

Utility and Pitfalls of Dose Ranging Trials with Multiple Study Objectives Fixed or Adaptive*

Sue-Jane Wang, Ph.D.

Office of Biostatistics, Office of Translational Sciences
Center for Drug Evaluation and Research, US FDA

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Clinical/Statistical review teams/divisions/offices adaptive design case sharing education, discussion, contribution

Outline

- ◆ What are study objective(s)
- ◆ Approaches: fixed vs adaptive
- ◆ Randomization strategies
- ◆ Utility and Pitfalls
- ◆ Concluding Remarks

Study Objective(s)

- ◆ Detect dose-response (POC)
- ◆ Identify clinical relevant response (CRR) \cong MCID ???
- ◆ Select dose(s) for Phase III planning
- ◆ Estimate the dose-response

- ◆ Screen drug for anti-disease activity
- ◆ Extend knowledge of toxicology and pharmacology of drug to hopefully benefit patients

Conventional Fixed Designs

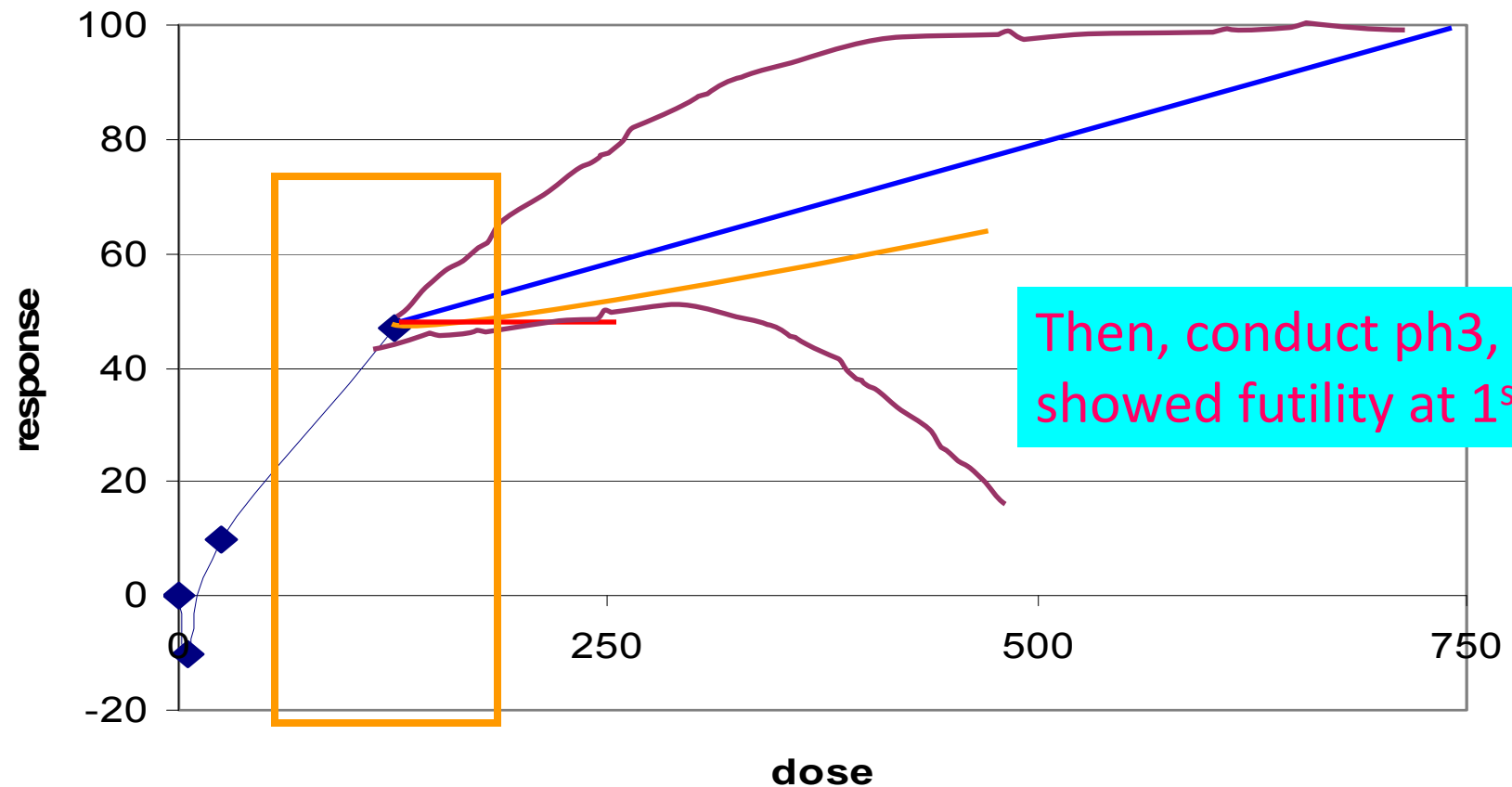
- ◆ Fixed randomization ratio
 - ◆ K-dose groups +/- placebo & dosing frequency
 - ◆ Sample size to detect dose-response, if significant, then, estimate target effect size
- ◆ Design fixed and analysis adaptive (selection)
 - ◆ MCPMod
 - ◆ Postulate a few plausible dose-response models
 - ◆ Dose-response characterization
 - ◆ Dunnett
 - ◆ Formally incorporates selection into decision process

