

Delivering Transformative Medicines to Patients by Leveraging Complex Innovative Designs

AMG 592 SLE Case Study

Amy Xia, May Mo, Tony Jiang
on Behalf of AMG 592 Team

Amgen

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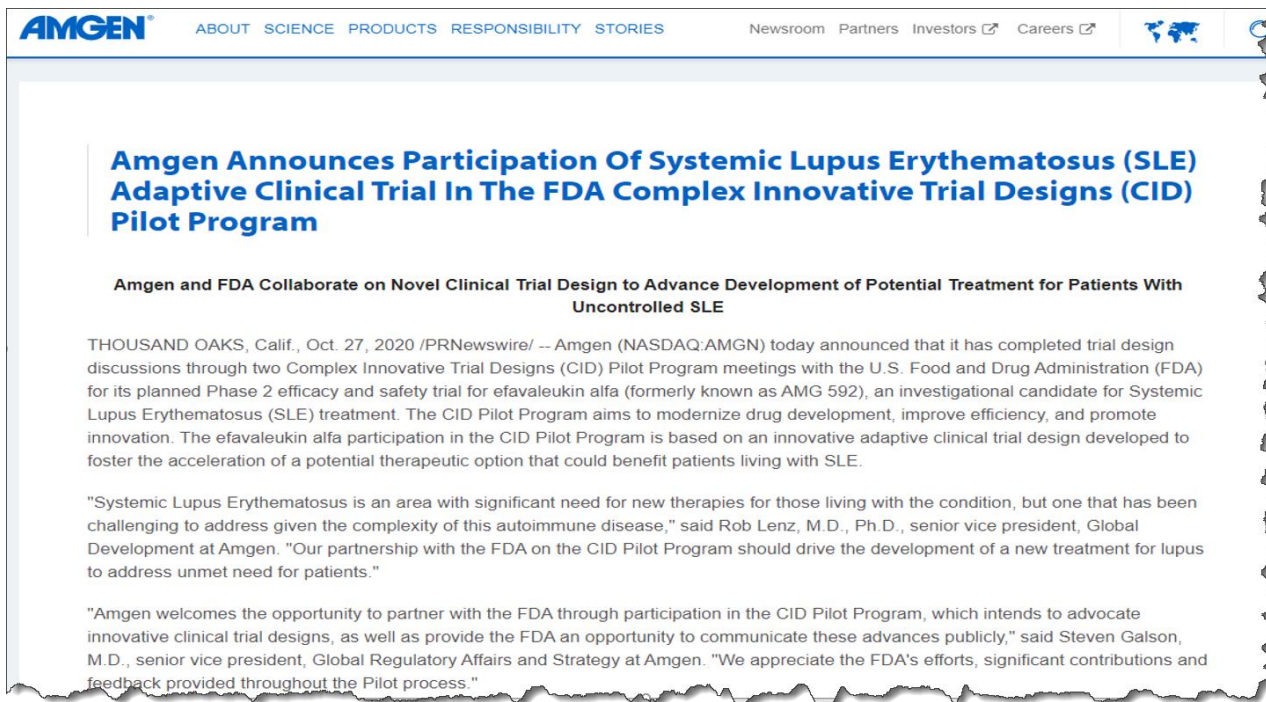


OUTLINE

- **Introduction and Background – Amy Xia**
- **Study Design and 4 Principles – May Mo**
- **Simulation and Shiny Tool – Tony Jiang**

FDA Complex Innovative Design (CID) Pilot Program

- Under PDUFA VI, FDA launched the CID pilot program in 2018, aiming to facilitate and advance the use of complex adaptive, Bayesian, and other novel clinical trial designs which often require simulations to determine the statistical properties of the trial
- The program provides two additional meetings to discuss a specific CID proposal
- FDA can select up to 2 CID proposals per quarter for 5 years
- This program will continue under PDUFA VII



Lupus is a Complex, Heterogeneous Autoimmune Disease

What is lupus?

Systemic Lupus Erythematosus (SLE), or lupus, is a chronic, inflammatory autoimmune disease¹ which affects approximately five million people globally²

5m



In SLE the body produces antibodies that attack its own healthy cells and tissues in addition to producing antibodies to protect against infection^{2,3}



BRAIN



HEART



KIDNEYS



MUSCULO-SKELETAL

Signs and symptoms

Symptoms can vary greatly. Some of the most common symptoms of lupus are:¹



Extreme fatigue

Skin rashes

Anaemia

Kidney problems

Painful and swollen joints (arthritis)



It is estimated that

70-90%

of lupus cases are in females

with the highest incidence during a woman's most productive childbearing and professional development years⁴



Persistent SLE disease activity is associated with a higher risk of organ damage and mortality⁵

Top Barriers to Lupus Drug Development

- Challenges in understanding the biology of the chronic autoimmune disease
- Heterogeneity of clinical symptomatology defining the patient population

Disease
Heterogeneity



- Lack of user-friendly, sensitive and accurate outcome measures
- Lack of stand-alone domain specific assessments of organ systems or symptoms

Outcome
Measures



- Under-represented disease populations and many competing trials
- Suboptimal outcome measures
 - High variability
 - High control response rate

Clinical Trial
Design



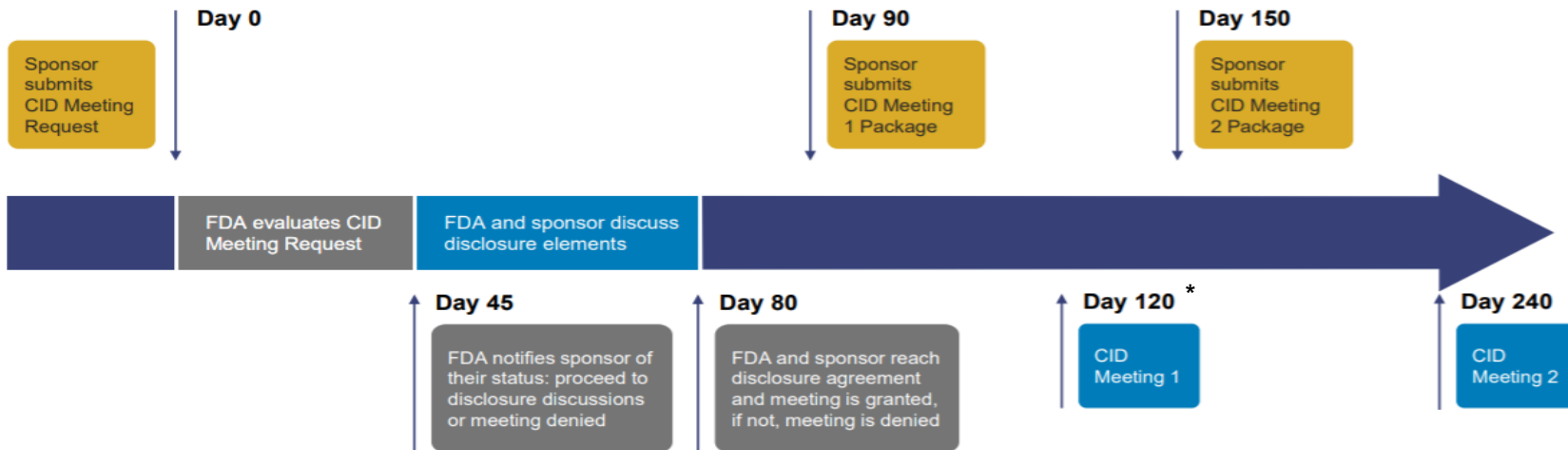
Heterogeneity of the disease is a foundational barrier

