

Head-to-head Comparisons using Real-World Data

Design and data source considerations from pilot investigations in CVD

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Speakers







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This is a part of a webinar series





Introduction: Target trial emulation



Randomized clinical trials

- Randomized trials are the standard for comparing effectiveness of interventions
 - Randomization reduces selection bias and confounding
 - Provide <u>causal</u> estimates of effects
- Decision-making and health policy need to be informed by causal knowledge about comparative effectiveness and safety

Treat with drug A, B or C?

Treat with drug B or C if no response to A? Treat now or later?

Which individuals to treat?



Randomized trials are not always feasible

- Possible reasons:
 - Costly
 - Untimely
 - Unethical
 - Impractical
- Trials may be limited in scope
 - E.g. eligibility criteria, outcomes measured
 - Real-world generalizability
- Decisions need to be made even in the absence of a randomized trial to address them – maintaining status quo is also a decision



Plan B: Emulate a randomized trial using RWD

- Try to emulate a hypothetical randomized clinical trial using observational ٠ data
- limitations: ٠
 - Can not emulate placebo, blinding, or force adherence ٠
 - Limited by data available (e.g. only approved therapies) ٠



Target trial framework

Step 1: Ask a causal question

 Equivalent to specifying the protocol of the analogous randomized trial explicitly

Step 2: Answer the causal question

- Identify a suitable RWD source
- Try and emulate the randomized trial using RWD
- Usually, it is necessary to cycle between steps 1 and 2 iteratively to tune them due to practical limitations, e.g. by restricting eligibility criteria



Step 1

Step 2





Problems with naïve analysis of RWD

- Asking the "wrong" or a poorly-defined causal question
 - Explicit representation of the causal question as a target trial prevents this
- Answering the causal question incorrectly
 - Not accounting for various biases that may arise
 - Not adjusting for relevant confounders/adjusting where we shouldn't
 - Not using the appropriate methods
- Result: biased effect estimates, wrong conclusions



Example: post-menopausal hormone therapy and CHD

• Effect of post-menopausal hormone therapy on risk for coronary heart disease?





Use cases for target trial emulation

- Generation of efficacy or safety evidence for conditional regulatory approval or post-market assessment
- Providing a comparison when network meta-analysis is not possible
- Expanding the scope (e.g. eligibility criteria) of a randomized trial
- Refining aspects of an existing treatment protocol
- Enabling a comparison to identify optimal treatment regimen



Pilot project: Head-to-head comparison of second-line therapies for type 2 diabetes

Considerations for protocol design and data source



Overview

- Patients with type 2 diabetes on metformin often fail to achieve/maintain glycemic control and require second-line therapy
- >6 classes of anti-diabetic drugs have been approved for use
 - A head-to-head comparison of effectiveness, safety is lacking
 - Choice of second-line agent varies across clinical practices
- There is a need for comparative effectiveness evidence to guide choice of second-line agent
 - To reduce detrimental effects, improve cost-effectiveness
 - Guide personalized therapy



Efficacy is the main driver in prescription practices globally



Gomes, Marilia B., et al. "Treatment of type 2 diabetes mellitus worldwide: baseline patient characteristics in the global DISCOVER study." *Diabetes research and clinical practice* 151 (2019): 20-32.



Step 1 – asking the causal question

- Framing the causal question explicitly in the form of a target trial protocol
- Considerations for protocol design:
 - Who needs to be involved?
 - What are the roles for those involved?
 - How do you define the data/variables necessary for emulation?
 - How do you refine the protocol given practical limitations (e.g. data)?





GRADE trial

- Ongoing open-label randomized trial for comparative effectiveness of second-line therapy in type 2 diabetes
 - Started in 2013
 - Est. primary completion in July 2021
- Outcomes
 - HbA1C (primary)
 - Cardiovascular, microvascular side effects, adherence-tolerability
- No SGLT-2 inhibitor arm



Nathan, David M., et al. "Rationale and design of the glycemia reduction approaches in diabetes: a comparative effectiveness study (GRADE)." *Diabetes care* 36.8 (2013): 2254-2261.



Data considerations for target trial emulation

- Data from combination of EMR/EHR, claims data, clinician notes
- Brief checklist:
 - "Big" data sufficiently large sample size, large number of variables
 - Representative sample covered
 - Bias/irregularities in capture of study variables
 - Long follow up
 - Database expertise
 - Access to clinical practitioners familiar with data
 - Data normalization, harmonization, de-identification
 - Cohort generation minimizing bias, false positives/negatives

Hall, Gillian C., et al. "Guidelines for good database selection and use in pharmac oepidemiology research." *Pharmacoepidemiology and drug safety* 21.1 (2012): 1-10.



