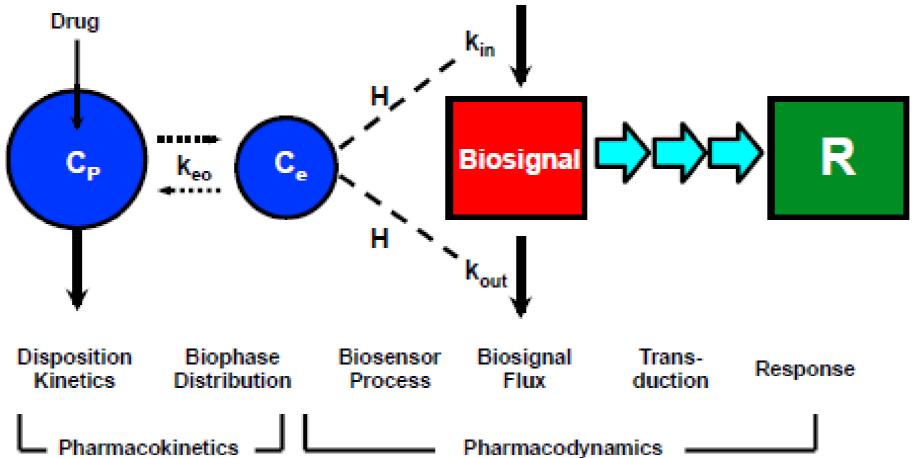
PKPD: PAST, PRESENT AND FUTURE

Applied to antibiotics - Prof. Dr. An Vermeulen - 23 Nov 2017



LABORATORY OF MEDICAL BIOCHEMISTRY AND CLINICAL ANALYSIS

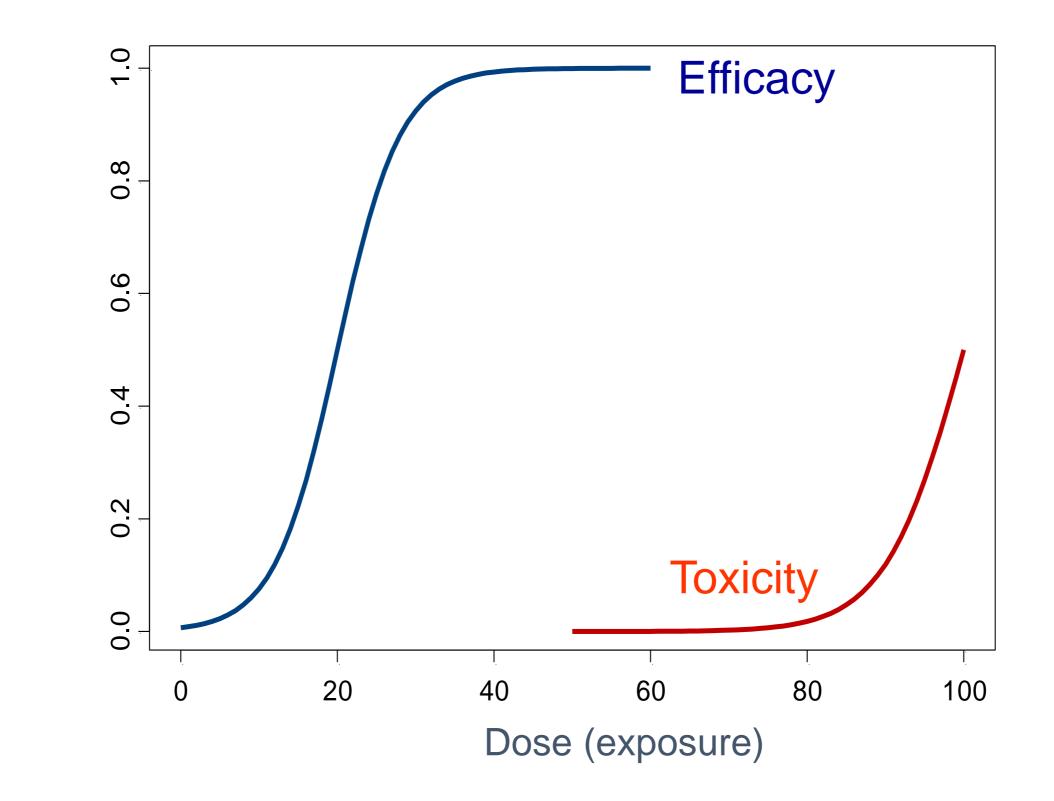
Components of Pharmacokinetic and Pharmacodynamic Systems



Adapted from Jusko et al., J Pharmacokinet Biopharm 23:5 (1995) Mini-review: Mager et al., Drug Metab Disp 31:510 (2003)



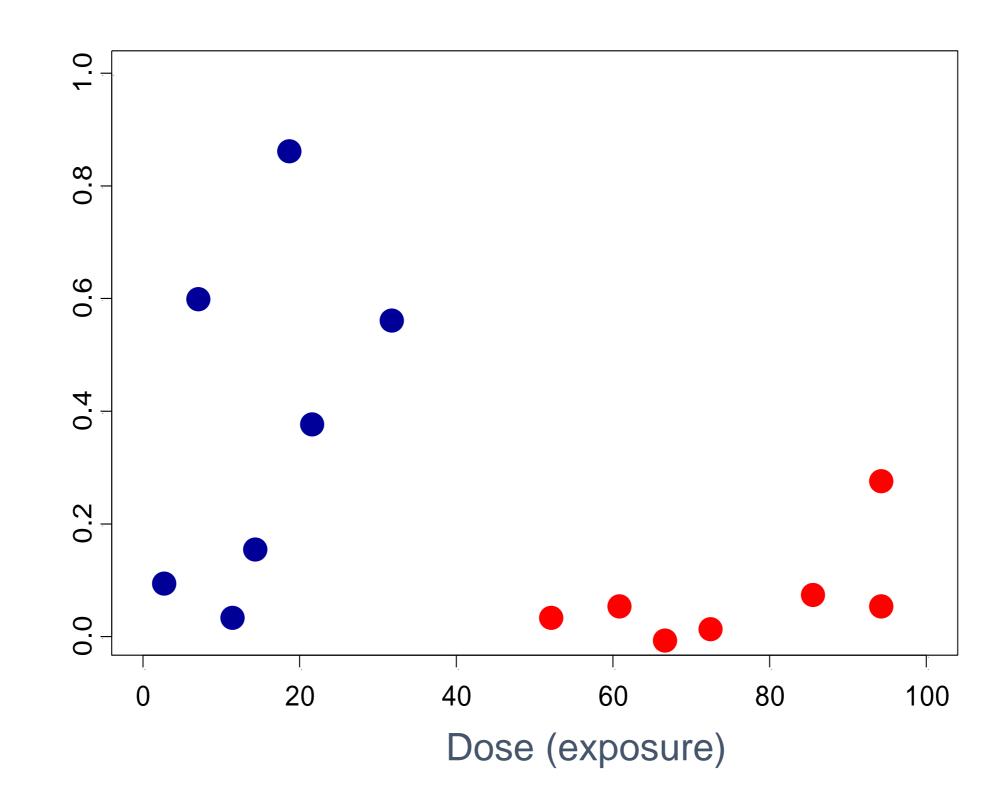
DOSE SELECTION MADE EASY





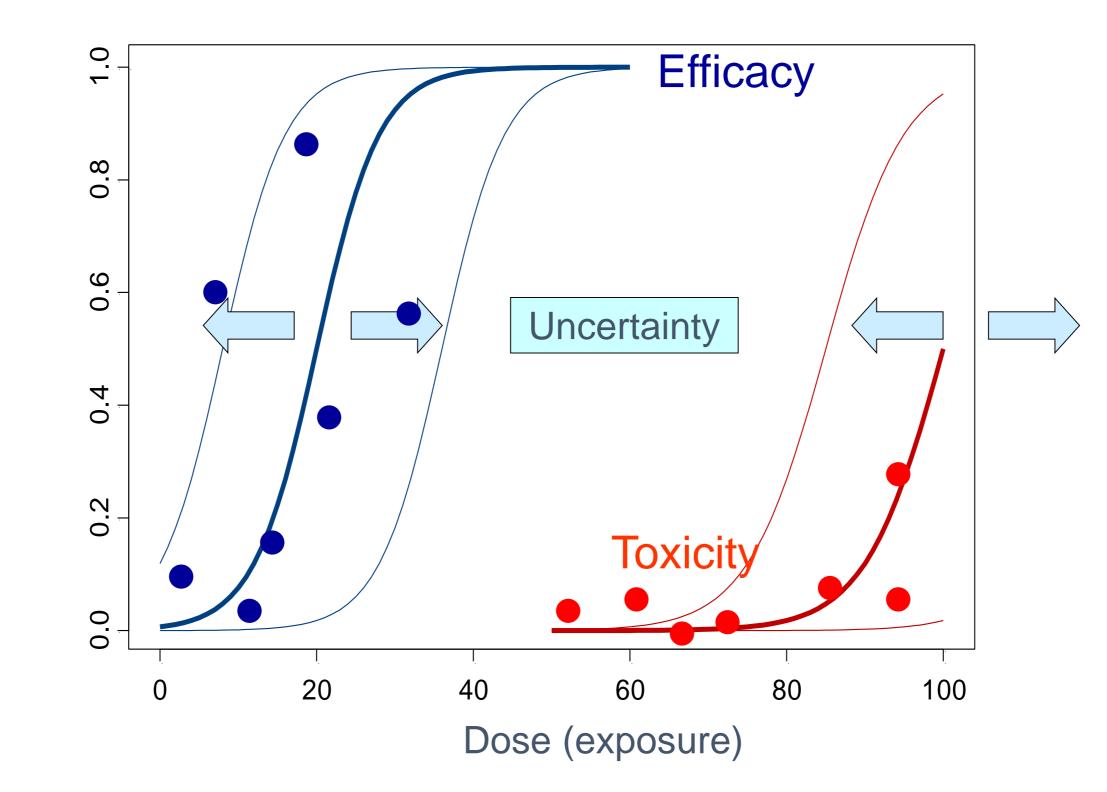


IN REALITY ...



