

Verwendung von Beobachtungsdaten in der Nutzenbewertung – Einfluss von Design and Analysetechniken auf den Bias von Effektschätzungen

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Disclaimer

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Ansichten und Meinungen sind meine
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Organisation.



Agenda

1. Concept Causal Inference
2. Challenges of Observational Studies
3. Target Trial
4. Potential Biases: Case Example
5. So What?

Concept Causal inference

Prediction and Causal Effect

A — B

Statistical Relation:
"If I learn ..."

- A is **related** to B
- If I **learn** about A, I know more about B than before (and vice versa)
- I can use A to (better) predict B and vice versa
- A and B are not independent
- Correlation coefficient $\neq 0$
- $P(A \text{ and } B) \neq P(A) \cdot P(B)$
- Information "flows" between A and B

A $\longrightarrow$ B

Causal Relation:
"If I change ..."

- A **causes** B
- If I **change** A, B changes (not vice versa)
- I can use interventions on A to improve B (not vice versa)
- $P(B|\text{do}(A=0)) \neq P(B|\text{do}(A=1))$
- Effect "flows" from A to B

Counterfactual Concept

Parallel Universes



Clearing The Air On 'NASA's Antarctic Parallel Universe' – Australian Research & Space Exploration

Action



Only actions and
consequences of the
actions change

Everything else stays
the same

- Time
- Climate
- Environment
- Social aspects
- Other actions
- ...

