

Model based optimization of dose-finding studies for drug-combinations

Theodoros Papathanasiou^{1,2}, Anders Strathe², Rune V. Overgaard²,
Trine M. Lund¹, Andrew C. Hooker³

1) Department of Drug Design and Pharmacology, University of Copenhagen, Copenhagen, Denmark

2) Novo Nordisk A/S, Quantitative Clinical Pharmacology, Søborg, Denmark

3) Department of Pharmaceutical Biosciences, Uppsala University, Sweden

Background

Combination pharmacotherapy

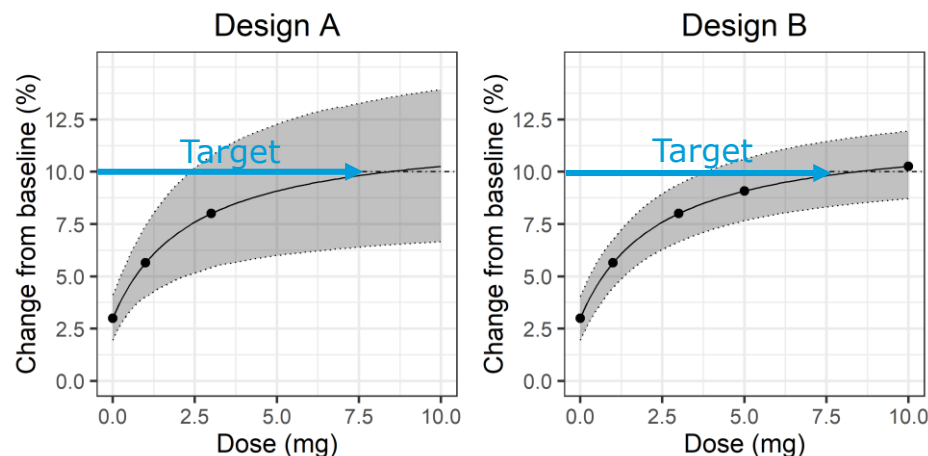
- Simultaneous administration of more than one drugs may
 - Enhance the efficiency of pharmacotherapy¹
 - Higher effect compared to the two mono-components alone
 - Lead to decreased side-effects¹
 - Smaller needed doses for each drug
- Used in various medical fields
 - Metabolic disease (Diabetes, Obesity)
 - Cancer
 - Infectious disease
 - Circulatory system disorders (Hypertension, Atherosclerosis)
 - Anaesthesiology



Background

Dose selection in clinical drug development

- Even for single drugs, **dose selection** is one of the most challenging steps in drug development^{1,2}
 - Poor dose selection is still an important cause of the high attrition rate in confirmatory trials^{1,2}
- Accurate delineation of the *Dose-Exposure-Response* (DER) relationship is a **key** aspect for rational dose selection^{1,2}

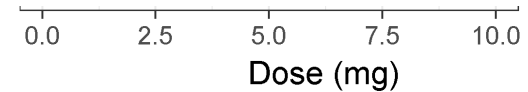


1. Musuamba FT, Manolis E, Holford N, Cheung SYA, Friberg LE, Ogungbenro K, et al. Advanced methods for dose and regimen finding during drug development: Summary of the EMA/EFPIA workshop on dose finding (London 4-5 December 2014). CPT Pharmacometrics Syst Pharmacol.
2. Report from Dose Finding Workshop European Medicines Agency, London, 04 – 05 December 2014

Background

Dose selection in drug combinations

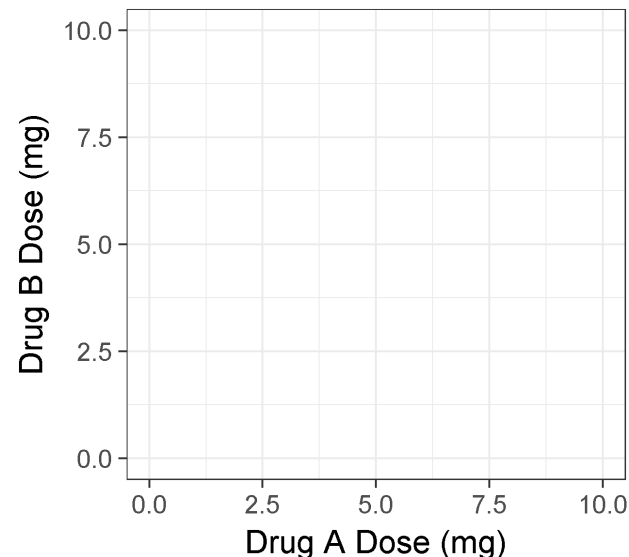
- Dose selection for single drugs is a single-dimensional problem



Background

Dose selection in drug combinations

- Dose selection for single drugs is a single-dimensional problem
- For drug combinations, there is an additional level of complexity
 - Many potential doses to explore
 - Complex Dose-Exposure-Response relationships
 - Especially when pharmacodynamic interactions are present
- Drug development challenge
 - Which combination doses should be explored in a dose-finding setting to
 - Maximize the collected information
 - Increase the probability to select a promising dose to bring forward to confirmatory trials



Aim

- Evaluate the added benefit of using **Optimal Design** for guiding the allocation of studied combination doses in a dose-finding setting
- **Compare** the optimized designs to a typical drug-combination dose-finding design in terms of **probability** to identify the most promising combination dose to bring forward to confirmatory trials

Methods

Drug characteristics and pharmacodynamic endpoint

- Two hypothetical compounds
 - Drug A: Well-established E-R relationship and approved dose
 - Drug B: Novel add-on, with unknown E-R relationship
 - Drugs administered as a 'loose' combination
 - Any combination dose could be considered
- Pharmacodynamic endpoint:
 - % change from baseline
 - Can be applied to any continuous clinical response endpoint
- Analysis method:
 - End-of-study, cross-sectional Exposure-Response (E-R) analysis

Methods

Drug exposure

- Pharmacokinetic (PK) assumptions:
 - Population PK models for both drugs developed prior to the E-R analysis
 - No PK interactions between the two drugs
 - Average steady state concentration (C_{ss} , ng/mL) following repeated dosing
 - Assuming standard linear pharmacokinetics for both drugs

$$C_{ss,x}(Dose_x) = \frac{Dose_x \cdot F_x}{CL_{x,i} \cdot \tau} \quad CL_{x,i} = \theta_x e^{\eta_{x,i}} \quad \eta_{x,i} \sim N(0, \omega_x^2)$$

CL: Drug clearance

F: Bioavailability

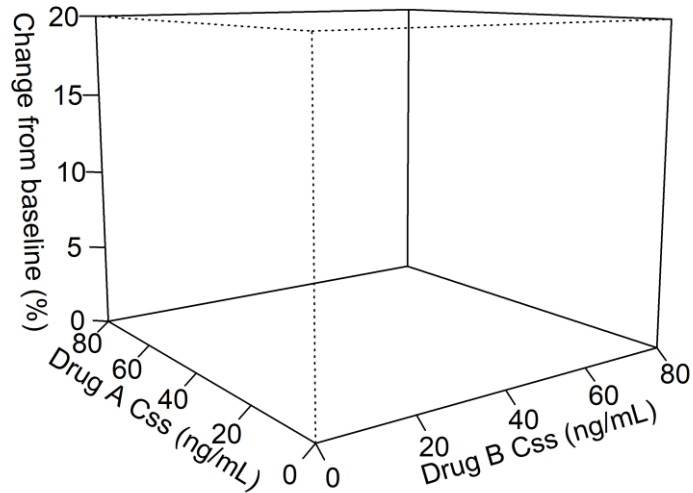
τ : Dosing interval

The apparent clearance (CL/F) and dosing interval for both drugs were considered to be equal (CL/F=10 L/h and τ =24h)

