## The open-source air pollution project openair

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#### The UK Air Quality Forecasting Seminar 16th July 2009

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## **Outline**

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## **Outline**

#### 1 [Introduction](#page-2-0)

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#### Opportunities and barriers Analysis of measurement and model output data

#### **Opportunities**

- $\blacksquare$  The analysis of air quality data can provide important insights into air pollution
- $\blacksquare$  There is a huge amount of data available
- Insightful analysis provides the evidence to support air quality management decisions

#### **Barriers**

- No consistent set of tools available to carry out analysis
- Tools can be spread across many different software applications
- **Many useful approaches are simply unavailable**
- Lack of time, money or ideas about what can be done

#### The openair project Summary of project

#### Key points

- 3-year NERC project to October 2011
- **Develop and make available open-source data analysis** tools to AQ community
- $\blacksquare$  Use **R** statistical software as the platform
	- $\blacksquare$  Highly capable software for "programming with data"
	- Develop a "package" of tools and progressively include advanced methods not widely available
	- $\blacksquare$  Transparency all code open to scrutiny

#### openair website Central resource for the project

- **Available at** [www.openair-project.org](http://www.openair-project.org)
- **popenair** package  $$ development version
- **All documentation**, data sets etc.
- **Mailing list and newsletter**
- Sister NERC project AirTrack at the University of Lancaster with complimentary aims
- Joint **openair**/AirTrack workshop, London, 1st October 2009



#### Data analysis How best to analyse data?

 $\blacksquare$  Data analysis is most useful when built around specific questions, however. . .

**Exploratory data analysis can be very insightful and is** under-used — but time consuming

John Tukey sums it up:

"The combination of some data and an aching desire for an answer does not ensure that a reasonable answer can be extracted from a given body of data."

## **Outline**

- 2 [Examples of openair functions](#page-7-0)
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#### Example analysis at a background site Thurrock — east of the M25

- Import a few years of data for a range of pollutants
- Examples of how some of the openair functions can be applied

#### Example

 $tk1 = import("d://mydata/thurrock.csv")$ 



### Quickly summarise data The summarise function

- **Always a good idea to look at** data first before doing anything more serious
- **The summarise function** provides a way to do this rapidly

#### Example

summarise(tk1)



### What do the met conditions look like? The wind.rose function

- **Plot a traditional wind rose** using the wind.rose function
- lots of options to control how the data are plotted

#### Example

wind.rose(tk1)





### Wind roses by year Plot by year, month, hour of the day...



$$
Example
$$
\n
$$
wind\text{.rose(tk1, type = "year")}
$$

#### How do concentrations vary in time? The time.variation function

- $\blacksquare$  The temporal variation in concentrations can provide important clues as to the source
- Sources can vary differently by hour of the day, day of the week and season
- **Enhanced with further information** 
	- **Traffic data**
	- **Meteorological data e.g. boundary layer height,** atmospheric stability
	- Excellent for model evaluation

#### Example

time.variation(tk1, pollutant  $=$  "nox")

#### How do concentrations vary in time? The time.variation function



#### How do concentrations vary in time? Two or more pollutants at once  $(SO<sub>2</sub>$  and  $NO<sub>X</sub>)$



Normalising the concentrations helps greatly when comparing different pollutants

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#### How do concentrations vary in time? Two or more pollutants at once  $(SO_2, NO_X \text{ and } O_3)$



Very different behaviours and underlying reasons for differences

### Polar plots and source identification Concentrations by wind speed and direction

Variation with wind speed and direction can help identify sources and source characteristics<sup>1</sup>

- Tall stack emissions vs. ground-level sources
- **Wind-blown sources e.g. particle** re-suspension
- Hot buoyant plumes e.g. aircraft jets
- **Local street canyon mixing**

#### Example

polar.plot(tk1, pollutant  $=$  "nox")



 $1$ Carslaw et al. (2006). Detecting and quantifying aircraft and other on-airport contributions to ambient nitrogen oxides in the vicinity of a large international airport. Atmos. Env., 40 (28), 5424-5434.

#### Polar plots and source identification Concentrations by wind speed and direction

- $\blacksquare$  The plot for  $SO<sub>2</sub>$  is markedly different to  $NO<sub>x</sub>$
- **Exidence of at least three sources** 
	- Can be shown to be a refinery, power station and industrial sources

#### Example

polar.plot(tk1, pollutant  $=$  "so2")



#### Temporal polar plots Concentrations by wind direction and time

#### **Plot as an annulus**

- Consider how concentrations vary by hour of the day, day of the week, season or trend by wind direction
- For  $NO<sub>x</sub>$  highest concentrations at night from north-west

#### Example

polar.annulus(tk1, pollutant  $=$  "nox")



#### Temporal polar plots Concentrations by wind direction and time

- $\blacksquare$  The SO<sub>2</sub> plot is markedly different to  $NO_{X}$
- Concentrations highest during daytime and from south-east

#### Example

polar.annulus(tk1, pollutant  $=$  "so2")

Gan choose type  $=$  "weekday", "season" and "trend"



#### **Trends** Trend analysis in openair

- $\blacksquare$  Trends are an important component of air quality analysis
- **Mann-Kendall** analysis often used for environmental time series
- Consider monthly trends with option to de-seasonalise the data
- **Use bootstrap simulation** techniques to estimate uncertainties and block bootstrap to deal with autocorrelation



#### Example

 $MannKendall(tk1, pollutant = "so2")$ 

#### **Trends** Can consider trends in many different ways



Trends can be 'conditioned' by many different variables-here by wind direction

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## **Outline**

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- <span id="page-22-0"></span>**3** [Outlook and concluding remarks](#page-22-0)

## **Developments**

- Reviewing scientific literature and will adopt promising approaches
	- Examples include, improved temporal characterisation e.g. Fourier analysis, change-point detection
	- $\blacksquare$  Better quantitative analysis
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- **3** The **openair** package
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## **Developments**

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- 2 Better support for model evaluation
	- Automate the process of evaluating models
	- Develop a range of metrics
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## **Developments**

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- **3** The **openair** package
	- Graphical-user interface (GUI)?
	- Remote repository with full version control and easier installation
	- **Develop documentation**
	- Reproducible analyses using Sweave,  $R$  and  $\angle$ ATEX

Thank you for you attention!

# Questions?

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