Pertusis challenge 1.1

Slim Fourati

2024-01-12

```
suppressPackageStartupMessages(library(package = "knitr"))
suppressPackageStartupMessages(library(package = "glmnet"))
suppressPackageStartupMessages(library(package = "edgeR"))
suppressPackageStartupMessages(library(package = "biomaRt"))
suppressPackageStartupMessages(library(package = "GSVA"))
suppressPackageStartupMessages(library(package = "agua"))
suppressPackageStartupMessages(library(package = "tidyverse"))
workDir <- "/Users/iew5629/Desktop/Projects/20231013_Pertussis/Workspace"</pre>
options(readr.show_col_types = FALSE)
agua::h2o_start()
# 2020
pts2020DF <- read_tsv(file</pre>
                                      = file.path(workDir, "documents/2020LD_subject.tsv"))
sample2020DF <- read_tsv(file</pre>
                                         = file.path(workDir, "documents/2020LD_specimen.tsv"))
ab2020DF <- read_tsv(file = file.path(workDir, "documents/2020LD_plasma_ab_titer.tsv"))
fcm2020DF <- read_tsv(file = file.path(workDir, "documents/2020LD_pbmc_cell_frequency.tsv"))</pre>
olink2020DF <- read_tsv(file = file.path(workDir, "documents/2020LD_plasma_cytokine_concentration.tsv")</pre>
ge2020DF <- read_tsv(file = file.path(workDir, "documents/2020LD_pbmc_gene_expression.tsv"))</pre>
# 2021
pts2021DF <- read_tsv(file = file.path(workDir, "documents/2021LD_subject.tsv"))</pre>
sample2021DF <- read_tsv(file = file.path(workDir, "documents/2021LD_specimen.tsv"))</pre>
ab2021DF <- read_tsv(file = file.path(workDir, "documents/2021LD_plasma_ab_titer.tsv"))
fcm2021DF <- read_tsv(file = file.path(workDir, "documents/2021LD_pbmc_cell_frequency.tsv"))</pre>
olink2021DF <- read_tsv(file = file.path(workDir, "documents/2021LD_plasma_cytokine_concentration.tsv")
ge2021DF <- read_tsv(file = file.path(workDir, "documents/2021LD_pbmc_gene_expression.tsv"))</pre>
# 2021
pts2022DF <- read_tsv(file = file.path(workDir, "documents/2022BD_subject.tsv"))</pre>
sample2022DF <- read_tsv(file = file.path(workDir, "documents/2022BD_specimen.tsv"))</pre>
ab2022DF <- read_tsv(file = file.path(workDir, "documents/2022BD_plasma_ab_titer.tsv"))
```

Challenge1.1

```
yDF <- ab2020DF %>%
filter(isotype %in% "IgG" & antigen %in% "PT") %>%
merge(y = sample2020DF, by = "specimen_id", all.x = TRUE) %>%
merge(y = pts2020DF, by = "subject_id", all.x = TRUE) %>%
filter(planned_day_relative_to_boost %in% c(0, 14)) %>%
select(subject_id, planned_day_relative_to_boost, MFI_normalised) %>%
pivot_wider(names_from = planned_day_relative_to_boost, values_from = MFI_normalised, names_prefix =
```

