme(axis.title.x=element_blank(),
        axis.text.x=element_blank(),
        axis.ticks.x=element_blank(),
        axis.title.y=element_blank(),
        axis.text.y=element_blank(),
        axis.ticks.y=element_blank())

p_MST <- p3 + annotation_custom(ggplotGrob(p3_i), xmin = -3.41, xmax = -1,

```

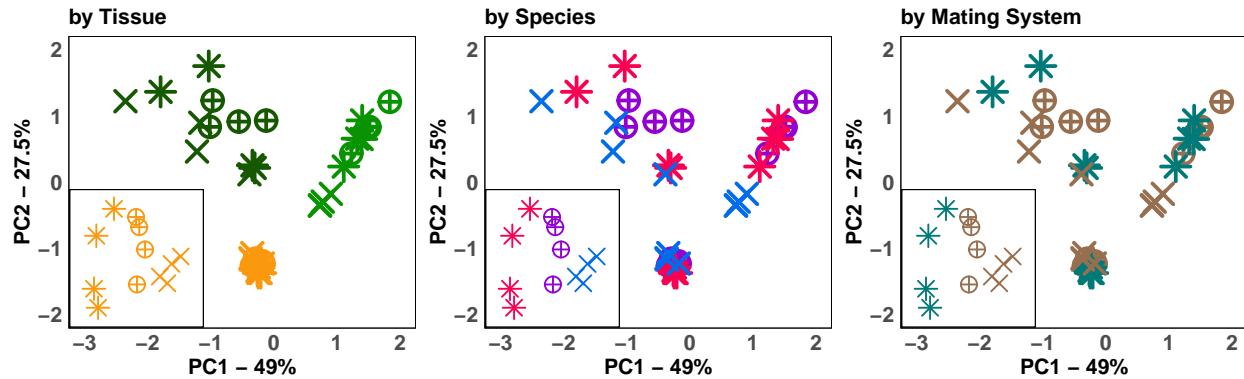
```

ymin = -2.35, ymax = 0)

grid.newpage()
ggarrange(p_tissue, p_species, p_MST, ncol=3, nrow=1) # set the common legend

```

Insert the flower plot into main figures.



PCA plot separate by tissues

Load data

```

setwd("/Users/zebinzhang/Desktop/My_Computer/Uppsala_University/Capsella_Project")
TMM.Flower <- read.table("InputData/TMM_Diploids_F.txt", header = T, sep = "\t")
TMM.Leaf <- read.table("InputData/TMM_Diploids_L.txt", header = T, sep = "\t")
TMM.Root <- read.table("InputData/TMM_Diploids_R.txt", header = T, sep = "\t")

```

```
head(TMM.Flower)[1:5]
```

PCA in Flowers

```

##           CG1_F    CG2_F    CG3_F    CG4_F    CR1_F
## Carubv10000018m.g 7684.806 6298.346 5793.3873 6236.9802 4275.497
## Carubv10000019m.g 1019.473 1690.816 401.1901 991.6462 357.008
## Carubv10000020m.g 3325.447 2731.951 2855.3056 3079.5167 3025.966
## Carubv10000021m.g 6097.419 5552.219 5613.8344 6340.5385 5897.512
## Carubv10000022m.g 5387.324 5080.437 5374.0826 5583.7381 5522.177
## Carubv10000023m.g 5601.544 5348.137 5554.6822 5333.5295 5517.350

```

```
dim(TMM.Flower)
```

```
## [1] 17307    12
```



```
dim(na.omit(TMM.Flower))
```

```
## [1] 17307    12
```

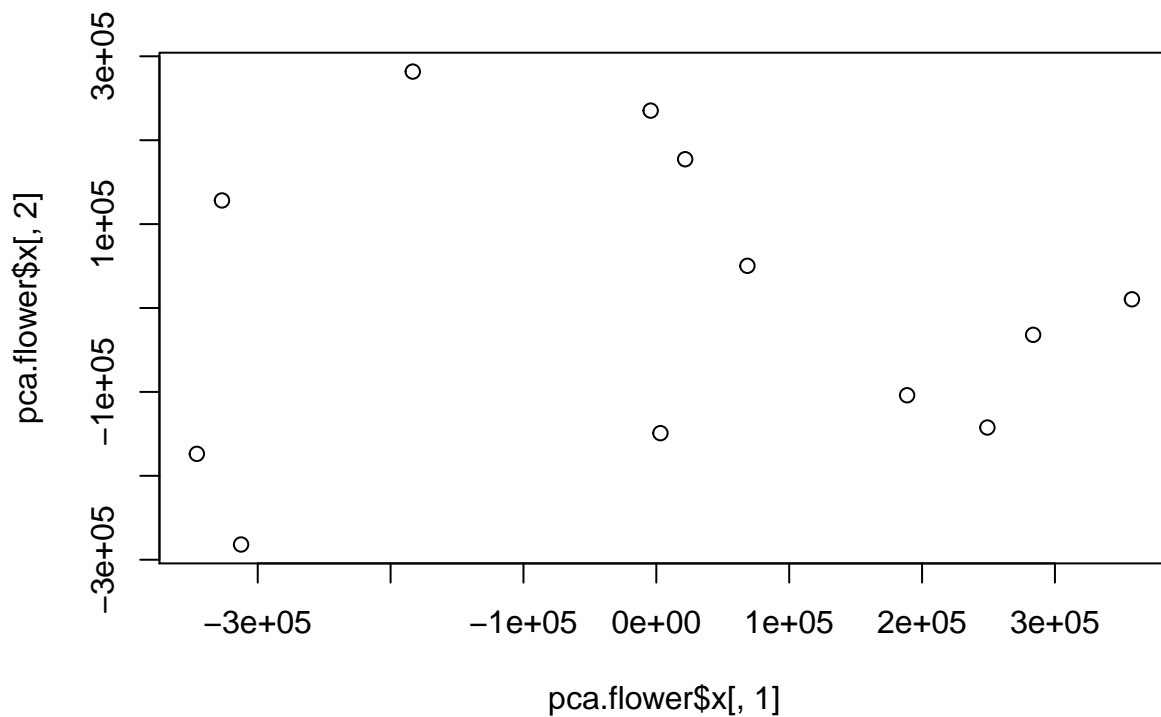
```
### perform PCA in Flower
```

