



# Data & Statistical Sciences

## Comparison of Adaptive Seamless Phase 2/3 Designs with Dose Selection and Multiple Endpoints

Mandy Jin, PhD, AbbVie

2024 University of Pennsylvania Conference on Statistical Issues in Clinical  
Trials

abbvie



# Disclaimer

The comments provided here are solely those of the presenter and are not necessarily reflective of the positions, policies or practices of presenter's employers.

This publication was neither originated nor managed by AbbVie, and it does not communicate results of AbbVie-sponsored Scientific Research.

# Outline

## ❖ FDA Guidance:

- Adaptive Designs for Clinical Trials of Drugs and Biologics
- Optimizing the Dosage of Human Prescription Drugs and Biological Products for the Treatment of Oncologic Diseases

## ❖ Adaptive Seamless Design with Dose Selection

- A seamless design which can allow for the mid-Term modification
- Adaptive Graph-based Multiple Testing Procedure (agMTP) Based on Conditional Error Rate
- Dunnett-adjusted Adaptive Test based On Ranked Dose Responses

## ❖ Q&A

# Advantage of Adaptive Design

- ❖ An Adaptive Design uses accumulating data to decide how to modify aspects of the study design based on pre-specified criteria
- ❖ Advantages of Adaptive Design:
  - Decrease development time
  - Decrease sample sizes (costs)
  - Reduce patient burden
- ❖ Assessment of Adaptive Design:
  - Power gain/total sample size saving
  - Type I error
  - Less operation challenges
  - More .....

# FDA Guidance: Adaptive Designs for Clinical Trials of Drugs and Biologics (FDA 2019)

## ❖ Four key principals:

- Chance of erroneous conclusions should be adequately controlled
- Estimation of treatment effects should be sufficiently reliable
- Details of the design should be completely prespecified
- Trial integrity should be appropriately maintained

# Types of Adaptive Designs (Included in FDA Guidance)

- ❖ Group sequential designs
- ❖ Adaptations to the patient population
- ❖ Adaptations to endpoint selection
- ❖ Adaptations to the sample size
- ❖ Adaptations to patient allocation
- ❖ Adaptations to treatment arm selection (e.g., seamless design)

# FDA Guidance: Optimizing the Dosage of Human Prescription Drugs and Biological Products for the Treatment of Oncologic Diseases (FDA 2023)

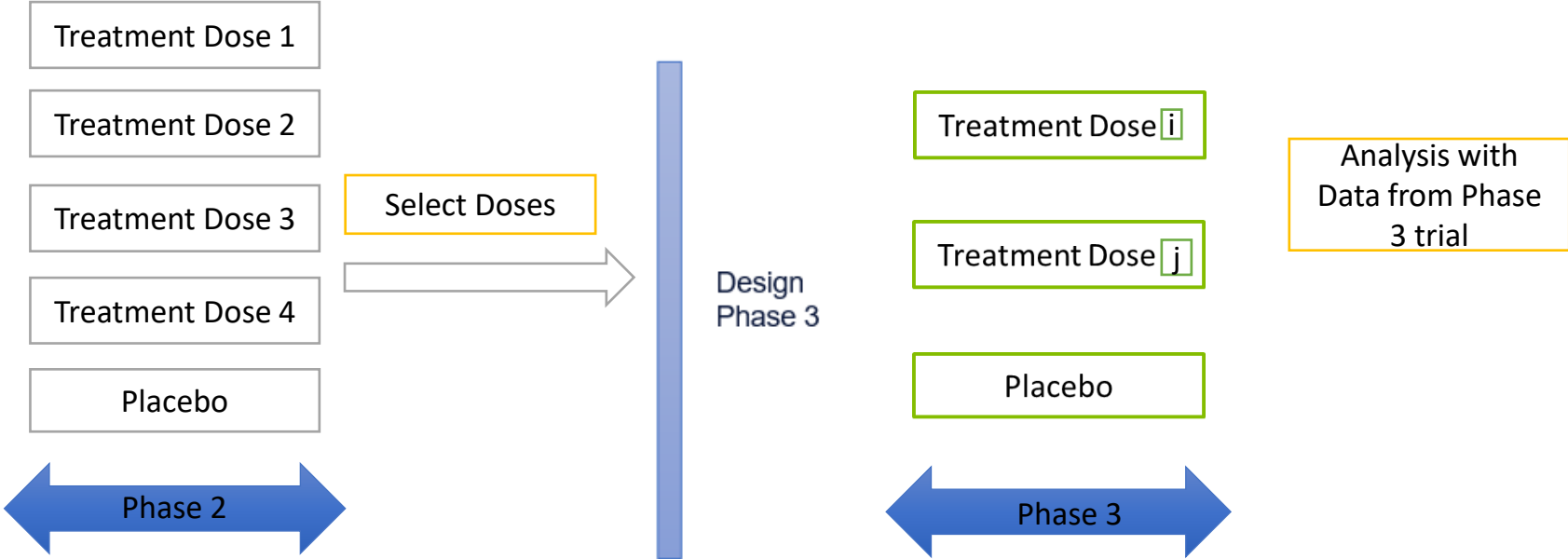
- ❖ Moves away from selecting the maximum tolerated dose (MTD)
- ❖ An optimal dose with similar efficacy and better safety profile based on comprehensive benefit risk assessment
- ❖ Should be applied to Adaptive Design with Dose Selection too