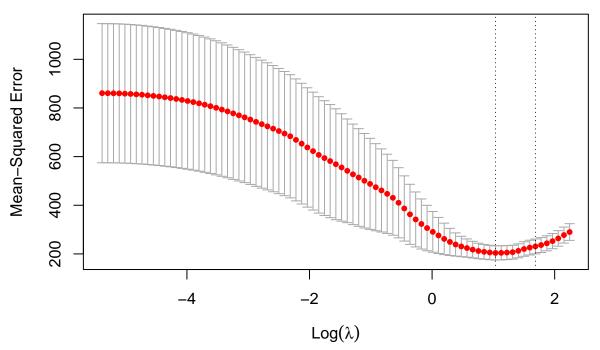
```
mutate(MFI_FC = MFI_14/MFI_0)
```

pts info

BL ab

```
xDF <- ab2020DF %>%
 merge(y = sample2020DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2020DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% 0 &
           grepl(pattern = "IgG", isotype)) %>%
  mutate(cname = paste0(isotype, "_", antigen),
         cname = make.names(cname)) %>%
  dplyr::select(subject_id, cname, MFI_normalised) %>%
  distinct() %>%
  pivot_wider(names_from = cname, values_from = MFI_normalised) %>%
  column_to_rownames(var = "subject_id") %>%
  slice(match(yDF$subject_id, table = rownames(.)))
x2DF <- ab2021DF %>%
  merge(y = sample2021DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2021DF, by = "subject_id", all.x = TRUE) %>%
 filter(planned_day_relative_to_boost %in% 0) %>%
  mutate(cname = paste0(isotype, "_", antigen),
         cname = make.names(cname)) %>%
  dplyr::select(subject_id, cname, MFI_normalised) %>%
  distinct() %>%
  pivot_wider(names_from = cname, values_from = MFI_normalised) %>%
  column_to_rownames(var = "subject_id")
xDF <- xDF[, intersect(colnames(xDF), colnames(x2DF))]</pre>
x2DF <- x2DF[, colnames(xDF)]</pre>
xDF <- scale(xDF)</pre>
x2DF <- scale(x2DF)</pre>
```

30 30 30 29 28 28 25 24 23 16 9 9 7 4 1 1



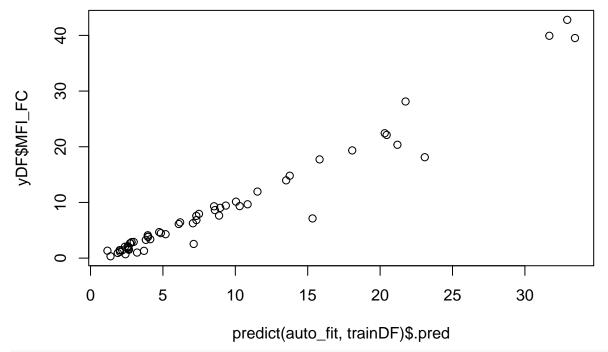
```
coef(fit, s = fit$lambda.min) %>%
  as.matrix() %>%
  as.data.frame() %>%
  rownames_to_column() %>%
  filter(s1 != 0) %>%
  kable()
```

rowname	s1
(Intercept)	29.5000000
IgG_PT	6.2777634
$IgG1_FIM2.3$	0.7633756
IgG1_OVA	0.4557108
$IgG2_FHA$	0.6514573
$IgG3_FIM2.3$	1.2691300
${\rm IgG4_FIM2.3}$	0.9510652

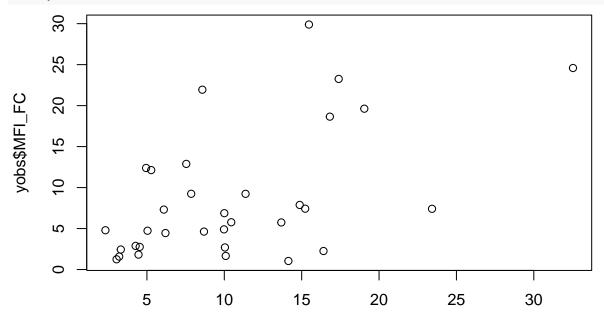
```
yhat <- predict(fit, newx = as.matrix(x2DF), s = fit$lambda.min, type = "response")

yobs <- ab2021DF %>%
  filter(isotype %in% "IgG" & antigen %in% "PT") %>%
  merge(y = sample2021DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2021DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% c(0, 14)) %>%
```

```
select(subject_id, planned_day_relative_to_boost, MFI_normalised) %>%
  pivot_wider(names_from = planned_day_relative_to_boost, values_from = MFI_normalised, names_prefix =
  mutate(MFI_FC = MFI_14/MFI_0) %>%
  slice(match(rownames(x2DF), table = .$subject_id))
plot(yhat[, "s1"], rank(-1 * yobs$MFI_FC))
                                                                                   0
                                                            0
                                     0
      30
                                                            0
                             0
                                     00
rank(-1 * yobs$MFI_FC)
                   0
     25
                     0
     20
                    00
               0
                             8
     15
      10
      2
                           0
      0
                  25
                            30
                                       35
                                                 40
                                                            45
                                                                      50
                                                                                55
                                           yhat[, "s1"]
cor.test(yhat[, "s1"], rank(-1 * yobs$MFI_FC))
##
##
    Pearson's product-moment correlation
##
## data: yhat[, "s1"] and rank(-1 * yobs$MFI_FC)
## t = 5.5282, df = 31, p-value = 4.723e-06
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
   0.4765408 0.8437866
## sample estimates:
##
       cor
## 0.70458
set.seed(seed= 3)
trainDF <- xDF %>%
  as.data.frame() %>%
  mutate(MFI_FC = yDF$MFI_FC)
auto_fit <- auto_ml() %>%
     set_engine("h2o", max_runtime_secs = 5) %>%
     set_mode("regression") %>%
     fit(MFI_FC ~ ., data = trainDF)
plot(predict(auto_fit, trainDF)$.pred,
     yDF$MFI_FC)
```



