Viscosity: function-on-scalar regression

Sarah Brockhaus¹

¹ Institut für Statistik Ludwig-Maximilians-Universität München Ludwigstraße 33, D-80539 München, Germany sarah.brockhaus@stat.uni-muenchen.de

The results of this vignette together with more explanations can be found in Brockhaus et al. (2015).

1 Descriptive analysis

Load FDboost package and write useful functions for plotting. Load data and choose the time-interval.

```
> # load("viscosity.RData")
> data(viscosity)
> str(viscosity)
List of 7
 $ visAll : AsIs [1:64, 1:132] 41.5 25.2 63.7 35.6 17.8 12.3 38.6 22 18.2 36 ...
 $ timeAll: num [1:132] 11 13 15 17 19 21 23 25 27 29 ...
 $ T_C : Factor w/ 2 levels "low", "high": 1 1 2 2 2 2 1 1 1 1 ...
 $ T_A : Factor w/ 2 levels "low", "high": 1 1 1 1 1 1 1 1 1 1 ...
 $ T_B : Factor w/ 2 levels "low", "high": 1 1 1 1 1 1 1 2 2 ...
$ rspeed : Factor w/ 2 levels "low", "high": 1 2 1 2 1 2 2 1 2 1 ... $ mflow : Factor w/ 2 levels "low", "high": 2 1 1 2 1 2 1 2 2 1 ...
> ## set time-interval that should be modeled
> interval <- "509"
> ## model time until "interval"
> end <- which(viscosity$timeAll==as.numeric(interval))</pre>
> viscosity$vis <- log(viscosity$visAll[,1:end])</pre>
> viscosity$time <- viscosity$timeAll[1:end]
> ## set up interactions by hand
> vars <- c("T_C", "T_A", "T_B", "rspeed", "mflow")</pre>
> for(v in 1:length(vars)){
    for(w in v:length(vars))
    viscosity[[paste(vars[v], vars[w], sep="_")]] <- factor(</pre>
       (viscosity[[vars[v]]]:viscosity[[vars[w]]]=="high:high")*1)
+ }
```

```
> #str(viscosity)
> names(viscosity)
 [1] "visAll"
                     "timeAll"
                                                     "T_A"
 [5] "T_B"
                     "rspeed"
                                     "mflow"
                                                     "vis"
 [9] "time"
                     "T_C_T_C"
                                     "T_C_T_A"
                                                     "T_C_T_B"
[13] "T_C_rspeed"
                     "T_C_mflow"
                                     "T_A_T_A"
                                                     "T_A_T_B"
                                                     "T_B_rspeed"
[17] "T_A_rspeed"
                     "T\_A\_mflow"
                                     {\it "T\_B\_T\_B"}
[21] "T_B_mflow"
                     "rspeed_rspeed" "rspeed_mflow"
                                                     "mflow_mflow"
> pdf("vis.pdf")
> par(mfrow=c(1,1), mar=c(3, 3, 1, 2), cex=1.5)
> mycol <- gray(seq(0, 0.8, 1=4), alpha=0.8)[c(1,3,2,4)]
> int_T_CA <- with(viscosity, paste(T_C,"-", T_A, sep=""))</pre>
> with(viscosity, funplotLogscale(time, vis,
                                     col=getCol2(int_T_CA, cols=mycol[4:1])))
> legend("bottomright", fill=mycol,
         legend=c("T_C low, T_A low", "T_C low, T_A high",
                   "T_C high, T_A low", "T_C high, T_A high"))
> dev.off()
null device
```

Plot the data

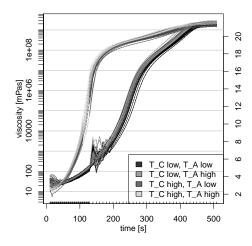


Figure 1: Viscostiy over time with temperature of tools (T_C) and temerature of resin (T_A) color coded.

2 Model with all main effects and interactions of first order

Fit model with all main effects and interactions.

```
> set.seed(1911)
> modAll <- FDboost(vis ~ 1</pre>
                    + bols(T_C) # main effects
                    + bols(T_A)
                    + bols(T_B)
                    + bols(rspeed)
                     + bols(mflow)
                     + bols(T_C_T_A) # interactions T_WZ
                    + bols(T_C_T_B)
                    + bols(T_C_rspeed)
                    + bols(T_C_mflow)
                    + bols(T_A_T_B) # interactions T_A
                    + bols(T_A_rspeed)
                    + bols(T_A_mflow)
                     + bols(T_B_rspeed) # interactions T_B
                     + bols(T_B_mflow)
                     + bols(rspeed_mflow), # interactions rspeed
                     timeformula=~bbs(time, lambda=100),
                    numInt="Riemann", family=QuantReg(),
                     offset=NULL, offset_control = o_control(k_min = 10),
                     data=viscosity, check0=FALSE,
                     control=boost_control(mstop = 100, nu = 0.2))
   Get optimal stopping iteration using bootstrap over curves.
> set.seed(1911)
> folds <- cv(weights=rep(1, modAll$ydim[1]), type="bootstrap", B=10)</pre>
> cvmAll <- suppressWarnings(validateFDboost(modAll, folds = folds,
                                      getCoefCV=FALSE,
                                     grid=seq(10, 500, by=10), mc.cores=10))
> mstop(cvmAll) # 180
> # modAll <- modAll[mstop(cvmAll)]</pre>
> # summary(modAll)
> # cvmAll
   Do model selection using stability selection.
> set.seed(1911)
> folds <- cvMa(ydim=modAll$ydim, weights=model.weights(modAll),</pre>
                type = "subsampling", B = 50)
> stabsel_parameters(q=5, PFER=2, p=16, sampling.type = "SS")
> sel1 <- stabsel(modAll, q=5, PFER=2, folds=folds, grid=1:100,
```

```
+ sampling.type="SS", mc.cores=10)
> sel1
> # selects effects T_C, T_A, T_C_T_A
```

