

MODEL-BASED DRUG DEVELOPMENT

8 June 2011

EFSPI Statistical Leaders Meeting

Introduction

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- As set out in the 1-pager, Sheiner, Senn, Lalonde and colleagues have highlighted an apparent slowness by statisticians to engage with model-based approaches to drug development

- Questions
 - ▣ do we agree with this negative verdict on our discipline?
 - ▣ is it true, as it seems, that the kineticists have stolen a march on us?
 - ▣ where is the leadership within our own profession to challenge this view?
 - ▣ What, if anything, can EFSPI do here?

Models exemplified

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- To motivate discussion I will highlight examples from different areas of our business
 - ▣ 1. Modelling clinical data
 - (i) Predicting relative efficacy in a new indication
 - (ii) Modelling competitor data
 - (iii) More efficient trial design
 - ▣ 2. Pre-clinical PK-PD modelling
 - ▣ 3. Biological systems modelling

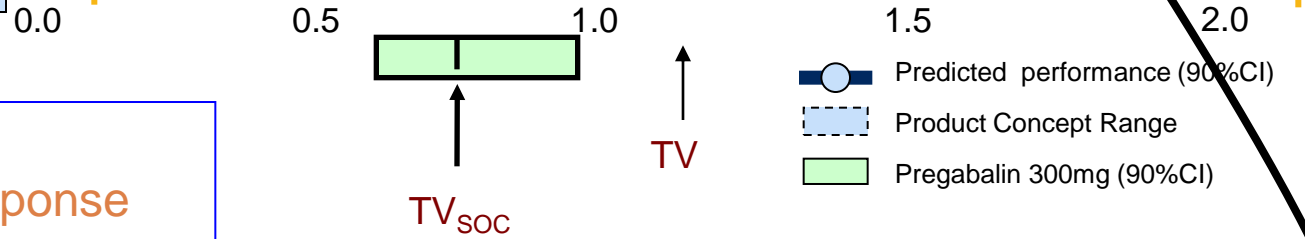
1. Modelling clinical data

(i) Predicting relative efficacy in a new indication

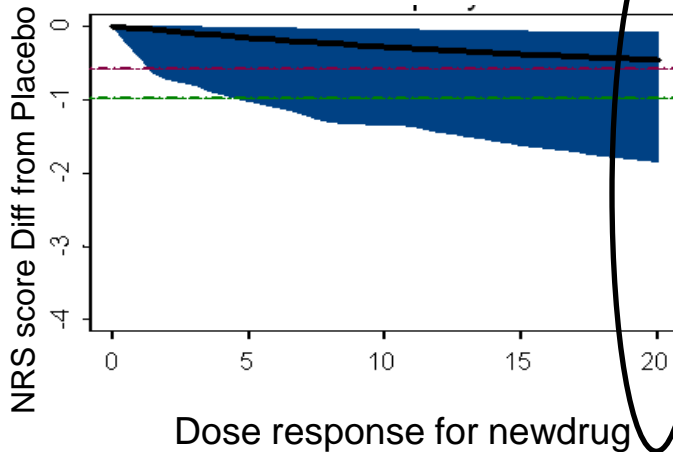
Predicted performance and 90%CI for newdrug

Efficacy

Placebo corrected improvement in NRS pain score (scale 0-10)



A prediction of the expected dose-response for *newdrug* using Model Based Meta-Analysis (MBMA)



- We have data on *newdrug* in PHN
- We wish to predict efficacy in DPN
- We have competitor data in both indications

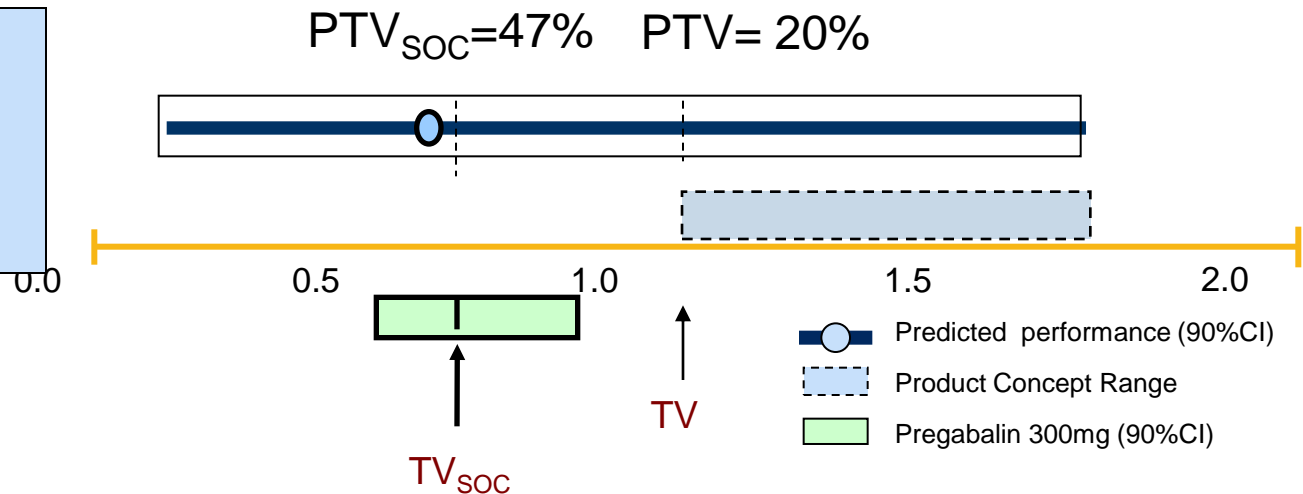
1. Modelling clinical data

(i) Predicting relative efficacy in a new indication

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Efficacy

Placebo corrected improvement in NRS pain score (scale 0-10)



