The Journal of Open Source Software

AQEval: R code for the analysis of discrete change in Air Quality time-series

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DOI: 10.xxxxxx/draft

Software

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Editor: Gabriele Bozzola

Reviewers:

- @eldemet
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Submitted: 02 June 2025
Published: unpublished

Summary

AQEval (Air Quality Evaluation) is an R package for the routine investigation of discrete changes in air quality time-series. The main functions use break-point/segmentation (BP/S) methods to detect, characterise and quantify change, while other functions build on these to provide a workflow to measure smaller changes and/or changes in more complex environments.

Statement of Need

Authorities responsible for air quality management are typically required to implement and evaluate air quality interventions they adopt (Bradley et al., 2019). These interventions are often costly, disruptive and unpopular (Glazener & Khreis, 2019), and the inherent variability monitoring data hinders impact assessments (Grange & Carslaw, 2019; Jones et al., 2012; Kelly et al., 2011; Pearce et al., 2011). Various methods have been developed to investigate discrete changes in a wide range of time-series (see e.g. Reeves et al., 2007; Truong et al., 2020) and several R (R Core Team & others, 2025) packages have been developed for their use, e.g. `bcp` (Erdman & Emerson, 2008), `changepoint` (Killick et al., 2016), `segmented` (Muggeo & others, 2008), and `strucchange` (Zeileis et al., 2002). Some have even been applied to air quality time-series (see e.g. Carslaw et al. (2006), Carslaw & Carslaw (2007)). However, many of those tasked with air quality policy assessment, although highly skilled in a wide range of monitoring activities, are unlike to be able to dedicate sufficient time and resources to the development of in-house expertise in such specialist analyses. `AQEval` was developed to address this skill gap. It aligns the inputs and outputs of a number of statistical methods to provide a one-package option for anyone interested in using R to routinely investigate change in air quality data. As many air quality professionals already use the R package `openair` (Carslaw & Ropkins, 2012; Ropkins & Carslaw, 2012) for more conventional analysis and data visualisation, `AQEval` has also been written using `openair` coding conventions to reduce the learning-curve typically associated with learning new software.

Sources

`AQEval` is freely available under General Public License (GPL):

- The latest (stable) release version of `AQEval` is on the Comprehensive R Archive Network (CRAN) https://CRAN.R-project.org/package=AQEval;
- The developers' version and code are publicly on GitHub https://github.com/karlropkins/AQEval, where issues or change requests can also be posted; and
- The project website is at https://karlropkins.github.io/AQEval/.

Analytical Rationale

The `AQEval` Break-Point/Segment (BP/S) methods involve three steps: finding possible 'points-of-change', testing these and quantifying 'regions-of-change' about the most likely:

1. Breaks-points are determined using the `strucchange` methods of Zeileis and colleagues (Zeileis et al., 2002, 2003). Here, a rolling-window approach is applied: a first subset (time-series window TW_0 in Figure 1a) is selected and linear regression modelled; the window advanced (TW_1 in Figure 1a) and a second model built, and so on through the time-series; then, likely points-of-change assigned by comparing the F-Stat scores of sequential models.
2. In addition to the standard Bayesian Information Criterion (BIC) testing used by `strucchange`, `AQEval` also checks all individual break-points are statistically valid ($p<0.05$), and down-scores less likely combinations.
3. Finally, the `segmented` methods of Muggeo and colleagues (Muggeo, 2003, 2017; Muggeo & others, 2008) are used to determine regions-of-change about break-points. Here, the confidence intervals for the selected break-points are used as start points, and segments assigned by random walk testing about these points as shown in Figure 1b.

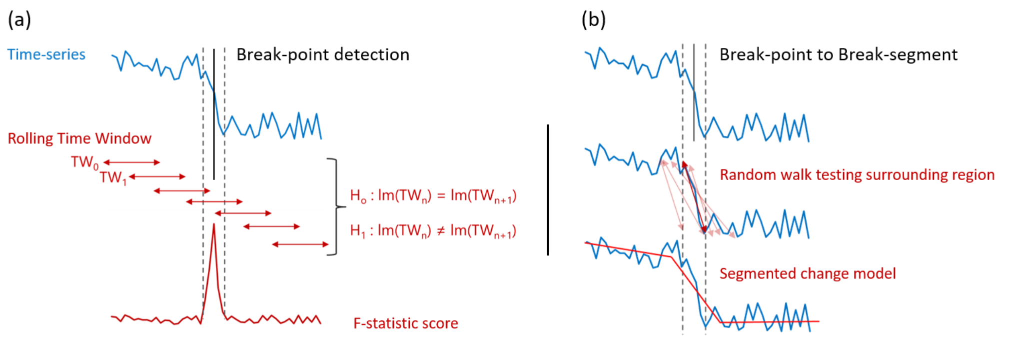

Figure 1: Break-point/segment scheme: (a) break-point, and (b) segment modelling about break-points.

