



**Karolinska  
Institutet**

# **Point-wise averaging approach in dose-response meta-analysis of aggregated data**

**XXVIIIth International Biometric Conference**

**Alessio Crippa**

July 14th 2016

# Acknowledgements

My co-authors:

- ▶ Ilias Thomas
- ▶ Nicola Orsini

Funding organization:

- ▶ Karolinska Institutet's funding for doctoral students
- ▶ Young Scholar Award from the Karolinska Institutet's Strategic Program in Epidemiology

# Dose–response meta–analysis

Summarize results from multiple studies on the relation between a quantitative exposure and the occurrence of a health outcome

## Research questions

- ▶ What is the shape of the association between the quantitative exposure and the outcome?
- ▶ What are the exposure values associated with the best or worst outcome?
- ▶ How heterogenous are the individual dose–response curves?

## Aggregated data

gday	dose	case	control	n	rr	lb	ub
Ref.	0	165	172	337	1.00	1.00	1.00
<2.5	2	74	93	167	0.80	0.51	1.27
2.5-9.3	6	90	96	186	1.16	0.73	1.85
>9.3	11	122	90	212	1.57	0.99	2.51

## Two-stage approach

### First stage:

Estimate a common dose–response model in each study

( $i = 1, \dots, K$ )

$$\mathbf{y}_i = \mathbf{X}_i \boldsymbol{\beta}_i + \boldsymbol{\varepsilon}_i \quad (1)$$

with  $\mathbf{y}_i$  vector of non-referent log relative risks

### Second stage:

Combine study–specific regression coefficients  $\hat{\boldsymbol{\beta}}_i$ ,  $\text{Var}(\hat{\boldsymbol{\beta}}_i)$

$$\bar{\boldsymbol{\beta}} = \frac{\sum_{i=1}^K \mathbf{W}_i \hat{\boldsymbol{\beta}}_i}{\sum_{i=1}^K \mathbf{W}_i} \quad (2)$$

with  $\mathbf{W}_i = \left( \text{Var}(\hat{\boldsymbol{\beta}}_i) + \boldsymbol{\Psi} \right)^{-1}$

Results are presented as pooled relative risks for selected dose levels  $x^*$

$$\log \widehat{RR} = \mathbf{X}^* \bar{\beta} \quad (3)$$

### General limitations:

- ▶ The dose–response model needs to be the same across studies:  
Poor fit in some of the study-specific dose-response analyses
- ▶ Pooling of  $\hat{\beta}_i$  discards the initial exposure range:  
Predictions may be affected by extrapolation

# Aims

To propose a point-wise averaging approach to take into account differential curves and exposure distributions

Specific aims:

- ▶ To introduce more flexibility in the dose-response analysis
- ▶ To allow each study to contribute to the overall curve based on the observed exposure distribution

## Point-wise averaging approach

Initially proposed for IPD meta-analysis, it consists of the following steps:

- 1 Estimation of study-specific dose–response curves
- 2 Study-specific predictions
- 3 Combining study-specific predictions by meta-analysis



## Estimation of study-specific curves

$$\mathbf{y}_i = f_i(\mathbf{x}_i; \beta_i) + \varepsilon_i \quad (4)$$

$f_i$  can differ across studies

Fractional polynomials of order 2

$$\mathbf{y}_i = \beta_{i1}\mathbf{x}_i^{p_{i1}} + \beta_{i2}\mathbf{x}_i^{p_{i2}} + \varepsilon_i \quad (5)$$

with  $\mathbf{x}_i^0 = \log(\mathbf{x}_i)$

Select  $(p_{i1}, p_{i2})$  in the set of values  $\{-2, -1, -0.5, 0, 0.5, 1, 2, 3\}$   
which minimize the AIC

## Estimation of study-specific curves (2)

Restricted cubic splines with 3 knots

$$\mathbf{y}_i = \beta_{i1} f_{i1}(\mathbf{x}_i) + \beta_{i2} f_{i2}(\mathbf{x}_i) + \varepsilon \quad (6)$$

$$f_{i1} = \mathbf{x}_i$$

$$f_{i2} = \frac{(\mathbf{x}_i - k_{i1})_+^3 - \frac{k_{i3} - k_{i1}}{k_{i3} - k_{i2}} (\mathbf{x}_i - k_{i2})_+^3 + \frac{k_{i2} - k_{i1}}{k_{i3} - k_{i2}} (\mathbf{x}_i - k_{i3})_+^3}{(k_{i3} - k_{i1})^2} \quad (7)$$

