

Time series regression: advancements in this new tool for epidemiological analyses.

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LSHTM Feb 2012

Broad themes covered

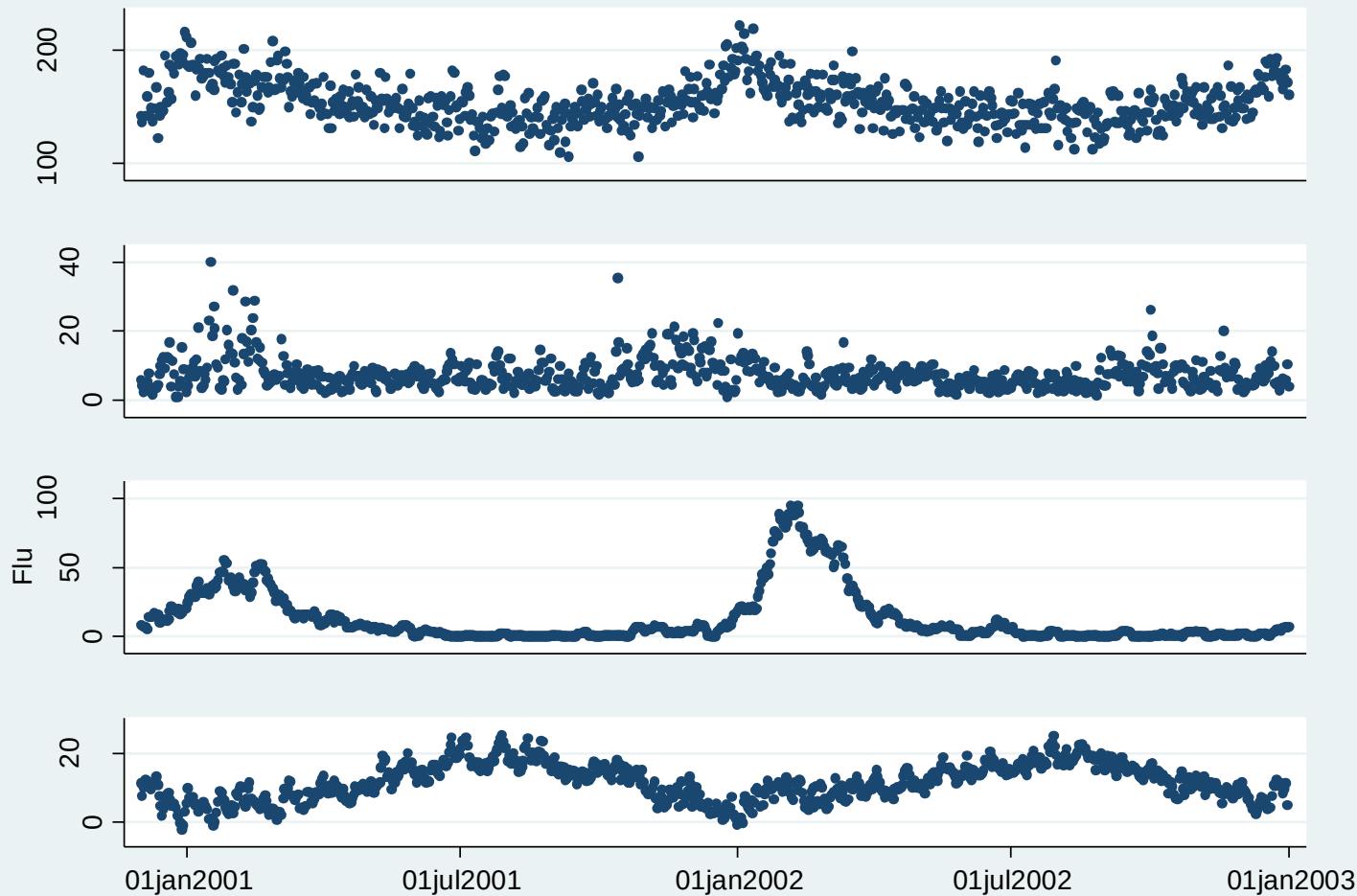
Part 1 (Ben)

1. Introducing time series regression
2. Introducing distributed lag non-linear models (DLNMs)

Part 2 (Antonio)

1. Exploring patterns in DLNMs (& complex patterns generally) across “cities”

Time series regression – the context



Time series regressions

What: (Poisson) regression with some time series elements

Why: Find acute effects of time-varying factors

Why not: Effect is chronic or exposure is time invariant

Statistical model needed to allow for:

- Measured time-varying risk factors (flu, pollution, ...)
- Overdispersion
- [Autocorrelation]
- **Unmeasured** smoothly varying risk factors ...
- **Lagged** effects (delay)

Model – general form

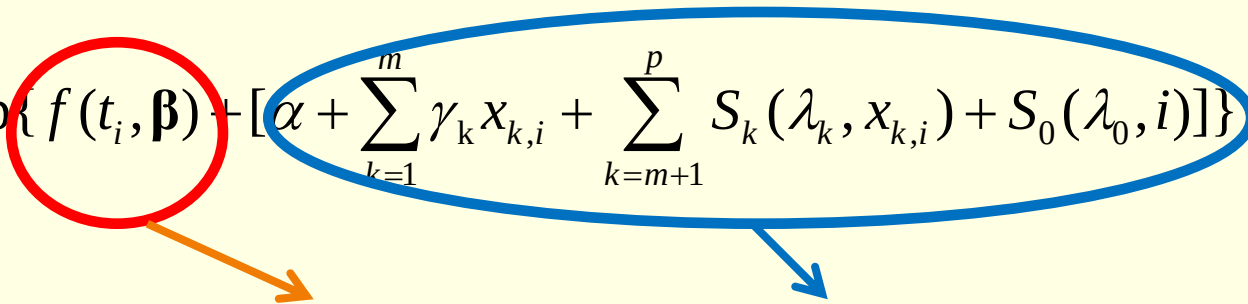
Quasi-Poisson regression model with:

$$E(Y_i) = \exp\{f(t_i, \boldsymbol{\beta}) + [\alpha + \sum_{k=1}^m \gamma_k x_{k,i} + \sum_{k=m+1}^p S_k(\lambda_k, x_{k,i}) + S_0(\lambda_0, i)]\}$$

t_i = temperature on day i

Y_i = mortality count on day i

Model – general form

$$E(Y_i) = \exp\{f(t_i, \boldsymbol{\beta}) + [\alpha + \sum_{k=1}^m \gamma_k x_{k,i} + \sum_{k=m+1}^p S_k(\lambda_k, x_{k,i}) + S_0(\lambda_0, i)]\}$$