Rationale for Proposed CID

These challenges have led to a high development failure rate of potential therapeutics and highlight the need for **innovative clinical trial design to improve development efficiency and probability of success** compared with the traditional development approach:

- make the most efficient use of clinical trial data to simultaneously inform dose selection, generate adequate and well-controlled evidence on efficacy and quality safety data
- reduce the probability of inconclusive trial, and enable early and accurate decision-making
- shorten the time to bring new therapies to patients

CID PILOT PROGRAM: *PROCESS AND TIMELINE*



***Note:** *If sponsor believes that feedback received at the first CID meeting is sufficient and does not want a second meeting before initiating a trial, the sponsor may choose to finalize the protocol, submit it to the IND, and begin enrolling patients*

AMGEN US FDA EXPERIENCE THROUGH THE CID PILOT PROGRAM

Amgen participated in two meetings with FDA to engage in scientific discussions and reach agreement on an innovative study design that is appropriate for a study supporting registration

Meeting Request	Meeting 1	Meeting 2
<ul style="list-style-type: none">➤ Requested discussion of the clinical relevance of the potential primary endpoints and formal definition of their estimands➤ Recommended removing some proposed adaptive elements to reduce the dimensions to be explored in simulation for feasibility and interpretability considerations➤ Suggested arm-dropping as an alternative to RAR*➤ Set expectations on operating characteristics, simulation replicates, and nuisance parameters to be explored	<ul style="list-style-type: none">➤ Discussed in detail the space of plausible nuisance parameters and combinations required to provide convincing evidence of type I error control and other operating characteristics➤ Confirmed that BHM* and RAR would not preclude the study from being registrational, however, requested evaluations against multiple alternative designs, analysis methods, and simulation scenarios to demonstrate advantages of the proposed design➤ Provided feedback on primary endpoint selection and recommended additional criteria to maintain trial conduct and integrity	<ul style="list-style-type: none">➤ Confirmed that Amgen had largely addressed concerns and implemented suggestions to demonstrate that the proposed study design was appropriate as a registrational study➤ Requested further comparison to alternative methods (NDLM, Dunnett) to establish BHM as the favorable method➤ Requested information to justify for range of control response rate and concordance between adjacent visits➤ Requested data access plan to be submitted

*RAR: Response Adaptive Randomization

BHM: Bayesian Hierarchical Model

Regulatory Guidance & 4 Principles

Adaptive Designs for Clinical Trials of Drugs and Biologics Guidance for Industry

U.S. Department of Health and Human Services
Food and Drug Administration
Center for Drug Evaluation and Research (CDER)
Center for Biologics Evaluation and Research (CBER)

November 2019
Biostatistics

An adaptive trial intended to provide substantial evidence of effectiveness should satisfy:

- 1. Adequate control of the chance of erroneous conclusions**
- 2. Sufficiently reliable estimation of treatment effects**
- 3. Pre-specification of trial planning**
- 4. Maintenance of trial integrity**

How We Benefitted from CID Regulatory Engagement



FEEDBACK

Direct feedback
from large
multidisciplinary
team from the
agency



KNOWLEDGE SHARE

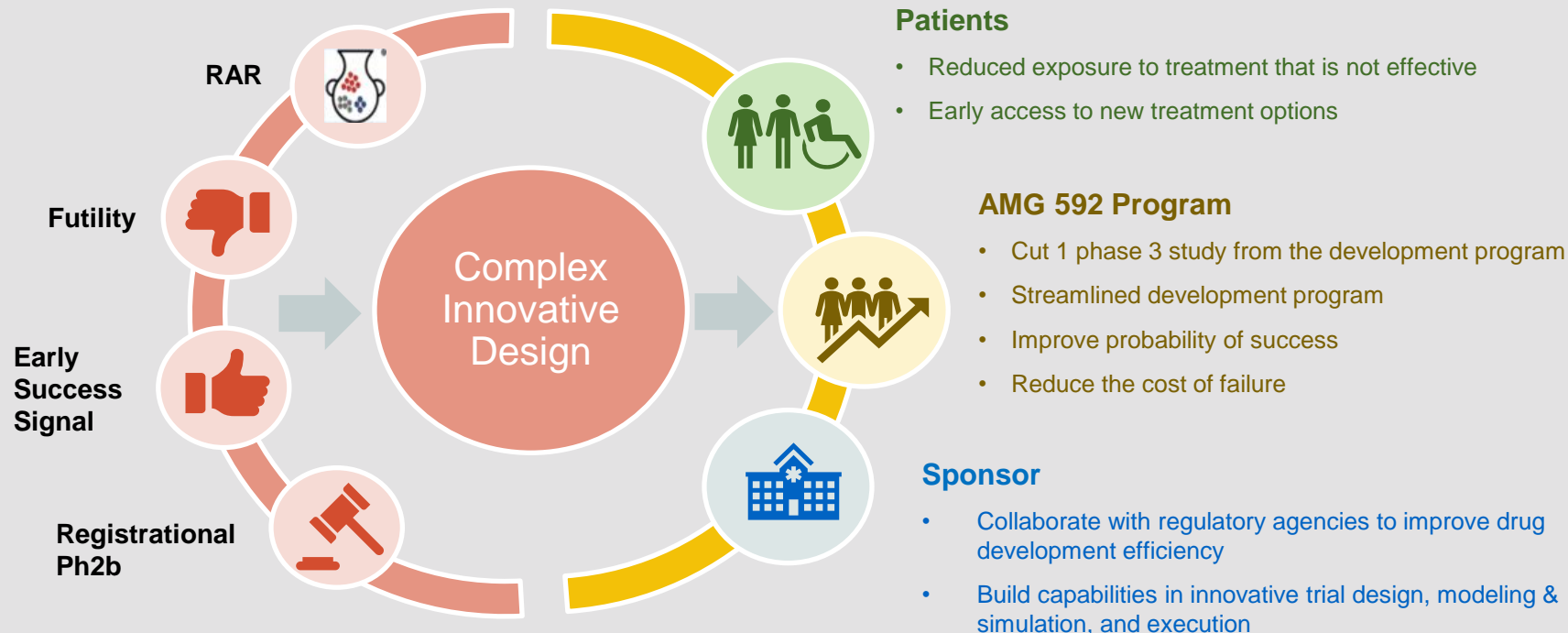
Opportunity to share
innovative tools to
evaluate complex
innovate designs



