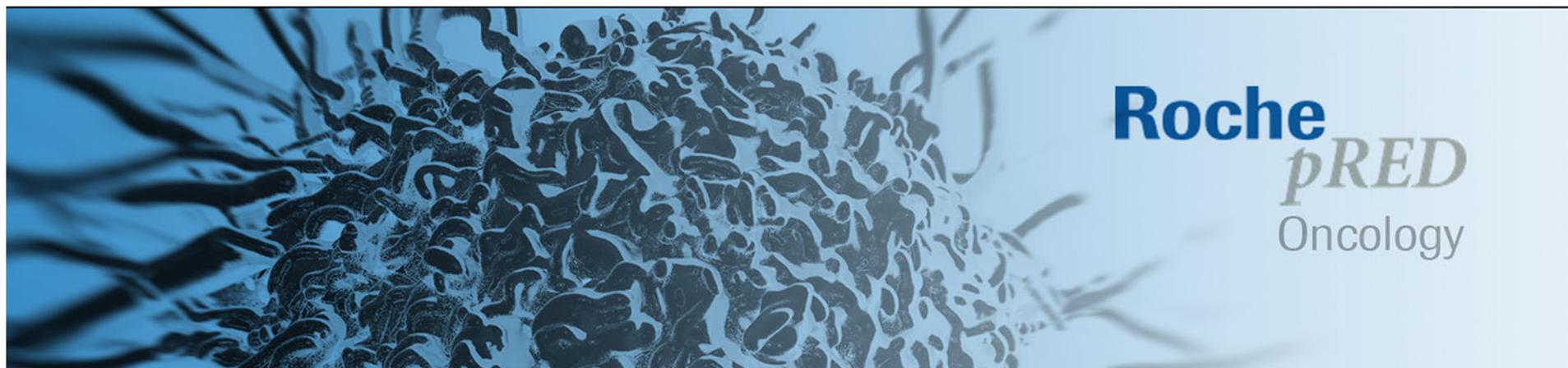

Bayesian Learning in Early Phase Cancer Immunotherapy: A Case Study

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Case study on an NME (“A”) in Oncology

Introduction

- New cancer immunotherapy (CIT) “A” for the treatment of solid tumors
- Mainly expected to work in combination with other drugs
- Frontrunner for a platform of similar biologics
- Entry-into-human phase 1 study started
- Combination A + B phase 1b study to start about one year later
- Follower NME C from same platform still to enter phase 1

This talk: How to use Bayesian statistics for answering the typical clinical development questions (in blue) in this CIT project

What is the maximum tolerated dose (MTD)?

