

Learning and Predicting Real-World Treatment Effect based on Randomized Controlled Trials and Registry Data: A CASE STUDY ON RHEUMATOID ARTHRITIS

BBS Seminar, Novartis Campus, Basel

January 13th, 2016

Eva-Maria Didden*

On behalf of GetReal Work Package WP4

* Institute of Social and Preventive Medicine, University of Berne, Switzerland



innovative
medicines
initiative

The research leading to these results has received support from the Innovative Medicines Initiative Joint Undertaking under grant agreement no [115303], resources of which are composed of financial contribution from the European Union's Seventh Framework Programme (FP7/2007-2013) and EFPIA companies' in kind contribution.
www.imi.europa.eu

ACKNOWLEDGEMENTS

- **Institute of Social and Preventive Medicine, University of Berne, Switzerland:**
Noemi Hummel, Matthias Egger, and Yann Ruffieux
- **Health Technology Assessment Group, F. Hoffmann-La Roche Ltd, Basel, Switzerland:**
Sandro Gsteiger
- **Hôpitaux Universitaires de Genève, HUG, Switzerland:**
Axel Finck
- **Inselspital, Department of Rheumatology, Immunology and Allergology, Berne, Switzerland:**
Stephan Reichenbach

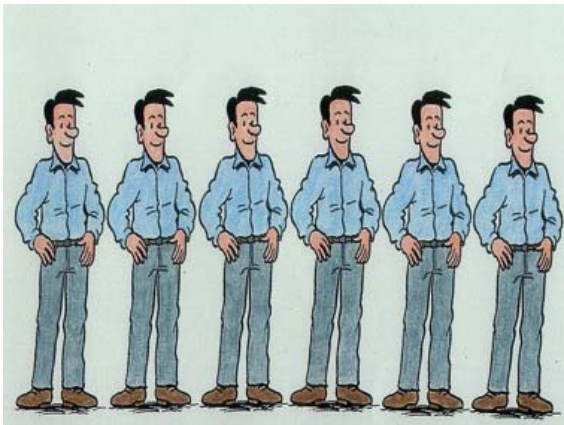
Outline

- **GetReal Project**
 - Overview
 - Work package 4
- **Rheumatoid Arthritis Case Study**
 - Research question
 - Predictive modelling framework
 - Discussion

GetReal – Background

Efficacy-effectiveness gap:

Differences between treatment effect in a clinical trial population and in daily clinical practice



GetReal – Objective

- Development of new methods for collecting and synthesizing real-world evidence (RWE)
 - Presentation of a guideline on how to adopt these methods into the early process of pharmaceutical research and development (R&D) and of healthcare decision making
- Close collaboration between companies, healthcare decision makers, academic institutions and other stakeholders
- Generation of a consensus on best practice in the use of RWE in regulatory and reimbursement decision-making