# Seamless Adaptive Seamless Phase 2/3 Designs with Dose Selection

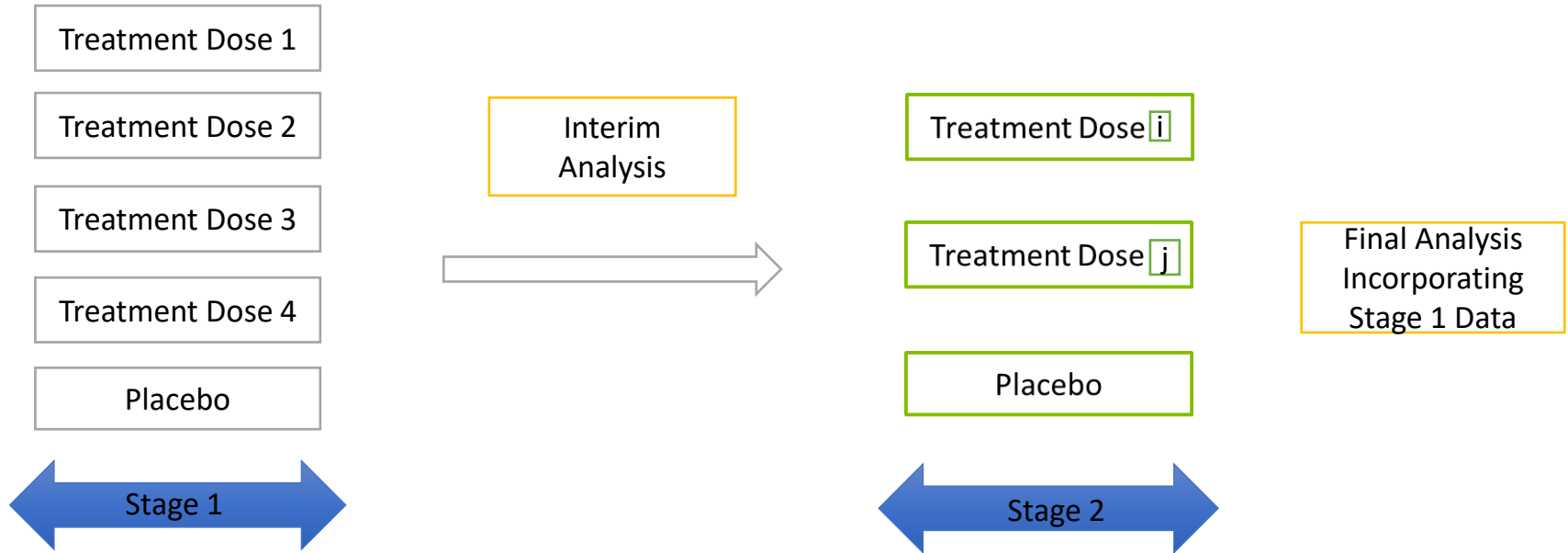
- ❖ Starts as a Phase 2, and select doses based on pre-specified criteria to expand to a confirmatory stage seamlessly
- ❖ Can incorporate both dose selection and confirmation of efficacy of a selected dose in one trial
- ❖ Proceeds in the same trial, but uses data from subjects enrolled in both stages in the final analysis
- ❖ Can expedite drug development compared to the conventional approach with conducting Phase 2 and Phase 3 trials sequentially



# Traditional Approach of Conducting Phase 2 and Phase 3 Trials Sequentially



# Adaptive Design with Dose Selection (Select One or Multiple Doses in Stage 2/Confirmatory Stage)



Criteria for dose selection should be based on efficacy and safety, and not necessary the dose with maximum efficacy.

# Inferential Adaptive Seamless Design with Dose Selection

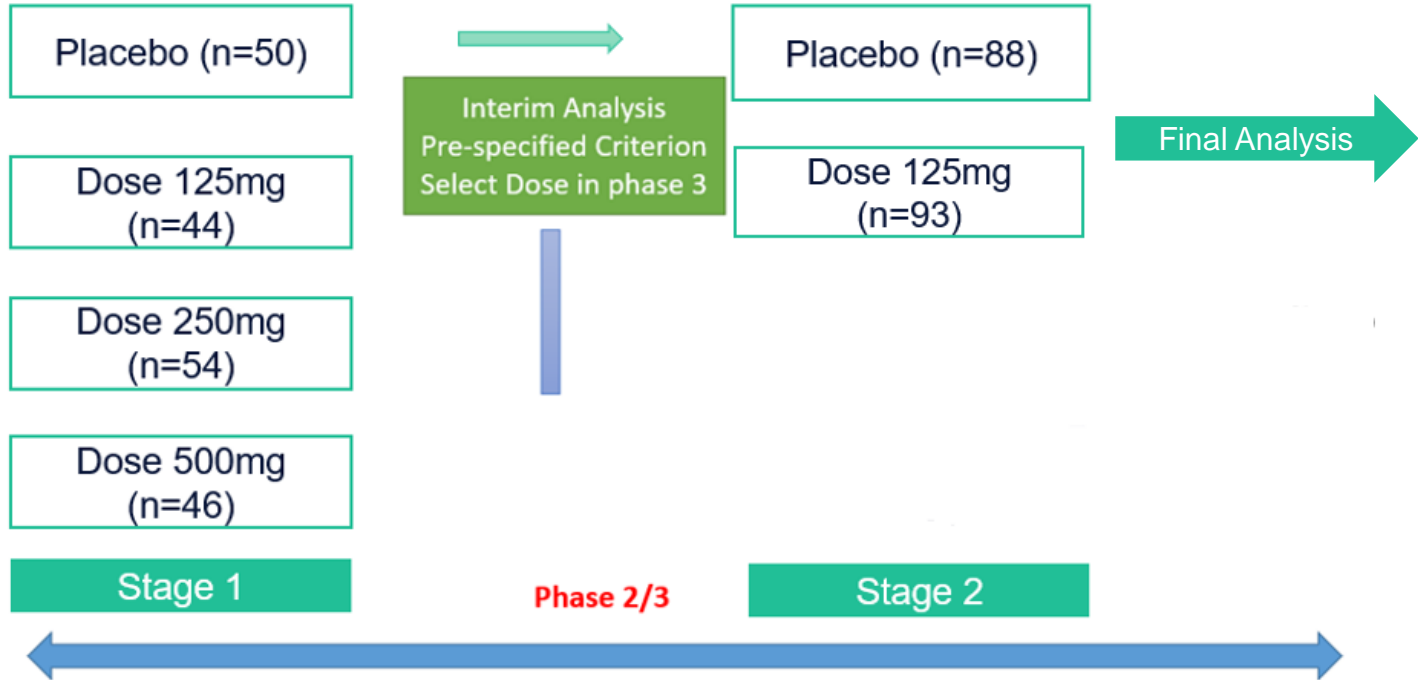
- ❖ An inferential seamless design combines the data from the Phase 2 component with the data from the Phase 3 component
- ❖ Requires control of Type I Error
- ❖ The Phase 2 component serves for dose selection, with typically one or two dose moving to Phase 3
- ❖ Dose selection can be based on surrogate endpoints which shows early efficacy signal, if long term endpoints are not mature
- ❖ Multiplicity adjustment should be applied if there are multiple hypotheses in the final analysis (multiple doses and/or multiple endpoints)