plot(predict(auto_fit, new_data = as.data.frame(x2DF))\$.pred,
 yobs\$MFI_FC)



predict(auto_fit, new_data = as.data.frame(x2DF))\$.pred

```
cor.test(predict(auto_fit, new_data = as.data.frame(x2DF))$.pred,
    yobs$MFI_FC)
```

```
##
## Pearson's product-moment correlation
##
## data: predict(auto_fit, new_data = as.data.frame(x2DF))$.pred and yobs$MFI_FC
## t = 3.5888, df = 31, p-value = 0.001128
```

```
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.2438128 0.7462781
## sample estimates:
         cor
## 0.5417736
cor.test(rank(-1 * predict(auto_fit, new_data = as.data.frame(x2DF))$.pred),
    rank(-1 * yobs$MFI FC))
##
## Pearson's product-moment correlation
## data: rank(-1 * predict(auto_fit, new_data = as.data.frame(x2DF))$.pred) and rank(-1 * yobs$MFI_FC)
## t = 3.1238, df = 31, p-value = 0.003854
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.1754721 0.7128648
## sample estimates:
##
         cor
## 0.4893048
```

FCM

```
xDF <- fcm2020DF %>%
  merge(y = sample2020DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2020DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% 0) %>%
  dplyr::select(subject_id, cell_type_name, percent_live_cell) %>%
  pivot_wider(names_from = cell_type_name, values_from = percent_live_cell) %>%
  column_to_rownames(var = "subject_id") %>%
  slice(match(yDF$subject_id, table = rownames(.)))
xDF[is.na(xDF)] <- 0</pre>
x2DF <- fcm2021DF %>%
  merge(y = sample2021DF, by = "specimen id", all.x = TRUE) %>%
  merge(y = pts2021DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% 0) %>%
  dplyr::select(subject_id, cell_type_name, percent_live_cell) %>%
  pivot_wider(names_from = cell_type_name, values_from = percent_live_cell) %>%
  column_to_rownames(var = "subject_id")
x2DF[is.na(x2DF)] \leftarrow 0
xDF <- xDF[, intersect(colnames(xDF), colnames(x2DF))]</pre>
x2DF <- x2DF[, colnames(xDF)]</pre>
xDF <- scale(xDF)</pre>
x2DF <- scale(x2DF)</pre>
fit <- cv.glmnet(x = as.matrix(xDF),</pre>
                 y = rank(-yDF[match(rownames(xDF), table = yDF$subject_id), "MFI_FC"]),
                 nfolds = nrow(xDF),
                 grouped = FALSE)
plot(fit)
coef(fit, s = fit$lambda.min) %>%
```

```
as.matrix() %>%
  as.data.frame() %>%
  rownames_to_column() %>%
  filter(s1 != 0) %>%
  kable()
yhat <- predict(fit, newx = as.matrix(x2DF), s = fit$lambda.min, type = "response")</pre>
yobs <- ab2021DF %>%
  filter(isotype %in% "IgG" & antigen %in% "PT") %>%
  merge(y = sample2021DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2021DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% c(0, 14)) %>%
  select(subject_id, planned_day_relative_to_boost, MFI_normalised) %>%
  pivot_wider(names_from = planned_day_relative_to_boost, values_from = MFI_normalised, names_prefix =
  mutate(MFI_FC = MFI_14/MFI_0) %>%
  slice(match(rownames(x2DF), table = .$subject_id))
plot(yhat[as.character(yobs$subject_id), "s1"], rank(-1 * yobs$MFI_FC))
cor.test(yhat[as.character(yobs$subject_id), "s1"], rank(-1 * yobs$MFI_FC))
```