We can calculate the probability of achieving a Target Value (PTV) for newdrug in the new indication

1. Modelling clinical data

(ii) Modelling competitor data

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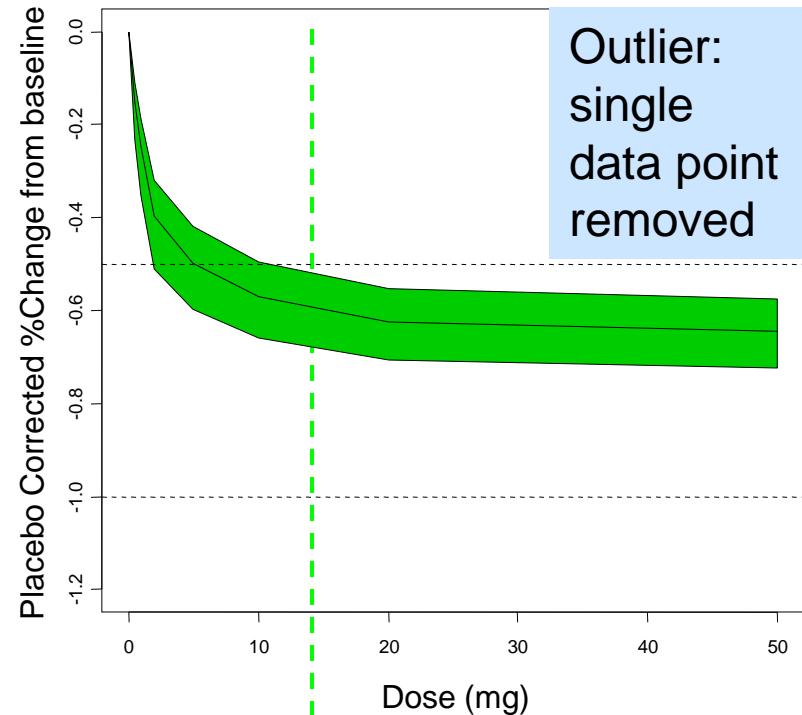
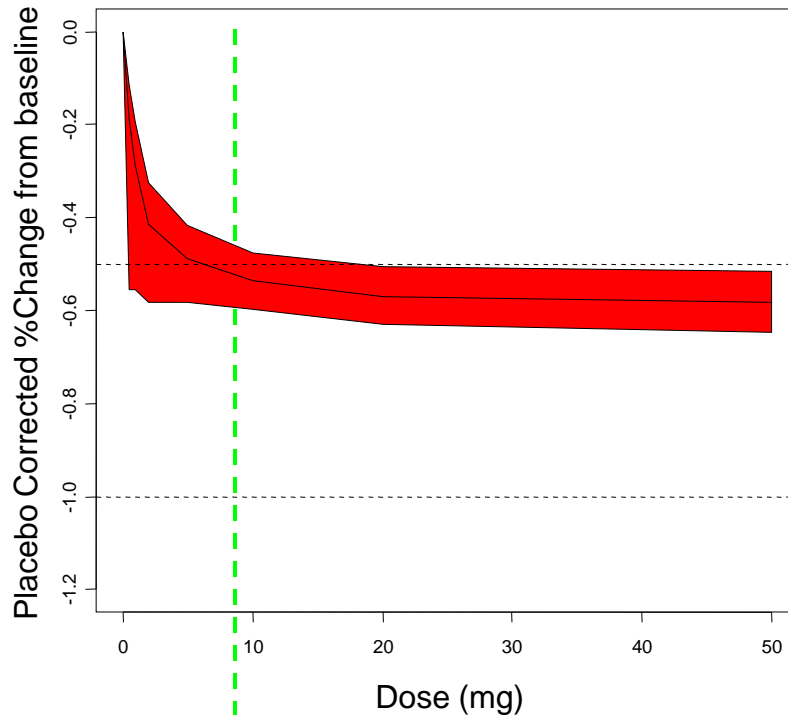
Product Concept	Diabetes agent providing weight loss and/or cardiovascular benefits
Mechanism of Action	X
Strategy	Accelerated development to be in the first wave for this MOA
Competitive Landscape	<i>Company</i> is behind several competitors
Key Gaps in Knowledge	How to differentiate from the leading competitor?