# Case Studies

# Case Study 1: ADVENT Trial- An Anti-Diarrhea Therapy in Patients with HIV Disease

- ❖ Chaturvedi, Antonijevic, and Mehta, 2014
- ❖ A two-stage seamless adaptive design
- ❖ Assess the efficacy and safety of three doses of crofelemer (125, 250, 500 mg) taken orally twice daily against placebo
- ❖ Dose selected at interim analysis with the pre-specified criteria based on efficacy and safety evaluations
- ❖ The first trial using this type of seamless adaptive clinical trial design that led to an FDA approval

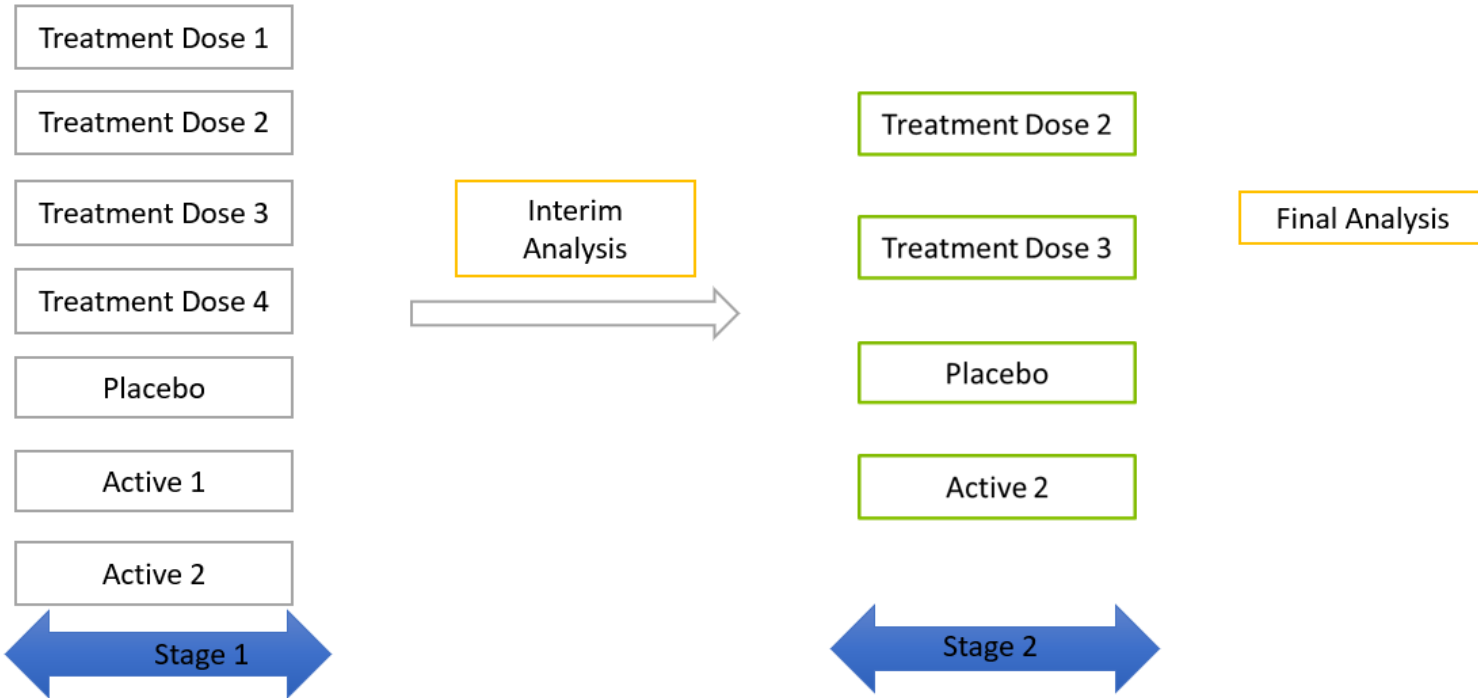
# ADVENT Trial



## Case Study 2: INHANCE Study for COPD

- ❖ A seamless adaptive Phase 2/3 : Sugitani and Bretz, 2016
- ❖ Multiple arms in Stage 1 (Phase 2), select 2 doses of indacaterol, active control, placebo in Stage 2 (Phase 3)
- ❖ Dose selected at interim analysis with the pre-specified criteria for efficacy and safety
- ❖ Mentioned in an earlier example