Compare consequences

$$Y_A \rightarrow Y_I$$

Challenges of observational studies

Challenges of Causal Analyses of RWD

Confounding

- Time independent and time dependent

Selection bias

- Selection by indication

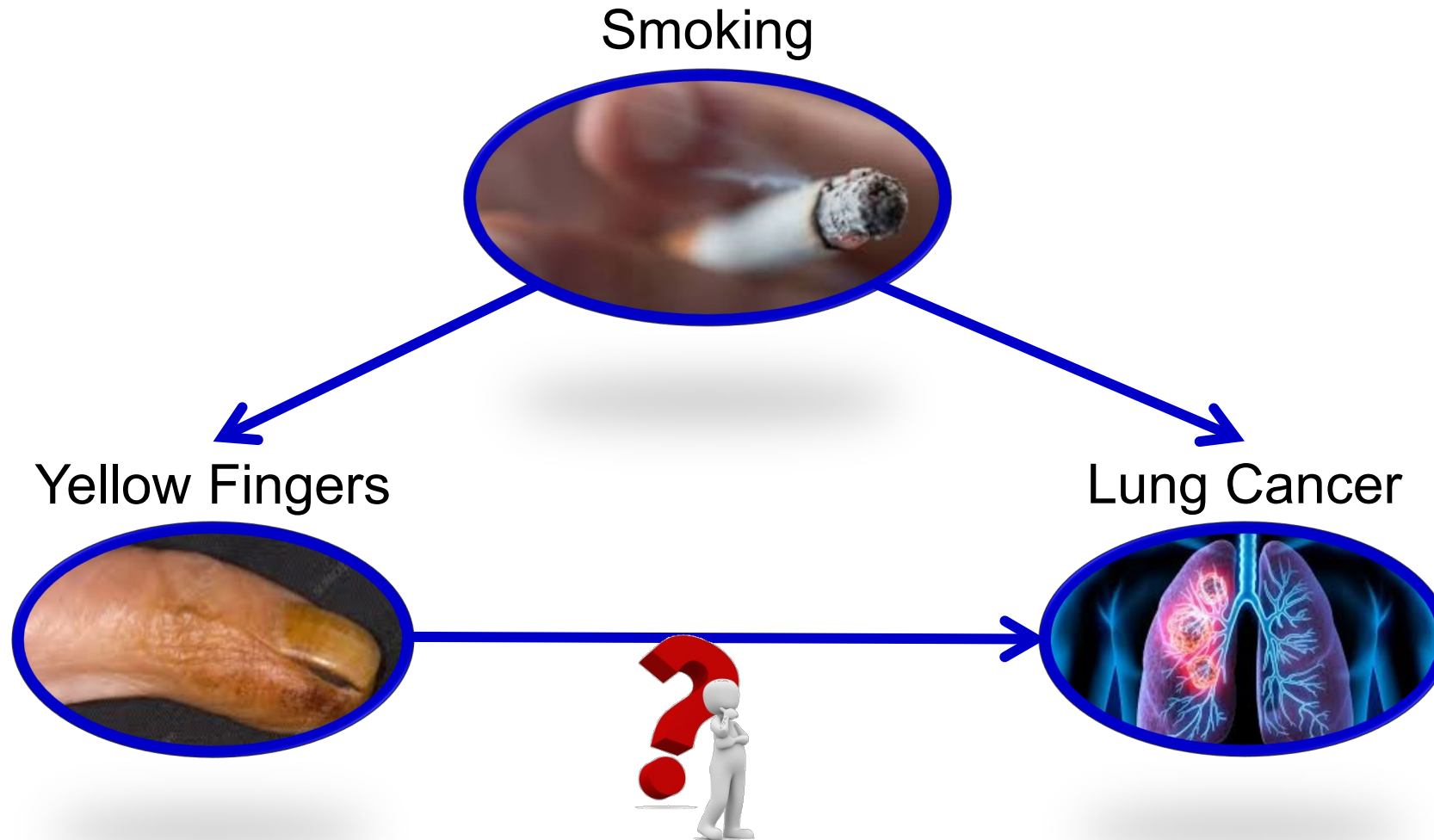
Immortal time bias