GetReal – Public-Private Partnership

11 Public partners:

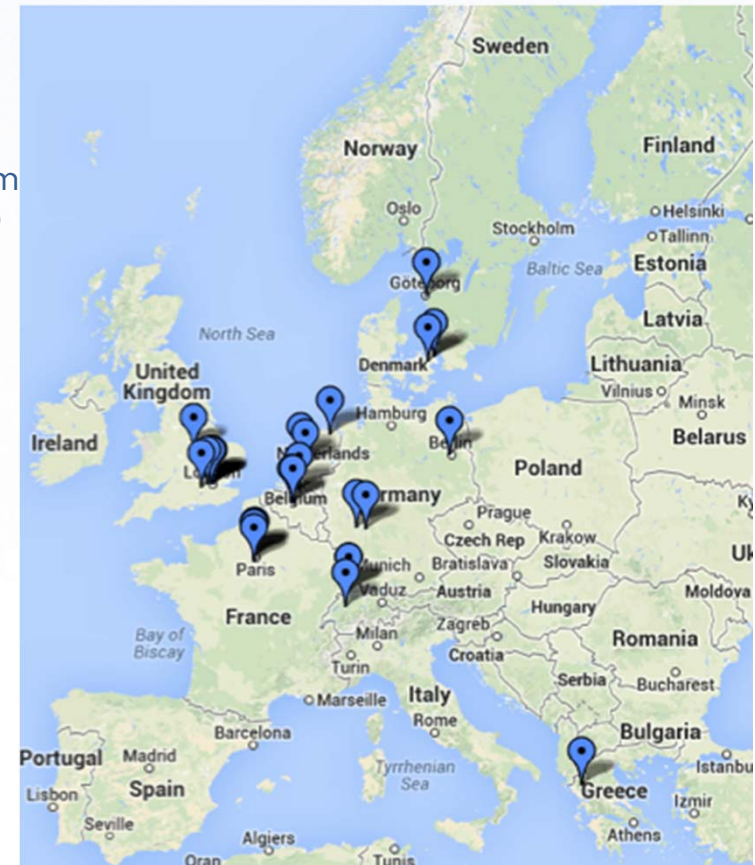
- University Medical Center Utrecht, the Netherlands
- University Medical Center Groningen, the Netherlands
- University of Ioannina, Greece
- University of Bern, Switzerland
- University of Leicester, UK
- University of Manchester, UK
- European Organisation for Research and Treatment of Cancer, Belgium
- Zorginstituut Nederland, the Netherlands
- Haute Autorité de Santé, France
- National Institute for Health and Care Excellence, UK
- European Medicines Agency, UK

15 EFPIA companies:

- GlaxoSmithKline
- Amgen
- AstraZeneca
- Bayer
- Boehringer Ingelheim
- Bristol Myers Squibb
- Eli Lilly
- Janssen
- LASER
- Merck Serono
- MSD
- Novartis
- Novo Nordisk
- Roche
- Sanofi
- Takeda

Patients' organizations:

International Alliance of Patients' Organizations



GetReal - Work Packages

WP1:

Collaborate with key stakeholders in medicine development to assess

- the acceptability and usefulness of RWE
- approaches to the analysis of RWE, i.e. the effectiveness of new therapies

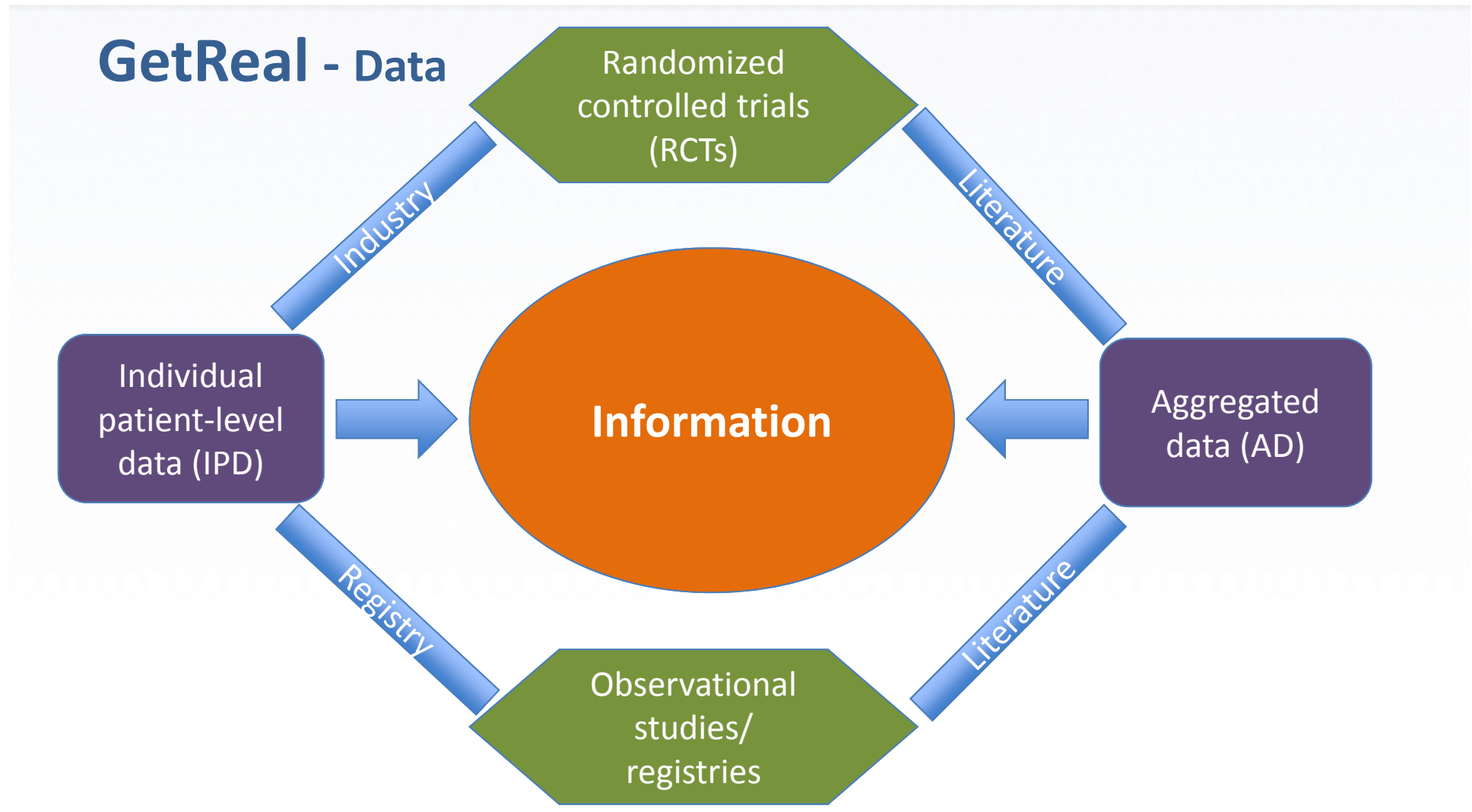
WP2:

Study the scientific validity of RWE study designs and explore analytical approaches to better inform pharmaceutical R&D and healthcare policymakers

WP3:

- Identify the operational challenges of performing RWE studies early in the medicine development process
- Develop practical solutions to better inform the planning and delivery of RWE studies

WP5: Consortium Project Management



WP4 - Tasks

- Develop best practices for evidence synthesis
 - Network meta-analysis based on AD
 - IPD meta-analysis
 - Mathematical modelling to predict relative effectiveness from RCT efficacy data→ 3 systematic reviews
- Develop and investigate methods using IPD
→ 4 case studies using IPD from RCTs and observational studies
- Develop a mathematical modelling framework to predict relative effectiveness from RCT efficacy → *see below*
- Develop user-friendly evidence synthesis software and relevant training material to support best practice

Case Study on Rheumatoid Arthritis (RA)

1. Research question
2. Predictive modelling framework
3. Results
4. Discussion



Research Question

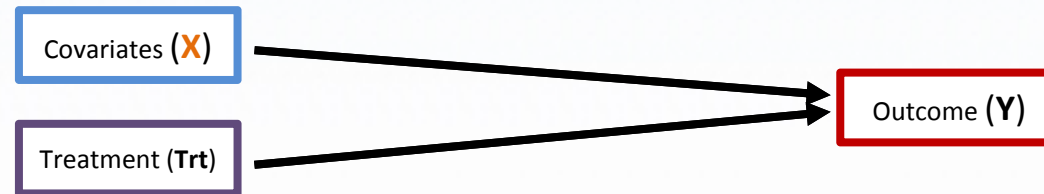
How can we - based on randomized controlled trial (RCT) and observational data - set up a mathematical model that allows us to predict treatment effect in patients with *Rheumatoid Arthritis* (RA)?

