# Application Of Causal Inference Methods for Analyzing Randomized Controlled Trial Data Combined with Real-Word Data

<sup>1\*</sup>Alfredo E. Farjat, <sup>2</sup>Kaisa Laapas, <sup>3</sup>James Potts

<sup>1</sup>Bayer BV, Hoofddorp, Netherlands; <sup>2</sup>Bayer Oy, Espoo, Finland; <sup>3</sup>Bayer Corporation, Whippany, USA \*corresponding author: Alfredo E. Farjat, <u>alfredo.farjat@bayer.com</u>.

#### Introduction

- Randomized controlled trials (RCTs) augmented with real-world data (RWD) can provide high-quality evidence to evaluate safety and effectiveness for new medical products in regulatory settings
- Several methods have been developed for creating statistically comparable groups and evaluating treatment effect
- However, poor balance between the baseline characteristics of comparing groups can be a critical limitation

# **Objectives**

 Compare different matching/weighting methods for the evaluation of treatment effect in a RCT augmented with RWD in the presence of poor balance of baseline characteristics

## Methods

- RCT Phase II: Participants were randomized to one of two experimental arms or an internal control arm (ICA) and were evaluated for bleeding events within three months
- External control arm (ECA): The Finnish healthcare records was used to develop an ECA, fulfilling eligibility for the RCT
- Matching and weighting methods based on 28 known confounders were used to create statistically comparable groups between RCT participants and those eligible from RWD
- After **propensity score (PS)** trimming<sup>1</sup>, inverse probability of treatment weighting (**IPTW**)<sup>2</sup>, G-computation (**G-comp**)<sup>3</sup>, Augmented IPTW (**AIPTW**)<sup>2</sup>, Targeted Maximum Likelihood Estimation (**TMLE**)<sup>3</sup>, and Overlapping Weights (**OW**)<sup>1</sup> methodologies were applied for comparing the ICA to ECA, and then to evaluate the treatment effect of the combined ICA+ECA to the RCT experimental arm
- In addition, treatment effect was evaluated using the propensity score composite likelihood (PSCL)<sup>4</sup> approach

Active Treatment Arm 1

RND
N=755

Active Treatment Arm 2

Internal Control Arm (ICA)

RWD
eligible

External Control Arm (ECA) (after PS trimming) N=2847

Figure 1: RCT design and Finnish anonymized RWD

#### Results

- The RCT had n=505 participants randomized to the experimental arms and n=250 to the ICA
- A pool of eligible RWD Finnish subjects (n=3327) was established, of which n=2847 were included after trimming (480 excluded)
- Figure 4 shows no difference in bleeding outcomes between the ICA and the matched ECA across methods
- **Figure 5** shows treatment effect by method for RCT and RCT with augmented control arm (ICA n=250 + ECA n=2847 = 3097 control subjects). Precision is increased with external data augmentation but point estimates moved towards zero difference

