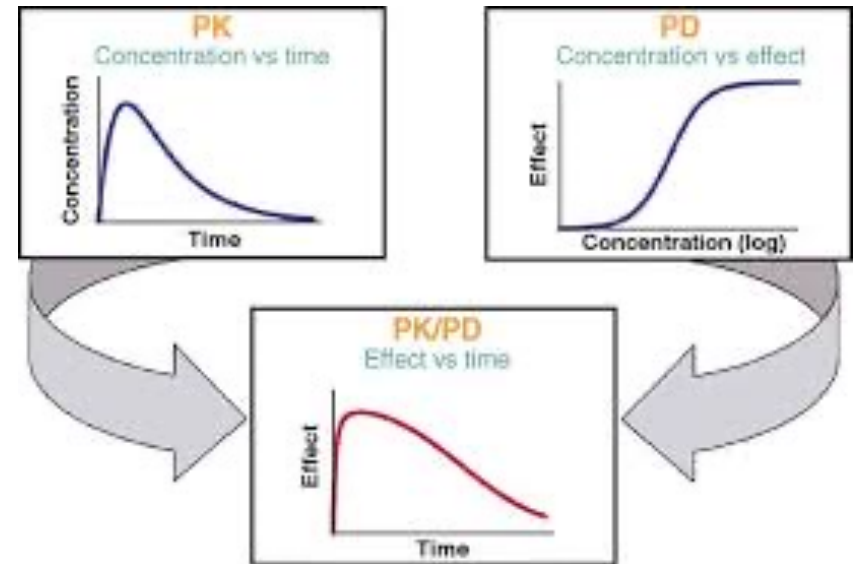


Innovative trial designs including Bayesian approaches

Brian Anderson PhD, FANZCA, FCICM

Professor Anaesthesiology
University of Auckland
New Zealand



Learning Points

- Dose is based on PK, PD and adverse effects
- Variability affects response to medicines
- Major sources of usual variability are maturation, size, drug interactions and genes
- Modelling and simulation, using Bayesian approaches, can be used to predict dose, demonstrate effect and inform future studies

New Zealand: a land of sheep

4,000,000 people

40,000,000 sheep



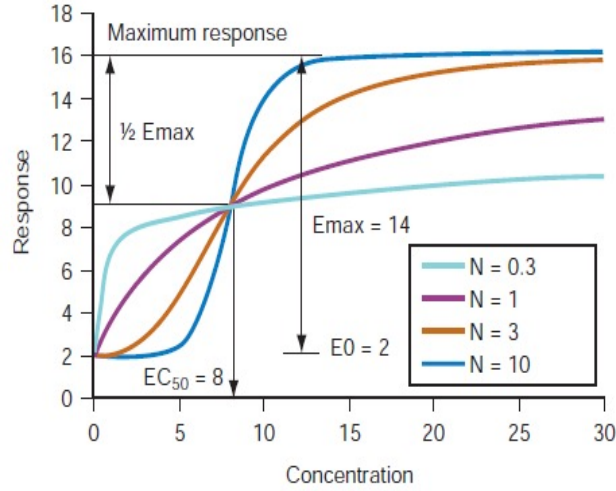
The Students in NZ



Assumptions

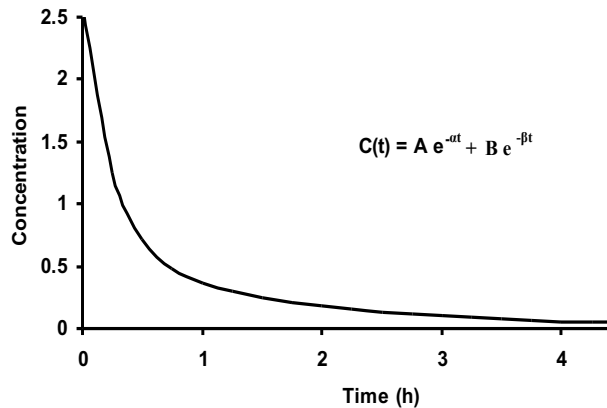
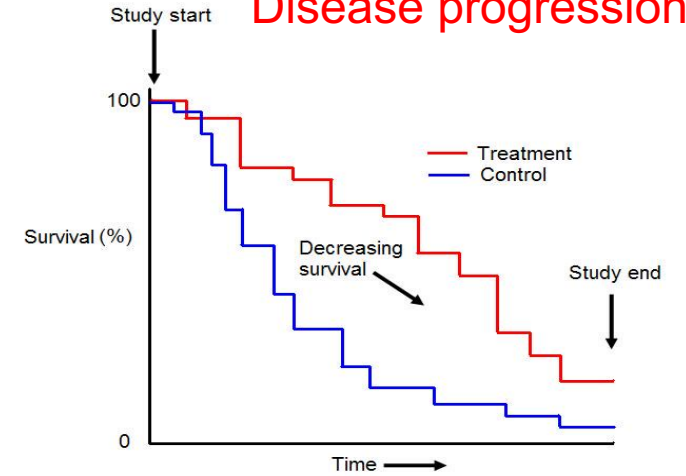
- Basic understanding of population modelling
 - Underlying theme is variability and prior knowledge (Bayesian)
 - Amy Cheung - PKPD
- Aware difficulties of study in children < 2 years

Pharmacodynamics

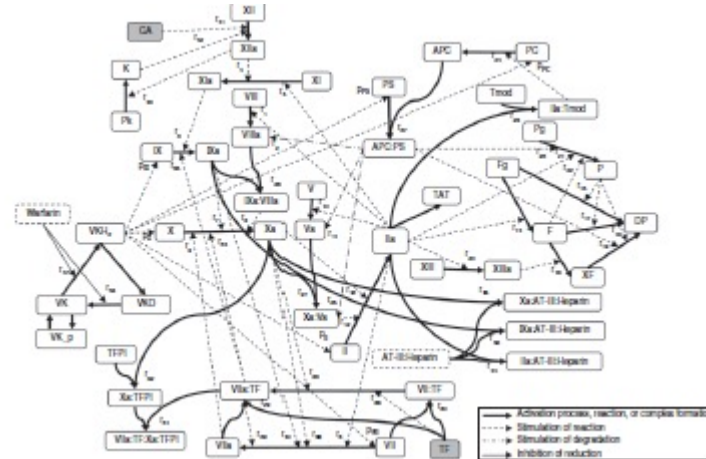


Pharmacometrics

Survival analysis Disease progression



Pharmacokinetics



Quantitative Systems Pharmacology

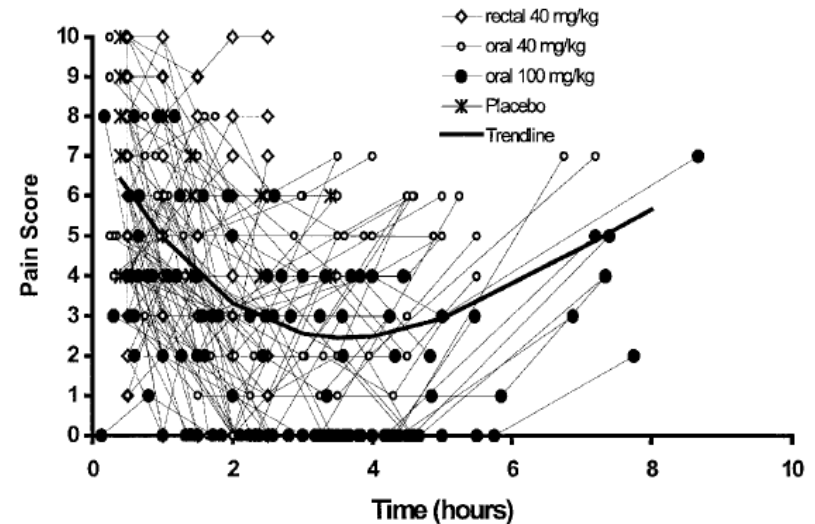
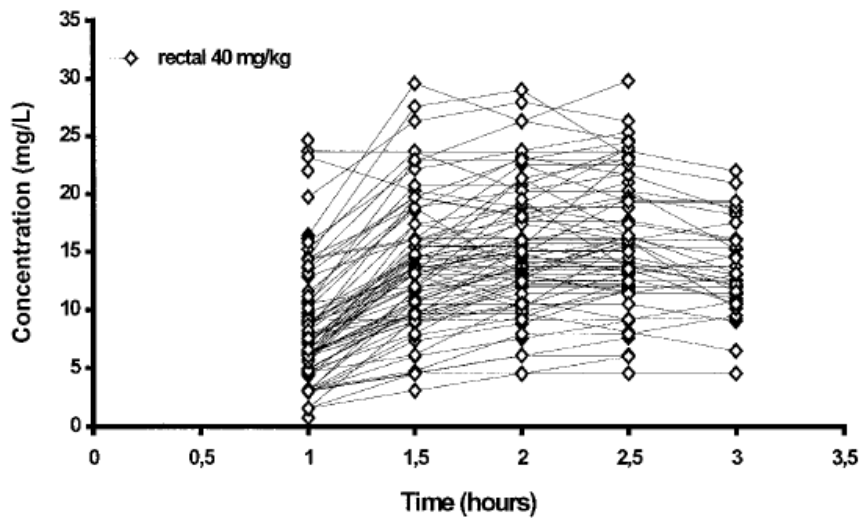
Population Approach