Utility and Pitfall with Fixed Designs

- ◆ Learn from experience to design properly → can be A&WC
 - ◆ E2 = E3 with knowledge of successful similar developments
 - ◆ Sufficiently powered to detect T-effect → Dunnett's useful
 - ◆ Can also characterize DR relationship with good precision
- ◆ Learning mode without existing experience → Sufficient 'n' ?
 - ◆ What's endpoint? E2 = Early time point of E3; E2 <> E3
 - ◆ E2 correlated with E3? E2 predictive of E3 ?
 - ◆ Little prior to size the trial
- ◆ Model dose-response relationship – lack precision
- ◆ Dunnett can be inefficient
- ◆ Dose choice difficult, though may have apparent DR pattern

Danger of extrapolation without formal investigation

Figure. Dose-response curve for efficacy biomarker



Then, conduct ph3,
showed futility at 1st IA

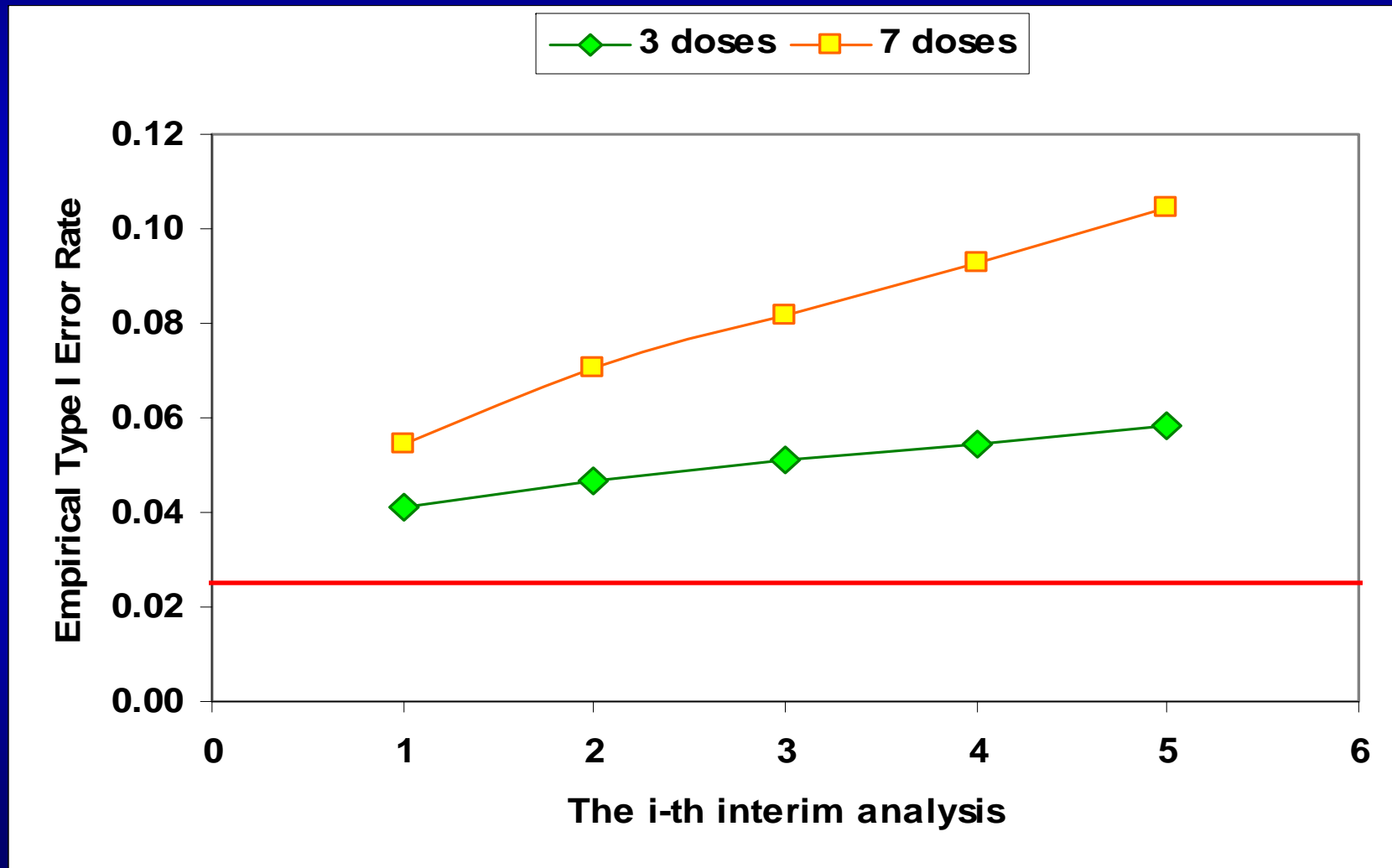
Utility and Pitfall with Hybrid Designs

- ◆ Selection, sequential, fixed randomization ratio both stages
- ◆ Single-stage randomization selection w/o pbo
 - ◆ Ranking and selection with prob (correct selection)
 - ◆ No hypothesis test, rely on existence of best dose
 - ◆ False+ (impressive arm) rate \uparrow as $K \uparrow$ unless large difference
 - ◆ Left to mandatory follow-on in ph3 where α -controlled
- ◆ Two-stage randomization selection \rightarrow (prelude of adaptive)
 - ◆ control conditional type I error when w/o pbo
 - ◆ strong control of overall type I error when include pbo
- ◆ SPRT to identify 1-arm $> C$ by elimination of inferior arms
 - ◆ $pr(\text{correct selection} | H_0)$ vs $pr(\text{correct selection} | H_a)$
 - ◆ also, strong control of overall type I error; allow stop futile trial?
- ◆ Properly recognized as Phase 2 trials