Predictive Modelling - Procedure

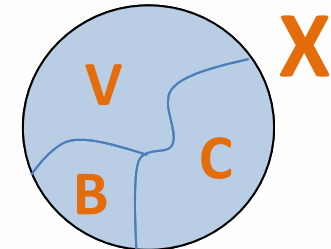
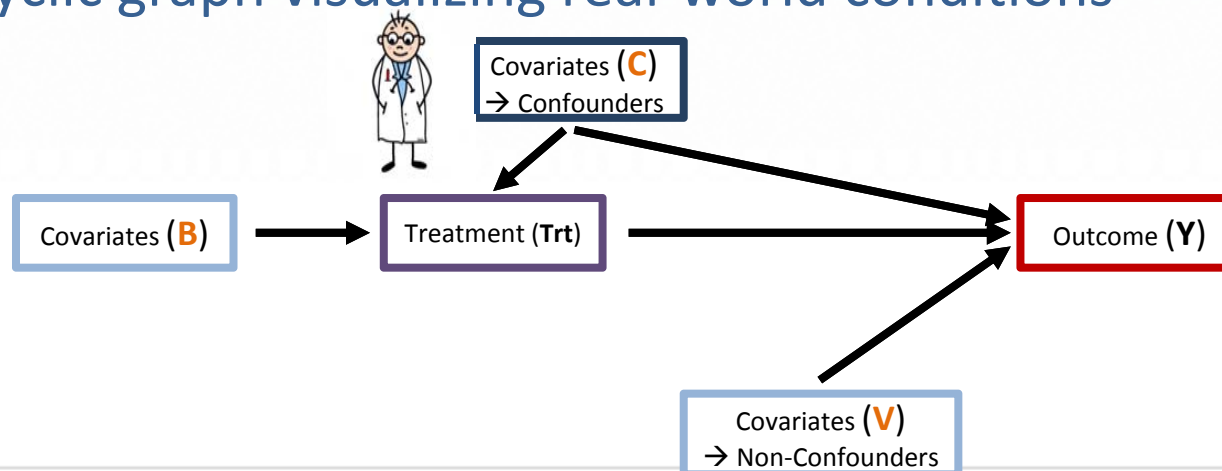
1. Selection of a simple linear regression model for data from RCTs
2. Development of a *marginal structural model* (MSM) for observational data, to adjust for potential confounders
3. Incorporation of insights from both modelling approaches into a Bayesian inference framework
4. Prediction of treatment effect for a new real-world population, possibly under new study conditions

Predictive Modelling – Graphical model representation

- Acyclic graph visualizing RCT conditions

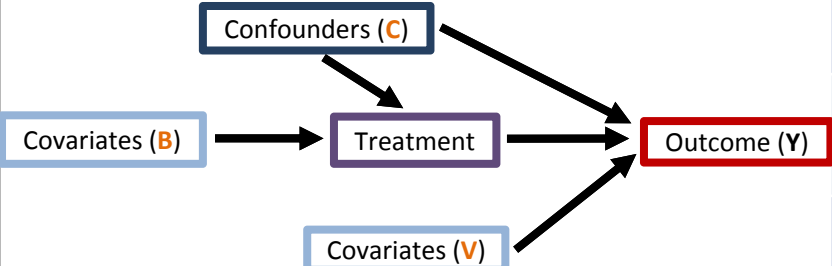


- Acyclic graph visualizing real-world conditions



Predictive Modelling – Variable selection

Outcome: Change in	RCT DATA Covariates X	OBSERVATIONAL/REGISTRY DATA			E x p e r t (RA) Stats Not selec- ted
		Covariates B	Covariates V	Confounders C	
DAS28	gender	calendar year	BMI/obesity	age	}
HAQ	seropositivity	hospital (y/n)	gender	disease duration	
EQ5D	baseline DAS28	socio-economics	steroid intake	seropositivity	
ACR	baseline HAQ-DI	# [concomitant DMARDs]	smoking	
CDAI	# [previous anti-TNF agents]		baseline HAQ-DI	# [previous anti-TNF agents]	}
RDAI		type of concomitant DMARDs	baseline DAS28	
.....				comorbidities	}
				# [previous DMARDs]	
				



```

graph LR
    B[Covariates (B)] --> T[Treatment]
    C[Confounders (C)] --> T
    T --> Y[Outcome (Y)]
    V[Covariates (V)] --> Y
    
```


Predictive Modelling – Formal model representation

- Linear model (LM) for RCT data:

α : Intercept, β : Treatment effect
 γ : (non-confounding) Covariate effect

$$Y_{\text{rct}} \sim N(\alpha_{\text{rct}} + \beta_{\text{rct}} \text{Trt} + \gamma_{\text{rct}} X_{\text{rct}}, \sigma_{\text{rct}}^2 I)$$

