# Class 17: Analyzing Sequence Data in the Cloud

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# **Downstream Analysis**

Import Kallisto results using the tximport() function.

```
library(tximport)

# setup the folder and filenames to read
folders <- dir(pattern="SRR21568*")
samples <- sub("_quant", "", folders)
files <- file.path( folders, "abundance.h5" )
names(files) <- samples

txi.kallisto <- tximport(files, type = "kallisto", txOut = TRUE)</pre>
```

1 2 3 4

Look at the transcript count estimates:

```
head(txi.kallisto$counts)
```

#### SRR2156848 SRR2156849 SRR2156850 SRR2156851 ENST00000539570 0.00000 0 0 ENST00000576455 2.62037 ENST00000510508 0 0.00000 0 1.00000 0 0 ENST00000474471 ENST00000381700 0 0.00000 0 ENST00000445946 0.00000 0

Look at the total number of transcript counts in each sample by adding the column:

```
colSums(txi.kallisto$counts)

SRR2156848 SRR2156849 SRR2156850 SRR2156851
2563611 2600800 2372309 2111474
```

Look at how many transcripts are found by adding the total of the rows:

```
sum(rowSums(txi.kallisto$counts)>0)
```

[1] 94561

Remove the transcripts that have no reads in the data:

```
# add the rows and keep the data that are greater than zero
to.keep <- rowSums(txi.kallisto$counts) > 0

# create a new dataset
kset.nonzero <- txi.kallisto$counts[to.keep,]</pre>
```

```
# keep the data that change between the samples and remove the data that do not keep2 \leftarrow apply(kset.nonzero,1,sd)>0

# create a new dataset x \leftarrow kset.nonzero[keep2,]
```

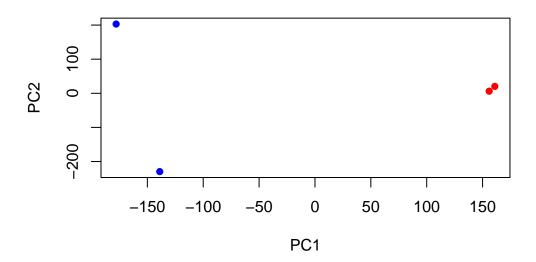
# **Principal Component Analysis**

```
# transpose the x dataset and scale it
pca <- prcomp(t(x), scale=TRUE)</pre>
```

```
summary(pca)
```

#### Importance of components:

```
PC1 PC2 PC3 PC4
Standard deviation 183.6379 177.3605 171.3020 1e+00
Proportion of Variance 0.3568 0.3328 0.3104 1e-05
Cumulative Proportion 0.3568 0.6895 1.0000 1e+00
```



Q. Use ggplot to make a similar figure of PC1 vs PC2 and a seperate figure PC1 vs PC3 and PC2 vs PC3.

First create a dataframe for grouping the control and treatment groups:

```
# create a dataframe and group the samples into control and treatment
# use factor() to turn the characters into a factor for easier coloring using discrete value
colors <- data.frame(group=factor(c('control','control', 'treatment','treatment')))
# make the rownames the sample names
rownames(colors) <- rownames(pca$x)
colors</pre>
```

group SRR2156848 control SRR2156849 control SRR2156850 treatment SRR2156851 treatment

Add the group dataframe as another column into a PCA dataframe:

```
# convert the pca$x into a dataframe to add the group column
new_pca <- as.data.frame(pca$x)

# add the group column
new_pca$group <- colors$group
new_pca</pre>
```

```
        PC1
        PC2
        PC3
        PC4
        group

        SRR2156848
        -177.9368
        203.031882
        -4.507483
        0.8660196
        control

        SRR2156849
        -138.9188
        -229.558755
        8.656814
        0.8659919
        control

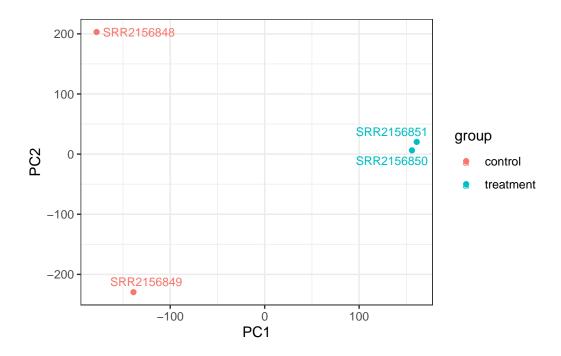
        SRR2156850
        155.8981
        6.206921
        -211.755452
        0.8660168
        treatment

        SRR2156851
        160.9486
        20.312009
        207.599341
        0.8660462
        treatment
```

PC1 vs PC2

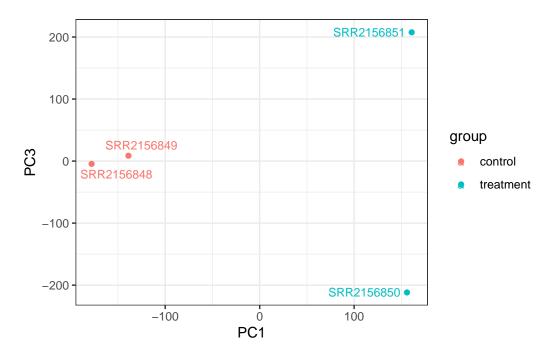
```
library(ggplot2)
library(ggrepel)
```

```
ggplot(new_pca) +
  aes(PC1, PC2, label=rownames(new_pca), col=group) +
  geom_point() +
  geom_text_repel(size=3) +
  theme_bw()
```



# PC1 vs PC3

```
ggplot(new_pca) +
  aes(PC1, PC3, label=rownames(new_pca), col=group) +
  geom_point() +
  geom_text_repel(size=3) +
  theme_bw()
```



## PC2 vs PC3

```
ggplot(new_pca) +
  aes(PC2, PC3, label=rownames(new_pca), col=group) +
  geom_point() +
  geom_text_repel(size=3) +
  theme_bw()
```