The variability in clearance was assumed to be log-normally distributed with standard deviation 25%

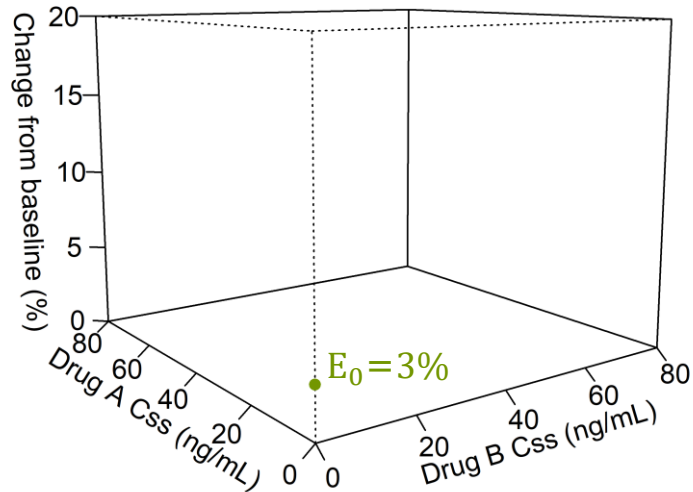
Methods

Combination Dose-Exposure-Response model



Methods

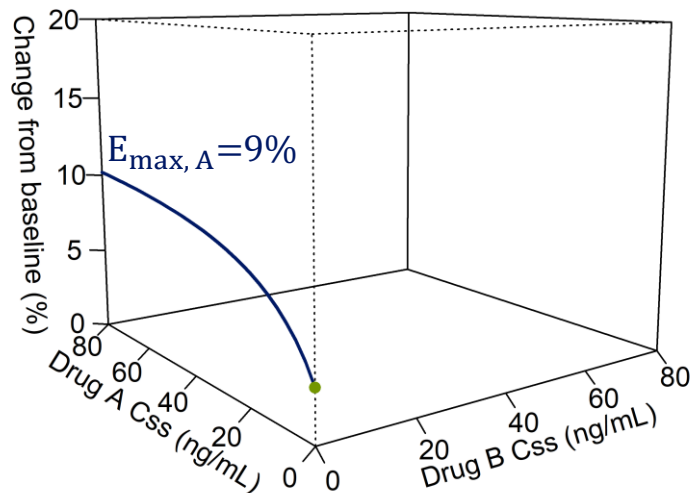
Combination Dose-Exposure-Response model



$$E = E_0$$

Methods

Combination Dose-Exposure-Response model

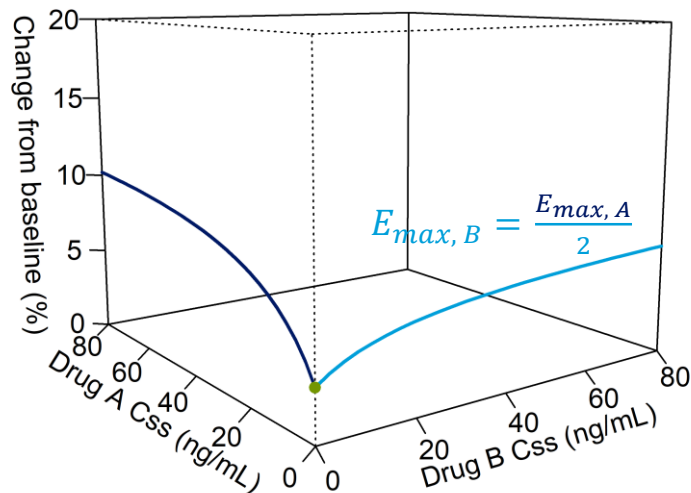


$$E_A = \frac{E_{\max, A} \cdot C_{ss, A}^{\gamma_A}}{EC_{50, A}^{\gamma_A} + C_{ss, A}^{\gamma_A}}$$

$$E = E_0 + E_A$$

Methods

Combination Dose-Exposure-Response model



$$E_A = \frac{E_{max,A} \cdot C_{ss,A}^{\gamma_A}}{EC_{50,A}^{\gamma_A} + C_{ss,A}^{\gamma_A}} \quad E_B = \frac{E_{max,B} \cdot C_{ss,B}^{\gamma_B}}{EC_{50,B}^{\gamma_B} + C_{ss,B}^{\gamma_B}}$$

$$E = E_0 + E_A + E_B$$



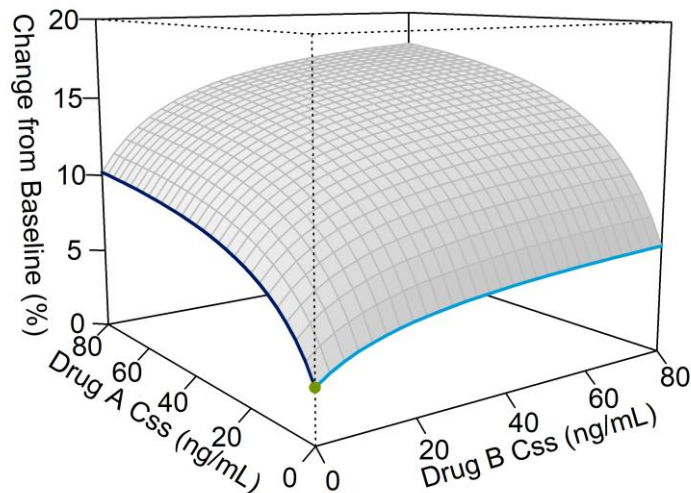
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Methods

Combination Dose-Exposure-Response model



$$E_A = \frac{E_{Max,A} \cdot C_{ss,A}^{\gamma_A}}{EC_{50,A}^{\gamma_A} + C_{ss,A}^{\gamma_A}} \quad E_B = \frac{E_{max,B} \cdot C_{ss,B}^{\gamma_B}}{EC_{50,B}^{\gamma_B} + C_{ss,B}^{\gamma_B}}$$

$$E = E_0 + E_A + E_B + \alpha E_A E_B + \varepsilon, \quad \varepsilon \sim N(0, \sigma^2)$$

$$\alpha = \begin{cases} \text{Positive Interaction,} & \alpha > 0 \\ \text{Lack of Interaction,} & \alpha = 0 \\ \text{Negative Interaction,} & \alpha < 0 \end{cases}$$