or $E(Y_i) = \exp\{< \text{temperature effect} > + < \text{Other effects} > \}$

t_i = temperature on day i

Y_i = mortality count on day i

Model – general form

Measured Linear Measured non-linear Unmeasured smooth in time

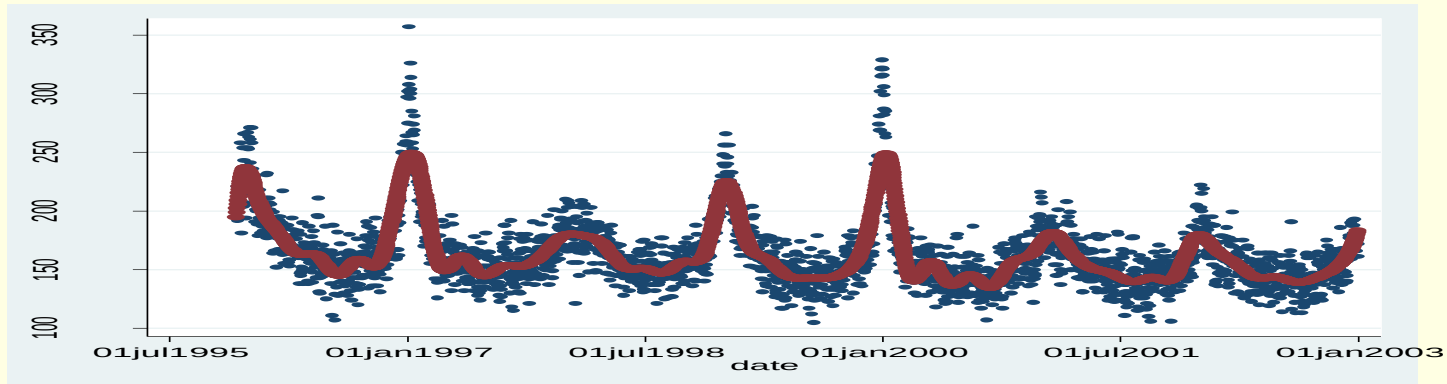
$$E(Y_i) = \exp\{f(t_i, \boldsymbol{\beta}) + [\alpha + \sum_{k=1}^m \gamma_k x_{k,i} + \sum_{k=m+1}^p S_k(\lambda_k, x_{k,i}) + S_0(\lambda_0, i)]\}$$

or $E(Y_i) = \exp\{< \text{temperature effect} > + < \text{Other effects} > \}$

t_i = temperature on day i

Y_i = mortality count on day i

... Control for smoothly varying risk factors

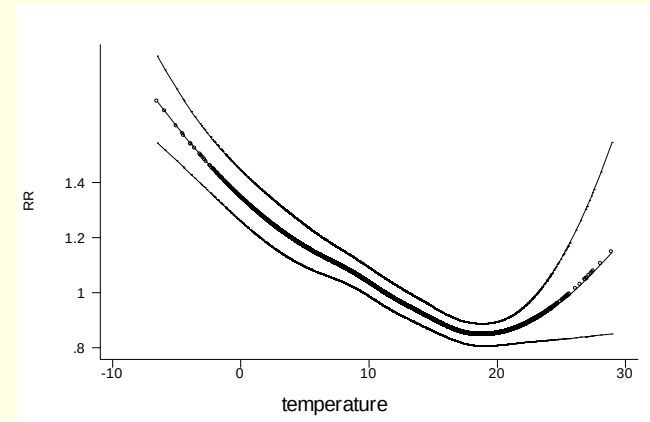


Options:

- Explicitly seasonal
 - Stratification: month (1-12) or week (1-52) indicators
 - Fourier (sine/cosine) functions
- General smooth functions of time
 - Stratification: month(1-72) or week (1-312)
 - Linear
 - Polynomial
 - **Cubic spline (illustrated above with 7 df/y)**
 - Non-parametric (GAM: LOWESS, Penalised spline ...)

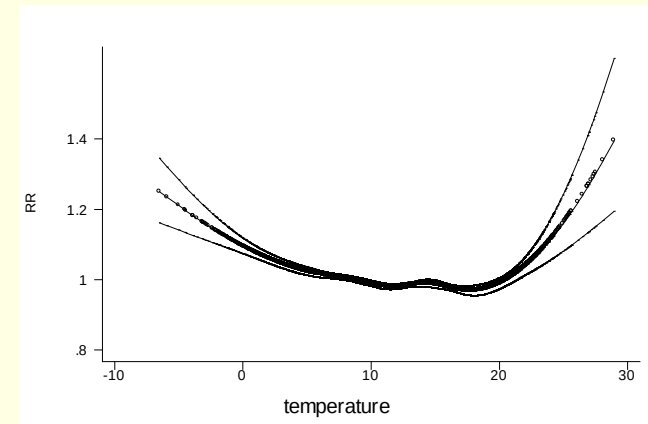
Impact of confounding

Unadjusted:



Adjusted:

- Time/season smooth curve ,
 - Pollution
 - Flu
 - Day of week
- [Air pollution benchmark model]