- Time zero bias

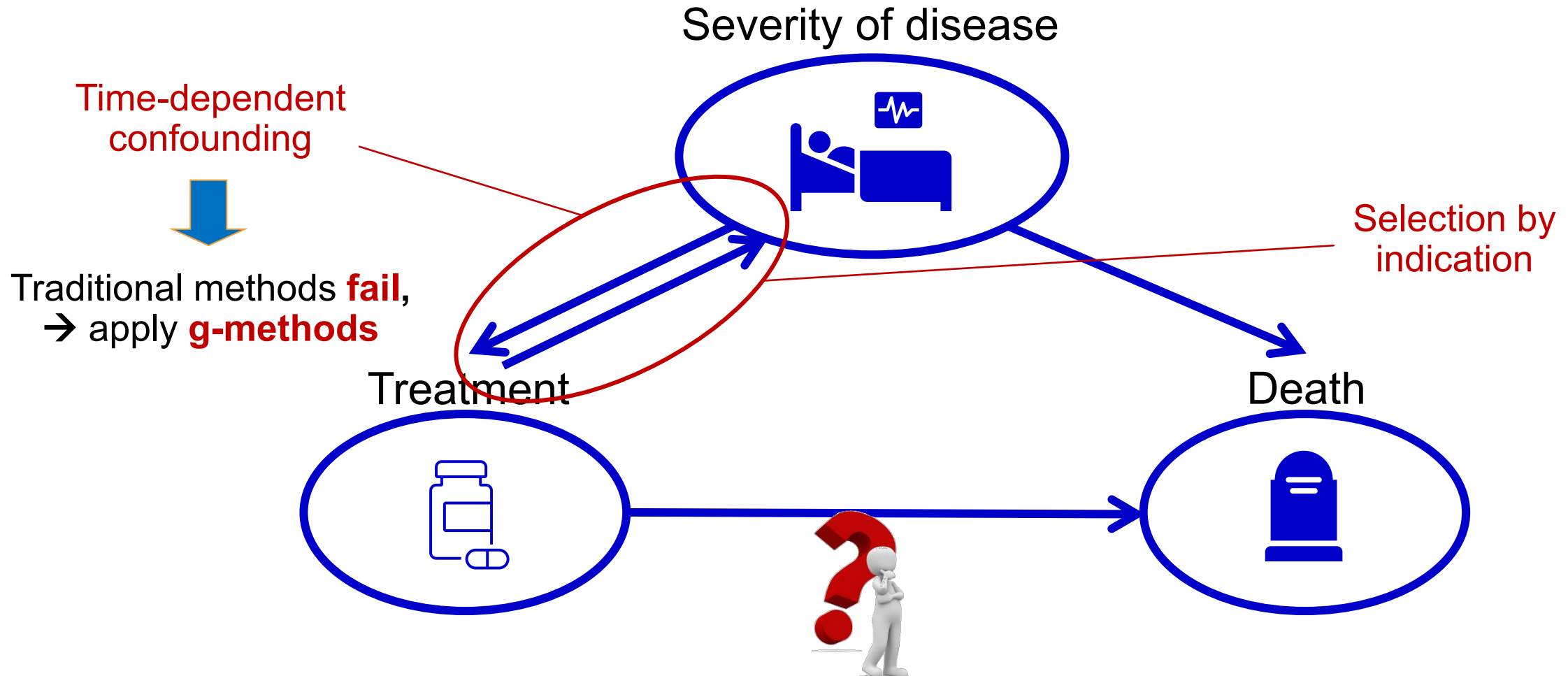
Adequate data

- Reliability
- Extent
- Accessibility

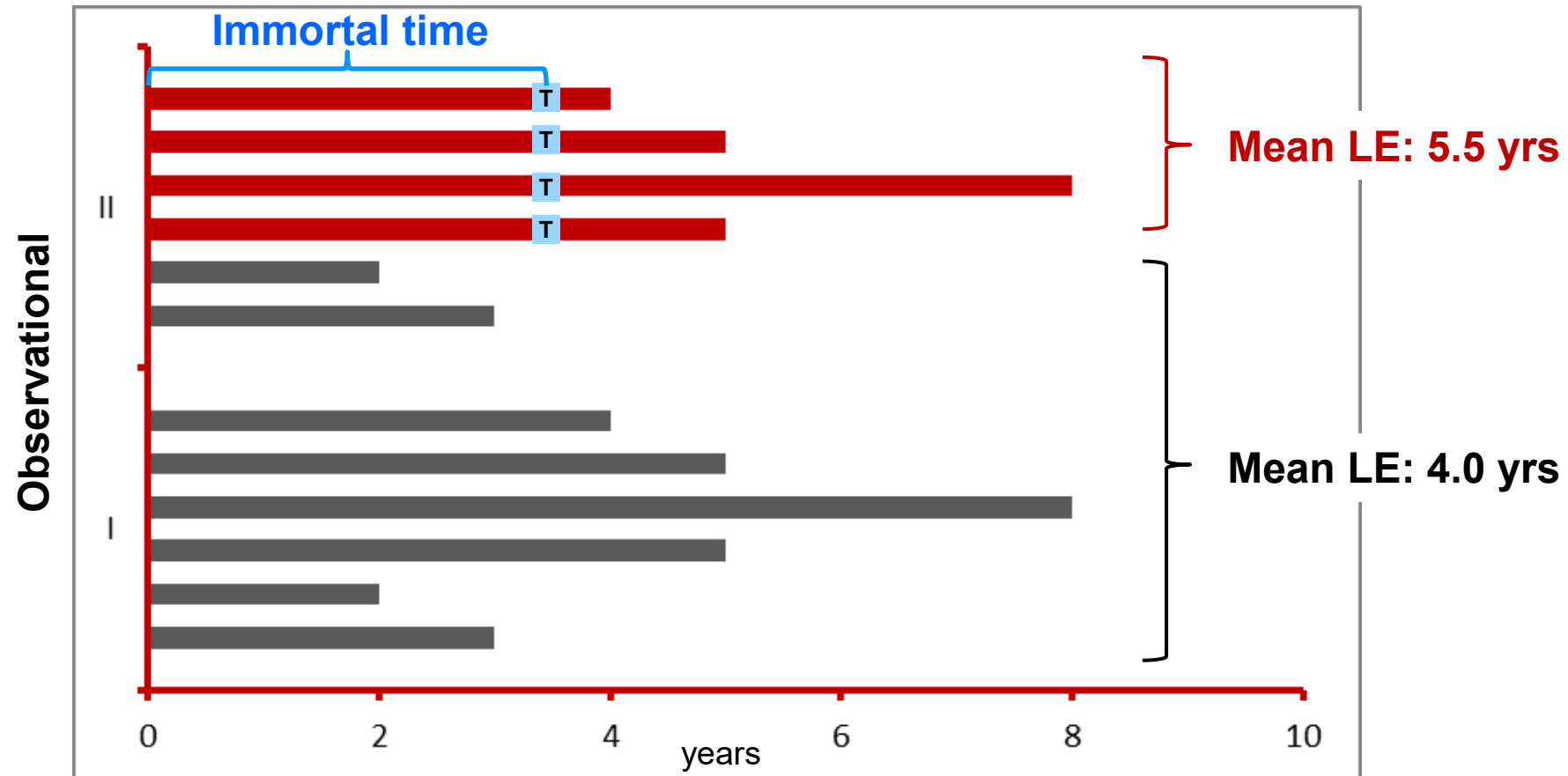
Confounding



Confounding

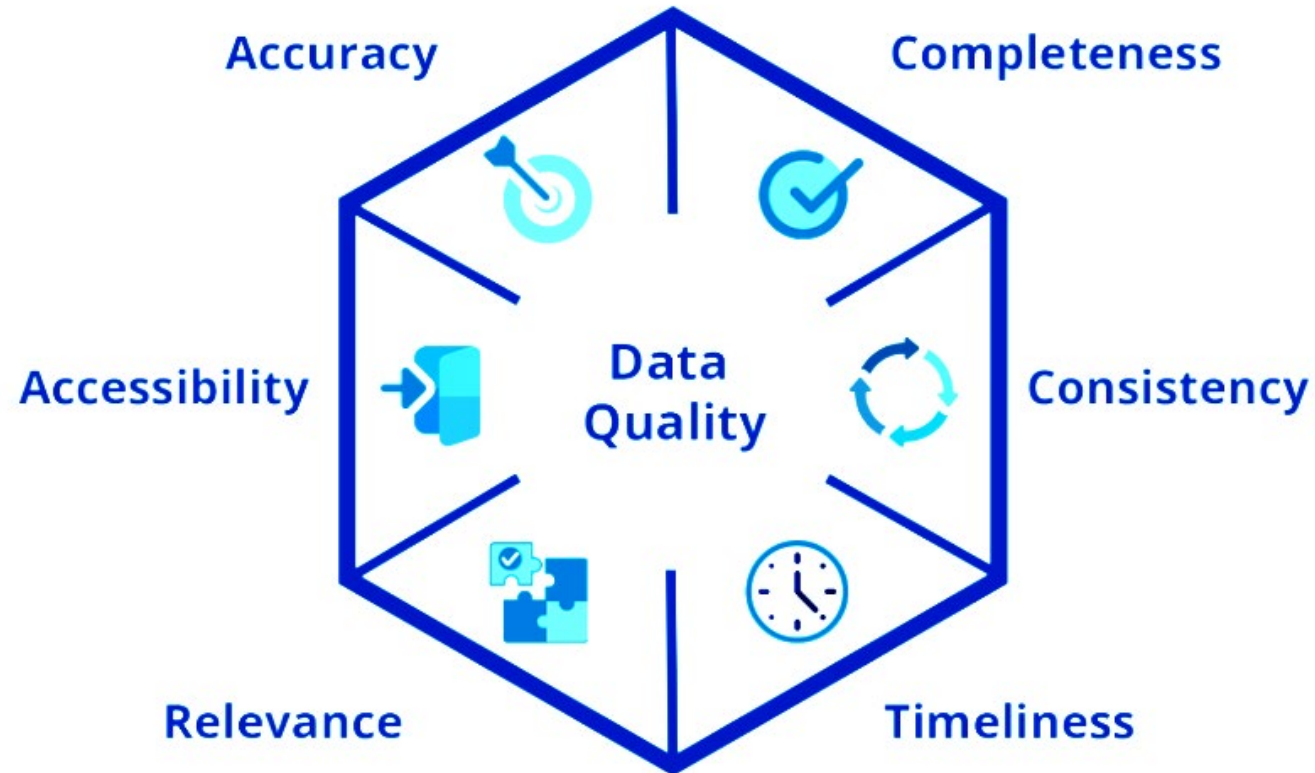


Immortal Time Bias



Tx effect: HR < 1 BIAS!