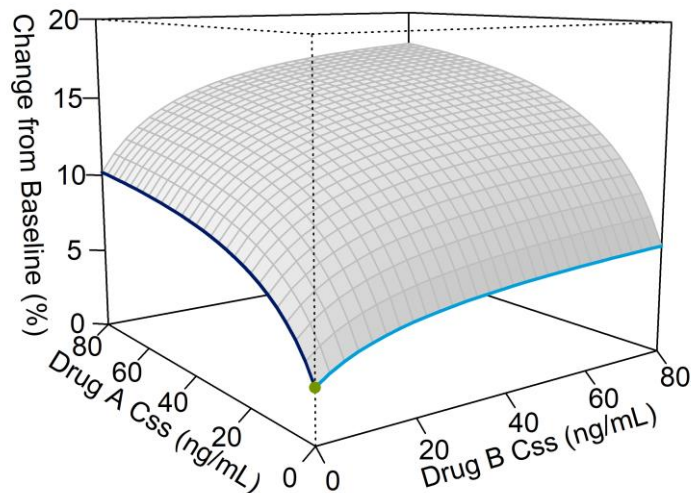
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Methods

Combination Dose-Exposure-Response model



$$E_A = \frac{E_{Max,A} \cdot C_{ss,A}^{\gamma_A}}{EC_{50,A}^{\gamma_A} + C_{ss,A}^{\gamma_A}} \quad E_B = \frac{E_{max,B} \cdot C_{ss,B}^{\gamma_B}}{EC_{50,B}^{\gamma_B} + C_{ss,B}^{\gamma_B}}$$

$$E = E_0 + E_A + E_B + \alpha E_A E_B + \varepsilon, \quad \varepsilon \sim N(0, \sigma^2)$$

$$\alpha = \begin{cases} \text{Positive Interaction,} & \alpha > 0 \\ \text{Lack of Interaction,} & \alpha = 0 \\ \text{Negative Interaction,} & \alpha < 0 \end{cases}$$

$$\alpha = 0.15$$



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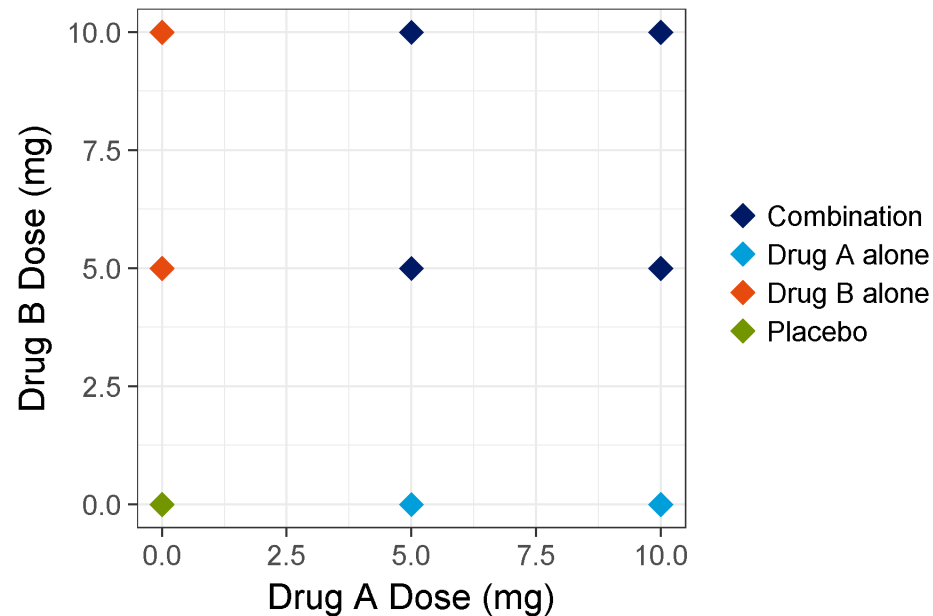
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Methods

Reference design (3x3 factorial)

- Most comprehensive design found in the literature¹
 - Simple construction
 - Ignores potential differences in the information in the design space
- $N_{\text{trial}}=540$ subjects (60 subjects per arm)
 - Power calculation using a two-sided t-test
 - Reflects the most common method for obtaining the sample size in dose finding trials
 - Powered to detect a $\Delta=5\%$ from placebo

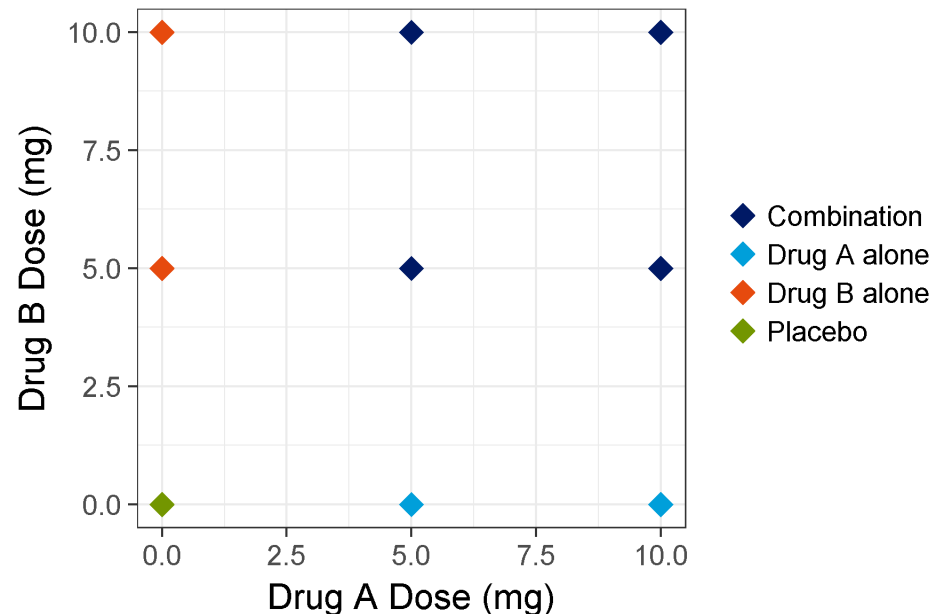


Level of variance: 6% sd. 95% power. 95% significance level of $\alpha = 0.05$

Methods

Design optimizations

- Design optimizations was performed with respect to dose allocation.
 - Monotherapy and combination doses were allowed to vary.
 - Optimizations initiated from the reference design.
 - Search grid from **0 to 10 mg** with 0.1 mg resolution
 - Maximum combination dose based on safety information from phase I
- Drug A parameters were kept fixed
 - Most common situation in clinical drug development for drug combinations¹
- E_0 assumed to be of little interest
 - Ds optimization family
- PopED was used for the evaluation and optimization of all designs²



1. Nøhr-Nielsen A, De Bruin ML, Thomsen M, Pipper CB, Lange T, Bjerrum OJ, et al. Body of evidence and approaches applied in the clinical development program of fixed-dose combinations in the European Union from 2010-2016. Br J Clin Pharmacol. 2019