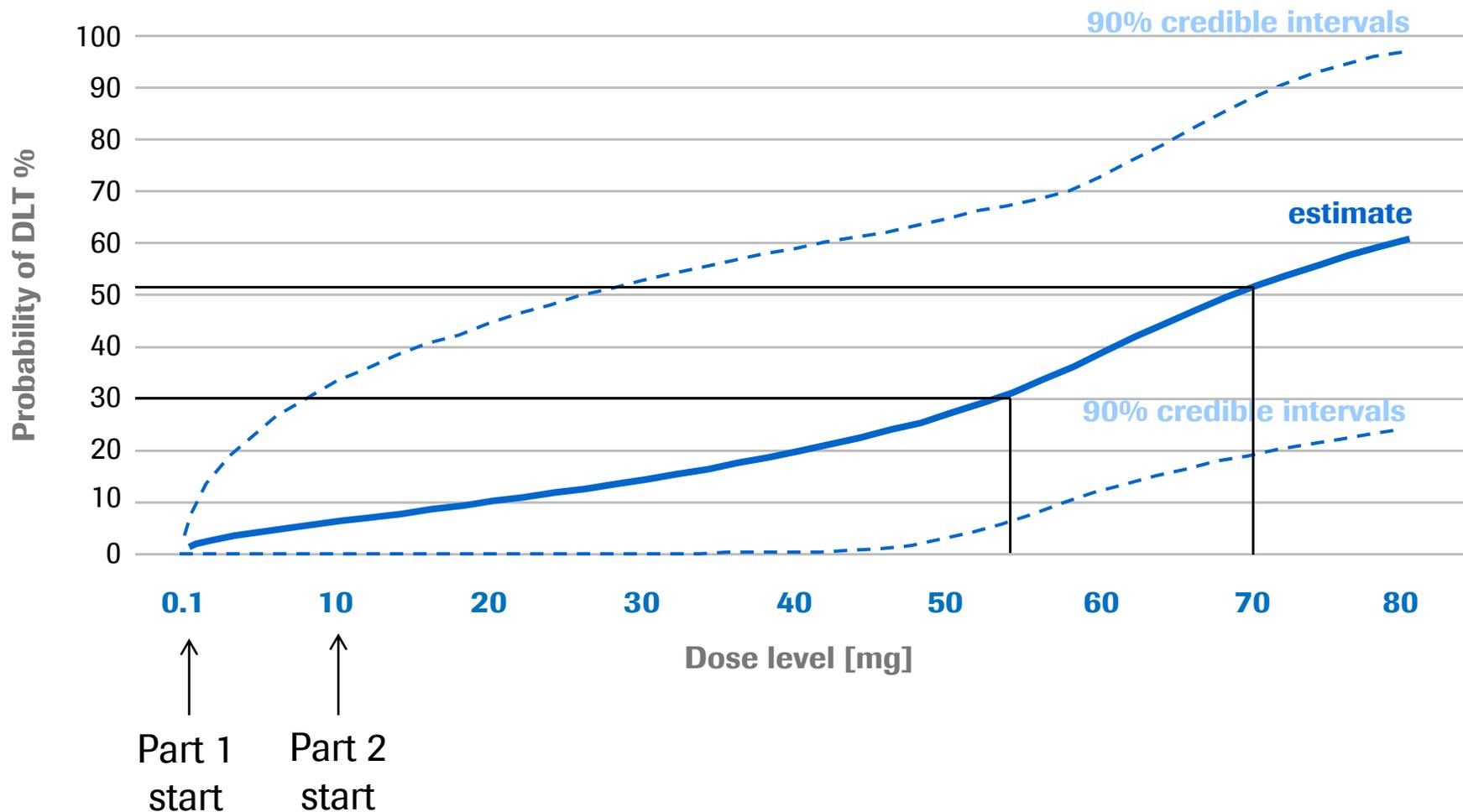
Approach used in phase 1 study

- Two-part dose escalation for our NME A
- First part: single patient cohorts in very low dose range (starting from minimum anticipated biological effect level [MABEL] x 4)
 - Reason: no good animal toxicity model exists (only one with homologous molecule)
 - Safety rules for closing this part early
 - Rule-based escalation: step up dose “staircase”
- Second part: modified Continual Reassessment Method (mCRM) design
 - Dose-limiting toxicities (DLTs) defined in protocol (3 weeks period)
 - Logistic regression for modelling DLT rate in relation to dose
 - Start with biweekly (Q2W) schedule evaluation, afterwards Q3W
- Prior information which was used for mCRM setup:
 - In-vitro assays
 - In-vivo non-human primates study with a homologous molecule

Prior dose-DLT rate model

Logistic regression with bivariate normal prior

$$\text{logit}[p(x)] = \alpha + \beta \cdot \log\left(\frac{x}{x^*}\right), (\alpha, \log(\beta)) \sim N_2(\mu, \Sigma)$$



Dose escalation with mCRM

Standard design to find MTD

- 1. Gather DLT data (binary) for each patient
2. Update the dose-DLT rate model with Bayesian inference (get posterior distribution of parameters with MCMC)
3. Use updated model to estimate the next best dose:
 - **Priority No. 1: overdose control**
Dose must have $< 25\%$ risk that DLT rate is $> 35\%$
 - **Priority No. 2: “target toxicity window”**
Dose should maximize chance that DLT rate $\in (20\%, 35\%)$
4. Are prespecified stopping criteria met? (relating to precision reached)
 - Then define MTD = next best dose.
 - Otherwise continue to step 5.
5. Treat next cohort of 3 patients at next best or a lower dose (clinical judgement has always final call!)