Adequate Data



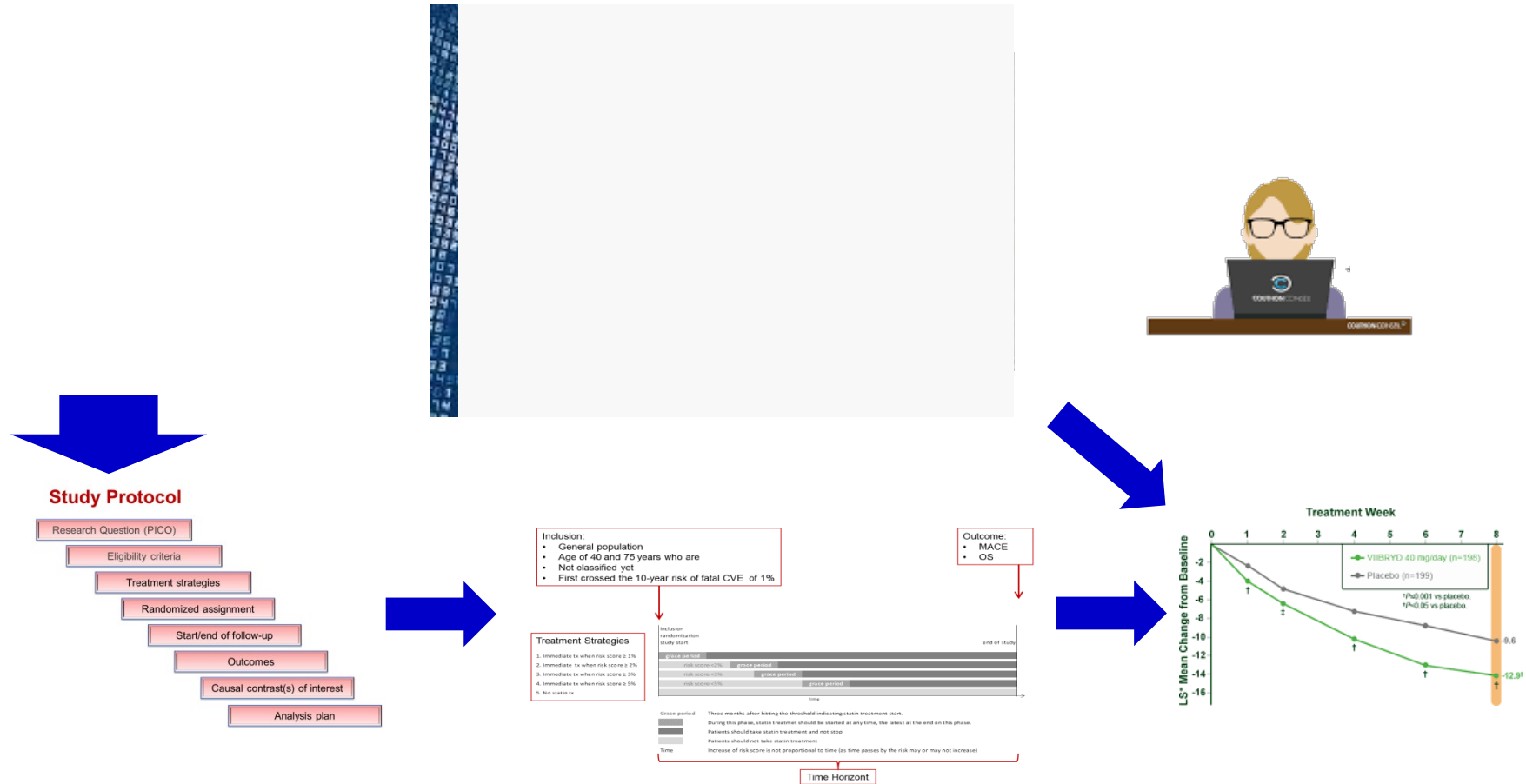
Target Trial Emulation



Target Trial

- Structural approach mimicking the counterfactual approach
 - Develop the [protocol for a hypothetical randomized trial](#) (“target trial”) that would address the research question of interest
- Target Trial can be used to
 - Avoid self-inflicted biases (time-zero bias, immortal time bias)
 - Control for time-dependent confounding
 - Useful for big data analysis
 - When-to-start-treatment questions

Target Trial



Potential biases: Case example

Asses Potential Biases in RWD: Compare traditional and causal methods



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**Journal of
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ORIGINAL ARTICLE

Causal analyses with target trial emulation for real-world evidence removed large self-inflicted biases: systematic bias assessment of ovarian cancer treatment effectiveness

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Aim

Assess potential biases in causal analysis of large real-world database

Type

Impact

Case of ovarian cancer

Aggressive disease with poor survival outcomes

Treatment

Surgery followed by chemotherapy

No standards for subsequent treatment lines



PICO

Population:

Ovarian cancer patients progressed after 1st-line treatment

Intervention:

With 2nd-line treatment (LOT2)

Comparison:

Without 2nd-line treatment

Outcome:

Overall survival (OS)

Methods

Retrospective observational study

IMS Oncology electronic medical records (EMR)

Oncology practices and comprehensive cancer centers in US

Causal graphs

Identify potential confounding and other biases

Estimate direction of biases

Analyses

Completely Crude

Traditional Baseline Adjustment

Causal Methods

Effect measure:

Hazard ratios (HR)

95% confidence intervals (95%CI)

Methods

Compare to Reference Case

Early versus delayed treatment of relapsed ovarian cancer (MRC OV05/EORTC 55955): a randomised trial

Gordon J S Rustin, Maria E L van der Burg, Clare L Griffin, David Guthrie, Alan Lamont, Gordon C Jayson, Gunnar Kristensen, César Mediola, Corneel Coens, Wendi Qian, Mahesh K B Parmar, Ann Marie Swart, for the MRC OV05 and EORTC 55955 investigators*

Summary

Background Serum CA125 concentration often rises several months before clinical or symptomatic relapse in women with ovarian cancer. In the MRC OV05/EORTC 55955 collaborative trial, we aimed to establish the benefits of early treatment on the basis of increased CA125 concentrations compared with delayed treatment on the basis of clinical recurrence.

Methods Women with ovarian cancer in complete remission after first-line platinum-based chemotherapy and a

	Hazard ratio (95% CI)	Log-rank p value
Unadjusted	0.98 (0.80–1.20)	0.85
Adjusted		
For stratification factors*	0.99 (0.80–1.22)	..
For prognostic factors†	0.98 (0.79–1.21)	..
For stratification and prognostic factors	1.01 (0.82–1.25)	..
Sensitivity analyses‡	1.01 (0.82–1.23)	0.96

*Age, International Federation of Gynecology and Obstetrics stage, first-line chemotherapy, time from completion of first-line chemotherapy to doubling of CA125 concentration, and country. †Histology, WHO performance status, and time from doubling of CA125 concentration to randomisation. ‡Sensitivity analyses of non-curtailed data (all follow-up data received, not curtailed at 5 years for MRC OV05 and 3 years for EORTC 55955).

