

# Integration of physiological and biochemical concepts into the development of biopharmaceuticals

Philip Lowe PhD

Modelling & Simulation, Novartis Pharma AG  
Basel, Switzerland

EU Bioanalysis Forum, Open Symposium – Less is More  
18<sup>th</sup> November, 2011, Barcelona

# Abstract

- The goal of this talk is to give those who are based in a more laboratory based environment some examples of how the results of ligand binding assays are interpreted and used within the context of preclinical and clinical drug development. The majority of the talk will focus on pharmacokinetic-pharmacodynamic model based analyses, as there are serious interpretation issues if classical non-compartmental analyses are blindly applied to the nonlinear behaviours common with biopharmaceuticals
- By incorporating aspects of mechanism into PKPD models, one gains a deeper understanding of the basis of drug disposition and pharmacology, greater biological realism and thereby better extrapolatability, or prediction, of patient responses. For biopharmaceuticals especially, mechanistic aspects of physiology and biochemistry can be linked by explicitly describing drug distribution and elimination together with drug target production (expression) and turnover. Drug-target binding is modelled enabling target capture to be quantitated and correlated with clinical responses. Examples will include:
  - Omalizumab capturing IgE thereby alleviating the signs and symptoms of allergic asthma
  - Non-clinical to clinical translation of a monoclonal antibody with highly nonlinear pharmacokinetics characteristic of target mediated drug disposition (TMDD)
  - Canakinumab capturing interleukin-1 $\beta$  thereby alleviating the pain and inflammation experienced by patients with cryopyrin associated periodic syndromes
- The overall objective of a PKPD modelling and simulation process could be realised, to deliver physiologically and biochemically reasonable predictions of the effectiveness of the drugs. Through simulation from the models, suitable posologies – doses and regimens – for specific groups of patients were suggested

# Biochemistry of binding

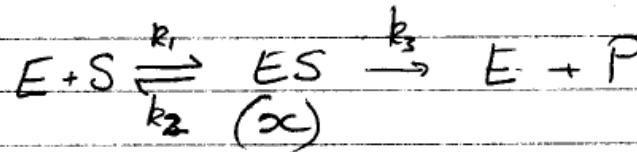
Drug and target related by an *equilibrium* reaction



Add drug, mass balance “pushes” reaction to the right

Enzyme kinetics

Michaelis-Menton



assumptions

①  $e \ll S$  usually at least 100×

②  $v = k_3 [ES]$   $v =$  overall rate

③  $k_2 \gg k_3$  Michaelis-Menton assumption  
so that  $E + S \rightleftharpoons ES$  is in eqm.

# “Test-tube” scenario Drug-target binding is a function of target loading or expression

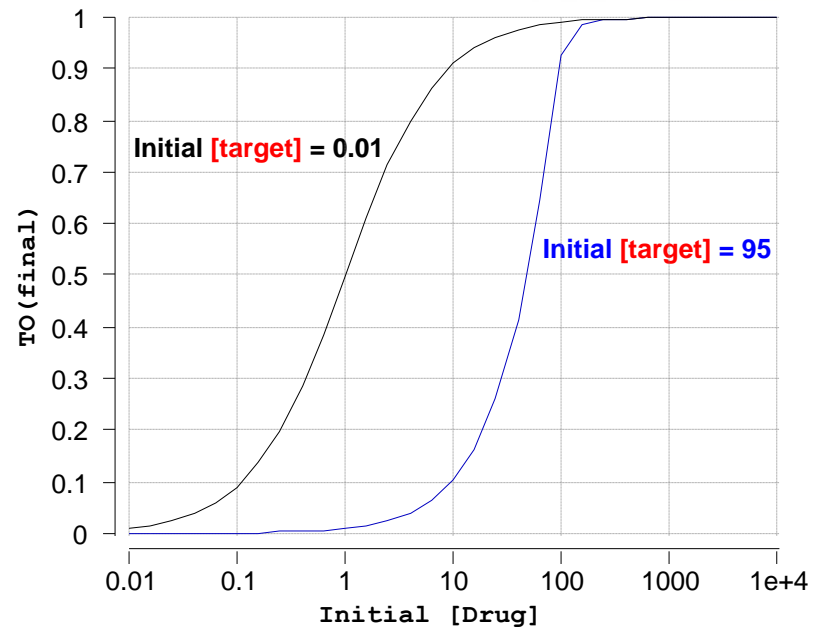
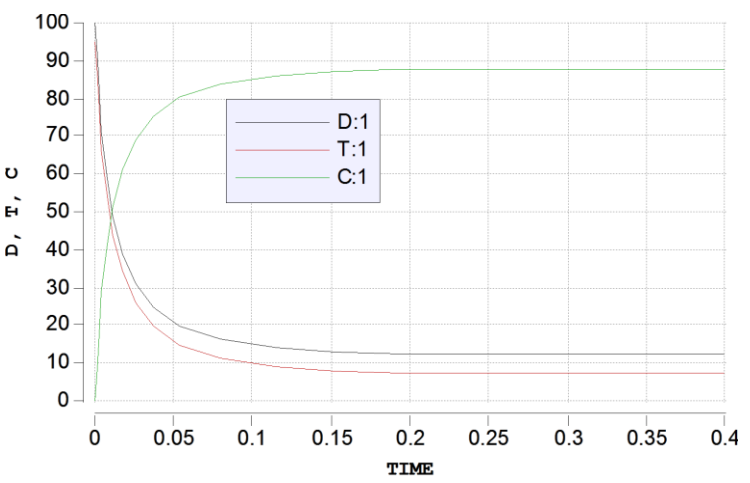


$$\begin{aligned}
 d/dt(D) &= -k_{on} * D * T + k_{off} * C && ; \text{Drug} \\
 d/dt(T) &= -k_{on} * D * T + k_{off} * C && ; \text{Target} \\
 d/dt(C) &= +k_{on} * D * T - k_{off} * C && ; \text{Complex}
 \end{aligned}$$

$$TO = C / (C + T) \quad ; \text{Target occupancy}$$

```

init(D) = 100      ;initial [drug]
init(T) = 95      ;initial [target]
init(C) = 0       ;initial [complex]
  
```

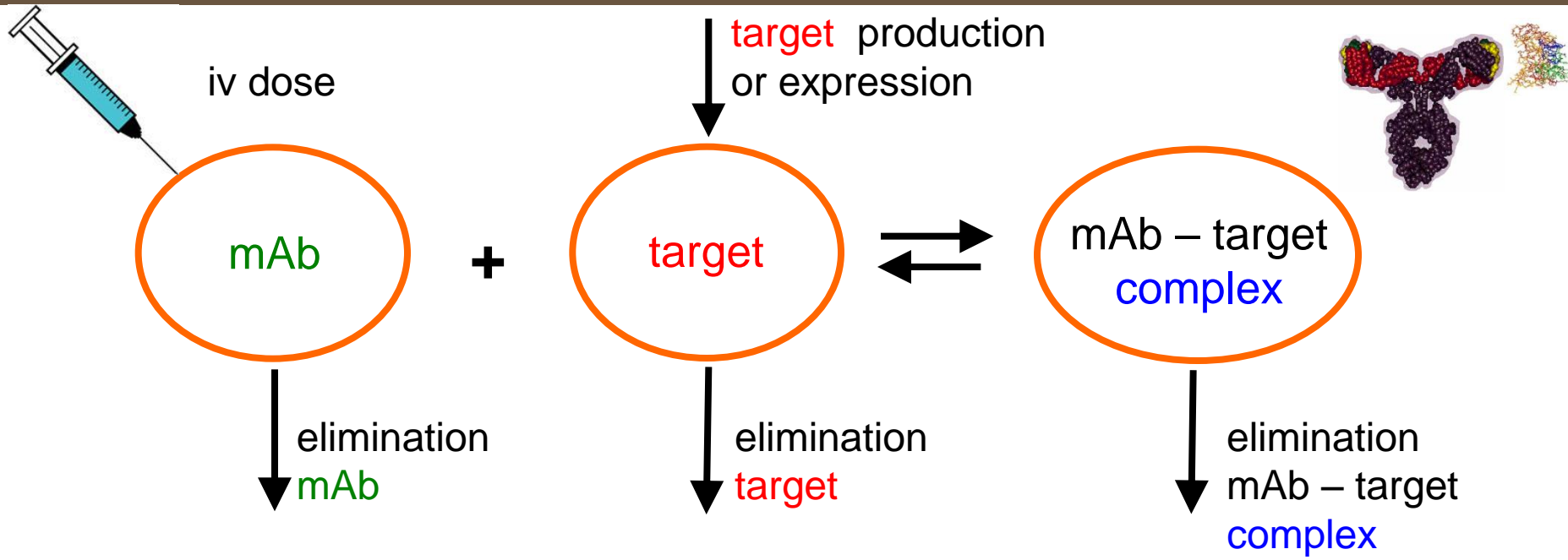


$K_d$  is equal to  $EC_{50, \text{added drug}}$  only when **[target]** << **[drug]**