1. Modelling clinical data

(ii) Modelling competitor data

The leading competitor: dose-response for HbA1c%

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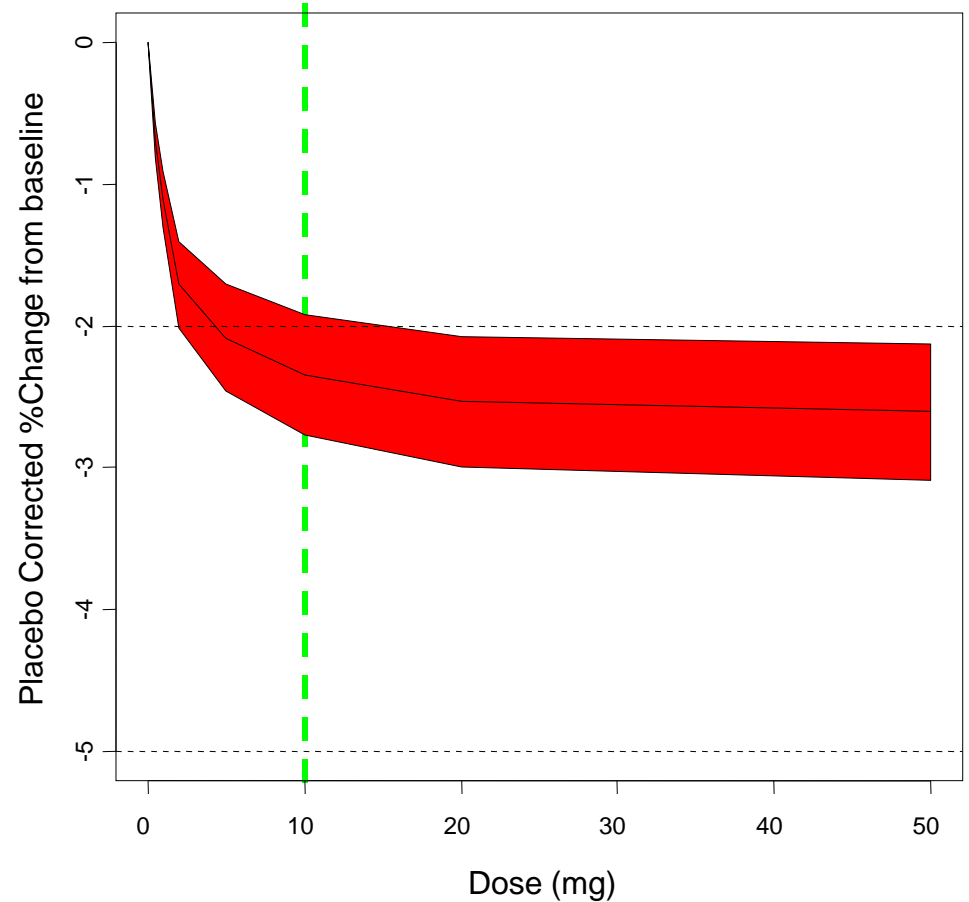
- Estimated $E_{\max} = 0.59 \pm 0.1\%$ with an ED_{50} of 1.05 ± 0.8 mg for HbA1c effect
- Potential outlier at 20 mg dose in diabetic naive study (-0.55% HbA1c) if removed yields
 - $E_{\max} = 0.67 \pm 0.04$, ED_{50} 1.7 ± 0.3 mg
 - variability significantly reduced, especially on ED_{50}

1. Modelling clinical data

(ii) Modelling competitor data

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The leading competitor: dose-response for % weight loss