GUIDANCE

Clear guidance on
missing pieces of the
evaluation

What Do We Gain from the CID Program?



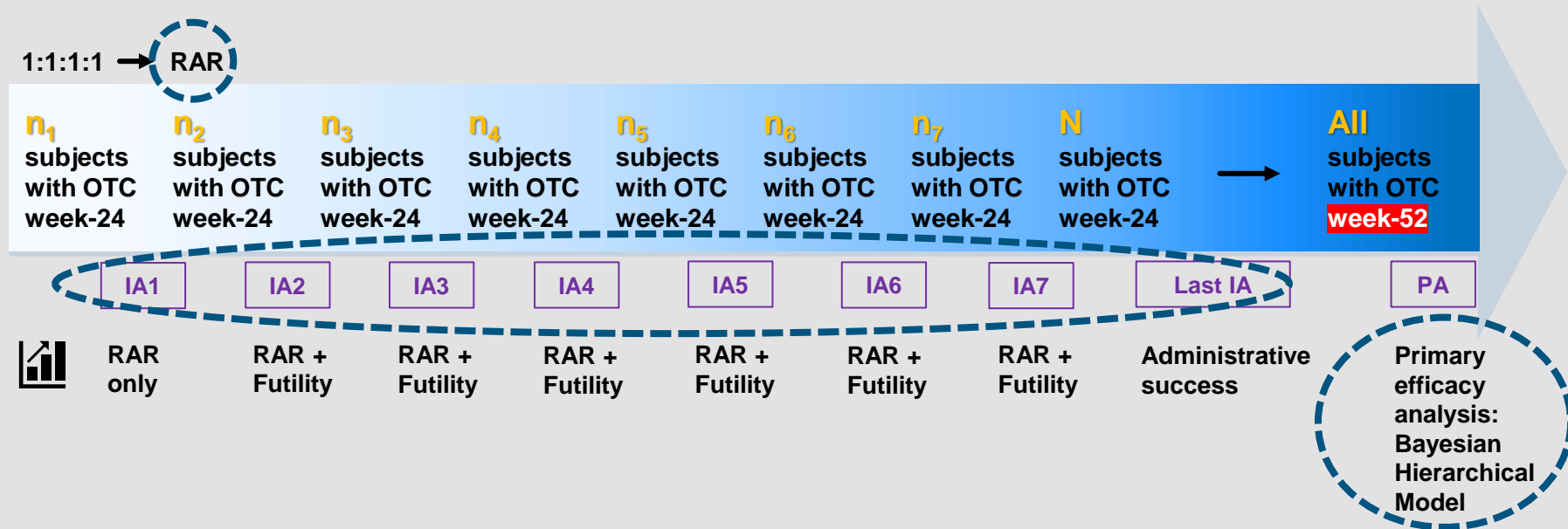
Study Schema



Objectives

1. Dose Selection
2. Qualify as an adequate and well-controlled study

Interim Analysis Schedule



OTC: opportunity to complete

Rationale for the Adaptive/Innovative Design Features



**Response Adaptive
Randomization**

- Learn from accumulating data from ongoing trial
- Patient centric: reduce exposure to less effective treatment
- Increase efficacy & safety data collection on effective treatment



**Interim Analyses
for futility**

- Stop patient exposure to non-effective treatment
- Reduce the cost of failure / shorten development timeline
- Redirect resources to other promising programs



**Final Analysis
Bayesian Hierarchical Model**

- Dynamic borrowing across the active treatment arms improves estimation of treatment effect
- No underlying dose-response assumptions to reduce bias

Response Adaptive Randomization

- The randomization ratio to **each active treatment group** is based on the posterior probability that each group has the highest response rate at **week 52** among the three active treatment groups.

$$Allocation_d \propto \Pr\left(p_d = \max_c p_c \mid \text{interim data}\right) \quad c, d \in \{low, medium, high\}$$

- The posterior probability is calculated based on the **Bayesian independent model**

$$X_d \sim \text{Binomial}(p_d, N_d)$$

$$\log\left(\frac{p_d}{1 - p_d}\right) = \alpha_d$$

for $d \in \{low, medium, high\}$

Bayesian Hierarchical Model

Leverage information across all doses without a prior understanding of expected dose response

- The number of responders in each group is modeled using a binomial distribution:

$$X_d \sim \text{Binomial}(p_d, N_d)$$

where p_d is the **week 52** response rate in group d .

- Each response rate is modeled independently using a logistic model:

$$\log\left(\frac{p_d}{1 - p_d}\right) = \alpha_d$$

- The log-odds of response in the treatment groups is modeled using a hierarchical prior:

$$\alpha_d \sim \mathcal{N}(\alpha_{\text{treatment}}, \sigma^2) \text{ for } d \in \{\text{low}, \text{medium}, \text{high}\}$$

- **BHM** is used in **futility and primary efficacy analyses**

Principle 1: Control of Erroneous Conclusion

Type 1 Error (Global Null)

What is Global Null?

- There is no treatment effect in any of the 3 treatment arms
- There is no treatment effect at any timepoint during the 52-week treatment period

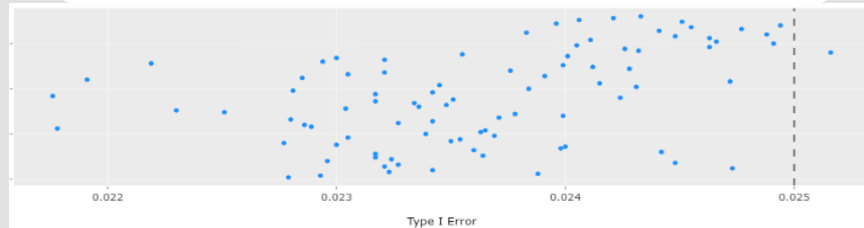
How is Type 1 Error Defined?