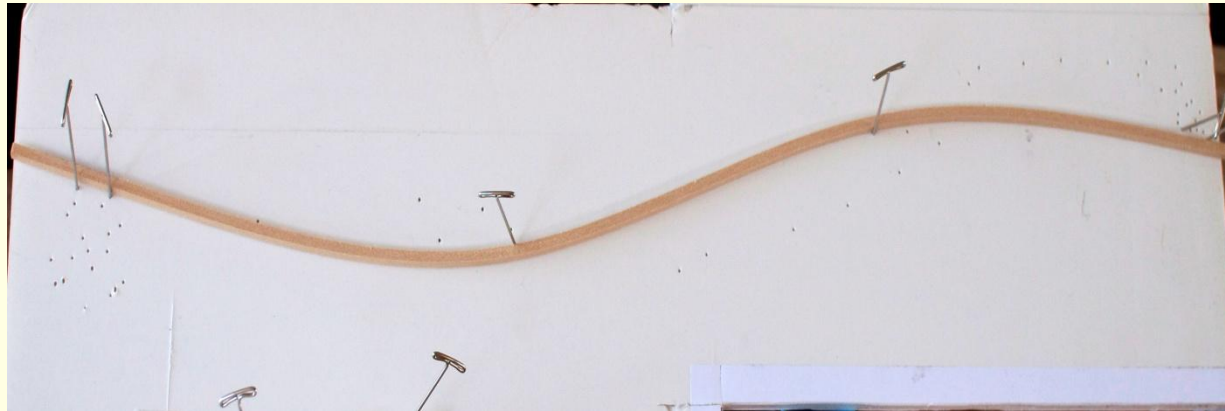
That's all on confounder control

Functional form of temperature –mortality association

$$(f(t_i, \beta))$$

Splines: eg natural cubic spline: NCS

1940's



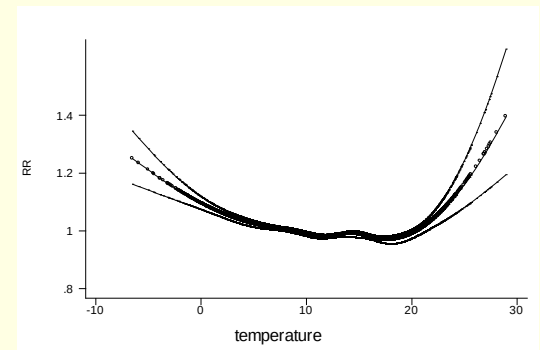
1990's

$$f(t_i, \boldsymbol{\beta}) = \sum_{b=1, k} \beta_b t_{b,i}$$

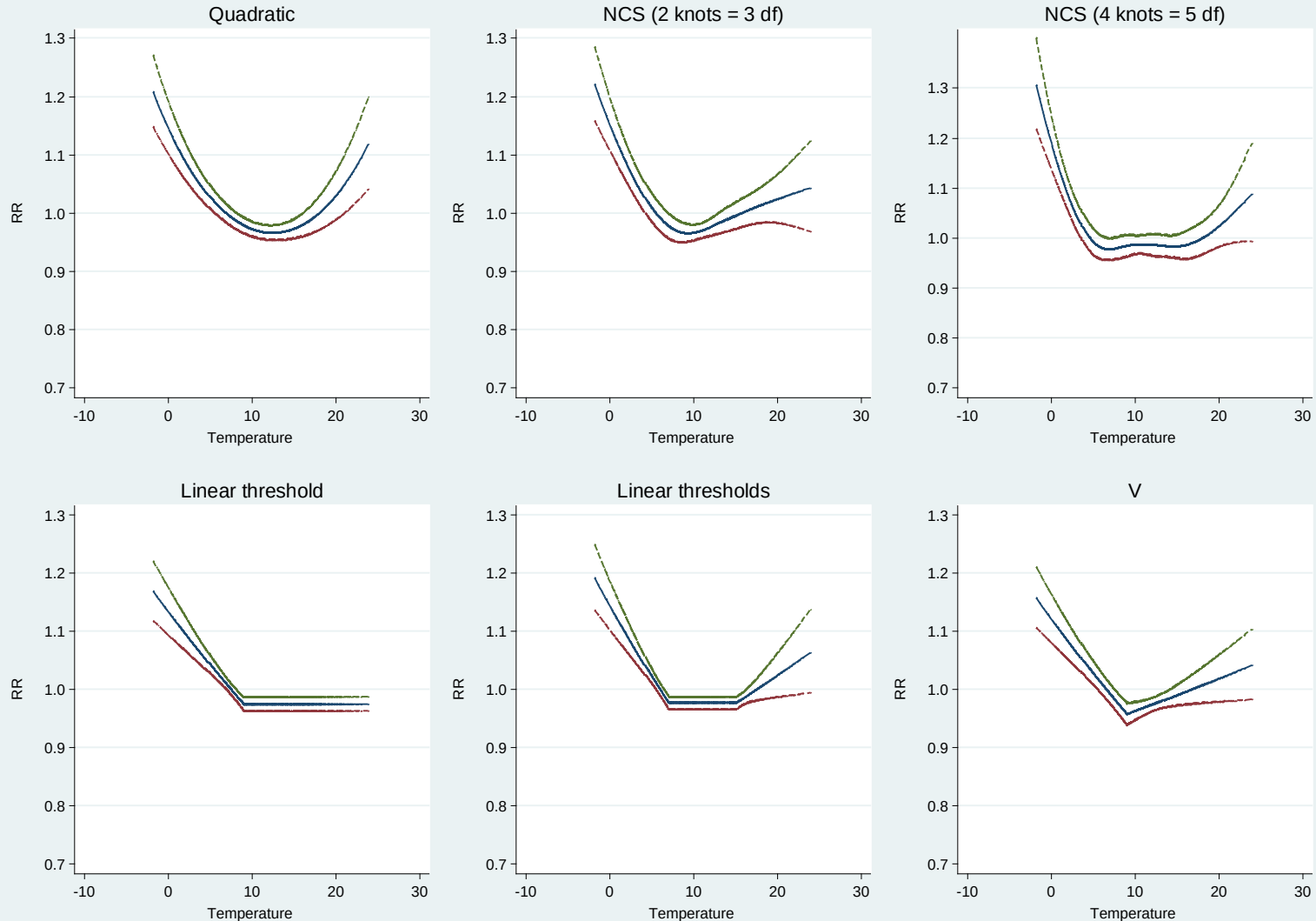
$$t_{b,i} \equiv [(t_i - \{knot\}_b)^+]^3$$

“Spline **basis**” variables

(OK geeks – not quite right – but close enough)



Var. of interest: Non-linear options



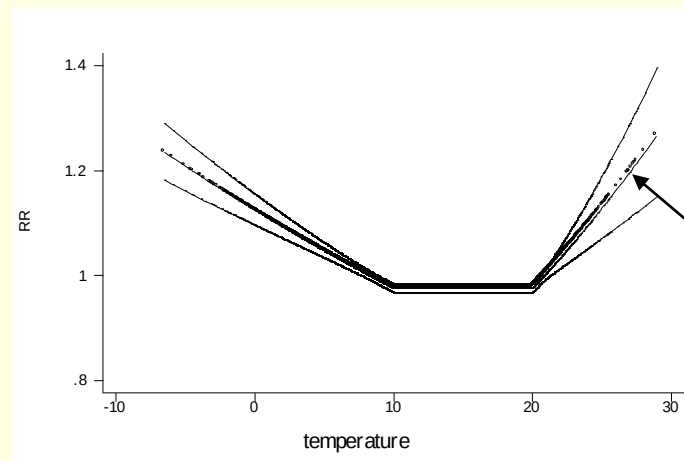
See additional practical for more detail