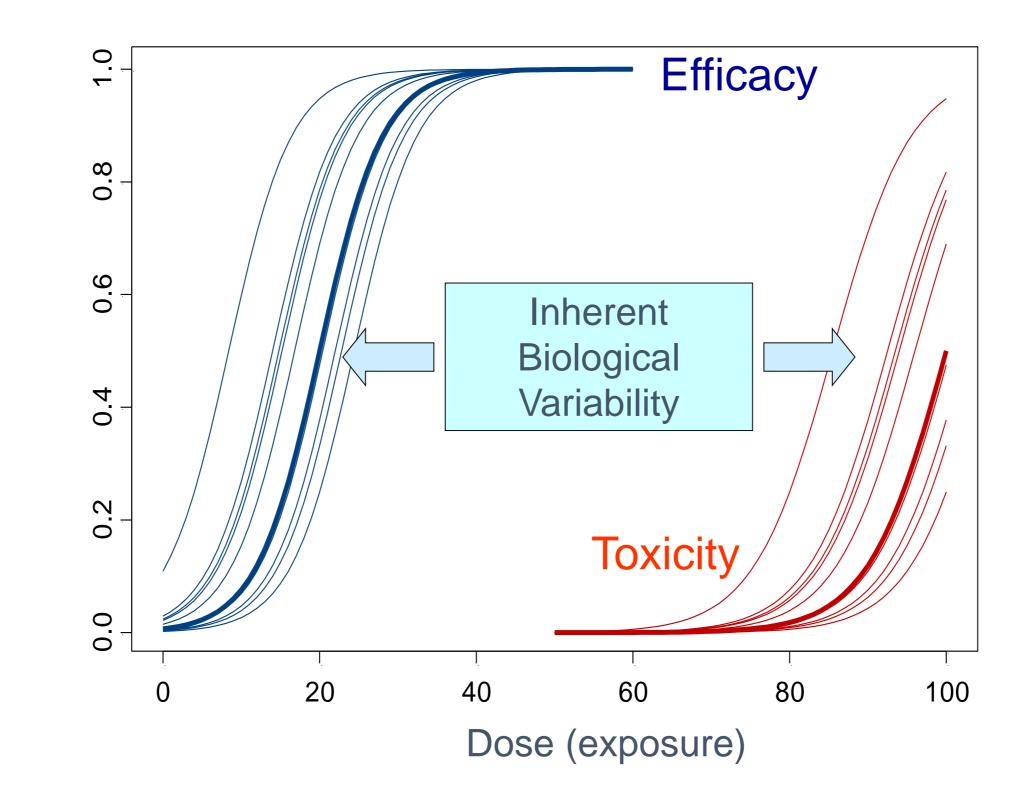








ALSO, ALL PATIENTS ARE DIFFERENT





PKPD MODELING

- We try to establish relationships between exposure (mostly plasma concentrations) and effects
 - In a way, if we focus on PK, it is because we believe it is a biomarker for effects, and that understanding the PK profile helps to understand the PD (effects) profile
 - Most of the times, we can not measure the concentrations at the effect site, and we sample plasma concentrations instead (easy matrix to sample and to measure concentrations in)
 - The relationship between plasma and effect site concentrations is then again linked mathematically
- Once a model is developed, it allows simulations of scenarios we have not studied/for which we do not have information
 - Time points w/o sampling
 - Additional doses/dose regimen
 - Repeated dosing (using single dose data)
 - At population & individual level
 - For different covariate subgroups (pediatrics, CrCL, E



PKPD MODEL APPLICATIONS

- **Conceptualization and quantification**
- In vitro \rightarrow In vivo extrapolation
- **Species scaling**
- **Clinical trial design**
- **Design of new dosage forms**
- **Comparative efficacy/toxicity**
- Patent enhancement
 - Labeling advice





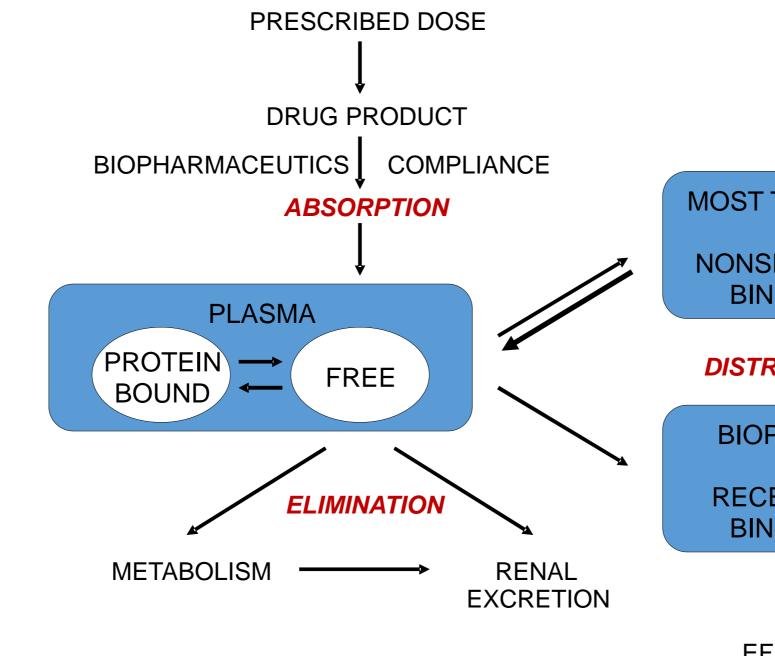
PHARMACOKINETICS





PHARMACOKINETICS

Pharmacokinetics is the study and mathematical characterization of the time course of drug absorption, distribution, metabolism, and excretion (**ADME**) processes that determine the time-course of drug action.





Adapted from Atkinson, Principles of Clinical Pharmacology (2001)

