lstr_dat2

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preparing data

```
zz <- read.csv("OxfordManRealizedVolatilityIndices.csv", stringsAsFactors = FALSE, skip = 3)</pre>
grep("[.]rv$",colnames(zz), value=TRUE)
## [1] "SPX2.rv"
                       "FTSE2.rv"
                                     "N2252.rv"
                                                    "GDAXI2.rv"
                                                                  "RUT2.rv"
## [6] "AORD2.rv"
                       "DJI2.rv"
                                     "IXIC2.rv"
                                                    "FCHI2.rv"
                                                                  "HSI2.rv"
                                     "SSMI.rv"
## [11] "KS11.rv"
                       "AEX.rv"
                                                    "IBEX2.rv"
                                                                  "NSEI.rv"
## [16] "MXX.rv"
                       "BVSP.rv"
                                     "GSPTSE.rv"
                                                    "STOXX50E.rv" "FTSTI.rv"
## [21] "FTSEMIB.rv"
rr <- zz %>% select(DateID, RUT2.rv) %>% mutate(rv = 100 * sqrt(252*RUT2.rv))
rv1 <- na.omit(rr$rv)
rv2 <- mls(rv1, 1:20, 1)
dat <- na.omit(cbind("y" = rv1, rv2))</pre>
```

lstr function

```
lstr_m <- function(y, x, z=NULL, weight, startx, method=c("Nelder-Mead", "BFGS")){</pre>
  d \leftarrow ncol(x)
  # weight function
  all_coef <- function(p) {</pre>
    weight(p,d)
  GG <- function(p){</pre>
  b3 \leftarrow p[4]; b4 \leftarrow p[5]
    xm <- x%*%(all_coef(p))
     (1 + \exp(-b3 * (xm - b4) / sd(x)))^{-1}
  # midasr
  lstr <- function(p){</pre>
    b0 \leftarrow p[1]; b1 \leftarrow p[2]; b2 \leftarrow p[3]
    xm <- x%*%(all_coef(p))
    mod \leftarrow b0 + b1*xm*(1 + b2 * GG(p))
    sum((y-mod)^2)
  }
  cc <- optimx(par=startx, lstr, method=method)</pre>
  ccmin <- which.min(cc$value)</pre>
  cc <- cc[ccmin, ]</pre>
  par <- as.numeric(cc[ ,1:length(startx)])</pre>
  fitt <- x%*%all_coef(par)</pre>
  ress <- y - x%*%all_coef(par)
```

```
out1 <- list("data" = cbind(y,x),</pre>
               "coef" = par,
               "mid_coef" = all_coef(par),
               "RSS" = cc$value,
               "fitted" = fitt,
               "residuals" = y-fitt,
               "mse" = mean((y-fitt) ^ 2))
  # Galvao
  lstr <- function(p){</pre>
   b0 <- p[1]; b1 <- p[2]; b2 <- p[3]
    xm <- x%*%(all_coef(p))
    mod \leftarrow b0 + b1*xm + (b2-b1)*GG(p)
    sum((y-mod)^2)
  }
  cc <- optimx(par=startx, lstr, method=method)</pre>
  ccmin <- which.min(cc$value)</pre>
  cc <- cc[ccmin, ]</pre>
  par <- as.numeric(cc[ ,1:length(startx)])</pre>
  fitt <- x\*\*all_coef(par)
  ress <- y - x\*\all_coef(par)
  out2 <- list("data" = cbind(y,x),</pre>
                "coef" = par,
                "mid_coef" = all_coef(par),
                "RSS" = cc$value,
                "fitted" = fitt,
                "residuals" = y-fitt,
                "mse" = mean((y-fitt)^2)
 list("midasr" = out1, "galvao" = out2)
}
mr \leftarrow lstr m(dat[,1], dat[,-1], weight = nbeta, startx = c(1,1,1,1,1))
mrs \leftarrow midas_r simple(dat[,1], dat[,-1], weight = nbeta, startx = c(1,1,1))
mr$midasr[-c(1,5:6)]
## $coef
## [1] 0.68753696 0.91262298 1.03012965 -0.01202779 1.31263117
##
## $mid_coef
## [1] 0.375429552 0.020787070 0.019531799 0.018817531 0.018314765
## [6] 0.017923832 0.017601211 0.017323896 0.017078111 0.016854782
## [11] 0.016647397 0.016450850 0.016260699 0.016072548 0.015881328
## [16] 0.015680101 0.015457364 0.015189005 0.014800977 0.005434139
##
## $RSS
## [1] 116776.3
##
## $mse
## [1] 48.83344
```

```
mr$galvao[-c(1,5:6)]
## $coef
## [1] 0.9499780 0.9734991 5.4808635 -4.7007380 3.4097913
##
## $mid_coef
## [1] 4.296564e-01 1.402685e-01 1.065986e-01 8.037271e-02 5.973168e-02
## [6] 4.358885e-02 3.112160e-02 2.165327e-02 1.461037e-02 9.502342e-03
## [11] 5.909966e-03 3.477608e-03 1.907327e-03 9.539475e-04 4.206019e-04
## [16] 1.544673e-04 4.248743e-05 6.894802e-06 3.083114e-07 1.193503e-71
##
## $RSS
## [1] 112526.9
##
## $mse
## [1] 25.7438
list("coef"=mrs$coefficients, "mid_coef"=mrs$midas_coefficients, "mse"=mean((mrs$model[,1]-mrs$model[,-
## $coef
## [1] 0.9840919 0.9691233 4.4716877
##
## $mid_coef
## [1] 4.260296e-01 1.270793e-01 1.020000e-01 8.161261e-02 6.465359e-02
## [6] 5.053327e-02 3.885057e-02 2.928513e-02 2.156088e-02 1.543066e-02
## [11] 1.066881e-02 7.067281e-03 4.433591e-03 2.589786e-03 1.372074e-03
## [16] 6.309736e-04 2.319516e-04 5.665662e-05 5.098026e-06 6.336356e-56
##
## $mse
## [1] 25.44749
mse
mr$midasr$mse
## [1] 48.83344
mr$galvao$mse
## [1] 25.7438
mean((mrs$model[,1]-mrs$model[,-1]%*%mrs$midas_coefficients)^2)
## [1] 25.44749
data
summary(mr$midasr$data[ ,1])
##
     Min. 1st Qu. Median
                              Mean 3rd Qu.
                                              Max.
    0.000 8.897 11.766 14.030 16.351 121.404
```

residuals

summary(mr\$midasr\$residuals) ## V1 ## Min. :-12.733 ## 1st Qu.: 1.325 ## Median : 3.281 ## Mean : 4.381 ## 3rd Qu.: 6.076 ## Max. : 85.969 summary(mr\$galvao\$residuals) ## V1 ## Min. :-33.3899 ## 1st Qu.: -1.8088 ## Median : 0.1253 ## Mean : 0.7006 ## 3rd Qu.: 2.5923 ## Max. : 68.0537 summary(mrs\$residuals) Min. 1st Qu. Median Mean 3rd Qu. Max. ## -33.7126 -2.2995 -0.3146 0.2217 2.1374 67.6913