Lags

Lag = delay (in days) between
exposure (heat or cold) and death

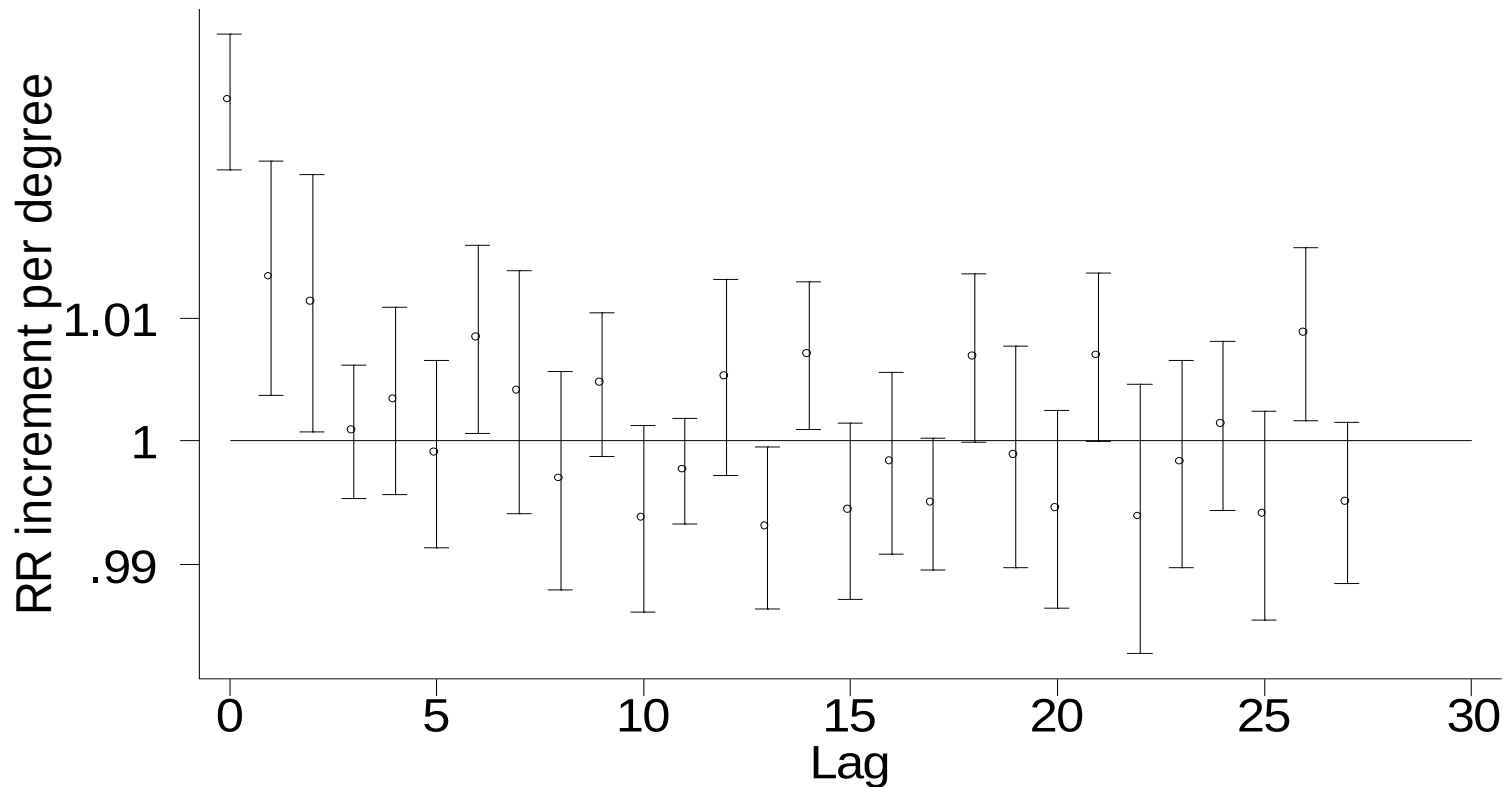
Exploring lagged effects 1:

- Use a linear threshold model
- Fit a **distributed lag model (DLM)** to the heat slope



Heat effects:
Different slope
for each lag

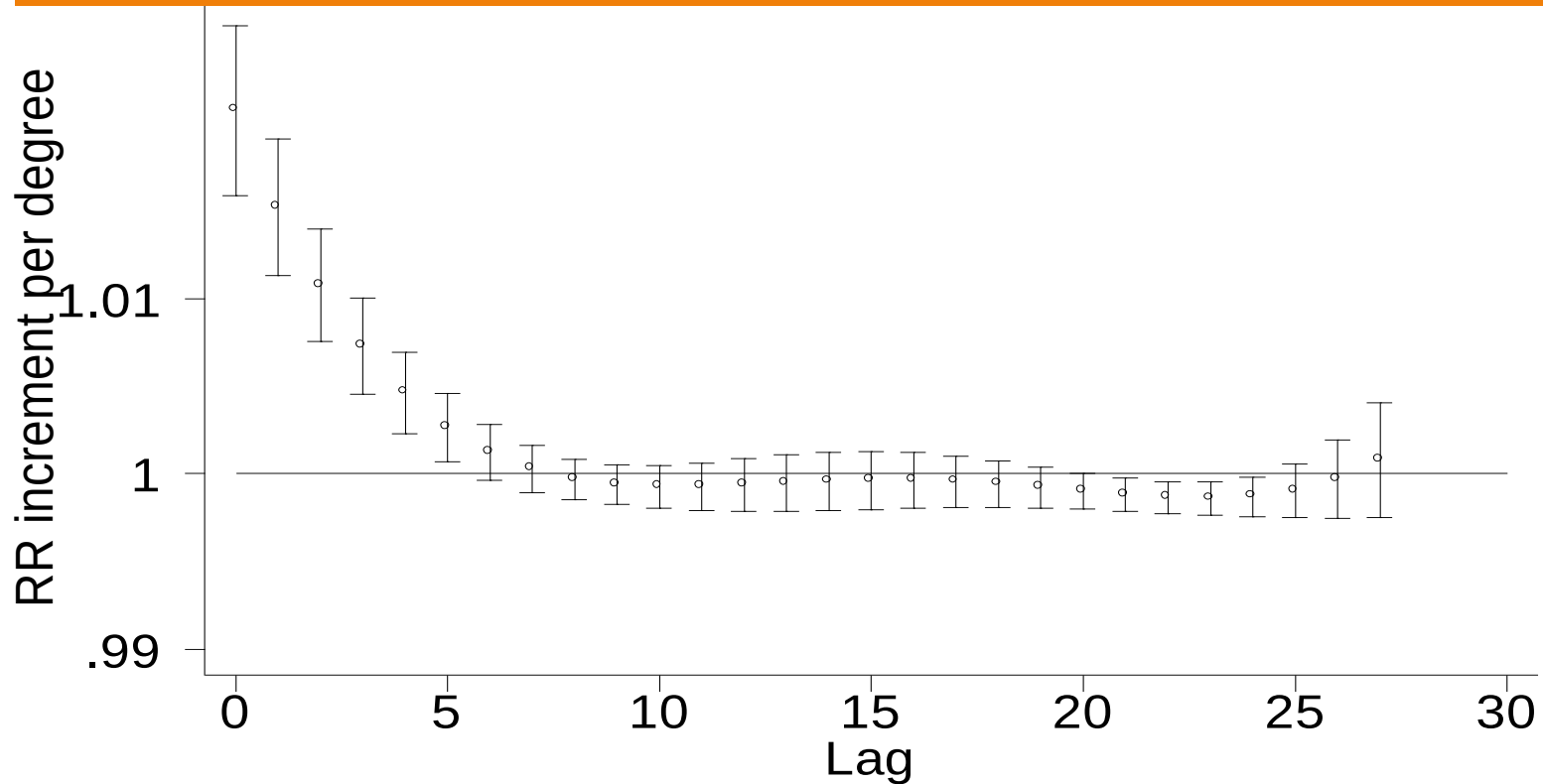
Distributed lags with linear effects – heat (degrees above 20°C)