Table 3: Hazard ratios for overall survival

Rustin GJ, et al. Early versus delayed treatment of relapsed ovarian cancer (MRC OV05/EORTC 55955): a randomised trial. Lancet. 2010 Oct 2; 376(9747):1155-63. doi: 10.1016/S0140-6736(10)61268-8. PMID: 20888993.

Analytic Strategies

Analytic Strategy		Intervention	Comparator	Controlling for Bias				
				Confounding Baseline	Time-dep	Unmeasured	Immort Time	Selection
Crude <								

Treatment Assignment

Treatment strategies:

Second line treatment (T)

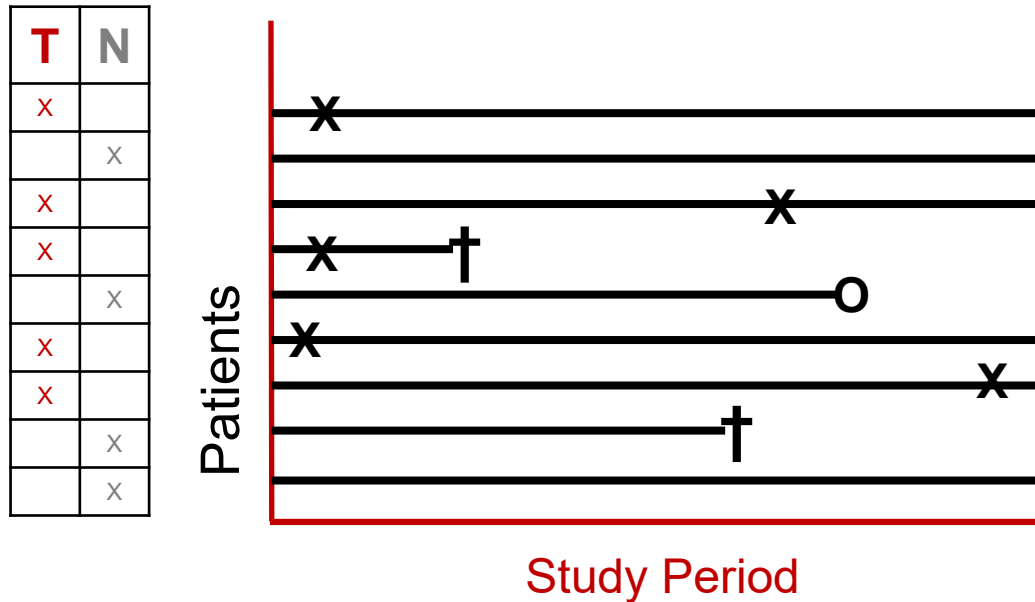
Never treat (N)

X Start of treatment

† Death

O Censored

Progression



Treatment Assignment

Treatment strategies:

Second line treatment (T)

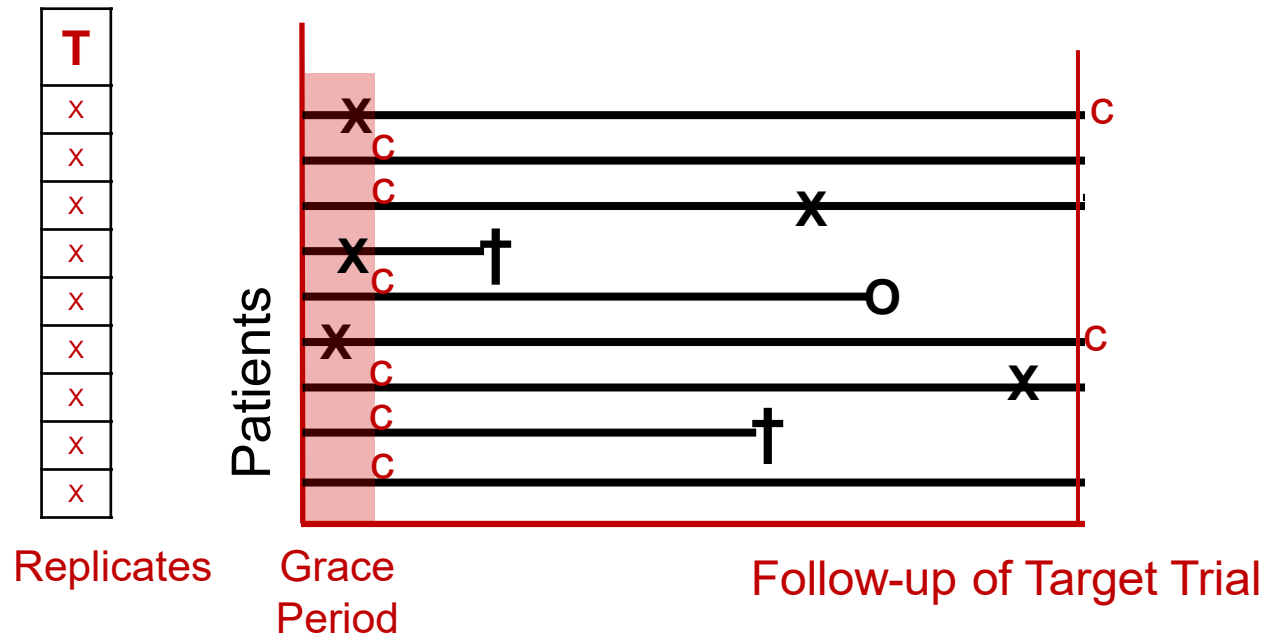
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Treatment Assignment

Treatment strategies:

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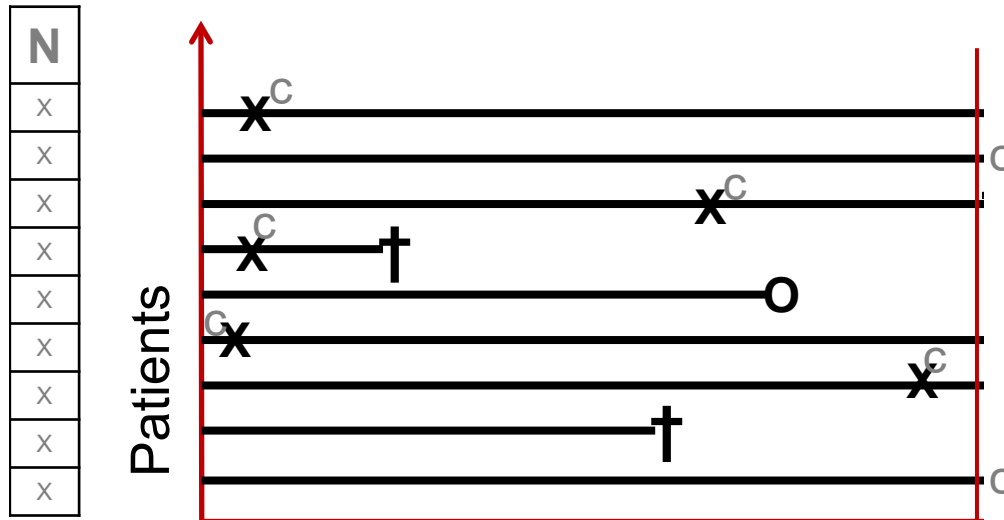
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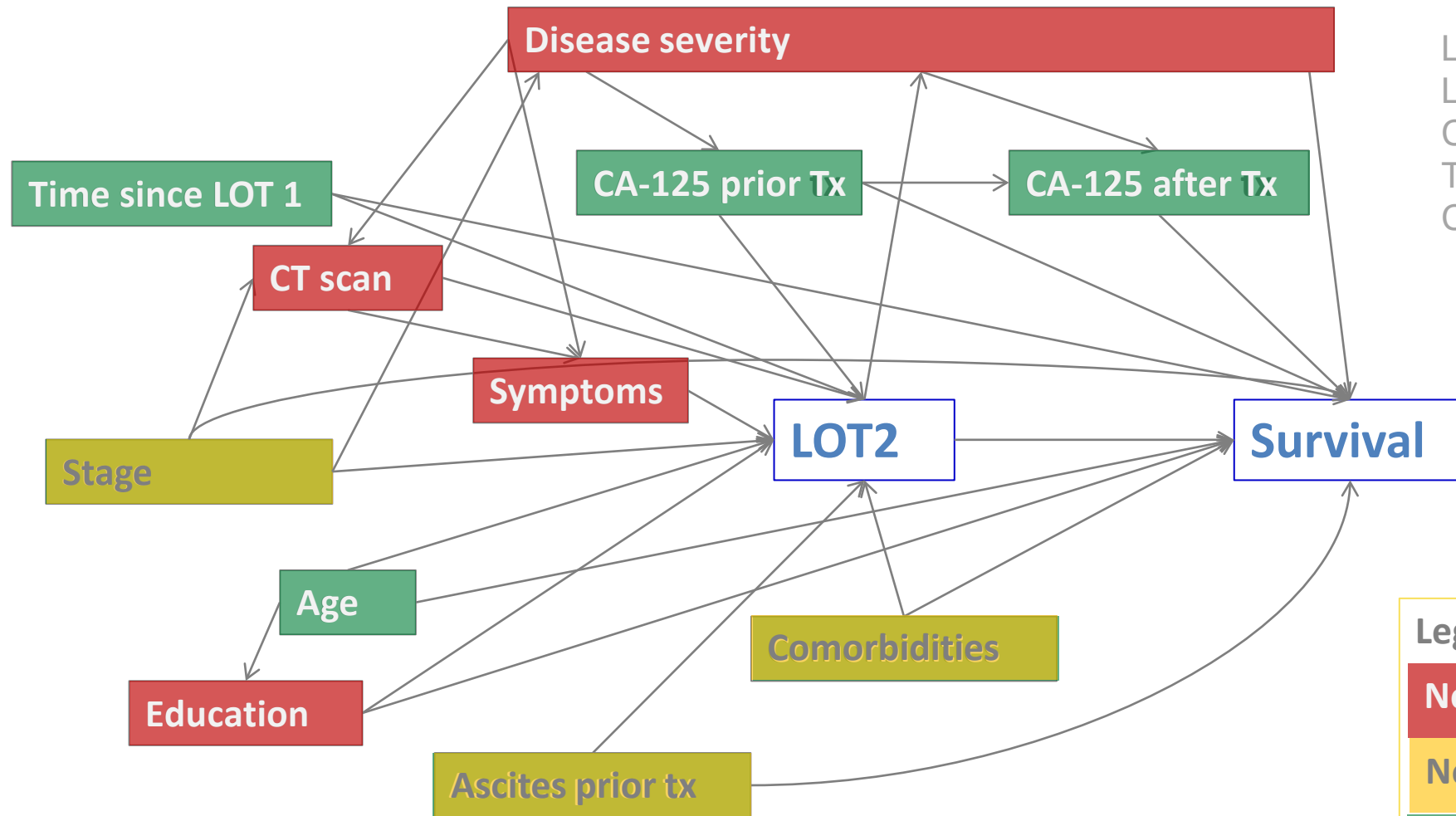


Replicates

Follow-up of Target Trial

**Censoring needs
proper adjustment**

Directed Acyclic Graph (DAG)



LOT1: 1st Line of Treatment;
LOT2: 2nd Line of Treatment;
CA: Cancer-Antigen;
Tx: Treatment;
CT: Computer Tomography;

Legend:

Not available

Not reliable

Available and reliable

Expert Panel Assessment of Assumed Bias Direction

Immortal time bias
Underestimates the treatment HR

Confounding
Estimation of bias direction using the DAG and techniques described by VanderWeele