- MSM for the observational data:

$$\text{Trt} = \begin{cases} 1, & \text{biological agent} \\ 0, & \text{control treatment} \end{cases}$$

→ weighted linear regression model

$$Y_{\text{ob}} \sim N(\alpha_{\text{ob}} + \beta_{\text{ob}} \text{Trt} + \gamma_{\text{ob}} V_{\text{ob}}, \sigma_{\text{ob}}^2 W_{\text{ob}}^{-1}), \quad W_{\text{ob}} \propto \frac{1}{f(\text{Trt} | C_{\text{ob}})}$$

«If both models are sufficiently well specified and further MSM assumptions hold, the estimated treatment effects should be similar.»

Predictive Modelling – Formal model representation

Likelihood: Gaussian MSM of the form

$$Y|\Theta \sim N(\alpha + \beta Trt + \gamma V, \sigma^2 W^{-1}) ; \quad \Theta = \{\alpha, \beta, \gamma, \sigma^2, \Theta(W)\}$$

Priors:

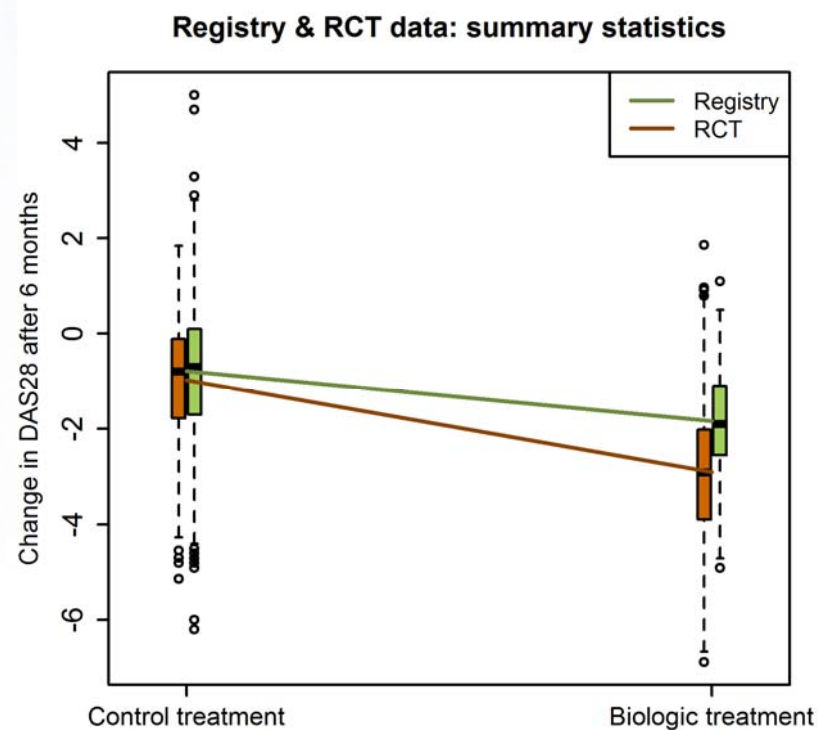
- Set $\beta \sim N(\hat{\beta}_{rct}, \tau^2)$, where τ^2 must be carefully determined
- For the remaining parameters, choose suitable non- or weakly-informative priors

Predictions:

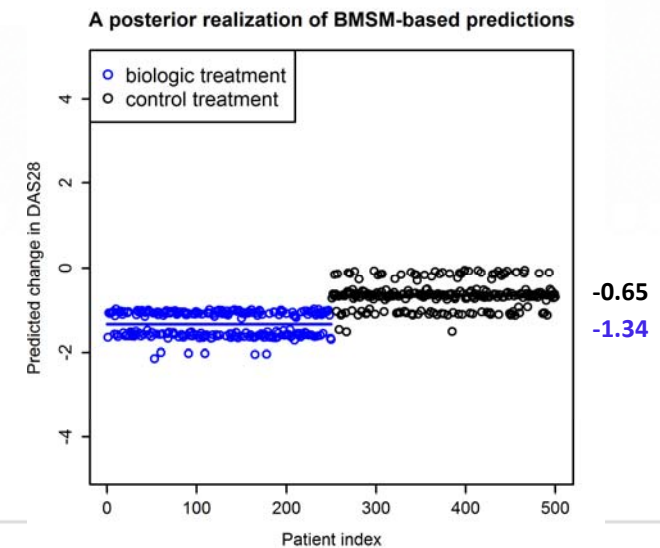
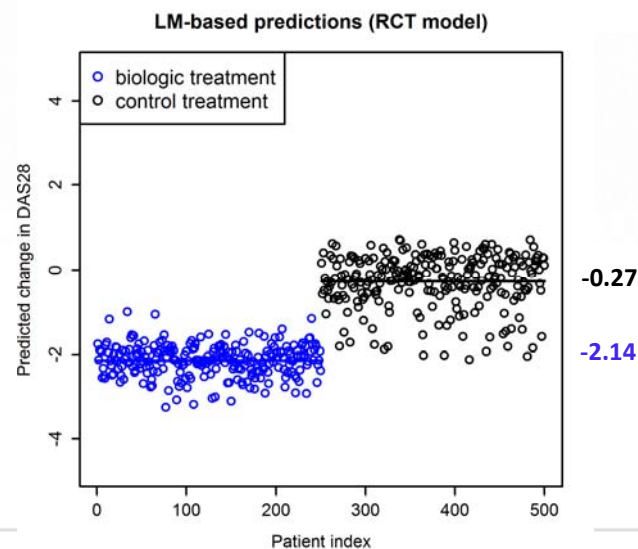
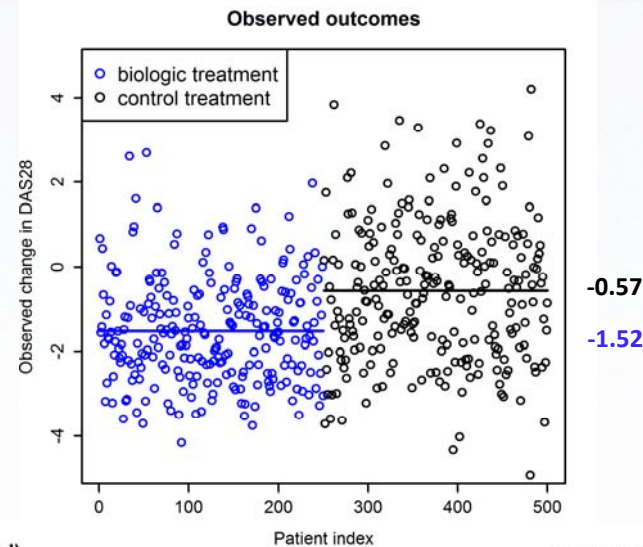
1. Take the previously selected MSM structure and variables as a modelling basis
2. Estimate the posterior distributions of all unknown parameters
3. For any new set of observational data Y , draw posterior realizations (predictions)

$$\hat{Y}_{BMSM} = \{\hat{Y}^{(1)}, \hat{Y}^{(2)}, \dots\} \text{ from the according posterior predictive distribution}$$