$$DLM(heat): \sum_{l=0}^{27} \gamma_l t_{heat,i-l}$$

London CVD 1976-1993

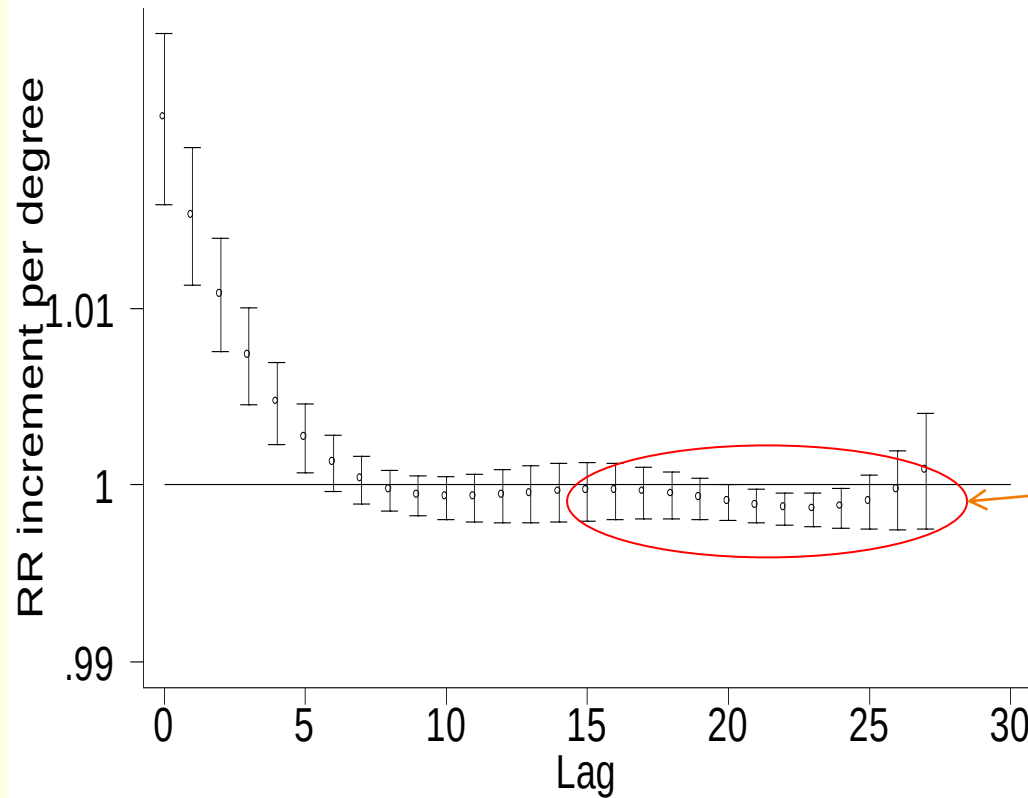
Constrained Distributed lags (degrees above 20°C) Quartic polynomial [Schwartz 2001]



$$DLM(heat): \sum_{p=1}^P \gamma_p \overline{t_{heat,p,i}}, \text{ where } \overline{t_{heat,p,i}} = \sum_{l=0}^L l_p t_{heat,i-l}$$

Lag basis variable

Distributed lag models and “harvesting”



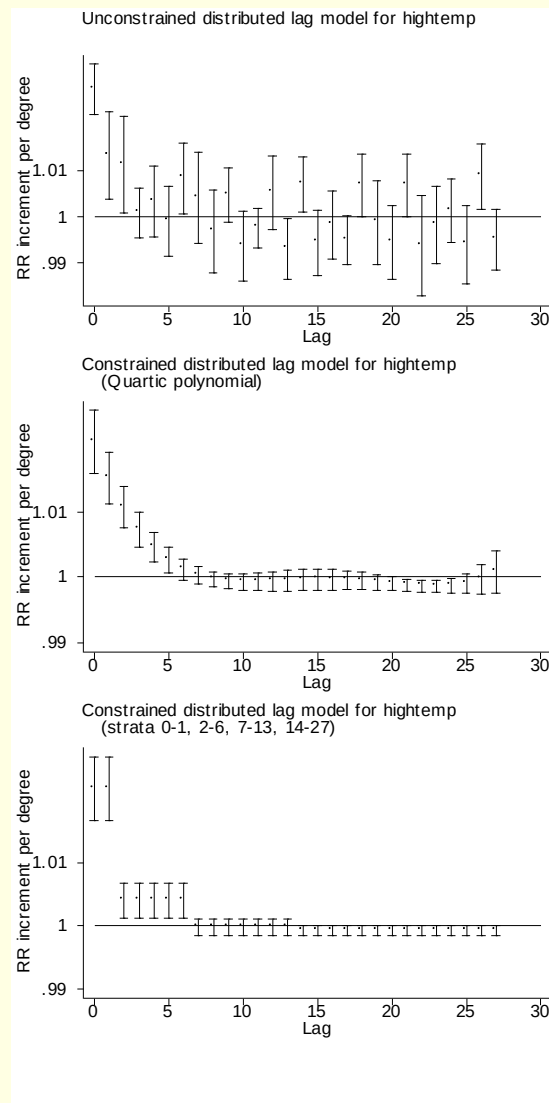
Deficit –
“harvesting”?