Describing the Signal and the Noise

... and it is the noise (variability) that is important

$$C = \text{Dose} \cdot F / V \times e^{-CL/V \times t}$$

$$\text{Effect} = \frac{E_{\text{max}} \times C^N}{C_{50}^N + C^N}$$



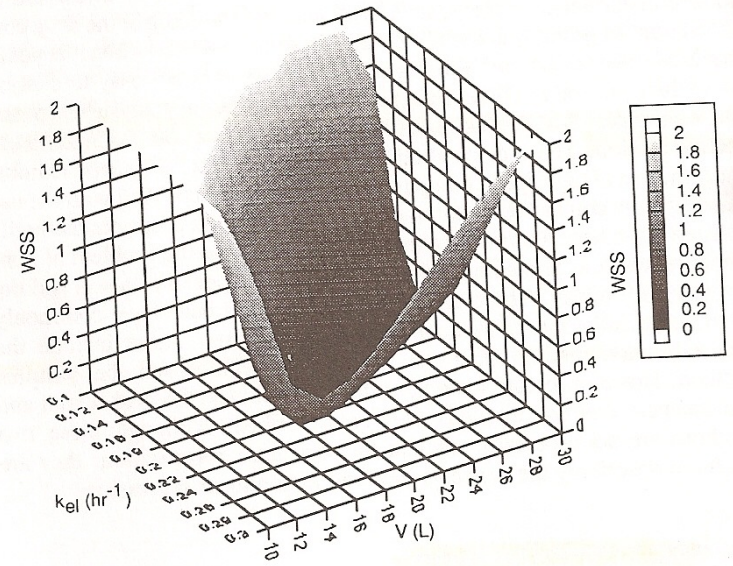
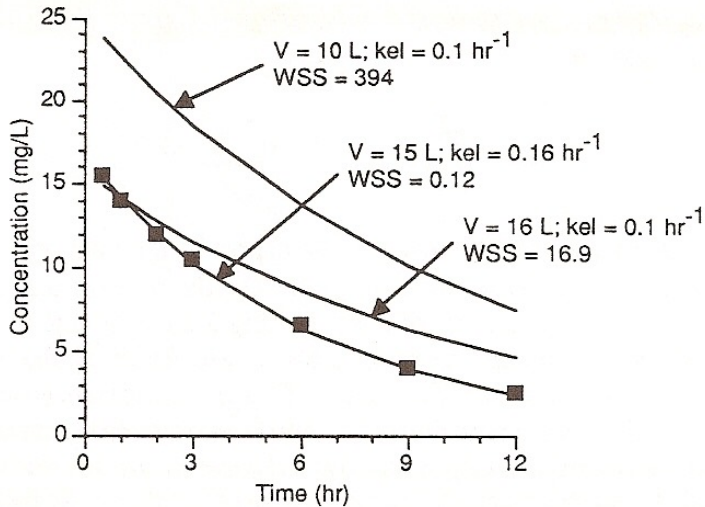
Mixed Effects Modelling

Fixed Effects (predictable variability)

Covariates and parameters
e.g. renal function and clearance

Random Effects (unpredictable variability)

Parameter variability e.g. in clearance
Residual error e.g. measurement error, process noise, model misspecification, assay error, transcription



Minimisation successful

Population Modeling – a logical processor



A green rectangular sign with white text, mounted on two wooden posts. The sign lists statistics for New Cuyama. The background shows a rural landscape with a field, a utility pole, and buildings in the distance under a cloudy sky.

NEW CUYAMA	
Population	562
Ft above sea level	2150
Established	<u>1951</u>
TOTAL	4663

Paediatric Studies Difficult



HHS Public Access

Author manuscript

Pain. Author manuscript; available in PMC 2019 February 01.

Published in final edited form as:

Pain. 2018 February ; 159(2): 193–205. doi:10.1097/j.pain.0000000000001104.

Clinical trial designs and models for analgesic medications for acute pain in neonates, infants, toddlers, children, and adolescents: ACTION recommendations

Gary A. Walco^{a,b}, Ernest A. Kopecky^{c,d}, Steven J. Weisman^e, Stevens^d, Paul J. Desjardins^f, Charles B. Berde^g, Elliot J. Krane^h, Myron Yasterⁱ, Carlton D. Dampier^k, Robert H. Dworkin^l, Ian Krane^h, Lynne G. Maxwellⁿ, Srinivasa Raja^j, Bernard Schachtel^o, and William T. Zempsky^m

Pediatric Analgesic Clinical Trial Designs, Measures, and Extrapolation: Report of an FDA Scientific Workshop

The Lancet Child & Adolescent Health Commission

Addressing practical challenges in clinical research sponsored by the US Food and Drug Administration on aspects of clinical trial design, standard parallel-placebo trials, and how to address ethical and practical challenges of subjects experiencing acute pain.



AUTHORS: Charles B. Berde, MD, PhD,^{a,b} Gary A. Walco, PhD,^{cd} Elliot J. Krane, MD,^{ef} K. J. S. Anand, MBBS, DPhil,^{g,h} Jacob V. Aranda, MD, PhD,ⁱ Kenneth D. Craig, PhD,^k Carlton D. Dampier, MD, CPI,^{lm} Julia C. Finkel, MD,^{no} Martin Grabois, MD,^{aq} Celeste Johnston, RN, PhD,^r John Lantos, MD,st Alyssa Lebel, MD,^{ab} Lynne G. Maxwell, MD,^{uv} Patrick McGrath, OC, PhD,^{wx} Timothy F. Oberlander, MD,^{yz} Laura E. Schanberg, MD,^{aa} Bonnie Stevens, RN, PhD,^{bb} Anna Taddio, BScPhm, MSc, PhD,^{bb} Carl L. von Baeyer, PhD,^{cc} Myron Yaster, MD,^{dd} and William T. Zempsky, MD^{ee}

Delivering transformative action in paediatric pain: a Lancet Child & Adolescent Health Commission



Christopher Eccleston, Emma Fisher, Richard F Howard, Rebecca Slater, Paula Forgeron, Tonya M Palermo, Kathryn A Birnie, Brian J Anderson, Christine T Chambers, Geert Crombez, Gustaf Ljungman, Isabel Jordan, Zachary Jordan, Cairiona Roberts, Neil Schechter, Christine B Sieberg, Dick Tibboel, Suellen M Walker, Dominic Wilkinson, Chantal Wood

Clinical studies difficult and Sheiner proposed alternative approach

Clin Pharm Ther 1991

The intellectual health of clinical drug evaluation

Lewis B. Sheiner, MD San Francisco, Calif.

"We have to trust the scientific judgement of the scientists . . . Statistics should be their handmaidens, not their jailers."

—D. Salsburg¹

Let me introduce my topic by presenting three symptomatic examples of an intellectual illness from which I believe clinical research in general, and drug trials in particular, is suffering.

1. An irrelevant analysis. Fig. 1 is redrawn from a recent article in our Society's journal, CLINICAL PHARMACOLOGY & THERAPEUTICS.² The error bars are drawn on only one point on each curve, but these are typical of the rest of the points.

The figure shows a characteristic time course of drug action, independent of dose, and a clear progression of pain relief as the dose of bromfenac increases. Rather than stressing this compatibility with expectation and rather than deriving a dose-response curve from which the (time course of) response to any dose could be predicted, why do the authors confine themselves to pairwise comparisons of the mean (over time) response between doses and deem it relevant to tell us, for example, that "No significant distinction was achieved between the 10 and 25 mg bromfenac doses"²

2. A great waste. It seems that Smith Kline & French Laboratories (Philadelphia, Pa.) studied about

surely many tens, if not hundreds, of millions of dollars. Although proving unequivocally that the drug was effective, all those studies apparently did not reveal the minimum effective dose. This is not just an isolated instance. Temple³ has pointed out that for many drugs the dose approved initially is considerably greater than that ultimately found to be adequate. This has been true, for example, for many β -blockers and more recently for zidovudine.

3. An excessive conservatism. Imagine a double-blind crossover placebo-controlled study of a new drug for the prophylaxis of angina. Fig. 2 shows both the placebo and active drug data on one hypothetical patient. Here the intended dose of the new drug was one per day, and if the patient had taken the new drug as intended, there would be no reason to think it was superior to placebo on the basis of the similar frequency of anginal attacks on both treatments.

But what if a medication monitor had been used,⁴ and the pattern of dosage was the one listed as *Actual* in Fig. 2? It seems the patient did not take any drug at all for the first 2 days and had anginal attacks on days 2 and 3. He then took the drug for 3 days without angina, stopped taking the drug, and had an attack the next day. The same thing happened again after 1 week of full compliance without angina. Surely, if there were multiple records like this one and the placebo records did not show similar patterns, we would be

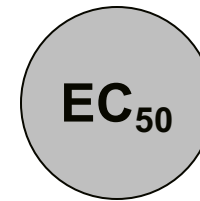
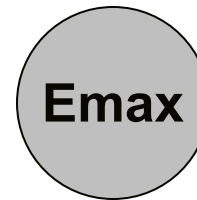
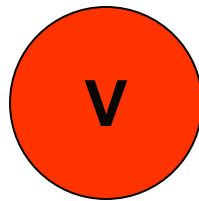
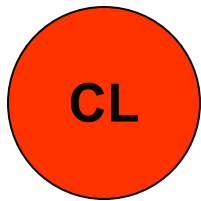
An approach around these difficulties

What do anaesthesiologists do?