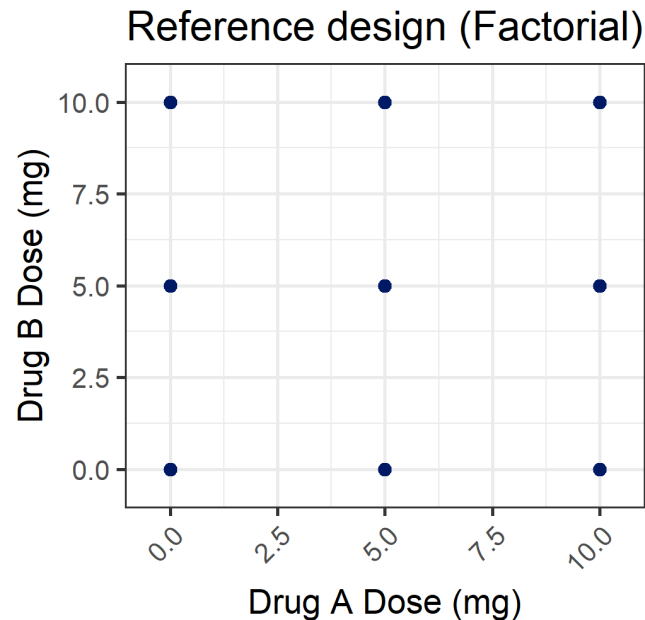
2. Nyberg J, Ueckert S, Strömberg EA, Hennig S, Karlsson MO, Hooker AC. PopED: an extended, parallelized, nonlinear mixed effects models optimal design tool. Comput Methods Programs Biomed. 2012;108(2):789-805.

Results

Reference design (3x3 factorial)

- Evaluation of reference design
 - Low expected overall parameter precision¹
 - Very little information on $EC_{50,B}$ and the interaction parameter α

Parameter Precision (RSE (%))		
Parameter	Value	Factorial
E_0 (%)	3	14.9
$E_{max,B}$ (%)	4.5	42.4
$EC_{50,B}$ (ng/mL)	20	95.8
α (unitless)	0.15	50.7
Average RSE (%)	-	51
Ds-Efficiency (%)	-	100



N/Arm • 60

FIM (Fisher Information Matrix) predicted RSEs(%)

N/Arm: Number of subjects per arm

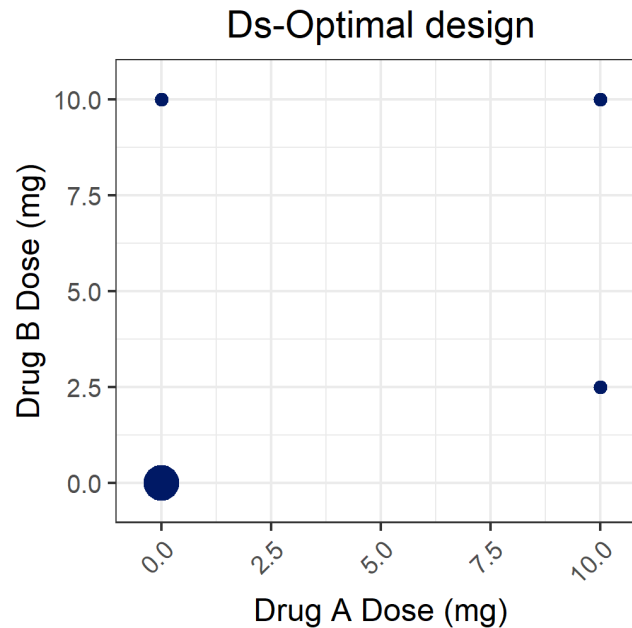
1. Papathanasiou, T., Strathe, A., & Hooker, A. C. (2018). Feasibility of Exposure-Response Analyses for Clinical Dose-Ranging Studies of Drug Combinations

Results

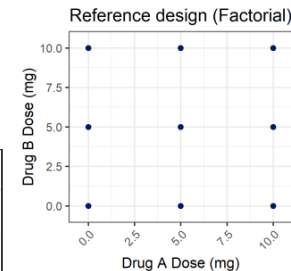
Ds-Optimal design

- 41% gain in Ds-efficiency
 - Same information content with reference with as little as **324** subjects
- Simpler than reference design
 - Four arms were shown to be adequate for parameter estimation

Parameter	Parameter Precision (RSE (%))			
	Value	Factorial		Ds-Optimal
E_0 (%)	3	14.9	■	14.9
$E_{\max,B}$ (%)	4.5	42.4	▲	28.1
$EC_{50,B}$ (ng/mL)	20	95.8	▲	66.8
α (unitless)	0.15	50.7	▲	45.2
Average RSE (%)	-	51	▲	38.8
Ds-Efficiency (%)	-	100	▲	141.2



N/Arm ● 120 ● 180



FIM (Fisher Information Matrix) predicted RSEs(%)

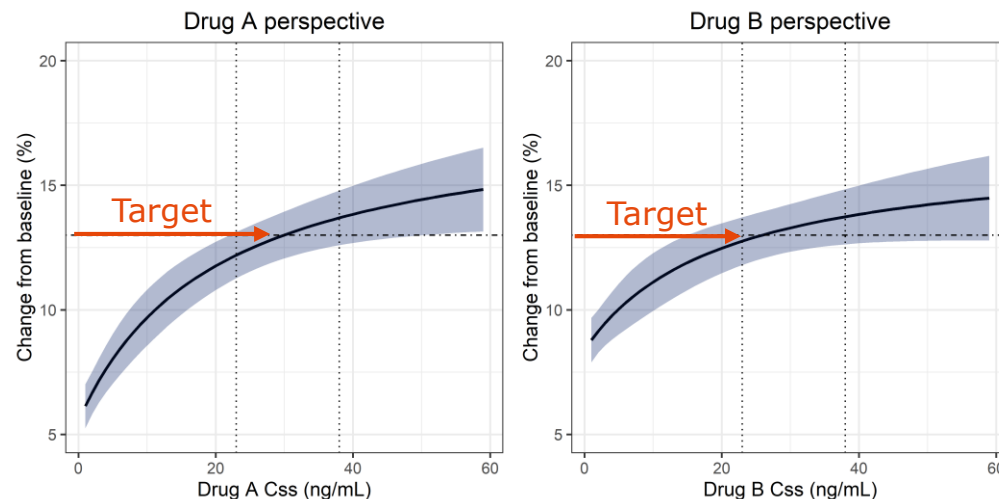
N/Arm: Number of subjects per arm

Same weights across trials (i.e. N/arm) for all doses: N = 60/arm. Some arms are replicates

Results

Ds-Optimal design – model predictions

- Performing clinical trials simply to obtain accurate parameters estimates is somewhat implausible
- Dose-finding trials objective
 - Maximize the confidence of dose-selection for confirmatory trials
 - Minimize the prediction variance around a desired (pre-specified) effect level
- D-optimal designs
 - Maximize information in the parameter space
 - Potentially suboptimal for predictions at a desired effect level



Results

Compound Ds/V-Optimal design

- V-Optimal designs
 - Focus on minimizing the model prediction variance over a range of concentration of drug A and drug B¹
- **Advantages**
 - **Very good** prediction for a wanted area of the E-R curve
- **Disadvantages**
 - **Very imprecise** parameter estimates (expected)¹
 - **Implausible** clinical trial design
- **Solution:** A combination of Ds and V
 - Equal contribution of Ds- and V-optimality criteria

$$Ds/V(\xi) = \kappa \cdot \log Eff_{Ds}(\xi) + (1 - \kappa) \cdot \log Eff_V(\xi)$$