The effects T_A , T_B and their interaction are selected into the model.

3 Model with selected effects

Estimate the model containing only the selected effects T_C , T_A , and their interaction.

```
> set.seed(1911)
> mod1 < - FDboost(vis ~ 1 + bols(T_C) + bols(T_A) + bols(T_C_T_A),
                   timeformula=~bbs(time, lambda=100),
                   numInt="Riemann", family=QuantReg(), checkO=FALSE,
                   offset=NULL, offset_control = o_control(k_min = 10),
                   data=viscosity, control=boost_control(mstop = 200, nu = 0.2))
> mod1 <- mod1[430]
  Find the optimal stopping iteration.
> set.seed(1911)
> folds <- cv(weights=rep(1, mod1$ydim[1]), type="bootstrap", B=10)</pre>
> cvm1 <- suppressWarnings(validateFDboost(mod1, folds = folds,</pre>
                                      getCoefCV=FALSE,
                                      grid=seq(10, 500, by=10), mc.cores=10))
> mstop(cvm1) # 430
> mod1 <- mod1[mstop(cvm1)]</pre>
> # summary(mod1)
```

Center all coefficient functions at each timepoint, yielding the following model:

```
\operatorname{median}\{\log(\operatorname{vis}_{i}(t))|x_{i}\} = \beta_{0}(t) + T_{Ai}\beta_{A}(t) + T_{Ci}\beta_{C}(t) + T_{ACi}\beta_{AC}(t),
```

where $vis_i(t)$ is the viscosity of observation i at time t, T_{Ai} and T_{Ci} are the temperatures of resin and of tools, respectively, each coded as -1 for the lower and 1 for the higher temperature. The interaction T_{ACi} is 1 if both temperatures are in the higher category and -1 otherwise.

```
> intercept <- intercept + colMeans(pred2)</pre>
> pred2 <- t(t(pred2)-intercept)</pre>
> ## effect of T_A
> pred3 <- predict(mod1, which=3, newdata=newdata)</pre>
> intercept <- intercept + colMeans(pred3)</pre>
> pred3 <- t(t(pred3)-colMeans(pred3))</pre>
> ## interaction effect T_C_T_A
> pred4 <- predict(mod1, which=4, newdata=newdata)</pre>
> intercept <- intercept + colMeans(pred4[3:4,])</pre>
> pred4 <- t(t(pred4)-colMeans(pred4[3:4,]))</pre>
> # offset+intercept
> smoothIntercept <- mod1$predictOffset(newdata$time) + intercept
  Plot the centered coefficient functions.
> pdf("visMod.pdf")
> par(mfrow=c(1,1), mar=c(3, 3, 1, 2), cex=1.5)
> mycol <- gray(seq(0, 0.5, 1=3), alpha=0.8)
> funplotLogscale(mod1$yind, pred2[3:4,], col=mycol[1], ylim=c(-0.5,6), lty=2, lwd=2)
> lines(mod1$yind, pred3[2,], col=mycol[2], lty=3, lwd=2)
> lines(mod1$yind, pred4[4,], col=mycol[3], lty=4, lwd=2)
> legend("topright", lty=2:4, lwd=2, col=mycol,
         legend=c("effect T_C high","effect T_A high","effect T_C, T_A high"))
> dev.off()
null device
```

References

Brockhaus S, Scheipl, F., Hothor, T., and Greven, S. (2015), The functional linear array model, *Statistical Modelling*, 15(3), 279-300.

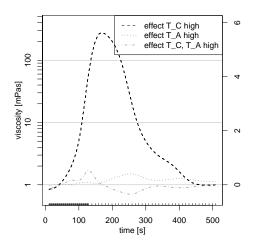


Figure 2: Viscosity over time and estimated coefficient functions. On the left hand side the viscosity measures are plotted over time with temperature of tools (T_C) and temperature of resin (T_A) color-coded. On the right hand side the coefficient functions are plotted.