Clinical Pharmacology

Pharmacokinetics

Pharmacodynamics



Concentration

Effect

Dose

$$C = \text{Dose} / V \cdot e^{-CL/V \cdot t}$$

$$\text{Effect} = \frac{E_{\max} \cdot C^N}{EC_{50}^N + C^N}$$

Example: What dose propofol in infants

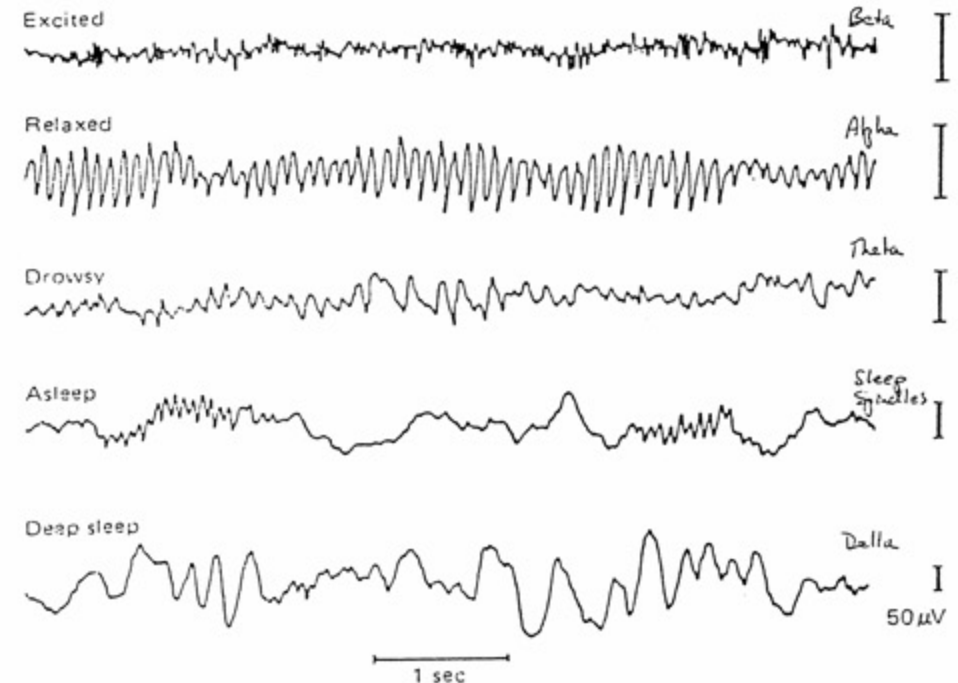


- PKPD parameters incorporated into pump

ANAESTHESIA IS NOT...

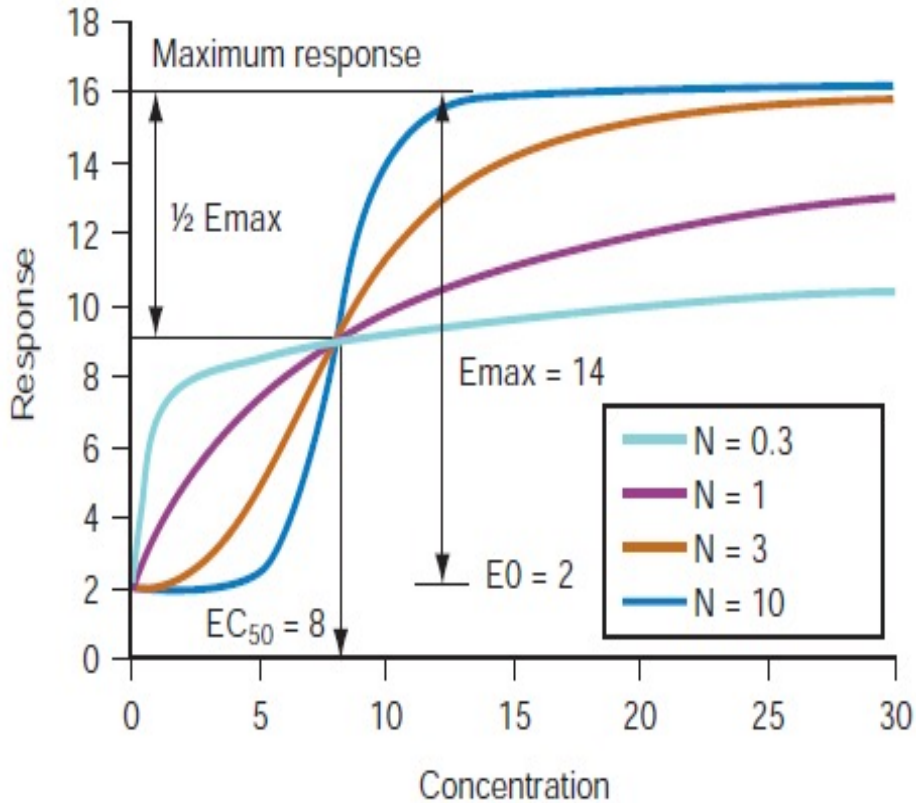
the half-asleep watching the half-wake being half-murdered by the half-witted

Malcolm Fisher

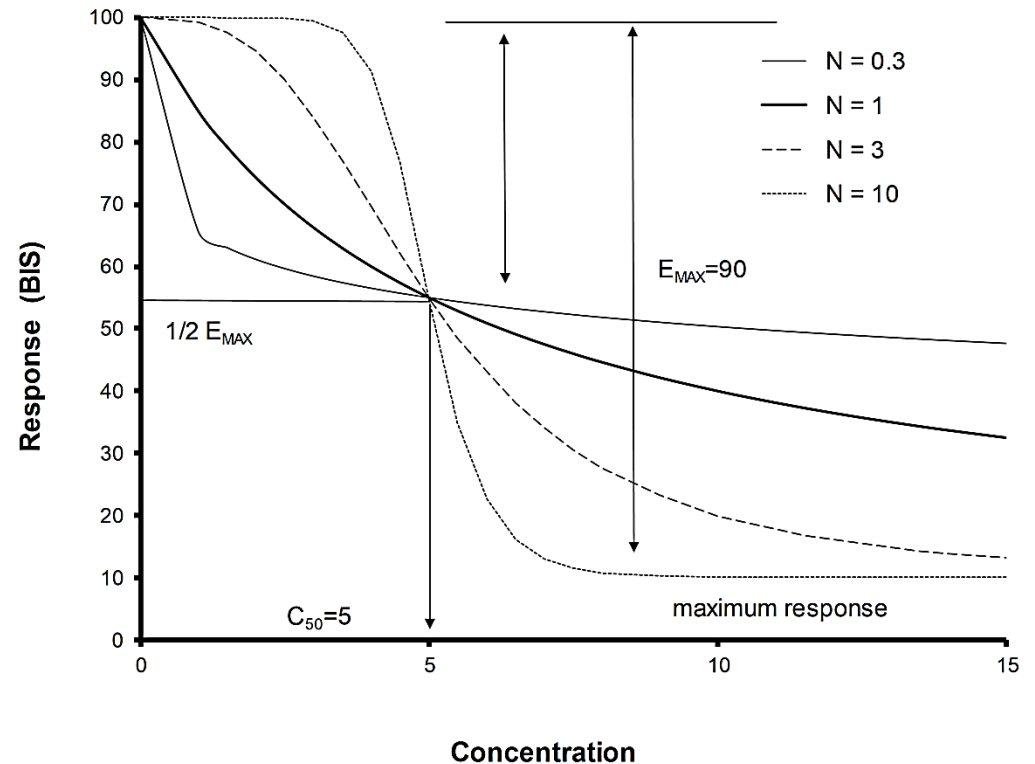


Emax Model Upside Down

Most Drugs



Bispectral Index

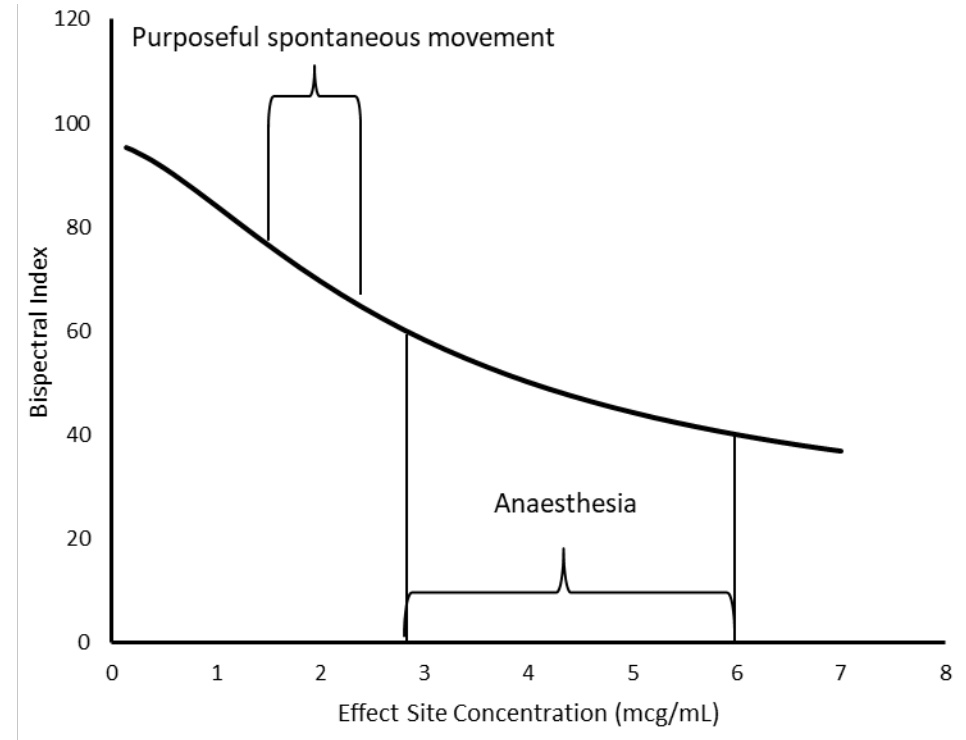


$$Effect = E0 + \frac{EMAX \times Ce^N}{(C_{50}^N + Ce^N)}$$

$$Bispectral\ Index = BIS_0 - \frac{EMAX \times Ce^N}{(C_{50}^N + Ce^N)}$$

Target Concentration (PD)