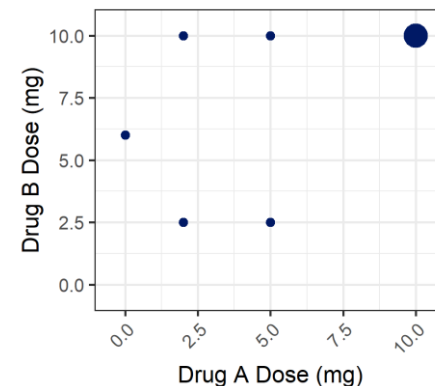
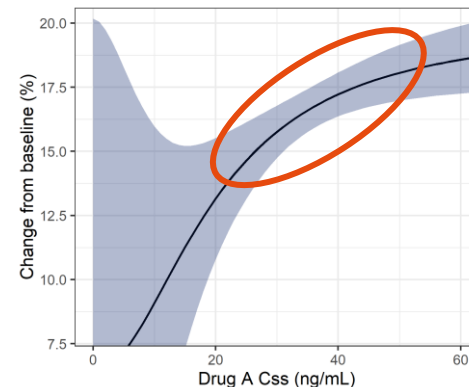
ξ : Design variable

Eff_D : D-efficiency

Eff_V : V-efficiency

κ : integer ($0 \leq \kappa \leq 1$)

Controls how much each design criterion influences the final design



N/arm • 60 • 240



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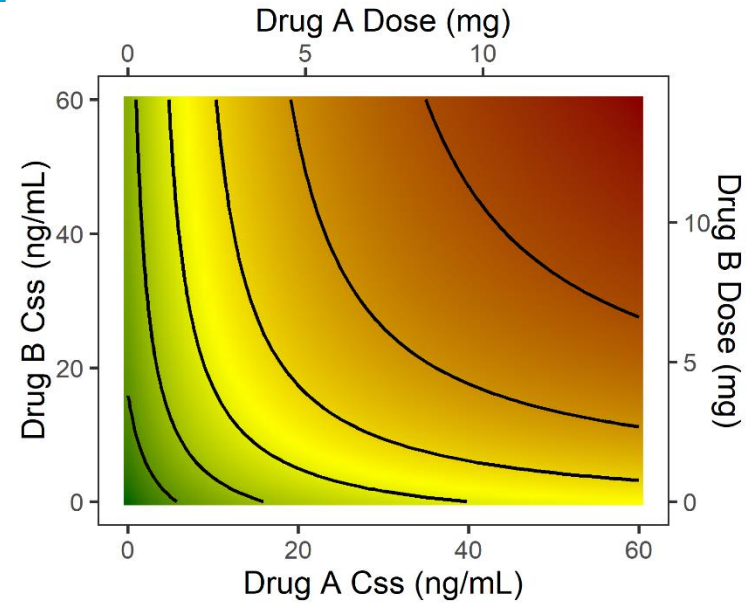


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1. Miller F, Guilbaud O, Dette H. Optimal designs for estimating the interesting part of a dose-effect curve. J Biopharm Stat. 2007;17(6):1097-115

Results

Scenario for correct dose identification



Change from baseline (%)

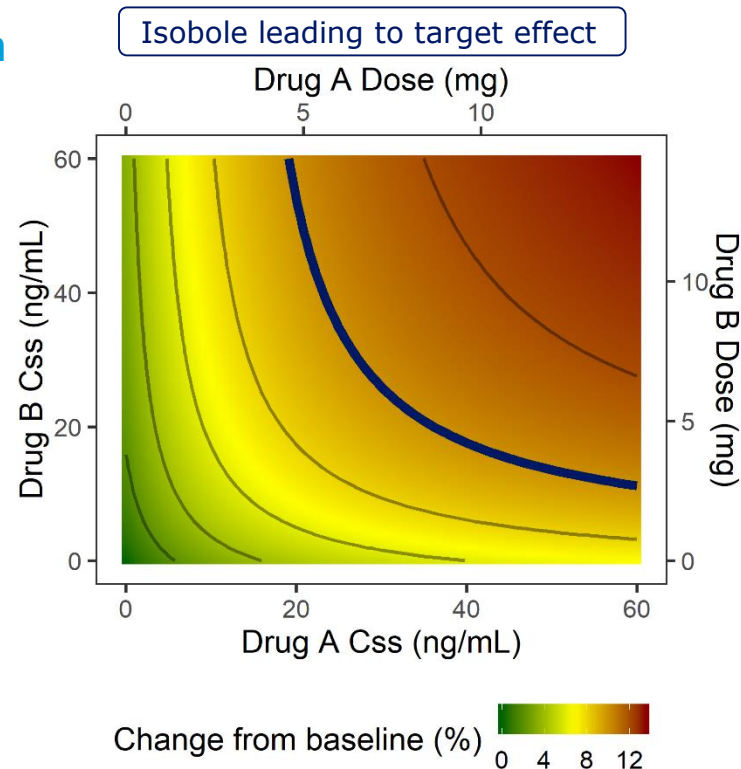


0 4 8 12

Results

Scenario for correct dose identification

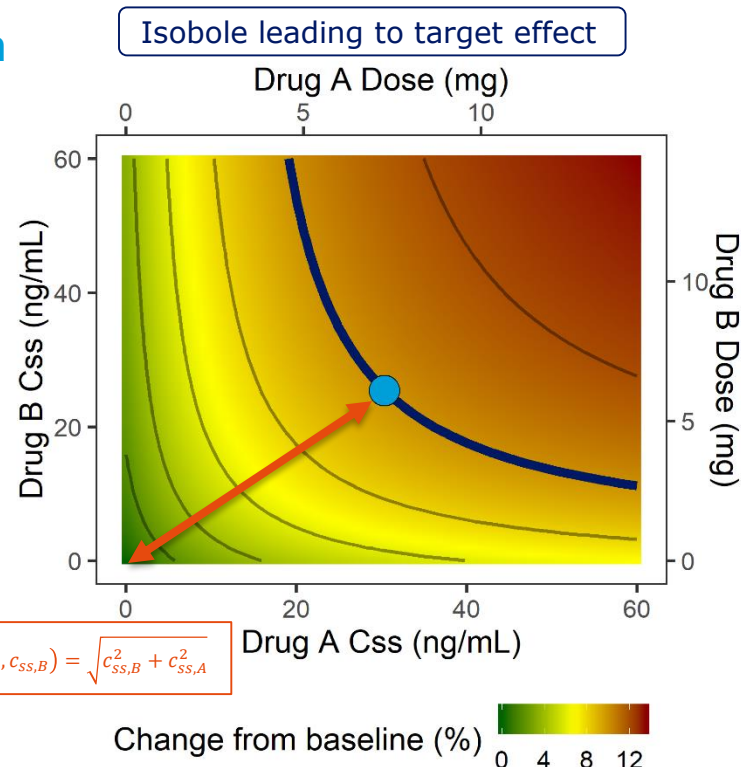
- Target effect: $\Delta_{\text{Target}} = 10\%$
 - 3% CFB for placebo
 - Dark blue line represent the true 13% (10%+3%) isobole