Shift depends on amount of target in system, i.e. expression or load level

# Target binding incorporated into a PKPD model

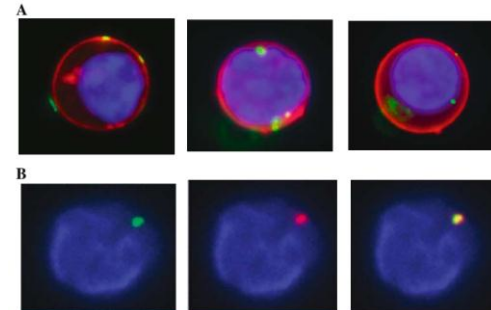
*Drug input/output and target production and turnover added to binding reaction*



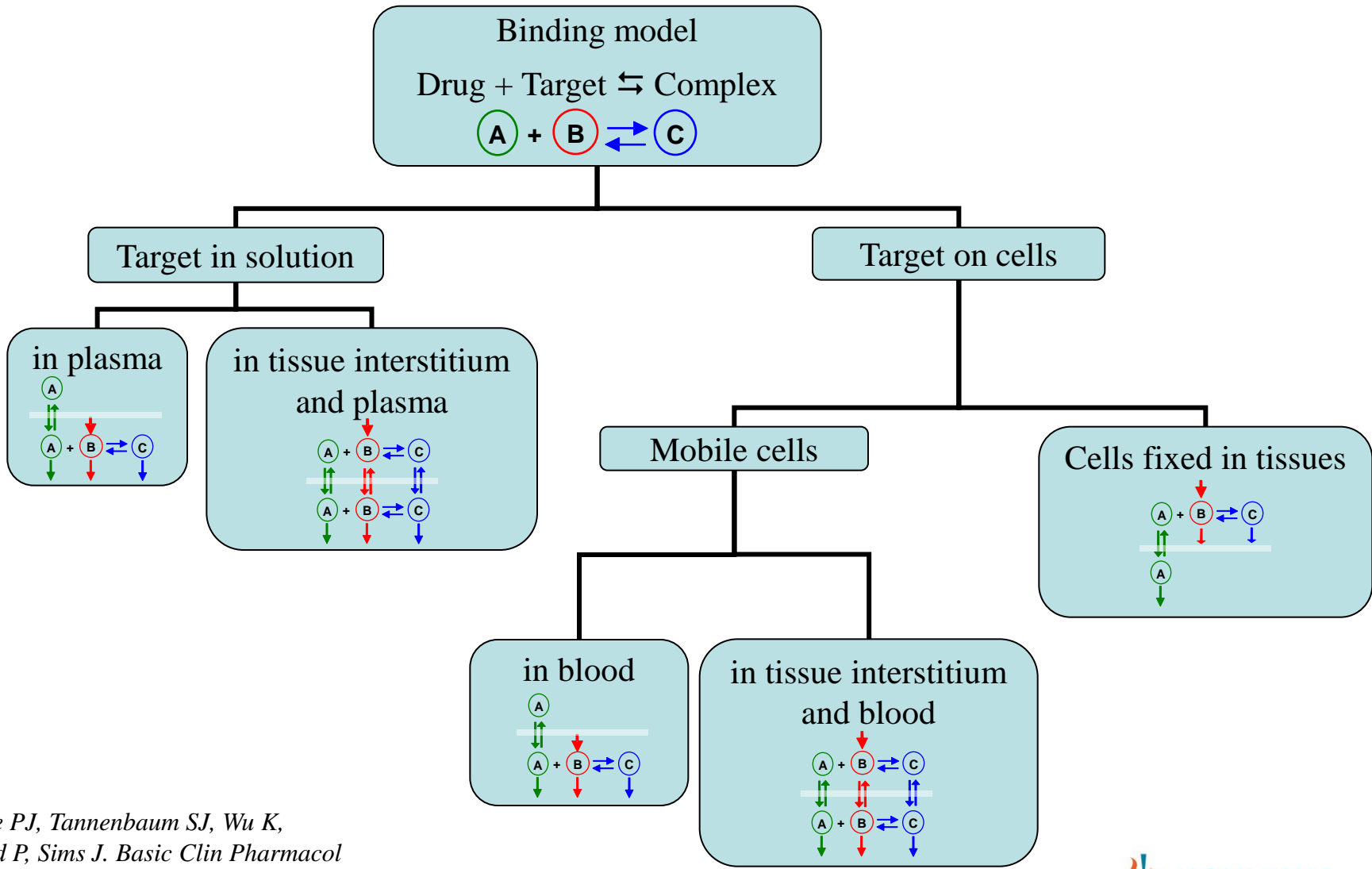
slow clearance  
 $V \sim 3 \text{ L}$  central, for 70 kg  
 $t_{1/2} \sim 3\text{-}4 \text{ w}$

**general rule:**  
 faster than a mAb for most (not all) proteins

**general rule:**  
 soluble target – slow  
 cell surface target – faster



# Taxonomy of drug-target binding physiological localisations



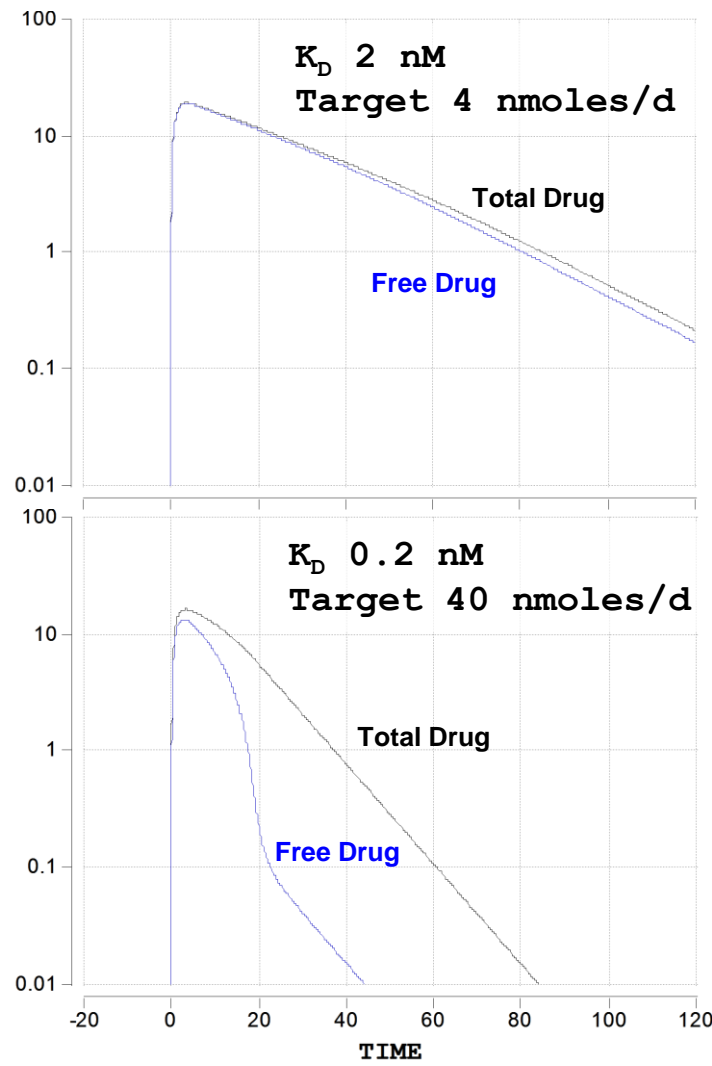
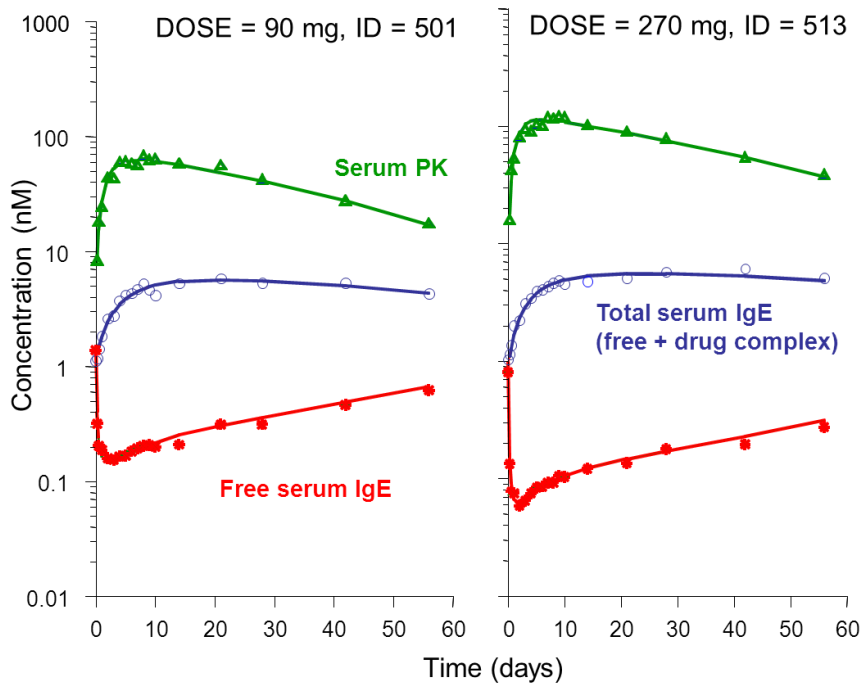
*Lowe PJ, Tannenbaum SJ, Wu K, Lloyd P, Sims J. Basic Clin Pharmacol Toxicol 2009; 106:195-209*