A summary of net effect from distributed lag models

Net RR increment (%)
over 28 days
per degree above 20
(95% CI):

5.0(3.7,6.3)



4.9(3.6,6.2)

4.9(3.7,6.2)

Distributed lag non-linear model

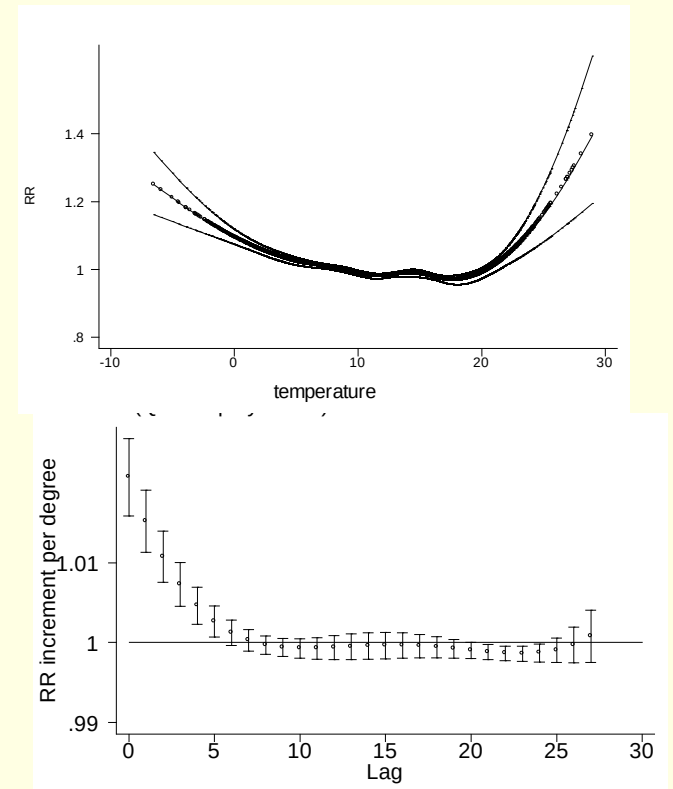
$f(t_i, \beta)$ tweaked again

To capture curved associations with non-trivial lag dependence:

Need combined virtues of:

(a) Smooth temp-mort curves

(b) Smooth effects of lag.



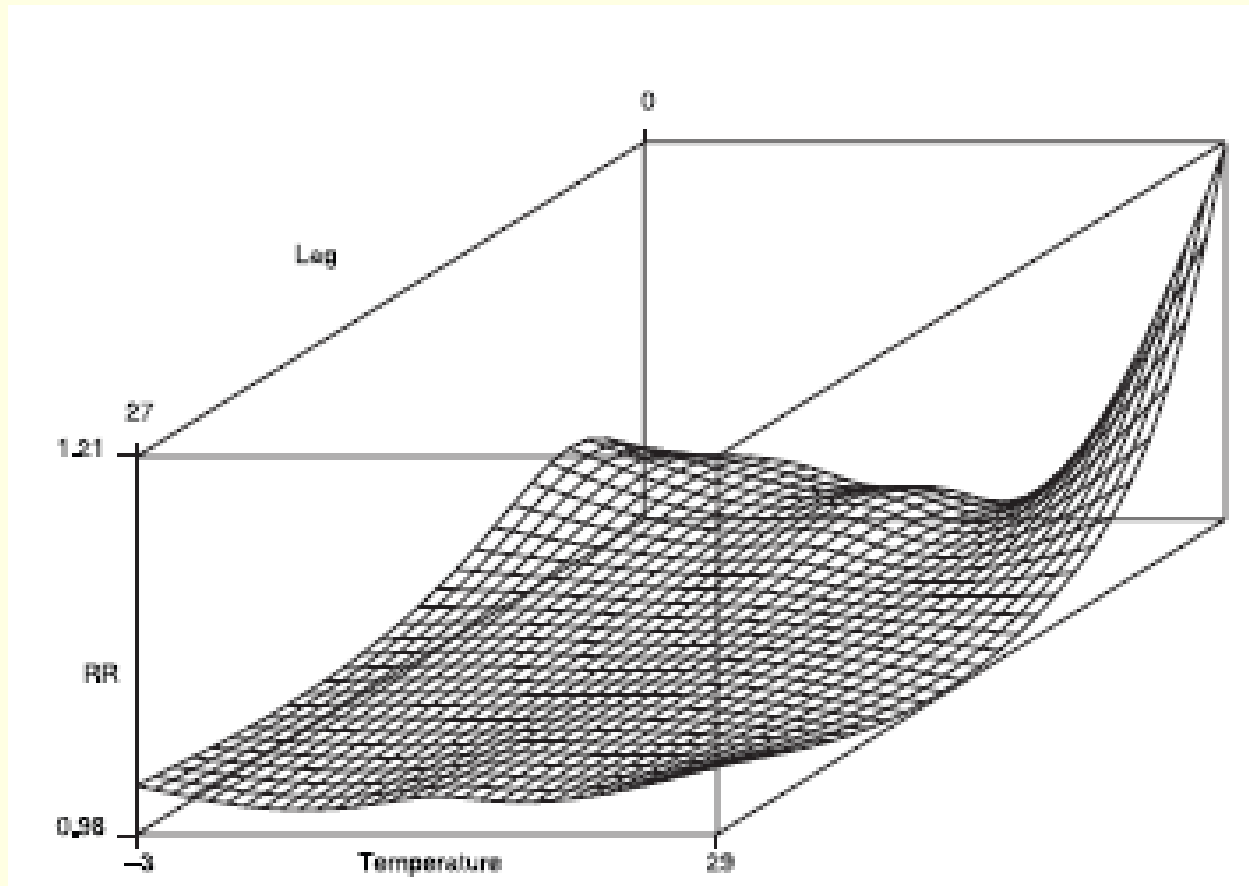
Distributed lag non-linear models (**DLNM**)

Unrestricted choice of:

Shape of temperature-mortality graph	X	Shape of change with lag
• Step changes • Thresholds • Cubic splines • ...		• Step changes • Polynomial • Cubic splines • ...