```
pca.flower <- prcomp(t(na.omit(TMM.Flower)))
```

```
# 1) X
```

```
## plot PC1 and PC2
```

```
plot(pca.flower$x[,1],pca.flower$x[,2])
```



```
# 2) sdev
```

```
pca.flower.var <- pca.flower$sdev^2
```

```
head(pca.flower.var)
```

```
## [1] 60837790941 31752477312 19214354990 7725694762 4081414720 3116208228
```

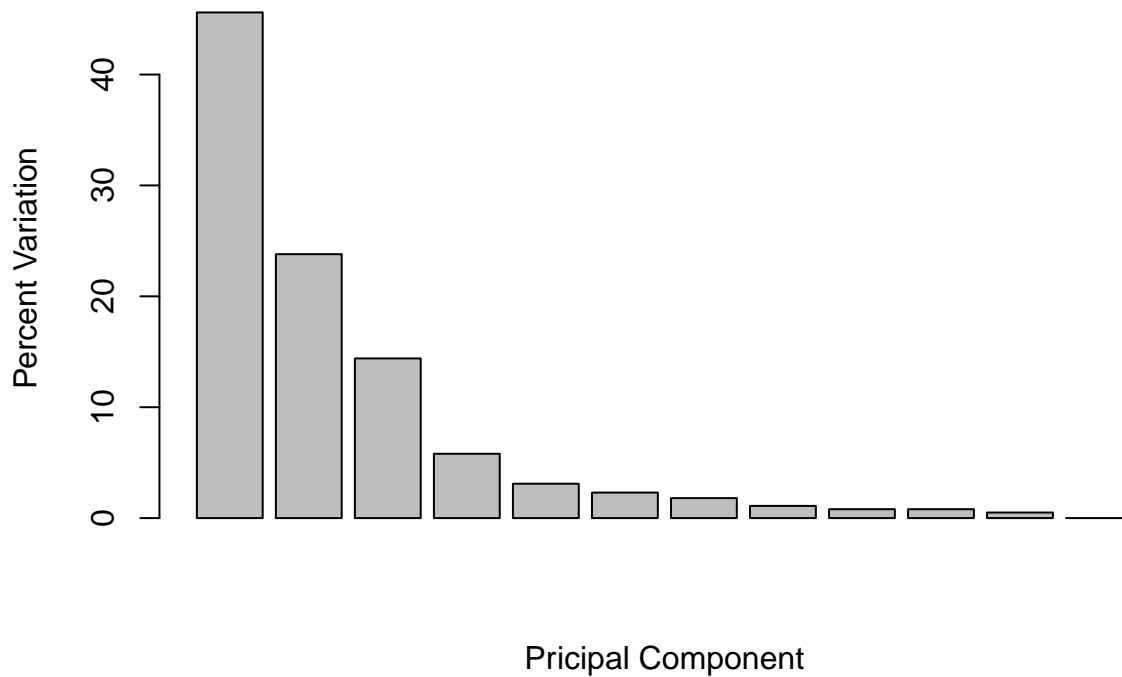
```
## calculate the percentage..
```

```
pca.flower.var.per <- round(pca.flower.var/sum(pca.flower.var)*100,1)
```

```
## Plotting the percentage is easy with barplot()
```

```
barplot(pca.flower.var.per, main = "Contribution of PCs in Flowers", xlab = "Principal Component", ylab = "Percentage of Variance Explained")
```

Contribution of PCs in Flowers



```
# 3) rotation -- the loading scores rotation
```

```
### Draw graph in Flowers
```

```
# Format the data the way ggplot2 likes it
```

```
dim(pca.flower$x)[1]
```

```
## [1] 12
```

```
pca.flower.data <- data.frame(ID = 1:dim(pca.flower$x)[1])
```

```
for (i in 1:dim(pca.flower$x)[1]){
```

```
  pcs <- paste0("PC", i)
```

```
  pca.flower.data[[pcs]] <- pca.flower$x[,i]
```

```
}
```

```
pca.flower.data$ID <- row.names(pca.flower$x)
```

```
pca.flower.data$ID <- row.names(pca.flower.data)
```

```
pca.flower.data$tissue <- "Flower"
```

```
pca.flower.data$Species <- c(rep("CG",4), rep("CR",4), rep("CO", 4))
```

```
pca.flower.data$Species <- factor(pca.flower.data$Species, levels = c("CG","CR","CO"))
```

```
#pca.flower.data$species <- c(c("CR","CG","CO"))
```

```
ZscoreFlower <- pca.flower.data
```

```
head(ZscoreFlower)
```

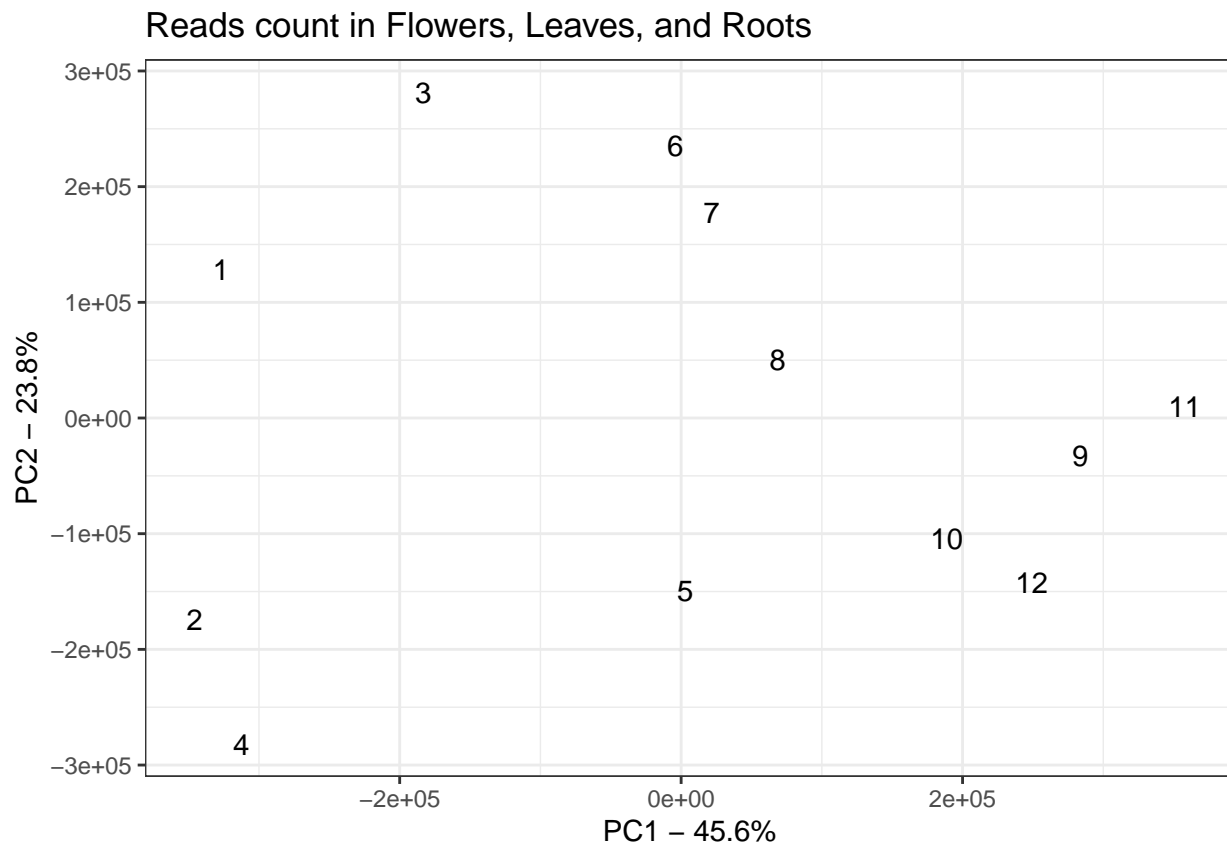
```
##      ID      PC1      PC2      PC3      PC4      PC5      PC6
## 1  1 -326902.058 128200.6 155351.0038 8347.959 -33883.23 -65092.43
## 2  2 -345819.621 -173820.0 62148.3607 6351.159 137084.95 49138.03
## 3  3 -183308.300 281754.4 108521.1338 -113651.996 -69034.07 37027.22
## 4  4 -312534.713 -281851.5 -878.2356 70210.075 -64291.80 -28275.65
## 5  5 3081.972 -149151.9 -224566.3383 -174966.191 -16428.40 31686.95
## 6  6 -4257.591 235252.3 -100864.0258 97625.278 61320.68 11853.72
##      PC7      PC8      PC9      PC10      PC11      PC12 tissue
## 1 -8207.9039 82616.243 -43.30106 13936.050 -5044.585 -1.625375e-09 Flower
## 2 -45283.8653 -14082.197 -17223.44109 17818.393 2786.605 -4.885407e-10 Flower
## 3 -23164.8764 -61672.001 -9213.33212 -9187.486 10439.164 -9.453557e-10 Flower
## 4 77513.4889 -36549.057 11864.09877 -22810.445 2737.705 -8.174623e-10 Flower
## 5 -935.6943 46954.285 4967.81207 -20801.146 17560.981 2.672199e-10 Flower
## 6 -10080.8005 4860.015 48589.98730 -58728.041 -7841.190 5.280437e-10 Flower
## Species
## 1 CG
## 2 CG
## 3 CG
## 4 CG
## 5 CR
## 6 CR
```