MOST TISSUES

NONSPECIFIC BINDING

DISTRIBUTION

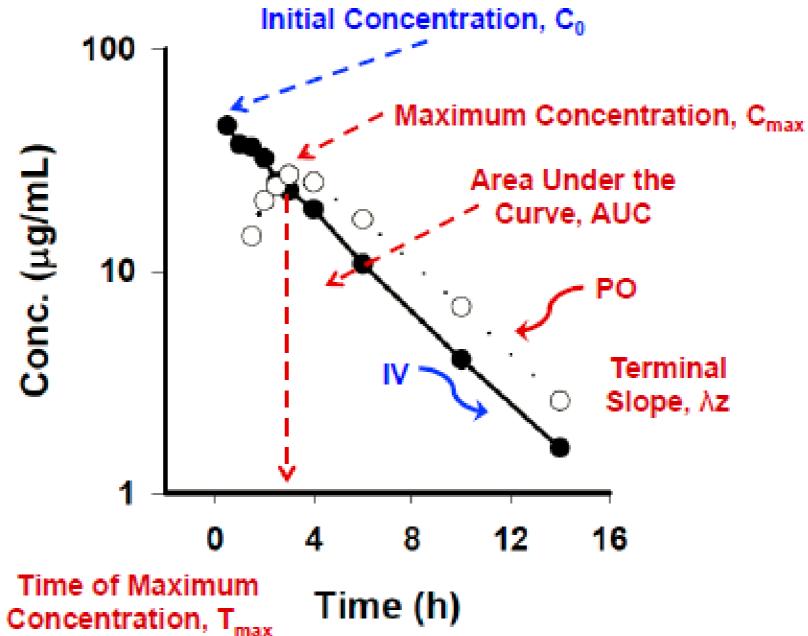
BIOPHASE

RECEPTOR BINDING

> • EFFECT

NONCOMPARTMENTAL ANALYSIS

Time-Course of Drug Exposure

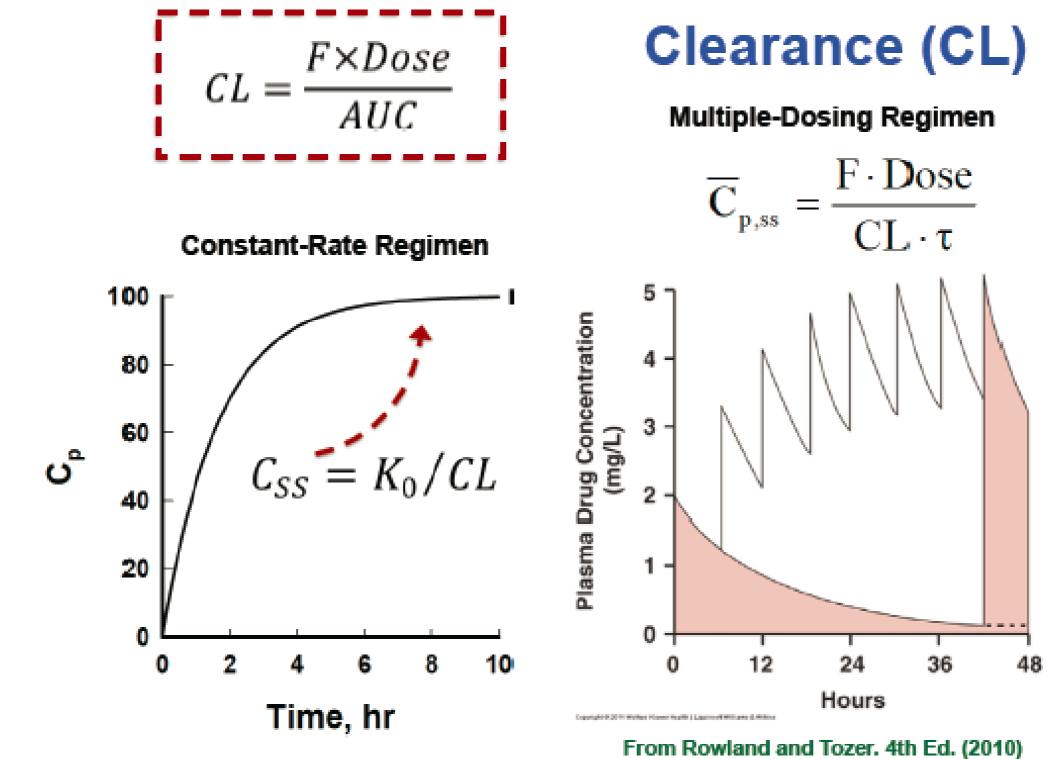






Terminal Slope, λz

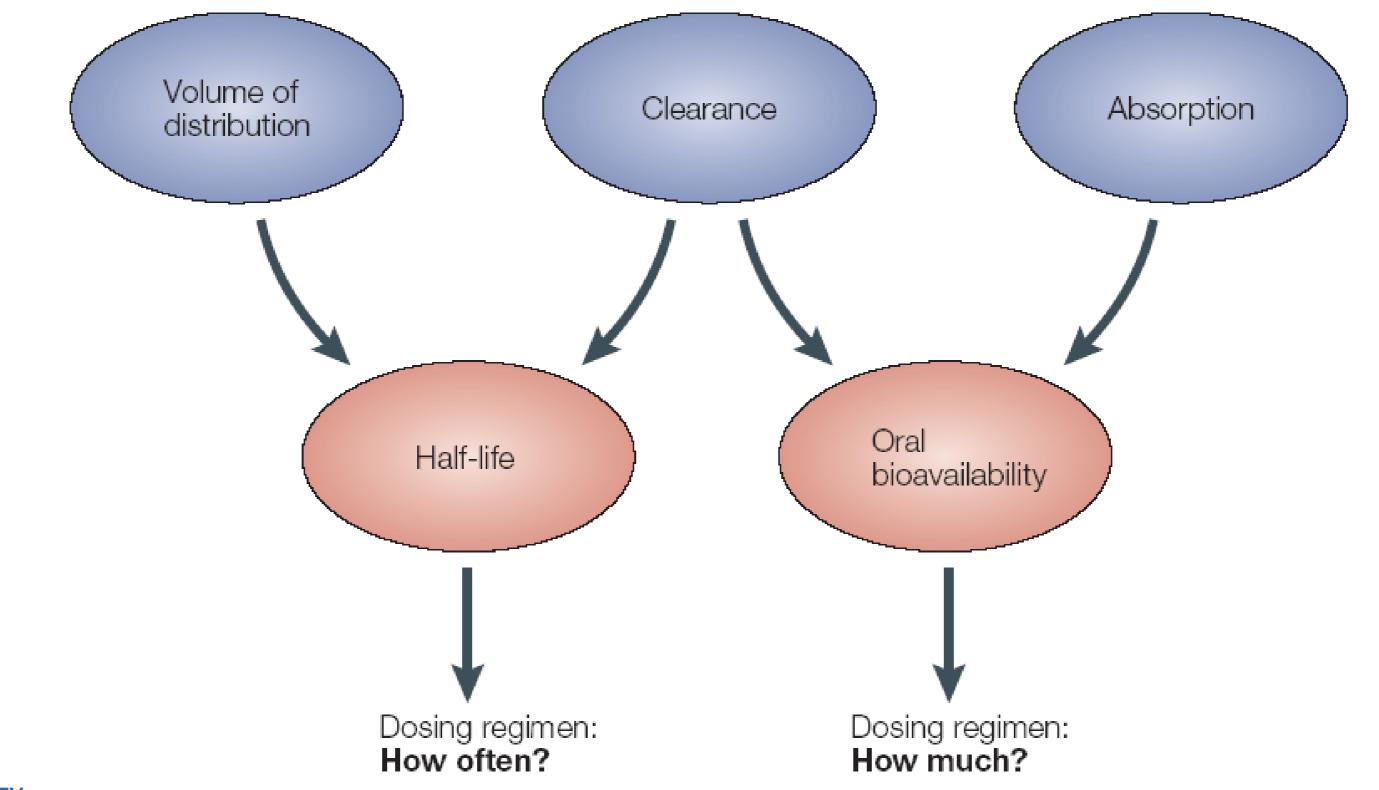
NONCOMPARTMENTAL ANALYSIS







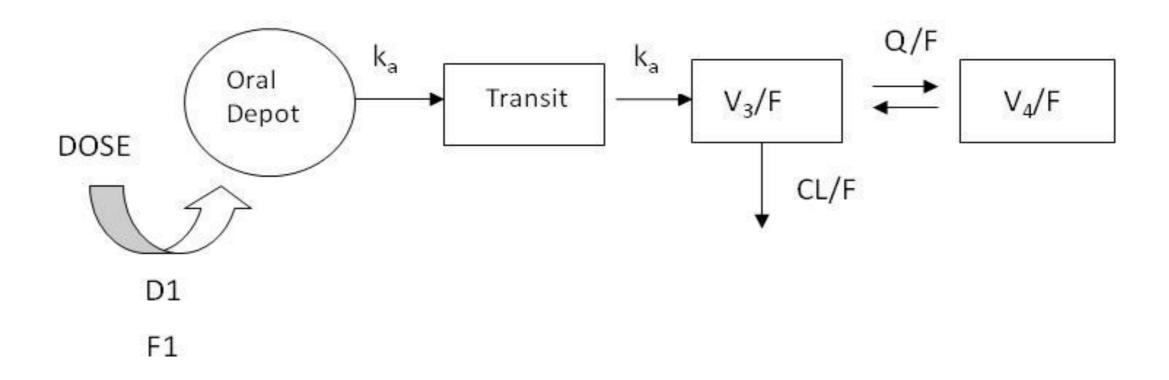
PRIMARY PHARMACOKINETIC PROPERTIES



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van de Waterbeemd and Gifford. Nature Rev Drug Disc. 2:192-204 (2003)

COMPARTMENTAL ANALYSIS

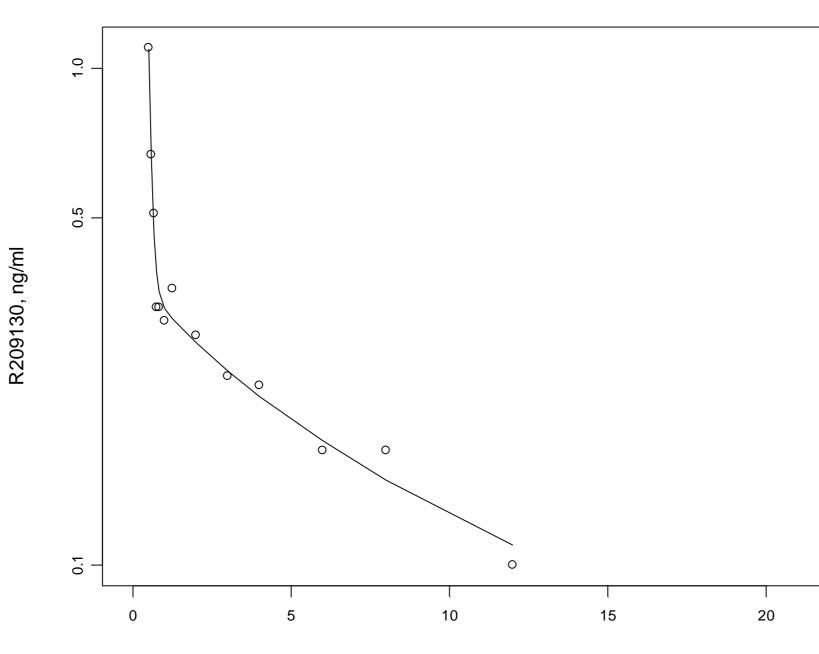






COMPARTMENTAL ANALYSIS

4 50001





Time, h



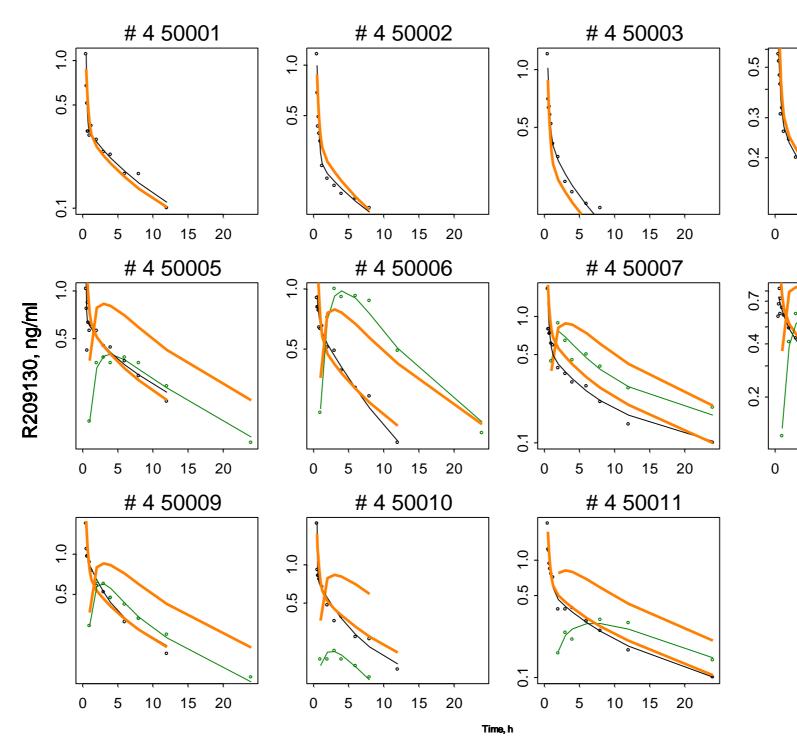
THE ASPECT VARIABILITY

- Modelling is essentially similar at the population level!
- But ..
 - At the population level, the aspect <u>variability</u> comes into play
 - In a population analysis, the focus is on
 - estimating the magnitude of this variability
 - trying to identify patient characteristics that can explain (part of) this variability
 - dose-adjustments in subsets? LABELLING impact !
 - dose-individualisation in individuals? (avoid TDM therapeutic drug monitoring)
- So ..
 - Three levels of variability have to be considered:
 - IIV or BSV: inter-individual or between-subject variability
 - IOV: inter- or between-occasion variability _
 - Intra-subject, within-subject or residual (unexplained) variability —

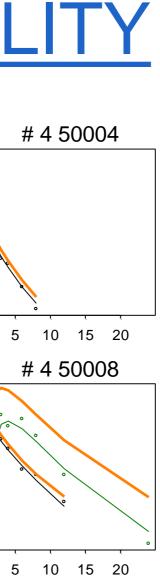




THE ASPECT VARIABILITY







POPULATION PK PRINCIPLES

- Nonlinear Mixed Effects Modelling (NONMEM)
- Mixed effects
 - Fixed effects (θ)
 - Random effects (η , ε)

Fixed effects

- Estimate mean PK(PD) parameters in the population (= typical values, central tendency, population average, ...)
- Identify factors that influence the PK (PD) parameters (demographics, lab values, PGx, smoking habits etc...)

Random effects _

- Estimate unexplained variability
 - => IIV, IOV and residual variability

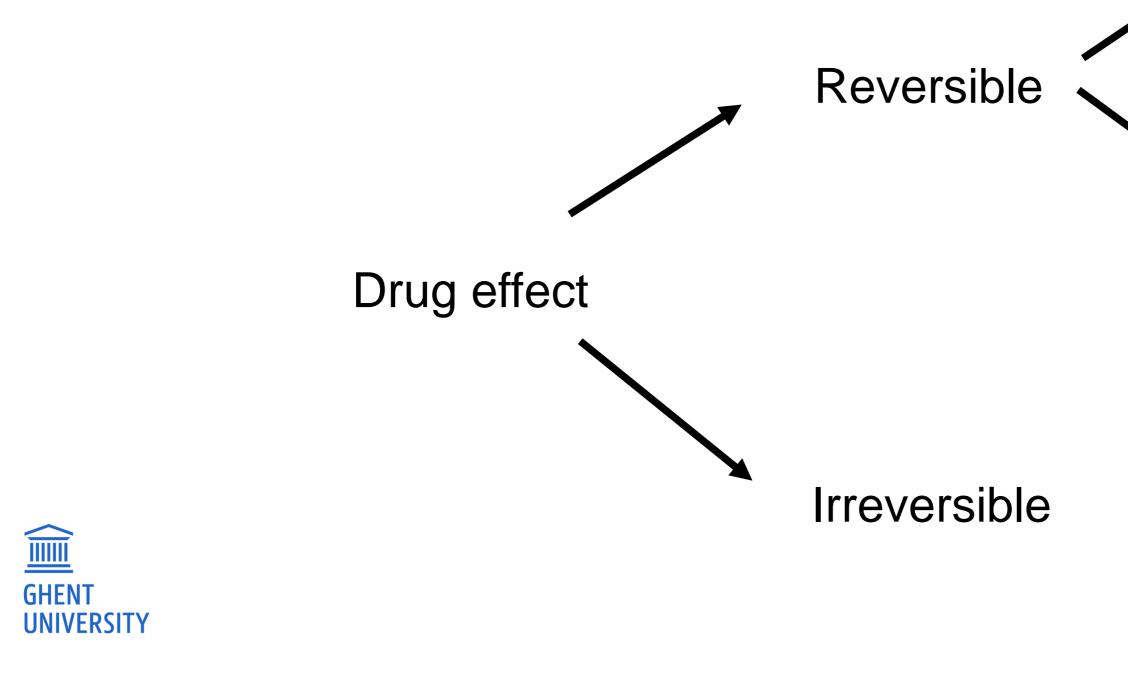


PHARMACODYNAMICS





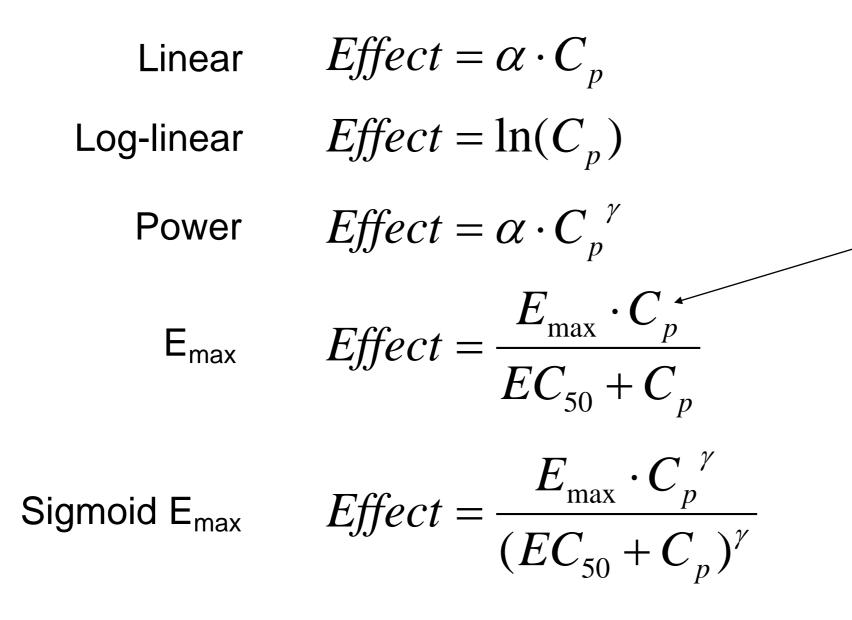
PK-PD RELATIONSHIPS: TYPE OF EFFECTS



Direct

Indirect

REVERSIBLE EFFECTS – DIRE CONCENTRATION-EFFECT MODELS

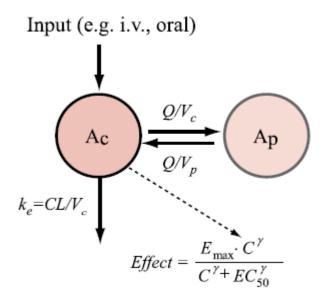




 EC_{50}/IC_{50} values: the potency of the drug E_{max}/I_{max} values: the efficacy of the drug γ : sigmoidicity factor