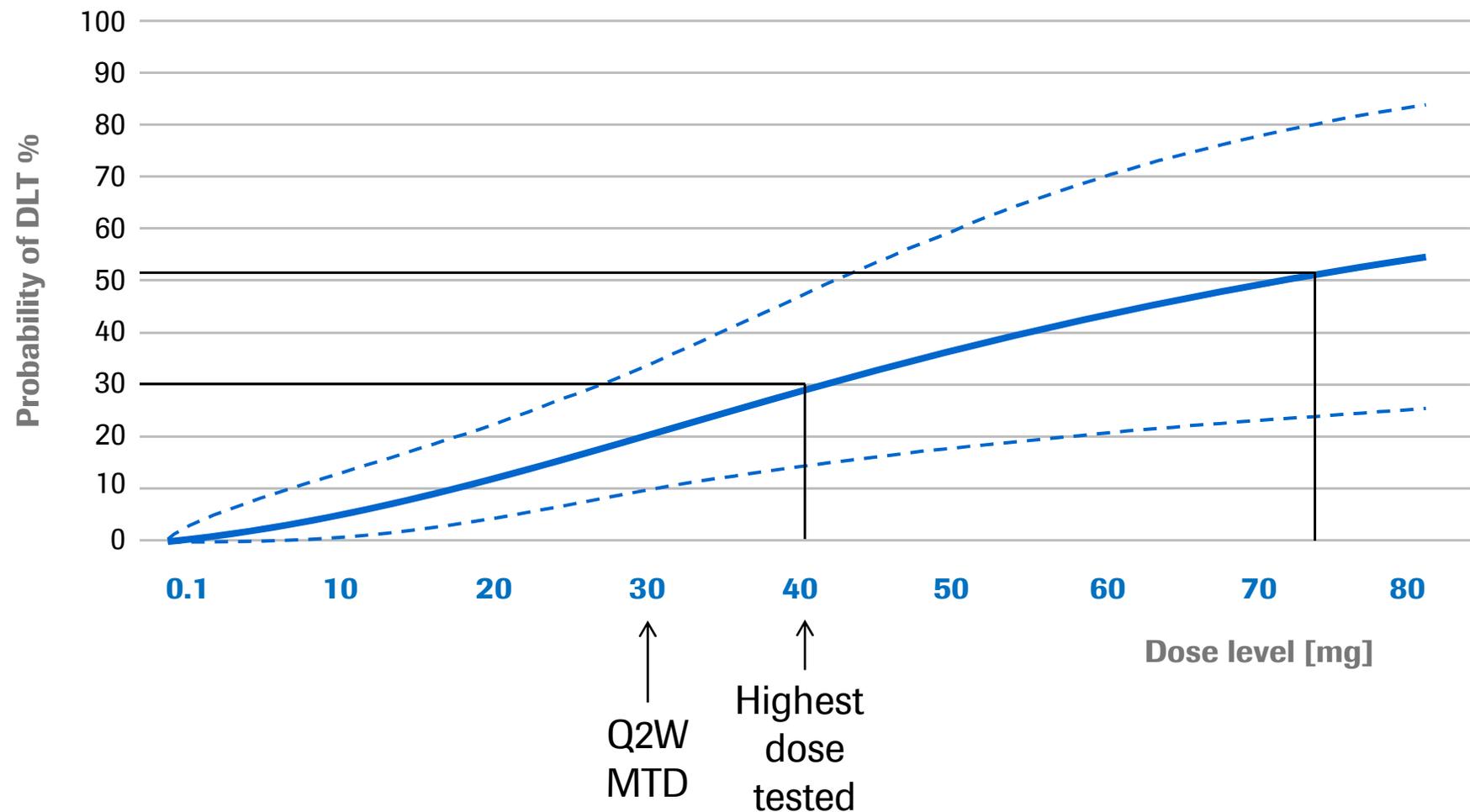
Practical experience in the case study

Benefits, challenges and opportunities

- High flexibility of the mCRM design:
 - Number of patients per cohort could be increased as needed (e.g. to get additional pharmacodynamic data – biopsies are important in CIT)
 - Parallel enrollment at different doses possible (e.g. from an imaging substudy for biodistribution data generation)
 - Later detected DLTs could be incorporated (e.g. due to immune-mediated adverse events)
- Some practical challenges encountered:
 - Clinicians were more conservative due to non-DLT adverse events (AEs)
 - Change from originally planned Q3W to QW schedule during the study
 - Newly proposed escalation in cycle 2 or cycle 3 instead of only in cycle 1
- Opportunities for extensions of mCRM:
 - Ordinal AEs / multiple schedules / stepwise escalation / ...

