## Validação - Somente COVID-19

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#### Variáveis selecionadas

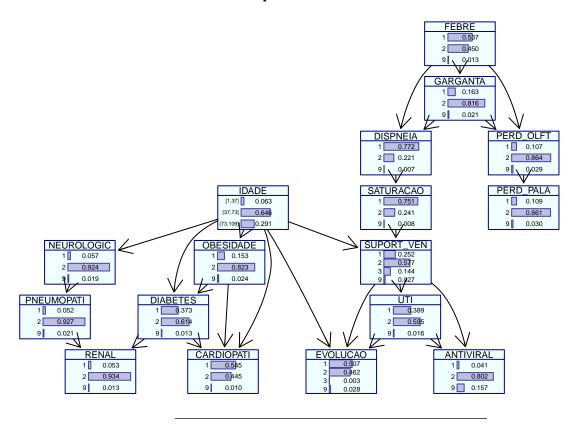
```
## Rows: 76,666
## Columns: 17
         <fct> "(37,73]", "(37,73]", "(73,109]", "(37,73]", "(73,109]", "(~
## $ IDADE
## $ FEBRE
         <fct> 2, 2, 2, 1, 2, 1, 1, 1, 1, 2, 1, 2, 1, 2, 2, 1, 1, 2, 2, 2, ~
## $ GARGANTA
         ## $ DISPNEIA
         <fct> 1, 2, 2, 1, 2, 2, 2, 2, 1, 1, 1, 1, 2, 1, 2, 1, 2, 1, 1, 2,~
## $ SATURACAO <fct> 1, 1, 2, 2, 2, 2, 1, 2, 2, 1, 1, 1, 1, 1, 2, 2, 2, 1, 2, 2,~
## $ EVOLUCAO
         <fct> 2, 1, 1, 1, 2, 1, 2, 2, 1, 2, 1, 2, 2, 2, 1, 1, 1, 1, 1, 1, 1, 1
## $ RENAL
         ## $ DIABETES
         <fct> 2, 1, 2, 1, 2, 2, 2, 2, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, ~
## $ OBESIDADE <fct> 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 1, 2, 2, 1, 2, 2, 9, 2,~
## $ UTI
         <fct> 1, 1, 2, 1, 2, 2, 1, 2, 2, 2, 2, 2, 1, 1, 2, 2, 1, 2, 2, 2,~
## $ CARDIOPATI <fct> 2, 2, 2, 1, 1, 2, 2, 1, 1, 1, 1, 1, 1, 1, 2, 2, 1, 1, 1, 1, -
## $ SUPORT_VEN <fct> 2, 2, 2, 9, 3, 3, 2, 2, 3, 2, 2, 3, 1, 1, 3, 3, 2, 3, 9, 3,~
```

#### Imposição de estrutura com arcos que fazem sentido clínico (White list)

## Rede Causal (DAG, Directed Acyclic Graphic)

```
#par(mfrow=c(2,2))
graphviz.plot(bn1, shape='rectangle', highlight = list(arcs = w1), main = '...')
                                                             FEBRE
                                                           GARGANTA
                                                   DISPNEIA
                                                                    PERD_OLFT
                                                                    PERD_PALA
                              IDADE
                                                  SATURACAO
                                                 SUPORT_VEN
NEUROLOGIC
                      OBESIDADE
                 DIABETES
 PNEUMOPATI
                                                      UTI
                                           EVOLUCAO
                                                           ANTIVIRAL
         RENAL
                          CARDIOPATI
fitted.1 <- bn.fit(bn1, s1)</pre>
#par(mfrow=c(2,2))
graphviz.chart(fitted.1, scale = c(2, 3), type = "barprob", col = "darkblue", bg = "azure", bar.col =
```

## Rede de probabilidades

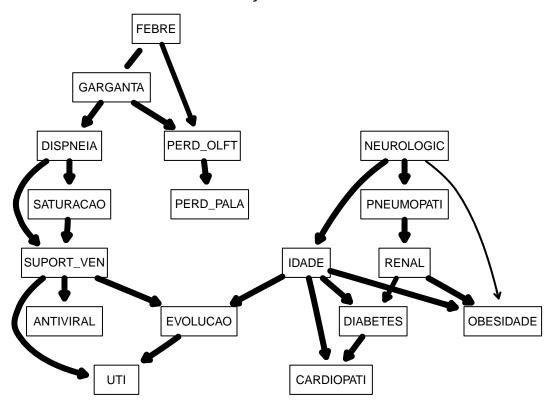


#### **Bootstrapping**

```
boots.trap <- 400
str.diff = suppressMessages(boot.strength(s1, R = boots.trap, algorithm = "mmhc"))
cat(paste('Threshold: ', attr(str.diff, "threshold")))

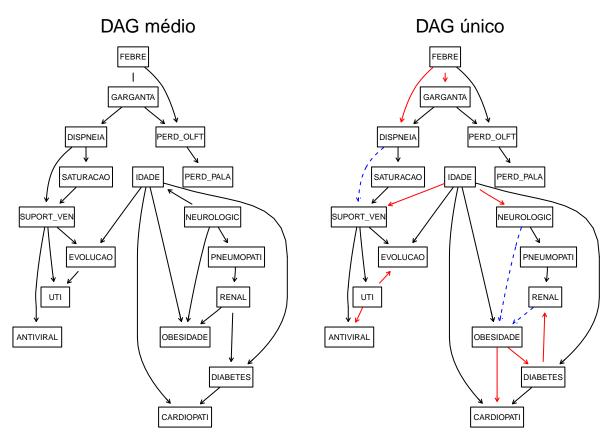
## Threshold: 0.5
avg.diff = averaged.network(str.diff)
strength.plot(avg.diff, str.diff, shape = "rectangle", main = paste("Iterações = ", boots.trap))</pre>
```

## Iterações = 400



How can we compare the averaged network (avg.diff) with the network we originally learned in from all the data? The most qualitative way is to plot the two networks side by side, with the nodes in the same positions, and highlight the arcs that appear in one network and not in the other, or that appear with different directions.

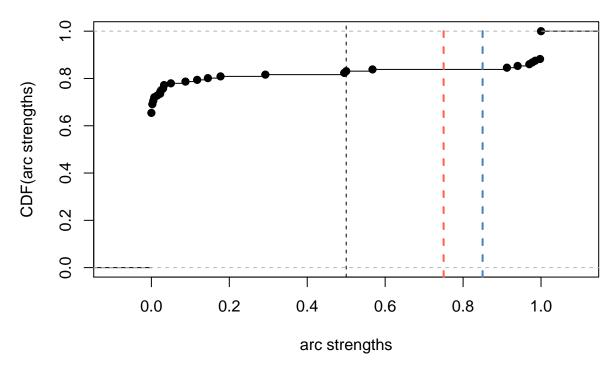
```
par(mfrow = c(1, 2))
graphviz.compare(avg.diff, bn1, shape = "rectangle", main = c("DAG médio", "DAG único"))
```



It is also a good idea to look at the threshold with respect to the distribution of the arc strengths

```
plot(str.diff)
abline(v = 0.75, col = "tomato", lty = 2, lwd = 2)
abline(v = 0.85, col = "steelblue", lty = 2, lwd = 2)
```

## threshold = 0.5



The simpler network we obtain by setting threshold = 0.8 in averaged.network() is shown below; it is certainly easier to reason with from a qualitative point of view.

# Iterações = 400 Thr = 0.5

# Iterações = 100 Thr = 0.75