# Components of the PKPD model

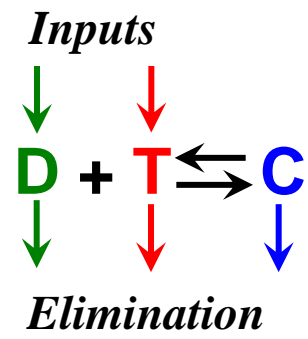
- Pharmacokinetics of the monoclonal antibody:
  - species differences often well understood and easily characterised
  - good prediction to man, either by scaling and/or by reference to prior similar antibodies
- Binding affinity to the target ligand:
  - species differences understood during *in vitro* characterisation
- Localisation, expression and turnover of target, clearance of drug-target complex
  - species differences often not well characterised
- Potentially, competition with other binding proteins
  - Such that observed *in vivo* binding not same as isolated *in vitro* system

# Example 1: Target turnover and binding model

- i) Model compares well with omalizumab, total and free IgE data
- ii) What if assay measures free drug?      iii) what if expression &  $K_D$  different?



- Assays must be correctly specified, free or total
- If higher affinity binding and higher target expression, major differences between free and total



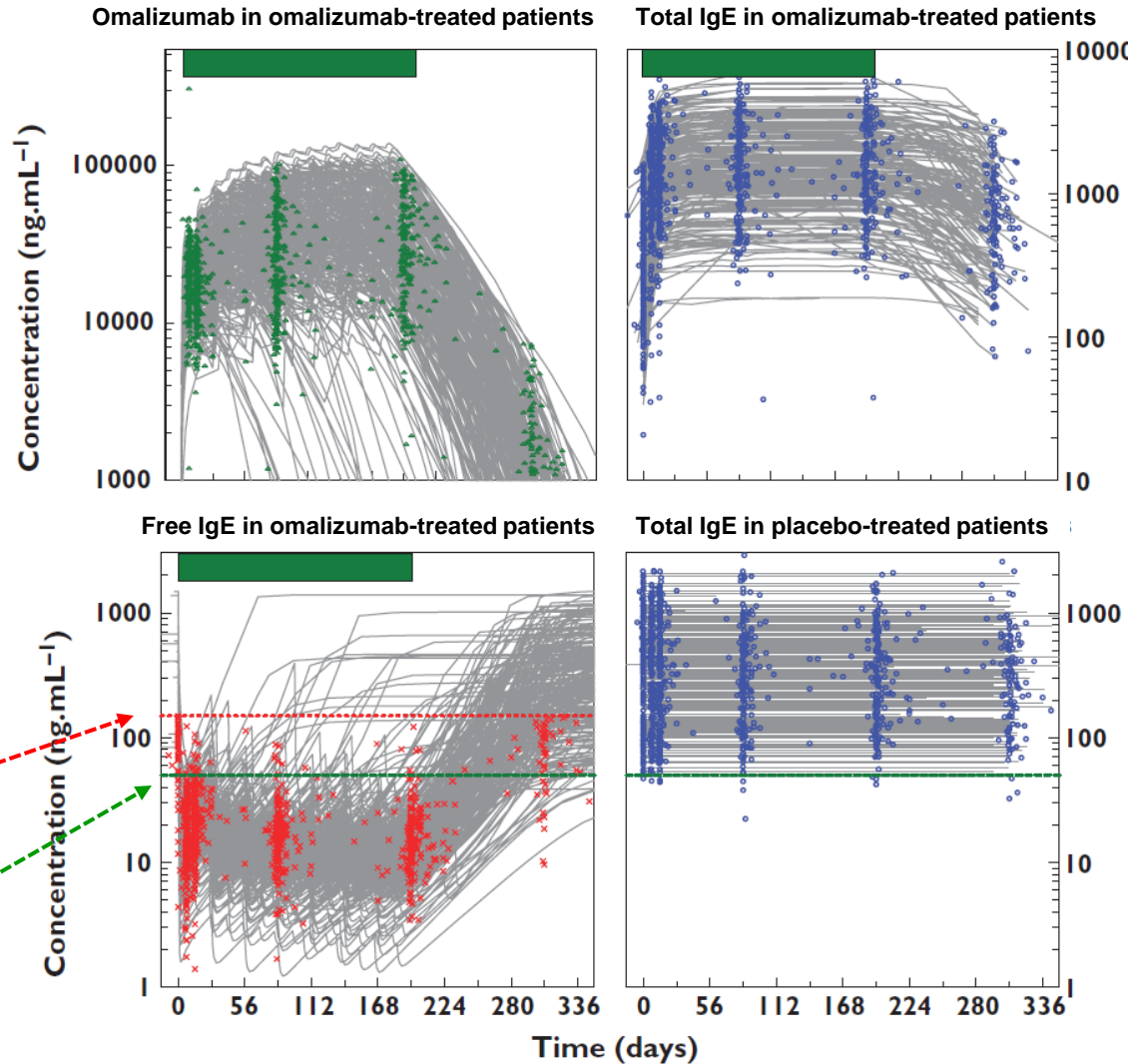


# Xolair-IgE model allowed reconstruction of free IgE versus time profiles for individual patients

- Model applied to data from the INNOVATE study patients
- Randomised, double-blind, placebo-controlled study over 28 weeks
- Omalizumab: n=226  
Placebo: n=214
- Model-simulated omalizumab regimens derived through dosing table to achieve at least  $0.016 \text{ mg.kg}^{-1}$  per  $\text{IU.mL}^{-1}$  IgE at baseline

Upper limit of quantification for free IgE

Target level of free IgE (50 ng/mL)



