



MRC  
Biostatistics  
Unit



UNIVERSITY OF  
CAMBRIDGE

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## **Short course on Response-Adaptive Methods for Clinical Trials**

*Lecture 4: Applications of RAR: Backfill in  
dose finding and RAR Repository*

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# Outline

## 1. Implementing and assessing Bayesian response-adaptive randomisation for backfilling in dose-finding trials

Introduction

Method

Simulations

Discussion

## 2. The RA-ClinicalTrials Repository

Motivation

The Repository

Analysis

Call to Action

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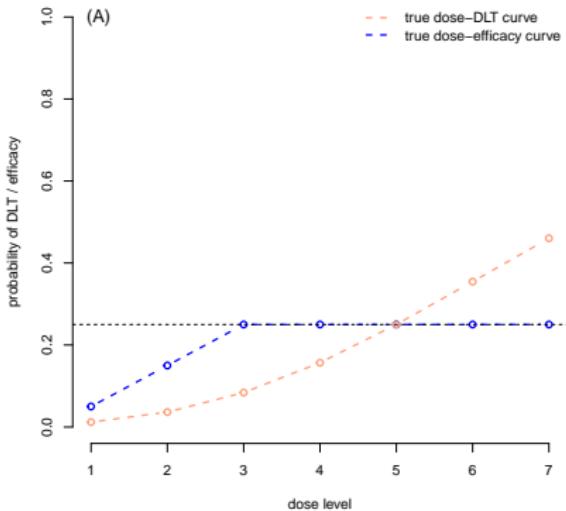
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# Motivation

- Phase-II Dose-Finding Study
- Efficacy Plateau
  - ▶ in oncology with targeted agents or immunotherapy
- **Identify** the lowest dose level that maximises efficacy whilst remaining tolerable and **allocating** patients close to that dose level
  - ▶ Backfill + RAR



# Previous Work

- Continual Reassessment Method (CRM) to identify *Maximum Tolerated Dose (MTD)* [O'Quigley et al (1990)]
- Backfill with Equal Randomization (ER) under MTD to identify *Recommended Phase 2 Dose (RP2D)* [Dehbi et al (2021)]
- Backfill with Response Adaptive Randomization (RAR) under MTD to identify RP2D and allocated patients better [Pin et al (2024)]

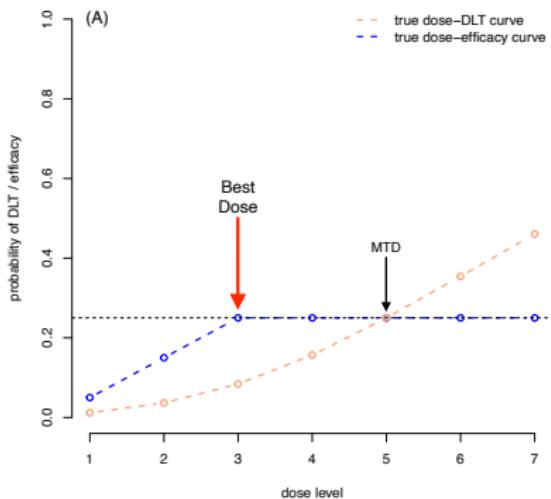


Figure: Dose 3 (RP2D) < Dose 5 (MTD)

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# Notation

- Multi-armed trial: 1,...,K different dose levels
- Fix number of patients in the trial:  $n = \sum_{k=1}^K n_k$ 
  - ▶  $n_k$  is the number of patients on dose level k
- Potential efficacy outcomes:  $Y_{ki} \stackrel{\text{iid}}{\sim} Bern(p_k)$ 
  - ▶  $p_k$  is success probability
- Binary indicator  $a_{ki}$  s.t.  $\sum_{k=1}^K a_{ki} = 1 \forall i$
- Response-adaptive-randomization:  $P(a_{ki} = 1 | a^{(i-1)}, Y^{(i-1)})$

# How does Backfill work?

**Algorithm** backfill( $n = 57, c = 3$ )

$MTD \leftarrow 1$

**while**  $MTD = 1$  **do**

    allocate  $c$  patients to dose level 1

$MTD \leftarrow$  update through CRM

**end while**

**while** Number of allocated patients  $< n$   
**do**

    allocate  $c$  patients to MTD

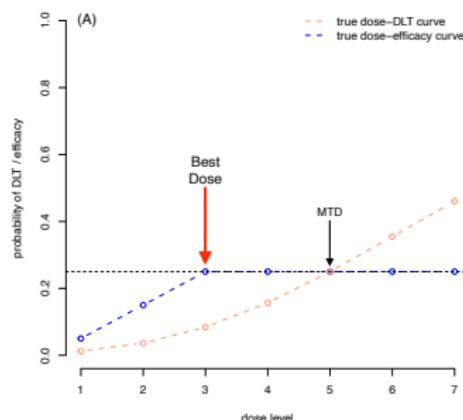
    randomise  $c$  backfill-patients with ER or RAR to dose levels 1, ...,  $MTD - 1$