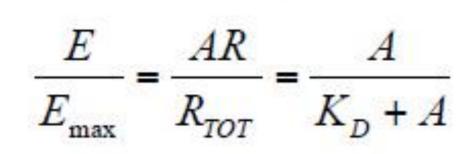


Effect directly related to plasma concentration



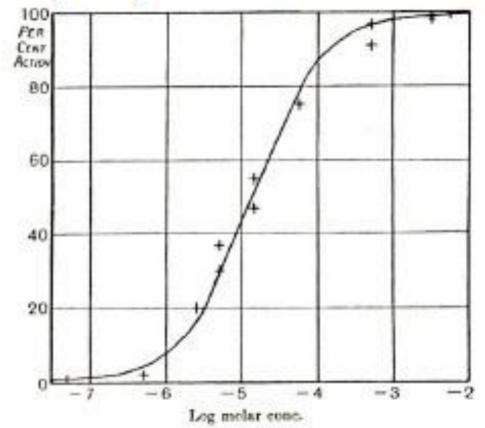
Occupancy Theory

- A. V. Hill
 - Law of mass action for occupancy
- A. J. Clark



Hill Equation

$$E = \frac{E_{\max} \cdot C^{\gamma}}{EC_{50}^{\gamma} + C^{\gamma}}$$





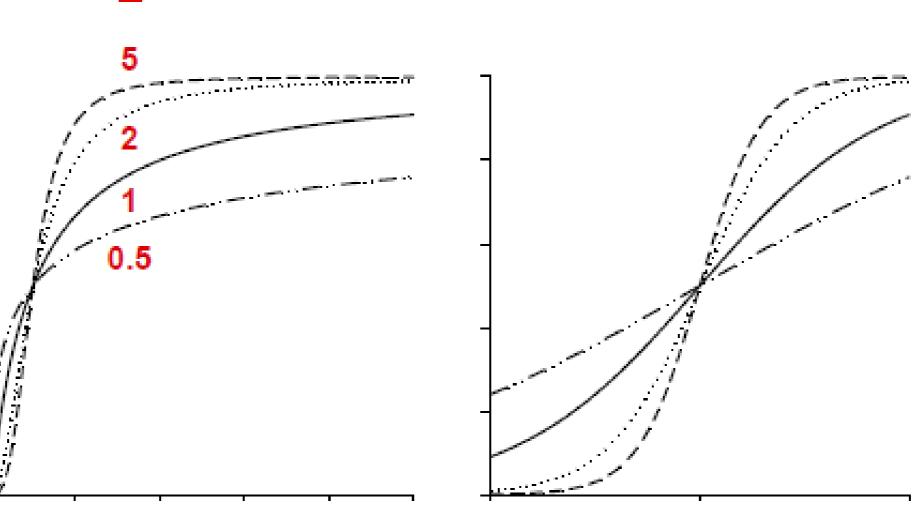
ACh induced contraction of frog rectus abdominis muscle. Clark, J Physiol. (1926)





Concentration-Effect Relationship

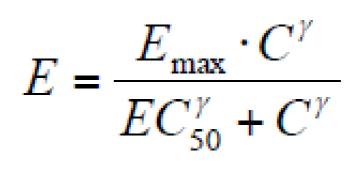
Effect (%)

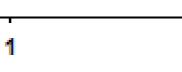




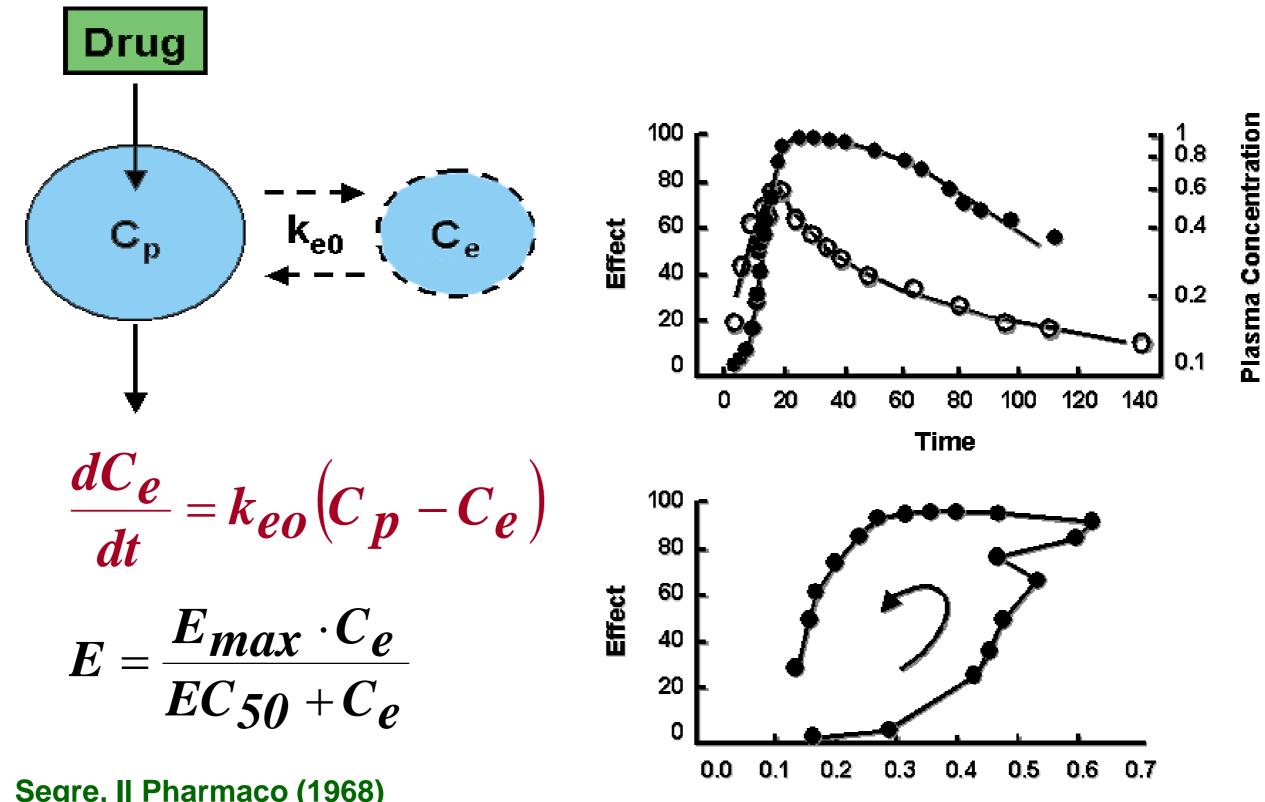
Concentration

0.1





EVERSIBLE EFFECTS – INDIRECT **MODELING A BIOPHASE**



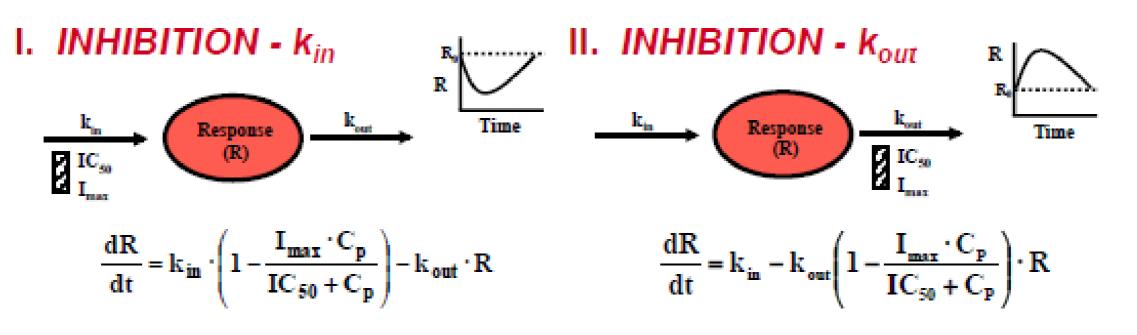


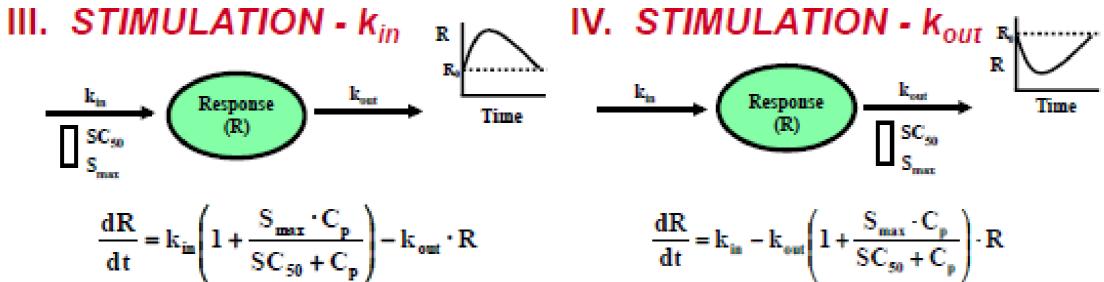
Segre, II Pharmaco (1968) Sheiner et al., CPT (1979)