```
for(i in 2:13){
  ZscoreFlower[,i] <- scale(ZscoreFlower[,i])
}
head(ZscoreFlower)
```

```
##      ID      PC1      PC2      PC3      PC4      PC5      PC6
## 1  1 -1.32535108 0.7194510 1.120730417 0.09497549 -0.5303706 -1.1660503
## 2  2 -1.40204810 -0.9754631 0.448349586 0.07225771 2.1457762 0.8802470
## 3  3 -0.74318239 1.5811823 0.782891211 -1.29302909 -1.0805829 0.6632968
## 4  4 -1.26710190 -1.5817271 -0.006335752 0.79878641 -1.0063528 -0.5065232
## 5  5 0.01249516 -0.8370280 -1.620062437 -1.99060626 -0.2571520 0.5676324
## 6  6 -0.01726145 1.3202164 -0.727651440 1.11069165 0.9598461 0.2123446
##      PC7      PC8      PC9      PC10      PC11      PC12 tissue
## 1 -0.16804663 2.1562284 -0.001302233 0.4326155 -0.1960269 -1.8059882 Flower
## 2 -0.92713085 -0.3675359 -0.517976651 0.5531347 0.1082844 -0.5477301 Flower
## 3 -0.47427205 -1.6095977 -0.277081153 -0.2852062 0.4056543 -1.0533369 Flower
## 4 1.58699232 -0.9539058 0.356800138 -0.7081024 0.1063842 -0.9117834 Flower
## 5 -0.01915718 1.2254753 0.149401658 -0.6457280 0.6824002 0.2887523 Flower
## 6 -0.20639186 0.1268431 1.461292130 -1.8230889 -0.3046999 0.5774343 Flower
## Species
## 1 CG
## 2 CG
## 3 CG
## 4 CG
## 5 CR
## 6 CR
```

```
# graph parttern 1
ggplot(data = pca.flower.data, aes(x=PC1, y=PC2, label=ID)) +
  geom_text() +
  xlab(paste("PC1 - ", pca.flower.var.per[1], "%", sep = "")) +
```

```
ylab(paste("PC2 - ", pca.flower.var.per[2], "%", sep = "")) +
theme_bw() +
ggtitle("Reads count in Flowers, Leaves, and Roots")
```



```
# graph parttern 2
# PC1 vs PC2
p1 <- ggplot(ZscoreFlower, aes(PC1, PC2, col = Species, fill = Species)) +
  geom_point(size = 6, stroke = 2, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#f90052","darkviolet","#006beb")) +
  scale_fill_manual(values = c("#f90052","darkviolet","#006beb")) +
  # stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
  xlab(paste("PC1 - ", pca.flower.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.flower.var.per[2], "%", sep = "")) +
  stat_ellipse(geom="polygon", aes(fill = Species), # add frame
    alpha = 0.05,
    show.legend = FALSE,
    level = 0.90) +
  scale_x_continuous(breaks=seq(-2, 2, 1), limits=c(-2.1, 2.1)) +
  theme_minimal() +
  theme(panel.grid = element_blank(),
    panel.border = element_rect(fill= "transparent")) +
  theme(axis.text=element_text(size=16,face="bold"),
    axis.title=element_text(size=16,face="bold")) +
  theme(legend.position = "none") +
```

```

ggtitle("Flower") +
theme(plot.title = element_text(size=16,face="bold"))

#####
## PCA in Leaves
### Define leaf
head(TMM.Leaf)[1:5]

##
##          CG1_L    CG2_L    CG3_L    CG4_L    CR1_L
## Carubv10000018m.g 5411.7256 8093.0227 9968.4895 8261.5091 5948.4891
## Carubv10000019m.g 318.1181 747.9534 339.2023 958.7714 182.9841
## Carubv10000020m.g 1943.3791 2520.3253 2816.4044 3657.3709 3404.2495
## Carubv10000021m.g 2959.6804 5365.9697 4522.8324 4894.9462 4799.9591
## Carubv10000022m.g 3701.1556 6682.6636 4724.8480 5168.4485 5436.1021
## Carubv10000023m.g 4053.9332 4449.8566 5459.5006 6262.2343 5360.7120

dim(TMM.Leaf)

## [1] 17307    12

dim(na.omit(TMM.Leaf))