	Estimate	95%CI	Sh %	CV (%)
BIS₀	94	86.8, 94.6	48.1	5
E_{MAX}	0.81	0.70, 0.93	47.8	14.8
C_{50,PROP} (mg/L)	2.99	2.45, 3.66	47.1	35.4
T_{1/2keo,PROP} (min)	2.38	4.21, 14.6	45.9	67.6
Hill	1.55	1.31, 2.37	44.6	44.5
Additive residual Error (BIS units)	5.9	5.1, 6.8	-	η_{RUV} 0.363

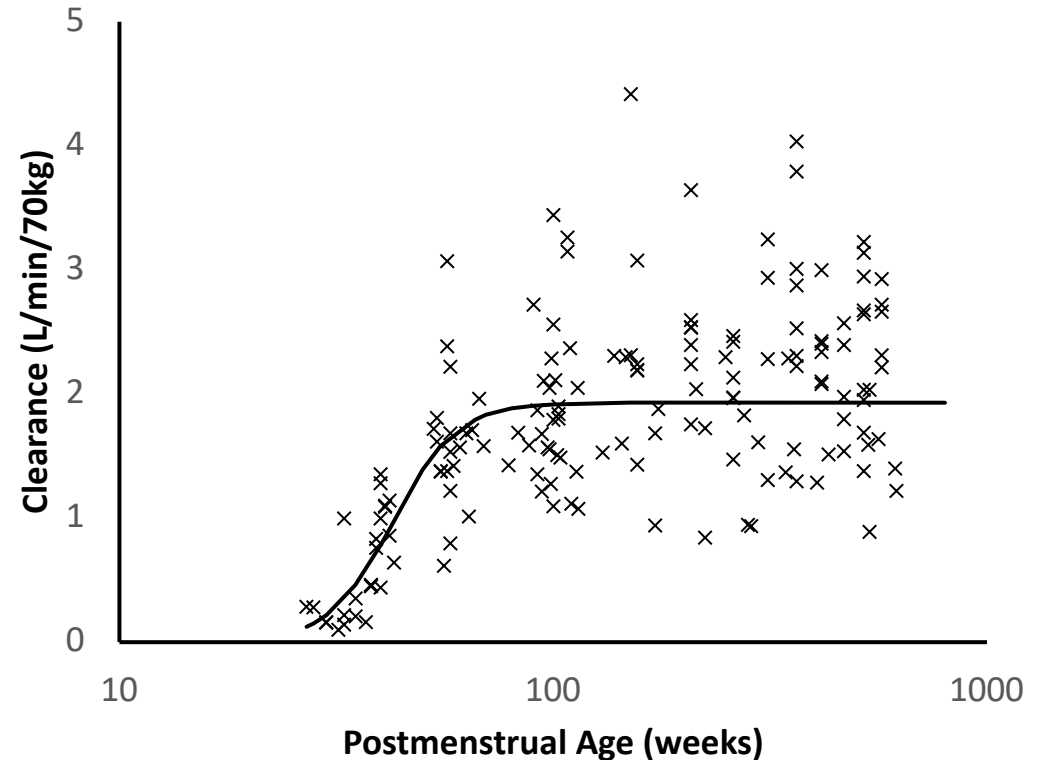


Target Effect BIS=50
Target concentration 3 mg/L

Fuentes Pediatric Anesth 2018

Propofol Clearance Variability (PK)

	Estimate	95%CI	Sh %	PPV (%)
V1 (L/70 kg)	18.5	5.2, 23.8	8.5	41.1
V2 (L/70kg)	41.1	29.2, 58.1	9.7	23.3
V3 (L/70 kg)	230	178, 390	14.5	50.3
CL (L/min/70 kg)	1.93	1.74, 2.19	2.9	40.7
Q2 (L/min/70kg)	3.82	3.24, 7.64	11.1	47.4
Q3 (L/min/70 kg)	0.837	1.09, 1.65	6.2	69.6
TM50	42.6		-	-
Hill	5.88			-
Additive residual Error ($\mu\text{g}\cdot\text{mL}^{-1}$)	0.012	0.0002, 0.0184		η_{RUV} 0.56
Proportional Residual Error (%)	16.9	12.5, 28.3		



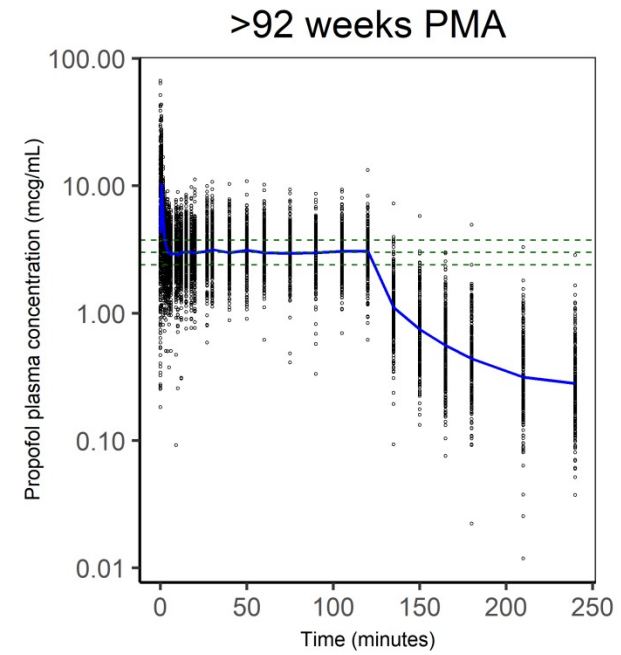
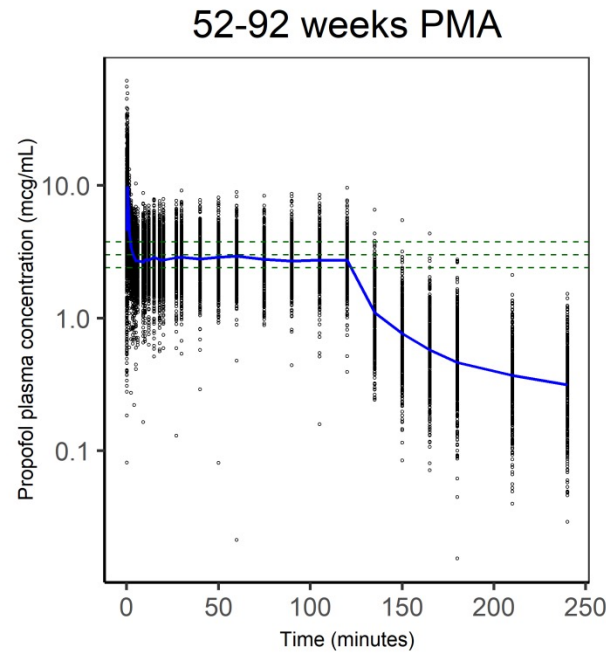
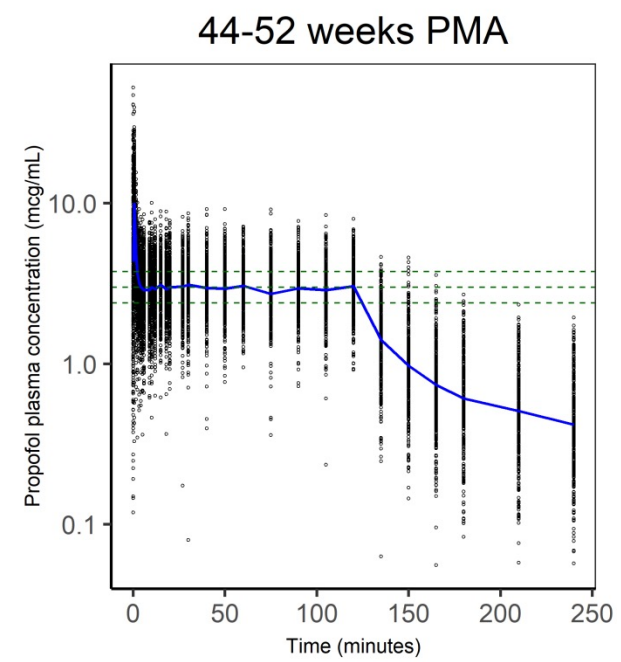
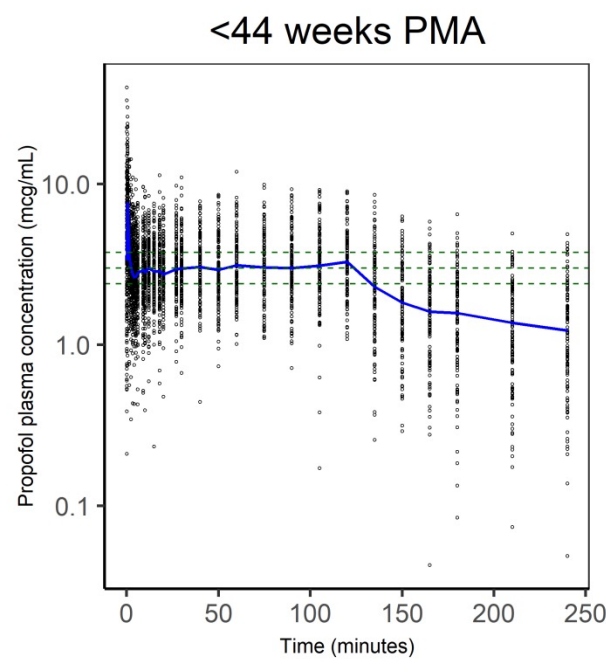
$$MF = \frac{PMA^{Hill}}{TM_{50}^{Hill} + PMA^{Hill}}$$

$$CL_{GRP} = CL_{STD} \cdot \left(\frac{WT}{WT_{STD}} \right)^{3/4} \cdot MF$$

Propofol Dosing Regimen

Age	Induction dose (mg/kg)	0-15 min	15-30 min	30-60 min	60 – 120 min
Target plasma concentration 3 $\mu\text{g}\cdot\text{mL}^{-1}$					
27-44 PMA weeks	2	9	7	6	5
44-52 PMA weeks	2.5	11	10	9	8
3-12 months	2.5	12	11	10	9
1-3 years	2.5	13	12	11	10

