# Sequential Rematching Randomization and Adaptive Monitoring with the Second-Generation p-value to increase the efficiency and efficacy of Randomized Clinical Trials

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### **RCT Costs and Benefits**

### 2015-2016 Pivotal Trials accepted by FDA?

- ▶ 138 Trials to approve 59 novel therapeutic drugs
- ► Enrolled on median 488 patients (IQR: 230, 740)
- Cost per patient on median \$41K (IQR: \$32K \$82K)

#### **Outline**

Three papers to increase study efficiency and efficacy.

- ▶ Paper 1: Sequential Rematched Randomization
- ▶ Paper 2: Adaptive Monitoring Using the Second Generation p-value
- ▶ Paper 3: Estimation of operating characteristics of AM with SGPV

Motivating Example: REACH Trial

#### **REACH Trial**

### Rapid Education/Encouragement And Communications for Health

Population: Adults with Type 2 Diabetes (DM)

Purpose: Increase glycemic control and adherence to medications

Main Intervention: Text message-delivered diabetes support for 12 months

Outcome: Glycemic control (A1c) compared to control

Multi-site enrollment: 512 patients from Vanderbilt and Non-Vanderbilt Clinics

Clinical Relevance: Change clinical practice – Decrease in A1c of 0.5 or more



Did you take all of your diabetes meds today, Sun, 09/05? Please reply Y or N.



### **Key Baseline Covariates**

# Biological Factors

- ▶ Baseline A1c\*
- Age at baseline
- Race / Ethnicity
- DM Rx type\*
- ▶ Time since DM dx\*
- \* Post-hoc analysis: Greater association with outcome

### Socio-economic Factors

- Yrs of education
- Income level
- Insurance type

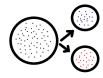
### Stratified Block Randomization

- ▶ Limited in number of covariates creating strata
- ▶ 4000+ Strata Stratifying on all covariates\*
- ▶ 2 \* 3 \* 3 \* 2 = 36 Strata Stratifying on:
  - 2 Sites
  - ▶ 3 Baseline A1c levels
  - 3 DM Rx Types
  - 2 Time since DM Dx levels

# Matched Randomization [?]



Observational Study: Bipartite (2 group) matching



Randomization Trial: Non-bipartite matching

Refined stratification into pairs of patients using distance matrix

- ▶ Pairs selected for smallest total distance
- Allows for any number of continuous and categorical covariates
- All participants known before randomization

# Matching OTF, some additional notes [?]

# Pre-specification:

- Initial reservoir size
- Threshold to denote degree of similarity

Fixed Threshold: Match is better than [20%] of random matches

- ▶ Mahalanobis Distance of random pairs scales to  $F_{(p,n-p)}$
- Assumes normally distributed baseline covariates

### Reservoir:

- Not required to deplete
- May result in unequal treatment group sizes

### **Extension 1: Dynamic Threshold**

Dynamic percentile of random match distances

- 1. Chance of matching now versus later
- 2. Random match distances (from data, ie not assumed normal)
- 3. Remove threshold at end to ensure all patients match

Patients enroll in b enrollment blocks

 $||U_b||$ : Number of unmatched patients at  $b^{th}$  enrollment

 $||R_b||$ : Number remaining patients after  $b^{th}$  enrollment

$$Q_b = rac{||U_b|| - 1}{||U_b|| + ||R_b|| - 1}$$
 
$$Threshold_b = egin{cases} F_b^{-1}(Q_b) & ||U_b|| < ||R_b|| \ ext{best match(es)} & ||U_b|| \ge ||R_b|| \end{cases}$$

 $F_b$  is estimated by  $\hat{F_b}$  by bootstrap sampling random matches from distance matrix.

### **Extension 2: Sequential Re-Matching**

# Sequential Matching

► Formed matches remain through study

### Sequential Re-Matching

- ▶ Matches are allowed to break if a better match enters
- ▶ However, participants keep original treatment assignment
  - ► New participant: Match to anybody
  - Former enrolled participant: Match to anyone of opposite treatment or new participant

Case study using REACH Data to comparing randomization schemes

Balance of baseline covariates Efficiency estimating treatment effect

Results will vary based on baseline covariates

- Number of patients
- Number of covariates
- Distribution of covariates
- ► Adjusted *R*<sup>2</sup> of covariates

### Simulations to evaluate Sequential Rematching Randomization

Generate 20K MCMC replicates under potential outcomes framework:

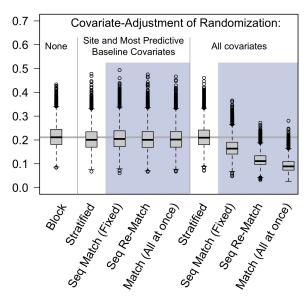
- Bootstrap sample 512 observations from REACH trial
- Outcomes
  - Y(0) = Predicted three month A1c + random residual
  - Y(1) = Y(0) 0.5
- Generate randomization scheme under
  - Block Randomization
  - Stratified Block Randomization
  - Matched Randomization
- Observe
  - Balance: Average Standardized Mean Difference among all covariates
  - Efficiency: CI Width of 1) fully-adjusted linear model and 2)
     Permutation distribution

Where Stratified and Matched randomization adjusts for

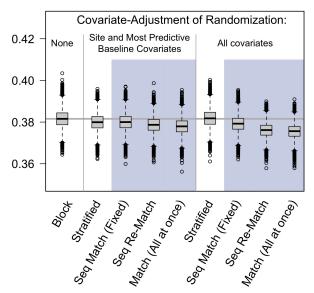
- ▶ Site, Baseline A1c, DM Rx Type, and Time since DM Dx
- All baseline covariates



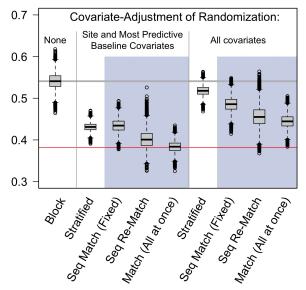
# Maximum Standardized Mean Difference (all baseline covariates)



# Covariate-Adjusted OLS Confidence Interval Width of estimated treatment effect



# Permutation Confidence Interval Width of estimated treatment effect



### Recommendations regarding balance and efficiency

### Conclusions

- ▶ Greatest overall balance by adjusting for all baseline covariates
- Greatest linear gains in efficiency
  - Model efficiency: adjusting randomization to all baseline covariates
  - ▶ Permutation efficiency: adjusting randomization to priority covariates
- Permutation efficiency achieves nearly same efficiency as fully-adjusted model with Block Randomization

### Practical Recommendations

- Recommend adjusting to all baseline covariates
- Adaptively monitor until reaching a clear clinical conclusion

Paper 2: Adaptive Monitoring Using the Second Generation p-value

### Prematurely Ending Clinical Trial(s)

Towards a Revolution in COPD Health (TORCH) (?)

Primary Aim: Establish whether beta-agonist (salmeterol plus fluticasone propionate) has survival benefit in patients with chronic obstructive pulmonary disease

### 2007 6112 patients

- ► HR 0.825 (95% CI: 0.681-1.002, p-adjusted=0.052)
- Awkward Conclusion: primary outcome did not reach statistical significance, yet 'significant benefits in all other outcomes.'

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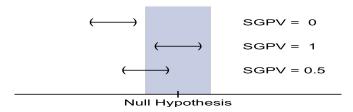
Primary Aim: Establish whether beta-agonist (salmeterol plus fluticasone propionate) has survival benefit in patients with chronic obstructive pulmonary disease

### 2007 6112 patients

- ► HR 0.825 (95% CI: 0.681-0.998, p-adjusted=0.0498)
- Awkward Conclusion: primary outcome did not reach statistical significance, yet 'significant benefits in all other outcomes.'

### Second Generation p-value (SGPV; ?)

What proportion of interval overlaps with trivial effects?

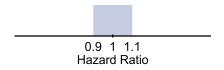


Interpretation of SGPV $_T$  (T for trivial effects)

- ightharpoonup SGPV<sub>T</sub> = 0.0: Evidence to rule out trivial effects
- ▶ SGPV<sub>T</sub> = 1.0: Evidence to rule out non-trivial effects
- ▶ SGPV<sub>T</sub> = 0.5: Inconclusive, need more data

### Second Generation p-value (SGPV; ?)

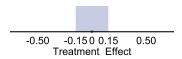
Example of indifferenze zone with TORCH



If HR were 0.825 (95% CI: 0.681-0.899) Interpretation of SGPV  $_T=0$ : Rule out trivial effects

### **REACH**

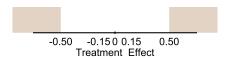
Rapid Education/Encouragement And Communications for Health



A decrease is HbA1c of 0.15 (ex. 7.5 to 7.35) is considered essentially no change

### **REACH**

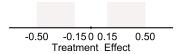
Rapid Education/Encouragement And Communications for Health



A decrease is HbA1c of 0.5 (ex. 7.5 to 7.0) would highly suggest adopting intervention

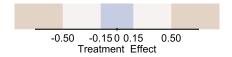
### **REACH**

Rapid Education/Encouragement And Communications for Health



A decrease is HbA1c of between 0.15 and 0.5 is interesting it may suggest adopting intervention