Results

Scenario for correct dose identification

- Target effect: $\Delta_{\text{Target}} = 10\%$
 - 3% CFB for placebo
 - Dark blue line represent the true 13% (10%+3%) isobole
- Optimal dose-combination for target effect
 - Light blue dot: the smallest combination of both drugs leading to the target effect



$$MEC(c_{ss,A}, c_{ss,B}) = \sqrt{c_{ss,B}^2 + c_{ss,A}^2}$$

$MEC_{A,B}$: Combined minimum effective concentration



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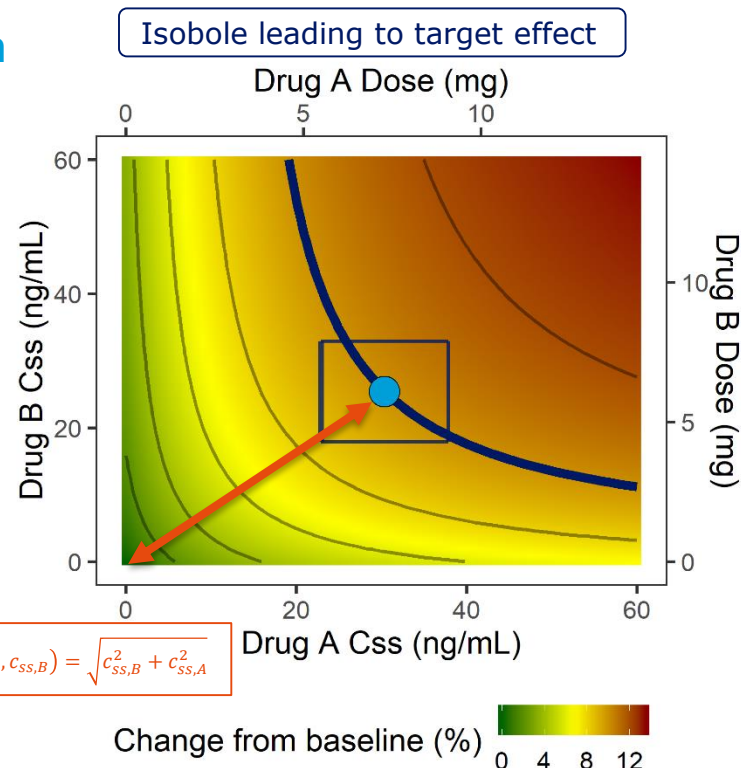
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Results

Scenario for correct dose identification

- Target effect: $\Delta_{\text{Target}} = 10\%$
 - 3% CFB for placebo
 - Dark blue line represent the true 13% (10%+3%) isobole
- Optimal dose-combination for target effect
 - Light blue dot: the smallest combination of both drugs leading to the target effect
- Square represents the area over which the integration for the Ds/V-Optimality criterion was performed
 - Square around the optimal dose-combination with a length of each side L=15 ng/mL (chosen arbitrarily)



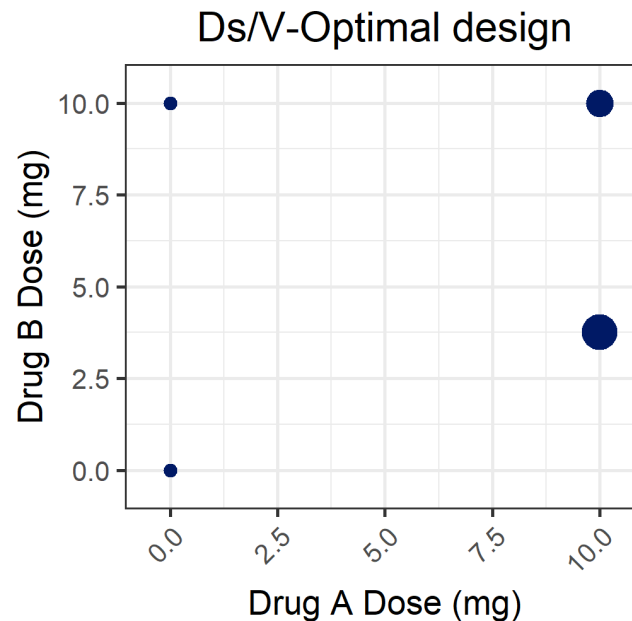
Results

Compound Ds/V-Optimal design

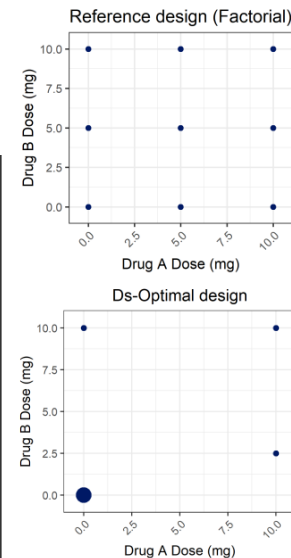
- Ds-efficiency
 - Some loss when compared to the Ds-optimal design
 - More Ds-efficient than the reference design
- Dose allocation
 - Four arms were shown to be adequate
 - Less clustering around placebo

Parameter Precision (RSE (%))

Parameter	Value	Factorial	Ds-Optimal	Ds/V-Optimal
E_0 (%)	3	14.9	14.9	25.8
$E_{\max,B}$ (%)	4.5	42.4	28.1	34
$EC_{50,B}$ (ng/mL)	20	95.8	66.8	60.8
α (unitless)	0.15	50.7	45.2	61.4
Average RSE (%)	-	51	38.8	45.5
Ds-Efficiency (%)	-	100	141.2	107.5



N/Arm • 60 • 180 • 240



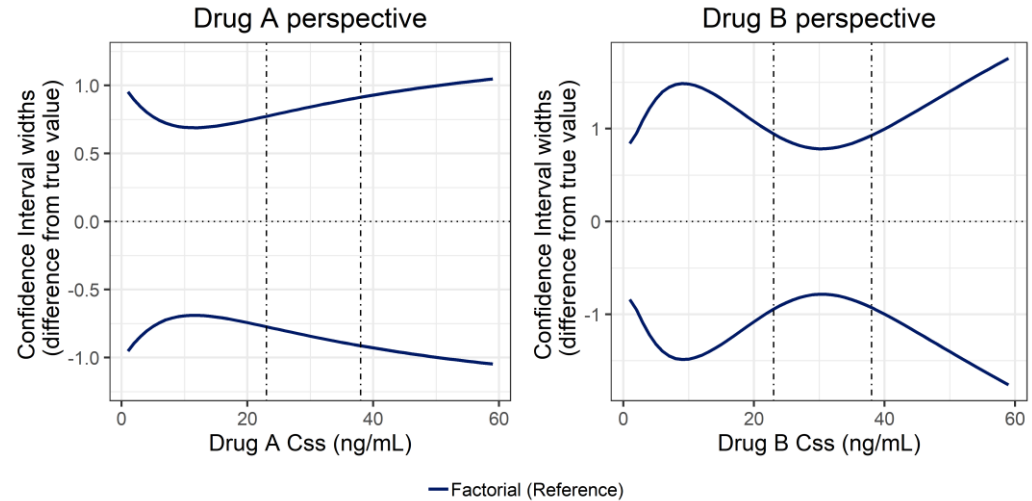
FIM (Fisher Information Matrix) predicted RSEs(%)

N/Arm: Number of subjects per arm

Same weights across trials (i.e. N/arm) for all doses: N = 60/arm. Some arms are replicates