$MTD \leftarrow$  update through CRM

**end while**

Choose Model with(out) Plateau

**return** RP2D



# ER vs. RAR

- **Equal randomisation (ER)** allocated backfill patients with equal probability ( $1/(\text{size of backfill set})$ ) to one of the dose level in the backfill set
- **Bayesian RAR (BRAR)** adjusts probabilities based on efficacy data from dose-escalation and backfill patients
  - ▶ BRAR design: use a posterior probability of interest to determine assignment **probabilities** [Thompson (1933)].
  - ▶ BRAR allocation probability:  $P(a_{k,i} = 1 | \mathbf{A}^{(i-1)}, Y^{(i-1)}) = P(\max_{1 \leq h \leq MTD-1} p_h = p_k | \mathbf{A}^{(i-1)}, Y^{(i-1)}) = q^{(i-1)}$

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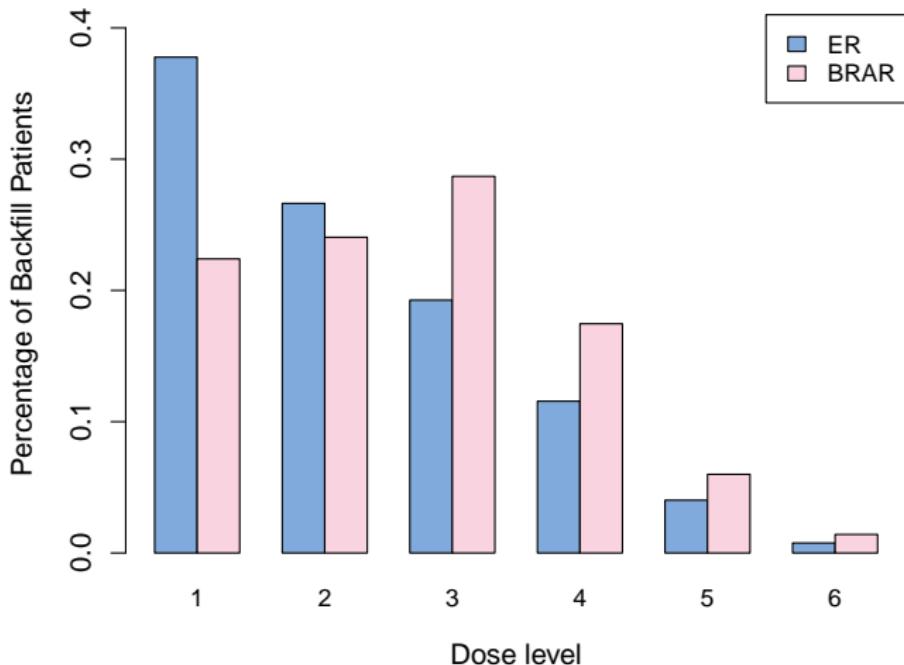
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# Allocation of Backfill Patients

Scenario A



# Recommendation of Dose Level

**Scenario A: % of recommendations per dose**

Dose Level	CRM	Backfill & ER	Backfill & BRAR
1	0%	0%	0%
2	0%	0.3%	0.1%
3	0.2%	27.3%	30.8%
4	15.7%	43.7%	44.7%
5	62.0%	19.4%	16.9%
6	20.7%	7.3%	5.9%
7	1.4%	2.0%	1.6%
RP2D	0.2%	27.3%	30.8%
[RP2D, MTD)	15.9%	70.9%	75.5%
[RP2D, MTD]	78.0%	90.4%	92.4%
(MTD , 7]	22.1%	9.3%	7.5%

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# Final Thoughts

1. Whether backfill is beneficial depends on true DLT and Efficacy curves (7 other scenarios in [Pin et al (2024)]).
2. If Backfill is beneficial, using BRAR leads to *allocating more patients close to RP2D and an increased probability of recommending the correct dose level at the end of the trial.*
3. Future Research
  - 3.1 Impact of different cohort sizes on the design
  - 3.2 Other dose-escalation methods: 2-parameter CRM, BOIN
  - 3.3 Continuous endpoints → nonparametric RAR