Simulation of 1000 individuals reveals unexplained variability



Half of the predictions were in the range 80-125% of the target concentration between 5 and 90 minutes of infusion duration

Target Concentration Strategy

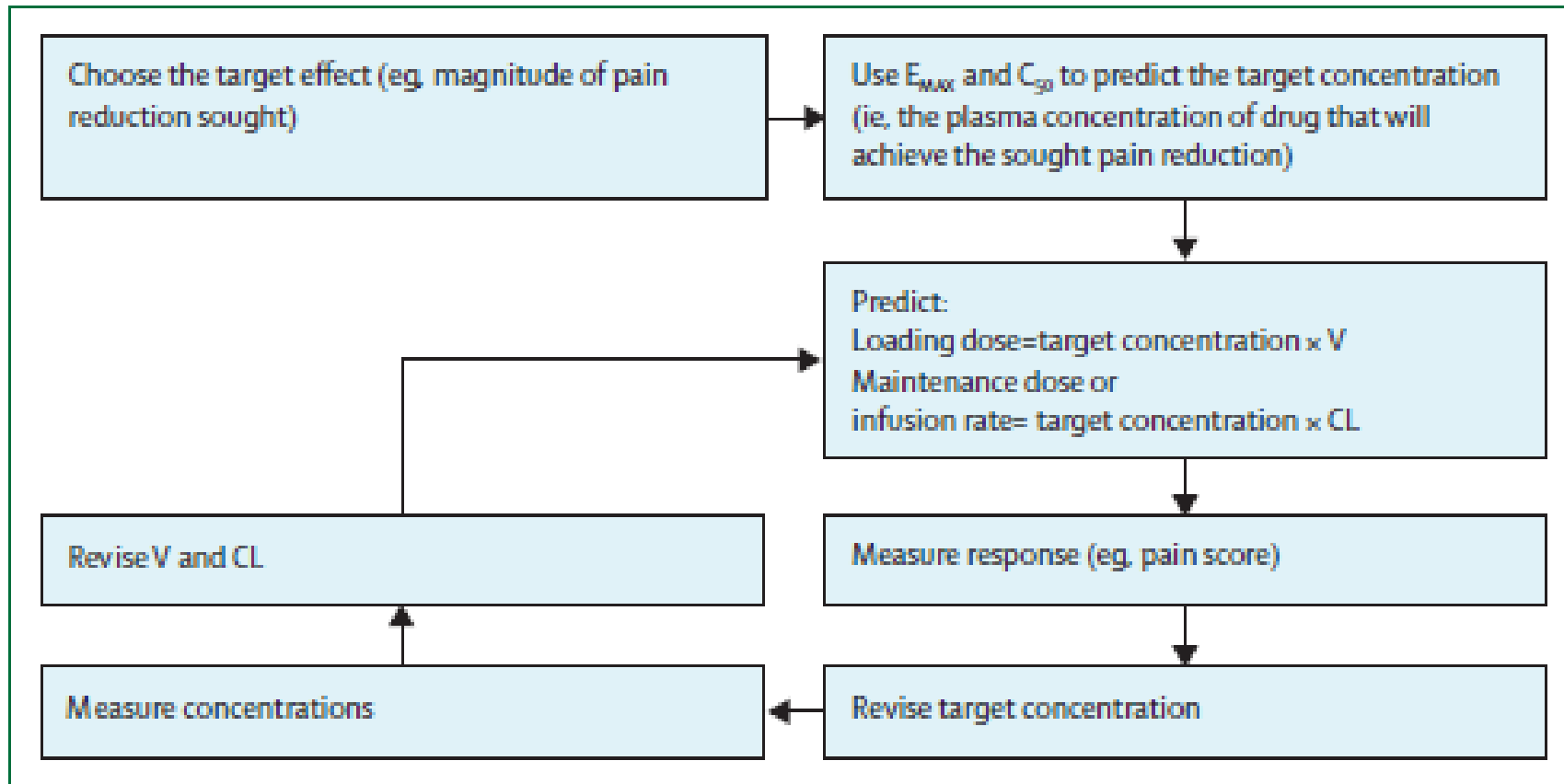


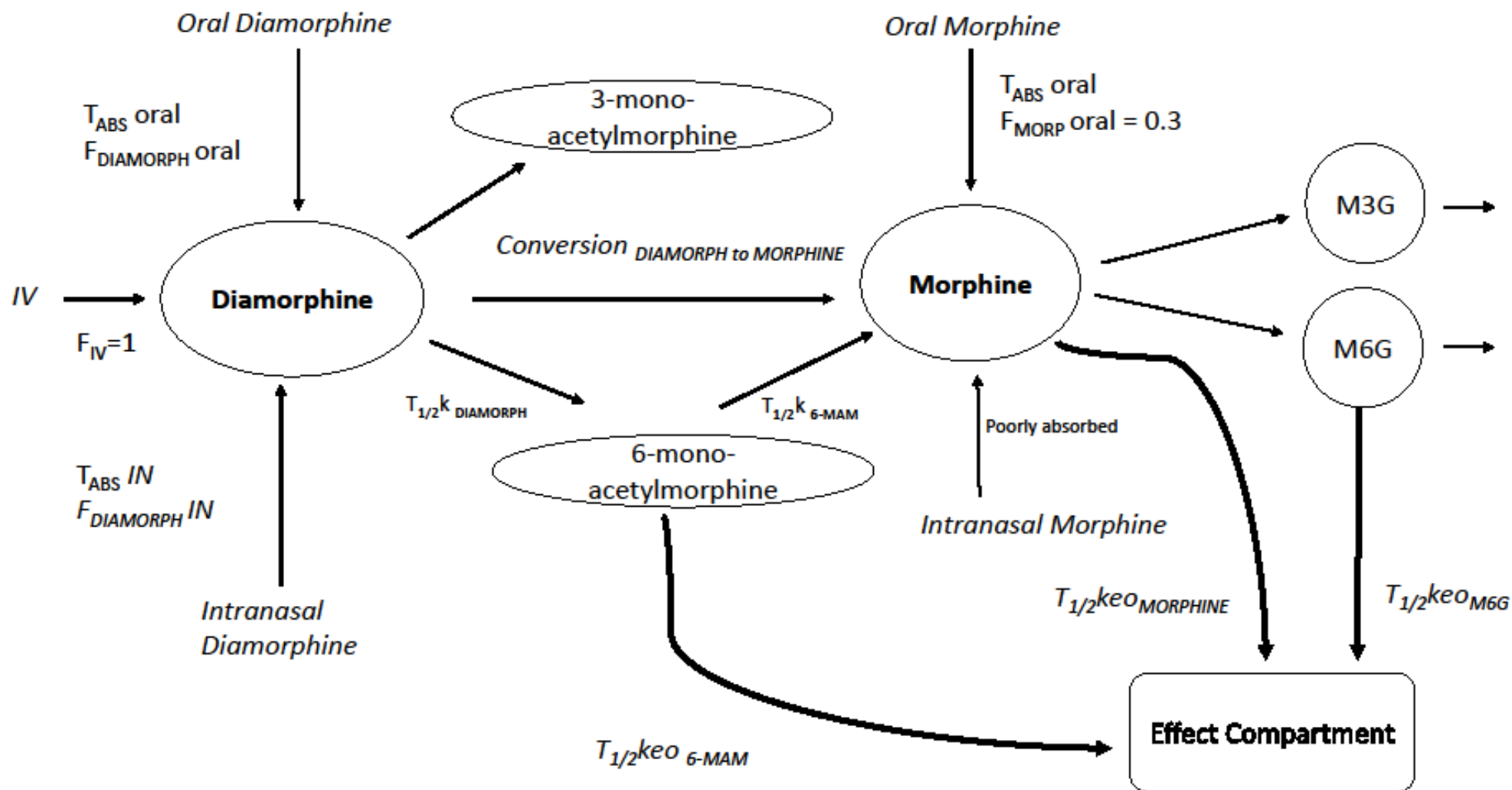
Figure 6: Strategy to determine the correct dose of pharmacological treatment

C_{50} =concentration at 50% of maximum response. CL =clearance. E_{max} =maximum effect. V =volume of distribution.