# INHANCE Study

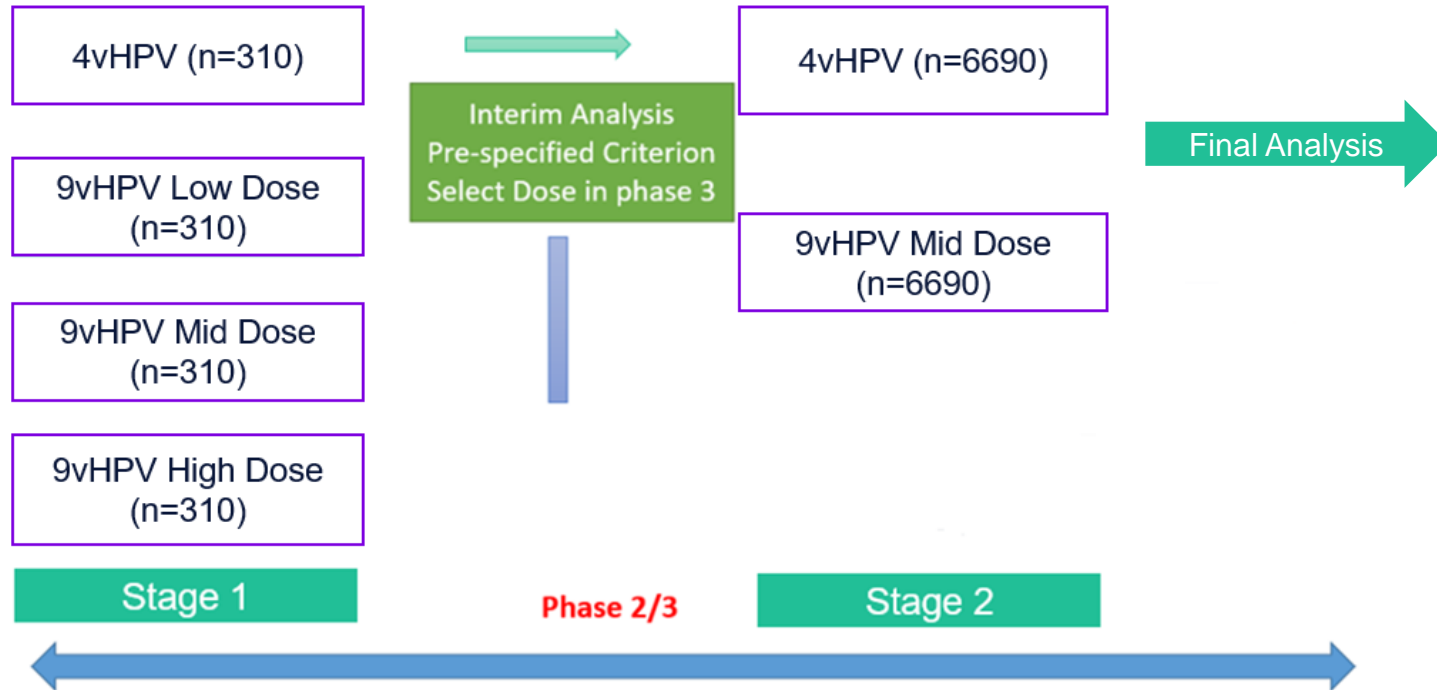




# Case Study 3: The 9-valent HPV Vaccine Trial (An Example in FDA Guidance)

- ❖ Chen, Gesser, and Luxembourg, 2015
- ❖ A seamless Phase 2b/3 design
- ❖ In Phase 2b stage, subjects randomized to three doses of the 9vHPV vaccine or the 4vHPV (active control)
- ❖ 9vHPV Mid dose was selected at interim analysis based on phase 2b data for comparative immunogenicity and safety data

# 9-valent HPV Vaccine Trial



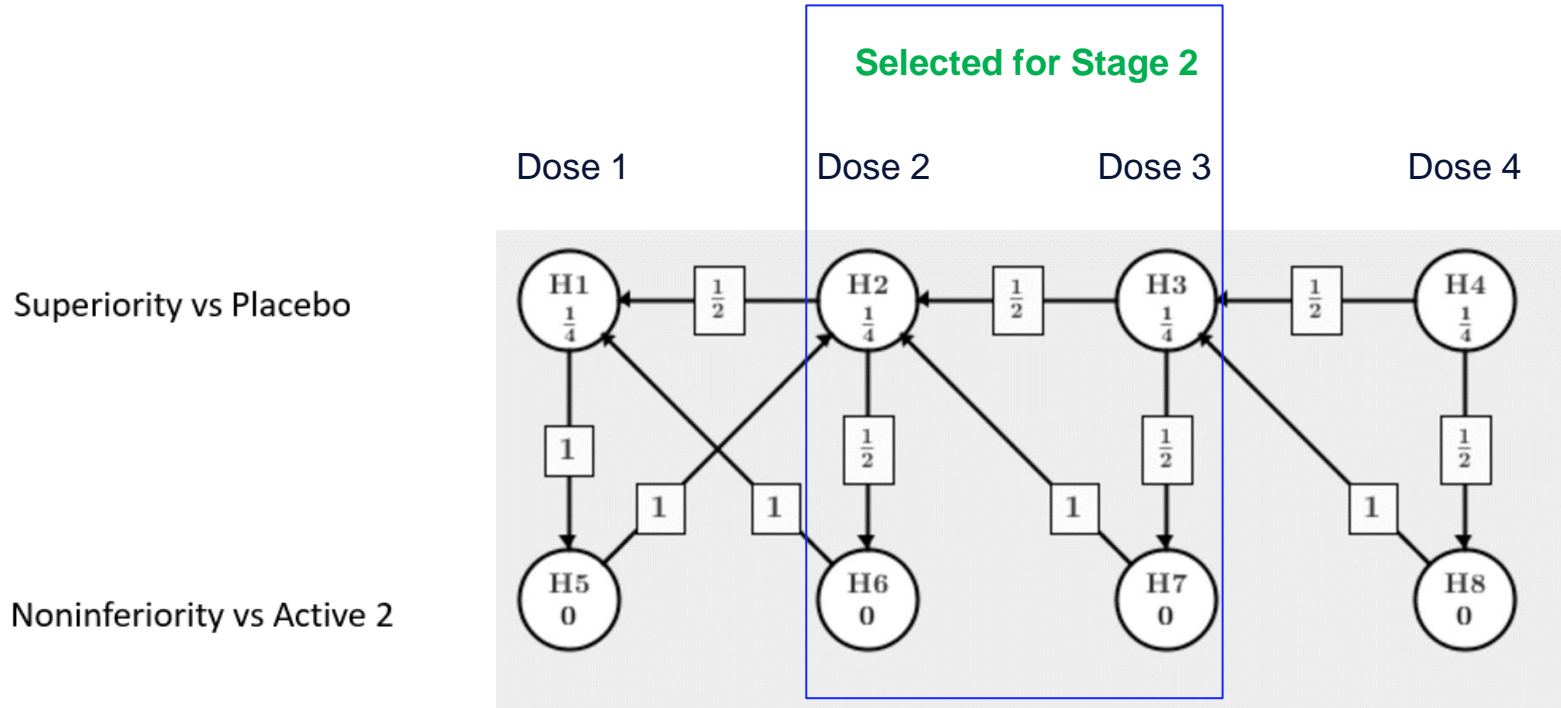
# Test Procedure in Adaptive Design with Dose Selection