Plasma Concentration

REVERSIBLE EFFECTS – INDIRECT

Family of Indirect Response Models







Dayneka et al., JPB 21:457 (1993); Sharma and Jusko. BJCP. 45: 229 (1998)

$$\frac{S_{max} \cdot C_p}{SC_{50} + C_p} \cdot R$$

IRREVERSIBLE EFFECTS

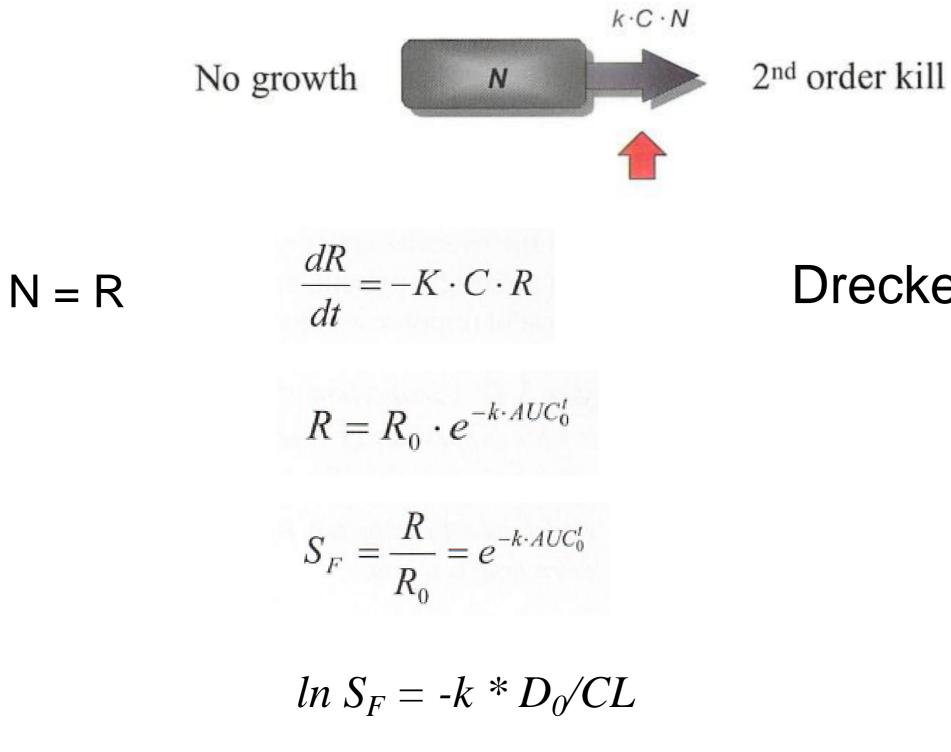
- Chemotherapeutic effects
 - Simple cell killing
 - Killing and regrowth
 - Cell cycle effects
 - Clinical therapy
- Irreversible enzyme inhibition
- Reactive drug metabolites
- Simple drug toxicity



Carcinogenicity models

26

SIMPLE CELL KILLING





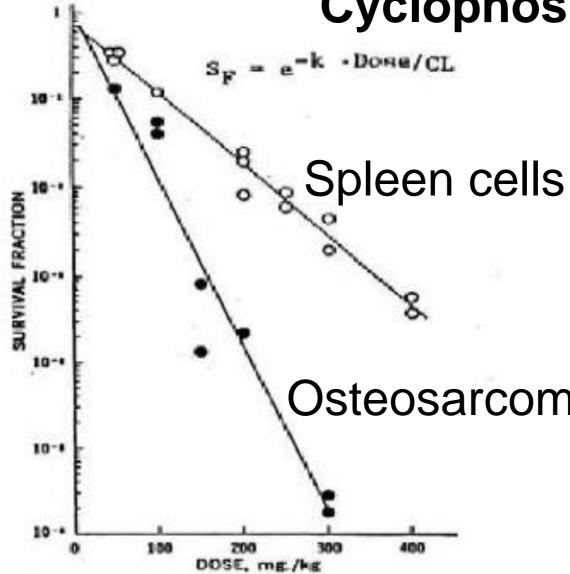
J. Pharm. Sci. <u>60</u>:892 (1971)

Drecker function

SIMPLE CELL KILLING

- Simple irreversible bimolecular . interactions produce log S_F/linear dose cell survival curves.
- The $\dot{R} = -k \cdot C \cdot R$ function predicts AUC to be the major PK determinant of total response.





Pharmacodynamics of Chemotherapeutic Effects: Dose-Time-Response Relationships for **Phase-Nonspecific Agents**

WILLIAM J. JUSKO

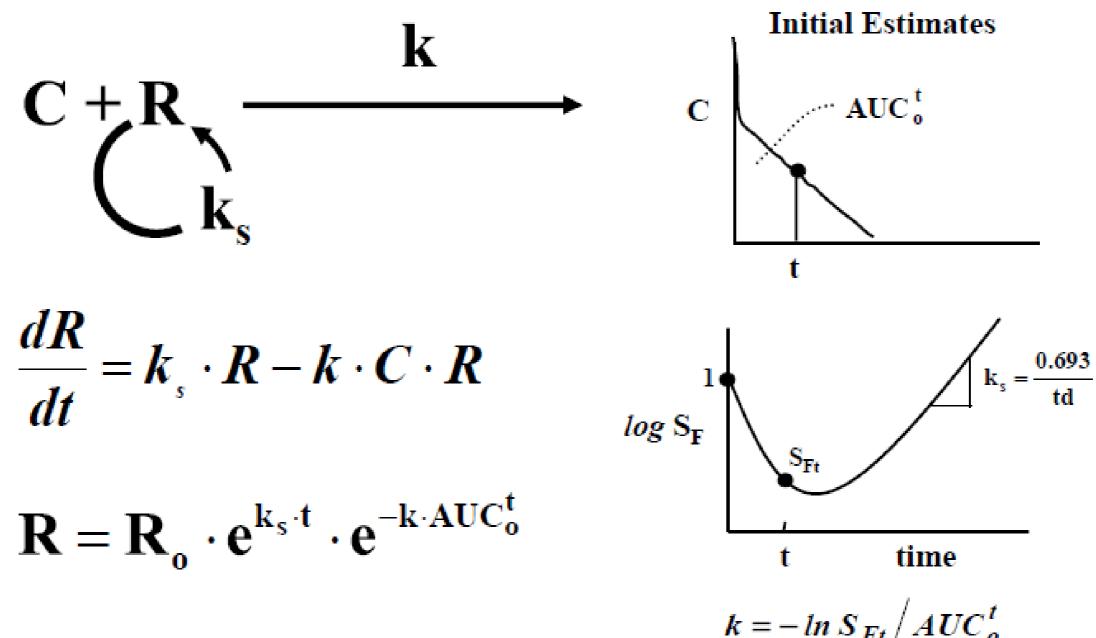


Cyclophosphamide - mice

Osteosarcoma cells

J. Pharm. Sci. (1971)

CELL KILLING – CELL GROW



Jusko, J. Pharm. Sci. 60:892 (1971)

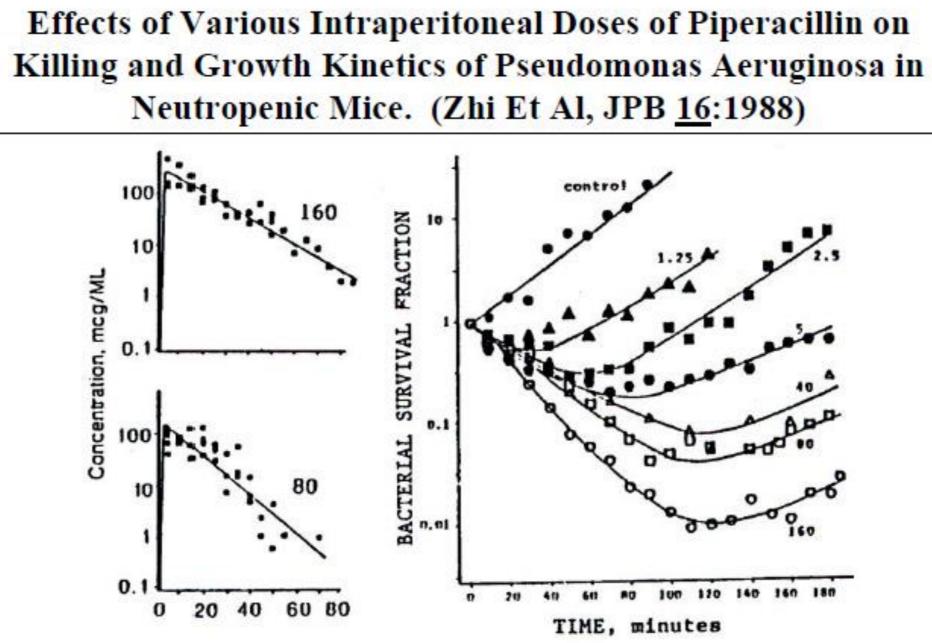




$$S_{Ft} / AUC_o^t$$

td = Doubling Time

<u>CELL KILLING – CELL GROWTH</u>

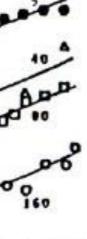


The joint effect of bimolecular drug/cell interaction and the ability of cell to grow produces biphasic survival curves with an initial phase of cell killing and a later phase of regrowth when all of the chemotherapeutic agent has been eliminated.









ANTIBACTERIAL EFFECTS





PKPD: PAST, PRESENT AND FUTURE ...

Classical PKPD: Compound selection

- Understanding time-concentration-effect relationship
- *Focus on dose predictions and therapeutic index
- Optimising design and interpretation of in vivo studies
- Empirical data driven

Mechanism-based PKPD: Target Validation

- Understanding target pharmacology
- *Focus on setting project targets
- Enabler for biomarker selection and translational strategy
- Data integration

Systems Pharmacology: Target Selection

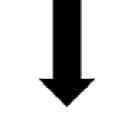
- Understanding pathway
- *Focus on target identification and selection
- Disease level, cross-target
- Knowledge/hypothesis driven



Pharm Res (2011) 28:1460-1464



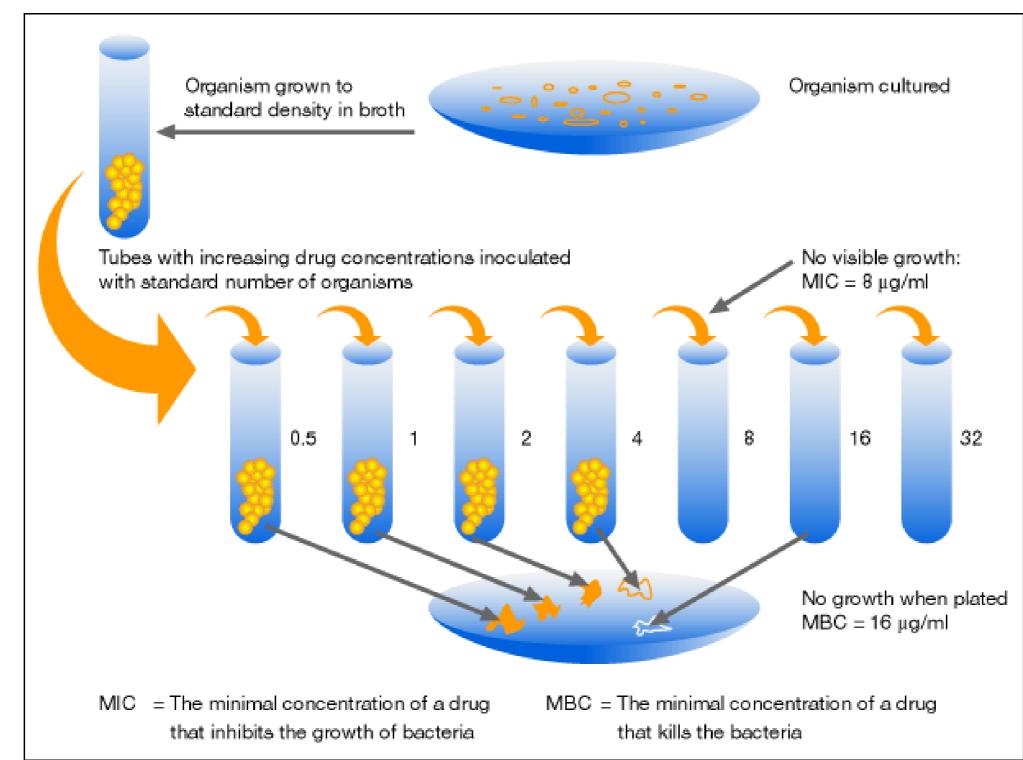
Biomarkers



Biomeasures



Determination of MIC (here: broth ditution test)