- Reject null for any of the treatment arms using Bayesian Hierarchical Model (BHM) and longitudinal modeling for:
 - any plausible control response rate
 - any plausible enrollment rate
 - any plausible correlation (concordance) within subject over time by treatment arms

Nuisance Parameters

- 3 Nuisance Parameters:
 - Enrollment rate: a plausible range
 - Control response: 30%, 40%, 50%
 - Correlation (concordance) patterns:
 - 0.5-0.9 same or different across visits by arms
- Full factorial combinations simulated 100K each

Type 1 Error Rate



Type I Error is controlled across the plausible Global Null scenarios

Principle 1: Control of Erroneous Conclusion

Type 1 Error (Local Null)

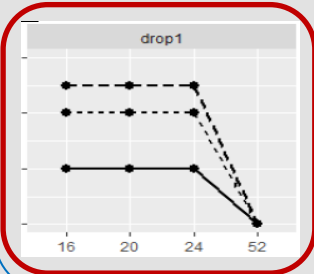
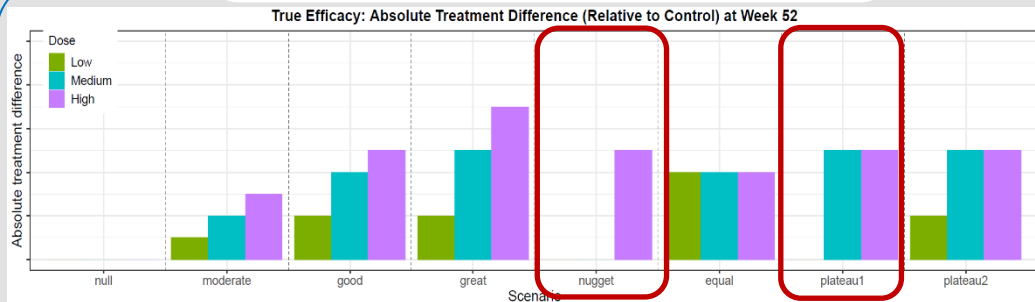
What is Local Null?

1. There is no treatment effect for 1 or 2 treatment arms, while at least one is effective
2. There is no treatment effect at week 52 assessment for the primary endpoint, while there are treatment effects at earlier visits (ie, “Drop1” Scenario)

Are these Type I error?

1. Reject null for any of the ineffective dose levels and select the dose for phase 3
2. Reject null for any treatment arm in “Drop 1” scenario

Scenarios Simulated



1. “Nugget” and “Plateau1” with either 1 or 2 ineffective dose(s) are evaluated across the nuisance parameter factorial combinations
2. The “Drop1” scenarios with efficacy at week 16-24 and none at week 52 is evaluated for selected nuisance parameter combinations

Each scenario was simulated 100k each

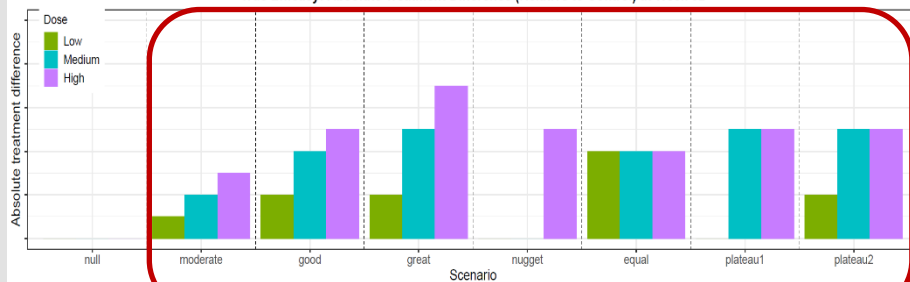
Type I Error is controlled across the plausible Local Null scenarios

Principle 1: Control of Erroneous Conclusion

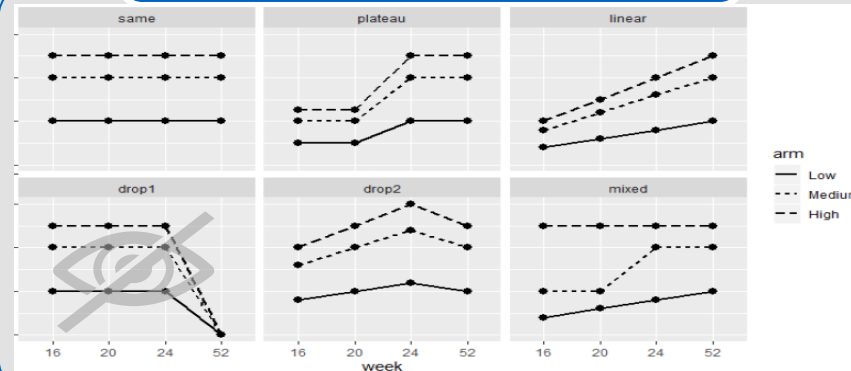
Type II Error

Efficacy Scenarios (n=7)

True Efficacy: Absolute Treatment Difference (Relative to Control) at Week 52



Longitudinal Patterns (n=5)



How is Type II Error Evaluated

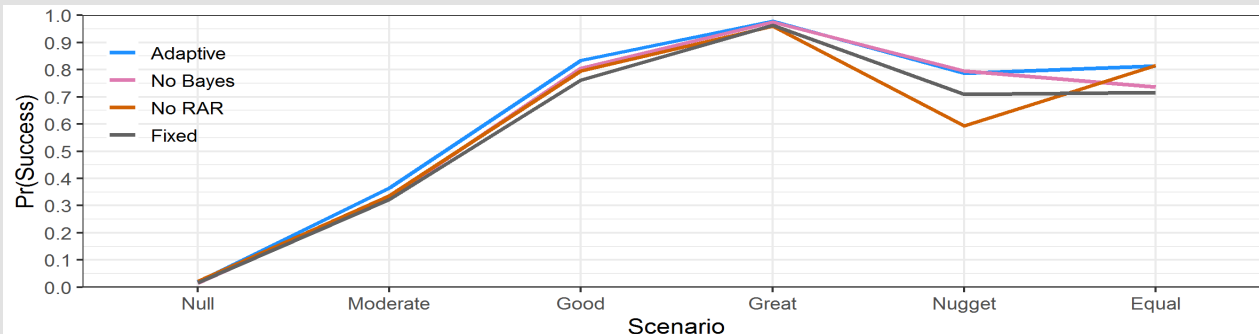
- The nuisance parameter space is reduced by fixing 2 parameters at the most plausible value, and varying the third parameter univariately across its plausible range resulting in reduced set of combinations
- Each of the 12 combinations is then evaluated across the 7 efficacy scenarios and 5 longitudinal patterns, which results in hundreds of total efficacy factorial combinations