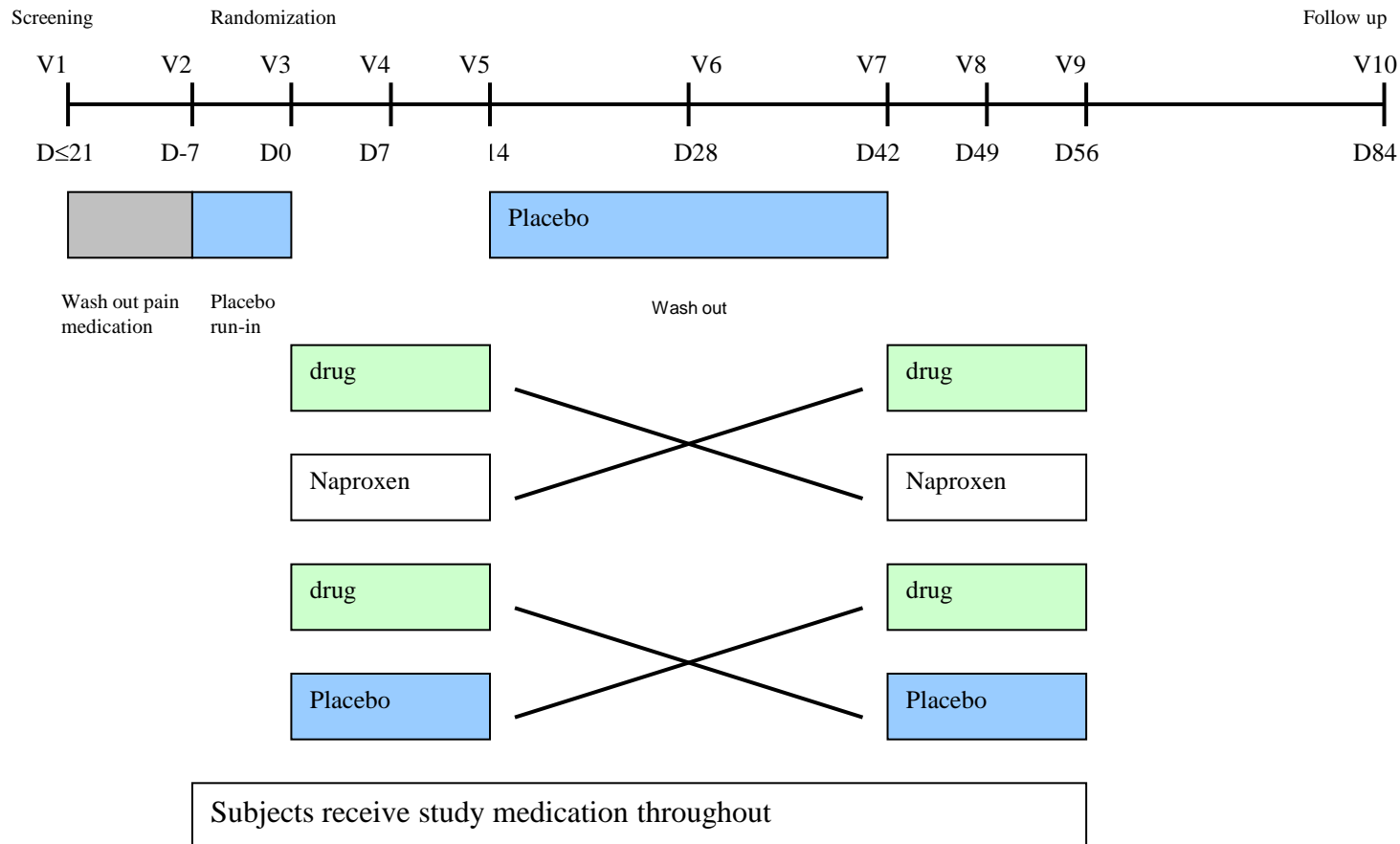


1. Modelling clinical data

(iii) More efficient trial design

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Trial to investigate pain relief following two weeks treatment with *drug* in patients with knee OA



1. Modelling clinical data

(iii) More efficient trial design

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Bayesian Study Design

- Use informative prior for naproxen vs. placebo

- Use elicited priors for:
 - ▣ drug vs. placebo
 - ▣ drug vs. naproxen

1. Modelling clinical data

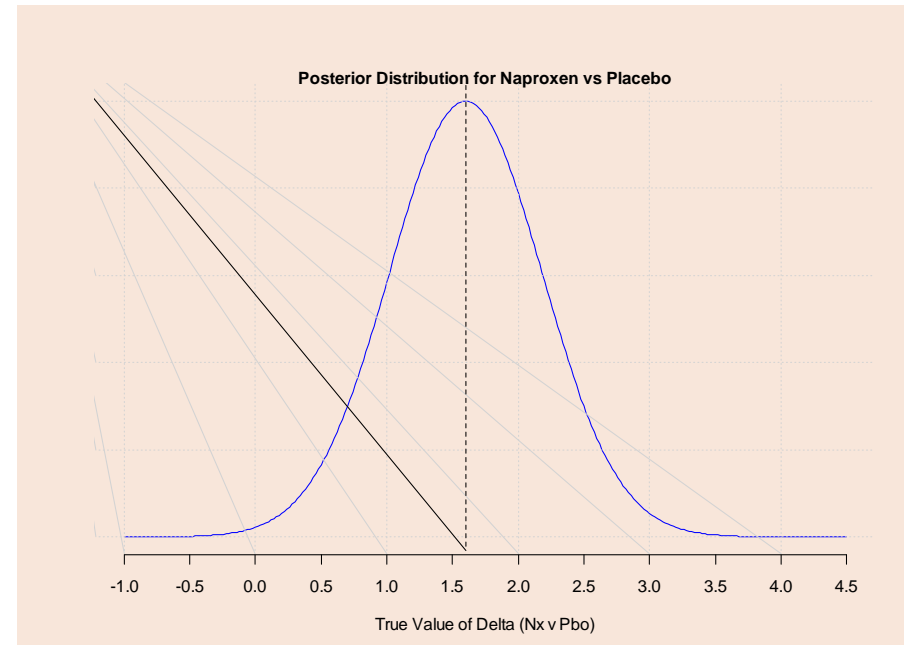
(iii) More efficient trial design

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- Prior for the effect of Naproxen vs. Placebo

$$\Delta \approx N(1.6, 0.58^2)$$

Study_Ref	Diff	SED	Variance
AAAAAAAAAA	2.0	0.33	0.11
BBBBBBBBBB	1.6	0.30	0.09
CCCCCC	2.0	0.66	0.44
DDDDDDDDDD	2.1	0.83	0.68
EEEEEEEEEE	1.1	0.35	0.12
FFFFFFFF	1.1	0.51	0.26



- Using this prior is equivalent to N = 54 subjects on naproxen – placebo → significant efficiency