Results

Surface 95% Confidence Intervals around target effect

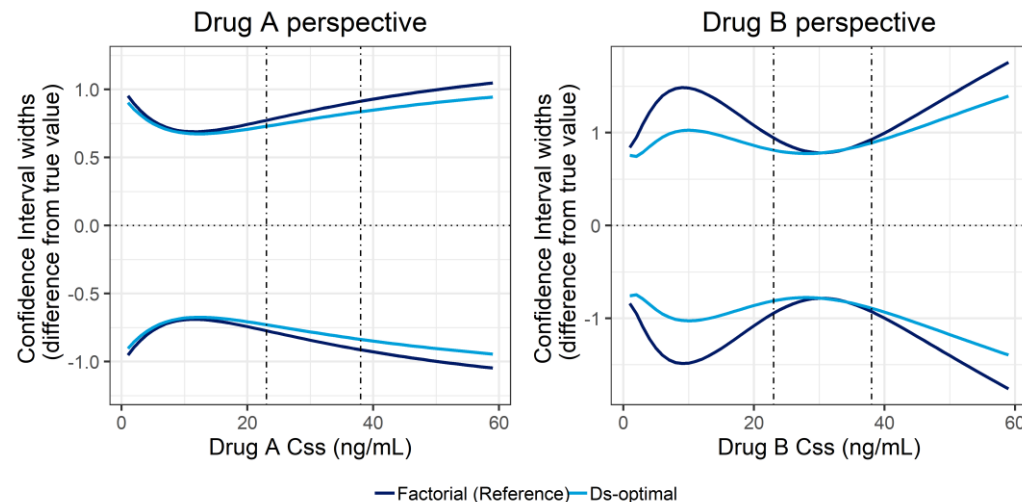


Model 95% CIs calculation was based on the Delta method

Results

Surface 95% Confidence Intervals around target effect

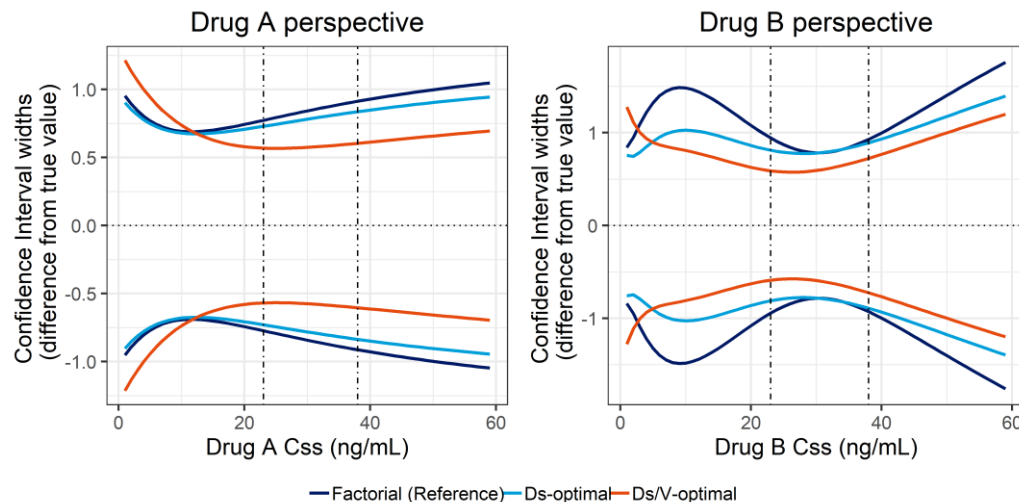
- Little gain in prediction certainty for the Ds-optimal design



Results

Surface 95% Confidence Intervals around target effect

- Little gain in prediction certainty for the Ds-optimal design
- Ds/V-optimal design lead to the highest prediction certainty around the target effect



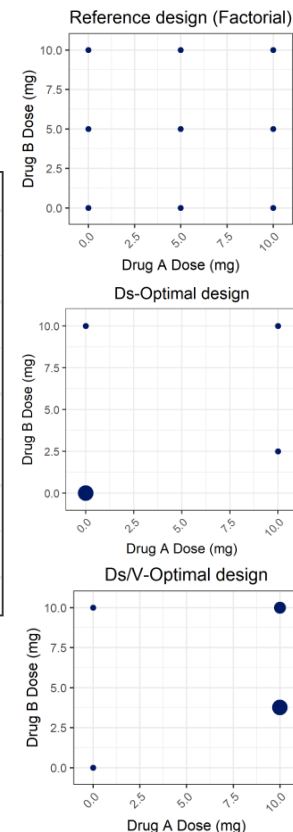
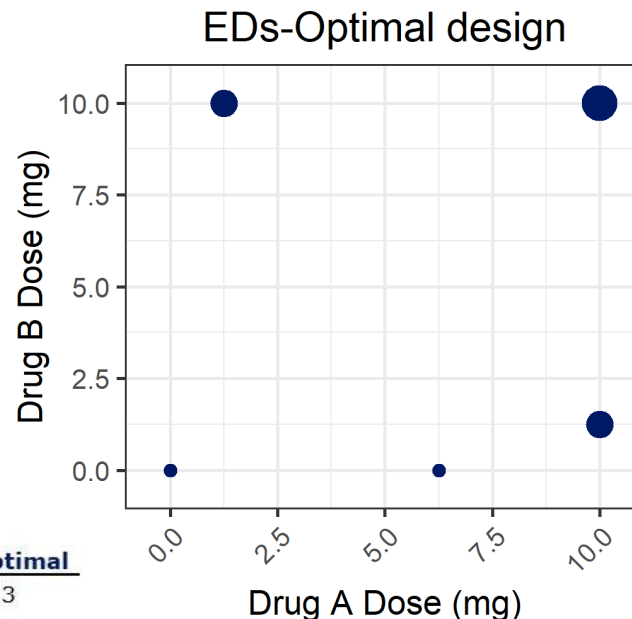
Results

EDs-optimal design

- For nonlinear models, designs are optimal only for the evaluated parameter vector¹
 - Design with uncertainty in parameter space
 - Uncertainty around **all** parameters
- Similar efficiency as compared to the reference
 - More generalizable design
 - Fewer arms as compared to the reference

Parameter Precision (RSE (%))

Parameter	Value	Distribution	Factorial	Ds-Optimal	EDs-Optimal
E_0 (%)	3	U(2.5, 3.5)	14.9	14.9	18.3
$E_{\max,B}$ (%)	4.5	U(3, 9)	42.4	28.1	33.9
$EC_{50,B}$ (ng/mL)	20	U(5, 35)	95.8	66.8	82
α (unitless)	0.15	U(-0.075, 0.175)	50.7	45.2	64.9
Average RSE (%)	-		51	38.8	49.8
Ds-Efficiency (%)	-		100	141.2	98.7



FIM (Fisher Information Matrix) predicted RSEs(%)

N/Arm: Number of subjects per arm

Same weights across trials (i.e. N/arm) for all doses: N = 60/arm. Some arms are replicates

1. Dodds et al., Robust Population Pharmacokinetic Experiment Design. Journal of Pharmacokinetics and Pharmacodynamics, 2005.