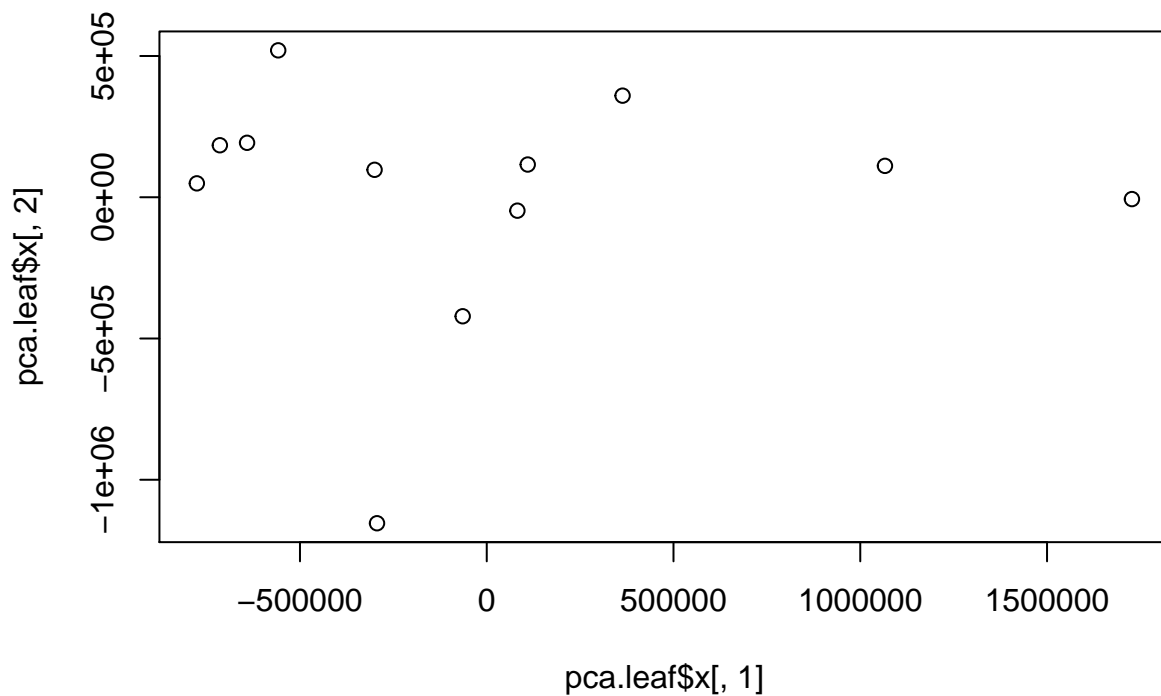
## [1] 17307    12

### perform PCA in Leaf
pca.leaf <- prcomp(t(na.omit(TMM.Leaf)))
summary(pca.leaf)

## Importance of components:
##
##          PC1      PC2      PC3      PC4      PC5
## Standard deviation 7.561e+05 4.285e+05 3.107e+05 2.216e+05 1.622e+05
## Proportion of Variance 5.838e-01 1.875e-01 9.855e-02 5.015e-02 2.687e-02
## Cumulative Proportion 5.838e-01 7.712e-01 8.698e-01 9.199e-01 9.468e-01
##
##          PC6      PC7      PC8      PC9      PC10
## Standard deviation 1.211e+05 1.069e+05 1.021e+05 7.987e+04 7.133e+04
## Proportion of Variance 1.496e-02 1.167e-02 1.065e-02 6.510e-03 5.200e-03
## Cumulative Proportion 9.618e-01 9.734e-01 9.841e-01 9.906e-01 9.958e-01
##
##          PC11      PC12
## Standard deviation 6.413e+04 1.163e-09
## Proportion of Variance 4.200e-03 0.000e+00
## Cumulative Proportion 1.000e+00 1.000e+00

# 1) X
## plot PC1 and PC2
plot(pca.leaf$x[,1],pca.leaf$x[,2])

```

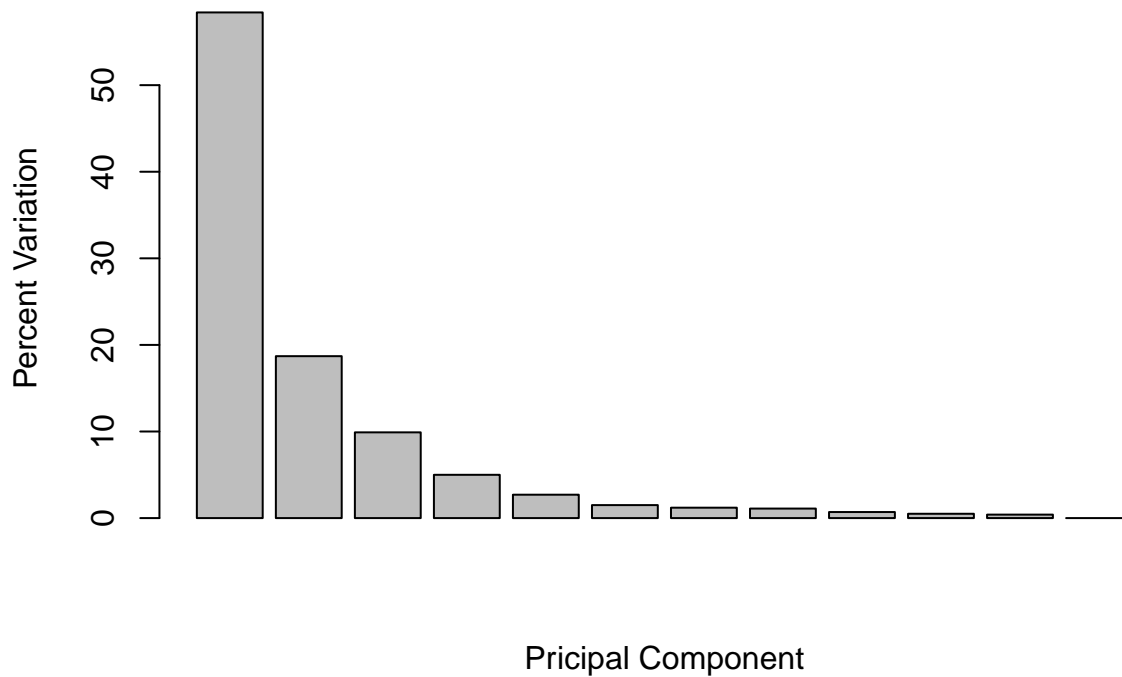


```
# 2) sdev
pca.leaf.var <- pca.leaf$sdev^2
head(pca.leaf.var)
```

```
## [1] 571670638623 183599311207 96505957875 49116163347 26310749196
## [6] 14654120331
```