Application of principles to Opioids: Example, Diamorphine (Heroin) age < 2 years

- Use restricted to a few countries
- Paediatric dose acute and palliative care unknown
- Renewed interest in the drug, clinical trials advocated
- Metabolism complex, never described in children (PK)
 - Maturation
 - Size
 - Physiological functions (renal system clears some metabolites)
- Effect attributed to metabolite, morphine (PD)
 - Other metabolites also have effect
 - No concentration-response relationship for morphine
 - Drug interactions
 - Pharmacogenomics

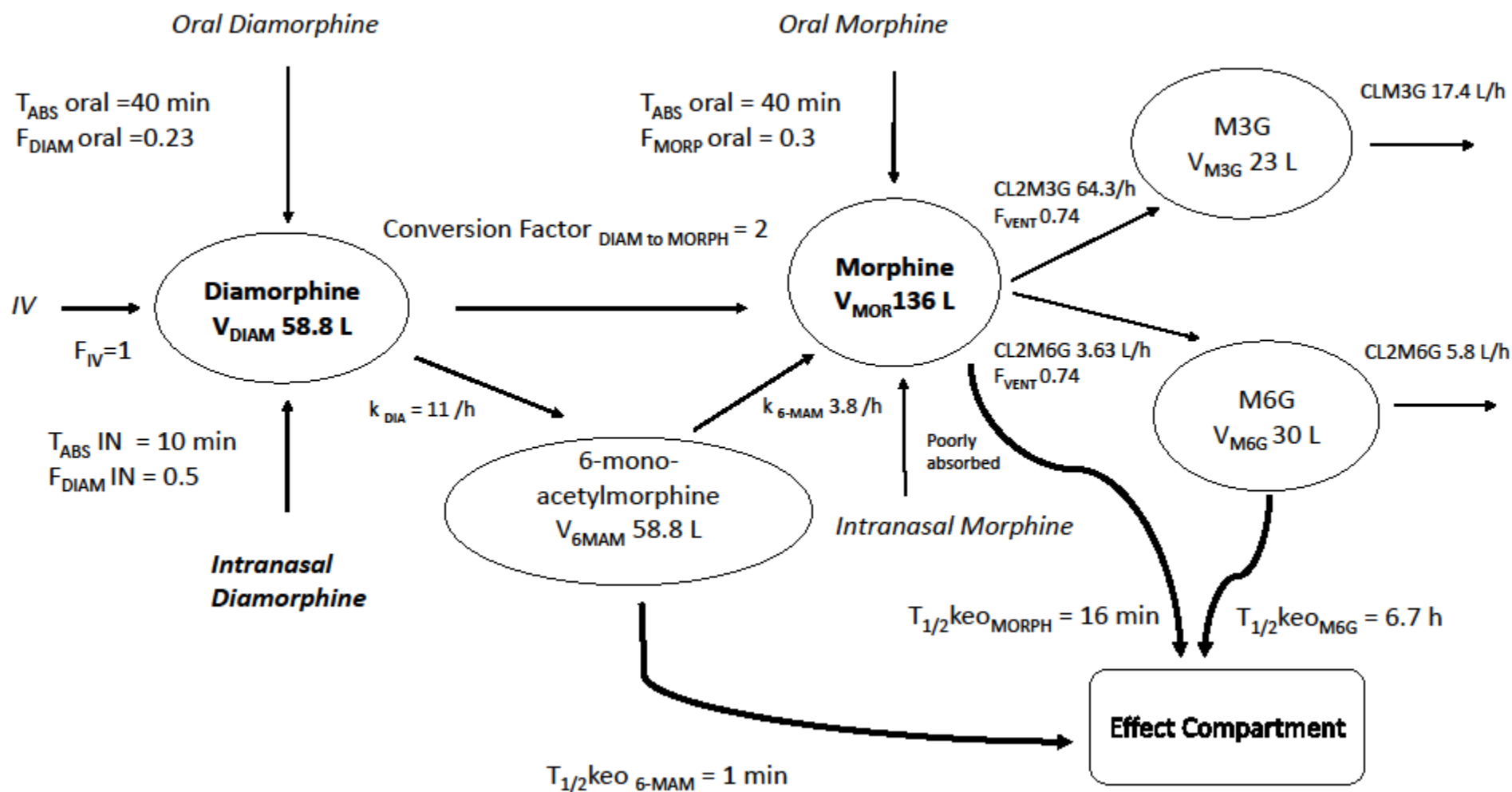
Diamorphine Metabolism



Diamorphine Pharmacokinetics

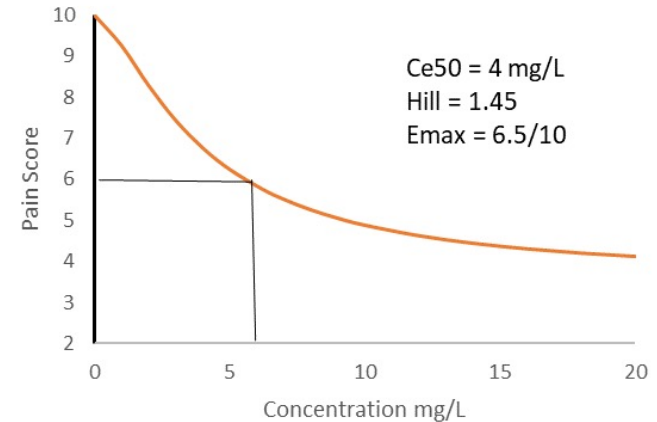
- Adult rate constants known
- Pediatric morphine PK described
- Formulation bioavailability assumed
- PK maturation models known (premature neonate to adult)
 - Premature neonatal data (Barrett 1991-6, ventilation)
- Renal function maturation known
- Size factors assumed (allometry)
- PBPK modelling supportive

Diamorphine Metabolism



Diamorphine Pharmacodynamics

- Link parameters known ($T_{1/2}$ keo)
- No Concentration-response
 - unlike NSAIDs, tramadol, acetaminophen
- Maturation receptors poorly defined
- Metabolite interactions (6-MAM, M6G)



Hannam JA, *Pediatr Anesth* 2018

Anderson BJ. *Pediatr Anesth* 2019

BUT

Target concentration defined

morphine 10 mcg/L

respiratory concentration-response same neonates-adults

Target Concentration Strategy

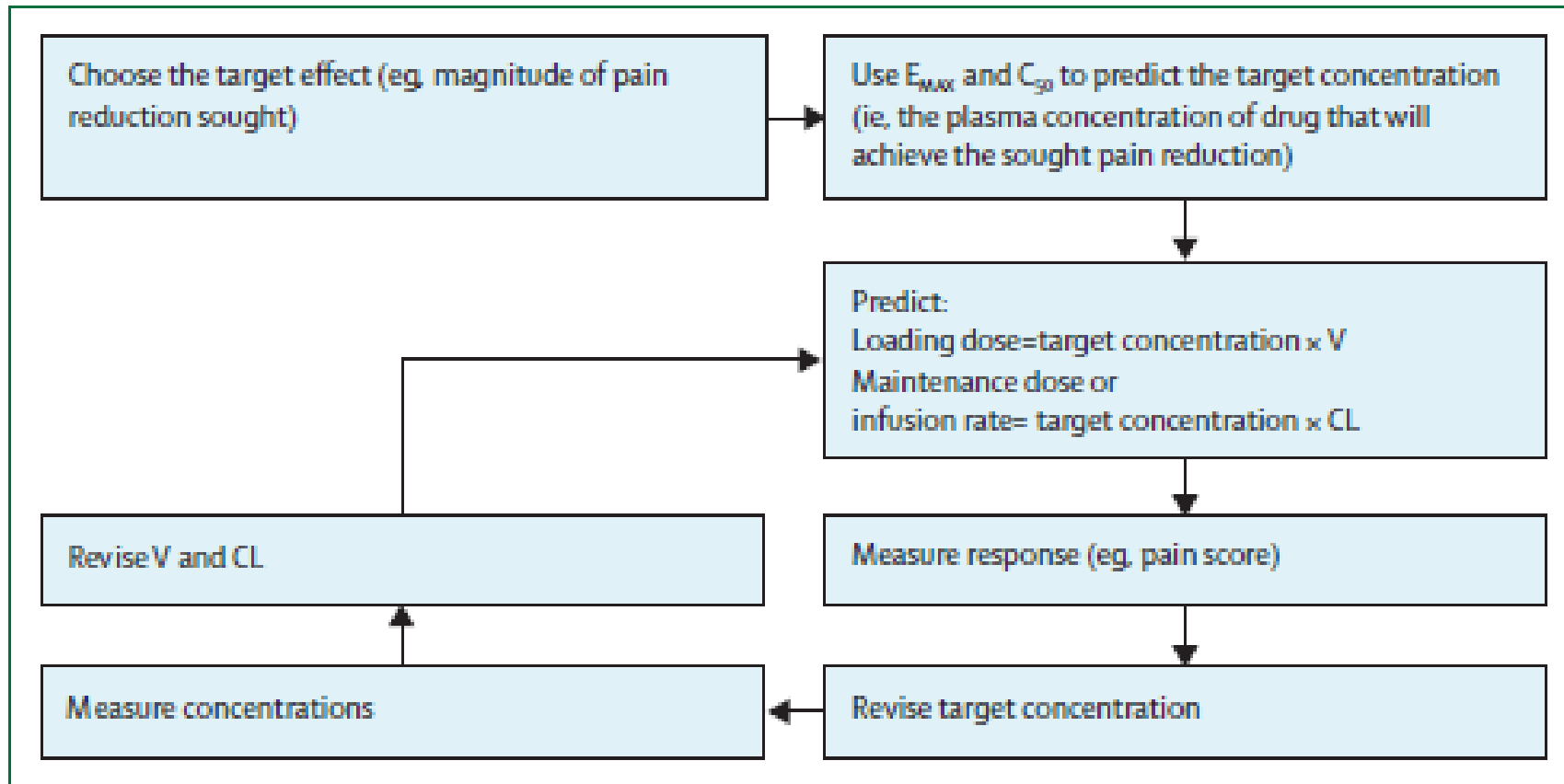


Figure 6: Strategy to determine the correct dose of pharmacological treatment

C_{50} =concentration at 50% of maximum response. CL =clearance. E_{max} =maximum effect. V =volume of distribution.