Figure 2: Overlap of propensity scores' distributions

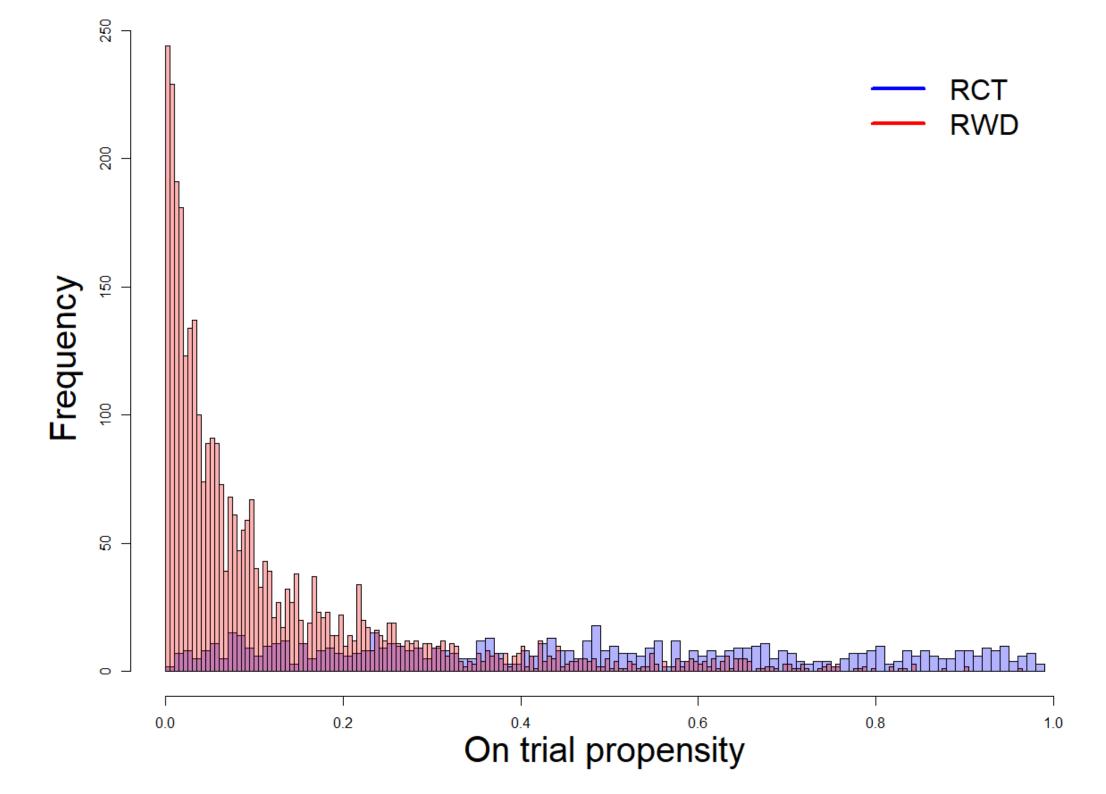


Figure 3: Absolute standardized mean difference between RCT and RWD