Principle 1: Control of Erroneous Conclusion

Type II Error

Power – $\Pr(\text{Success})$ of Any of the Treatment Arms

- Power estimates from 4 study designs are compared below based on the most plausible nuisance parameters and longitudinal pattern



- The operating characteristics (OCs) across simulated scenarios provide sensitivity analysis of the robustness of the study design to the underlying assumptions and identify worst case scenarios

Some Other OCs

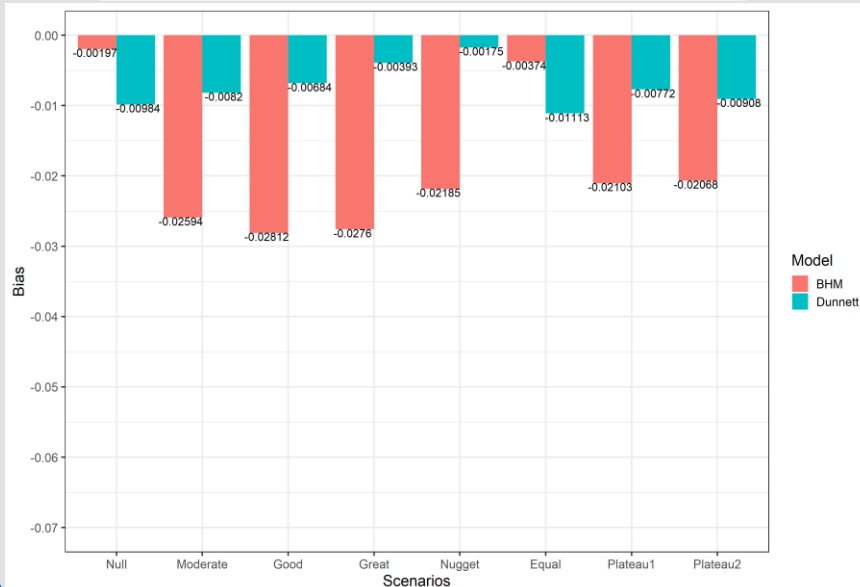
- $\Pr(\text{Futility})$ - probability of stopping the study early for futility
- $\Pr(\text{Adm.Success})$ - probability of achieving the administrative success criteria
- $\Pr(\text{Select Best Dose})$ - probability of selecting the best dose
- Avg.Randomized - Average randomized subjects across all treatment groups

Proposed design elements (RAR and BHM) improve study power

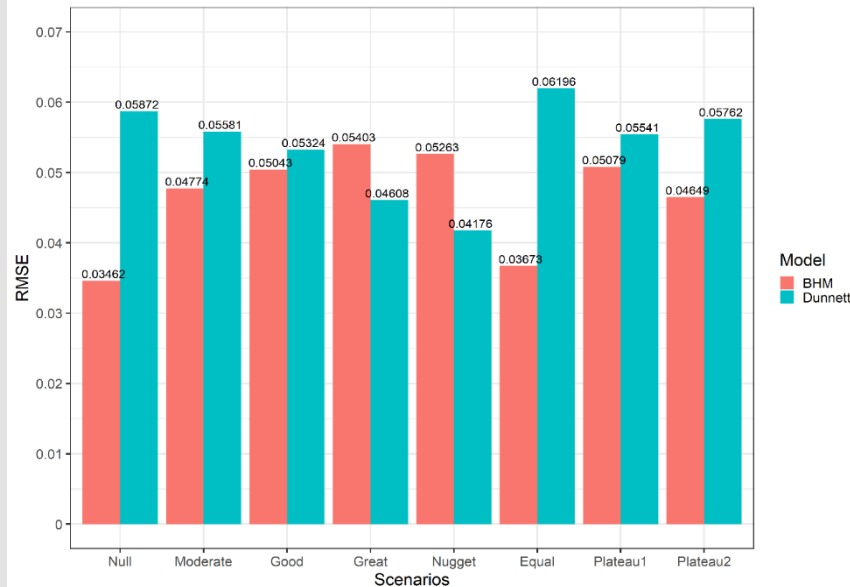
Principle 2:

Sufficiently Reliable Estimation of Treatment Effects

Bias



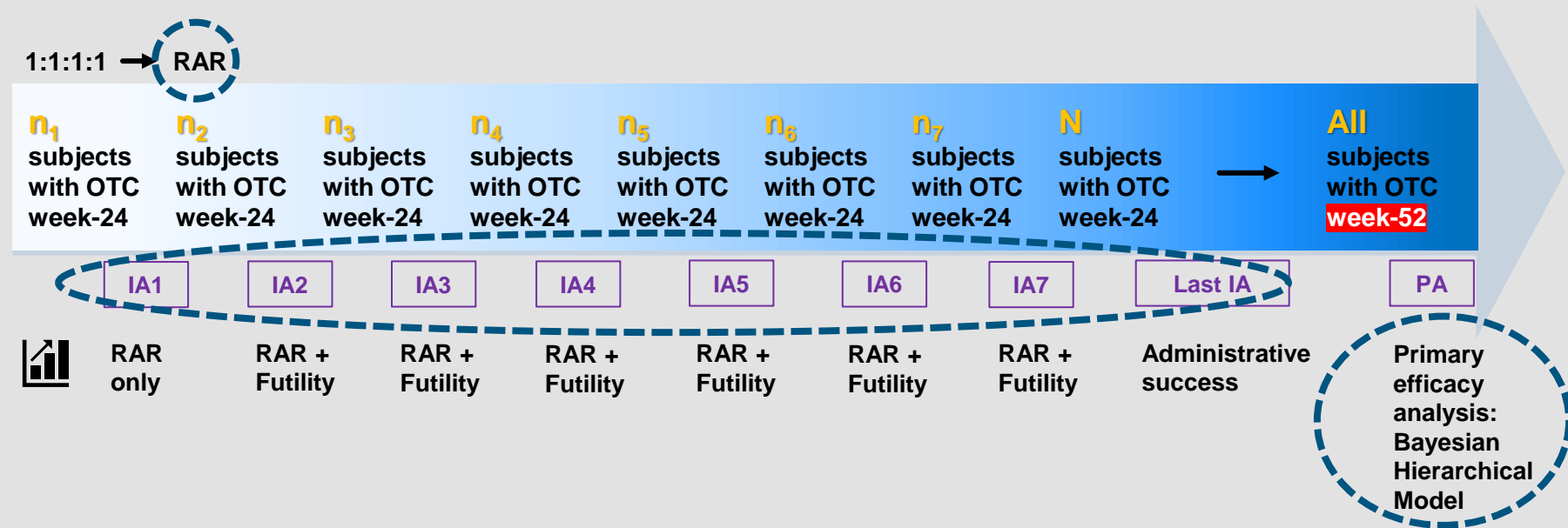
RMSE



BHM has lower RMSE of treatment effect estimates than Dunnett's

Principle 3:

Pre-Specification of Interim Analysis and RAR Algorithm



OTC: opportunity to complete

Principle 3:

Pre-Specified Decision Rules

- ***Futility Stopping***

Enrollment to the study may be stopped for futility if

$$\max \Pr(p_d - p_{placebo} > \text{target treatment effect} \mid \text{Interim Data}) < \text{low value threshold}, d \in \{low, medium, high\}$$

- ***Administrative Success***

BHM will be fit to compute the predictive probability of success in a hypothetical, future phase 3 study, with a frequentist final analysis tested at the 2.5% one-sided level. The threshold of administrative success is the predictive probability of success in this hypothetical future study is larger than a cutoff value.

- ***Primary Analysis Success***

The null hypothesis will be rejected if the posterior probability of superiority in any group is above a threshold:

$$\Pr(p_d > p_{placebo} \mid \text{Data}) > \text{high value threshold}, \text{ for any } d \in \{low, medium, high\}$$

Principle 4:

Maintaining Trial Conduct & Integrity

Adaptive design adds logistical challenges to trial conduct and trial integrity

- Limit access to comparative interim results provides confidence in design modification and assurance of quality trial conduct
 - **External:** Independent Data Monitoring Committee (DMC)
 - Implement a **carefully designed and prespecified adaptation plan**, in addition to its primary responsibility to maintain patient safety and trial integrity
 - **Internal:** Data Access Plan (DAP) to document limited access of sponsor
 - Individuals to perform interim analysis or access interim results
 - Procedures to control access and evaluate compliance
 - Processes for adaptive decision making and dissemination
- Ensure high-quality interim data for adaptive decision-making

SIMULATION EXPERIENCE

- Amgen team, in consultation with the FDA, has conducted an extensive simulation study to evaluate the CID design
- A comprehensive simulation report along with full results and code files have been submitted to the FDA according to the adaptive design guidance recommendation

SIMULATION REPORT – TABLE OF CONTENTS

1. Simulation Objectives	2.4 Operating Characteristics.....
2. Simulation Specifications	3. Simulation Results
2.1 Study Design Options	3.1 Summary of Operating Characteristics
2.2 Model Setup	3.1.1 Type I Error Evaluation
2.2.1 Virtual Data Generating Model.....	3.1.1.1 Global Null Scenario
2.2.1.1 Clinical Scenarios	3.1.1.2 Type I Error Under Non-global Null Scenarios.....
2.2.1.2 Sensitivity Scenarios	3.1.2 Efficacy Scenario Operating Characteristics.....
2.2.2 Analysis Model	3.1.3 Prior Consideration on the Hierarchical Variance Parameter
2.2.2.1 Primary Efficacy Analysis Model	3.1.4 Comparisons to Other Designs
2.2.2.2 Interim Analysis Model	3.2 Example Trials
2.2.2.3 Missing Data	3.2.1 Example Study 1: Success
2.3 Analyses and Decision Rules.....	3.2.2 Example Study 2: Futility.....
2.3.1 Interim Analysis Schedule.....	4. Summary and Recommendations.....
2.3.2 Response Adaptive Randomization.....	
2.3.3 Futility Stopping.....	
2.3.4 Administrative Success	
2.3.5 Primary Analysis Success	

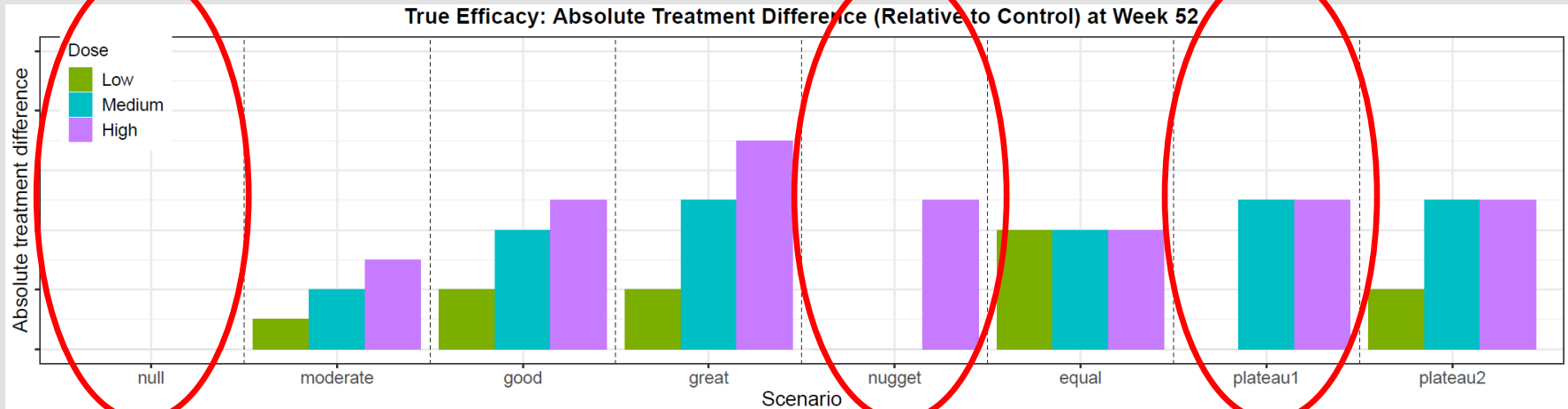
Adaptive Designs for Clinical Trials of Drugs and Biologics Guidance for Industry