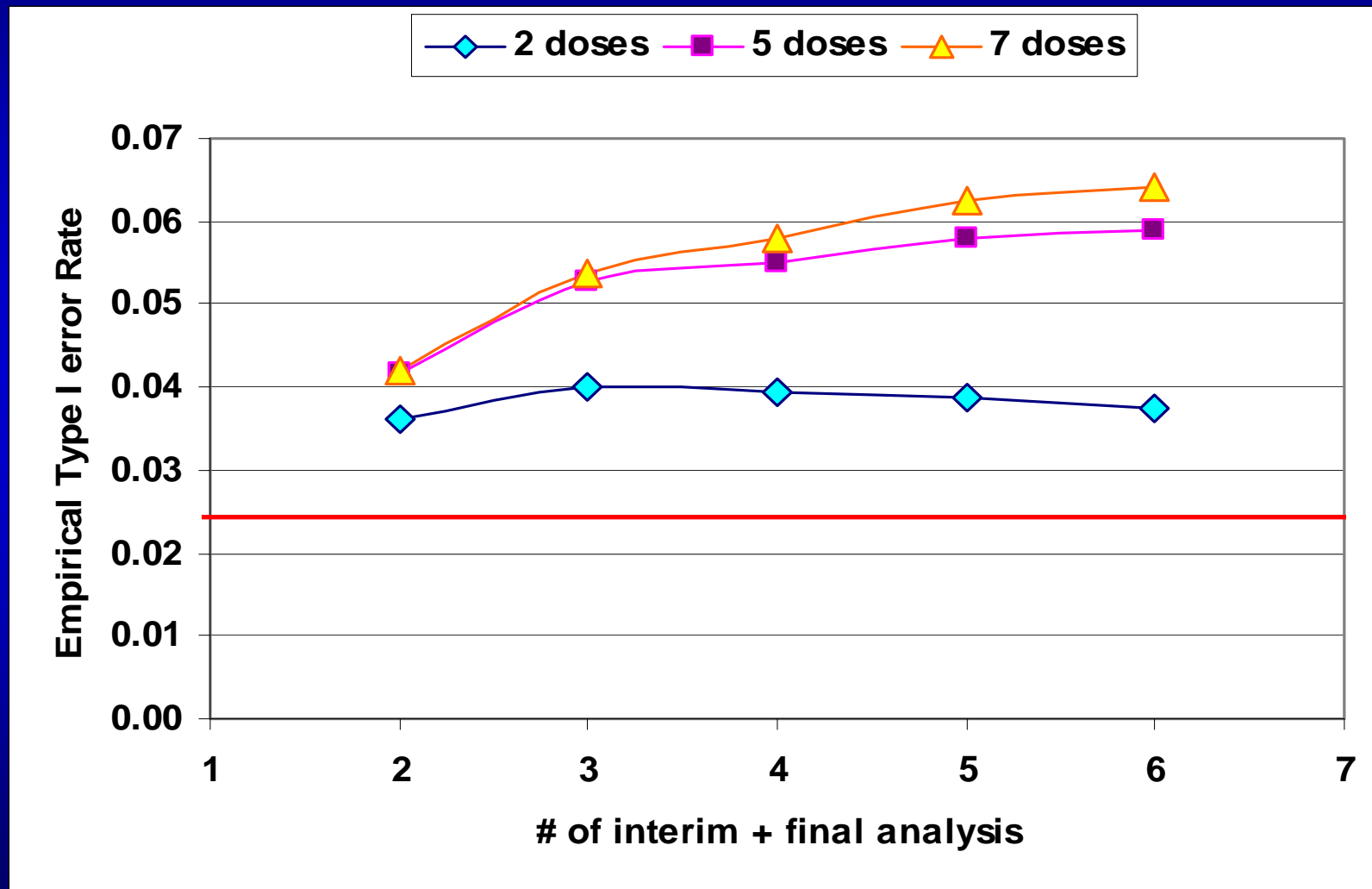
When ignoring multiplicity

- ◆ The type I error rate based on the adaptation process as a two-stage randomized selection in group sequential setting
- ◆ Selection based on clinical endpoint
- ◆ Selection based on biomarker for testing clinical endpoint

