



## Analysing observational datasets with treatment switches or non-adherence – emulating a Target Trial

ISPOR Barcelona, 2018

Dr Nicholas Latimer ([n.latimer@sheffield.ac.uk](mailto:n.latimer@sheffield.ac.uk))

Reader in Health Economics, School of Health and Related Research, University of Sheffield, UK

Nicholas Latimer is funded by a National Institute for Health Research Post-doctoral Fellowship. These slides presents independent research funded by the National Institute for Health Research (NIHR). The views expressed are those of the author(s) and not necessarily those of the NHS, the NIHR or the Department of Health.



### Analysing observational datasets

- Ideally we use RCTs to estimate comparative effectiveness
- If we can't run an RCT, we can try to emulate one using observational data
- **“Target Trial”**: Framework for analysing observational data recently introduced in the epidemiological literature ([Hernan MA, Robins JM. American Journal of Epidemiology;183\(8\) 2016](#))
- **Who is familiar with this approach? Has anyone ever conducted a Target Trial analysis?**



## The Target Trial approach (Hernan & Robins, 2016)

- There are 7 key components of the target trial protocol:
  1. Eligibility criteria
  2. Treatment strategies
  3. Assignment procedures
  4. Follow-up period
  5. Outcome
  6. Causal contrasts of interest
  7. Analysis plan

3



## The Target Trial approach (Hernan & Robins, 2016)

- There are 7 key components of the target trial protocol:
  1. Eligibility criteria
    - Only include people that would satisfy the eligibility criteria of our Target Trial
    - Likely to need data on things like line of therapy, performance status
  2. Treatment strategies
    - Likely to only be able to emulate a pragmatic trial
    - Assign eligible people to the 'trial' treatment strategy that is consistent with their baseline data
    - "New user" design, to avoid bias associated with selection of individuals who meet eligibility criteria after initiation of a treatment

4



## The Target Trial approach (Hernan & Robins, 2016)

- There are 7 key components of the target trial protocol:

### 3. Assignment procedures

- Cannot emulate blinding
- To emulate random assignment at baseline, must adjust for confounding factors to assure comparability of treatment groups
- Standard regression adjustment sufficient to adjust for baseline confounders
- **Can only successfully emulate trial if no unmeasured confounding at baseline**

5



## The Target Trial approach (Hernan & Robins, 2016)

- There are 7 key components of the target trial protocol:

### 4. Follow-up period

- Need to define time zero (baseline) at which eligibility criteria are met and after which outcomes begin to be counted
- Usually when an eligible individual starts a treatment strategy

### 5. Outcome

- Use the dataset to identify if and when the outcome occurs
- Can't emulate systematic and blind outcome ascertainment. Death is one of the least problematic outcomes to analyse

6



## The Target Trial approach (Hernan & Robins, 2016)

- There are 7 key components of the target trial protocol:

### 6. Causal contrast of interest

- ITT analysis is awkward in an observational data – “initiating” ≠ “assignment”
- Usually estimate per-protocol (PP) effect (effect if treatment strategy defined was adhered to)
- Note, the causal contrast of interest will depend upon the **estimand** of interest

7



## The Target Trial approach (Hernan & Robins, 2016)

- There are 7 key components of the target trial protocol:

### 7. Analysis plan

- Time-dependent confounding is important in observational analyses when:
  - Treatment switching occurs
  - Adherence changes over time
  - People are censored when they stop adhering to a defined treatment strategy (i.e. PP analysis)

#### Example: metastatic disease variable ( $m$ )

Treatment decreases  $pr(m)$   
 $m$  increases  $pr(\text{switch})$   
 $m$  is prognostic for survival

- Can't include  $m$  as time-dependent variable in standard regression because part of treatment effect is through  $m$
- Can't not adjust for  $m$ , due to confounding by indication
- Need advanced techniques, like inverse probability weighting or g-estimation



## Practical applications

- Harvard team working with SEER-Medicare data to demonstrate best practices for comparative effectiveness with observational data  
(Petito LC, Garcia-de-Albeniz X, Hernan MA. Assessing comparative effectiveness of cancer treatments in the SEER-Medicare linked database, StatFest 2018)
- First task: demonstrate whether approach “works” (given data available)  
→ Conduct analyses in an area where an RCT exists and compare results
- First case study: adjuvant chemotherapy for Stage II Colorectal Cancer.
  - Target Trial analysis replicated findings from RCT, though precision unstable
  - SEER-Medicare contains lots of information on confounders, but authors concluded some unmeasured confounding was likely to remain

9



## Practical applications

- My plan:
  - Assess feasibility of Target Trial approach using the UK Systemic Anti-cancer Therapy (SACT) dataset
  - Particularly valuable if possible, given use of this dataset to resolve uncertainty over effectiveness/cost-effectiveness of drugs placed in the Cancer Drugs Fund (see poster PCN317 on Wednesday morning)

10



## Discussion

- Target Trial approach → neat way of formulating observational analyses
- Used correctly, can allow appropriate adjustments to be made for treatment switches or non-adherence in observational data
- **Data collection is critical.** Can only be successful if good quality data are available on baseline and time-dependent confounding factors
  - **Suggestions of suitable datasets?**
- Target Trial facilitates better observational data analyses
  - **Does not mean RCTs not needed.** But if we're collecting observational data, we should use it to its full potential