U.S. Department of Health and Human Services
Food and Drug Administration
Center for Drug Evaluation and Research (CDER)
Center for Biologics Evaluation and Research (CBER)

November 2019
Biostatistics

SIMULATION SCOPE – CLINICAL SCENARIOS

Efficacy Scenarios on week 52 primary endpoint



Global Null

Local Null

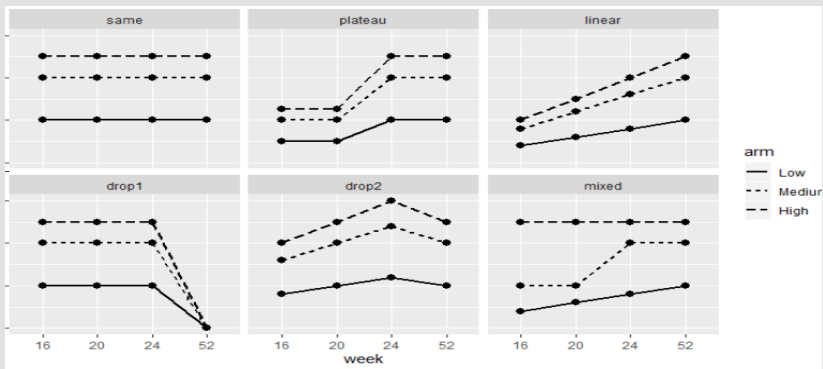
SIMULATION SCOPE – NUISANCE PARAMETERS

Control Arm Response

- A meta-analysis was done on historical SLE studies
- Based on the meta-analysis result, control response rates of (30%, 40%, 50%) is considered the most plausible scenarios

Longitudinal Pattern

- Longitudinal response pattern with respect to W52 response



Accrual Rate

- Three scenarios are included to cover the plausible range of accrual rate
- For all simulations, enrollment is assumed to have a 10-week ramp-up period followed by a constant accrual rate

Concordance

- Concordance of SRI-4 Response Between Adjacent Visits

Scenario	Value											
Same concordance across visits and treatment arms												
#1	0.5											
#2	0.6											
#3	0.7											
#4	0.8											
#5	0.9											
Different across visits, but same across treatment arms												
#6	Week 16-20					Week 20-24				Week 24-52		
	0.8					0.8				0.5		
Same across visits, but different across treatment arms												
#7	Control Arm			Low Dose			Med Dose			High Dose		
	0.5			0.6			0.7			0.8		
Different across visits and treatment arms												
#8	Control Arm			Low Dose			Med Dose			High Dose		
	16-20	20-24	24-52	16-20	20-24	24-52	16-20	20-24	24-52	16-20	20-24	24-52
	0.5	0.5	0.5	0.6	0.6	0.5	0.7	0.7	0.6	0.8	0.8	0.7

SIMULATION ITERATIONS

Type I Error

- For type I error evaluation, the results were based on 100k simulations per scenario, which provides type I estimation accuracy of approximately ± 0.001 with 95% confidence

Power and Estimation / Bias

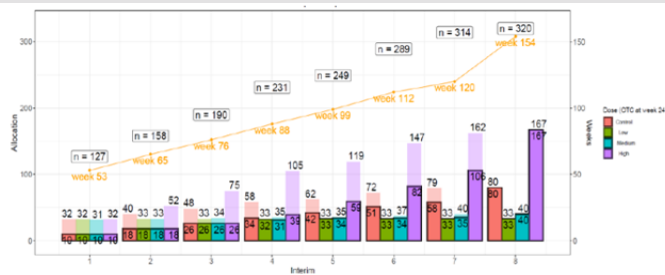
- For efficacy scenarios, the results are based on 10k simulations per scenario, which provides estimation accuracy of ± 0.01 for probabilities and ± 2 subjects for subject allocation with 95% confidence

Consistent with FDA adaptive design guidance

EXAMPLE TRIAL

- Example trials have been submitted in the simulation report and a shiny app has been created to visualize example trials with complete transparency

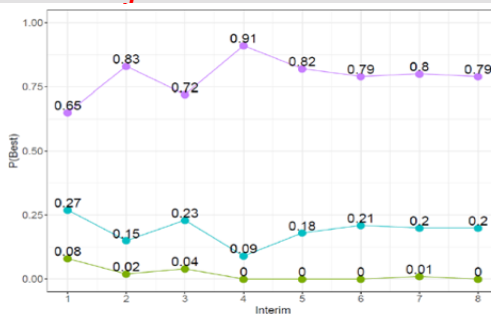
Sample Size and Allocation



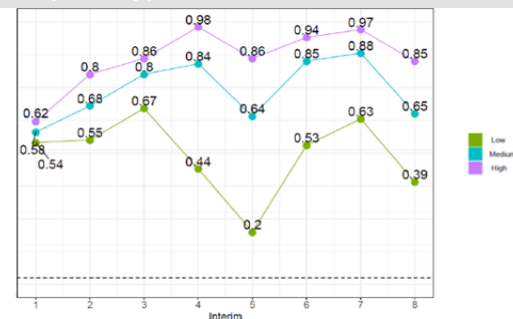
Legend number in black box shows the total sample size.

Prob. of Being Best Dose

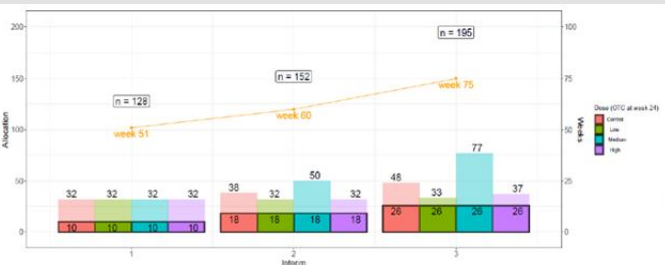
Example trial 1: a successful trial



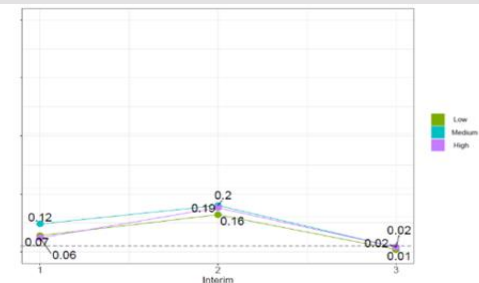
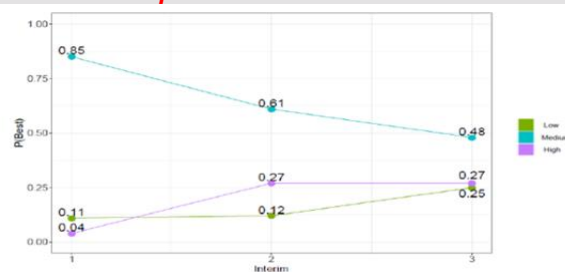
Posterior Probability (Futility)