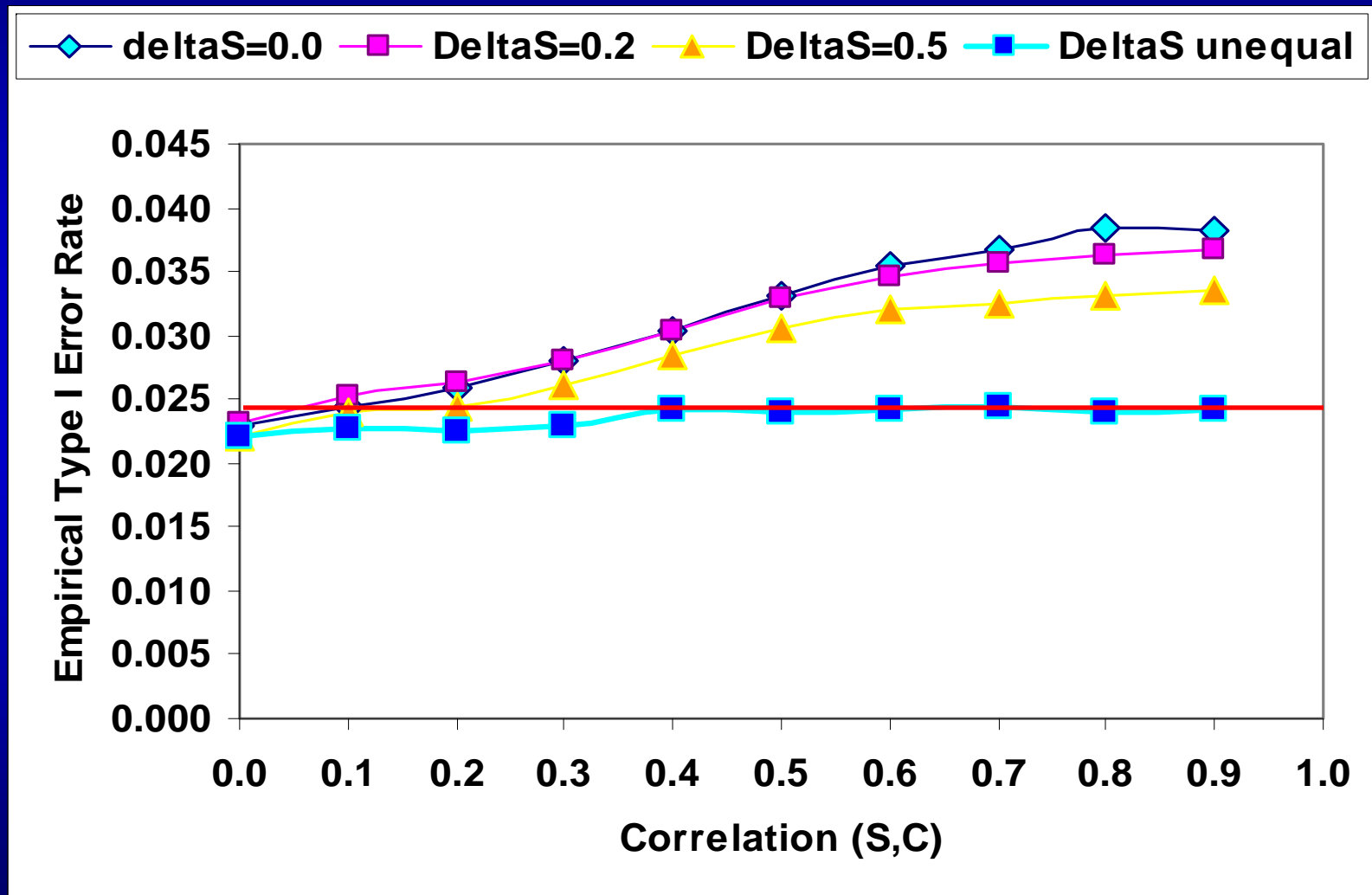
Select among doses most promising at k-th interim, e.g. $\text{Max}(Z_1, \dots, Z_k)$



Select among TRTs most promising at k -th interim, if $Z_{\max} - Z_j > 1$ for all j



Select b/t 2-Doses at $\tau=0.5$ based on Efficacy Biomarker, if $D^*Z1 < Z2$, select T2, ow T1



Adaptive Dose Ranging Designs #1

(White Paper, 2007)

- ◆ GADA: Bayesian dynamic dose allocation (non-para model)
- ◆ D-optimality adaptive allocation (parametric model)
- ◆ Design fixed, analysis adaptive
 - ◆ Dunnett's multi-arm test (select dose at study end)
 - ◆ MCPMod (identify dose-response model at study end)
 - ◆ Multiple Trend Tests (from a class of sigmoid Emax model, U, L, M DR)
 - ◆ Strong control of overall type I error for its primary study objective
- ◆ Bayesian model averaging; Non-para dose-response modeling