Total serum IgE (free + drug complex) ○

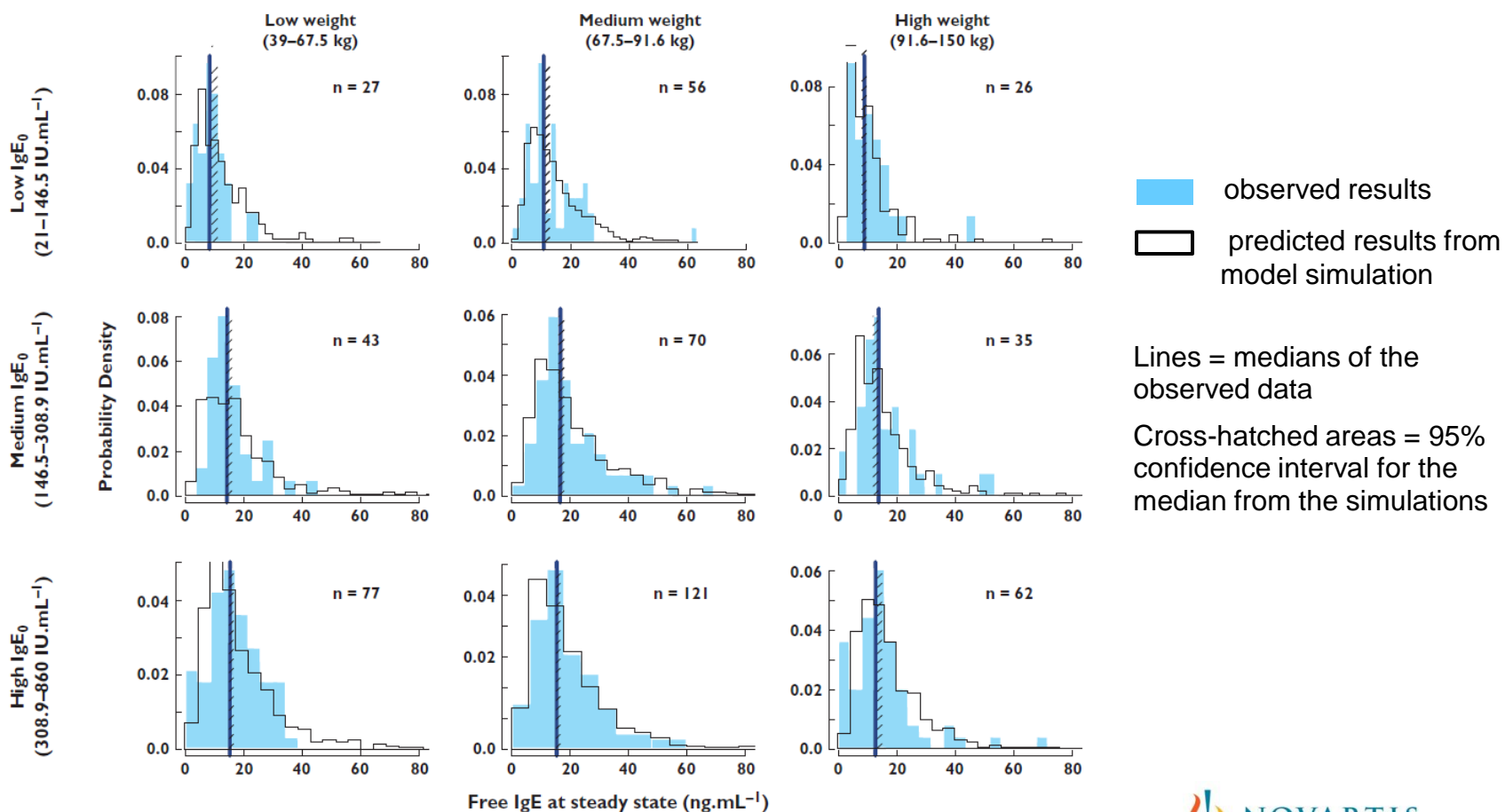
Serum PK Δ

Free serum IgE x

# Predictive check of the omalizumab–IgE model on to the distribution of free IgE at steady state for different regions of the dosing table

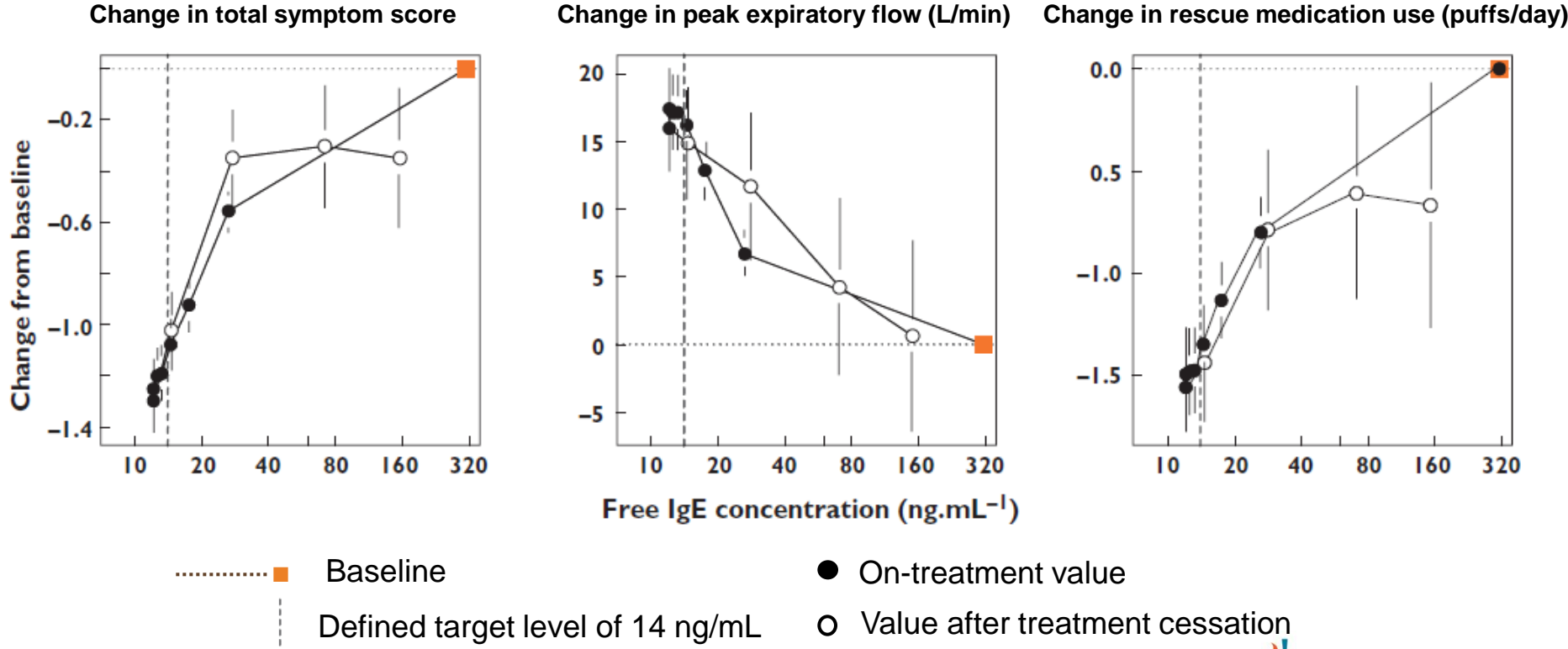
Comparison of:

- model simulation of a total of 19,330 virtual patients, split into nine subsets of the dosing table
- observed data from the Phase III studies for the numbers of patients indicated

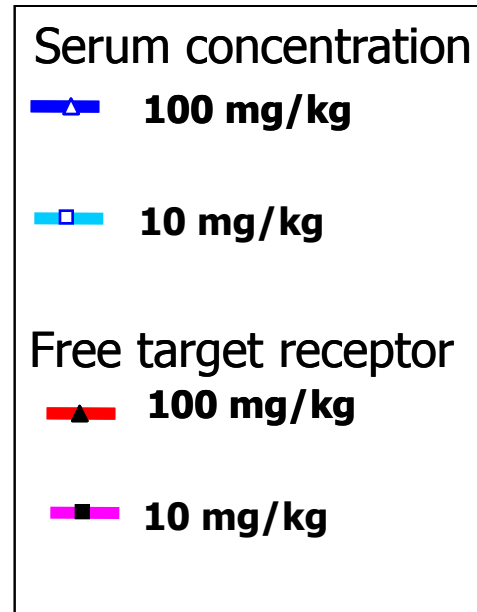
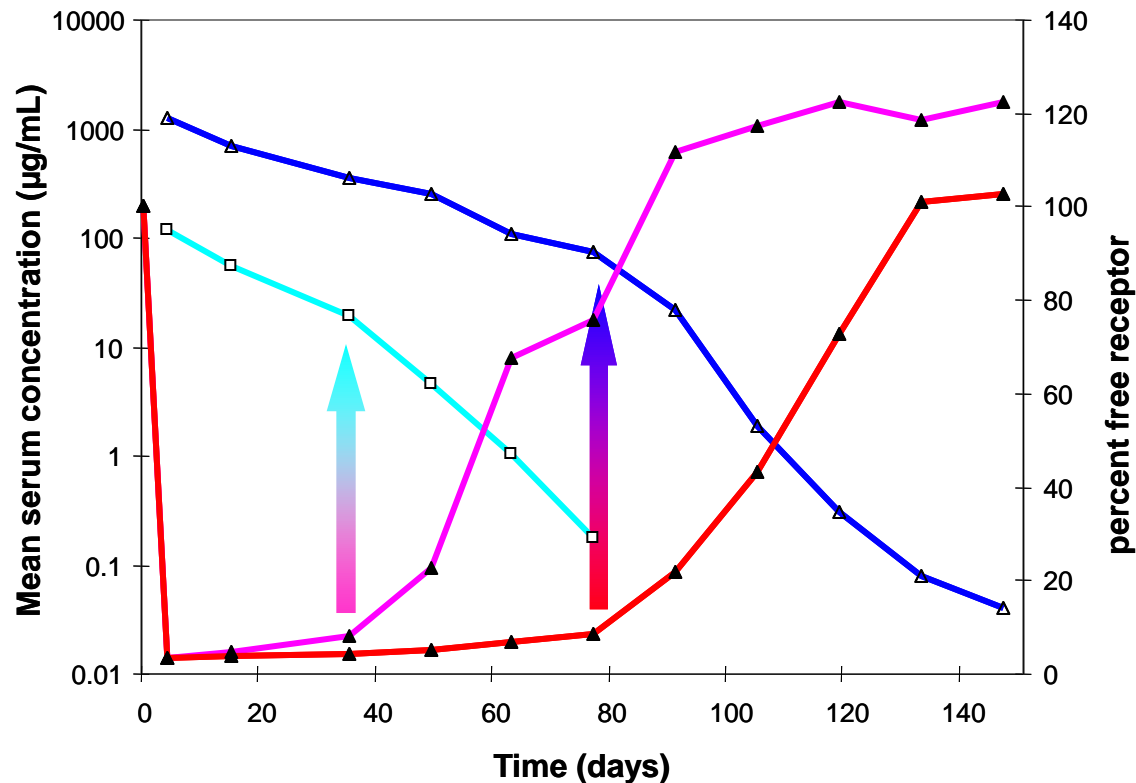