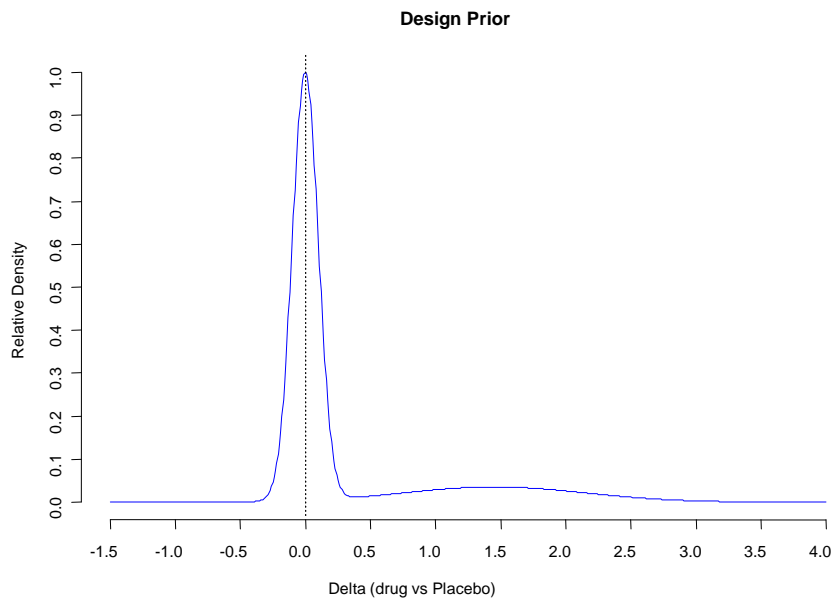
1. Modelling clinical data

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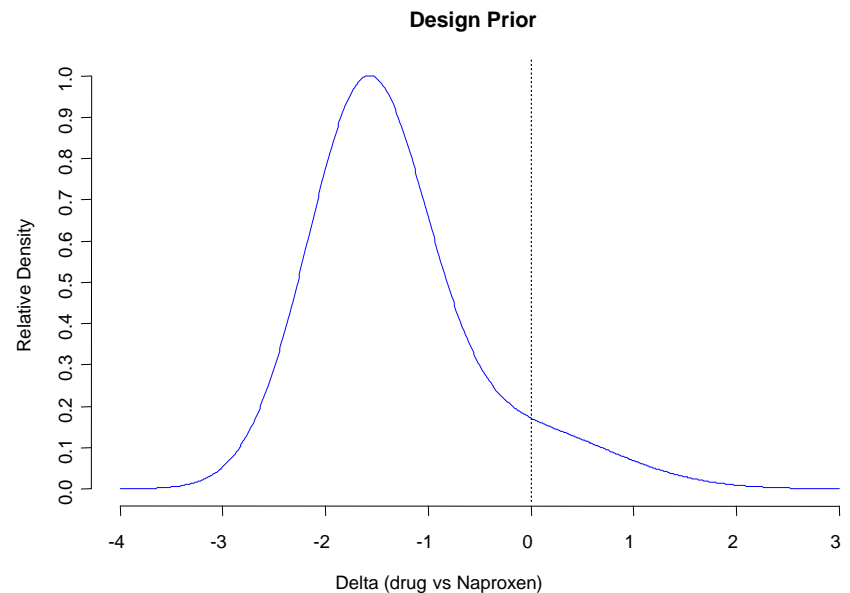
Elicited Prior Belief for drug vs. Placebo Effect

drug vs. Placebo



$$0.8 \cdot N(0, 0.1^2) + 0.2 \cdot N(1.45, 0.7^2)$$

drug vs. Naproxen



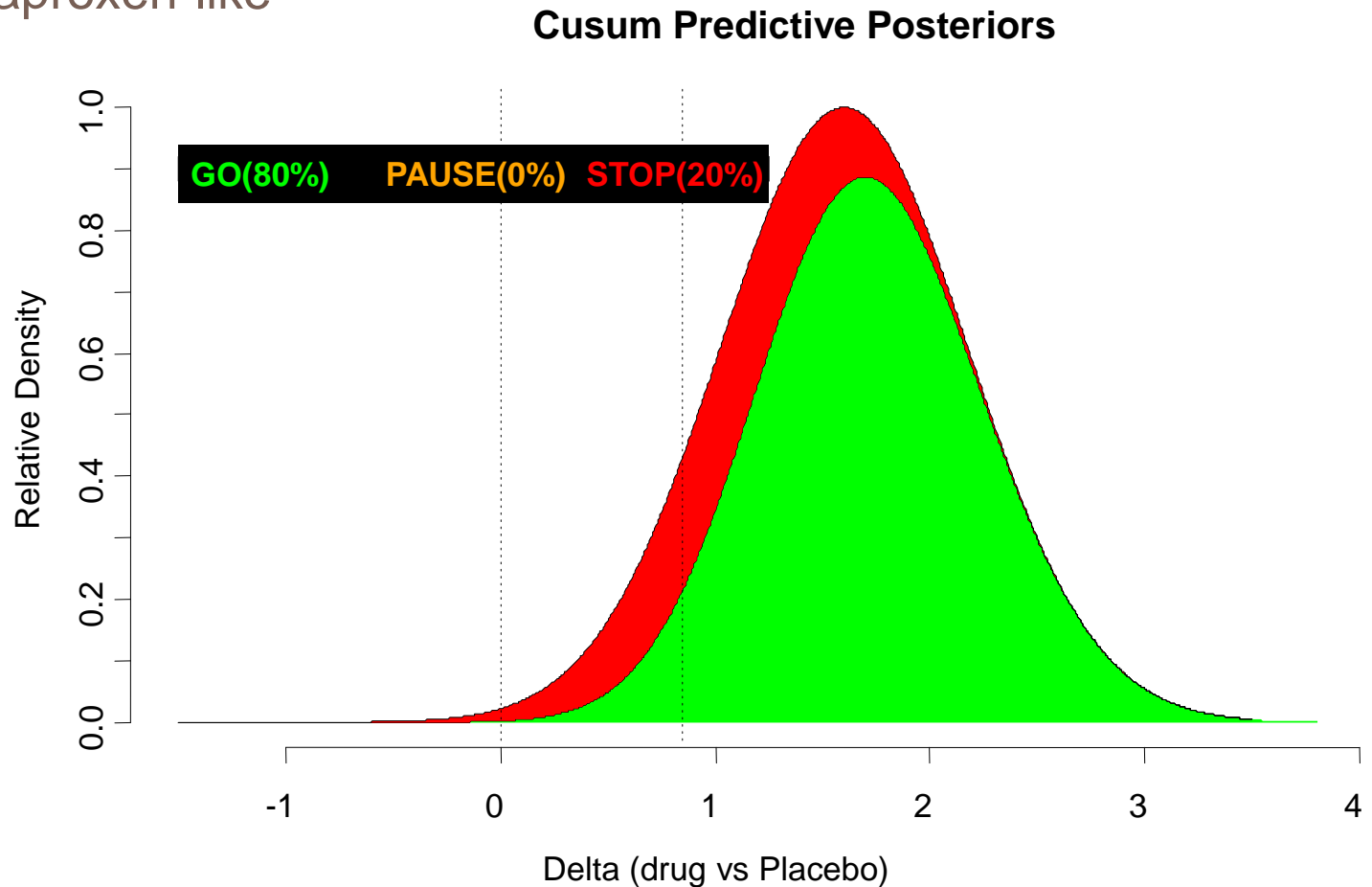
$$0.8 \cdot N(-1.6, 0.58^2) + 0.2 \cdot N(-0.15, (0.7^2 + 0.58^2))$$