- ❖ In many-to-one comparison:  $H_i: \mu_i = \mu_0, i = 1, \dots, m \text{ doses}, I = \{1, \dots, m\}$
- ❖ Global hypothesis:  $H_I = \cap_{i \in I} H_i$
- ❖ After IA, select doses  $J \subset I$ , final tested hypotheses is  $H_J = \cap_{j \in J} H_j$
- ❖ Multiple testing procedure or adjustment should be applied to  $H_J$  in the adaptive design setting with dose selection

# Design 1: A Seamless Design with Mid-Term Modification

- ❖ A seamless design by Sugitani and Bretz, 2016
- ❖ A multi-stage design ( $T \geq 2$  stages) with  $m$  doses ( $m$  hypotheses:  $H_1 \dots H_m, I = \{i \leq m\}$ )
- ❖ Example for a 2-stage design, at stage  $t \leq 2, p_{j,t}, H_j, j \leq m$ 
  - Inverse normal combination test for hypothesis  $j$  is:
  - $Z_{j,t} = [\sqrt{\eta_1} \phi^{-1}(1 - p_{j,1}) + \dots + \sqrt{\eta_t} \phi^{-1}(1 - p_{j,t})] / \sqrt{E_t} (E_t = \eta_1 + \dots + \eta_t)$
  - Drop a set of doses  $D_t \subset I$  (e.g., drop some doses, and select doses  $I \setminus D_t$ )
  - Inverse normal combination test:  $q_{j,t} = 1 - \phi(Z_{j,t})$  for  $j \in I \setminus D_t$
  - At Stage  $t, q_{j,t}$  is compared with incremental spent level  $\alpha_{j,t}^*(\alpha w_j(J))$  for  $H_j$
  - Remove unselected doses from the Graph-based multiple testing procedure, keep the original weights and transition fractions
- ❖ The design controls Type I error

# Testing Strategy in Sugitani and Bretz 2016



# Design 2: Adaptive Graph-based Multiple Testing Procedure (agMTP) Based on Conditional Error Rate

- ❖ Conditional Error rate: Koenig, Brannath, Bretz, Posch, 2008
- ❖ Corresponding conditional error function of  $\phi$  conditioning on the first-stage data ( $\boldsymbol{x}_1$ ) for each treatment group is given by

$$A(\boldsymbol{x}_1) = E_H(\phi = 1|\boldsymbol{x}_1)$$

$\phi = 0$  or  $1$ : *accept or reject H*

- ❖ After IA, option to
  - Complete the trial as initially planned, or select doses based on pre-specified criteria
  - Choose any other test for  $H$  at level  $A(\boldsymbol{x}_1)$  for the second-stage
- ❖ If adaptations are performed, the null hypothesis  $H$  is rejected based on the second-stage p-value  $q$  whenever  $q \leq A(\boldsymbol{x}_1)$

# Design 2: agMTP Based on Conditional Error Rate

- ❖ agMTP: Klinglmueller, Posch, and Koenig, 2014
- ❖ A multi-stage design ( $T \geq 2$  stages) with  $m$  doses (hypotheses:  $H_1 \dots H_m, I = \{i \leq m\}$ )
- ❖ Example for a 2-stage design, at stage  $t \leq 2, p_{j,t}, H_j, j \leq m$

➤ Conditional Error Rate for each intersection hypothesis  $H_J, J \subset I, .$  for all  $j \in J$

$A_{j,J}(w_{j,J}\alpha) = E_{H_J}[\mathbf{1}\{p_{j,1} \leq w_{j,J}\alpha\} | \chi_1], B_J(\alpha) = \sum_{j \in J} A_{j,J}(w_{j,J}\alpha); (\chi_1 \text{ denoted Stage 1 data})$

➤ Second stage p-values  $\mathbf{q} = (q_1, \dots, q_m)$

✓ Dose  $j$  selected,  $q_j = p_{j,2}$

✓ Dose  $k$  not selected,  $q_k = 1$

➤ Define  $v_J = (v_{1,J}, \dots, v_{m,J})$  with  $v_{i,J} = 0$  for all  $i \notin J$  and  $\sum_{j \in J} v_{j,J} \leq 1$  ( $v_J$  may be chosen arbitrarily for each  $J \subset I$  but the choice of weights will have an impact on the power of the procedure)

➤ Adaptive test: 
$$\tilde{\phi}_J(q, B_J) = \begin{cases} \max_{j \in J} \mathbf{1}(q_j < v_{j,J} B_J), & \text{if } B_J \leq 1 \\ 1, & \text{otherwise} \end{cases}$$