# Direct correlation between mean free IgE and change from baseline in signs and symptoms of asthma

- Combined data from four Phase III trials (n=898 omalizumab-treated patients)
- As free IgE suppressed by binding with omalizumab, mean total asthma symptom scores and rescue medication use decreased and morning peak expiratory flow increased



## Example 2: Nonclinical to clinical translation: An antibody against a cell surface target in cynomolgus monkey

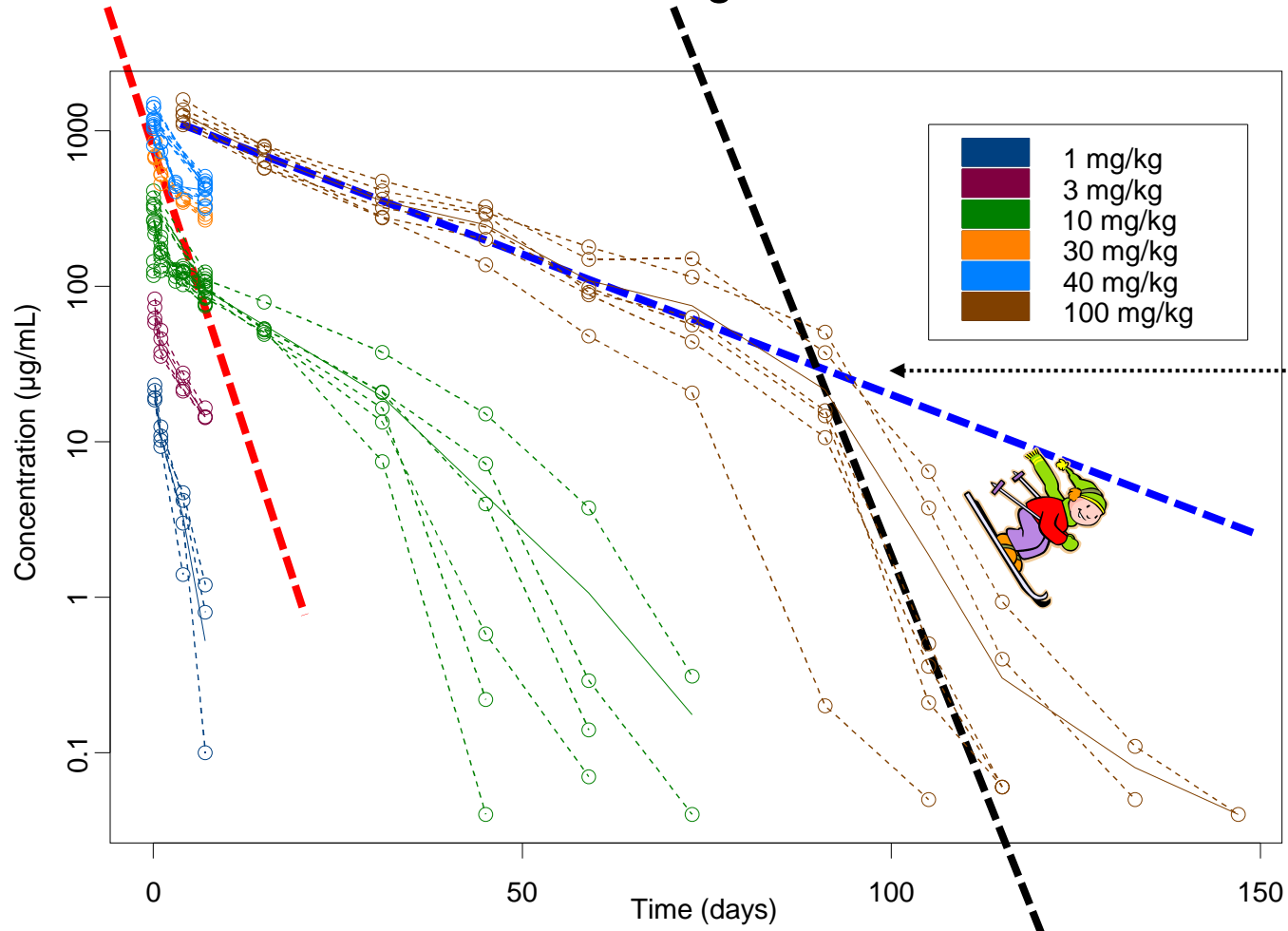


- Faster elimination when saturation less than 90%
- Independent saturation assay correlates with break points in PK

# Clear target mediated disposition with dose and time

## *Cynomolgus monkey pharmacokinetics of free monoclonal*

- Rapid elimination at lower doses and concentrations
- Slower elimination when target saturated



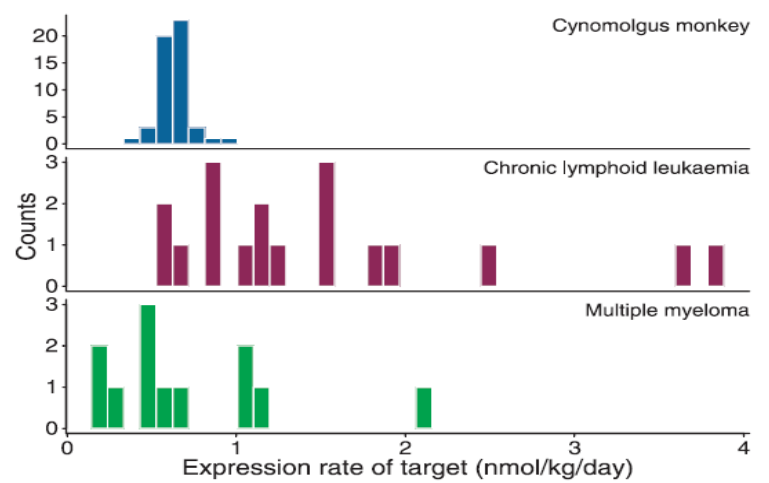
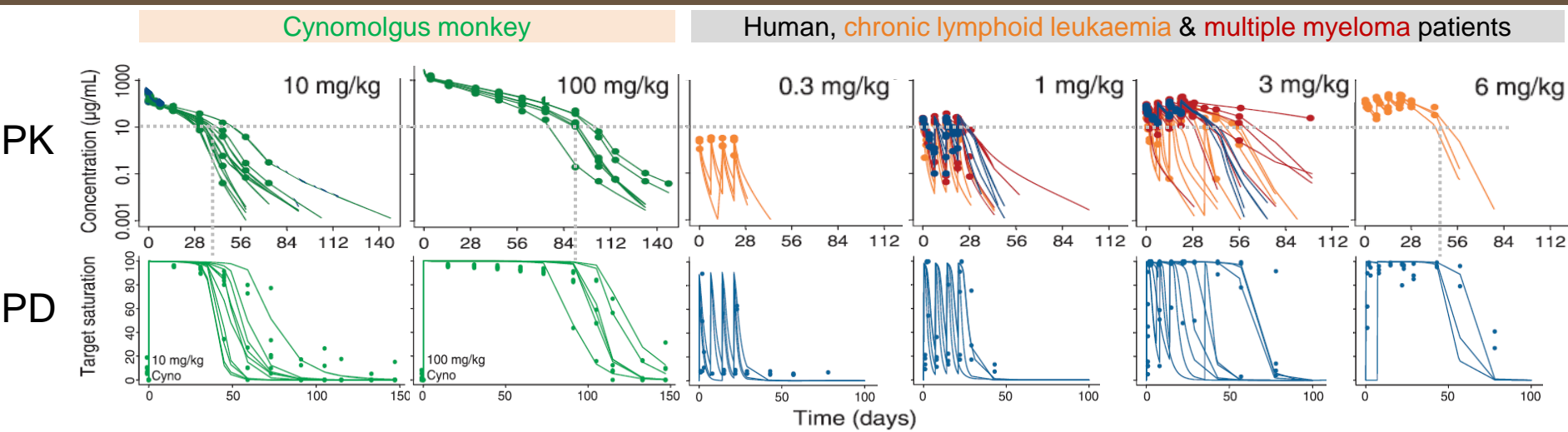
- Which half-life?
- Alpha half-life, distribution to tissues
  - Beta half-life, when target saturated
  - Gamma or TMDD half-life, when not saturated

Saturation level  
 “Over the cliff”  
*Varies from subject to subject*

Same data as in Lowe PJ, et al. *Basic Clin Pharmacol Toxicol* 2009; 106: 195-209



14 **Monoclonal antibody binds and occupies cell surface target**  
*Nonlinear PK correlates with target saturation; need to assess target expression and replenishment; level of receptor or target can vary between species and between diseases*



- Target expression consistent between cynomolgus monkeys
- Humans vary by more than order of magnitude
- More target → higher dose to achieve saturation
- Model used to advise on dose and regimen for next cancer indication

