

Detection of spatiotemporal changepoints in air quality – a generalised additive model approach

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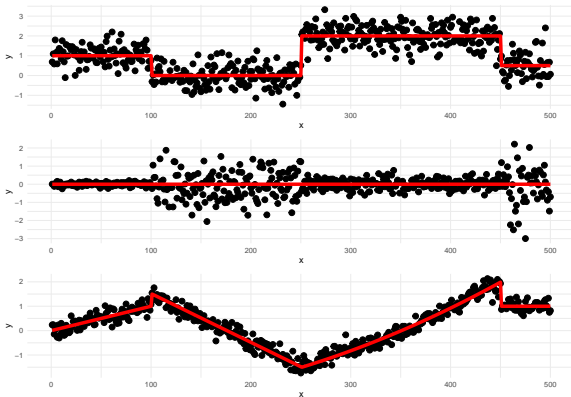
- What are changepoints?
- Fitting multiple changepoint models
- Spatiotemporal changes
- Application to Air Quality

What are changepoints?

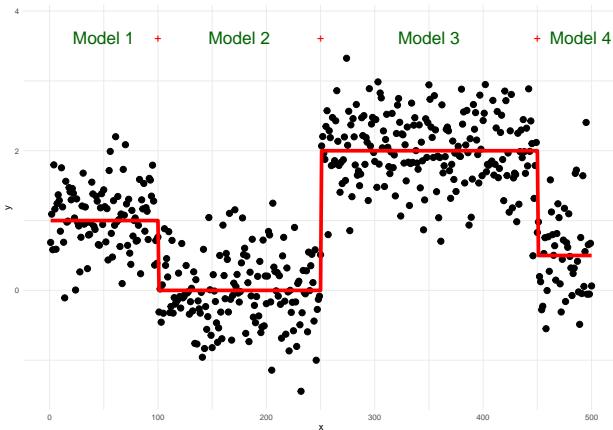


For data y_1, \dots, y_n , if a changepoint exists at τ , then y_1, \dots, y_τ differ from $y_{\tau+1}, \dots, y_n$ in some way.

There are many different types of change.



Problem

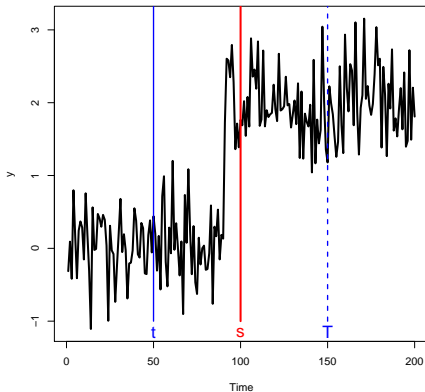


- How many changes?
- Where are the changes? 2^{n-1} possible solutions!

PELT in a nutshell



- Dynamic programming allows us to only worry about the location of the *last* change.
- Pruning means that as we go through the data we are smart about which locations are potential last change locations.



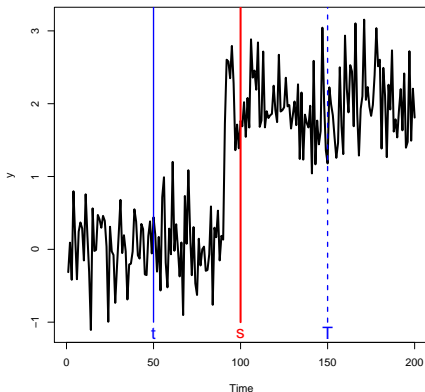


Let $0 < t < s < T$, if

$$F(t) + \mathcal{C}(y_{(t+1):s}) < F(s)$$

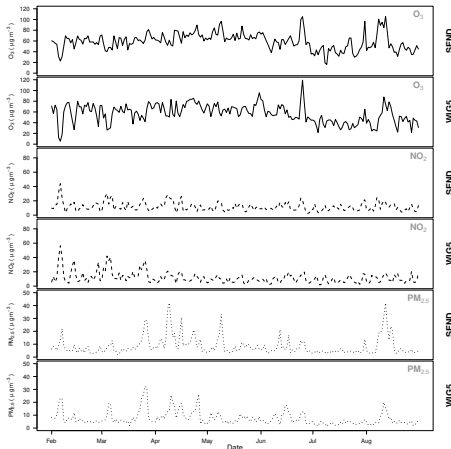
then at any future time $T > s$, t
can never be the optimal last
changepoint prior to T .