Population Modelling

Quantify the exposure-response relationship

Provide clarity and insight

Enable extrapolation beyond the observed data

Provide scientific rationale to dose selection

A knowledge management tool to capture and integrate pooled data from studies

Drive decision making during drug development

Hypothesis generating – the learning phase of drug development

Give a mechanistic understanding of the drug effect - theory enrichment

CLINICAL PHARMACOLOGY & THERAPEUTICS

VOLUME 61 NUMBER 3

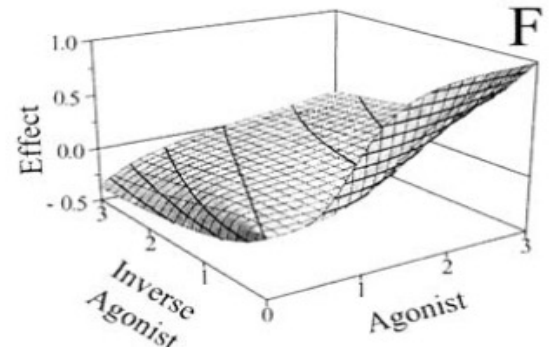
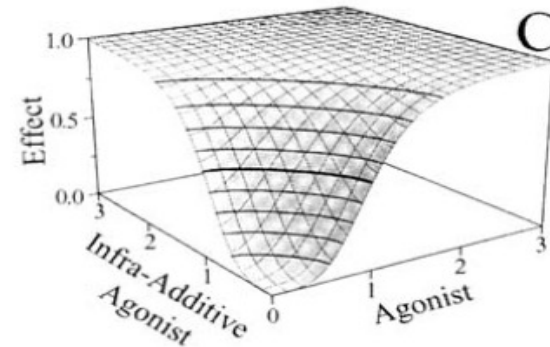
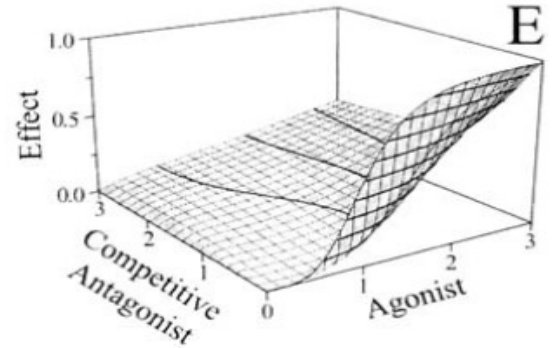
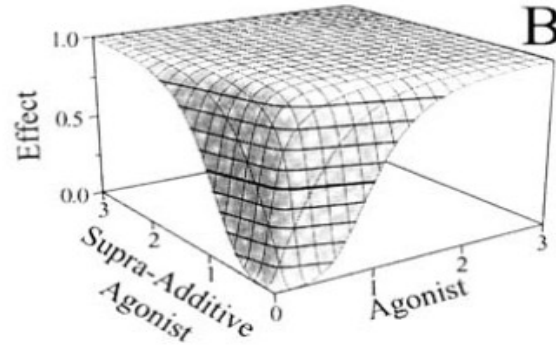
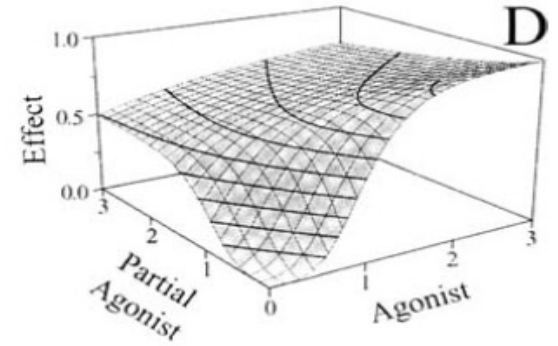
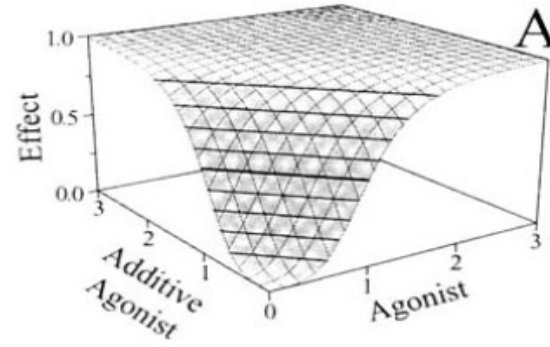
MARCH 1997

COMMENTARY

Learning versus confirming in clinical drug development

Lewis B. Sheiner, MD *San Francisco, Calif.*

Drug Interactions: surface responses



Opioid Drug Interactions

- Opioid drug interactions well described with anaesthesia agents
 - e.g., propofol-remifentanyl
- Simple analgesic interactions in older children described
 - e.g., Hannam JA, Anderson BJ, Potts A. Acetaminophen, ibuprofen and tramadol analgesic interactions after adenotonsillectomy. *Pediatr Anesth* 2018; 28: 841-851

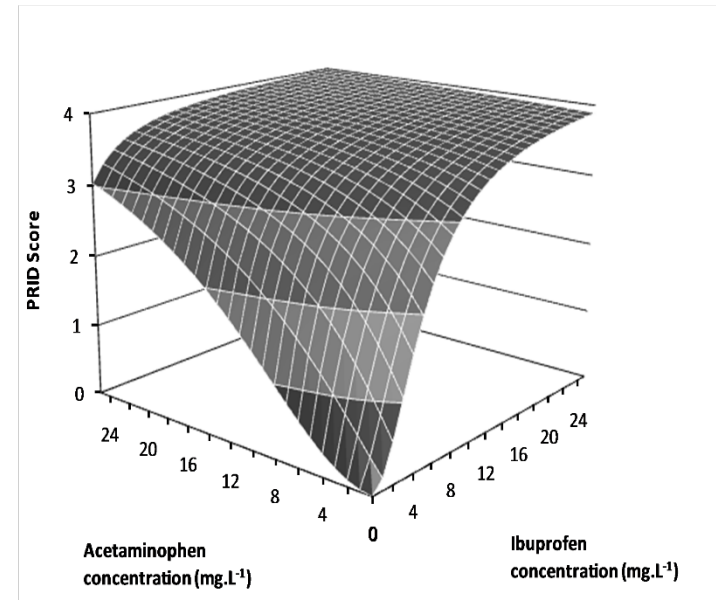
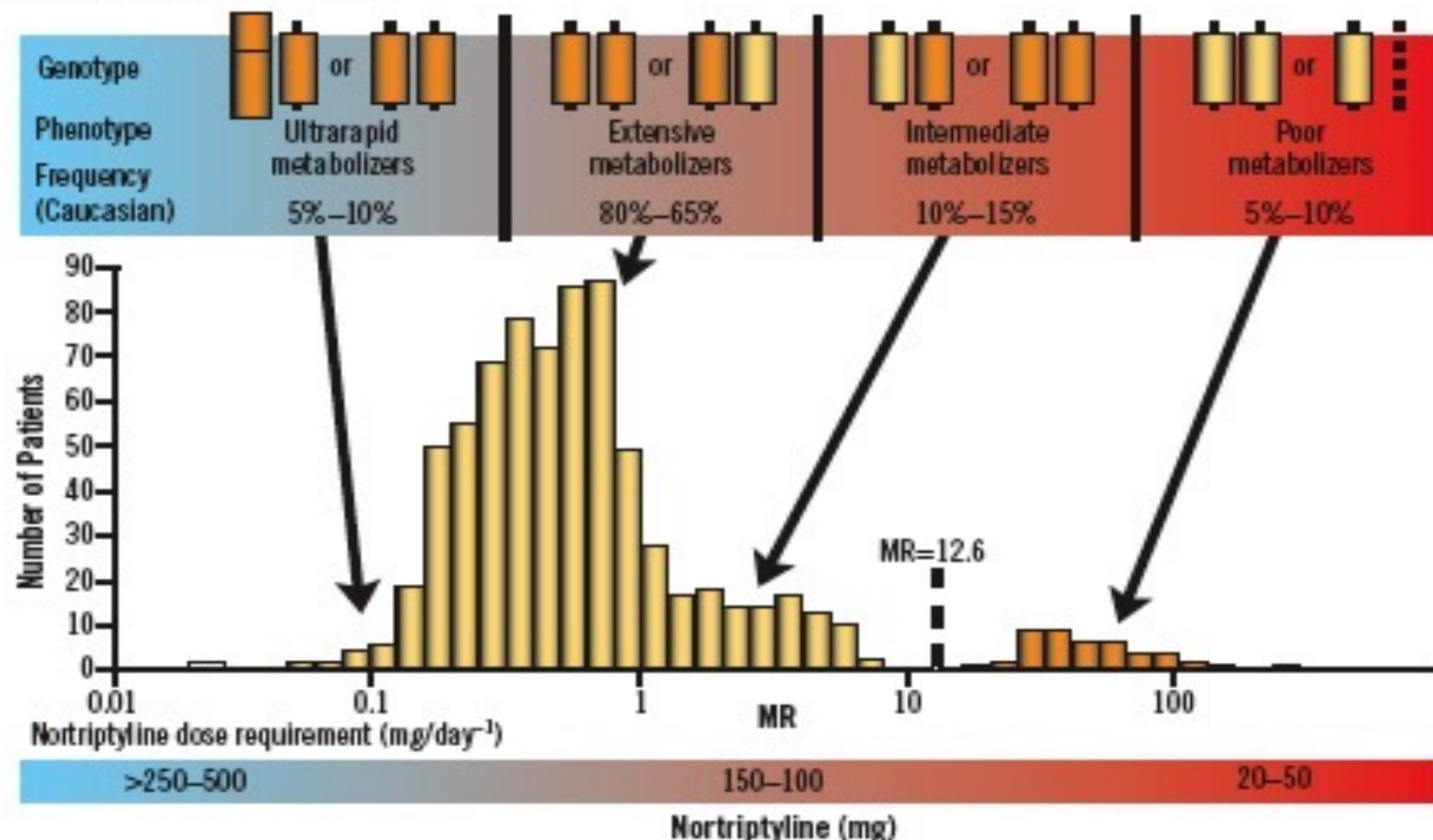