1. Modelling clinical data

(iii) More efficient trial design

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Cumulative Predictive Posteriors for superiority to placebo when drug is Naproxen-like



2. Pre-clinical PK-PD modelling

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- Pre-clinical modelling is used to inform the expected dose or exposure needed to demonstrate efficacy in humans
- Defining and achieving 'efficacious exposure' (C_{eff}) is essential to create confidence that we have tested the mechanism and can walk away from a negative result in man
- Surprisingly, there is little agreement on how to define C_{eff}
- What it is not: the lowest drug concentration to yield a statistically significant difference from negative control (vehicle)

2. Pre-clinical PK-PD modelling

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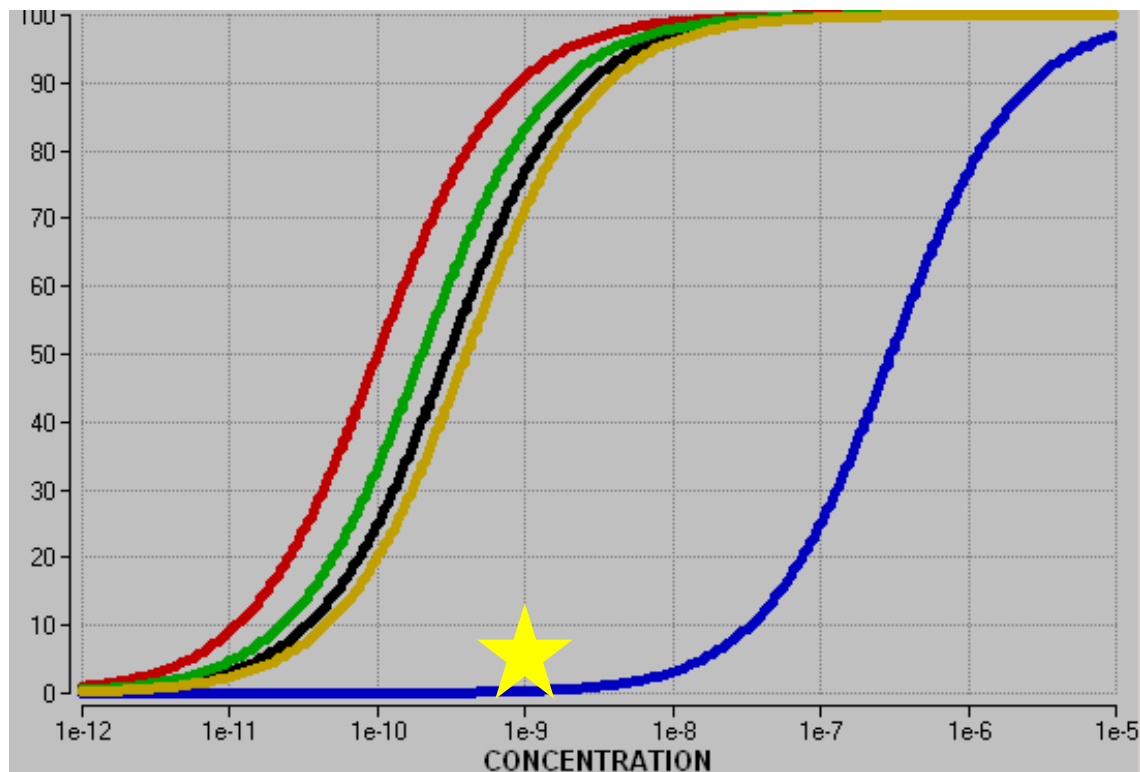
Human in vitro

