## BNLEARN

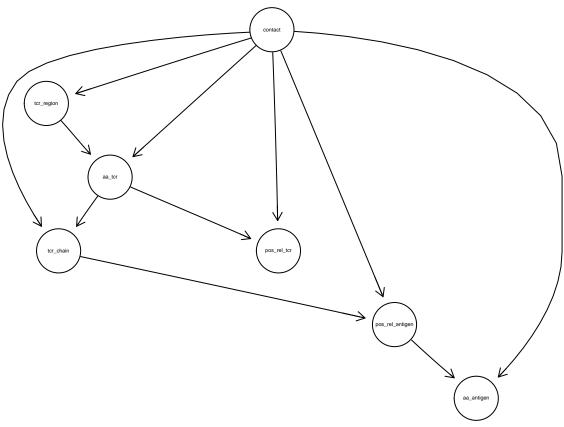
Load data and filter

```
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
library(RColorBrewer)
library(gplots)
##
## Attaching package: 'gplots'
## The following object is masked from 'package:stats':
##
##
       lowess
library(reshape2)
library(ggplot2)
library(bnlearn)
df <- read.table("structure.txt", header = T, sep="\t") %>%
  filter(mhc_type == "MHCI") %>%
  mutate(tcr_chain = as.factor(substr(as.character(tcr_v_allele), 1, 3)),
         pos_tcr = as.numeric(pos_tcr),
         len_tcr = as.numeric(len_tcr),
         pos_antigen = as.numeric(pos_antigen),
         len_antigen = as.numeric(len_antigen)) %>%
  select(tcr_region, tcr_chain, pos_tcr, len_tcr, aa_tcr, pos_antigen, len_antigen, aa_antigen, energy)
  mutate(contact = as.factor(energy < 0),</pre>
         pos_rel_tcr = cut(pos_tcr / (len_tcr - 1), 10),
         pos_rel_antigen = cut(pos_antigen / (len_antigen - 1), 10)) %>%
  select(tcr_region, tcr_chain, pos_rel_tcr, aa_tcr, pos_rel_antigen, aa_antigen, contact)
df$contact[is.na(df$contact)] <- "FALSE"</pre>
head(df)
     tcr_region tcr_chain pos_rel_tcr aa_tcr pos_rel_antigen aa_antigen
##
## 1
                                                   (-0.001, 0.1]
           CDR1
                      TRA (-0.001, 0.1]
                                             D
                                                                         T.
## 2
           CDR1
                      TRA (-0.001, 0.1]
                                             D
                                                      (0.1, 0.2]
                                                                         L
## 3
           CDR1
                      TRA (-0.001, 0.1]
                                             D
                                                      (0.2, 0.3]
                                                                         F
## 4
           CDR1
                      TRA (-0.001,0.1]
                                             D
                                                                         G
                                                      (0.3, 0.4]
           CDR1
                                                                         Y
## 5
                      TRA (-0.001, 0.1]
                                             D
                                                      (0.4, 0.5]
           CDR1
                      TRA (-0.001,0.1]
                                             D
## 6
                                                      (0.6, 0.7]
```

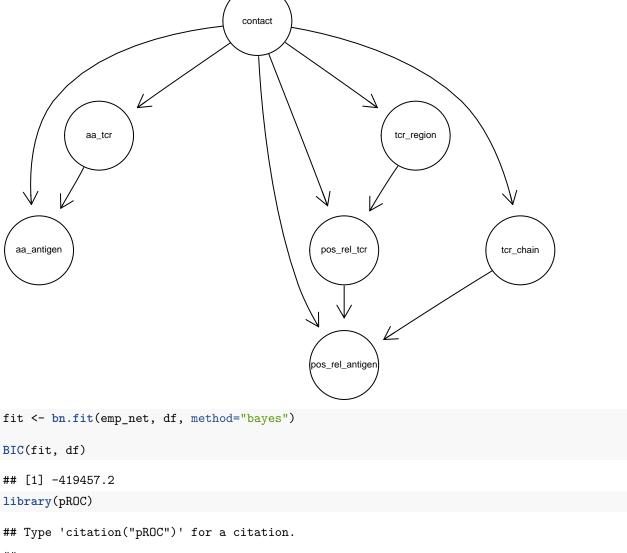
```
##
     contact
## 1
        TRUE
       FALSE
## 2
## 3
       FALSE
## 4
       FALSE
## 5
       FALSE
## 6
       FALSE
summary(df)
  tcr_region
                 tcr_chain
                                    pos_rel_tcr
                                                       aa_tcr
##
  CDR1: 9111
                 TRA:17301
                              (-0.001,0.1]:5769
                                                  S
                                                         : 4663
  CDR2: 6040
                 TRB:17293
                              (0.9,1]
                                          :5593
                                                  G
                                                          : 3981
## CDR3:19443
                              (0.4, 0.5]
                                          :4303
                                                          : 2443
                                                  Α
                                          :3259
##
                              (0.3, 0.4]
                                                  F
                                                          : 2367
##
                              (0.1, 0.2]
                                          :3192
                                                  Y
                                                          : 2309
##
                              (0.7, 0.8]
                                          :3059
                                                          : 2103
##
                              (Other)
                                          :9419
                                                   (Other):16728
##
                           aa_antigen
                                           contact
        pos_rel_antigen
##
   (-0.001,0.1]: 4220
                                          FALSE: 29819
                                : 5499
   (0.4, 0.5]
               : 3862
                         G
                                 : 3468
                                          TRUE : 4775
##
##
   (0.9,1]
                : 3862
                         V
                                 : 2893
##
   (0.1, 0.2]
                : 3627
                         Y
                                 : 2830
## (0.2,0.3]
                : 3627
                         F
                                 : 2420
   (0.7, 0.8]
                : 3627
                                 : 2381
                         Α
   (Other)
##
                :11769
                          (Other):15103
Inferred model
tb <- tree.bayes(df, training = "contact")</pre>
graphviz.plot(tb)
```