MIC: Minimum Inhibitory Concentration

No cell growth:

$$\frac{dR}{dt} = k_g \cdot R - k \cdot C \cdot R = 0$$

$$k_g = k_s$$

$$k_g = k \cdot C = k \cdot MIC$$

$$MIC = k_g/k$$

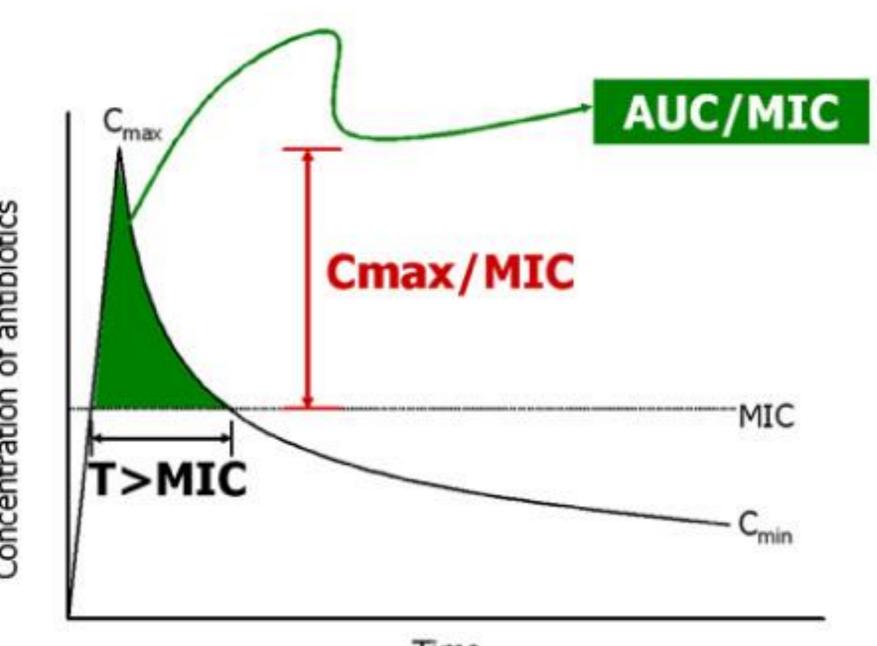
Cautions:

MIC is a dependent variable fraught with experimental uncertainties.





SUMMARY PK MEASURES



Time







35

ANTIBACTERIAL EFFECTS

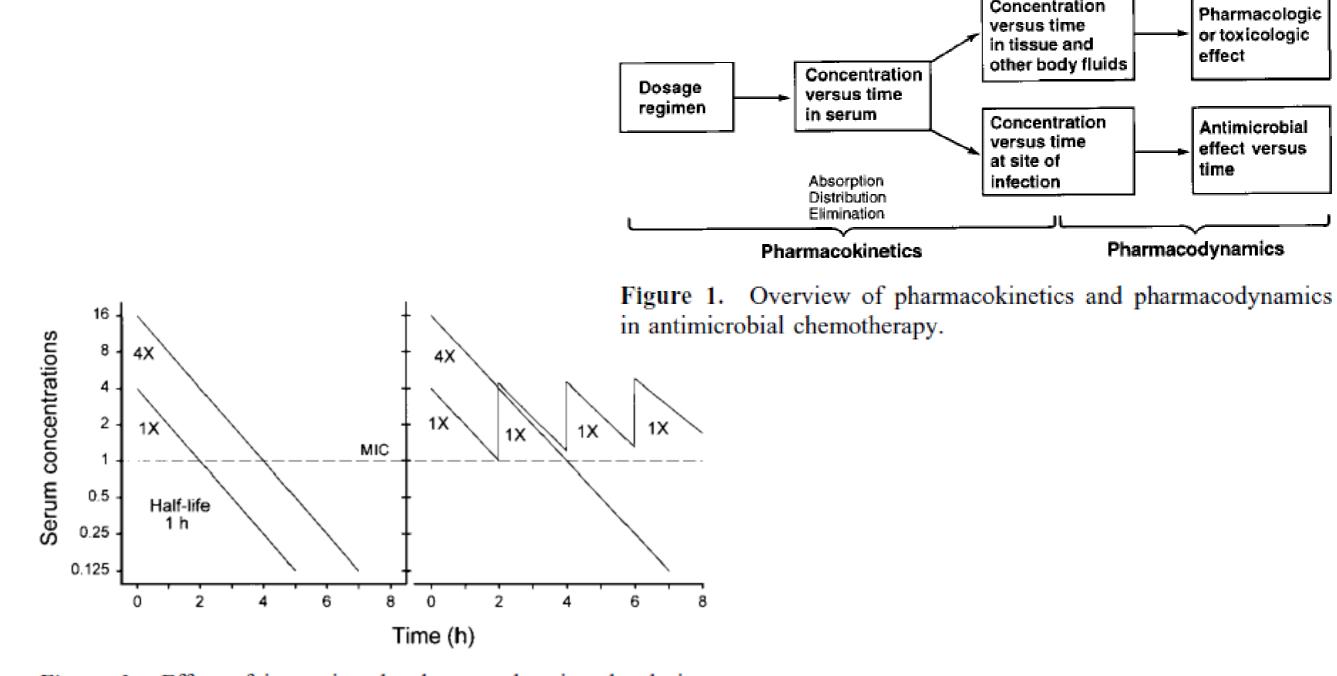
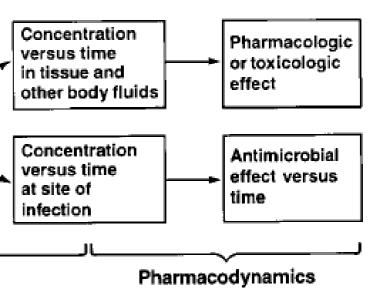




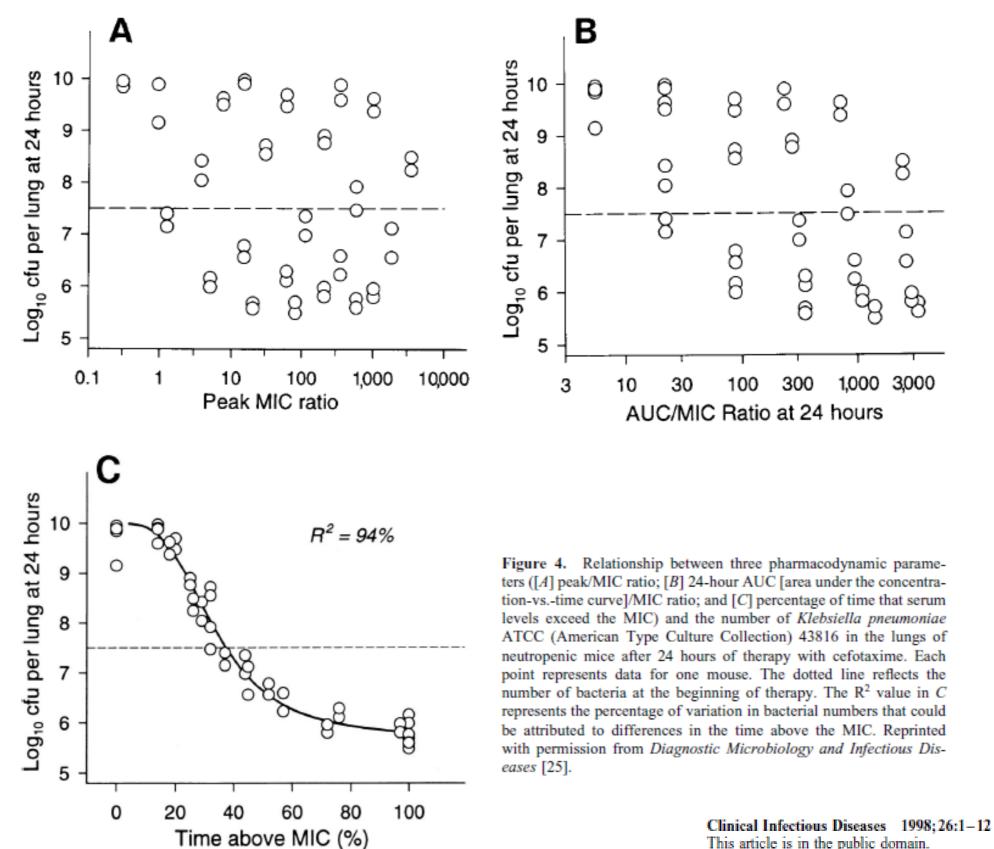
Figure 3. Effect of increasing the dose or changing the dosing regimen of a hypothetical drug on peak/MIC ratio, AUC (area under the concentration-vs.-time curve)/MIC ratio, and duration of time that serum levels exceed the MIC. Reprinted with permission from Diagnostic Microbiology and Infectious Diseases [25].

Clinical Infectious Diseases 1998;26:1–12 This article is in the public domain.