Partnership with Nashville Biosciences for RWD

- Subsidiary of Vanderbilt University Medical Center (VUMC)
- Leverages a large de-identified EMR, biospecimen repository BioVU®
 - E.g. 163,000 EMRs for type 2 diabetes
- Also provide scientific, clinical expertise



Source of data

 6100+ providers and 68 hospital locations accepting ~2.5 million patients annually





Target trial protocol – checklist

- Eligibility criteria
- Treatment strategies
- Causal contrasts of interest
- Confounders for assignment to treatment strategies
 - Other confounding variables will need to be defined depending on treatment strategies and causal contrasts (e.g. for adherence)
- Outcomes
- Follow-up
 - Start and end time
 - Defining time zero
- Analysis plan (will be discussed in next webinar)



Eligibility criteria

| Criterion | Available in NB data? | Source |
|---|--|---------------------------------------|
| T2DM diagnosis | Yes | ICD, ICD+meds+labs |
| Age (>45 years) | Yes | DOB |
| Antihyperglycemic monoTx with metformin for >3 months in 2014/later | Yes | Rx records, self-reported drug use |
| Suboptimal glycemic control after metformin initiation (>7% HbA1c) | Yes | Labs |
| No suspected T1D or other reasons for hyperglycemia (rare disorder, pancreatic surgery) | Yes (?) – depends on start of care at VUMC relative to treatment start date | |
| Not pregnant | Yes | ICD, CPT |



Treatment strategies

- Initiation of second-line therapy within 12 months of persistent inadequate HbA1c control with metformin use from one of:
 - GLP-1 RA
 - SGLT-2 inhibitor
 - DPP-4
 - Sulfonylureas
 - (TZD, insulin)
- Other strategies depend on data:
 - Sustained treatment with, addition of a third-line therapy
- Limitations:
 - Limited to drugs prescribed at/dispensed at VUMC pharmacy



Causal contrasts

- Intention-to-treat effect
 - Effect of being assigned to treatment strategy at baseline
 - Of interest in the GRADE trial
 - Requires adjustment for non-random allocation at baseline + non-random loss to follow-up
- Per-protocol effect
 - Effect of adhering to assigned treatment strategy
 - Also requires adjustment for non-adherence (pre + post-baseline factors, e.g. treatment discontinuation)
 - May be feasible depending on availability of sufficient relevant data on adherence



Confounders for treatment assignment at baseline*

| Predictor | Available in NB data? | Source/comments |
|---|---|--|
| Type of medical insurance | Yes | For most |
| History of CVD | Yes | |
| Physician comfort level/perception of drug | Unknown | May need help of treating physician or their Rx histories |
| HbA1c levels | Yes | Labs |
| Contraindications | Yes | Rx records |
| BMI | Yes | |
| Frailty/tolerability | (?) – Depends on how we define this, may involve practitioners | |

*How can we tell if we have adjusted sufficiently, appropriately?



Outcomes

- Metabolic
 - HbA1C levels (may not be a patient-important outcome)
 - BMI
- Macrovascular
 - MACE
 - Heart failure
- Microvascular
 - Glomerular filtration rate (eGFR)
 - Retinopathy

Gandhi, Gunjan Y., et al. "Patient-important outcomes in registered diabetes trials." Jama 299.21 (2008): 2543-2549.



Follow up

- Defining time zero
 - Time of treatment prescription
- Patients are followed from time zero until diagnosis of outcome or loss to follow-up due to various factors
 - E.g. due to discontinuation of pharmacy benefits, medical insurance etc.
 - Ideally, minimize loss to follow-up
 - Possible ways include only including patients with a history of regular visits



Next steps

- Cohort extraction and sample size estimation
- Iterative tuning of protocol given data limitations
- Definition of final protocol and analysis plan
- Trial emulation
 - Results survival curves, relative risk estimates
 - Subgroup analyses
- Sensitivity analysis
- Benchmarking against results from GRADE if possible



Conclusion

- In the absence of a randomized clinical trial, we can try to emulate a hypothetical randomized trial using RWD
 - Can also complement existing trials
 - Fraction of time and cost
- Considerations discussed today:
 - Trial protocol design
 - Suitable high-quality RWD for emulation
- Trial design may need to be refined iteratively due to practical constraints
- Multiple parties are needed to successfully define protocol and interact with big data (e.g. for data validation, database expertise)



Next webinar