Adaptive Dose Ranging Designs #2

(2010)

- ◆ **AMCPMod**: Optimal design + Model (uncertainty) + Bayesian statistics (Bornkamp, Bretz, Pinheiro, Dette)
- ◆ **DCoD**: D-optimal followed by C-optimal design based on sigmoid Emax model (Dragalin, Padmanabhan)
- ◆ **IntR** – Bayesian design minimizing average variance of all LS-estimates for “interesting part” of dose-response curve (Miller)
- ◆ **MultObj** – Multi-objective optimal design incorporating 2nd order moments and based on inverse quadratic model (Smith)
- ◆ **T-Stat** – Dose-adaptive design: t-statistics (Patel, Perevozskaya)

Utility/Pitfall - Adaptive Dose Selection

- ◆ Properly recognized as Ph2 learning trials
- ◆ Attractive because
 - Interested in a Ph2-like sample size
 - Can learn from several more doses if response adaptive used
 - Better precision to estimate doses of greater interests
 - Efficient learning of DR relationship for faster decision making on dose selection, though DR may not be fully characterized
- ◆ Adaptive learning trial – other design elements, e.g., multiple patient (sub)populations, from small to large and vice versa, efficacy endpoint not understood? Biomarker – how predictive? effect size if primary endpoint known

Two-Stage Adaptive Dose Selection #1

Learning

Stage I (2 weeks?)

Dose 1

Dose 2

Dose ...

Dose ...

Dose 15

Placebo

Active Drug

Response adaptive randomization

Longitudinal modeling; Predictive Prob.

Validating

Stage II (24 weeks)