Bias	Direction of Bias (in favor/against LOT2) Estimation of HR		
	HR - in favor of LOT2	HR ± either	HR + against LOT2
Confounding			
- Unmeasured (disease severity, CT scan, symptoms)			X
(education)	X		
- Time-independent (ascites, stage)			X
(age, comorbidities, time since LOT1)		X	
- Time-dependent (CA-125)			X
Immortal-time Bias	X		
Selection Bias / Confounding by indication			X

HR: Hazard Ratio

HR -: underestimation of HR

HR ±: either under- or overestimation of HR

HR +: overestimation of HR

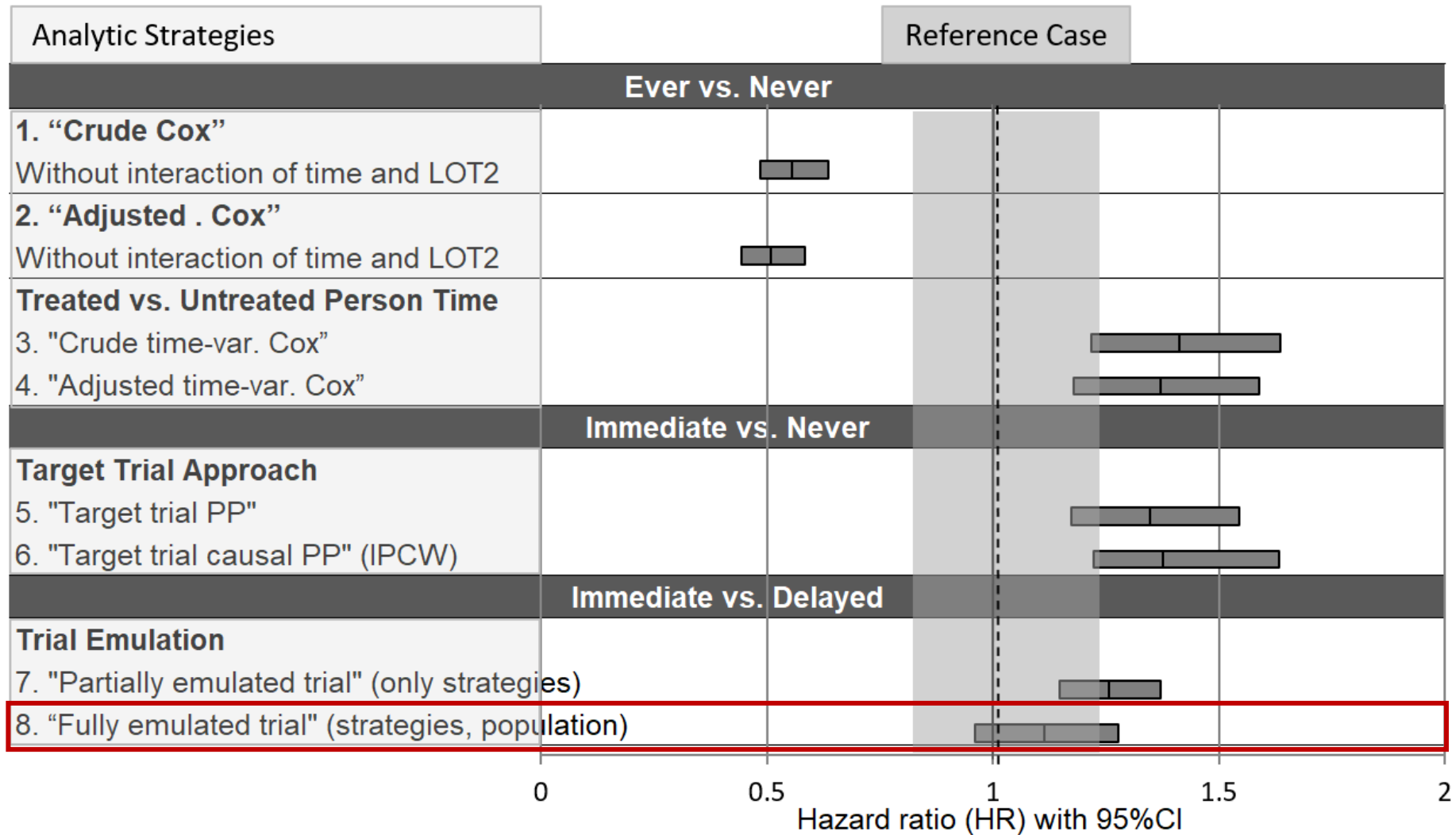
CA: Cancer-Antigen;

CT: computer tomography

LOT1: first line of treatment

LOT2: second line of treatment

Results



LOT2: 2nd Line of treatment;
 vs: versus;
 time-var: time-varying;
 PP: per protocol;
 IPCW: Inverse Probability of
 Censoring Weighting;
 CI: Confidence Interval;