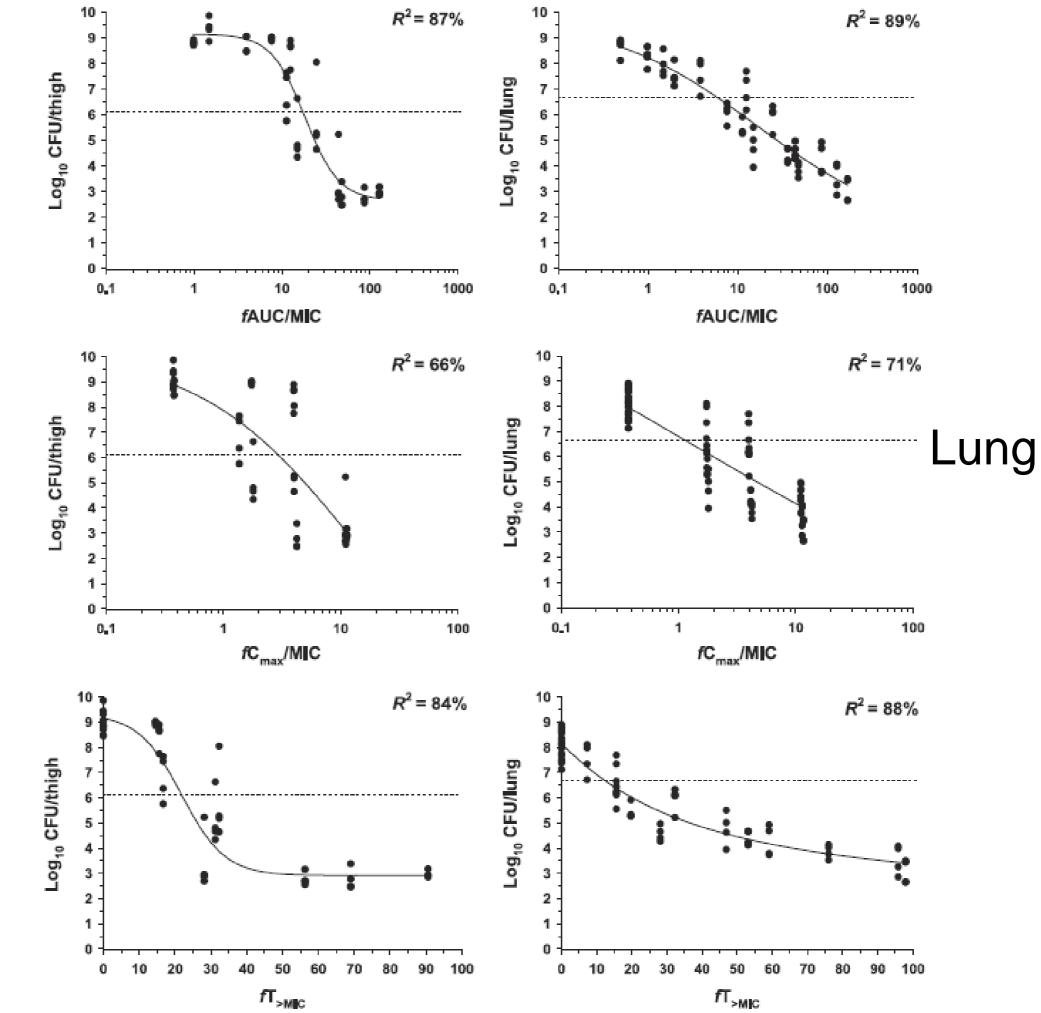


ANTIBACTERIAL EFFECTS













MIC DISTRIBUTION

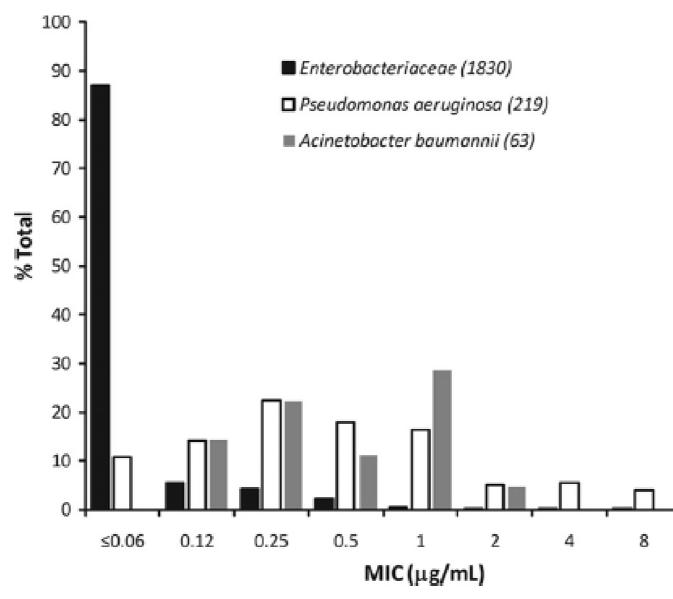
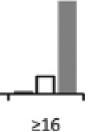


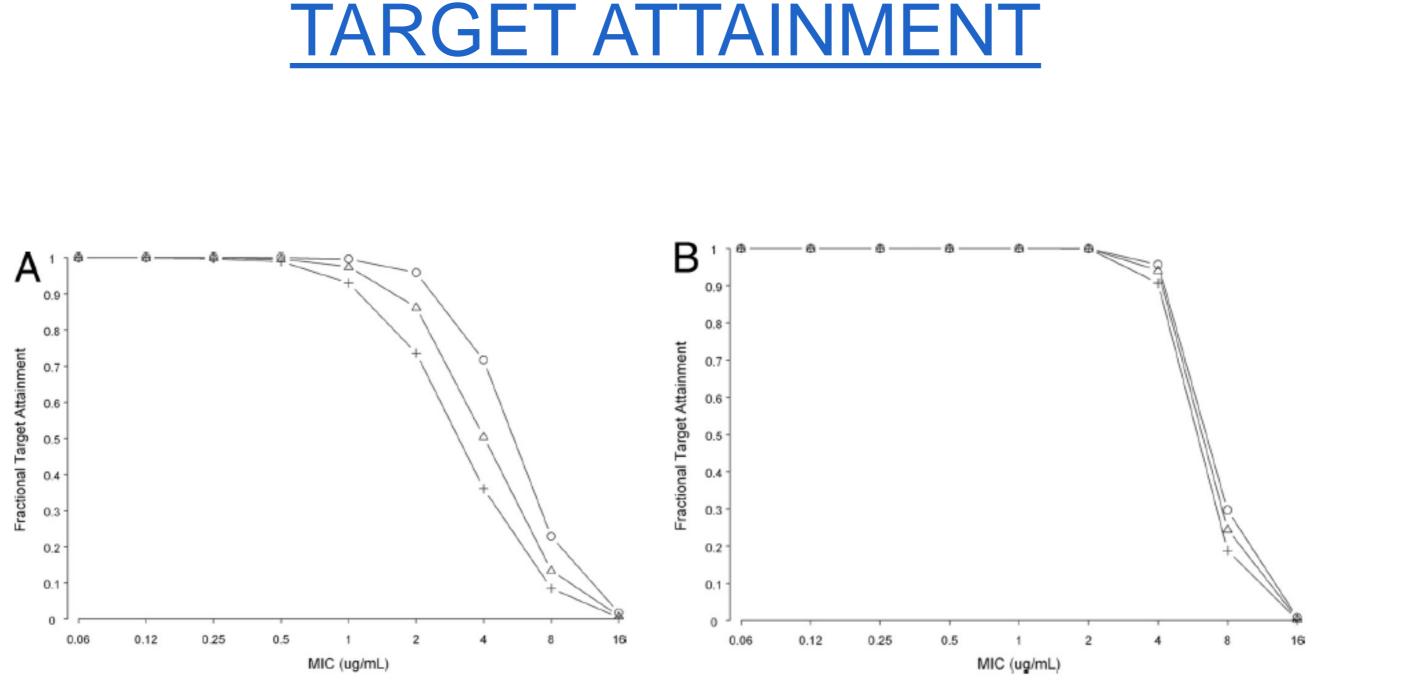
FIG. 1. Doripenem MIC distribution for selected Gram-negative pathogens from phase 3 clinical studies.



ANTIMICROBIAL AGENTS AND CHEMOTHERAPY, June 2010, p. 2360-2364







sion.



ANTIMICROBIAL AGENTS AND CHEMOTHERAPY, June 2010, p. 2360-2364

FIG. 2. Target attainment results for doripenem at 500 mg every 8 h infused over 1 h and 4 h over a wide range of creatinine clearances observed in phase 1, 2, and 3 studies (O, 25% T>MIC; \triangle , 30% T>MIC; +, 35% T>MIC). (A) One-hour infusion; (B) 4-hour infu-

TARGET ATTAINMENT

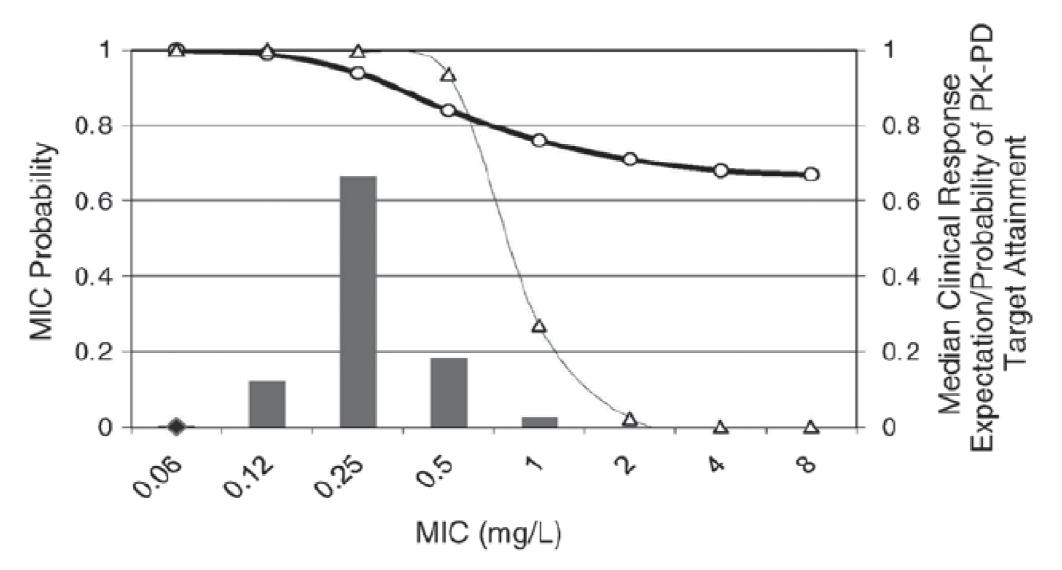


Fig. 8.1 Probability of target attainment (PTA, open triangles) based on AUC_{ss,24h}/MIC ratio, clinical response expectation (open circles), and tigecycline MIC distribution (bars), showing a trend of decreasing PTA and median clinical response expectation in increasing MIC. (Image from Ambrose et al. 2009; used with permission)





CLINICAL AND MICROBIOLOGICAL CURE

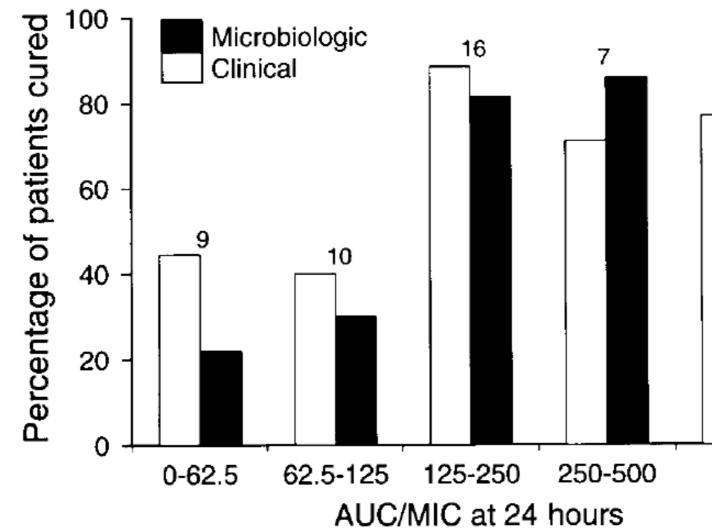


Figure 8. Relationship between the 24-hour AUC (area under the concentration-vs.-time curve)/MIC ratio and the microbiological and clinical efficacy of ciprofloxacin in 64 patients with serious bacterial infections. The 24-hour AUC/MIC is the sum of the AUCs for all doses administered every 24 hours divided by the MIC. Data are from [52].

