Sample Size Considerations for the Analysis of Continuous Repeated-Measures Outcomes in Sequential **Multiple-Assignment Randomized Trials**

Dynamic Treatment Regimes

A dynamic treatment regime (DTR) is a sequence of pre-specified decision rules which guides the delivery of an individualized sequence of treatments. This sequence is tailored based on ongoing information about the individual's progress in treatment.



Sequential Multiple-Assignment Randomized Trials

A sequential multiple-assignment randomized trial (SMART) is an experimental design which can provide data that informs the construction of an effective DTR (Murphy, 2005). Some or all participants are randomized more than once. Each randomization corresponds to a critical question regarding the development of a DTR.

The ENGAGE Trial

The ENGAGE study (J. McKay, PI; N = 500) is a SMART aimed at developing a DTR to increase motivation to attend an intensive outpatient treatment program (IOP) among alcoholand cocaine-dependent patients.

Figure 1: Diagram of the ENGAGE SMART. Circled R indicates randomization, boxes indicate treatments. MI-IOP corresponds to two motivational interviews encouraging participation in the IOP; MI-PC, two motivational interviews offering patients a choice of treatment modalities; NFC is no further contact.



• The outcome of interest is **treatment readiness**, a measure of a patient's willingness and ability to commit to active participation in a substance abuse treatment program.

- Treatment readiness was assessed using an 8-item questionnaire scored from 0 to 40 and coded such that higher scores are better. We consider measurements taken at baseline and at weeks 8 and 24.
- There are 4 **embedded DTRs**, indexed by first-stage treatment and second-stage treatment for continued non-engagers.

Table 1: Embedded DTRs in ENGAGE

		Stage 2 Treatment	
(a_1,a_2)	Stage 1 Treatment	Engagers	Ctd. Non-Engage

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Marginal Mean Model

We are interested in $E\left[Y_t^{(a_1,a_2)} \mid \mathbf{X}\right]$, the marginal mean of $\mathbf{Y}^{(a_1,a_2)}$ at time t under DTR (a_1,a_2) conditional on baseline covariates \boldsymbol{X}

• We impose a modeling assumption:

 $E[Y_t^{(a_1,a_2)} \mid \boldsymbol{X}] = \mu_t^{(a_1,a_2)}(\boldsymbol{X};\boldsymbol{\theta}),$

- $oldsymbol{ heta} = (oldsymbol{\eta},oldsymbol{\gamma})_{\cdot}$
- $\mu_t^{(a_1,a_2)}(\boldsymbol{X};\boldsymbol{\theta})$ should account for the design of the SMART.
- An example model for ENGAGE is

$$\mathcal{H}_t^{(a_1,a_2)}(oldsymbol{X};oldsymbol{ heta}) = oldsymbol{\eta}^+oldsymbol{X} + \gamma_0 \ + \mathbbm{1}$$
 for each (24 t + 24 t

$$+ \gamma_4(t-8)a_1 + \gamma_5(t-8)a_1 + \gamma_5(t-8)a_1$$

Figure 2: Plot of treatment readiness vs. time using data from ENGAGE.



Estimation of Model Parameters

The estimate $\hat{\theta}$ of θ is the solution to the following the estimating equations:

Estimating Equations

$$oldsymbol{0} = rac{1}{n} \sum_{i=1}^{n} \sum_{(a_1,a_2)} \left[W^{(a_1,a_2)} \left(A_{1,i}, R_i, A_{2,i}
ight)
ight.
onumber \ \cdot oldsymbol{D}^{(a_1,a_2)} (oldsymbol{X}_i)^{ op} oldsymbol{V}^{(a_1,a_2)} (oldsymbol{X}_i)^{-1} \left(oldsymbol{Y}_i - oldsymbol{\mu}^{(a_1,a_2)} (oldsymbol{X}_i;oldsymbol{ heta})
ight],$$