Implemented in R package `dlm` AG (and `clutzy` stata code - BA)

RR by temperature (NCS) and lag(NCS)



Temperature-mortality models: recap

$$f(t_i, \boldsymbol{\beta}) =$$

$$\text{Thresholds: } \beta_{\text{cold}} t_{\text{cold},i} + \beta_{\text{heat}} t_{\text{heat},i}$$

$$\text{Spline}(\text{gen}): \sum_{b=1}^B \beta_b t_{b,i}$$

$$\text{DLM}(\text{heat}): \sum_{p=1}^P \gamma_p \overline{t_{\text{heat},p,i}}, \text{ where } \overline{t_{\text{heat},p,i}} = \sum_{l=0}^L l_p t_{\text{heat},i-l}$$

Temperature-mortality models: +1

$$f(t_i, \beta) =$$

$$\text{Thresholds} : \beta_{\text{cold}} t_{\text{cold},i} + \beta_{\text{heat}} t_{\text{heat},i}$$

$$\text{Spline}(\text{gen}) : \sum_{b=1}^B \beta_b \mathbf{t}_{b,i}$$

Temperature **basis**
(B df)

+

$$\text{DLM}(\text{heat}) : \sum_{p=1}^P \gamma_p \overline{\mathbf{t}_{\text{heat},p,i}}, \text{ where } \overline{t_{\text{heat},p,i}} = \sum_{l=0}^L l_p t_{\text{heat},i-l}$$

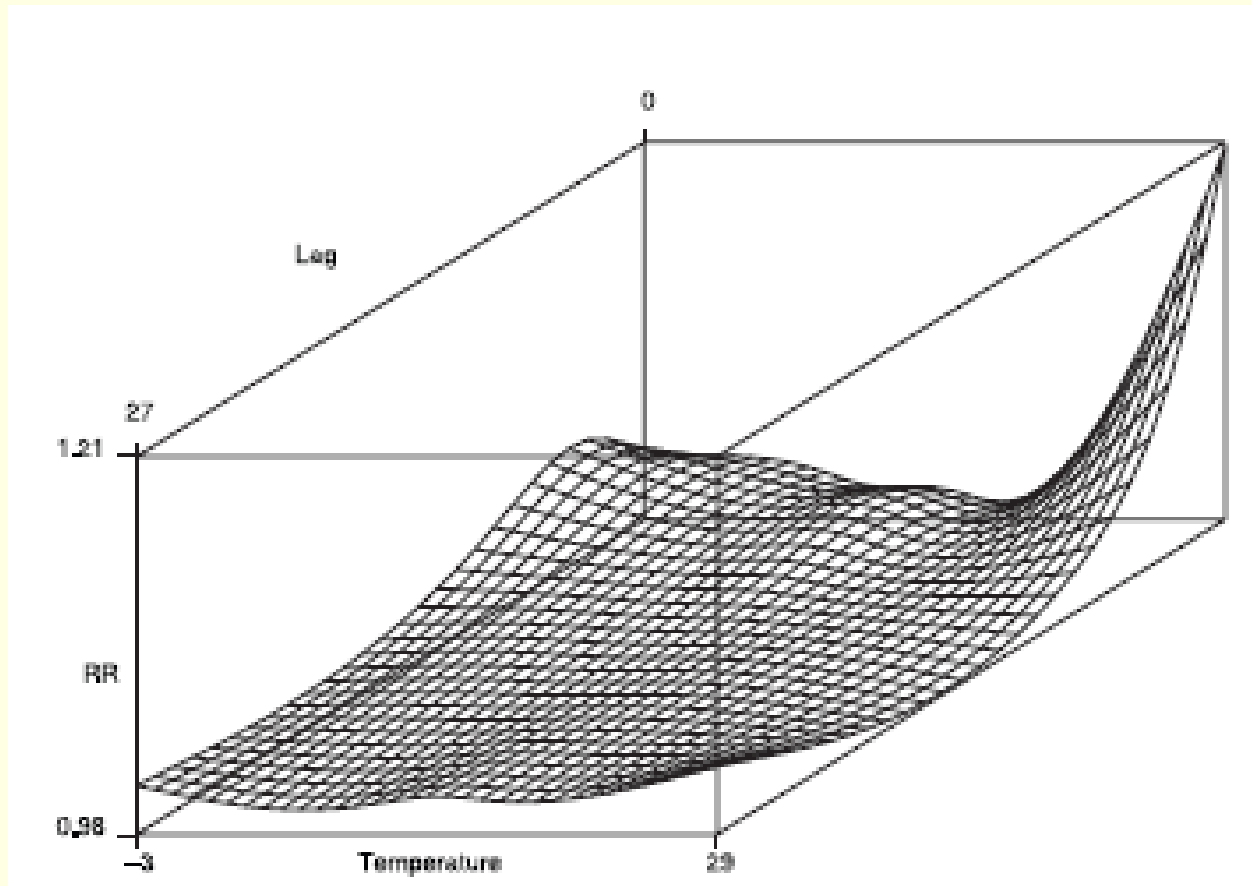
Lag **basis**
(P df)

=

$$\text{DLNM}(\text{gen}) : \sum_{b=1}^B \sum_{p=1}^P \eta_{b,p} \overline{\overline{\mathbf{t}_{b,p,i}}}, \text{ where } \overline{\overline{t_{b,p,i}}} = \sum_{l=0}^L l_p t_{b,i-l}$$