Study specific knots location

## Prediction and pooling

Limit prediction to study-specific range:  $\mathbf{x}^* \in \text{range}(\mathbf{x}_i)$

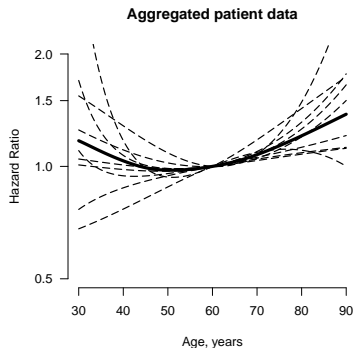
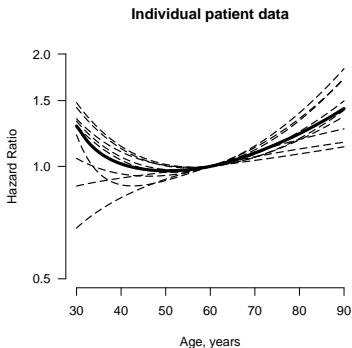
$$\begin{aligned}\log \widehat{RR}_i &= \mathbf{X}^* \beta_i \\ \text{Var}(\log \widehat{RR}_i) &= \text{diag}(\mathbf{X}^{*\top} \text{Var}(\beta_i) \mathbf{X}^*)\end{aligned}\quad (8)$$

Pool the study-specific effects (log RR)

$$\log \bar{RR}|_{\mathbf{x}^*} = \frac{\sum_{i=1}^K W_i \log \widehat{RR}_i I(\mathbf{x}^* \in \text{range}(\mathbf{x}_i))}{\sum_{i=1}^K W_i I(\mathbf{x}^* \in \text{range}(\mathbf{x}_i))}\quad (9)$$

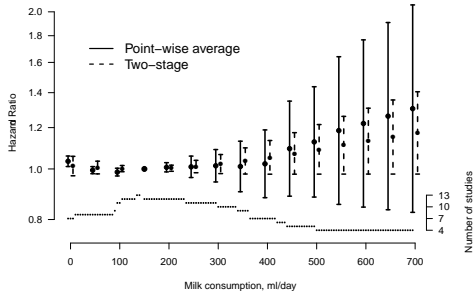
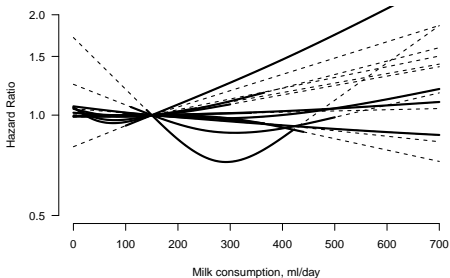
## Comparison with IPD meta-analysis

Based on breast cancer patients in the SEER program  
(<http://seer.cancer.gov/>)



# Comparison with two-stage meta-analysis

Re-analysis of a dose-response meta-analysis between milk and mortality



## Conclusions

- ▶ A point-wise averaging approach can properly address differential dose–response curves and exposure distribution, and limit the impact of extrapolation
- ▶ Results can be presented graphically for a grid of exposure values
- ▶ Differences with a two-stage approach may depend upon exposure distributions and strategies used in the dose–response analysis

# References I

- ▶ Crippa A, Orsini N. Multivariate Dose–Response Meta–Analysis: the dosresmeta R Package. 2016. J Stat Softw. In press.
- ▶ Sauerbrei W, Royston P. A new strategy for meta–analysis of continuous covariates in observational studies. *Statistics in medicine*. 2011 Dec 10;30(28):3341–60.
- ▶ Gasparrini A, Armstrong B, Kenward MG. Multivariate meta–analysis for non-linear and other multi-parameter associations. *Statistics in medicine*. 2012 Dec 20;31(29):3821–39.
- ▶ Orsini N, Li R, Wolk A, Khudyakov P, Spiegelman D. Meta–analysis for linear and nonlinear dose–response relations: examples, an evaluation of approximations, and software. *American journal of epidemiology*. 2012 Jan 1;175(1):66–73.
- ▶ Rota M, Bellocco R, Scotti L, Tramacere I, Jenab M, Corrao G, La Vecchia C, Boffetta P, Bagnardi V. Random–effects meta–regression models for studying nonlinear dose–response relationship, with an application to alcohol and esophageal squamous cell carcinoma. *Statistics in medicine*. 2010 Nov 20;29(26):2679–87.