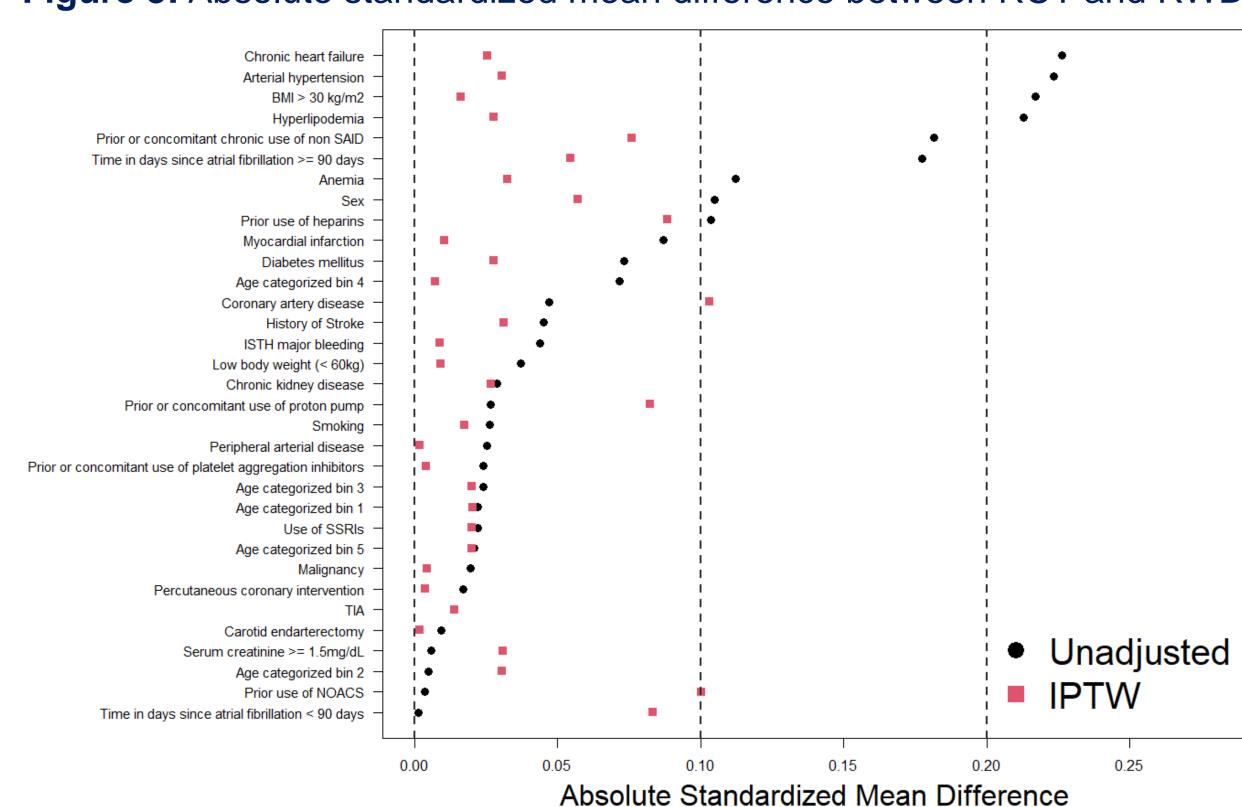


Figure 4: Major bleeding events rate difference between ICA and ECA by method

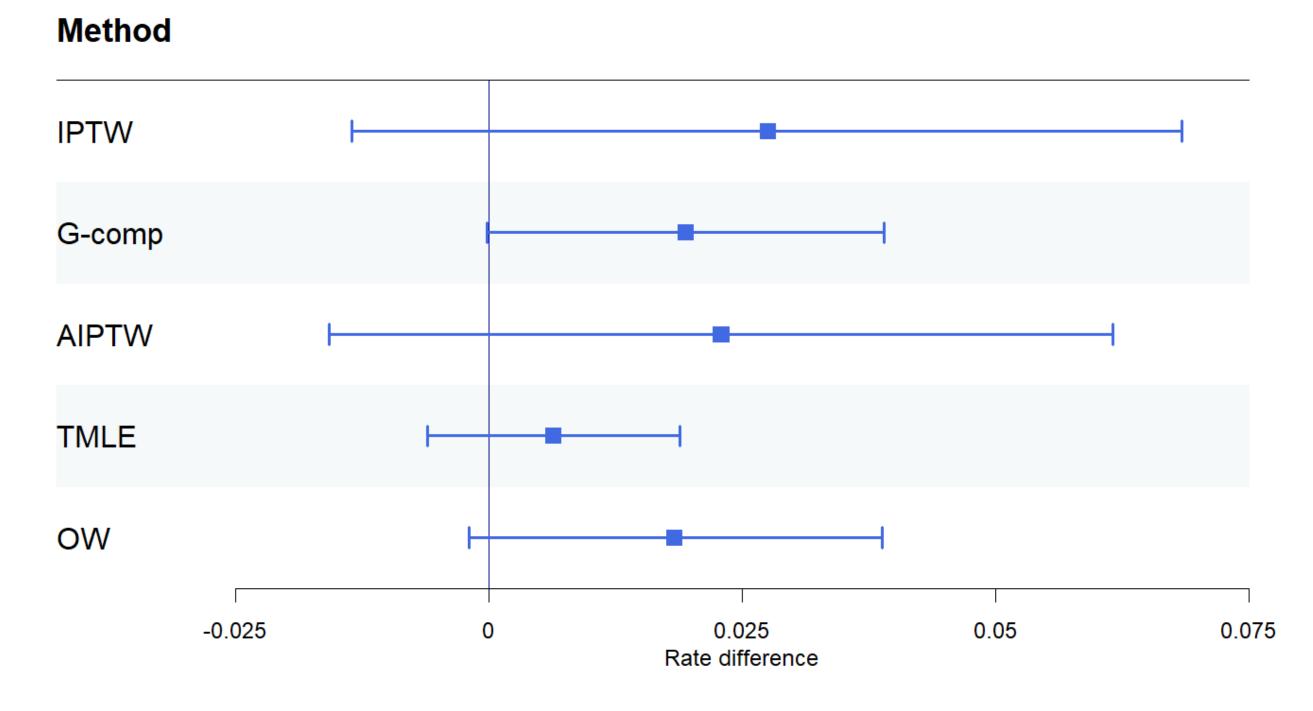
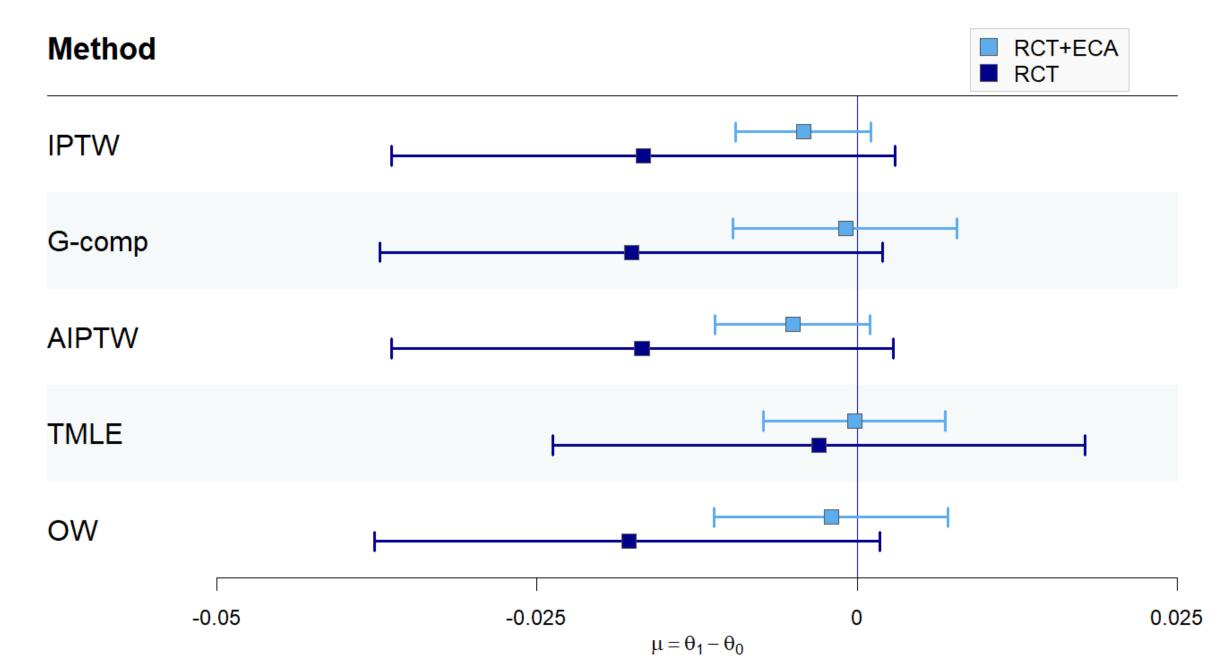