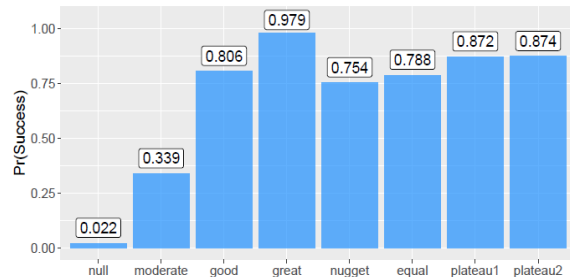
Example trial 2: a futile trial



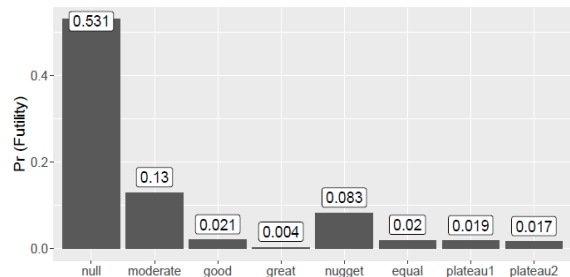
Legend number in black box shows the total sample size.



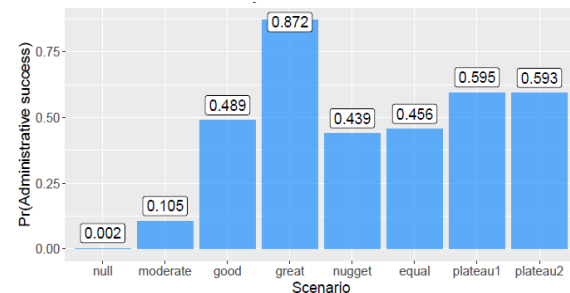
Probability of Success



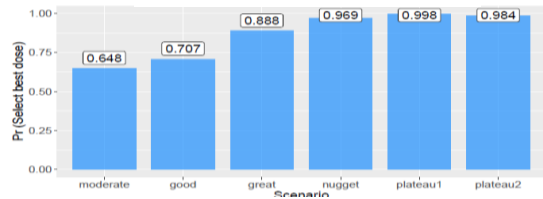
Probability of Futility



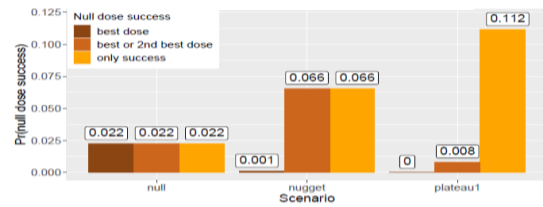
Probability of Administrative Success



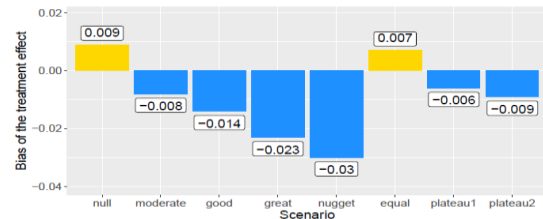
Probability of Selecting Best Dose



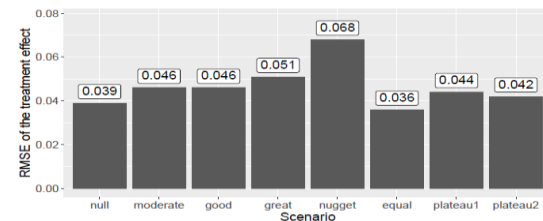
Probability of Null Dose Success



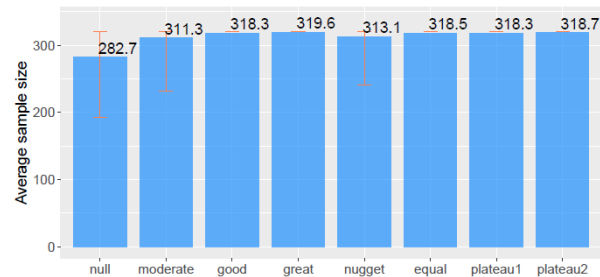
Bias of the Treatment Effect Estimate on Best Selected Dose



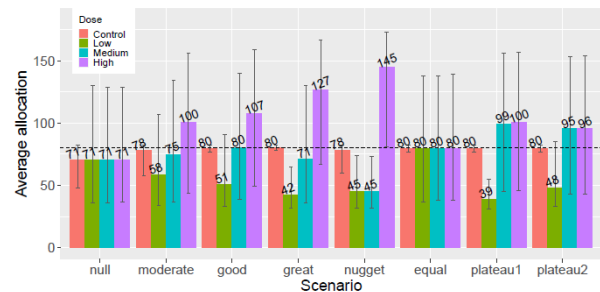
RMSE of the Treatment Effect Estimate on Best Selected Dose



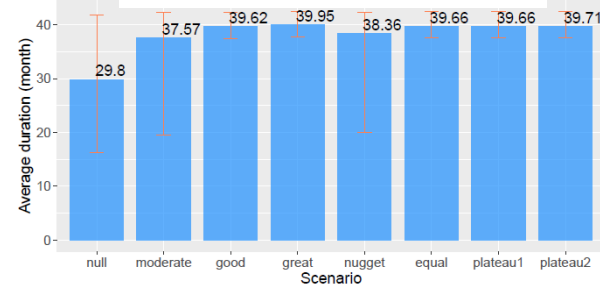
Average Total Sample Size in Each Scenario



Average Allocation by Dose in Each Scenario



Average Duration (Month) in Each Scenario



Closing Remarks

- ✓ PDUFA VI and 21st Century Cures Act provide exciting opportunities for industry to collaborate with regulatory agencies in promoting use of CIDs and providing the FDA an opportunity to communicate these advances publicly
- ✓ CIDs can help improve efficiency in clinical programs throughout the drug development cycle
- ✓ Our partnership with the FDA on the SLE CID Pilot Program should drive the development of a new treatment for lupus to address unmet need for patients
- ✓ We appreciate the FDA's efforts, significant contributions and feedback provided throughout the Pilot process

Acknowledgement

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THANK YOU!