```
## calculate the percentage..
pca.leaf.var.per <- round(pca.leaf.var/sum(pca.leaf.var)*100,1)
## Plotting the percentage is easy with barplot()
barplot(pca.leaf.var.per, main = "Contribution of PCs in Leaves", xlab = "Principal Component", ylab =
```

Contribution of PCs in Leaves



```
# 3) rotation -- the loading scores rotation
```

```
### Draw graph in Leaves
```

```
# Format the data the way ggplot2 likes it
```

```
dim(pca.leaf$x)[1]
```

```
## [1] 12
```

```
pca.leaf.data <- data.frame(ID = 1:dim(pca.leaf$x)[1])
```

```
for (i in 1:dim(pca.leaf$x)[1]){
```

```
  pcs <- paste0("PC", i)
```

```
  pca.leaf.data[[pcs]] <- pca.leaf$x[,i]
```

```
}
```

```
pca.leaf.data$ID <- row.names(pca.leaf$x)
```

```
pca.leaf.data$tissue <- "Leaf"
```

```
pca.leaf.data$Species <- c(rep("CG",4), rep("CR",4), rep("CO", 4))
```

```
pca.leaf.data$Species <- factor(pca.leaf.data$Species, levels = c("CG","CR","CO"))
```

```
ZscoreLeaf <- pca.leaf.data
```

```
head(ZscoreLeaf)
```

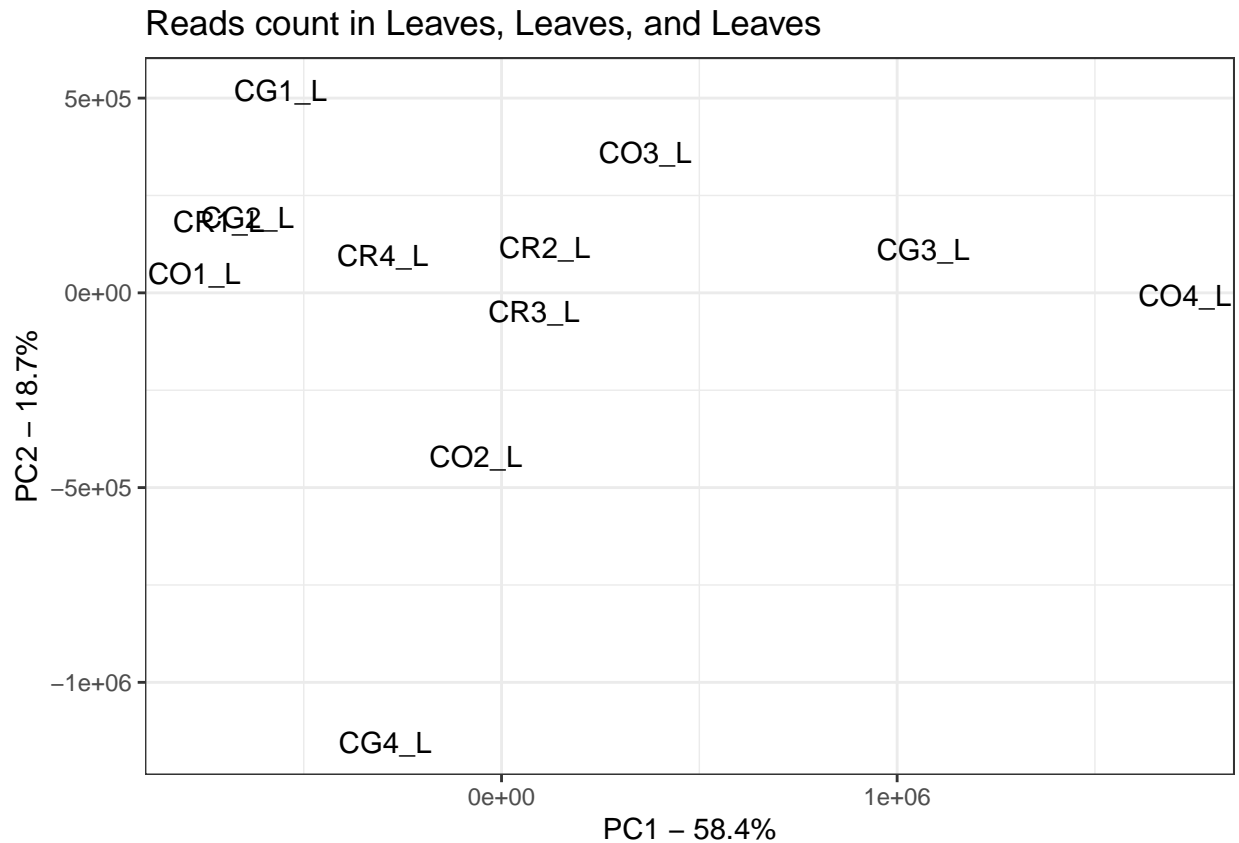
```
##      ID      PC1      PC2      PC3      PC4      PC5      PC6
## 1 CG1_L -558284.7  519933.8 -15655.37 -5317.096 186978.37 -79680.01
```

```
## 2 CG2_L -641695.2 192732.1 238297.58 25806.270 -80235.10 -175014.60
## 3 CG3_L 1066130.9 111055.4 -372594.89 383516.138 49494.19 -138165.97
## 4 CG4_L -293763.0 -1153923.5 -127065.44 10758.389 77744.59 -112676.68
## 5 CR1_L -714465.7 184021.7 -502423.39 -397331.844 32316.65 21645.42
## 6 CR2_L 109772.2 115744.7 -27783.09 32344.135 -457428.60 -13100.84
##      PC7      PC8      PC9      PC10      PC11      PC12 tissue
## 1 17894.53 -137590.77 11961.75 28958.102 -132516.543 2.164350e-09 Leaf
## 2 -79804.84 -21891.00 -41636.17 -155047.414 50658.583 1.570166e-10 Leaf
## 3 76916.97 123072.10 32904.87 -15638.491 -6782.283 -3.445514e-09 Leaf
## 4 -15757.88 -93994.23 -12251.03 47439.694 18083.597 -2.094976e-10 Leaf
## 5 148881.52 67233.20 -25503.16 -9553.748 50148.379 1.549079e-09 Leaf
## 6 34998.94 -53352.01 -47415.13 73062.446 -29658.038 1.306159e-09 Leaf
## Species
## 1 CG
## 2 CG
## 3 CG
## 4 CG
## 5 CR
## 6 CR
```

```
for(i in 2:13){
  ZscoreLeaf[,i] <- scale(ZscoreLeaf[,i])
}
head(ZscoreLeaf)
```

