Fedorczyk_hm_lab4

load data

pdf ## 2

```
data(bladderdata)

# sample info
pheno = pData(bladderEset)
# expression data
edata = exprs(bladderEset)
row.variances <- apply(edata, 1, function(x) var(x))
edata <- edata[row.variances < 6,]
edata.log <- log2(edata)</pre>
```

Homework Problem 1: Create a table to show the batch effects (refer to Figure 1 in Gilad and Mizrahi-Man, 2015). There are 5 batches (pheno\$batch); how are biological variables and other variables related to study design are distributed among those 5 batches? Explain what could be a problem. Prepare this into a PDF file.

```
batch_table <- pheno[, c('batch', 'outcome', 'cancer')] %>% unique()

batch_table <- batch_table[order(batch_table$batch),] %>% as.data.table()

batch_table <- batch_table %>% group_by(batch)

batch_table <- batch_table %>% mutate(result = paste0('(', outcome, ', ', cancer, ')', collapse = ", ")

batch_table <- batch_table[, c('batch', 'result')] %>% unique()

result <- data.frame(batch_table$result %>% t)

colnames(result) <- paste0('batch', ' ', batch_table$batch)

rownames(result) <- '(outcome, cancer)'

pdf("Fedorczyk_problem1.pdf", height = 15, width = 15)

grid.table(result)

dev.off()</pre>
```

Homework Problem 2: Make heatmaps, BEFORE and AFTER cleaning the data using ComBat, where columns are arranged according to the study design. You must sort the columns such that 5 batches are shown. Cluster the rows, but do not cluster the columns (samples) when drawing a heatmap. The general idea is that you want to see if the Combat-cleaned data are any improvement in the general patterns.

```
pheno_ordered <- pheno[order(pheno$batch, decreasing = FALSE),]</pre>
samples_ordered <- as.array(rownames(pheno_ordered))</pre>
edata_ordered <- edata[, samples_ordered]</pre>
combat_edata = ComBat(dat=edata, batch=pheno$batch, mod=model.matrix(~1, data=pheno), par.prior=TRUE, p
## Found5batches
## Adjusting forOcovariate(s) or covariate level(s)
## Standardizing Data across genes
## Fitting L/S model and finding priors
## Finding parametric adjustments
## Adjusting the Data
           Density Plot of First Batch γ
                                                           Normal Q–Q Plot of First Batch \hat{\gamma}
                                                  Sample Quantiles
Density
                                                       0.5
                                                       -1.5
     0.0
                         0
                                  1
                                           2
                                                                     -2
                                                                             0
                                                                                     2
                -1
           N = 22281 Bandwidth = 0.05758
                                                                   Theoretical Quantiles
           Density Plot of First Batch \hat{\delta}
                                                      Inverse Gamma Q-Q Plot of First Batch
                                                  Sample Quantiles
                                                                             ത്താ
Density
     0.8
                                                       \alpha
     0.0
                                                        0
                       2
                            3
                                        5
                                                                  2
           0
                 1
                                  4
                                                                                     10
                                                                                         12
            N = 22281 Bandwidth = 0.0534
                                                                   Theoretical Quantiles
combat_edata_ordered <- combat_edata[, samples_ordered]</pre>
my_palette <- colorRampPalette(c("blue", "white", "darkred"))(n = 299)
# pdf("Fedorczyk_problem2_before.pdf", height = 10, width = 10)
# heatmap.2(edata ordered,
```

main = "Bladder Cancer Data Clustered", # heat map title

```
#
            notecol="black", # change font color of cell labels to black
#
            density.info="none", # turns off density plot inside color legend
            trace="none",
#
                                  # turns off trace lines inside the heat map
            margins = c(12,9),
#
                                # widens margins around plot
#
            col=my_palette,
                                # use on color palette defined earlier
#
            dendrogram="none",
                                  # only draw a row dendrogram
#
            scale = "row",
#
            Colv = FALSE)
# dev.off()
# pdf("Fedorczyk_problem2_after.pdf", height = 10, width = 10)
# heatmap.2(combat_edata_ordered,
            main = "Bladder Cancer Data Cleaned by ComBat", # heat map title
#
#
            notecol="black",
                                  # change font color of cell labels to black
            density.info="none", # turns off density plot inside color legend
#
#
            trace="none",
                                 # turns off trace lines inside the heat map
#
            margins = c(12,9),
                                # widens margins around plot
                               # widens many was a county

# use on color palette defined earlier
#
            col=my_palette,
            dendrogram="none",
#
                                  # only draw a row dendrogram
            scale = "row",
#
            Colv = FALSE)
# dev.off()
```