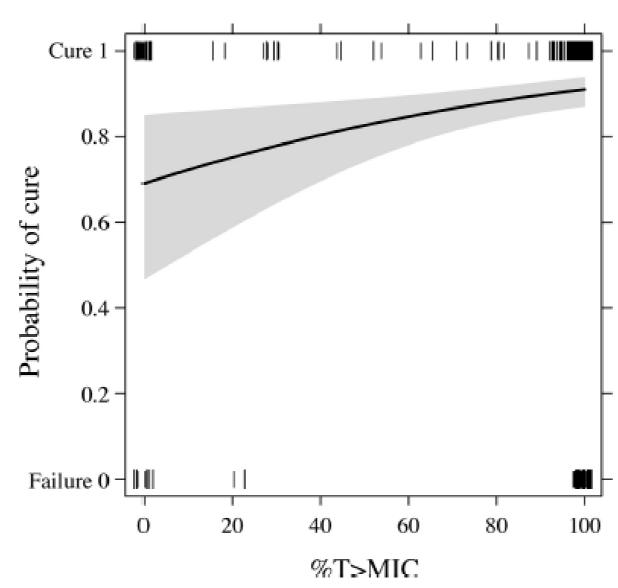
Clinical Infectious Diseases 1998;26:1-12 This article is in the public domain.

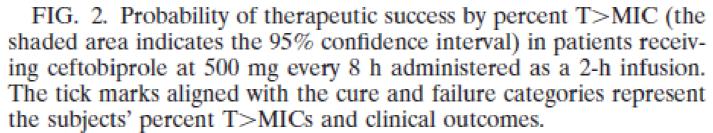


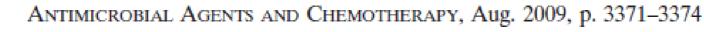


>500

CLINICAL CURE



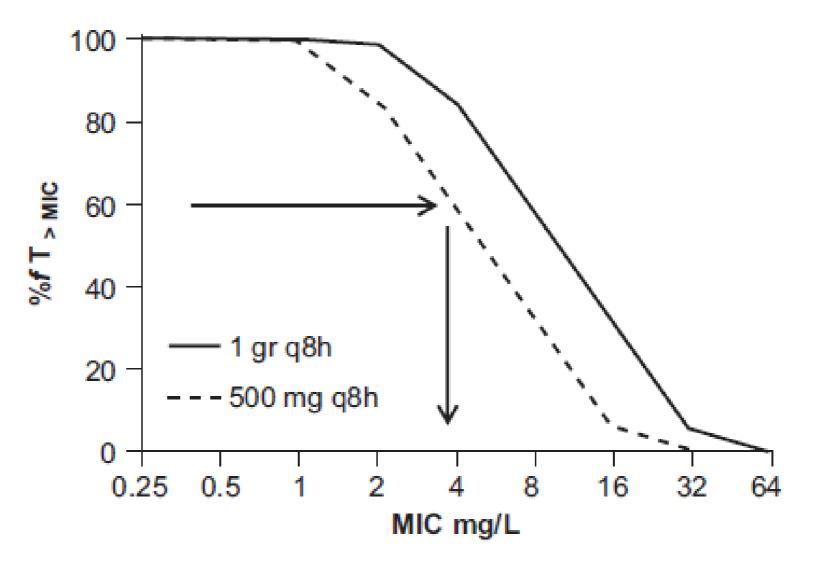






CLINICAL BREAKPOINT PKPD BREAKPOINT

Fig. 8.2 The percent of time that the free ceftazidime concentration is above MIC (% fT > MIC) for two dosing regimens of ceftazidime (1 g q8h vs. 500 mg q8h) against MIC to illustrate that clinical breakpoint is dependent on the dosing regimen. Arrows indicate that the pharmacodynamics target corresponding to 60% fT > MIC is 4 and 8 mg/L for 500 mg q8h and 1 g q8h, respectively. (Image from Mouton et al. 2012; used with permission)





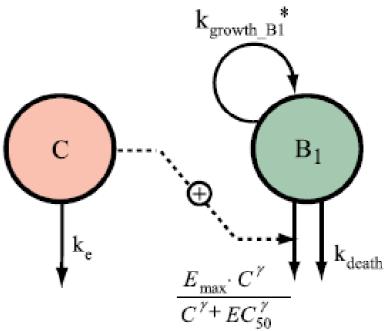


MODELING TIME COURSE OF NTIBIOTIC EFFECT

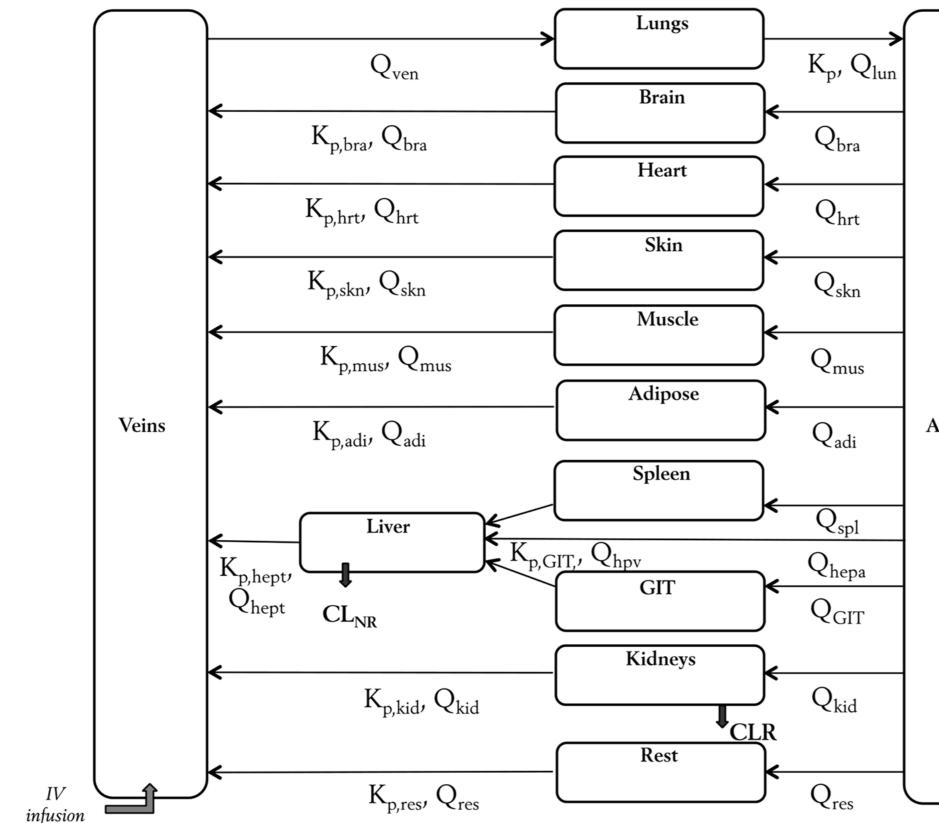
- Better to model time course of bacterial growth & killing
- **Differentiate between the** system (bacteria) and drug effects
- **Can address resistance development**
 - ... and the effects of the host's immune system



=> Use all available data: *in vitro*, animal, clinical



SYSTEMS PHARMACOLOGY: PBPK



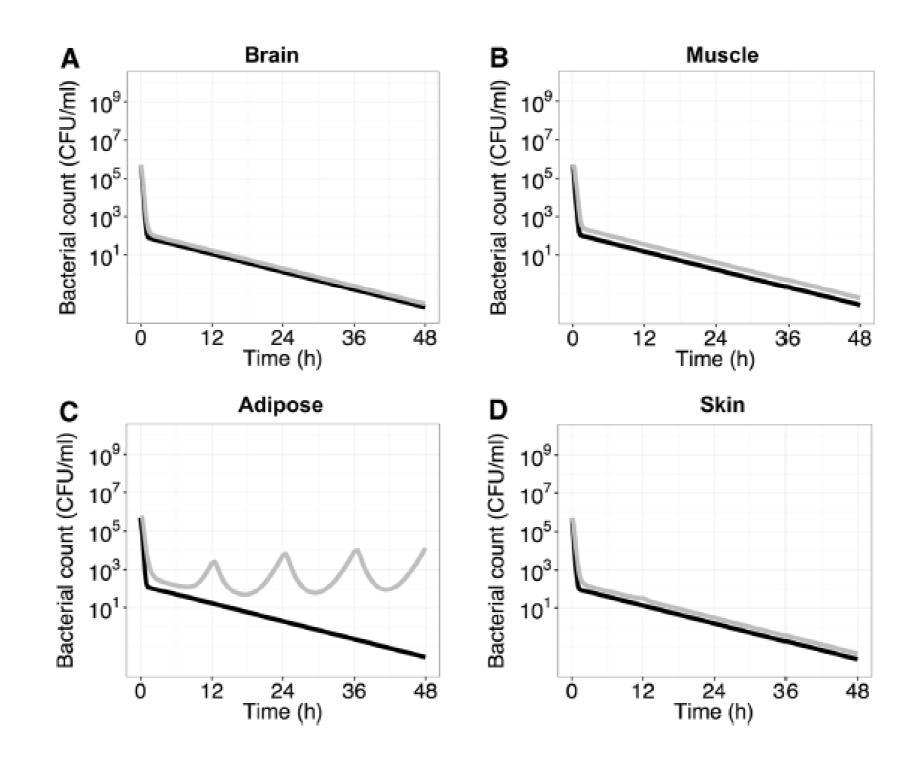




Arteries	
Arteries	

... COMBINED WITH PD MODEL AT INFECTION SITE

Fig. 5 Predictions of the time course of bacterial killing of E. coli strains LM347 (black) and LM625 (grey) in the extracellular compartment of different tissues (brain, muscle, adipose and skin) following a ciprofloxacin dose of 400 mg b.i.d



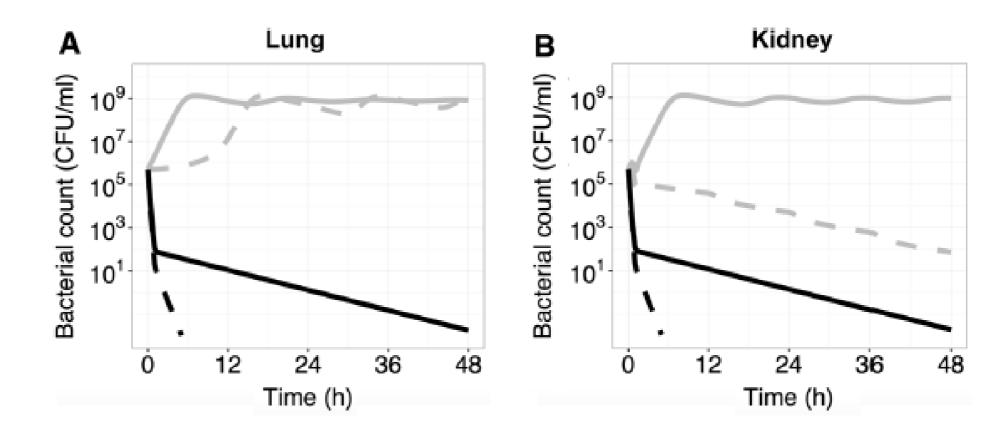


... ALLOWS TO DERIVE THE TISSUES WITH HIGHEST EFFECT

Fig. 6 Predictions of the time course of bacterial killing of E. coli strains LM347 (black) and LM707 (grey) in lung and kidney following administration of ciprofloxacin 400 mg b.i.d. with (dashed lines) and without (solid lines) addition of function for immune response

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$$\frac{dB}{dt} = k_{growth} \times B - k_{death} \times B - \left(\frac{E_{\max} \times C_{(t)}^{\gamma}}{EC_{50}^{\gamma} + C_{(t)}^{\gamma}}\right) \times B$$

 $dB/dt = \dots - Kkill_ANC \cdot (ANC/(ANC + ANC50))$ $(1 - B/(B + B50)) \cdot B.$



J Pharmacokinet Pharmacodyn (2017) 44:69-79

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