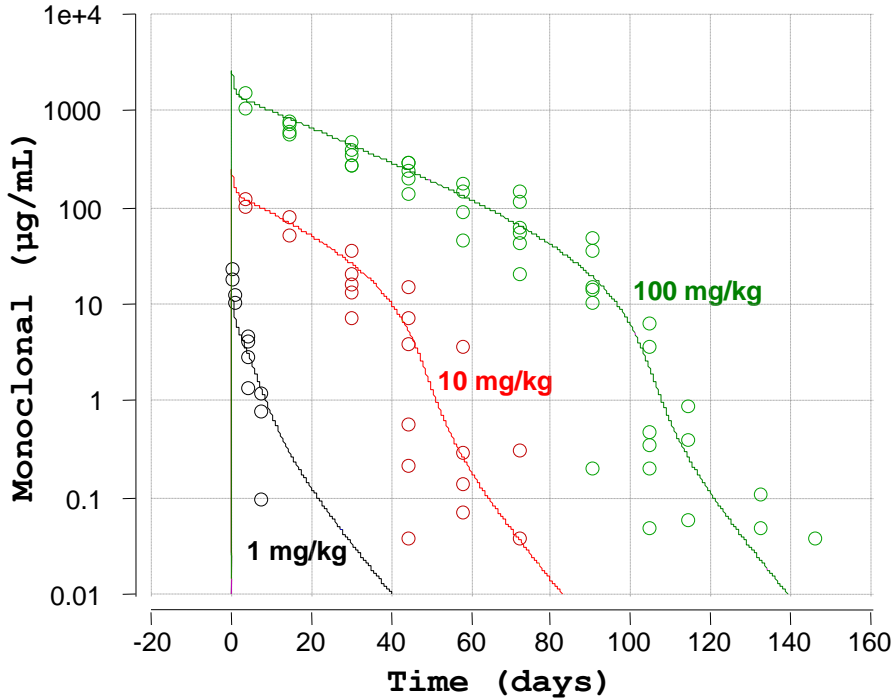
Fig. 6. Distribution of individual cynomolgus monkey and patient values for the turnover or expression rate of the target.

# Target binding model enables direct display of occupancy

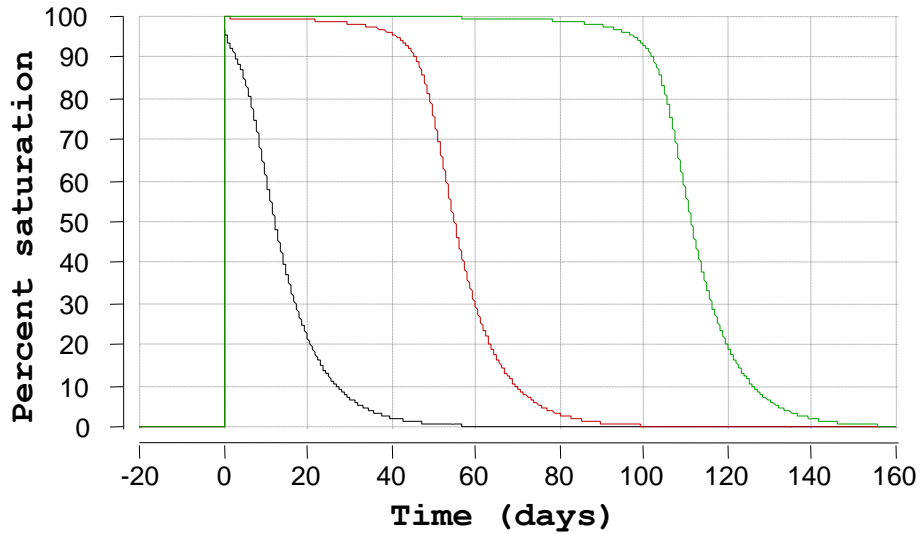
Calculated from quantities of mAb-target complex & total target

To scale adjust parameters to selected species using in vitro & in vivo data

## Pharmacokinetics (free drug)



## Saturation, or target occupancy



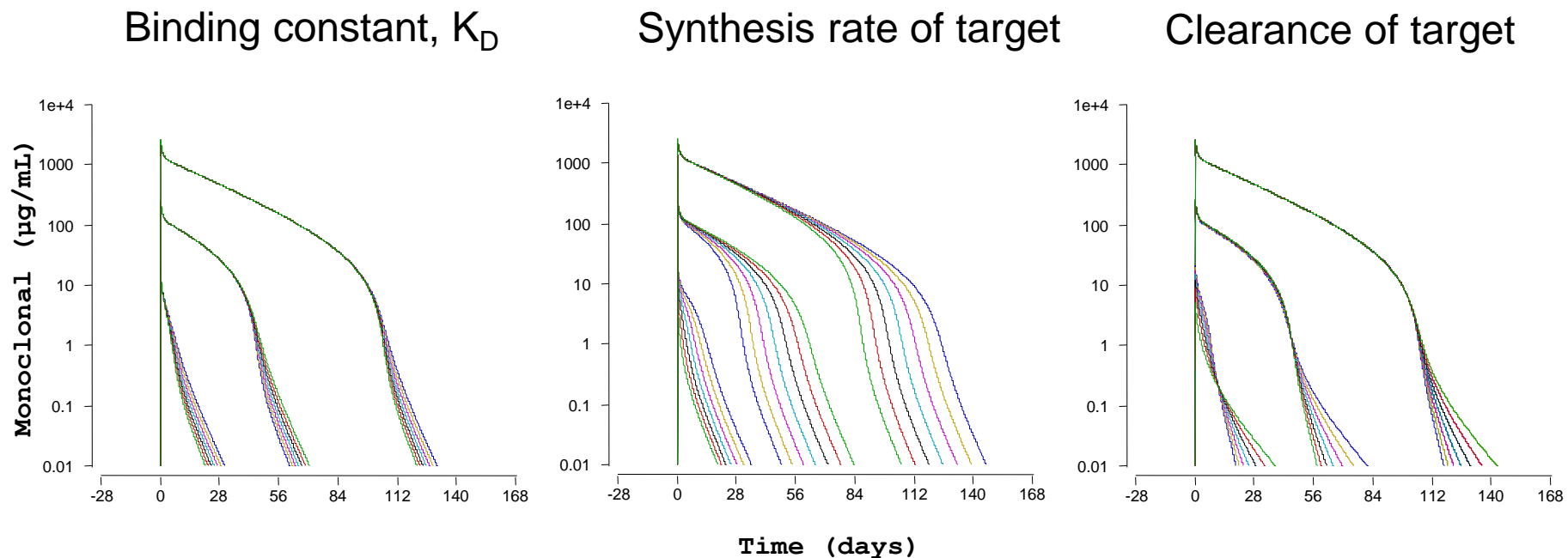
- Clearances (drug D, target T, complex C)
- Volumes (central, peripheral)
- Rate of target production or expression
- Binding constant (or  $K_M$ ,  $IC_{50}$ ,  $EC_{50}$ )

$$\begin{aligned}
 D/DT(TD) &= -FD \cdot CLD/V - CPLX \cdot CLC/V - Q \cdot (FD/V + DP/VP) \\
 D/DT(TT) &= RINB - FT \cdot CLT/V - CPLX \cdot CLC/V \\
 D/DT(DP) &= Q \cdot (FD/V - DP/VP) \\
 CPLX &= ((KD \cdot V + TD + TT) - ((KD \cdot V + TD + TT) ** 2 - 4 \cdot TD \cdot TT) ** 0.5) / 2 \\
 FD &= TD - CPLX \quad ; \text{FREE D} = \text{TOTAL D} - \text{COMPLEX} \\
 FT &= TT - CPLX \quad ; \text{FREE T} = \text{TOTAL T} - \text{COMPLEX} \\
 SAT &= 100 \cdot CPLX / TT
 \end{aligned}$$



# 16 Sensitivity to target turnover and binding

*Knowledge of synthesis rate of target (i.e. expression level) most important factor for predicting dose required for duration of target saturation*



- 4 fold range from highest to lowest in each batch of simulations
- 1, 10 and 100 mg/kg doses
- All other parameters as in prior slide



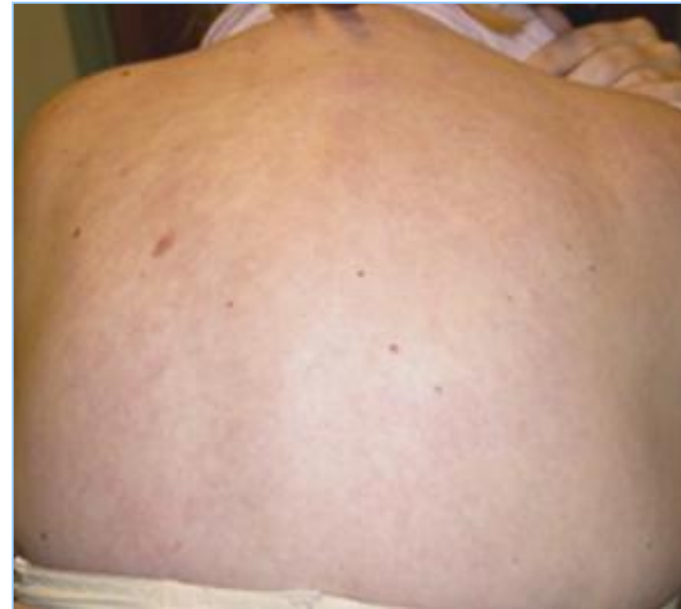
## Example 3: Ilaris™, canakinumab, ACZ885, a fully human monoclonal anti-interleukin-1 $\beta$ antibody for inflammatory disorders