Homework Problem 3: Make heatmaps of Pearson correlations statistics of samples. For example, see Figure 2 and 3 freom Gilad and Mizrahi-Man (2015) F1000Research: https://f1000research.com/articles/4-121. First, compute the correlation statistics among columns. Second, create a heatmap using heatmap.2(). Make sure to create or add labels for samples (cancer vs. normal; batch numbers; others)

pdf ## 2

Homework Problem 4: Apply two different Linear Models to the Bottomly et al. data. First, using a conventional approach, create a linear model with a genetic strain (biological variable)

and an experimental number (technical variable) on **uncorrected** gene expression data. Second, create a linear model with a genetic strain (biological variables) on **corrected** gene expression data from ComBat. Make a scatter plots of coefficients and a histogram of p-values as done in this notebook. Make sure that you are pulling out the correct coefficients, not any or all coefficients.

```
con <- url("http://bowtie-bio.sourceforge.net/recount/ExpressionSets/bottomly_eset.RData")
load(file=con)
close(con)
save(bottomly.eset, file="bottomly.Rdata")

load(file="bottomly.Rdata")

pheno <- pData(bottomly.eset)
edata <- exprs(bottomly.eset)
edata <- edata[rowMeans(edata) > 10, ]
edata <- log2(as.matrix(edata) + 1)</pre>
```

ComBat

```
combat_edata = ComBat(dat=edata, batch=pheno$experiment.number, mod=model.matrix(~1, data=pheno), par.p
## Found3batches
## Adjusting forOcovariate(s) or covariate level(s)
## Standardizing Data across genes
## Fitting L/S model and finding priors
## Finding parametric adjustments
## Adjusting the Data
#Model 1
mod1 = lm(t(edata) ~ as.factor(pheno$strain) + as.factor(pheno$experiment.number))
mod1_tidy <- tidy(mod1)</pre>
#histogram
pdf('Fedorczyk_problem4_p_value_model1.pdf')
ggplot(mod1_tidy %>% filter(term == "as.factor(pheno$strain)DBA/2J")) + geom_histogram(aes(x=p.value),
dev.off()
## pdf
##
```

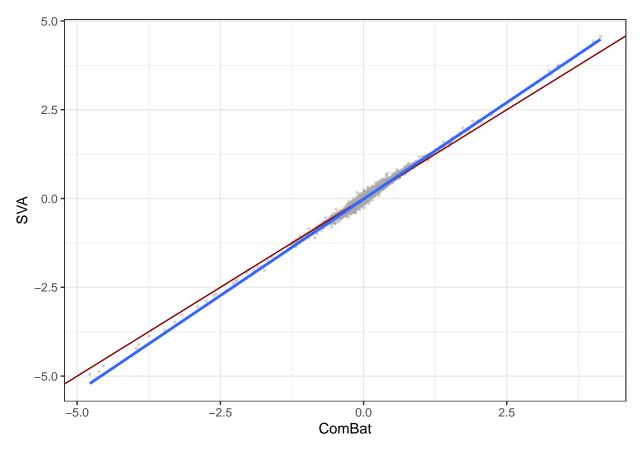
```
#Model 2
mod2 = lm(t(combat_edata) ~ as.factor(pheno$strain))
mod2_tidy <- tidy(mod2)</pre>
#histogram
pdf('Fedorczyk_problem4_p_value_model2.pdf')
ggplot(mod2_tidy %>% filter(term == 'as.factor(pheno$strain)DBA/2J')) + geom_histogram(aes(x=p.value),
dev.off()
## pdf
##
#Comparison
#filter : choose ROWS
#select : choose COLS
est_compare <- tibble(</pre>
  LinearModel = mod1_tidy %>% filter(term == "as.factor(pheno$strain)DBA/2J") %>% select("estimate") %>
  ComBat = mod2_tidy %>% filter(term == "as.factor(pheno$strain)DBA/2J") %>% select("estimate") %>% unl
pdf('Fedorczyk_problem4_compare.pdf')
ggplot(est_compare, aes(x=LinearModel, y=ComBat)) +
     geom_point(col="darkgrey", alpha=.5, size=.5) + geom_abline(intercept=0, slope=1, col="darkred") +
## 'geom_smooth()' using formula = 'y ~ x'
dev.off()
## pdf
##
     Homework Problem 5: Apply ComBat and SVA to the Bottomly et al. data. Make a scatter
     plots of coefficients and a histogram of p-values, comparing results based on ComBat and SVA.
     Assume that the biological variables in Bottomly et al data is the genetic strains. Make sure that
     you are pulling out the correct coefficients/pvalues, not any or all of them.
con <- url("http://bowtie-bio.sourceforge.net/recount/ExpressionSets/bottomly_eset.RData")</pre>
load(file=con)
close(con)
save(bottomly.eset, file="bottomly.Rdata")
load(file="bottomly.Rdata")
pheno <- pData(bottomly.eset)</pre>
edata <- exprs(bottomly.eset)</pre>
```