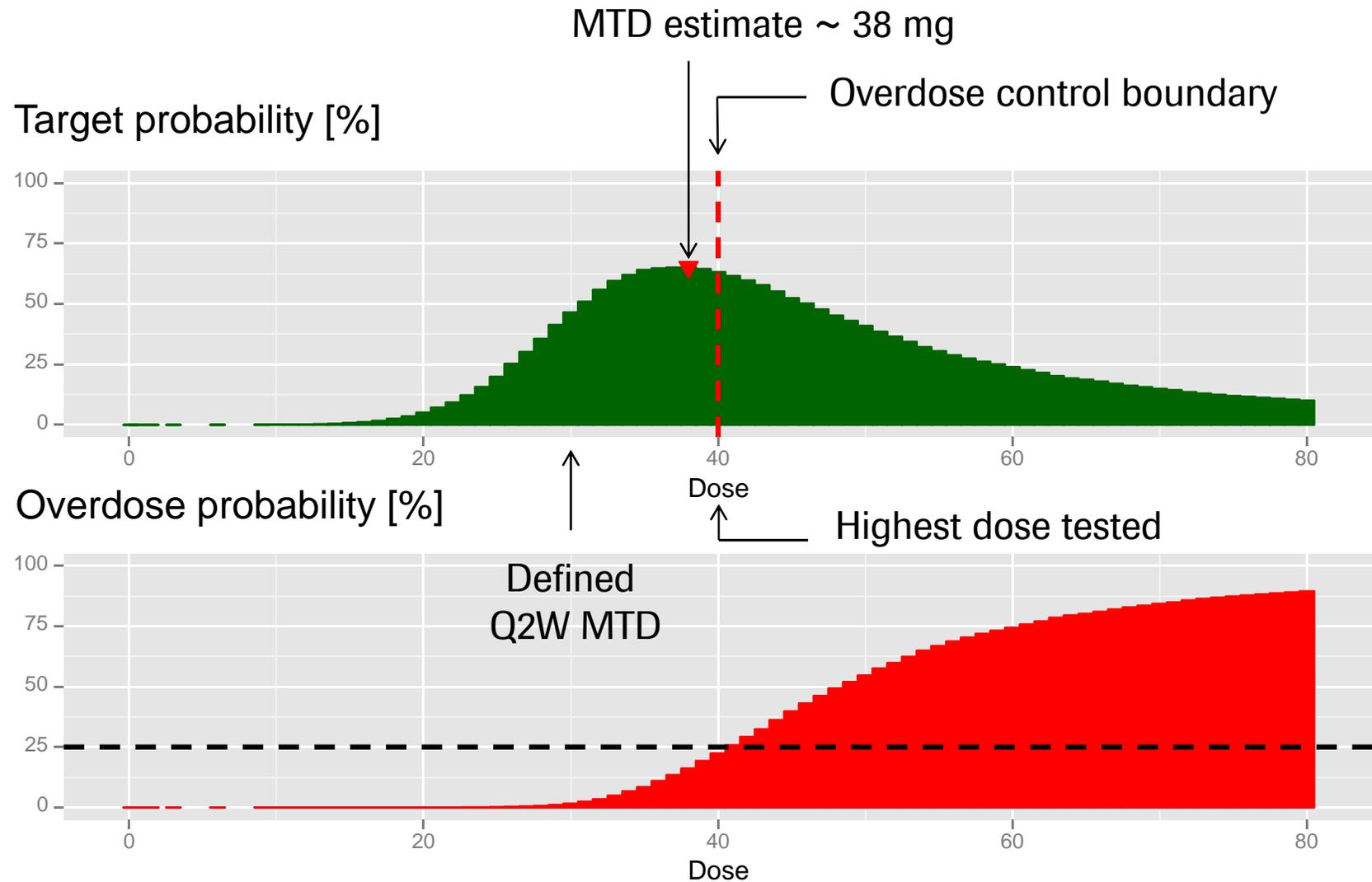
Posterior dose-DLT rate model

Example status after including 58 patients



Example target and overdosing probabilities

mCRM design would give higher MTD than defined



Combination study with NME B

Introduction

- No tumor shrinkage was observed so far in the phase 1 of our NME A
- As per clinical development plan, proceed to 2NME combination A + B with another CIT molecule “B”
- Synergistic toxicity cannot be excluded, especially due to concomitant administration of A and B
- Dose and schedule of NME B are given and should not be changed
- Therefore do single agent dose escalation only on our NME A
- How can we include the prior information from the phase 1 of A and the existing safety data from B?
- How can we decide whether the safety and efficacy of the 2NME combination A + B warrants further development? (“gating”)