### Rapid Education/Encouragement And Communications for Health



 $\mathsf{SGPV}_T$  to rule out trivial effects  $\mathsf{SGPV}_{HA}$  to rule out highly-actionable effects

 $SGPV_T = 0.0$ : Ruled out trivial effects  $SGPV_{HA} = 0.0$ : Ruled out highly actionable effects

```
Wait interval width stabilizes

Monitor interval and SGPV at desired looks

Alert SGPV_T = 0.0: Ruled out trivial effects

SGPV_{HA} = 0.0: Ruled out highly actionable effects

Affirm stop if same conclusion [40] patients later end of resources (n = 512)
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# Adaptive Monitoring with SGPV

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Monitor interval and SGPV at desired looks

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Report only the final interval when stopping

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# **Adaptive Monitoring with SGPV**

```
Wait interval width stabilizes
```

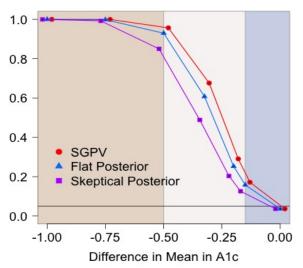
Monitor interval and SGPV at desired looks

Alert SGPV<sub>T</sub> = 0.0: Ruled out trivial effects SGPV<sub>HA</sub> = 0.0: Ruled out meaningful effects

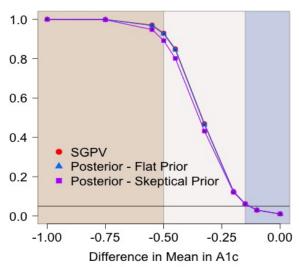
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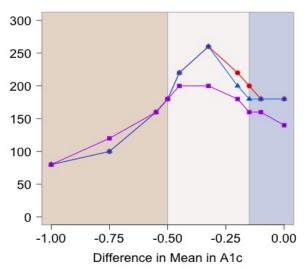




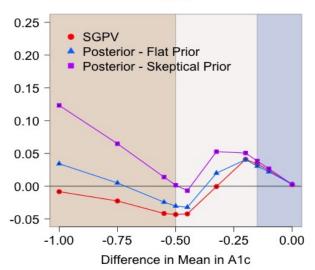




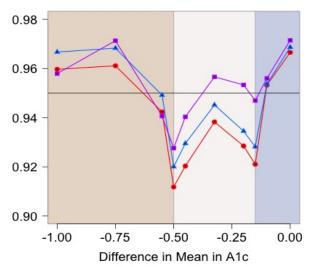












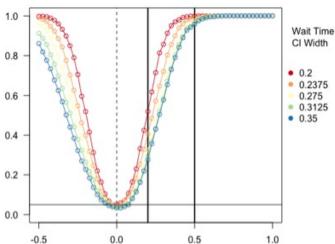
Paper 3: sgpvAM R package and practical guidance to control adaptive monitoring errors

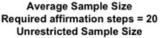
- 1. How to easily operationalize?
- 2. Encourage not changing clinical relevance
- 3. Under fully sequentially monitored study, how long to wait to ensure Type I Error < 0.05?
- 4. What are operating characteristics when observations are not observed immediately? When limited in sample size?

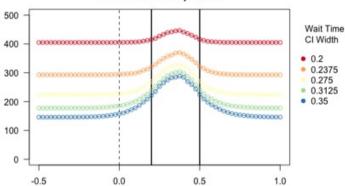
# sgpvAM package

- 1. Uses Valerie's spgv package on github
- 2. Vignette with examples
- Example 1-sided study with at most trivial effects = 0.2 and minimal highly actionable effect of 0.5, lag of 50 observations, will monitor no more frequently than 10 patients.

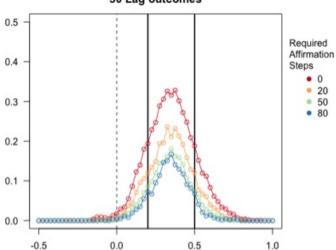
P( reject point null )
Required affirmation steps = 20
Unrestricted Sample Size



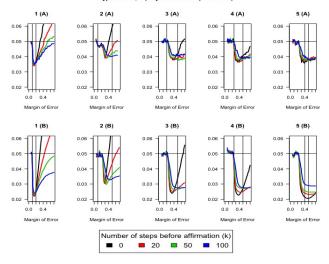




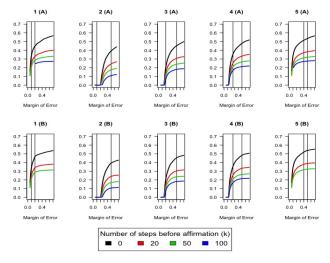
P( inconclusive )
Wait for expected CI width = 0.35
50 Lag outcomes



Type I Error ; P ( Reject Point Null | Theta = 0 )







# References