rat in vivo binding to target

Rat in vivo down-stream
pharmacology

Dog in vivo efficacy

TK limit, most sensitive non-
human species



Consistent pharmacology across species/models

PK-PD well characterised

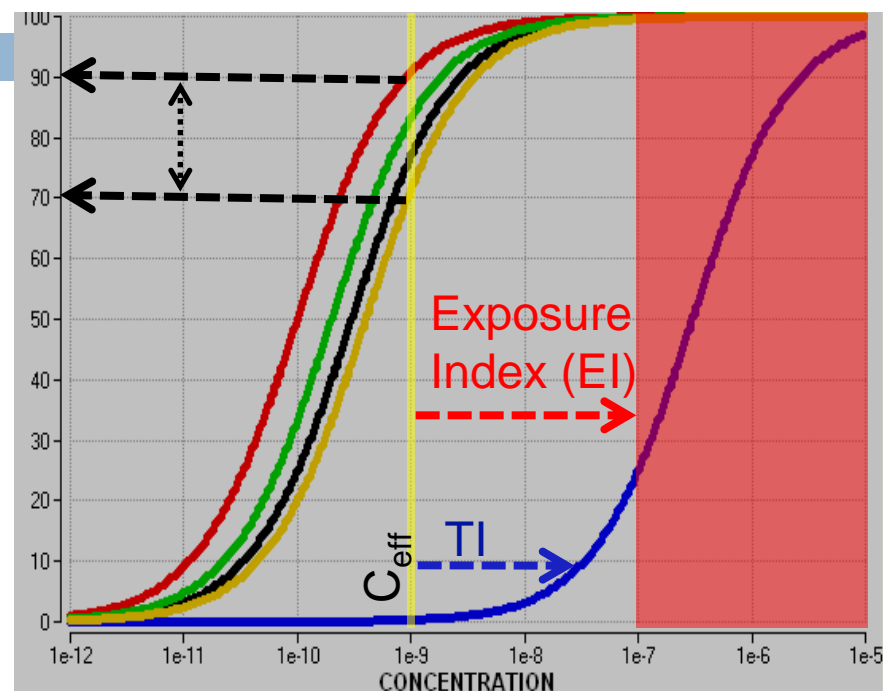
Large TI

C_{eff} will test the mechanism

2. Pre-clinical PK-PD modelling

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- in vitro and in vivo experiments provide estimates of IC_{50} / EC_{50} / K_i
- These are used to construct the E_{max} curves shown
- However, these estimates may not be completely robust



- How to estimate binding affinity (K_i) for a receptor antagonist in vitro?
 - Pooling across different salt forms of drug
 - Pooling of data from different labs
 - Inclusion / exclusion of data points from assay

2. Pre-clinical PK-PD modelling

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Salt form	Lab 1	Lab 2
1	3.30 nM (1.98-5.49 n=5)	3.39 nM
2		1.48 nM (0.796-2.77 n=8)
3	3.81 nM (3.04-4.78 n=28)	0.969 nM (0.614-1.53 n=5)
4		>11.1 nM (3.21E-7-3.86E8 n=2)

Efficacious
dose predicted
at 20mg: no TI

Efficacious
dose
predicted at
5mg: TI=2

- What is an appropriate estimator for K_i ?
- What is a 'no-regrets' dose? $10 \times K_i$