➤ Closed test procedure that rejects  $H_i$  ( $i \in J$ ):  $\tilde{\psi}_i = \min_{J \subset I, j \in J} \tilde{\phi}_J(\mathbf{q}, B_J)$

abbvie

➤ The design strongly controls FWER

# An Example in Klinglmueller, Posch, and Koenig, 2014

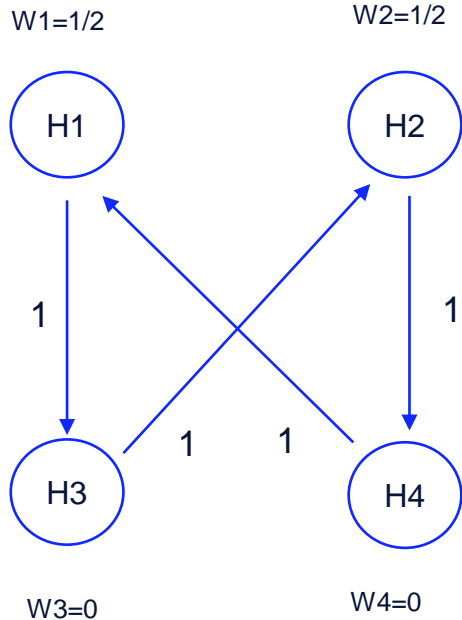
- ❖ Dose 1, 2 and Control
- ❖ Two endpoints E1 and E2

Before IA

E1

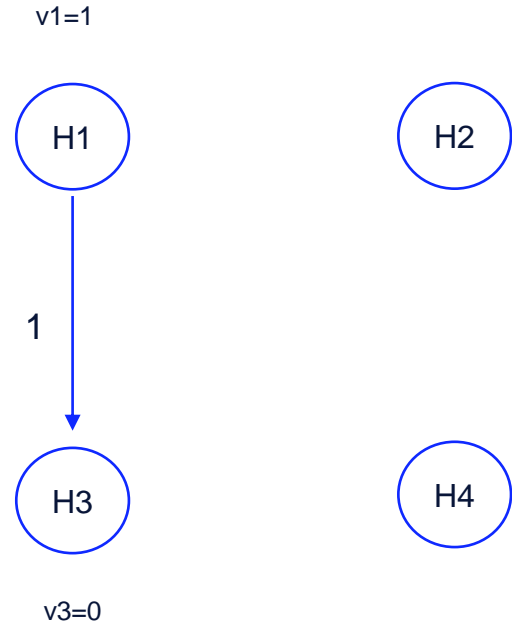
E2

abbvie



Pre-planned Adaptive Test After IA

Adaptive Test





# Design 3: Dunnett-adjusted Adaptive Test based On Ranked Dose Responses

- ❖ Dunnett-adjusted Adaptive Test based On Ranked Doses: Wang et al. 2023
- ❖ Extend Wang et al. 2023 to a design which can select multiple doses
- ❖ A multi-stage design ( $T \geq 2$  stages) with  $m$  doses ( $m$  hypotheses:  $H_1 \dots H_m$ )
- ❖ Example for a 2-stage design, at stage  $t \leq 2$ ,  $p_{j,t}$ ,  $H_j, j \leq m$ , ranked Stage 1 data
  - $X_{(1),1} \leq X_{(2),1} \leq \dots \leq X_{(m),1}$

# Design 3: Dunnett-adjusted Adaptive Test based On Ranked Dose Responses

❖ Dose  $s$  is selected based on the ranked response  $X_{(r),1}$  based on Stage 1 data,  $s \in J \subset I$

➤  $X_{(r),1} = X_{s,1}$

➤  $r$  is not necessarily the highest rank or highest dose performance

➤ Since  $X_{(r),1}$  on the selected dose  $s$  does not follow standard normal distribution, Dunnett adjustment to adjusted statistics following standard normal distribution

$$p_{s,adj}^1 = \text{Adjust\_Dunnett}(1 - \Phi(X_{s,1}))$$
$$X_{s,adj}^1 = \Phi^{-1}(1 - p_{s,adj}^1) \sim N(0,1)$$

➤ Wang et al (2023) improved the efficiency using Dunnett adjustment based on selected rank  $r$

➤  $X_{s,adj} = \sqrt{w_1}X_{s,adj}^1 + \sqrt{w_2}X_{s,2}$  ( $X_{s,2}$  is the Stage 2 data for Dose  $s \in J$ )

➤ Plan a graph-based multiple testing procedure on doses in  $J$

❖ The design by Wang et al. (2023) can be extended to a design which selects more than one doses and testing multiple endpoints