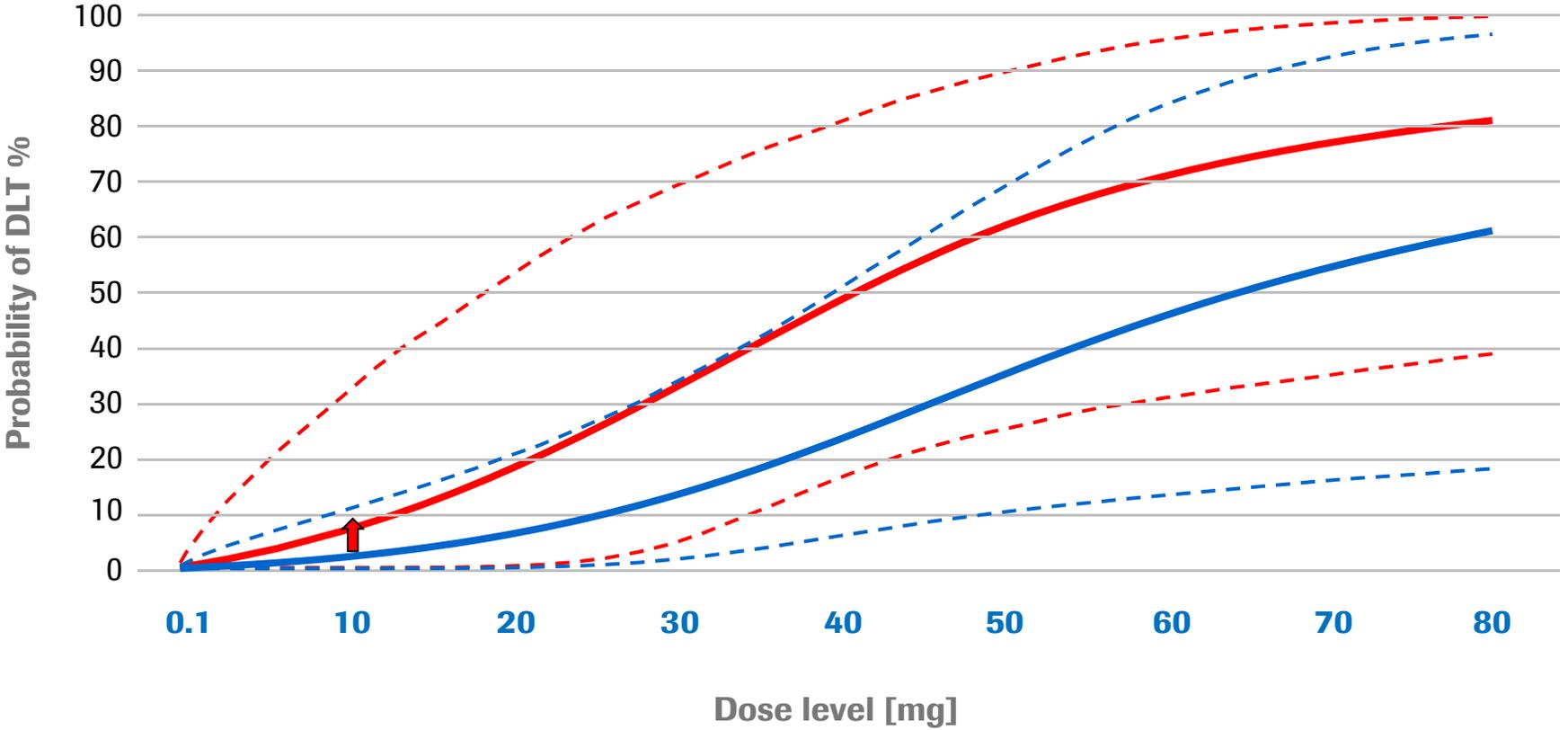
How to include prior information from A and B?