Olink

```
xDF <- olink2020DF %>%
  merge(y = sample2020DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2020DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% 0) %>%
  dplyr::select(subject_id, protein_id, protein_expression) %>%
  pivot_wider(names_from = protein_id, values_from = protein_expression) %>%
  column_to_rownames(var = "subject_id") %>%
  slice(match(yDF$subject_id, table = rownames(.)))
x2DF <- olink2021DF %>%
  merge(y = sample2021DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2021DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% 0) %>%
  dplyr::select(subject_id, protein_id, protein_expression) %>%
  pivot_wider(names_from = protein_id, values_from = protein_expression) %>%
  column_to_rownames(var = "subject_id")
x2DF[is.na(x2DF)] \leftarrow 0
xDF <- xDF[, intersect(colnames(xDF), colnames(x2DF))]</pre>
x2DF <- x2DF[, colnames(xDF)]</pre>
xDF <- scale(xDF)</pre>
x2DF <- scale(x2DF)</pre>
fit <- cv.glmnet(x = as.matrix(xDF),</pre>
                 y = rank(-yDF[match(rownames(xDF), table = yDF$subject_id), "MFI_FC"]),
                 nfolds = nrow(xDF),
                 grouped = FALSE)
plot(fit)
coef(fit, s = fit$lambda.min) %>%
as.matrix() %>%
```

```
as.data.frame() %>%
  rownames_to_column() %>%
  filter(s1 != 0) %>%
  kable()
yhat <- predict(fit, newx = as.matrix(x2DF), s = fit$lambda.min, type = "response")</pre>
yobs <- ab2021DF %>%
  filter(isotype %in% "IgG" & antigen %in% "PT") %>%
  merge(y = sample2021DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2021DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% c(0, 14)) %>%
  select(subject_id, planned_day_relative_to_boost, MFI_normalised) %>%
  pivot_wider(names_from = planned_day_relative_to_boost, values_from = MFI_normalised, names_prefix =
  mutate(MFI_FC = MFI_14/MFI_0) %>%
  slice(match(rownames(x2DF), table = .$subject_id))
plot(yhat[as.character(yobs$subject_id), "s1"], rank(-1 * yobs$MFI_FC))
cor.test(yhat[as.character(yobs$subject_id), "s1"], rank(-1 * yobs$MFI_FC))
```

GE

```
# extract d0 GE
dORawMat <- ge2020DF %>%
  merge(y = sample2020DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2020DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned day relative to boost %in% 0) %>%
  dplyr::select(subject_id, versioned_ensembl_gene_id, raw_count) %>%
  pivot_wider(names_from = subject_id, values_from = raw_count) %>%
  column_to_rownames(var = "versioned_ensembl_gene_id")
# TMM normalization
dge <- DGEList(counts</pre>
                            = dORawMat.
               remove.zeros = TRUE)
dge <- calcNormFactors(object = dge, method = "TMM")</pre>
normalizedCounts <- cpm(dge, normalized.lib.sizes = TRUE, log = TRUE)</pre>
# convert ensembl_qene_id to qene_symbol
human <- useMart("ensembl", dataset = "hsapiens_gene_ensembl")</pre>
id2symbol <- getBM(attributes = c("ensembl_gene_id_version", "hgnc_symbol"),</pre>
                   filters = "ensembl_gene_id_version",
                   values
                              = rownames(normalizedCounts),
                              = human)
                   mart
geneDOMat <- normalizedCounts %>%
  as.data.frame() %>%
  rownames to column(var = "ensembl gene id version") %>%
  merge(y = id2symbol, by = "ensembl_gene_id_version") %>%
  select(-ensembl_gene_id_version) %>%
  slice(order(apply(select(., -hgnc_symbol), MARGIN = 1, FUN = var), decreasing = TRUE)) %>%
  filter(!duplicated(hgnc_symbol) & hgnc_symbol != "") %>%
  column_to_rownames(var = "hgnc_symbol") %>%
  as.matrix()
```

```
# GSVA with BTM
load(file = "/Users/iew5629/Desktop/Projects/20231013_Pertussis/Workspace/documents/GeneSets.rda")
BTM.geneSets <- BTM.geneSets[!grepl(pattern = "TBA", names(BTM.geneSets))]
flag <- BTM.geneSets %>%
  stack() %>%
  filter(values %in% rownames(geneDOMat)) %>%
  group by(ind) %>%
  summarize(n = n()) \%
  arrange(desc(n)) %>%
 filter(n >= 3)
gsDOMat <- gsva(expr = geneDOMat, gset.idx.list = BTM.geneSets[flag$ind], verbose = FALSE)
# prediction
xDF <- t(gsDOMat)</pre>
fit <- cv.glmnet(x = as.matrix(xDF),</pre>
                 y = rank(-yDF[match(rownames(xDF), table = yDF$subject_id), "MFI_FC"]),
                 nfolds = nrow(xDF),
                 grouped = FALSE)
plot(fit)
```