We can prove that, under certain
regularity conditions, the expected
computational complexity will be
 $\mathcal{O}(n)$.





UK Defra Air Quality Expert group requested scientific evidence around COVID lockdown effects on air quality across the UK.

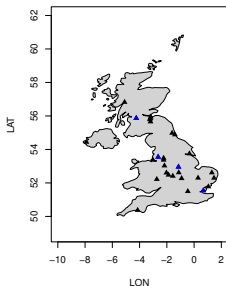




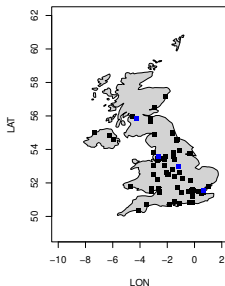
175 AURN active stations, 1st Feb - 31 Aug 2020 (213 days).

O_3 - 30 stations, NO_2 - 74 stations, $PM_{2.5}$ - 30 stations.

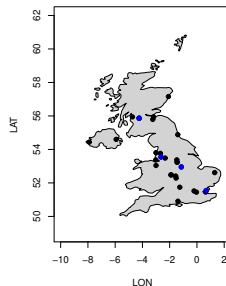
O_3 AURN Sites



NO_2 AURN Sites



$PM_{2.5}$ AURN Sites



OpenAir R package.

Generalised Additive Model

$$y_{s,t} = f_1(x_s) + f_2(x_t) + f_3(x_s, x_t) + \epsilon_{s,t}$$

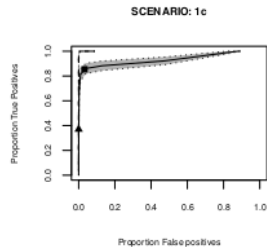
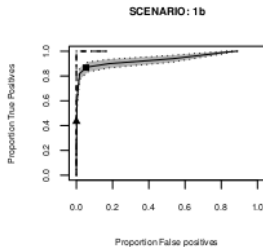
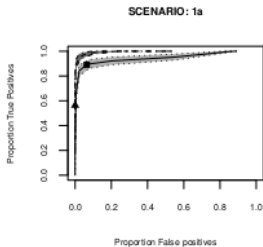
- f_1 is a 2D thin plate spline over space
- f_2 is a cubic regression spline over time
- f_3 is a tensor product for space-time interactions

Use `mgcv` in R or `GLMgam` in Python for GAM fit and likelihood.



Spatial: Constant, Random, Correlated

GAM-PELT: Dashed, Marginal: Solid

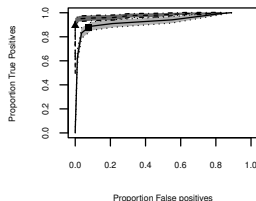




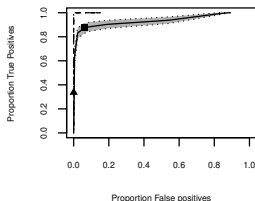
Spatial: Constant, Random, Correlated

GAM-PELT: Dashed, Marginal: Solid

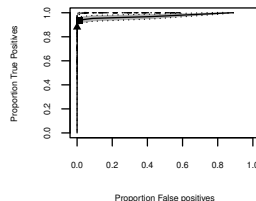
SCENARIO: 3a



SCENARIO: 3b



SCENARIO: 3c

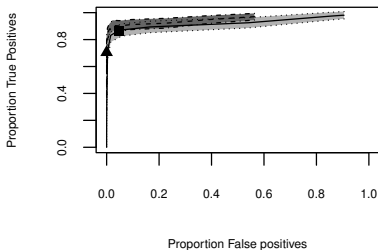




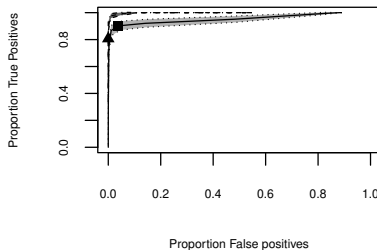
Random change in all. Left includes No change option.