```
##      ID      PC1      PC2      PC3      PC4      PC5      PC6
## 1 CG1_L -0.7383849 1.2134238 -0.05039485 -0.02399177 1.1527225 -0.6582177
## 2 CG2_L -0.8487032 0.4497990 0.76708337 0.11644290 -0.4946497 -1.4457540
## 3 CG3_L 1.4100600 0.2591817 -1.19938836 1.73049929 0.3051319 -1.1413562
## 4 CG4_L -0.3885296 -2.6930318 -0.40902550 0.04854394 0.4792957 -0.9307953
## 5 CR1_L -0.9449491 0.4294706 -1.61730820 -1.79283844 0.1992323 0.1788076
## 6 CR2_L 0.1451842 0.2701256 -0.08943417 0.14594302 -2.8200494 -0.1082229
##      PC7      PC8      PC9      PC10      PC11      PC12 tissue
## 1 0.1673878 -1.3472679 0.1497683 0.4059747 -2.0664447 0.89431569 Leaf
## 2 -0.7465050 -0.2143533 -0.5213098 -2.1736687 0.7899629 0.06359728 Leaf
## 3 0.7194915 1.2051033 0.4119887 -0.2192420 -0.1057620 -1.42728020 Leaf
## 4 -0.1474013 -0.9203772 -0.1533903 0.6650751 0.2819931 -0.08808157 Leaf
## 5 1.3926573 0.6583372 -0.3193148 -0.1339376 0.7820069 0.63969068 Leaf
## 6 0.3273847 -0.5224148 -0.5936659 1.0242902 -0.4624834 0.53916018 Leaf
## Species
## 1 CG
## 2 CG
## 3 CG
## 4 CG
## 5 CR
## 6 CR
```

```
# graph parttern 1
ggplot(data = pca.leaf.data, aes(x=PC1, y=PC2, label=ID)) +
  geom_text() +
  xlab(paste("PC1 - ", pca.leaf.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.leaf.var.per[2], "%", sep = "")) +
  theme_bw() +
  ggtitle("Reads count in Leaves, Leaves, and Leaves")
```

```
# graph parttern 2
# PC1 vs PC2
p2 <- ggplot(ZscoreLeaf, aes(PC1, PC2, col = Species, fill = Species)) +
  geom_point(size = 6, stroke = 2, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#f90052","darkviolet","#006beb")) +
  scale_fill_manual(values = c("#f90052","darkviolet","#006beb")) +
  #stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
  xlab(paste("PC1 - ", pca.leaf.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.leaf.var.per[2], "%", sep = "")) +
  stat_ellipse(geom="polygon", aes(fill = Species), # add frame
    alpha = 0.05,
    show.legend = FALSE,
    level = 0.85) +
  xlim(-4,4) +
  theme_minimal() +
  theme(panel.grid = element_blank(),
    panel.border = element_rect(fill= "transparent")) +
  theme(axis.text=element_text(size=16,face="bold"),
    axis.title=element_text(size=16,face="bold")) +
  theme(legend.position = "none") +
  ggtitle("Leaf")+
  theme(plot.title = element_text(size=16,face="bold"))
```

```
#####
```

```
## PCA in Roots
```

```
### Define root
```

```
head(TMM.Root)[1:5]
```

```
##          CG1_R    CG2_R    CG3_R    CG4_R    CR1_R
## Carubv10000018m.g 13582.814 18591.777 10782.824 9705.911 8412.2444
## Carubv10000019m.g 1750.108 2336.253 1175.333 1692.486 540.4639
## Carubv10000020m.g 5463.298 6814.073 4037.237 4050.582 5296.0009
## Carubv10000021m.g 7312.001 10293.684 6695.445 5861.066 7109.8276
## Carubv10000022m.g 7785.271 10254.409 6654.702 5701.691 7748.3278
## Carubv10000023m.g 7828.652 10194.108 7969.401 6606.896 8037.3918
```

```
dim(TMM.Root)
```

```
## [1] 17307    12
```

```
dim(na.omit(TMM.Root))
```

```
## [1] 17307    12
```

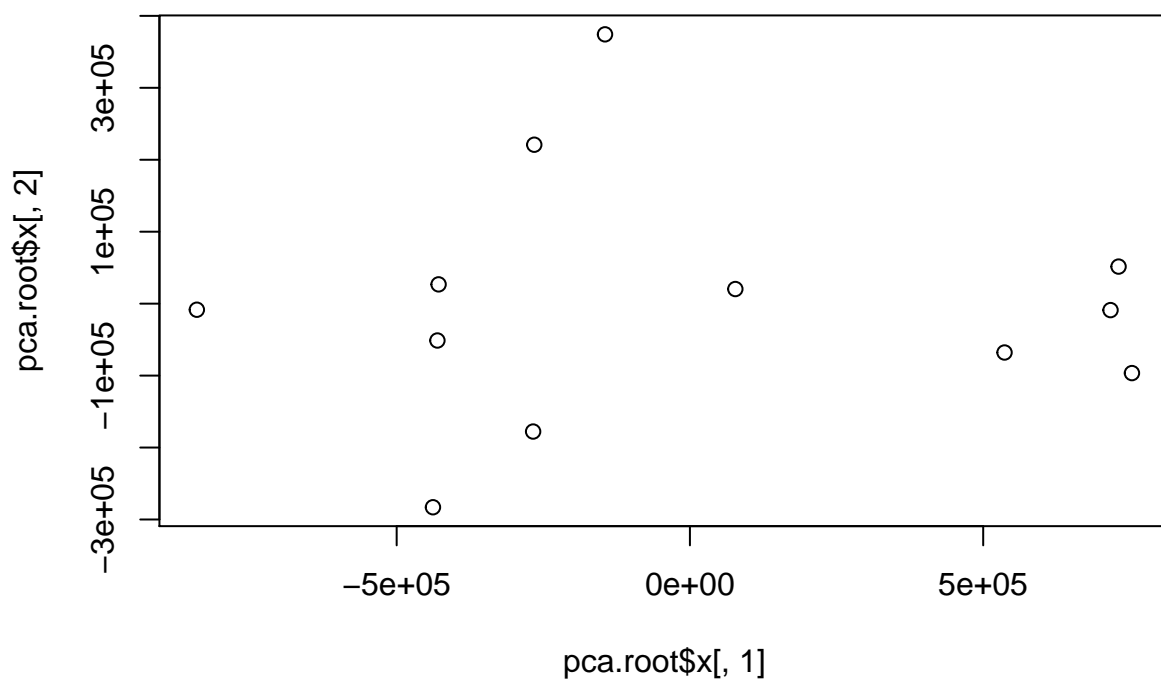
```
### perform PCA in Root
```