2022 test set

```
set.seed(seed= 3)
# train on 2020 and 2021 baseline ab only
yDF <- ab2020DF %>%
  filter(isotype %in% "IgG" & antigen %in% "PT") %>%
  merge(y = sample2020DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2020DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% c(0, 14)) %>%
  select(subject_id, planned_day_relative_to_boost, MFI_normalised) %>%
  pivot_wider(names_from = planned_day_relative_to_boost, values_from = MFI_normalised, names_prefix =
  mutate(MFI_FC = MFI_14/MFI_0)
xDF <- ab2020DF %>%
  merge(y = sample2020DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2020DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% 0 &
           grepl(pattern = "IgG", isotype)) %>%
  mutate(cname = paste0(isotype, "_", antigen),
         cname = make.names(cname)) %>%
  dplyr::select(subject_id, cname, MFI_normalised) %>%
  distinct() %>%
  pivot_wider(names_from = cname, values_from = MFI_normalised) %>%
  column_to_rownames(var = "subject_id") %>%
  slice(match(yDF$subject_id, table = rownames(.)))
x2DF <- ab2021DF %>%
  merge(y = sample2021DF, by = "specimen_id", all.x = TRUE) %>%
 merge(y = pts2021DF, by = "subject_id", all.x = TRUE) %>%
```

```
filter(planned_day_relative_to_boost %in% 0) %>%
  mutate(cname = paste0(isotype, "_", antigen),
         cname = make.names(cname)) %>%
  dplyr::select(subject_id, cname, MFI_normalised) %>%
  distinct() %>%
  pivot_wider(names_from = cname, values_from = MFI_normalised) %>%
  column_to_rownames(var = "subject_id")
yobs <- ab2021DF %>%
  filter(isotype %in% "IgG" & antigen %in% "PT") %>%
  merge(y = sample2021DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2021DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% c(0, 14)) %>%
  select(subject_id, planned_day_relative_to_boost, MFI_normalised) %>%
  pivot_wider(names_from = planned_day_relative_to_boost, values_from = MFI_normalised, names_prefix =
  mutate(MFI_FC = MFI_14/MFI_0) %>%
  slice(match(rownames(x2DF), table = .$subject_id))
x3DF <- ab2022DF %>%
  merge(y = sample2022DF, by = "specimen_id", all.x = TRUE) %>%
  merge(y = pts2022DF, by = "subject_id", all.x = TRUE) %>%
  filter(planned_day_relative_to_boost %in% 0) %>%
  mutate(cname = paste0(isotype, "_", antigen),
         cname = make.names(cname)) %>%
  dplyr::select(subject_id, cname, MFI_normalised) %>%
  distinct() %>%
  pivot_wider(names_from = cname, values_from = MFI_normalised) %>%
  column_to_rownames(var = "subject_id")
xDF <- xDF[, intersect(colnames(xDF), colnames(x2DF))]</pre>
x2DF <- x2DF[, colnames(xDF)]</pre>
x3DF <- x3DF[, colnames(xDF)]</pre>
xDF <- scale(xDF)
x2DF <- scale(x2DF)</pre>
x3DF <- scale(x3DF)</pre>
trainDF <- rbind(xDF, x2DF) %>%
  as.data.frame() %>%
  mutate(MFI_FC = c(yDF$MFI_FC, yobs$MFI_FC))
auto_fit <- auto_ml() %>%
     set_engine("h2o", max_runtime_secs = 5) %>%
     set_mode("regression") %>%
     fit(MFI_FC ~ ., data = trainDF)
yhat <- predict(auto_fit, new_data = x3DF)$.pred</pre>
rhat <- rank(-1 * yhat)</pre>
print(cbind(rownames(x3DF), rhat))
##
               rhat
## [1,] "97" "8"
## [2,] "98" "12"
```

[3,] "99" "10"

```
[4,] "100" "17"
##
   [5,] "101" "16"
   [6,] "102" "18"
   [7,] "103" "4"
##
   [8,] "104" "7"
## [9,] "105" "14"
## [10,] "106" "5"
## [11,] "107" "2"
## [12,] "108" "9"
## [13,] "109" "1"
## [14,] "110" "6"
## [15,] "111" "19"
## [16,] "112" "20"
## [17,] "114" "3"
## [18,] "115" "21"
## [19,] "116" "13"
## [20,] "117" "11"
## [21,] "118" "15"
```