Pragmatic approach (not the only or best one)

- Take the posterior model from phase 1 for NME A at the cutoff date, approximate parameter posterior with bivariate normal
- Add additional toxicity to be expected from the combination with NME B, by shifting the prior mean of the intercept
- Here: expect 5% additional DLT probability at dose of 10 mg (based on prevalence of potentially overlapping G2 toxicities in B)
- Some uncertainty about it taken into account, by increasing the prior variance of the intercept
- This leads to the prior dose-DLT rate model for the combo A + B
- With target toxicity window of 20-30%, the prior MTD estimate is 28 mg (22 mg with overdose control)

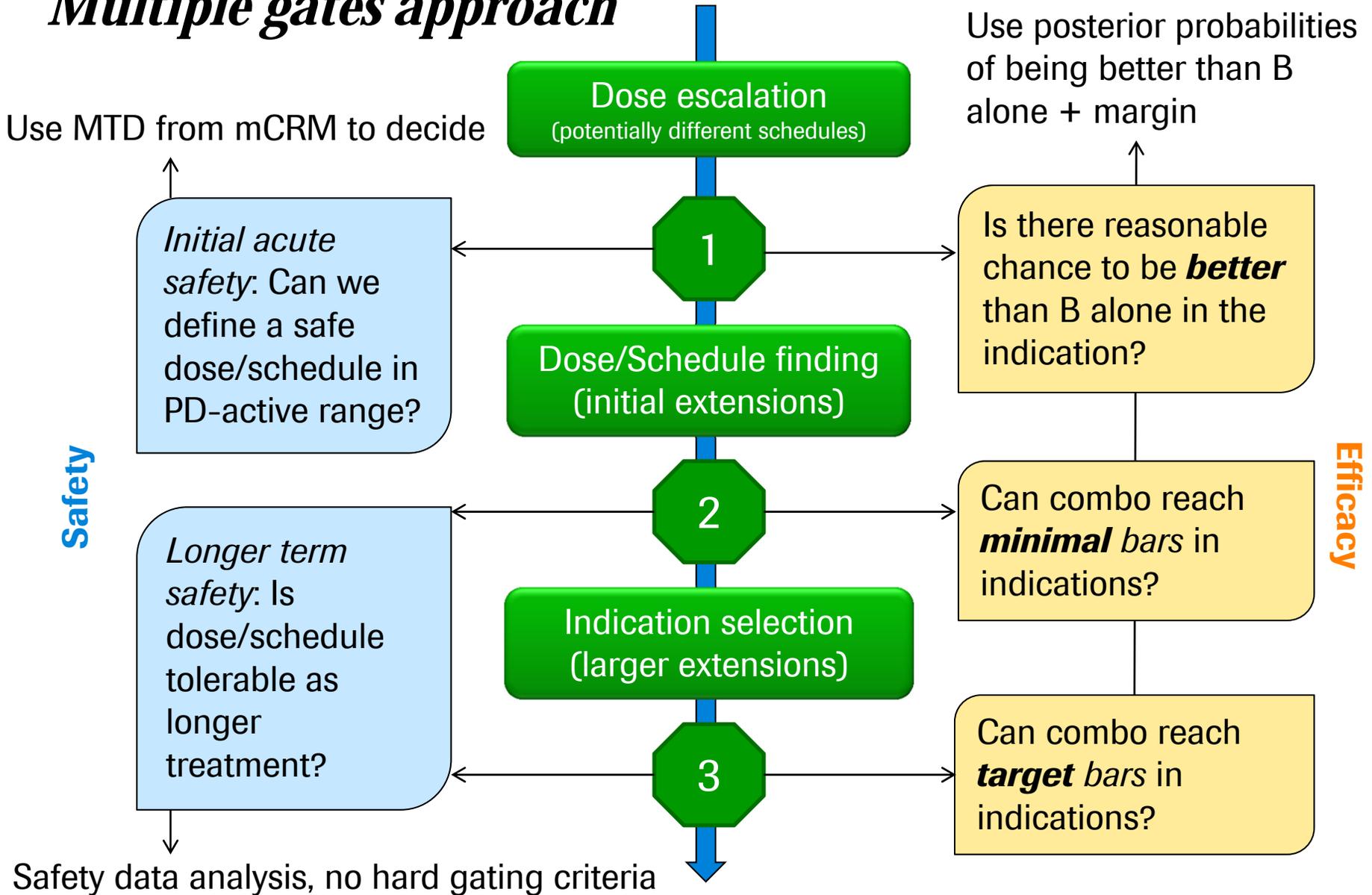


Comparison: **posterior for A** vs. **prior for A (+B)** *At cutoff date included in the protocol*



Should we continue developing combo A + B ?

Multiple gates approach



Use posterior probabilities for gating decisions

Example at the first efficacy gate



Is there reasonable chance to be **better** than B alone in the indication?