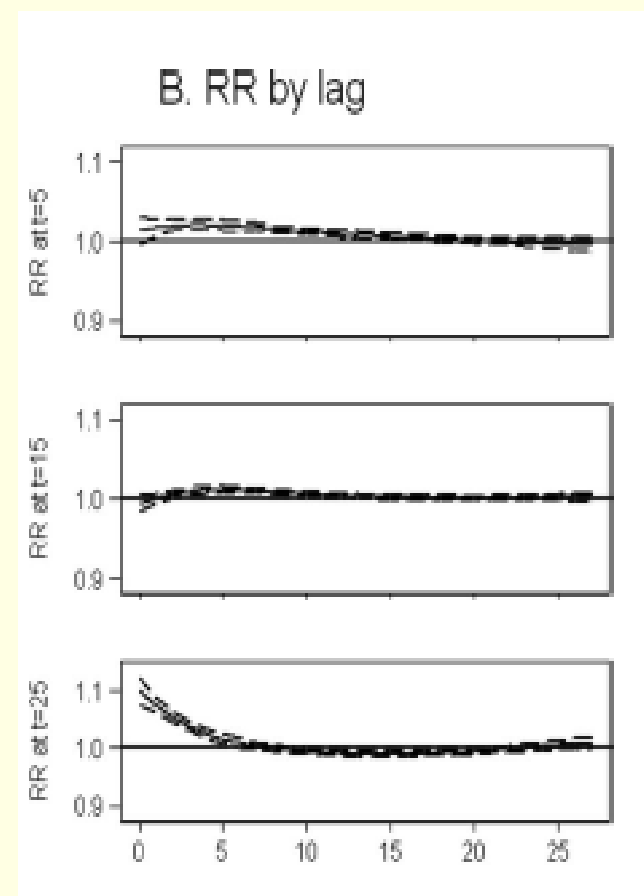
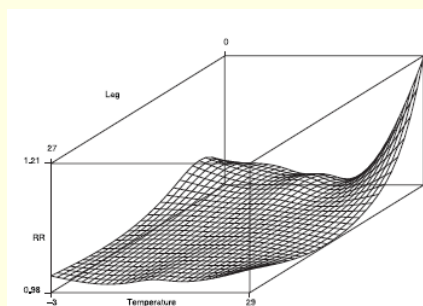
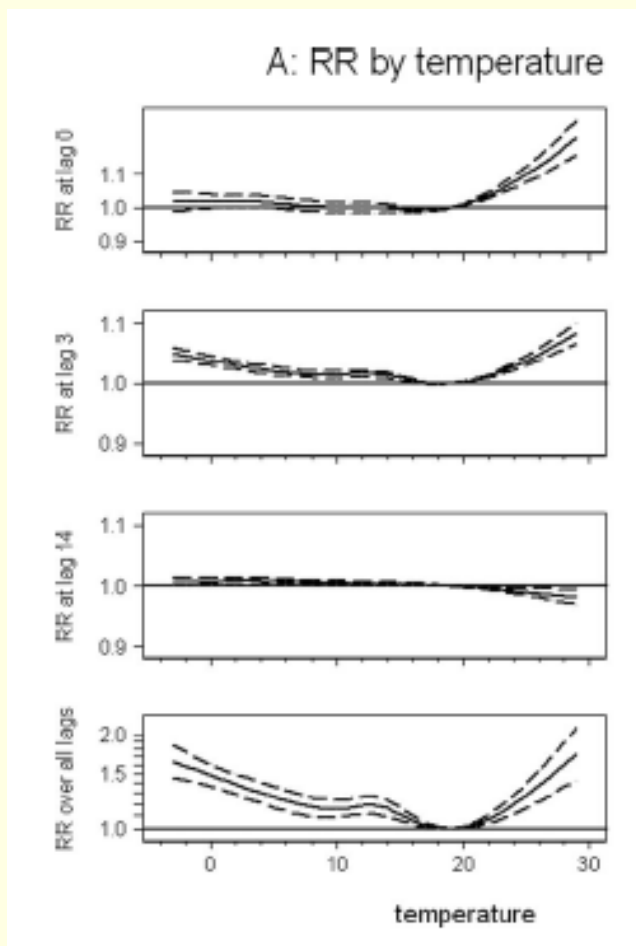
“Cross” **basis**
(BXP df)

RR by temperature (NCS) and lag(NCS)



5df temp X 4df lag = 20 df

Ways of looking at DLNMs

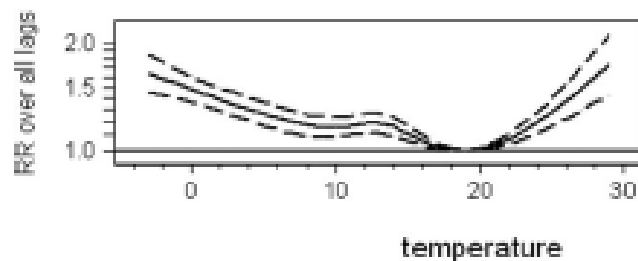
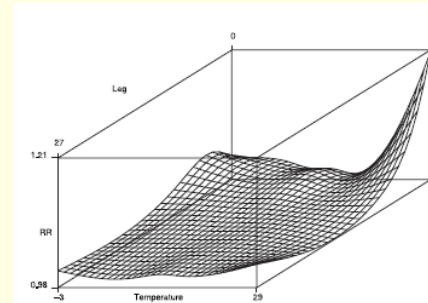
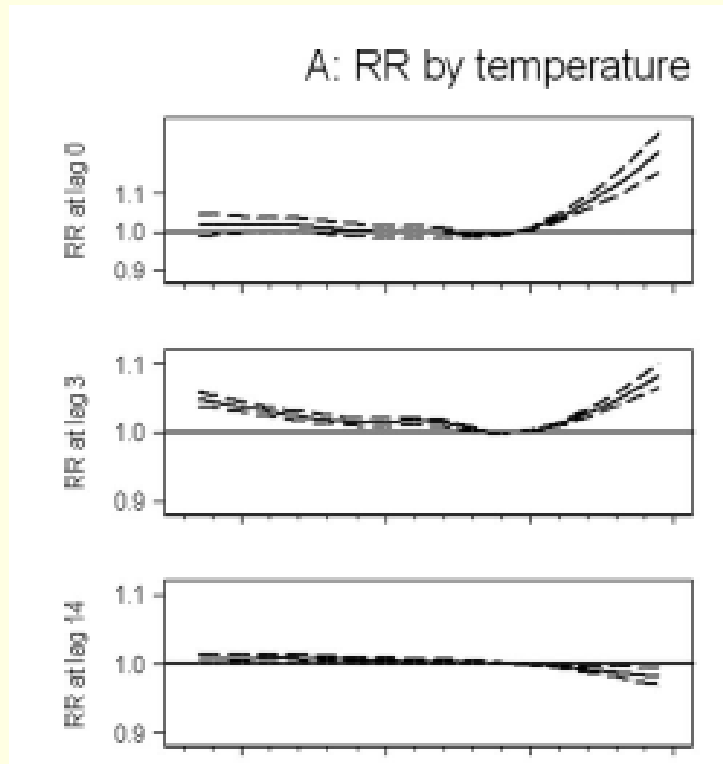


RR by temperature for given lag

RR vs 20°C

RR by lag for given temperature

- Adding graph of total (28 day) effect



Part 1 wrap-up

- Time series regression allows investigation of acute effects; many applications:
 - *Air pollution -> health outcomes*
 - *Weather -> health (incl. infectious diseases)*
 - *Circulating viruses -> mortality/admissions*
 - *Policy changes -> health (interrupted time series)*
 - *-...*
- DLNMs allow flexible modelling of delayed effects with possibly non-linear exposure-response.

Ecol TS: Counts, proportions, means;
+ Panel studies, cohorts, case-control, ...