Dose S1, n2

Dose S2, n2

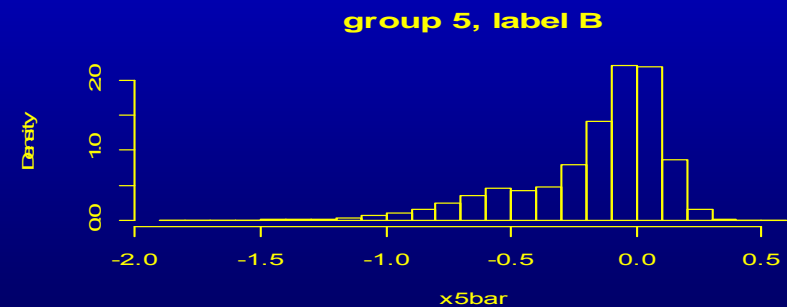
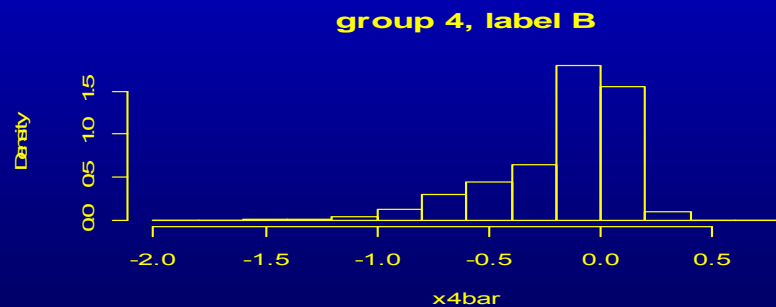
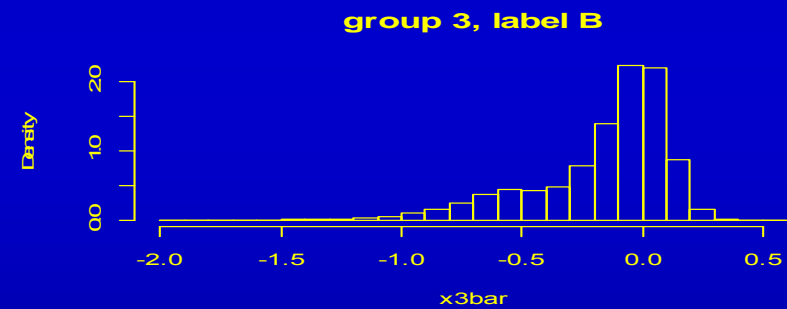
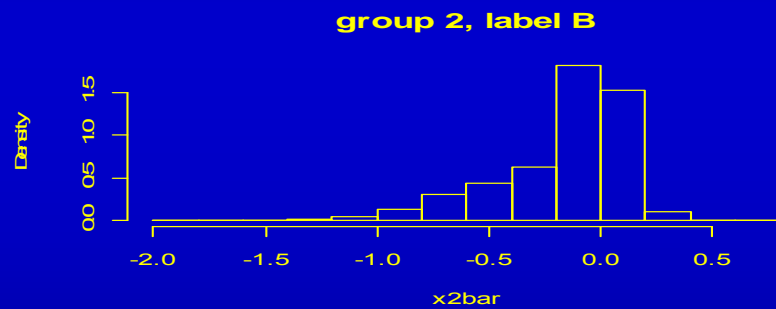
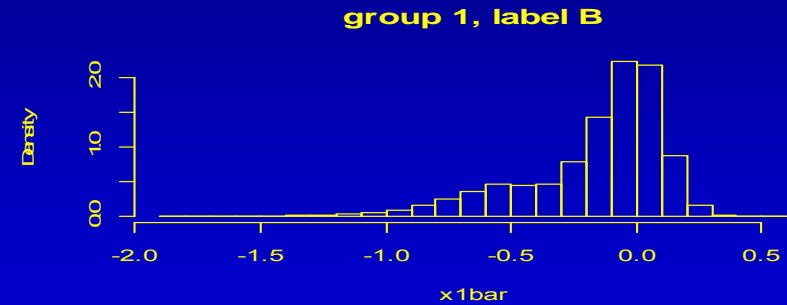
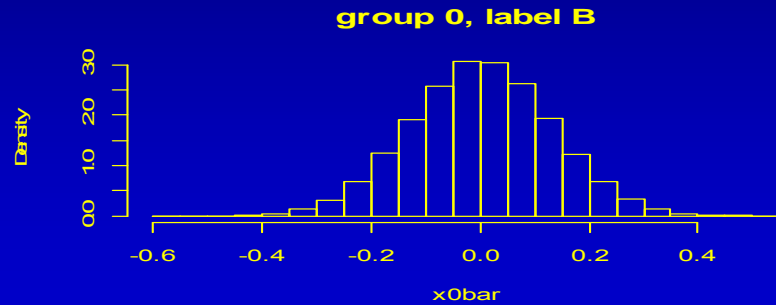
Placebo, n2

Active Drug, ?