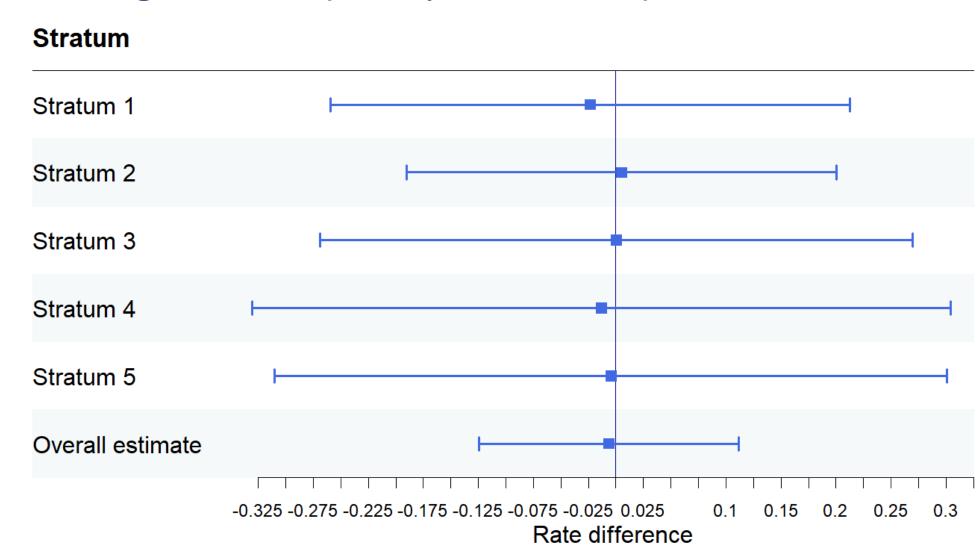
Figure 5: Treatment effect by method for RCT and RCT + ECA



# **Propensity Score Composite Likelihood**

- PS distribution was stratified into 5 strata
- In total, 255 subjects were borrowed from RWD to attain a 1:1 treatment-to-controls ratio
- Subjects were split proportionally to the distance in PS distributions in each stratum
- Figure 6 indicates no overall difference in bleeding rates

Figure 6: Propensity Score Composite Likelihood



### **Conclusions**

- Several causal inference methods are available for the analysis of RCT data combined with RWD
- Many of these methods were applied and showed similar results arriving to the same conclusion in the presence of poor propensity score distribution overlap

**References**: 1. Li et al. JASA. 2018, 113(521):390-400; 2. Kurz CF, Medical Decision Making 2021, 42(2):156-167; 3. Schuler MS and Rose S, American Journal of Epidemiology 2017, 185(1):65-73; 4. Chen WC et al. Journal of Biopharmaceutical Statistics 2020, 30(3):508-520.