## Loading required namespace: Rgraphviz

## Note: the specification for S3 class "AsIs" in package 'BiocGenerics' seems equivalent to one from p



```
emp_net <- model2network(paste(
    "[contact]",
    "[tcr_chain|contact]",
    "[tcr_region|contact]",
    "[pos_rel_antigen|pos_rel_tcr:tcr_chain:contact]",
    "[aa_antigen|aa_tcr:contact]",
    "[pos_rel_tcr|tcr_region:contact]",
    "[aa_tcr|contact]",
    sep =""))</pre>
graphviz.plot(emp_net)
```



```
BIC(fit, df)
## [1] -419457.2
library(pROC)
## Type 'citation("pROC")' for a citation.
##
## Attaching package: 'pROC'
## The following objects are masked from 'package:stats':
##
## cov, smooth, var
res <- predict(fit, node="contact", method="bayes-lw", data=df, prob=T)
p <- attributes(res)$prob
rocobj <- plot.roc(df[,"contact"], p[1,], ci=T)</pre>
```

```
Sensitivity

1.0 0.8 0.6 0.4 0.2 0.0

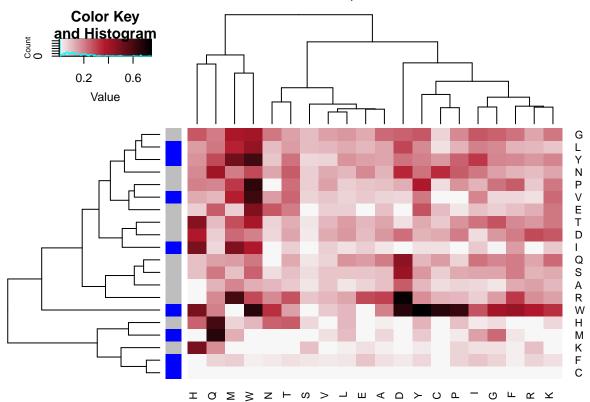
Specificity
```

```
rocobj
##
## Call:
## plot.roc.default(x = df[, "contact"], predictor = p[1, ], ci = T)
## Data: p[1, ] in 29819 controls (df[, "contact"] FALSE) > 4775 cases (df[, "contact"] TRUE).
## Area under the curve: 0.9105
## 95% CI: 0.9068-0.9142 (DeLong)
# df.cplx <- df %>% select(pdb_id, tcr_chain, contact)
# df.cplx$p <- p[1,]
# df.cplx <- df.cplx %>%
    group_by(pdb_id, tcr_chain) %>%
    summarise(contacts = sum(as.logical(contact)),
#
              contacts.pred = sum(p))
#
# ggplot(df.cplx, aes(contacts, contacts.pred)) +
  geom_point(shape=21) +
    geom_abline(slope = 1, intercept = 0) +
   scale_x\_continuous(limits=c(0,200)) +
    scale_y_continuous(limits=c(0,200)) +
    theme_bw()
get_prob <- function(var_name) {</pre>
  .df <- as.data.frame(fit[[var_name]]$prob)</pre>
  colnames(.df) <- gsub("Var1", "contact", colnames(.df))</pre>
  colnames(.df) <- gsub("Freq", paste("Freq", var_name, sep="."), colnames(.df))</pre>
  .df
```

