# Multivariate Change point Models for NO2

#### Introduction

```
library(InspectChangepoint)
library(hdbinseg)
library(changepoint.geo)

source("../R/univaraite_nsp.R")
source("../R/utils.R")
source("../R/pre_process_clean_data.R")
source("../R/multivaraite_nsp/__init__.R")
```

Load time series data for the 400 largest cities in Europe.

## Competing methods

Assuming a pieicewise constsant mean did not seem appropriate for the nuivariate time series. Nevertheless below I apply changepoint methods appropriate for detecting changes in pieicewise constant means. The Inspect alforithm is overly sensitive to noise and returns a very large number of changeoint locations.

```
other.methods <- fit.multivaraute.changepoints(t(xx$whitened))
par(mfrow = c(2,2))

for (ii in 1:length(other.methods))
{
    df.plot.with.dates.axis(xx$whitened, main = names(other.methods)[ii], col = "grey")
    for (jj in other.methods[[ii]]) abline(v = jj, col = "blue")
}</pre>
```

Of the three methods that return a sensible number of changepoints there is some disagreement between the changepoint locations. Most European countries began implementing nationwide lockdowns in March and we would therefore expect to detect a change some time after.

```
paste("geometric mapping:", paste(rownames(xx$whitened)[other.methods$`geometric mapping`], collapse =
paste("double cusum:", paste(rownames(xx$whitened)[other.methods$`double cusum`], collapse = ", "))
paste("sparsified bs:", paste(rownames(xx$whitened)[other.methods$`sparsified binary segmentation`], co
```

We can apply the same algorithms to first differenced data.

```
yy <- apply(xx$whitened, 2, diff)
other.methods.diff <- fit.multivaraute.changepoints(t(yy))

par(mfrow = c(2,2))

for (ii in 1:length(other.methods.diff))
{
    df.plot.with.dates.axis(yy, main = names(other.methods.diff)[ii], col = "grey")
    for (jj in other.methods.diff[[ii]]) abline(v = jj, col = "blue")
}</pre>
```

Again check for agreements in the detected changepoints... there is none!

```
paste("inspect:", paste(rownames(yy)[other.methods.diff$inspect], collapse = ", "))
paste("geometric mapping:", paste(rownames(yy)[other.methods.diff$`geometric mapping`], collapse = ", "
paste("double cusum:", paste(rownames(yy)[other.methods.diff$`double cusum`], collapse = ", "))
paste("sparsified bs:", paste(rownames(yy)[other.methods.diff$`sparsified binary segmentation`], collapse
```

## Narrowest Significance Pursuit

#### NSP with pieiceise constant means

```
m.nsp.pcwsConst <- panel_nsp(t(xx$whitened), thresh = thresh, deg = 0)

df.plot.with.dates.axis(xx$whitened, col = "grey", main = "NSP with piecewise constant means")
draw_mnsp_rects(m.nsp.pcwsConst$nsp.out, c(-100,100))

m.nsp.pcwsConst

Try isolating only the L2 aggregation
m.nsp.pcwsConst.L2 <- panel_nsp(t(xx$whitened), thresh = c(Inf, thresh[2]), deg = 0)

df.plot.with.dates.axis(xx$whitened, col = "grey", main = "(L2 aggregation only)")
draw_mnsp_rects(m.nsp.pcwsConst.L2$nsp.out, c(-100,100))

m.nsp.pcwsConst.L2</pre>
```

#### NSP with piecewise linear means

```
m.nsp.pcwsLin <- panel_nsp(t(xx$whitened), thresh = thresh, deg = 1)

df.plot.with.dates.axis(xx$whitened, col = "grey", main = "NSP with piecewise constant means")
draw_mnsp_rects(m.nsp.pcwsLin$nsp.out, c(-100,150))
m.nsp.pcwsLin</pre>
```

And again isolating only the L2 aggregations

```
m.nsp.pcwsLin.L2 <- panel_nsp(t(xx$whitened), thresh = c(Inf, thresh[2]), deg = 1)

df.plot.with.dates.axis(xx$whitened, col = "grey", main = "(L2 aggregation only)")
draw_mnsp_rects(m.nsp.pcwsLin.L2$nsp.out, c(-100,150))

m.nsp.pcwsLin.L2</pre>
```