❖ The design with all the possible modifications controls Type I error

# Extensions to Adaptive Design with Dose Selection and Multiple Endpoints

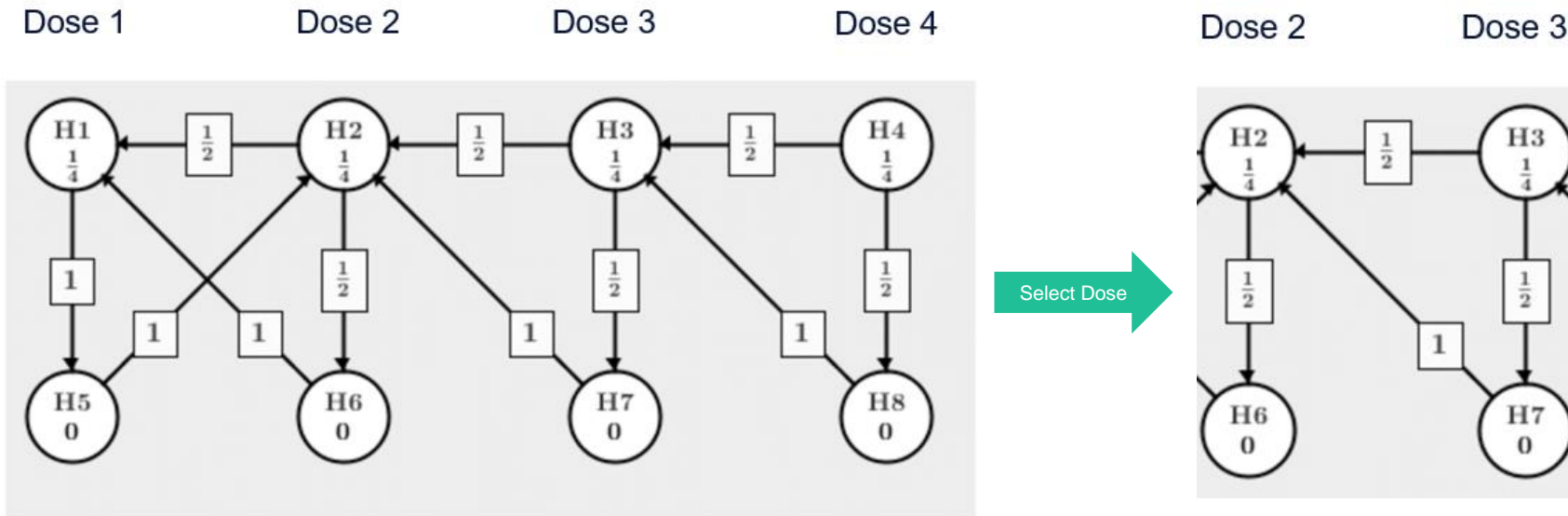
- ❖ The design can be extended to multiple doses and multiple endpoints
- ❖ Dose is selected based on Primary or Surrogate Endpoint E1, which shows short-term efficacy
- ❖ Advantage is that the endpoint with long term effect is not mature at IA
- ❖ E1 is usually correlated with another Endpoint E2 (such as clinical outcome)
- ❖ At the final analysis, plan a multiple testing procedure on the selected doses with Endpoints E1, E2, ...
- ❖ Design 1, 2, and 3 can be easily extended to a design for this goal

# Adaptive Design with Dose Selection Based on Primary Endpoint (For Simulation Purpose)

- ❖ Dose 1, 2, 3, 4, and Control
- ❖ Two endpoints E1 and E2, select 2 doses at IA (40% info) based on F1

Design 1

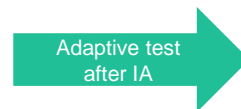
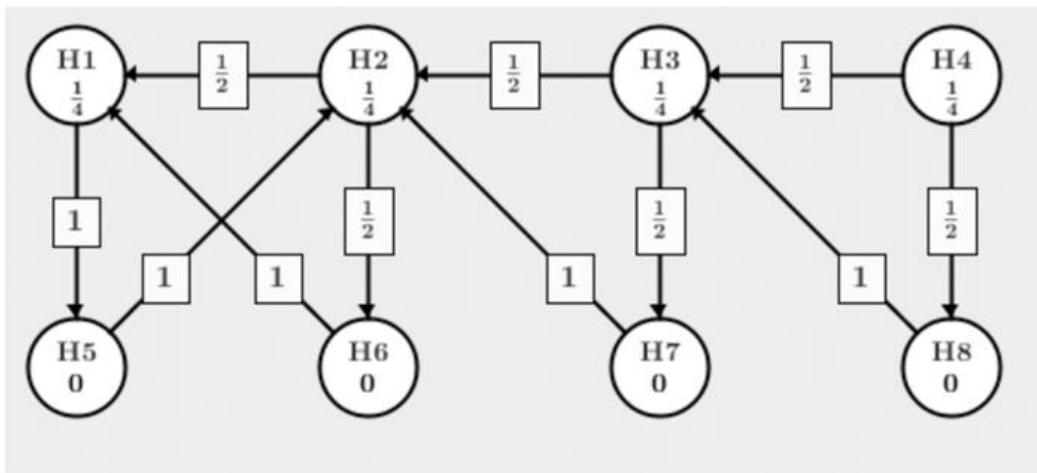
Selected for Stage 2



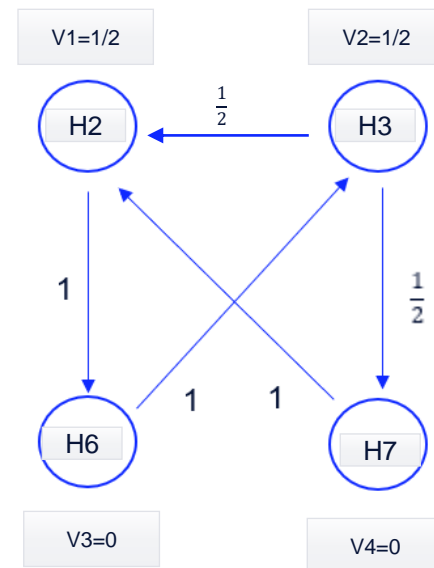
# Adaptive Design with Dose Selection Based on Primary Endpoint (For Simulation Purpose)

- ❖ Dose 1, 2, 3, 4, and Control
- ❖ Two endpoints E1 and E2, select 2 doses at IA (40% info) based on E1
- ❖ Calculate Conditional Error based on E1 and E2 using original planned graph
- ❖ Adaptive graph-based Multiple Test Procedure test after IA

Dose 1                      Dose 2                      Dose 3                      Dose 4