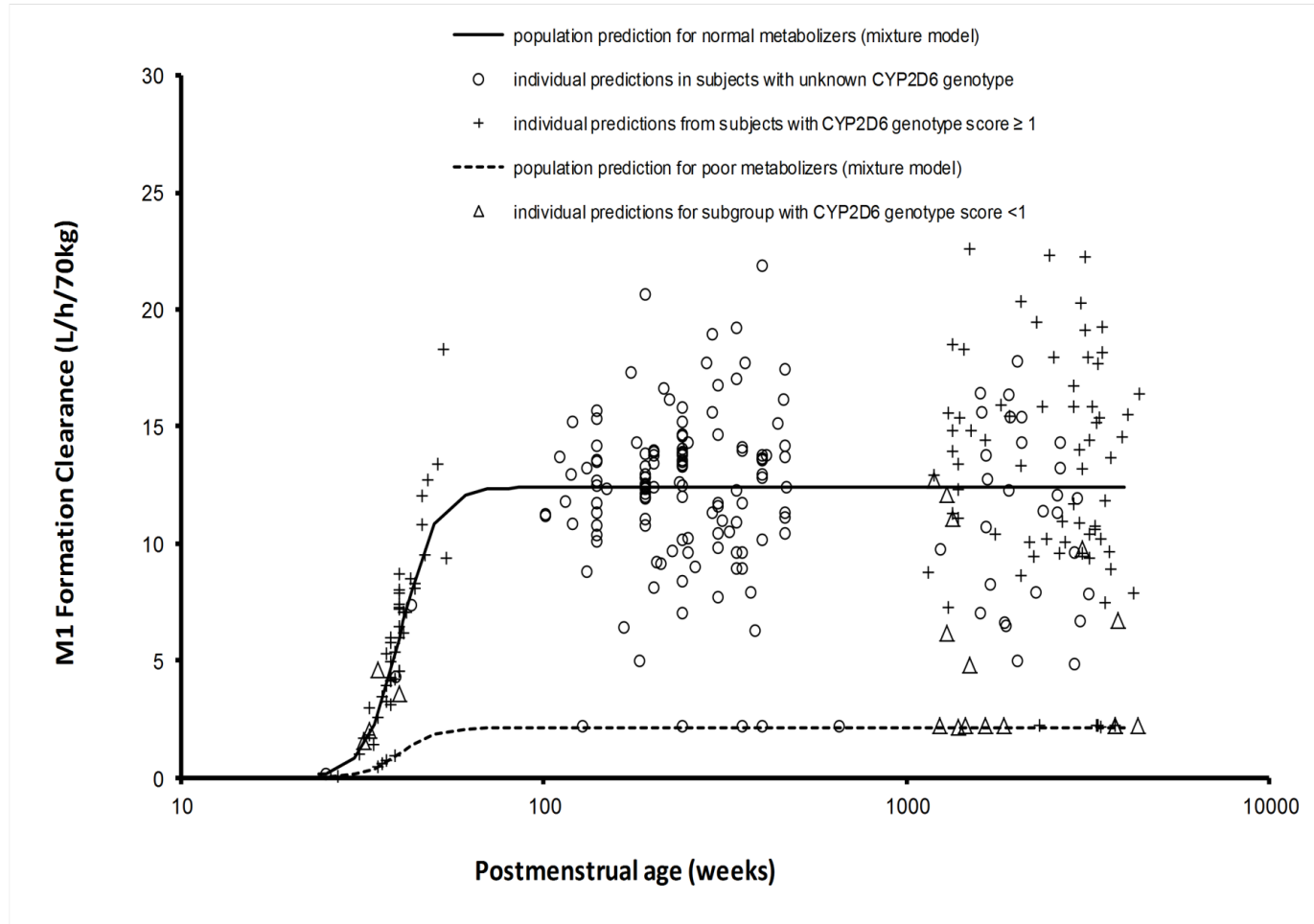
FIGURE 1

GENETICS OF CYP 2D6 METABOLIZING EFFECTS ON NORTRIPTYLINE



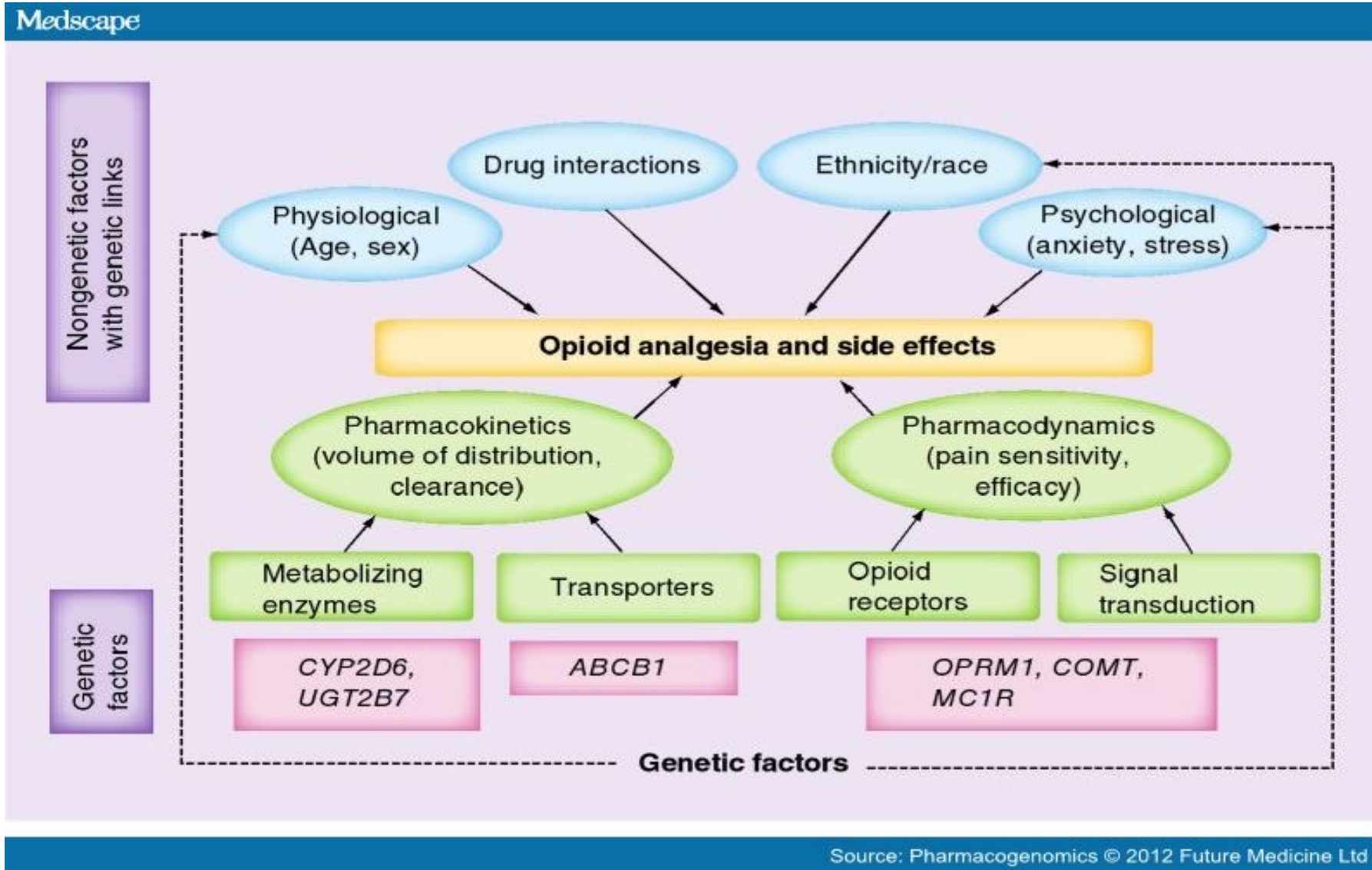
CYP=cytochrome P450; MR=metabolic ratio of parent debrisoquine ÷ metabolic OH-debrisoquine.

Impact of CYP2D6 on Tramadol Clearance



Holford SD et al. Parent-metabolite pharmacokinetic models for tramadol – tests of assumptions and predictions. J Pharm Clin Tox 2014;2(1):1023-34

Contributors to analgesic variability



The intellectual health of clinical drug evaluation

Lewis B. Sheiner, MD *San Francisco, Calif.*

Clin Pharm Ther 1991

"We have to trust the scientific judgement of the scientists . . . Statistics should be their handmaiden, not their jailer."

—D. Salsburg¹

surely many tens, if not hundreds, of millions of dollars. Although proving unequivocally that the drug was effective, all those studies apparently did not reveal the minimum effective dose. This is not just an isolated instance. Tamura^{2,3} has pointed out that for many drugs the

CLINICAL PHARMACOLOGY & THERAPEUTICS

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THE END