Figure 2 shows the break-point/segment analysis of an NO_2 time-series from a roadside site where a change event (*ca.* 25 μg.m^{-3}; 31%) is detected between 2003-01-11 and 2003-02-19.

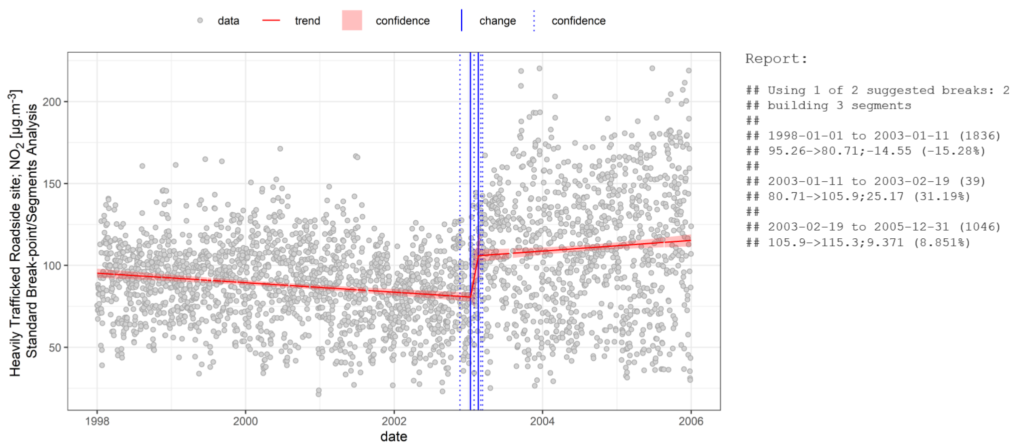

Figure 2: Standard AQEval break-point/segment analysis (graphical output and report) of an NO_2 1998-2005 time-series from a heavily trafficked roadside in the UK.

In some cases changes are small or local air inputs are complex, and time-series may require additional pre-processing to successfully isolate obscured change-events. For these, `AQEval` uses Generalized Additive Models (GAMs) (using `mgcv` methods, Wood, 2017, 2025) to subtract associated variance, by default:

$$[pollutant] = s_1(day\ of\ year) + s_2(hour\ of\ day) + te_1(wind\ speed, wind\ direction)$$

$$[pollutant]_{isolated} = ([pollutant] - [pollutant]_{predicted}) + mean(pollutant)$$

Where the investigated pollutant concentration, *[pollutant]*, is modelled as a function of *day of year*, *hour of day* and *wind speed* and *direction* using a combination of spline (s_1 and s_2) and tensor (te_1) fit-terms, and the unmodelled component, $[pollutant]_{isolated}$, is estimated as the mean-centred residual of this model.

Figure 3a shows the break-point analysis of NO_2 from a nearby but less heavily trafficked site where seasonality dominates the time-series, and Figure 3b shows the smaller (*ca.* 6.6 μg.m^{-3}; 13%) underlying change-event observed at a similar time to the large change observed at the more heavily trafficked site in Figure 2 using signal isolation and then break-point/segment analysis (2002-09-09 to 2002-12-21 compared with 2003-01-11 and 2003-02-19).