GAM-PELT: Dashed, Marginal: Solid

SCENARIO: 4a



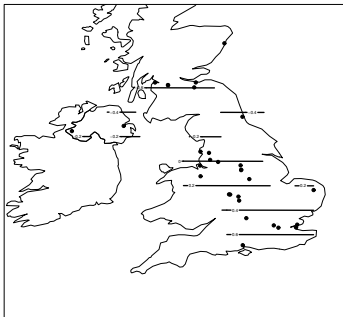
SCENARIO: 4b



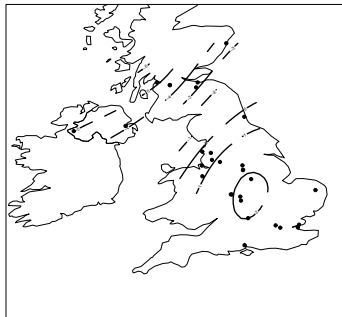


Changepoint at 21st March 2020

PM_{2.5} (pre-lockdown)



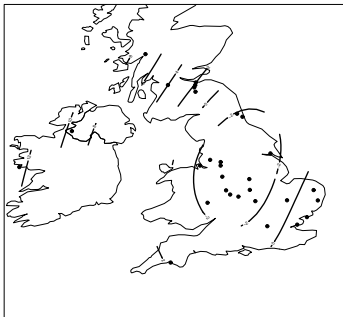
PM_{2.5} (lockdown)



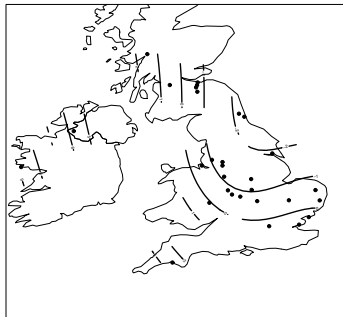


Changepoint at 26th March 2020

O_3 (pre-lockdown)



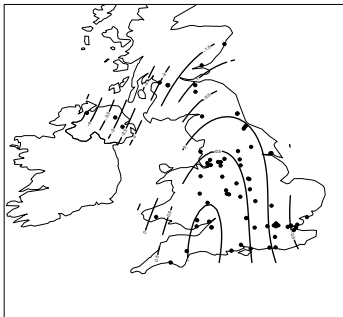
O_3 (lockdown)



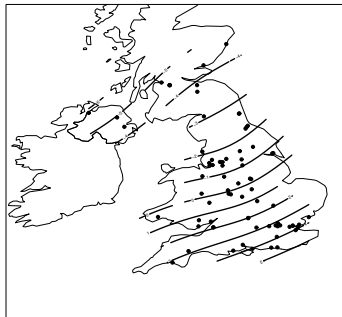


Changepoint at 27th March 2020

NO₂ (pre-lockdown)



NO₂ (lockdown)



- The GAM model is flexible enough to capture spatio-temporal patterns in air quality
- The addition of the PELT step for changepoint detection provides an easy-to-use extension to the model
- Early change for $PM_{2.5}$ correlates with early work from home in cities
- The decrease in O_3 and increase in NO_2 aligns with expectations from science around pollution dissipation
- Changepoints for phased return for schools in June are also seen.

- Introduced the PELT algorithm for identifying multiple changepoints optimally in a computationally efficient way
- Developed an approach to identify changepoints in spatiotemporal data with an application in air quality
- These and other extensions are useful for identify changes in a host of climatology applications