Results – Descriptive data analysis

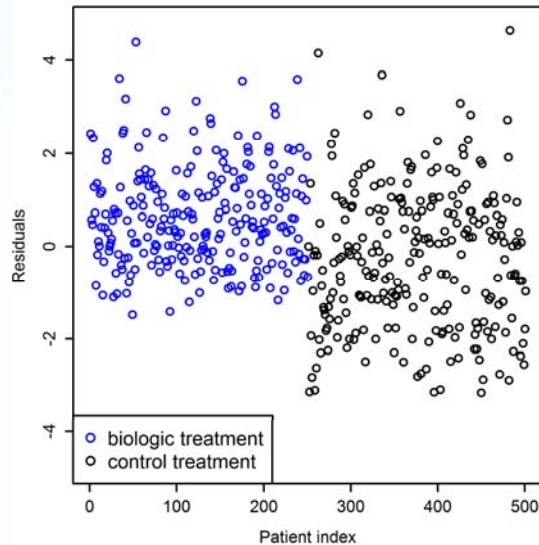


Results – Predictions for a new real-world population



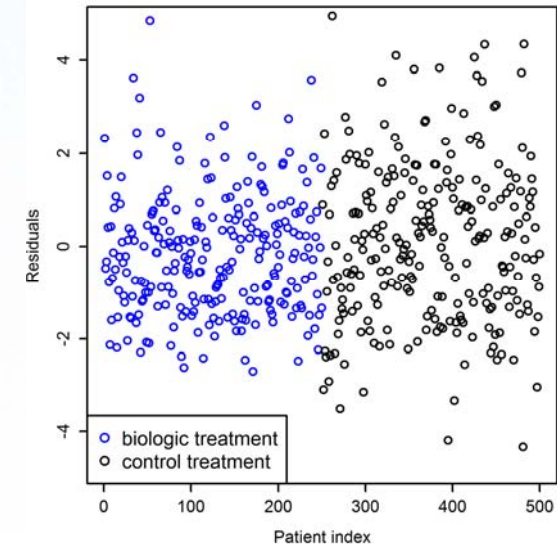
Results – Goodness-of-fit

Residuals $\hat{Y}_{lm} - Y$, under simple LM assumptions



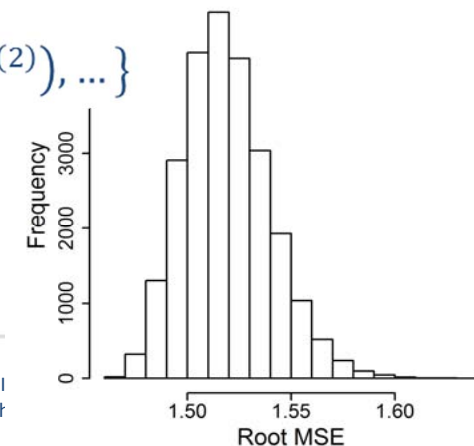
$$rMSE(\hat{Y}_{lm}) = 1.59$$

A posterior realization of residuals $\hat{Y}^{(\cdot)} - Y$, derived from our BMSM



Distribution of $\{rMSE(\hat{Y}^{(1)}), rMSE(\hat{Y}^{(2)}), \dots\}$

$$rMSE(\hat{Y}) = \sqrt{\frac{1}{N} \sum_{i=1}^N (\hat{Y}_i - Y_i)^2}$$



Discussion – Summary

- **Why use an MSM to adjust for confounding?**

- Flexibly applicable to different types of outcome and treatment data
- Easily extendable to settings with time-varying treatment and confounding

Most critical assumption: assumption of no unmeasured confounding

- **Why work within a Bayesian inference and prediction framework?**

- Inclusion of prior knowledge, possibly gained from multiple data sources
- Estimation of posterior and posterior predictive distributions, and derivation of all measures of interest (e.g. posterior modulus/mean of the parameters...)
- Relaxation of the missing data problem

Discussion – Work in progress

- Development of a framework to evaluate goodness-of-fit-and-prediction
- Consideration of dynamic treatment regimes with time-varying confounders and censoring
- Inclusion of covariate/confounder interactions
- Inclusion of additional patient records, e.g. from different countries
- Inclusion of AD from RCTs and observational studies
- Addressing the question whether and under which conditions the BMSM can be used to predict the real-world effect of a new drug, provided that only RCT data are available
- ...

REFERENCES

- <http://www.imi-getreal.eu/>
- A. Gelman, J. B. Carlin, H. S. Stern, and D. B. Rubin, *Bayesian data analysis*, Volume 2, London: Chapman & Hall/CRC, 2014.
- A. Gelman, and X.-L. Meng (eds.), *Applied Bayesian modeling and causal inference from incomplete-data perspectives*, John Wiley & Sons, 2004.
- J. M. Robins, M. A. Hernan, and B. Brumback, "Marginal structural models and causal inference in epidemiology." *Epidemiology*, Volume 11, Issue 5: 550-560, 2000.