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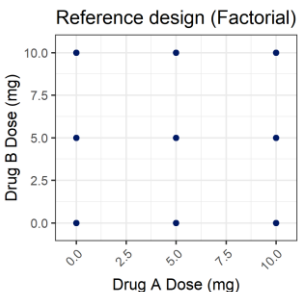
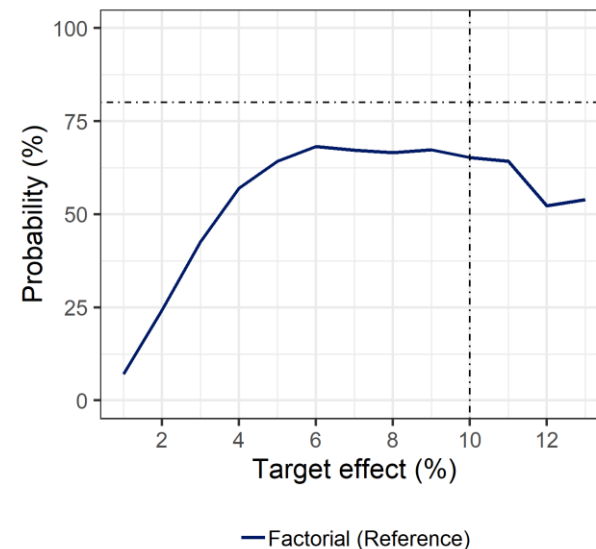
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Results

Probabilities for correct dose identification

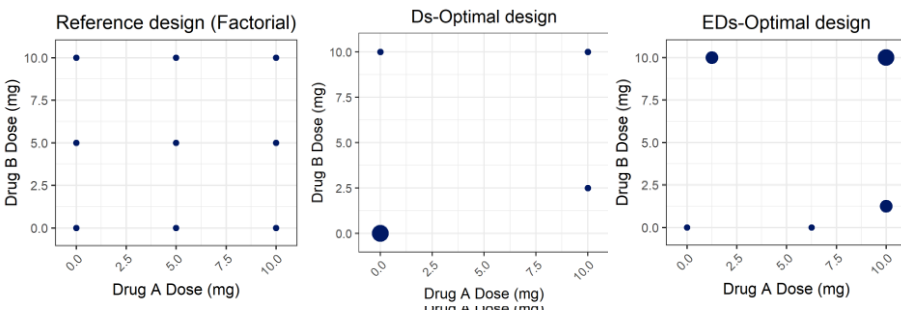
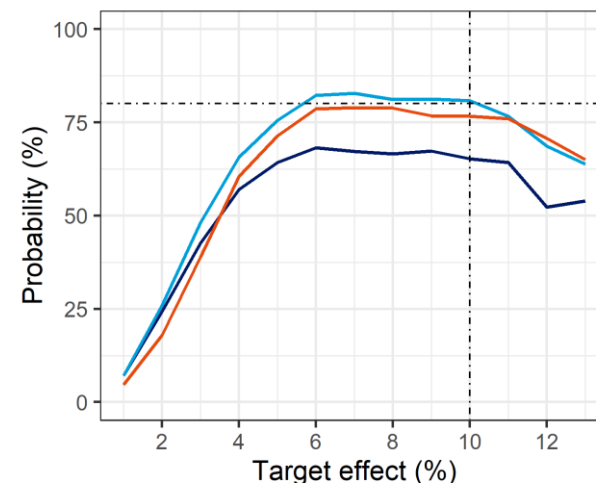
- Stochastic Simulation and Estimation (SSE) was performed based on the reference and optimized designs
 - 1000 SSE replicates
 - Probabilities that the estimated combination doses are within 20% of the true ones



Results

Probabilities for correct dose identification

- Stochastic Simulation and Estimation (SSE) was performed based on the reference and optimized designs
 - 1000 SSE replicates
 - Probabilities that the estimated combination doses are within 20% of the true ones
- Optimized designs consistently led to higher probability of correct dose-identification

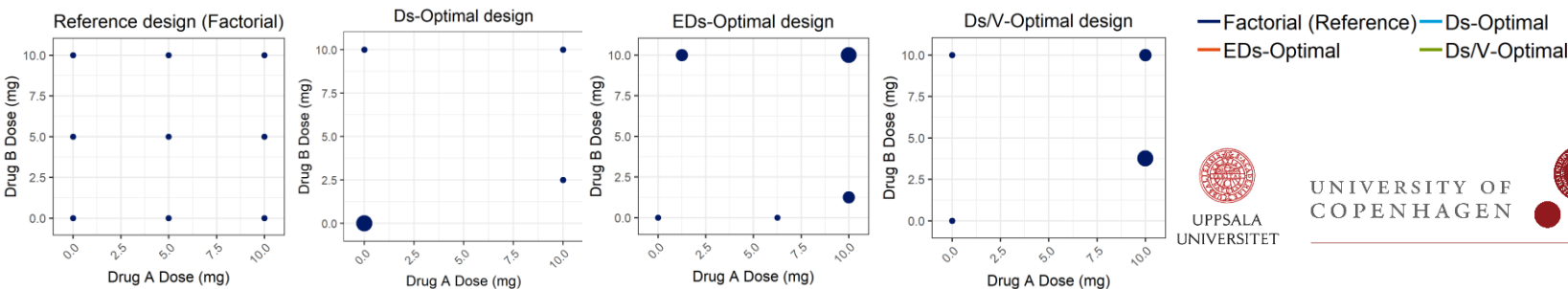
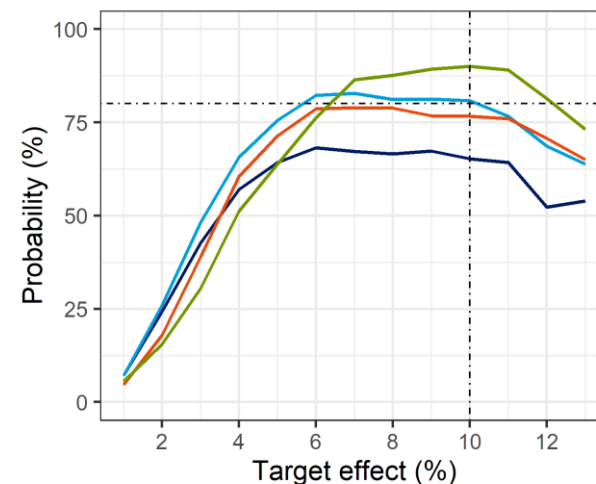


— Factorial (Reference) — Ds-Optimal
— EDs-Optimal

Results

Probabilities for correct dose identification

- Stochastic Simulation and Estimation (SSE) was performed based on the reference and optimized designs
 - 1000 SSE replicates
 - Probabilities that the estimated combination doses are within 20% of the true ones
- Optimized designs consistently led to higher probability of correct dose-identification
- Best performance is seen using the Ds/V design



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Conclusions and perspectives

- Optimized studies
 - Significantly improved the extracted amount of information
 - Allowed for higher confidence in decision making
 - Required smaller number of arms
- Compound D/V-criterion designs are a promising way forward for dose finding in combination therapy studies
- Future research should focus on
 - Expanding the methodology to include safety signals
 - Exploring the influence of uncertainty in the combination model structure

Acknowledgements

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