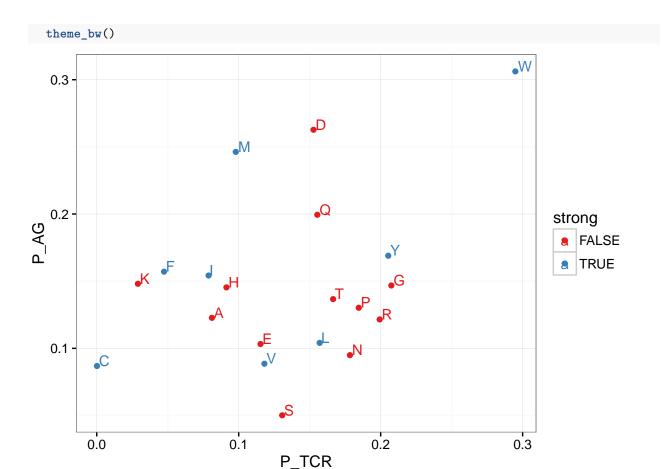
```
}
prob.tmp <- get_prob("contact")</pre>
for (var in colnames(df)[!(colnames(df) %in% c("contact", "pdb_id"))]) {
  prob.tmp <- merge(prob.tmp, get_prob(var))</pre>
prob.tmp$contact <- as.logical(prob.tmp$contact)</pre>
prob.tmp$P <- apply(prob.tmp[,which(grepl("Freq",colnames(prob.tmp)))], 1,</pre>
                 function(x) prod(x))
prob.aTaAC <- prob.tmp %>%
  group_by(aa_tcr, aa_antigen, contact) %>%
  summarise(P = sum(P)) %>%
  group_by(aa_tcr, aa_antigen) %>%
  summarise(P = P[which(contact)] / sum(P))
aa_pair_mat <- dcast(prob.aTaAC, aa_tcr ~ aa_antigen)</pre>
## Using P as value column: use value.var to override.
rownames(aa_pair_mat) <- aa_pair_mat$aa_tcr</pre>
aa pair mat$aa tcr <- NULL
aa_pair_mat <- as.matrix(aa_pair_mat)</pre>
aa_pair_mat[is.na(aa_pair_mat)] <- 0</pre>
strong_aa <- c("F", "I", "L", "M", "V", "W", "Y", "C")
strong_col <- ifelse(rownames(aa_pair_mat) %in% strong_aa, "blue", "grey")</pre>
names(strong_col) <- rownames(aa_pair_mat)</pre>
js_calc <- function(p, q) {</pre>
  p < -p/sum(p)
  q < -q/sum(q)
  m < -0.5 * (p + q)
  0.5 * (sum(p * log(p / m)) + sum(q * log(q / m)))
}
js_dist <- function(x) {</pre>
    mat <- x
    for(i in 1:nrow(mat)) {
        for(j in 1:nrow(mat)) {
            mat[i, j] <- js_calc(x[i, ], x[j, ])
    }}
    return(as.dist(mat))
}
heatmap.2(aa_pair_mat,
          hclustfun = function(x) hclust(x, method = "ward"),
           \#distfun = function(x) js_dist(x),
          RowSideColors = strong_col,
          #ColSideColors = strong_col,
          trace = "none",
          \#breaks = seq(-4, 0, length.out = 101),
```

```
col=colorpanel(100, "#f7f7f7", "#b2182b", "black"))
```

```
## The "ward" method has been renamed to "ward.D"; note new "ward.D2"
## The "ward" method has been renamed to "ward.D"; note new "ward.D2"
```



```
df.1 <- prob.tmp %>%
  group_by(aa_tcr, contact) %>%
  summarise(P = sum(P)) %>%
  group_by(aa_tcr) %>%
  summarise(P = P[which(contact)] / sum(P))
df.2 <- prob.tmp %>%
  group_by(aa_antigen, contact) %>%
  summarise(P = sum(P)) %>%
  group_by(aa_antigen) %>%
  summarise(P = P[which(contact)] / sum(P))
colnames(df.1) <- c("aa", "P_TCR")</pre>
colnames(df.2) <- c("aa", "P_AG")</pre>
df.1 <- merge(df.1, df.2)
df.1$strong <- ifelse(df.1$aa %in% strong_aa, T, F)</pre>
ggplot(df.1, aes(x=P_TCR, y=P_AG, color=strong)) +
  geom_point() +
  geom_text(aes(label=aa), vjust=0, hjust=-0.2) +
  scale color brewer(palette = "Set1") +
```



```
rf <- colorRampPalette(rev(brewer.pal(11, 'Spectral')))
r <- rf(32)

df.1 <- prob.tmp %>%
    group_by(pos_rel_antigen, pos_rel_tcr, tcr_chain, tcr_region, contact) %>%
    summarise(P = sum(P)) %>%
    group_by(pos_rel_antigen, pos_rel_tcr, tcr_chain, tcr_region) %>%
    summarise(P = P[which(contact)] / sum(P))

ggplot(df.1, aes(x=pos_rel_antigen, y = pos_rel_tcr, fill=P)) +
    geom_tile() +
    scale_fill_gradientn(colors=r) +
    facet_grid(tcr_region~tcr_chain)
```