ComBat

edata <- edata[rowMeans(edata) > 10,]
edata <- log2(as.matrix(edata) + 1)</pre>

```
combat_edata = ComBat(dat=edata, batch=pheno$experiment.number, mod=model.matrix(~1, data=pheno), par.p
## Found3batches
## Adjusting forOcovariate(s) or covariate level(s)
## Standardizing Data across genes
## Fitting L/S model and finding priors
## Finding parametric adjustments
## Adjusting the Data
modcombat = lm(t(combat_edata) ~ as.factor(pheno$strain))
modcombat_tidy <- tidy(modcombat)</pre>
sva
modsva = model.matrix(~as.factor(strain), data=pheno)
modsva0 = model.matrix(~1, data=pheno)
sva_output = sva::sva(edata, modsva, modsva0, n.sv=sva::num.sv(edata,modsva,method="leek"))
## Number of significant surrogate variables is: 1
## Iteration (out of 5 ):1 2 3 4 5
modsva = lm(t(edata) ~ as.factor(pheno$strain) + sva_output$sv)
modsva_tidy <- tidy(modsva)</pre>
comparison
#Coefficients
est_compare <- tibble(</pre>
 ComBat = modcombat_tidy %>% filter(term == "as.factor(pheno$strain)DBA/2J") %>% select("estimate") %>
 SVA = modsva_tidy %>% filter(term == "as.factor(pheno$strain)DBA/2J") %>% select("estimate") %>% unli
ggplot(est_compare, aes(x=ComBat, y=SVA)) +
  geom_point(col="darkgrey", alpha=.5, size=.5) + geom_abline(intercept=0, slope=1, col="darkred") + ge
```

'geom_smooth()' using formula = 'y ~ x'



```
#save to pdf
pdf("Fedorczyk_problem5_coef.pdf")
ggplot(est_compare, aes(x=ComBat, y=SVA)) +
    geom_point(col="darkgrey", alpha=.5, size=.5) + geom_abline(intercept=0, slope=1, col="darkred") + ge

## 'geom_smooth()' using formula = 'y ~ x'

dev.off()

## pdf
## 2

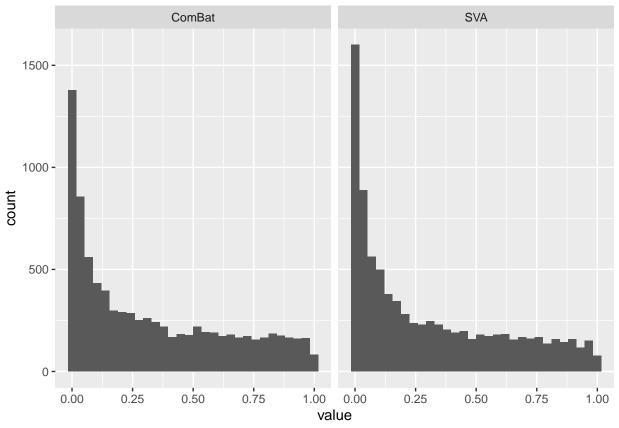
# ggsave('problem5_coef.png')
```

```
#pvalues
```

```
pvalues <- tibble(
   ComBat = modcombat_tidy %>% filter(term == "as.factor(pheno$strain)DBA/2J") %>% select("p.value") %>%
   SVA = modsva_tidy %>% filter(term == "as.factor(pheno$strain)DBA/2J") %>% select("p.value") %>% unlis

pvalues_gather <- gather(pvalues)
ggplot(pvalues_gather, aes(x=value)) + geom_histogram() + facet_wrap(~key)</pre>
```

'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



```
#save to pdf
pdf("Fedorczyk_problem5_pvalue.pdf")
ggplot(pvalues_gather, aes(x=value)) + geom_histogram() + facet_wrap(~key)
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```

dev.off()

pdf ## 2

ggsave('problem5_pvalue.png')