- Use objective response rate (ORR) as endpoint (note: all of this can be extended to coprimary / composite endpoints)
- Define “reasonable chance” = “at least 30% posterior probability”. Use uniform prior, and conjugate beta-binomial model for computations.
- Assume we have n=9 patients in an indication during dose escalation, (across all doses and schedule to avoid «cherry picking»)
- Assume with NME B alone there were 7/43 responses (obs. ORR: 16%) in this indication
- → Possible ORR / probability / decision outcomes for this indication:

Responses:	0	1	2	3	4	5	6	7	8	9
Obs. ORR [%]:	0	11	22	33	44	55	67	78	89	100
Probability to be better than B alone [%]:	17	47	73	89	97	99	100	100	100	100

STOP ↑ 30% threshold **GO**

Immunogenicity of A and pretreatment with X

Introduction

- Many of the phase 1 patients developed anti-drug antibodies (ADAs), not untypical for CIT molecules given interaction with immune system
- However, no safety impact and most of them also no PD impact
- Idea: Does pretreatment with drug “X” reduce the proportion of ADAs ?
- Proof-of-concept study with n=20 patients and binary ADA endpoint
- Patients will be randomized 3:1 to pretreatment, and control arm will be enlarged / have informative prior from the relevant phase 1 patients.

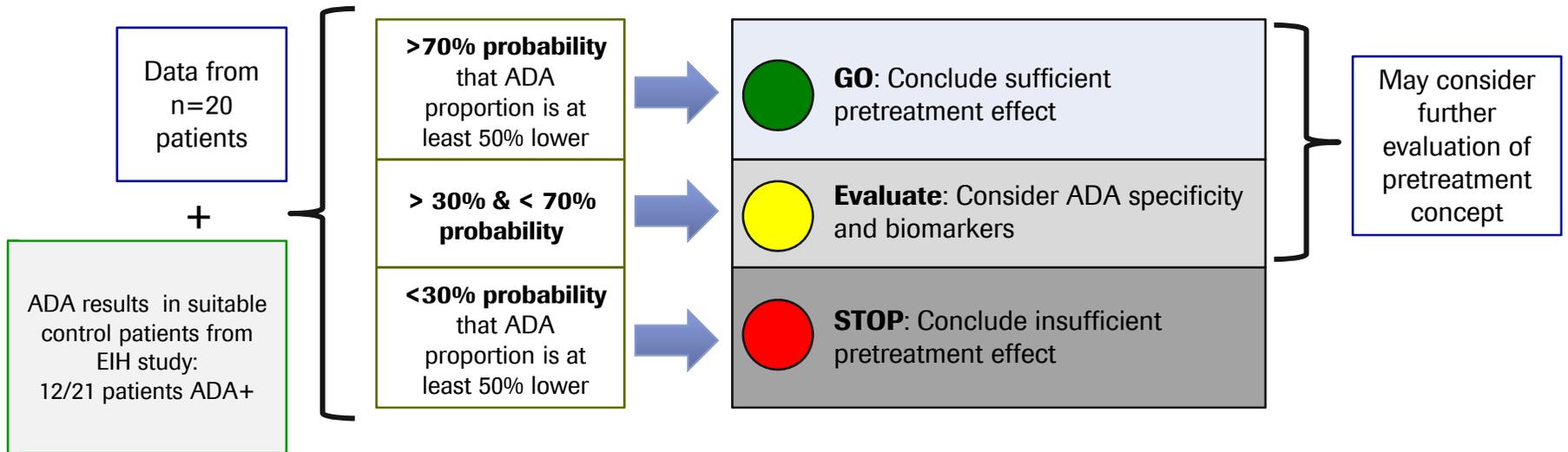
$\xrightarrow{\text{ADA prop. with pretreatment}}$
 $\xrightarrow{\text{Target reduction of ADA prop.}}$
 $\xleftarrow{\text{ADA prop. in control}}$

- Compute posterior probability $\mathbb{P}(\pi_1 < (1 - \delta)\pi_0 \mid \mathcal{D})$
- If $> 70\%$ \rightarrow success; if $< 30\%$ \rightarrow failure; otherwise \rightarrow inconclusive.
- Look here for large effect $\delta = 50\%$; reasonable power for detecting this.