Fixed randomization ratio

Final test statistic; simulated α -control

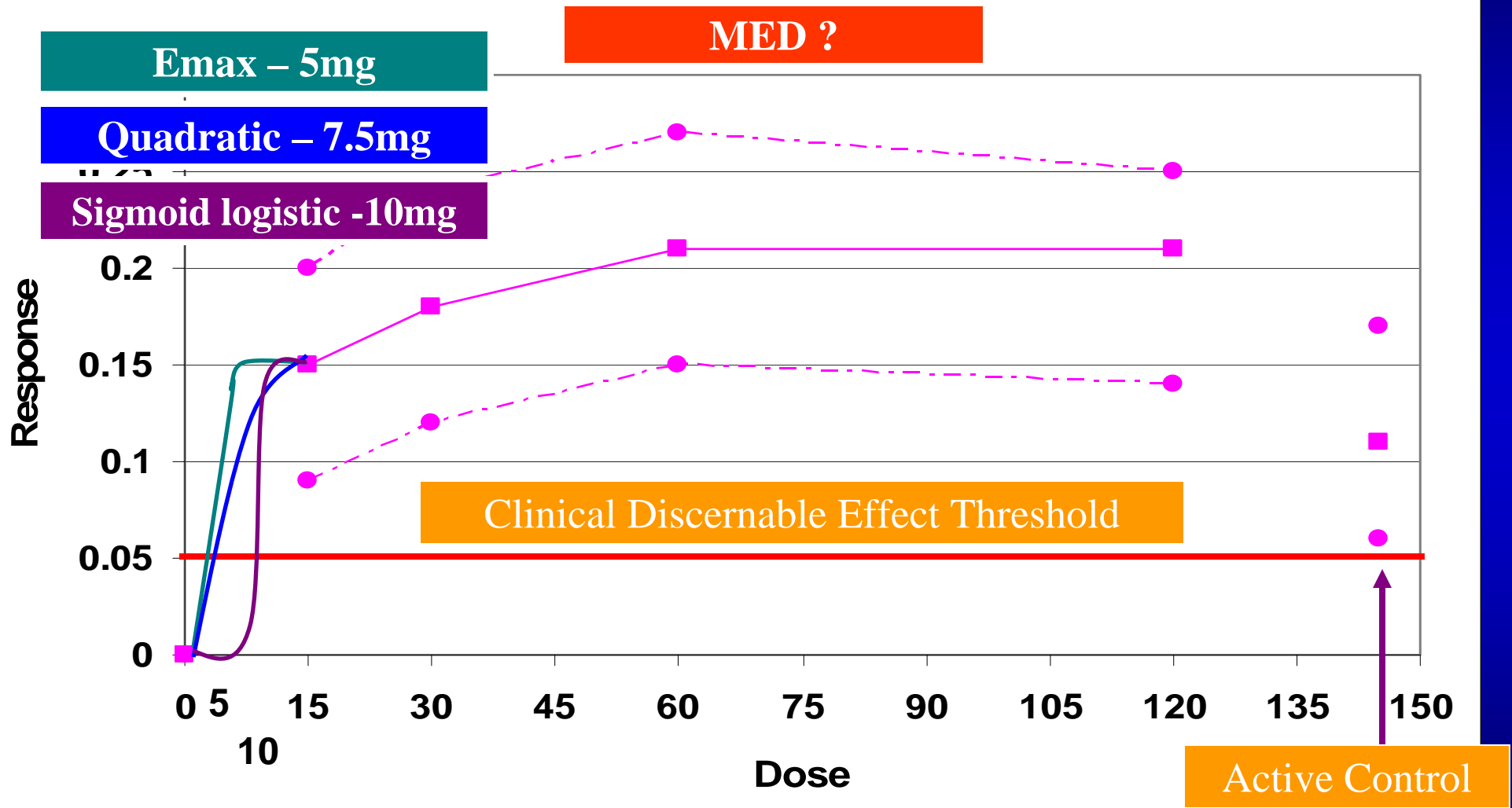
A Frequentist Version of response adaptive allocation



Discernable Effect

- Serve as a 'guide' in the exploratory stage
- Better than 'eye-balling visual' dose selection
- Can be very informative if safety issue is not a concern
- When benefit/risk profile can be a function of dosing frequency, dose-level
- Often efficacy biomarker, different from MCIE in confirmatory trial
- Even if clinical endpoint, MCIE is in the eyes of the beholder in exploratory studies
- When active control(s) are included

Figure. Estimated 95% DR Curve



Clinical Trial Designs

- Currently, clinical trial program in drug development mostly consists of
 - **Learning phase**
Phase 1 trial, POC trial, Phase 2 trial
 - **Confirming phase (or evidence setting)**
Phase 3 trial

Limitations of FD in Learning Trial

- Learning is often based on markers or endpoints other than 1^o clinical endpoints
- Sample size is limited, not powered for studying 1^o clinical endpoints
- Chance of very informative learning about key clinical endpoints is small
- Dose selection based on statistical significance is limited

Departing Remarks

- More opportunity for novel research/software developments
 - When dose-response exploration is based on biomarker
 - Inference based on response adaptive allocation procedure
 - Utility of response adaptive allocation: allow exploration of several more dose regimens than fixed design approach
- Willingness to include near ineffective or suboptimal doses
- Need to learn intelligently in exploratory adaptive trials, as fully commitment of resources may not be an option
- Strategic consideration of dose range, dose levels, frequency for adapting when safety is of potential concern
- US FDA draft guidance encourages use of AD in learning trials