where

- (a_1, a_2) specifies an embedded DTR,
- $W^{(a_1,a_2)}(A_{1,i},R_i,A_{2,i}) = 2 \cdot \mathbb{1} \{A_{1,i} = a_1\} (R_i + 2(1))$
- $\boldsymbol{D}^{(a_1,a_2)}(\boldsymbol{X}_i) = rac{\partial}{\partial \boldsymbol{\theta}} \boldsymbol{\mu}^{(a_1,a_2)}(\boldsymbol{X}_i; \boldsymbol{\theta})$
- $V^{(a_1,a_2)}(\boldsymbol{X}_i)$ is a working model for $Var\left(\boldsymbol{Y}^{(a_1,a_2)} \boldsymbol{\mu}^{(a_1,a_2)}(\boldsymbol{X}_i;\boldsymbol{\beta}) \mid \boldsymbol{X}_i\right)$

Assuming that $\mu^{(a_1,a_2)}(X_i;\theta)$ is correctly specified, $\hat{\theta}$ is consistent for the true parameter value, regardless of the choice of $V^{(a_1,a_2)}(X_i)$ (Lu et al., 2016).





6 Time 2

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where $\mu_t^{(a_1,a_2)}(\boldsymbol{X};\boldsymbol{\theta})$ is a marginal structural mean model with unknown parameters

 a_1t) + $\mathbb{1}_{\{t>8\}}$ ($8\gamma_1 + 8\gamma_2a_1 + \gamma_3(t-8)$) $a_2 + \gamma_6(t-8)a_1a_2$, t = 0, 8, 24

$$-R_i)\mathbb{1}\left\{A_{2,i}=a_2\right\}\right)$$

We developed a sample size formula for a SMART with a continuous repeated-measures outcome in which the primary aim is to compare two embedded DTRs (with different first-stage treatments) on the end-of-study measurement. To compare DTRs (1, 1) and (-1, 1), we size the trial based on a Wald test: $H_0: 16\gamma_2 + 32\gamma_4 + 32\gamma_6 = 0$ vs. $H_1: 16\gamma_2 + 32\gamma_4 + 32\gamma_6 \neq 0$.

We assume:

- $P(R = 1 \mid A_1 = 1) = P(R = 1 \mid A_1 = -1) = r$
- exchangeable correlation matrix with correlation ρ .

Suppose we want to detect a standardized effect size δ . The sample size for the SMART is



Below is a selection of minimum-required sample sizes for comparing two embedded DTRs in an ENGAGE-type SMART which start with different treatments. Sample sizes are based on a comparison of an end-of-study outcome, and vary with minimum-detectable standardized effect size and within-person correlation among the repeated measures.

Table 2: Example sample sizes for comparison of two embedded DTRs. r = 0.4, $\alpha = 0.05$ (two-sided), and $\beta = 0.2$.

	Within-Person Correlation			
Std. Effect Size	$\rho = 0$	$\rho = 0.3$	$\rho = 0.6$	
$\delta = 0.3$	559	508	358	
$\delta = 0.5$	201	183	129	
$\delta = 0.8$	79	72	51	

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Sample Size

1. The probability of response is the same for both first-stage treatments:

2. The variance of $(\mathbf{Y}^{(a_1,a_2)} - \boldsymbol{\mu}^{(a_1,a_2)}(\mathbf{X};\boldsymbol{\theta}))$ is unconditional on response:

 $\operatorname{Var}\left(\boldsymbol{Y}^{(a_{1},a_{2})} - \boldsymbol{\mu}^{(a_{1},a_{2})} \mid R = 1\right) = \operatorname{Var}\left(\boldsymbol{Y}^{(a_{1},a_{2})} - \boldsymbol{\mu}^{(a_{1},a_{2})} \mid R = 0\right)$

3. The true covariance structure of $(\mathbf{Y}^{(a_1,a_2)} - \boldsymbol{\mu}^{(a_1,a_2)}(\mathbf{X};\boldsymbol{\theta}))$ is $\sigma^2 \mathbf{R}(\rho)$, where $\mathbf{R}(\rho)$ is an

Sample Size Formula

$$\frac{z_{1-\alpha/2}+z_{1-\beta}}{\delta^2} \cdot 2\left(2-r\right) \cdot \left(1-\rho^2\right)$$

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References