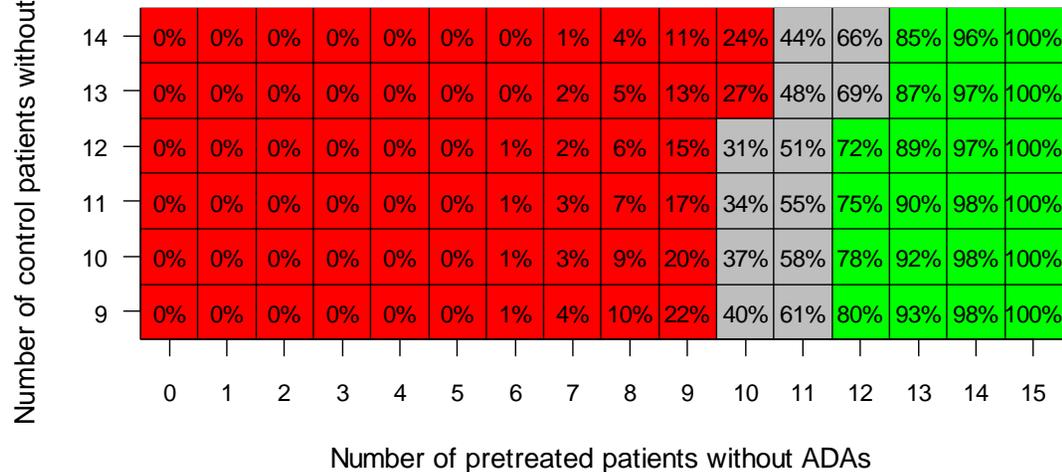
How to decide if pretreatment is effective?

Illustration of the Bayesian answer with example

Numbers just for illustration



Overview of potential outcomes with probabilities
 $\mathbb{P}(\pi_1 < (1 - 0.5)\pi_0 \mid \mathcal{D})$ and implied decisions



How to use the data for the follower NME C ?

General and statistical aspects

- Follower NME C from the same platform was still to enter-into-human.
- Safety and PK profile expected to be very similar to that of our NME A.
- Therefore, we should learn as much as possible from A to design the dose escalation trial for C as efficient as possible, for example general aspects:
 - Starting dose?
 - Schedule?
 - DLT period: could it be shortened? / account for later DLTs? Etc...
- Specifically, for the mCRM design, we would like to use the DLT data from A to specify an informative prior for C.
- Still, the design should be robust against surprises.

Robust mixture prior approach

Overview of the concept

- Take the posterior distribution of the model parameters and approximate it with a bivariate normal → informative component
- Construct a minimally informative prior → «neutral» component
- Mix the two components in a ratio such that the prior sample size is adequate (note: pragmatic approximations here)
- For example: If the informative prior is constructed from 60 patients, and prior sample size should be 10 patients, then choose 1:5 ratio (1/6 weight for informative component)
- The neutral component makes the design robust against surprising deviations from the informative component.
- Still, the dose escalation can be more efficient using the previous data.

Conclusions

Early phase cancer immunotherapy case study

- Dose escalation with new CIT agents needs to accommodate:
 - Very low starting doses, due to lack of adequate tox models
 - Flexible schedule is needed, due to a priori unclear optimal schedule
 - Late toxicities can occur (e.g. Immune-mediated AEs), to consider
- Many combinations are tested in the clinic:
 - Synergistic efficacy is hoped for, but also here preclinical models not conclusive
 - Efficient decision making is required and statisticians need to lead
 - Bayesian tools can help with well interpretable decisions
- Reverse translation to follow-on molecules / other combinations:
 - Bayesian inference ensures consistent borrowing of information
 - Optimally also the starting dose is informed by previous compound



Basel Biometrics Society Seminar 26 June 2017

Innovative model-based dose escalation designs

- **Date:** June 26, 2017, Time: 13.30-17.00
- **Venue: Here!**
(Roche IT Training Center / Aeschentor, Aeschenvorstadt 56, 4051 Basel)
- **Registration:** Free, please send an email to Barbora Martinec (barbora.martinec@roche.com)
- Overview of the current status of the application of such designs in the pharmaceutical industry (Roche, Novartis, Actelion)
- Current research areas for improving the methodology
- Moderated panel discussion around the question “What next?” about the upcoming challenges and opportunities in model-based dose escalation designs and the corresponding clinical trials

Clinicians welcome too!
Please come! 😊
CRM and more



Thank you! Questions?



Doing now what patients need next