# References I

-  O'Quigley, John and Pepe, Marcello and Fisher, Lloyd (1990) Continual Reassessment Method: A Practical Design for Phase 1 Clinical Trials in Cancer. *Biometrics*, 46(1): 33–43. ISSN 0006341X. doi: 10.2307/2531628.
-  Dehbi, Hakim M. and O'Quigley, John and Iasonos, Alexia (2021) Controlled backfill in oncology dose-finding trials. *Contemporary Clinical Trials*, 111. ISSN 15592030. doi: 10.1016/j.cct.2021.106605.
-  Pin, Lukas and Villar, Sofia S. and Dehbi, Hakim M. (2024) Implementing and assessing Bayesian response-adaptive randomisation for backfilling in dose-finding trials. *Contemporary Clinical Trials*. doi: 10.1016/j.cct.2024.107567.
-  Thompson, William R. (1933) On the likelihood that one unknown probability exceeds another in view of the evidence of two samples. *Biometrika*, 25(3-4): 285–294. ISSN 0006-3444. doi: 10.1093/biomet/25.3-4.285. URL .

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# The Gap: Theory vs. Practice

## The Current Perception:

- RAR is often dismissed as a "methodological curiosity."
- "Great theory, but nobody uses it."
- **Reality Check:** Is it truly absent, or just hard to find?

## The Problem

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*"The uptake of RAR in clinical trial practice remains disproportionately low..."*

—

Robertson et al. (2023)

**Our Solution:** A living, crowd-sourced evidence base.

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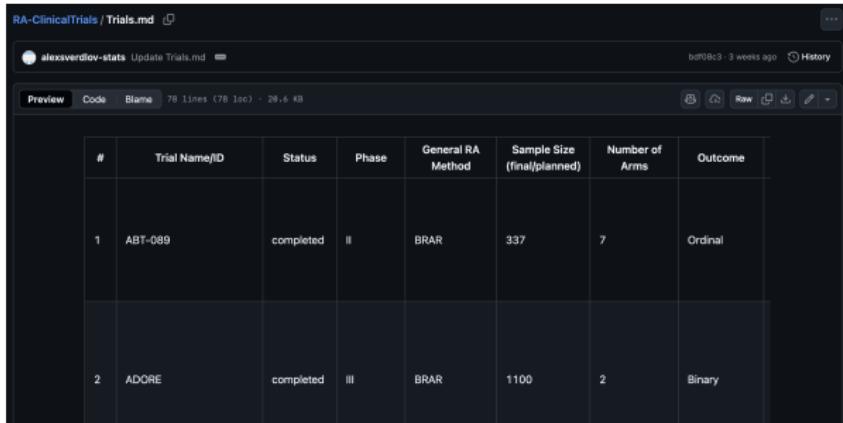
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# RA-ClinicalTrials: A Living Database



A screenshot of a GitHub repository titled "RA-ClinicalTrials / Trials.md". The repository was last updated by "alexsvendrov-stats" 3 weeks ago. The file contains a table with two rows of clinical trial data:

#	Trial Name/ID	Status	Phase	General RA Method	Sample Size (final/planned)	Number of Arms	Outcome
1	ABT-089	completed	II	BRAR	337	7	Ordinal
2	ADORE	completed	III	BRAR	1100	2	Binary

- **Open Source:** Hosted on GitHub.
- **Community Driven:** Additions via Pull Requests.

- **Quality Control:** Curated by MRC BSU.
- **Permanent:** Archived on Zenodo.

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# Analysis: Breaking Down the 74 Trials

## Trial Landscape (Nov 2025)

Category	Sub-Category	%
<b>Method</b>	<b>BRAR</b>	<b>83.8%</b>
	RPTW	2.7%
<b>Phase</b>	Phase II (Exploratory)	44.6%
	<b>Phase III (Confirmatory)</b>	<b>20.3%</b>
	Phase I/II or Seamless	6.8%

### Key Insights:

- **Dominance:** BRAR is the standard.
- **Scale:** Used in massive platform trials (e.g., REMAP-CAP).
- **Scope:** Not limited to early phase oncology.

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# Support this Project

Help us grow the evidence base

Have we missed a trial?



[github.com/lukaspinpin/  
RA-ClinicalTrials](https://github.com/lukaspinpin/RA-ClinicalTrials)

## How to Contribute

1. Fork the repository.
2. Add the trial to `trials.md`.
3. Open a **Pull Request**.

# How to Cite

## Official Citation (Zenodo)

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**Please use the DOI, not the GitHub link:**

Pin, L., Neubauer, M., Voller, C., Wilson, I., Deliu, N., Baas, S., Dimairo, M., Robertson, D., & Villar, S. (2025).

*Clinical Trials Using Response Adaptive Randomization (Version 1.1.1)*  
[Computer software].

Zenodo. <https://doi.org/10.5281/zenodo.17493900>

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The End

Thank you for your attention!

Any questions?

Please reach out:

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