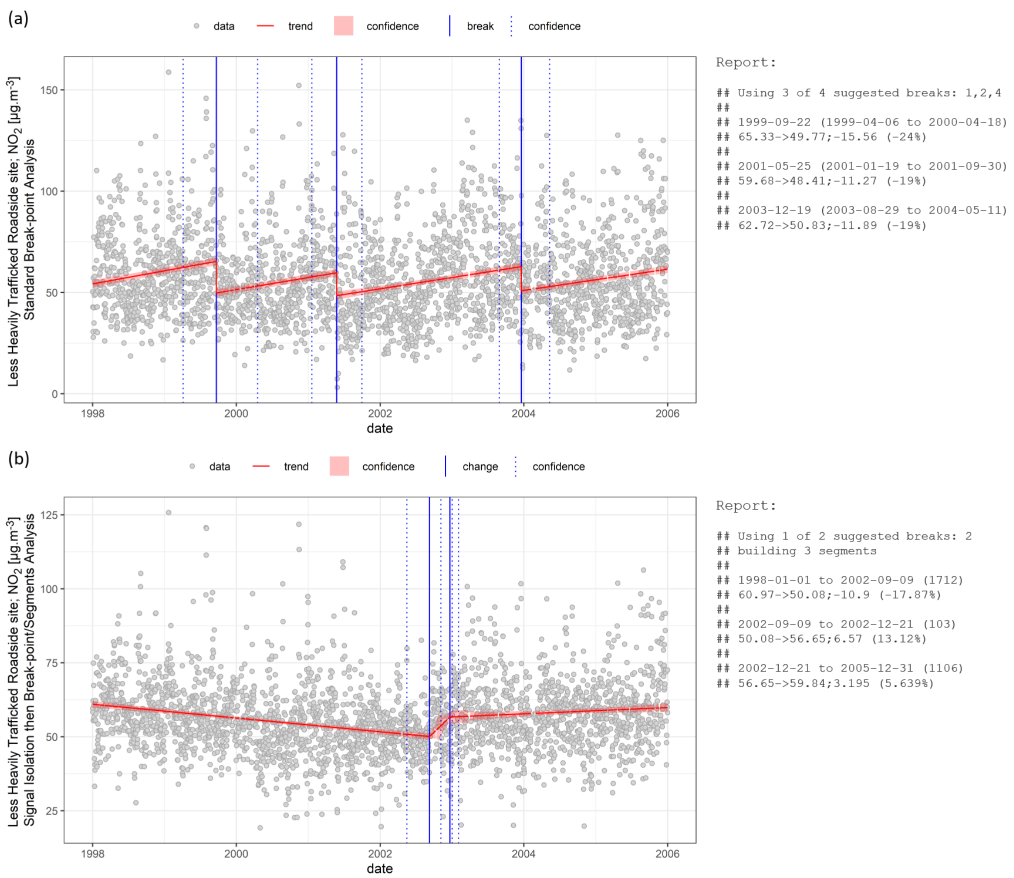

Figure 3: AQEval analysis of NO_2 1998-2005 time-series from a roadside site where: (a) standard (ambient air) break-point analysis exhibits a near-regular distribution of breaks typical of a site dominated by seasonal factors; and, (b) an underlying change-event is revealed using signal isolation and then break-point/segment analysis.

This default correction can also be modified to include other potential confounders, e.g. other frequency terms (e.g. day of week and/or week of year), background contributions (as local variance associated with trends at near-by site not affected by the investigated change), or proxies for other local contributors (e.g. other meteorological parameters like air temperature, markers for other sources, etc).

Related Outputs

The `AQEval` functions are described, along with worked examples of the code used to generate Figures 2 and 3, in the extended package introduction. Other work using `AQEval` include:

- Ropkins & Tate (2021), a peer-reviewed article on the multi-species AQEval analysis of air quality during the UK COVID-19 lockdown.
- Ropkins et al (2022), a peer-reviewed article on the use of `AQEval` to measure the NO_2 impact of bus fleet interventions.
- Also Clear Air Zone (CAZ) impact assessment reports include analyses using `AQEval`, see e.g.:
  - CAZ Baseline Study
  - First Year Report
  - (…)
  - Report archive

Acknowledgements

Initial `AQEval` development was funded by the UK Department for Environment, Food and Rural Affairs (Defra). The authors gratefully acknowledge contributions from colleagues at University of Leeds, Defra and IPSOS Mori, and internal review by the Defra/Department for Transport Joint Air Quality Unit (JAQU) Technical Independent Review Panel (T-IRP). The authors also gratefully acknowledge the work of the R core team and collaborators in developing and maintaining the open-source statistical language R and associated packages http://www.r-project.org/.

The views and opinions expressed herein by the authors are their own and do not necessarily reflect those of UK Government or any agency thereof.

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