Main Findings



Potential for several biases

May go in different directions



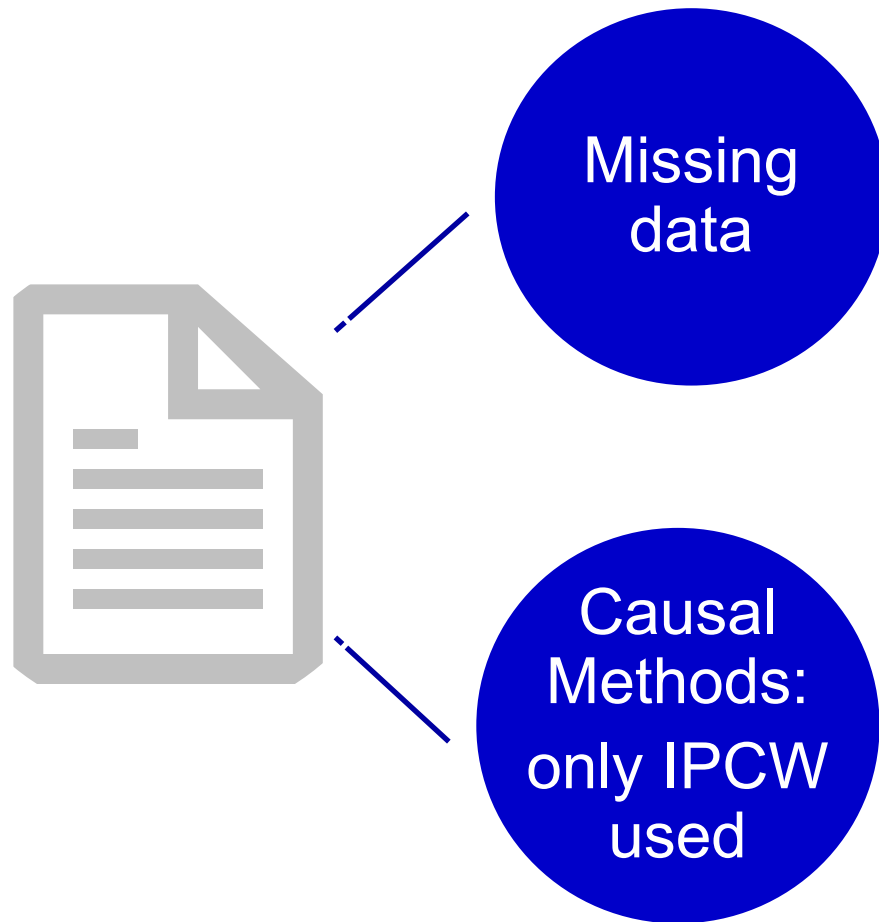
Several techniques exist to avoid biases

Visual: DAG to identify confounding
Structural: target trial approach
Statistical: careful consideration



Results were similar to those of the clinical trial when following the full causal approach

Limitations of this Study



Present in study

Comorbidities

TC scans

Symptoms

Could show in DAGs that

Missing data (in this study) → overestimation of HR

Other g-methods that could be considered

g-estimation,

Rank preserving structural failure time model (RPSFT),

Targeted likelihood estimation

Conclusion of this Study



Study design and methods have a major impact on the validity of the results



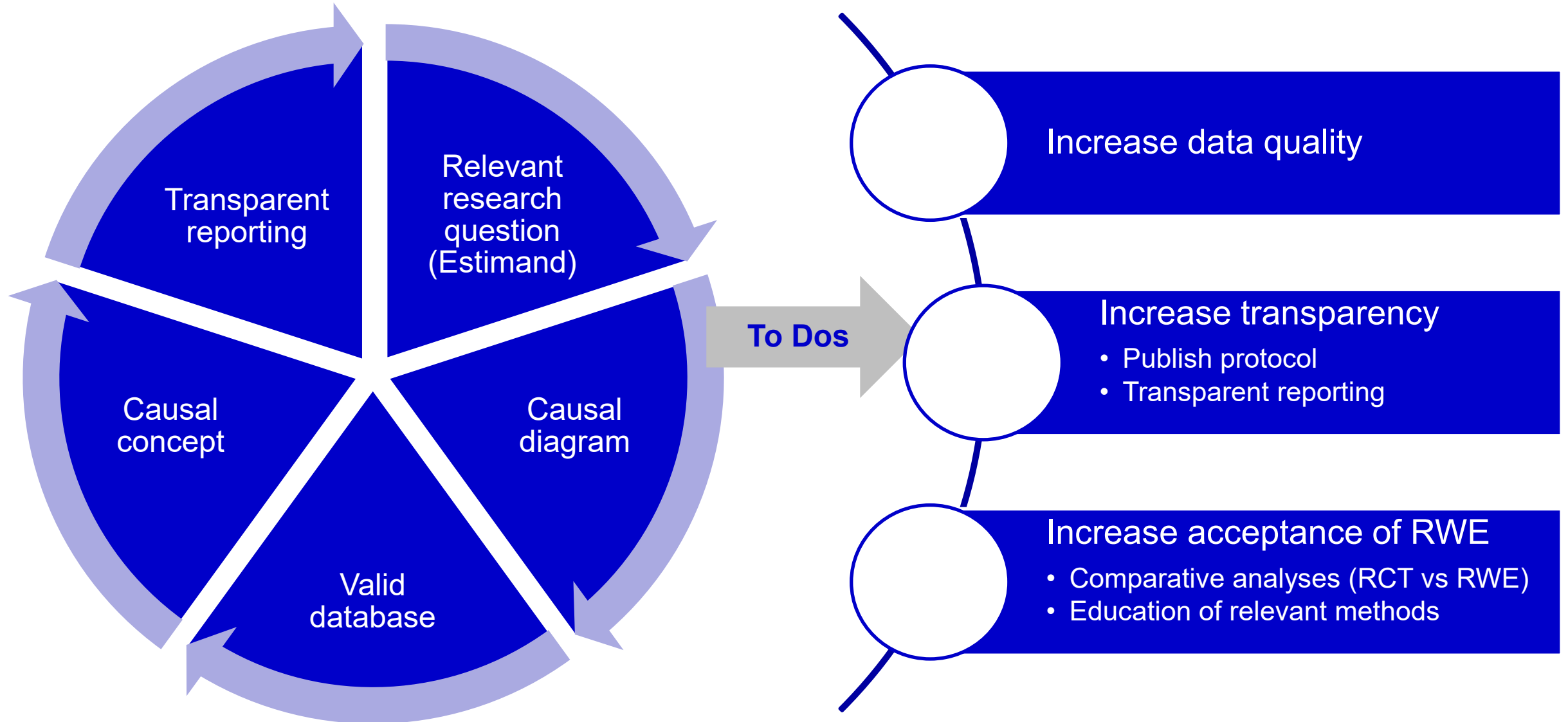
Target trial emulation is a structural approach mimicking the counterfactual approach

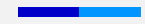


Target trial emulation may avoid self-inflicted biases

So what?

Observational Data Analysis





Thank You