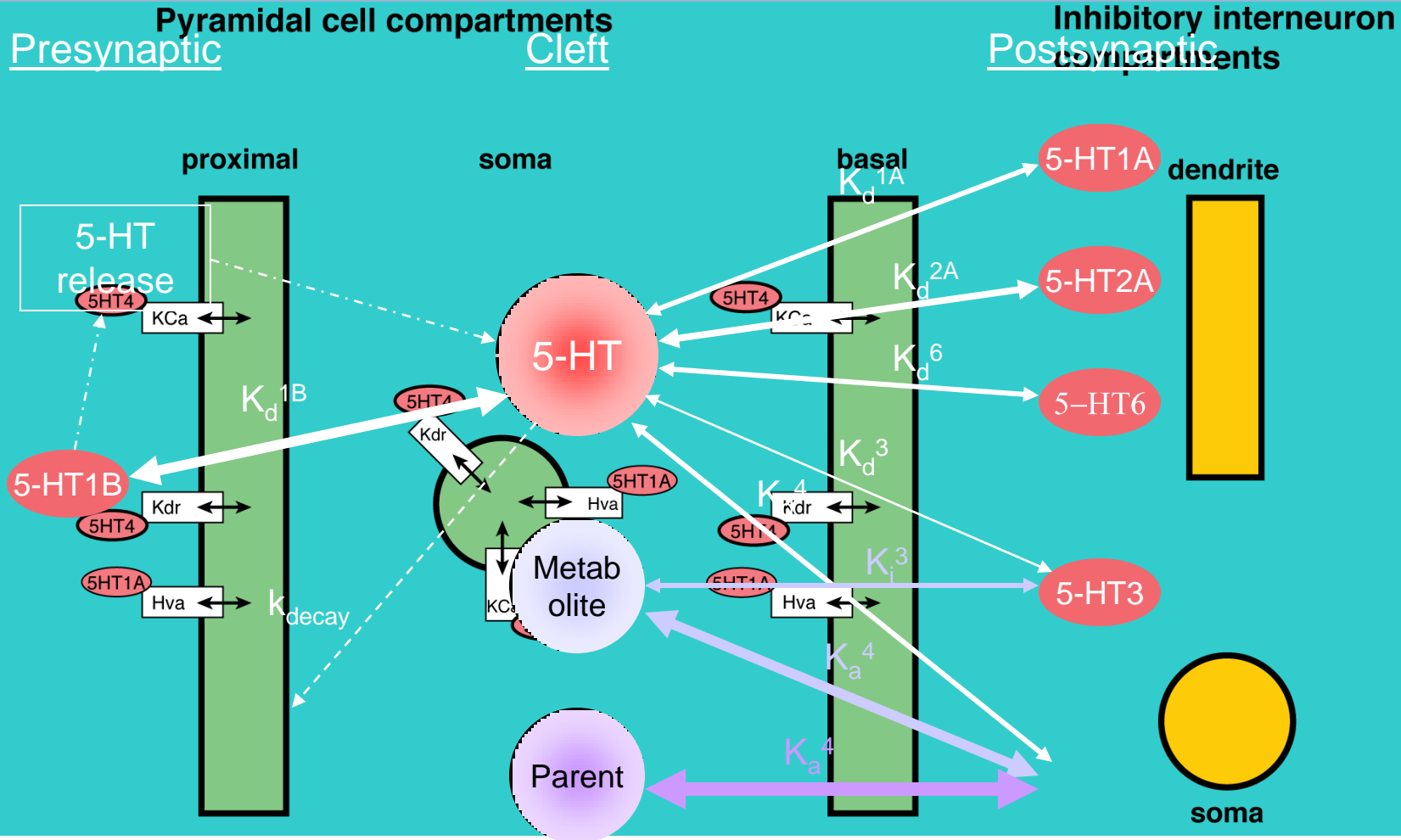
3. Biological systems modelling

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- Everything starts with target selection (human biological drug target)
- We are not biologists
 - ▣ Biologists and others are building complex models to describe basic human biology
 - ▣ Hypothesised cascade / pathway linking known biological processes
 - ▣ Suggests where to intervene to achieve desired pharmacology and avoid unwanted pharmacology
 - ▣ Rely on strong assumptions and typically take data from a variety of sources
 - ▣ Statisticians should be able to scrutinize these models
- The whole field of systems biology / pharmacology needs greater statistical scrutiny

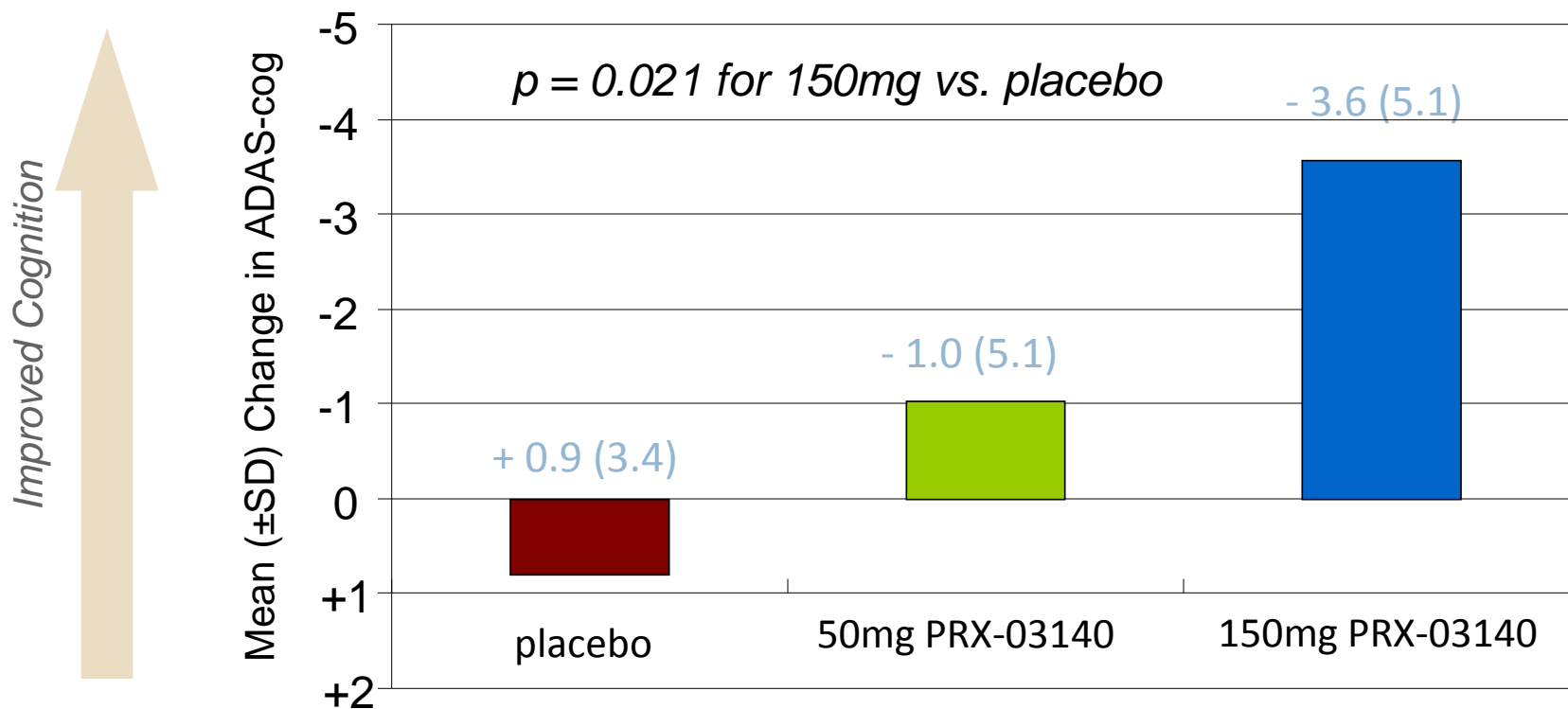
3. Biological systems modelling

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3. Biological systems modelling

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- After two weeks of dosing, mean ADAS-cog change for monotherapy (150mg) was 3.6 points
 - Approved Alzheimer's drugs typically show 3-4 point improvement after 12-24 weeks
- Statistically significant dose-response for 150mg vs. 50mg vs. placebo ($p=0.026$)

Model-based drug development: Questions

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