

A quick guide why not to use A+B designs

Adaptive Designs Working Group of the MRC Network of Hubs for Trials Methodology Research

This document summarises the main arguments why 3+3 and similar rule-based A+B designs [18,26] are inappropriate for phase I dose escalation studies. These designs are still widely used [17,23-25] although superior model-based designs such as the continual reassessment method (CRM) [9,21] are available.

Besides investigating a toxicity endpoint for one single drug, extensions to model-based designs for other practically relevant settings exist: e.g., two-drug combination therapies [6,11,13,14,32,33], seamless phase I/II designs that assess toxicity and efficacy together [2,4,28,31], and censored time-to-event endpoints (TITE-CRM) [10].

Why model-based¹ designs are preferable to rule-based² designs:

- The goal of an early-phase dose escalation study with cytotoxic oncology drugs is to estimate the maximum tolerated dose (MTD) i.e., the highest dose that “does not cause unacceptable side effects” [20], or more statistically speaking, the dose where the probability of dose-limiting toxicities (DLTs) is equal to some prespecified target level, typically between 0.25 and 0.33. Using model-based designs, the MTD at the end of the trial relates to a target probability of DLT that is specified before the trial begins. However, this is not the case with rule-based designs, and the expected probability of DLT at the MTD can vary greatly [7,15].
- More patients receive doses near the MTD with model-based than with A+B designs, where patients are more likely to be overdosed or treated at subtherapeutic doses [1].
- For instance, it was shown in simulations that only about 35% of patients are treated at the optimal dose with a 3+3 design compared to 55% for Bayesian adaptive designs [1].
- Model-based designs allow estimation of the MTD together with an informative measure of precision. However, under a 3+3 design there is a lot of uncertainty around the MTD selected at the end of a trial; the 95% confidence interval for the probability of DLT is either (0.004, 0.64) (for 1 in 6 DLTs) or (0, 0.71) (for 0 in 3 DLTs); if dose de-escalation is allowed in the design, it can also be (0, 0.46) (for 0 in 6 DLTs).
- A+B designs are commonly viewed as simple and straightforward, in opposition to the purported “black box” of model-based designs (where clinicians may feel they are not in control of their study), but there are user-friendly software tools (see below) to facilitate their implementation. Moreover, it is easy to make the “black box” transparent using the concept of “dose transition pathways” [30] that can help understand possible courses of action in a trial that uses a model-based design.

¹ There are model-free (or curve-free) designs that are recommendable, too, but these are not based on simplistic “rules” like A+B and related designs.

² Obviously model-based designs also involve certain “rules” e.g., how to update the model. We use the term rule-based here for A+B and related designs. Rule-based designs are also called algorithm-based designs.

- A+B designs are “memoryless” [22] and therefore inefficient: they only use the information from the last cohort of patients that entered the study to determine the next dose whereas model-based designs make use of all data accumulated.
- Although there are some extensions of rule-based designs (e.g., to drug combinations [5,12,16]), they also suffer from the same drawbacks as simple A+B designs, whereas model-based approaches are clearly more flexible and supersede rule-based designs also for more complex questions e.g., when assessing drug combinations [6,11,13,14,32,33] or toxicity together with efficacy [2,4,28,31].
- Where the goal of the dose escalation study is to understand a dose-response relationship (e.g., when looking into a pharmacodynamic marker and not just toxicity), only a model-based approach will be expedient, whereas A+B designs do not adequately characterise the dose-response curve and are therefore useless.

Table 1: Summary of some key features of rule-based and model-based designs.

	Rule-based designs	Model-based designs
<i>Target DLT rate</i>	unclear	clearly defined and can be flexibly chosen
<i>Patients treated at the optimal dose</i>	(relatively) few	(relatively) many
<i>Patients treated at subtherapeutic doses</i>	(relatively) many	(relatively) few
<i>Utilisation of available data</i>	poor	efficient
<i>Extension to more complex questions</i>	difficult & dubious	smooth & straightforward
<i>Deviations from the plan (e.g., other doses, different number of patients on a dose)</i>	hard or impossible to incorporate	easily accommodated

Whatever design is used, the model should be viewed as a guiding tool as there are a lot of factors not captured by only looking at DLTs. It provides recommendations for dose-escalation decisions, which can be combined with clinical expertise in order to reach a consensus about what the next cohort of patients will receive.

Some useful (and free) software tools:

- Properties of rule-based A+B designs can be calculated and visualised with this web app: <https://graham-wheeler.shinyapps.io/AplusB/>.
- CRM designs are easily calculated with the R packages `bcrm` [27], `CRM` [19], or `dferm` [8].
- Dual agent and multiple endpoints designs are included in the R package `crmPack` [3].
- The MD Anderson Cancer Center offers a software library (<https://biostatistics.mdanderson.org/softwaredownload>) with multiple tools for dose

escalation studies, specifically bCRM, BMA CRM, BOIN Design Desktop Program, CRM, EffTox, Dose Schedule Finder, ToxFinder, UAROET, and U2OET.

- A web-based dose-finding tool NextGen-DF [29] is available at <http://www.compgenome.org/NGDF>.
- The Center for Quantitative Sciences at Vanderbilt University's School of Medicine developed a web app to compare and select among different designs for dose escalation: <https://cqs.mc.vanderbilt.edu/shiny/AdaptiveDesignS>.

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