- **Interleukin-1- $\beta$** , a major player in inflammation, can cause
  - *Urticaria, malaise, joint pain, conjunctivitis, fatigue, frequent migraines, cold sensitivity, deafness*
- **Canakinumab** binds **IL-1 $\beta$**  to form inactive complexes: **D+T $\rightleftharpoons$ C**
  - So driving these cryopyrin associated periodic syndrome symptoms into remission

Baseline



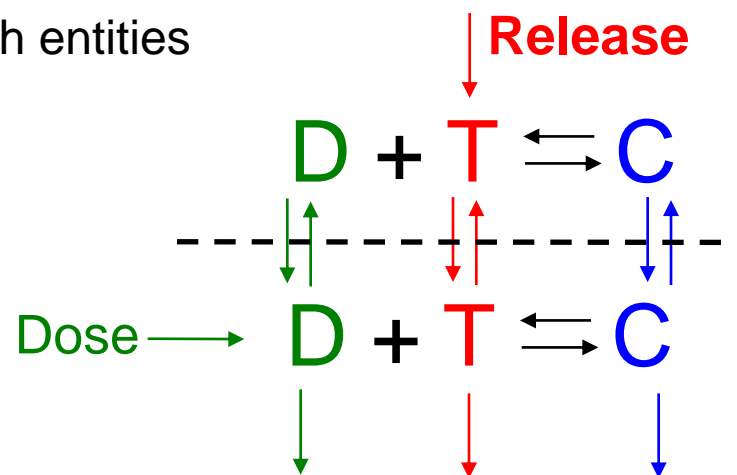
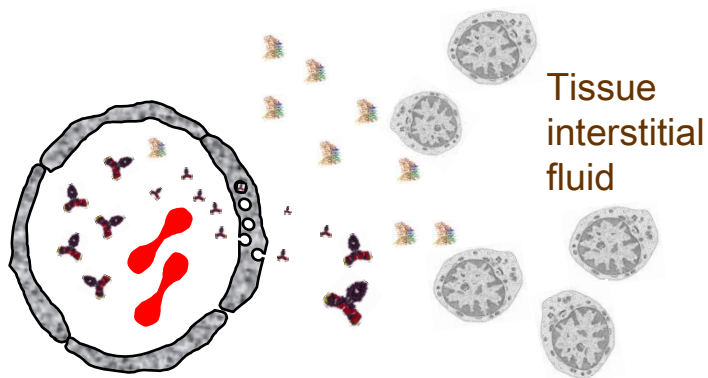
24 hours



# Binding model starts to link to an underlying multi-compartmental distributional physiology for the target



- IL-1 $\beta$  released into tissue interstitial fluids at sites of inflammation
- Canakinumab, as a protein, distributes between plasma and tissue interstitial fluid (lymph)
- Vascular endothelial permeation of both entities



Control subjects

(n=20)

Rheumatoid Arthritis patients

(n=20)

| Protein          | Lymph       | Serum     | Lymph      | Serum      |
|------------------|-------------|-----------|------------|------------|
| IgG (mg/dL)      | 238 ± 32    | 890 ± 53  | 384 ± 45   | 1708 ± 104 |
| Cytokine (pg/mL) |             |           |            |            |
| IL-1 $\beta$     | 1.50 ± 0.25 | 3.4 ± 1.0 | 14.8 ± 3.9 | 0.8 ± 0.3  |
| TNF $\alpha$     | 4.4 ± 1.1   | 2.0 ± 0.1 | 9.9 ± 1.1  | 4.7 ± 0.64 |

From

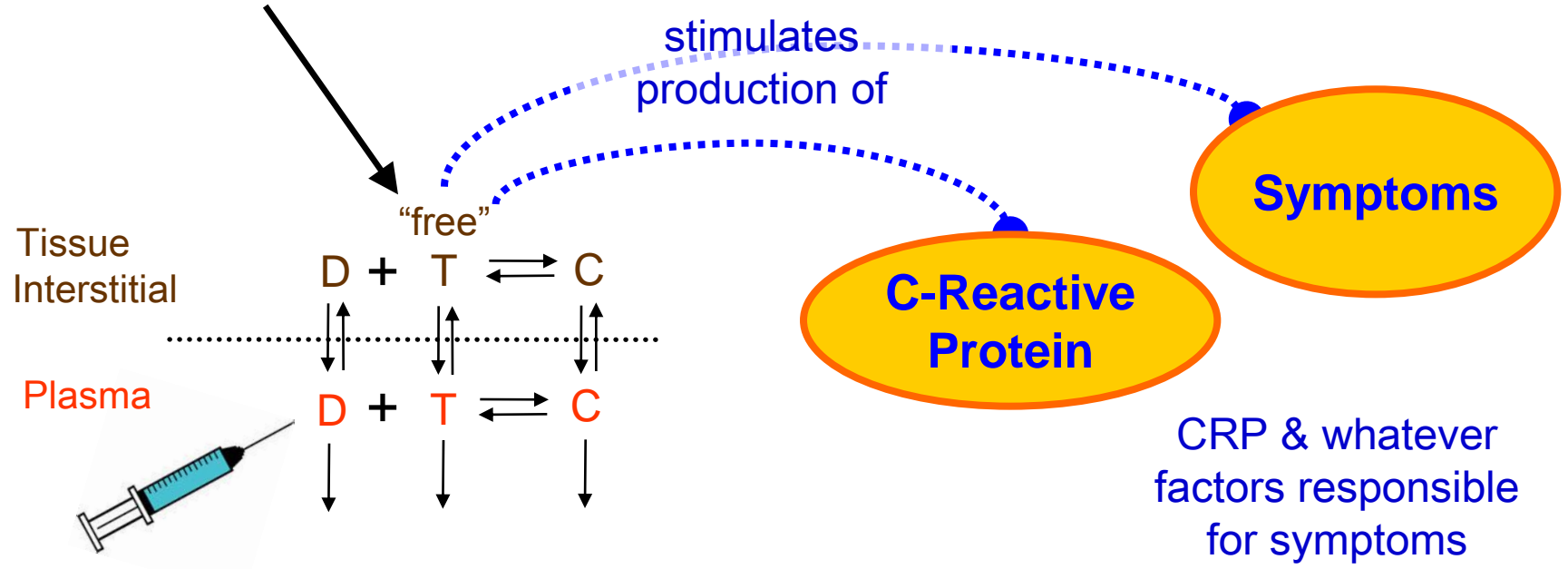
Olszewski et al (2001)

Arthritis &amp; Rheumatism

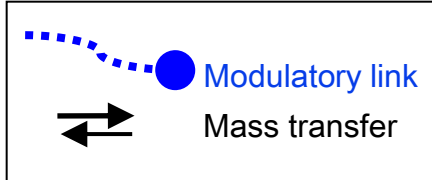
44(3) 541-549

# Free IL-1β linked to clinical measurements of inflammation and symptoms of disease

Cytokine released in tissues



- All captured in a drug and disease model
- Parameters estimated from patients
- Simulations for prediction of dose and regimen
- Then test...