```
pca.root <- prcomp(t(na.omit(TMM.Root)))
```

```
# 1) X
```

```
## plot PC1 and PC2
```

```
plot(pca.root$x[,1],pca.root$x[,2])
```



```
# 2) sdev
```

```
pca.root.var <- pca.root$sdev^2  
head(pca.root.var)
```

```
## [1] 303683761259 29189074778 17038370821 12707001775 11002447048
```

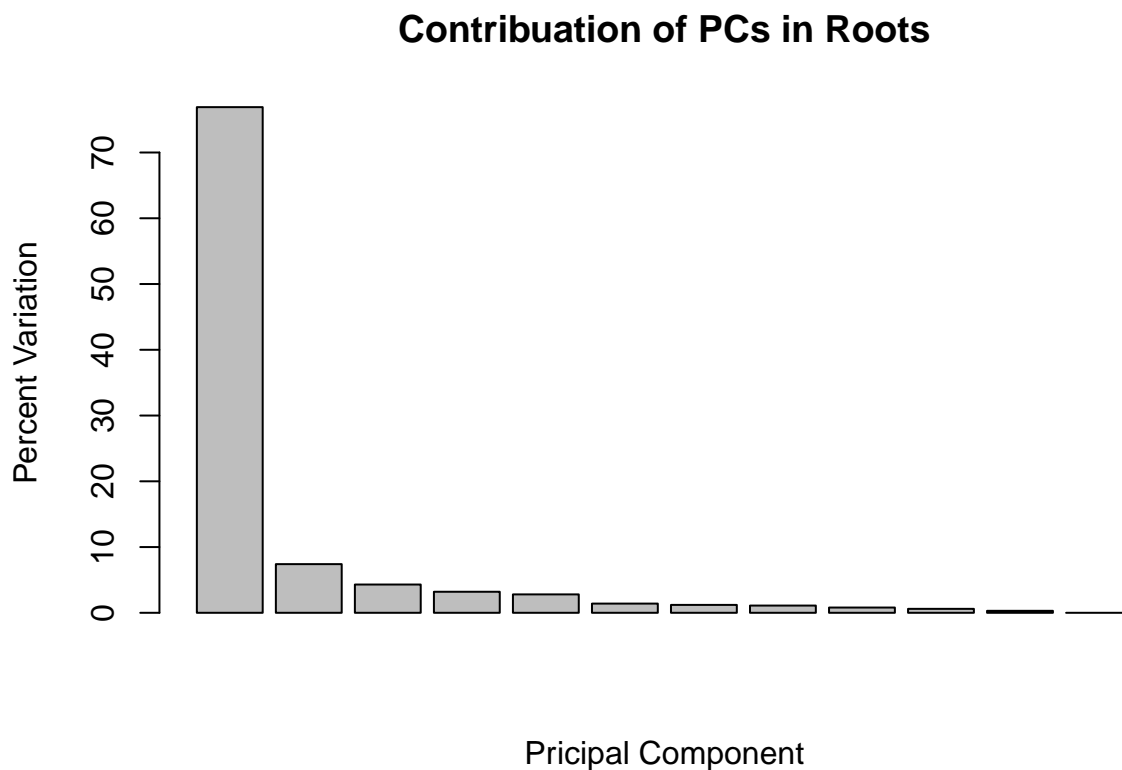
```
## [6] 5572785778
```

```
## calculate the percentage..
```

```
pca.root.var.per <- round(pca.root.var/sum(pca.root.var)*100,1)
```

```
## Plotting the percentage is easy with barplot()
```

```
barplot(pca.root.var.per, main = "Contribution of PCs in Roots", xlab = "Principal Component", ylab = "Percent Variation")
```



```
# 3) rotation -- the loading scores rotation
```

```
### Draw graph in Roots
```

```
library(ggplot2)
```

```
# Format the data the way ggplot2 likes it
```

```
dim(pca.root$x)[1]
```

```
## [1] 12
```

```
pca.root.data <- data.frame(ID = 1:dim(pca.root$x)[1])
```

```
for (i in 1:dim(pca.root$x)[1]){
```

```

pcs <- paste0("PC", i)
pca.root.data[[pcs]] <- pca.root$x[,i]
}

pca.root.data$ID <- row.names(pca.root$x)
pca.root.data$tissue <- "Root"
pca.root.data$Species <- c(rep("CG",4), rep("CR",4), rep("CO", 4))
pca.root.data$Species <- factor(pca.root.data$Species, levels = c("CG","CR","CO"))

ZscoreRoot <- pca.root.data
head(ZscoreRoot)

```

```

##      ID      PC1      PC2      PC3      PC4      PC5      PC6
## 1 CG1_R -265388.93  220897.27  30546.10  235041.65 -86107.91 -44112.952
## 2 CG2_R -428516.10  26781.77 -228283.11 -88752.86 -223122.16 -6918.289
## 3 CG3_R  77415.35   20300.63 -161616.25  137791.18  148007.42  50075.931
## 4 CG4_R -267337.64 -177765.01 -159511.42  35659.99  84211.33  85299.912
## 5 CR1_R -144773.23  374316.54 -14460.95 -164759.91  153728.95 -41870.567
## 6 CR2_R -438100.44 -282991.31  23207.28 -99606.69  67029.67 -85681.098
##      PC7      PC8      PC9      PC10      PC11      PC12 tissue
## 1  71586.426 -10617.60  58972.962 -36359.847  1773.9689  1.268819e-11 Root
## 2  -7389.171  37574.30 -52839.089  2823.173 -2634.8250 -2.225742e-09 Root
## 3 -14159.606 -74279.26 -99888.805 -12227.399  682.1711  5.467950e-10 Root
## 4 -43586.374  64996.32 109328.413  7006.303  3546.4037 -6.135187e-10 Root
## 5   6648.299  48202.61  9318.052  5584.389  3063.7982 -1.417067e-09 Root
## 6 110975.248 -28525.29 -3008.762 -55700.341 -11724.1136 -1.953455e-09 Root
## Species
## 1      CG
## 2      CG
## 3      CG
## 4      CG
## 5      CR
## 6      CR

```

