## R Notebook

Load the test datasets

```
path = "TestDatasets/"
# Load all files
11 = lapply( list.files(path,".feather",full.names = T), function(x) read_feather(x) )
# Name them
names(ll) <- file_path_sans_ext( list.files(path, ".feather") )</pre>
11 <- 11[which(!str_detect(string = names(11),pattern = "TwoBreak"))]</pre>
res <- data.frame()
# Do the loopie
for( i in names(ll) ){
  # Define time series
 xx <- as.data.frame( ll[[i]] )</pre>
  # Define time
  tt <- seq.Date(as.Date(as.yearmon(1982.01)),by = "1 month",length.out = 360)
  zz = zoo(xx$x,order.by = as.yearmon(tt),frequency = 12)
  # Run bfast01 model for single breaks
  rdist = 12/nrow(xx) # calculate h relative to total sample length
  bf1 = try(bfast01(as.ts(zz),formula = response ~ trend, test = c("OLS-MOSUM", "BIC"),aggregate=any, ba
   bf2 = try(bfast01(as.ts(zz),formula = response ~ trend + harmon, test = c("OLS-MOSUM", "BIC"),aggreg
    1 = list(bf1,bf2)
    # Get best model
    bf <- try( l[[which.min(lapply(1,AIC))]], silent = T )</pre>
    # Extract largest breakpoint
    if(class(bf)=="try-error")
      # Check if trend only converged and use this instead
      if(class(bf1)!="try-error"){
        bf = bf1 # Reset to trend only
      } else { next() }
    }
    if(bf$breaks > 0){
    # Time of largest change
    la_year <- bf$data$time[bf$breakpoints]</pre>
    res <- rbind(res, data.frame(SSBS = i, year = la_year,w = which.min(lapply(1,AIC))))
        res <- rbind(res, data.frame(SSBS = i, year = 0, w = which.min(lapply(1,AIC))))
}
```

The Result. It seems that the seasonal amplitude is quite affected by chance / random noise res

```
## 1 monotonic_10
                      0.000 2
## 2 monotonic 11 1993.917 2
## 3 monotonic_12 1993.917 2
## 4
     monotonic_13 2006.917 2
## 5
     monotonic 14
                      0.000 2
     monotonic 15 2000.750 2
## 6
## 7
     monotonic_16 1994.917 2
## 8
     monotonic 17
                      0.000 2
## 9
     monotonic_18
                      0.000 2
## 10 monotonic_19
                      0.000 2
## 11 monotonic_1
                      0.000 2
## 12 monotonic_20 1989.917 2
## 13 monotonic_2
                      0.000 2
## 14 monotonic_3 1999.917 2
## 15 monotonic_4 2006.917 2
## 16 monotonic_5 2005.917 2
## 17 monotonic 6 1997.917 2
## 18 monotonic_7 2007.083 2
## 19 monotonic_8 2004.917 2
## 20 monotonic_9 1992.917 2
## 21 OneBreak_10 2004.500 2
## 22 OneBreak_11 1988.917 2
## 23
      OneBreak_12 1985.250 2
## 24 OneBreak_13 1985.250 2
## 25 OneBreak_14 1990.750 2
## 26 OneBreak_15 2008.583 2
## 27
      OneBreak_16 2002.417 2
## 28 OneBreak_17 1993.167 2
## 29 OneBreak_18 2000.167 2
## 30
      OneBreak_19 2002.083 2
## 31
       OneBreak_1 1985.500 2
## 32
      OneBreak_20 1989.833 2
## 33
       OneBreak_2 2000.417 2
## 34
        OneBreak_3 2007.833 2
## 35
        OneBreak_4 1985.250 2
## 36
        OneBreak 5 1985.250 2
## 37
        OneBreak_6 2008.583 2
## 38
        OneBreak_7 2000.917 2
## 39
        OneBreak_8 1985.250 2
## 40
        OneBreak 9 1991.750 2
Fit a seperate model, but consider trend changes only
res2 <- data.frame()
# Do the loopie
for( i in names(ll) ){
  # Define time series
  xx <- as.data.frame( ll[[i]] )</pre>
  # Define time
  tt <- seq.Date(as.Date(as.yearmon(1982.01)),by = "1 month",length.out = 360)
```

##

SSBS

year w

zz = zoo(xx\$x,order.by = as.yearmon(tt),frequency = 12)

# Run bfast01 model for single breaks

```
rdist = 12/nrow(xx) # calculate h relative to total sample length
bf = try(bfast01(as.ts(zz),formula = response ~ trend, test = c("OLS-MOSUM","BIC"),aggregate=any, ban
if(bf$breaks > 0){
    # Time of largest change
    la_year <- bf$data$time[bf$breakpoints]

res2 <- rbind(res2, data.frame(SSBS = i, year = la_year,w = 1) )
} else {
    res2 <- rbind(res2, data.frame(SSBS = i, year = 0, w = 1 ))
}</pre>
```

Apparently the seasonal change detection (harmonic series) overfits on random generated data

res2

```
##
             SSBS
                      year w
## 1
     monotonic 10
                      0.000 1
## 2
     monotonic_11
                     0.000 1
## 3
     monotonic_12
                     0.000 1
## 4
     monotonic_13
                     0.000 1
## 5
     monotonic_14
                     0.000 1
## 6
     monotonic_15
                     0.000 1
## 7
     monotonic_16
                     0.000 1
## 8
     monotonic_17
                     0.000 1
## 9 monotonic_18
                     0.000 1
## 10 monotonic_19
                     0.000 1
## 11 monotonic_1
                     0.000 1
## 12 monotonic 20
                     0.000 1
## 13 monotonic_2
                     0.000 1
## 14 monotonic 3
                     0.000 1
## 15 monotonic_4
                     0.000 1
## 16 monotonic_5
                     0.000 1
## 17 monotonic_6
                     0.000 1
## 18 monotonic_7
                     0.000 1
## 19 monotonic_8
                     0.000 1
## 20 monotonic_9
                     0.000 1
## 21 OneBreak_10
                     0.000 1
## 22 OneBreak_11 1988.917 1
## 23 OneBreak_12 1983.833 1
## 24 OneBreak_13
                      0.000 1
## 25 OneBreak_14 1990.750 1
## 26
      OneBreak_15
                     0.000 1
## 27
      OneBreak_16 2002.417 1
## 28 OneBreak_17 1993.167 1
## 29
      OneBreak 18 2000.167 1
## 30
      OneBreak_19 2002.083 1
## 31
       OneBreak_1 1985.500 1
## 32 OneBreak_20 1989.833 1
## 33
       OneBreak_2 2000.417 1
## 34
       OneBreak_3 2007.833 1
## 35
       OneBreak_4 1985.083 1
## 36
       OneBreak_5 1985.083 1
## 37
       OneBreak_6 2011.083 1
## 38
       OneBreak_7 2000.917 1
```

## 39 OneBreak\_8 1983.333 1
## 40 OneBreak\_9 1991.750 1