# Causal chain is modelled allowing prediction beyond data to simulate and propose posology for canakinumab in patients

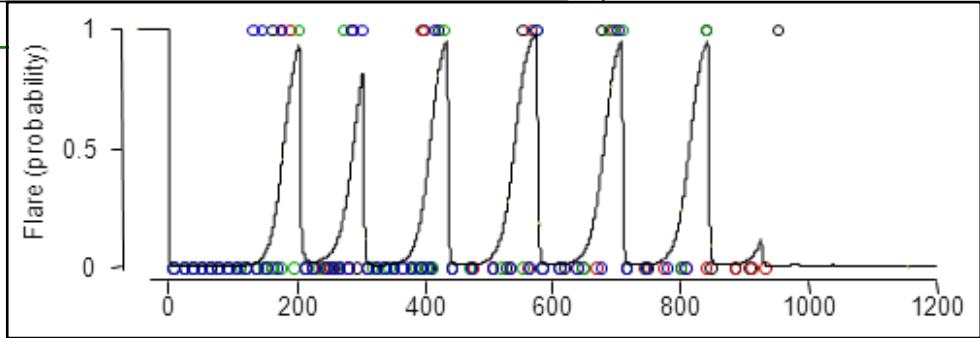
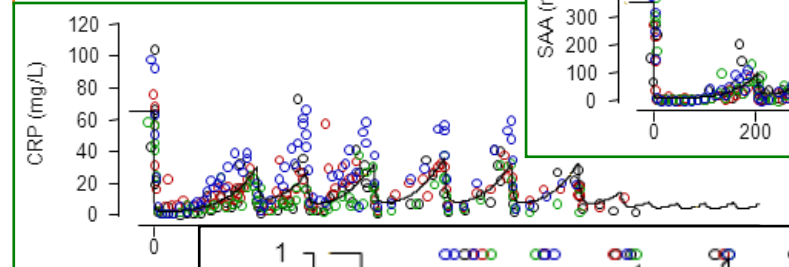
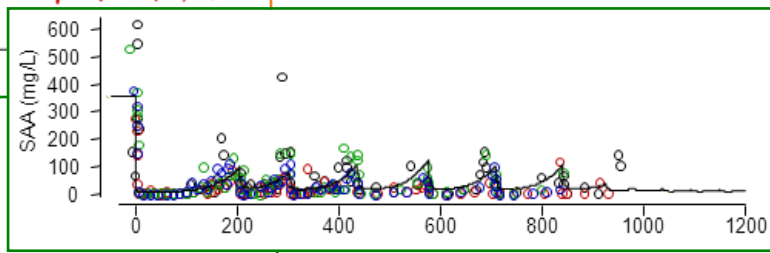
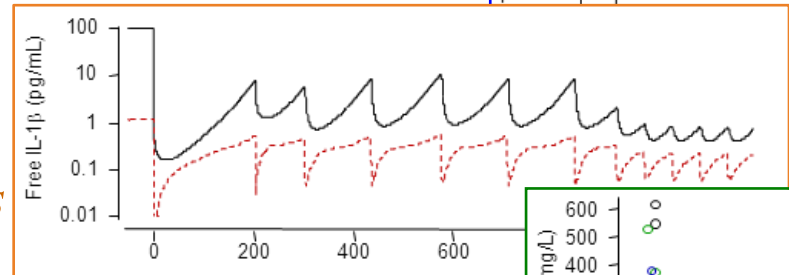
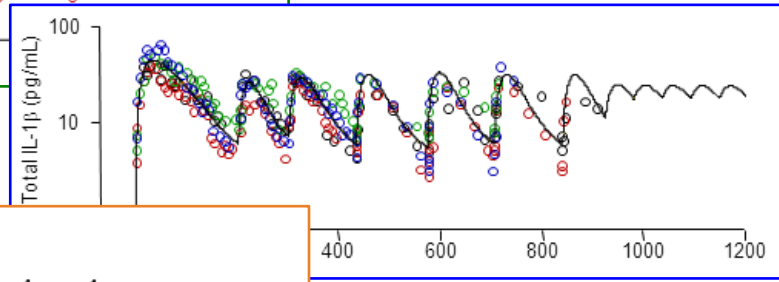
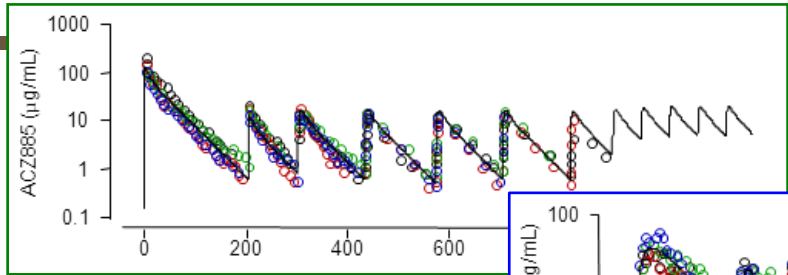
Canakinumab  $\uparrow$  (*dosed*)

Total IL-1 $\beta$   $\uparrow$   
*formation of slowly eliminated DL complexes*

Free IL-1 $\beta$  must  $\downarrow$   
*mass action, available IL-1 $\beta$  taken into complexes*

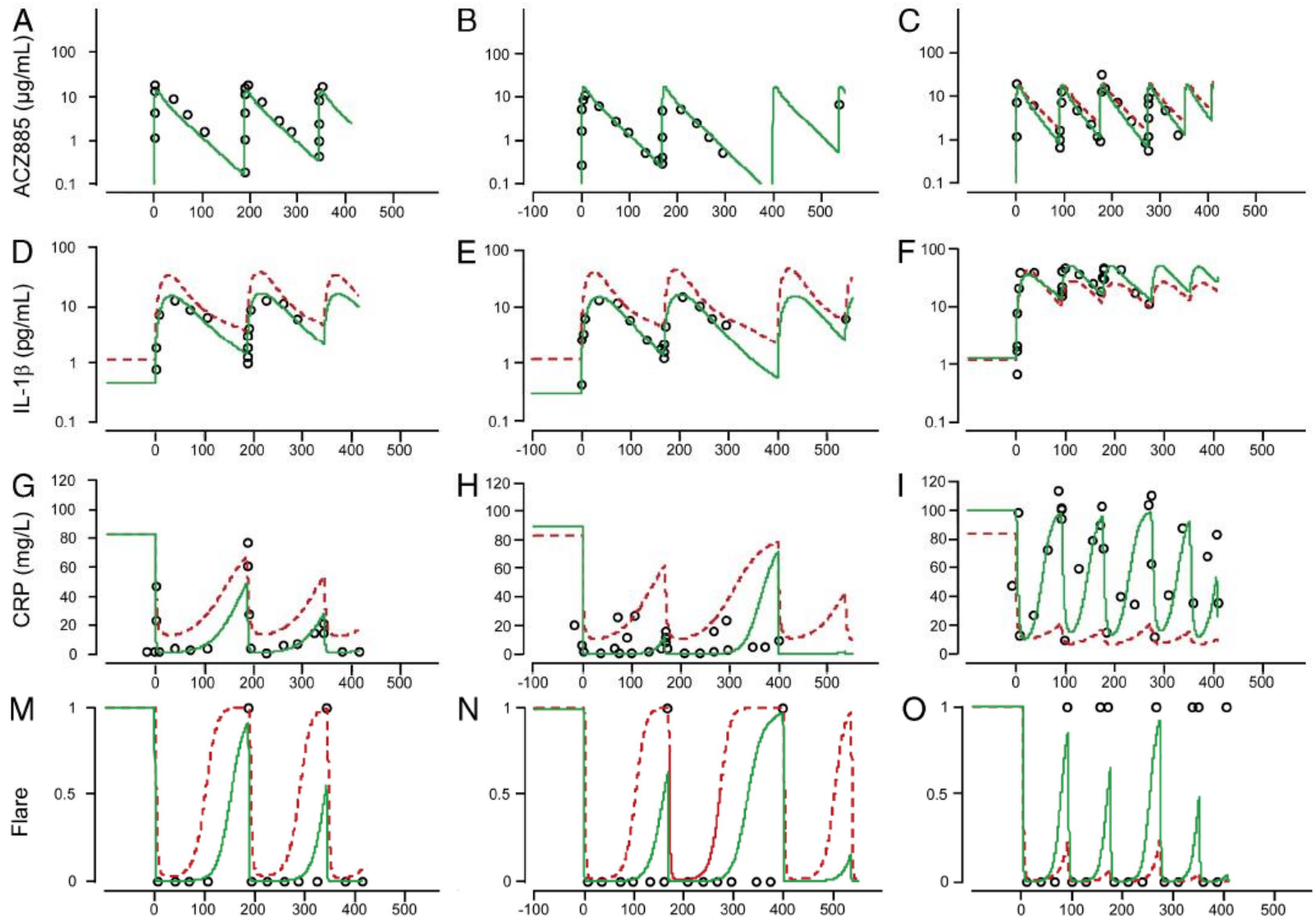
C-reactive protein  $\downarrow$   
Serum Amyloid A  $\downarrow$   
*Inflammatory biomarkers respond to  $\downarrow$  IL-1 $\beta$*

Symptoms  $\downarrow$   
*due to less stimulation by IL-1 $\beta$  receptor in tissues*



# Allowing prediction to the subsequent three patients

Green is fitted, red predicted from prior 4 patients



# Summary – drug-target binding models to quantitate capture or occupancy of ligands or receptors

- There is a hierarchy of drug-target binding and turnover models which describe physiologic location and nature of targets
- Closed, non-turning over (test-tube) models help understand differences between true binding constants ( $K_D$ ) and observable half-max effect concentrations ( $EC_{50}$ )
- Even simplest one compartment binding and turnover models useful for understanding capture of soluble targets located primarily in the central circulation
- Soluble targets primarily located in tissues require two or more compartments; increase in captured target in plasma enables whole body production estimate
- Binding of drug to target can, if molar dose in same range as molar amount target in body, give rise to dose, concentration and even time dependent nonlinear PK
- Such nonlinear or target mediated drug disposition can be used to estimate target occupancy without need for independent receptor occupancy assays
- Quantitation of nonlinear PK or capture of soluble targets can, if related to clinically relevant responses, be used to help set doses and regimens