```

for(i in 2:13){
  ZscoreRoot[,i] <- scale(ZscoreRoot[,i])
}
head(ZscoreRoot)

```

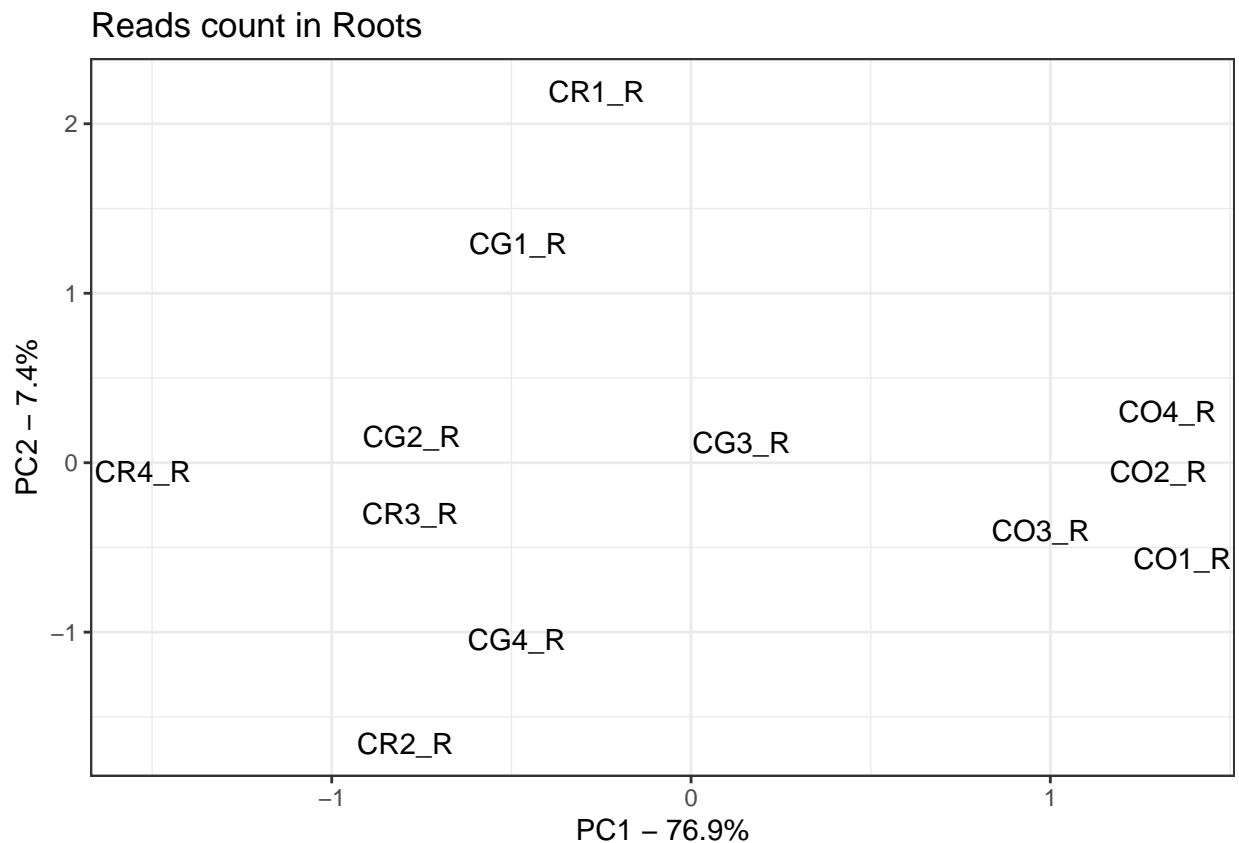
```

##      ID      PC1      PC2      PC3      PC4      PC5      PC6
## 1 CG1_R -0.4815840  1.2929454  0.2340140  2.0850827 -0.8209154 -0.59092171
## 2 CG2_R -0.7776002  0.1567578 -1.7488789 -0.7873373 -2.1271497 -0.09267498
## 3 CG3_R  0.1404806  0.1188227 -1.2381436  1.2223621  1.4110384  0.67079970
## 4 CG4_R -0.4851202 -1.0404857 -1.2220184  0.3163440  0.8028343  1.14264786
## 5 CR1_R -0.2627105  2.1909318 -0.1107855 -1.4616049  1.4655850 -0.56088350
## 6 CR2_R -0.7949923 -1.6563913  0.1777912 -0.8836229  0.6390318 -1.14775409
##      PC7      PC8      PC9      PC10      PC11      PC12 tissue
## 1  1.0211018 -0.1624989  1.05329609 -0.75690766  0.05369203  0.006187718 Root
## 2 -0.1053984  0.5750623 -0.94374106  0.05877036 -0.07974724 -1.246446404 Root
## 3 -0.2019712 -1.1368196 -1.78408010 -0.25453935  0.02064701  0.305075890 Root
## 4 -0.6217118  0.9947472  1.95267773  0.14585113  0.10733764 -0.344240029 Root
## 5  0.0948307  0.7377251  0.16642657  0.11625094  0.09273080 -0.793908781 Root
## 6  1.5829402 -0.4365702 -0.05373848 -1.15952124 -0.35484924 -1.094073557 Root

```

```
## Species
## 1      CG
## 2      CG
## 3      CG
## 4      CG
## 5      CR
## 6      CR
```

```
# graph parttern 1
ggplot(data = ZscoreRoot, aes(x=PC1, y=PC2, label=ID)) +
  geom_text() +
  xlab(paste("PC1 - ", pca.root.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.root.var.per[2], "%", sep = "")) +
  theme_bw() +
  ggtitle("Reads count in Roots")
```



```
# graph parttern 2
p3 <- ggplot(ZscoreRoot, aes(PC1, PC2, col = Species, fill = Species)) +
  geom_point(size = 6, stroke = 2, aes(shape = Species)) +
  scale_shape_manual(values=c(8,10,4))+
  scale_color_manual(values = c("#f90052","darkviolet","#006beb")) +
  scale_fill_manual(values = c("#f90052","darkviolet","#006beb")) +
  #stat_ellipse(aes(x= PC1, y=PC2, group = species), geom = "polygon", alpha = 0.1) +
  xlab(paste("PC1 - ", pca.root.var.per[1], "%", sep = "")) +
  ylab(paste("PC2 - ", pca.root.var.per[2], "%", sep = "")) +
```

```

stat_ellipse(geom="polygon", aes(fill = Species), # add frame
             alpha = 0.05,
             show.legend = FALSE,
             level = 0.85) +
xlim(-2,2) +
#labs(fill = "Species") +
theme_minimal() +
theme(panel.grid = element_blank(),
      panel.border = element_rect(fill= "transparent")) +
theme(axis.text=element_text(size=16,face="bold"),
      axis.title=element_text(size=16,face="bold"),
      legend.title = element_text(size=16, face = "bold"),
      legend.text = element_text(size=16)) +
theme(legend.position = "none") +
ggtitle("Root") +
theme(plot.title = element_text(size=16,face="bold"))

```

Combine all tissues into one figure

```

grid.newpage()
ggarrange(p3, p2, p1, ncol=3, nrow=1, common.legend = TRUE, legend="right") # set the common legend

```