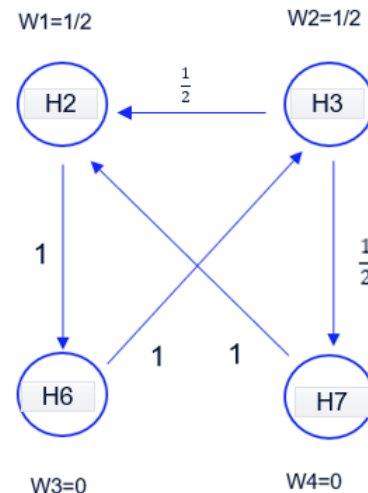
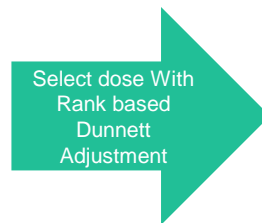
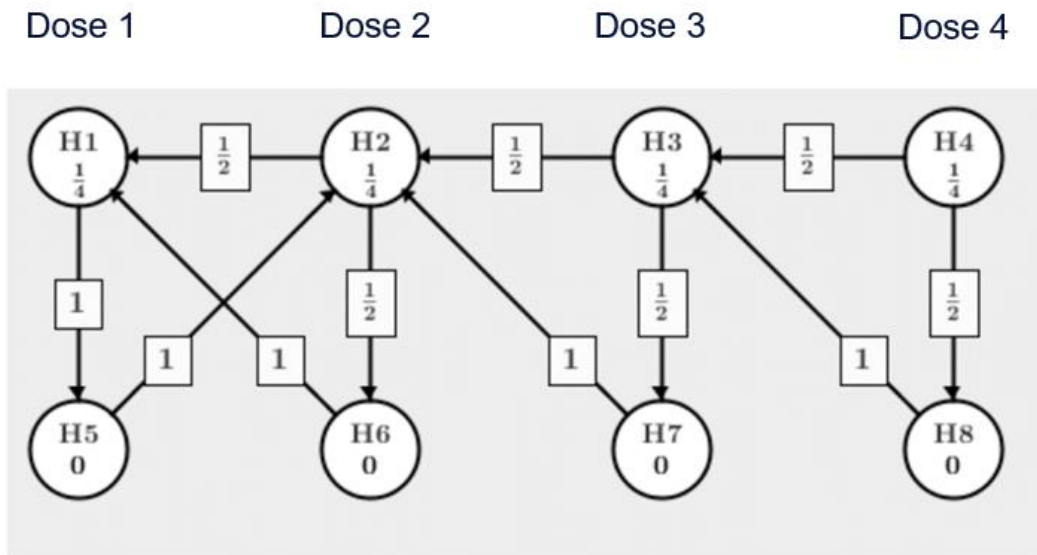
Design 2



# Adaptive Design with Dose Selection Based on Primary Endpoint (For Simulation Purpose)

- ❖ Dose 1, 2, 3, 4, and Control
- ❖ Two endpoints E1 and E2, select 2 doses at IA (40% info) based on E1
- ❖ Dunnett-adjusted test statistics from IA

## Design 3



# Notes to the Simulation Setting

- ❖ The selected doses can be any 2 doses, not necessarily the highest ranked doses
- ❖ Since the selected dose may or may not have the highest efficacy signal, multiplicity adjustment is required
- ❖ More simulation scenarios are considered to select 1 dose
- ❖ Another simulation setting with 3 doses are considered

# Type I Error and Power with 4 Doses

	Method	Select 2 Doses (Ranks)			Select 1 Dose (Rank)			
		(3,4)	(2,3)	(1,2)	4	3	2	1
Type I (%)	Design 1	2.0	0.8	0.4	1.6	0.8	0.3	0.1
	Design 2	2.3	0.88	0.58	2.0	1.0	0.7	0.2
	Design 3	1.9	0.5	0.2	1.8	0.9	0.3	0.1
Power (%)	Design 1	90.5	82.6	73.1	84.8	74.7	64.9	51.9
	Design 2	91.1	86.6	81.9	89.6	83.2	78.7	71.8
	Design 3	90.4	82.5	73.2	89.2	82.6	77.5	73.2

Design 1: Design 1: A Seamless Design with Mid-Term Modification

Design 2: Adaptive Graph-based Multiple Testing Procedure (agMTP) Based on Conditional Error Rate

Design 3: Dunnett-adjusted Adaptive Test based On Ranked Dose Responses



# Type I Error and Power with 3 Doses

	Method	Select 2 Doses (Ranks)		Select 1 Dose (Rank)		
		(2,3)	(1,2)	3	2	1
Type I (%)	Design 1	2.1	0.6	1.4	0.5	0.18
	Design 2	2.3	1.1	2.3	0.9	0.5
	Design 3	1.9	0.3	2.4	0.82	0.2
Power (%)	Design 1	89.9	80.3	85.5	74.7	59.1
	Design 2	91.3	84.8	88.5	81.6	74.5
	Design 3	86.6	79.3	87.8	81.3	76.7

Design 1: Design 1: A Seamless Design with Mid-Term Modification

Design 2: Adaptive Graph-based Multiple Testing Procedure (agMTP) Based on Conditional Error Rate

Design 3: Dunnett-adjusted Adaptive Test based On Ranked Dose Responses

# Reference

- FDA Guidance, 2019: Adaptive Designs for Clinical Trials of Drugs and Biologics.
- FDA Guidance, 2023. Optimizing the Dosage of Human Prescription Drugs and Biological Products for the Treatment of Oncologic Diseases.
- Koenig, Brannath, Bretz, Posch (2008). Adaptive Dunnett tests for treatment selection. *Statistics in Medicine*; 27(10):1612–1625.
- Klinglmueller, Posch, and Koenig (2014). Adaptive graph-based multiple testing procedures. *Pharmaceutical Statistics*; 345-356
- Sugitani T, Bretz F, Maurer W (2016). A simple and flexible graphical approach for adaptive group-sequential clinical trials. *Journal of Biopharmaceutical Statistics*.
- Wang, Chen, Chu, Fan, Chan. (2023). A rank-based approach to improve the efficiency of inferential seamless phase 2/3 clinical trials with dose optimization. *Contemporary Clinical Trials*.
- Chaturvedi, Antonijevic, and Mehta, “Practical Considerations for a Two-Stage Confirmatory Adaptive Clinical Trial Design and Its Implementation: ADVENT Trial”, 2014, Chapter 20, Book: Practical Considerations for Adaptive Trial Design and Implementation by He W, Pinheiro J and Kuznetsova O.
- Chen, Gesser, and Luxembourg, “A Seamless Phase IIB/III Adaptive Outcome Trial: Design Rationale and Implementation Challenges”, *Clinical Trials*, 2015, 12(1):84–90.

**Thank You!**