

Quantitative Systems Pharmacology Modeling Applications at Different Stages of Drug Development

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Quantitative systems pharmacology (QSP) modeling can be applied at different stages of drug development, from early research to late stage development

- Case studies:

- **Preclinical:** Mechanistic insights for AZD3458, a novel selective PI3K γ immuno-modulator
- **Late-stage clinical development and translation between patient populations:** Quantitative systems pharmacology model of potassium homeostasis for Lokelma



Case 1: Mechanistic insights for AZD3458, a novel selective PI3K γ immuno-modulator, using a quantitative systems model

Clinical Motivation

- In 2018, an estimated 1,735,350 new cases of cancer will be diagnosed in the United States and 609,640 people will die from the disease
- Cancer growth and progression are associated with immune suppression
- Monoclonal antibodies that target immune checkpoints provided an immense breakthrough in cancer therapy
- However, response rates remain relatively low; therefore, combination studies with small molecules, chemotherapy and other biologics are of interest to improve anti-tumor responses

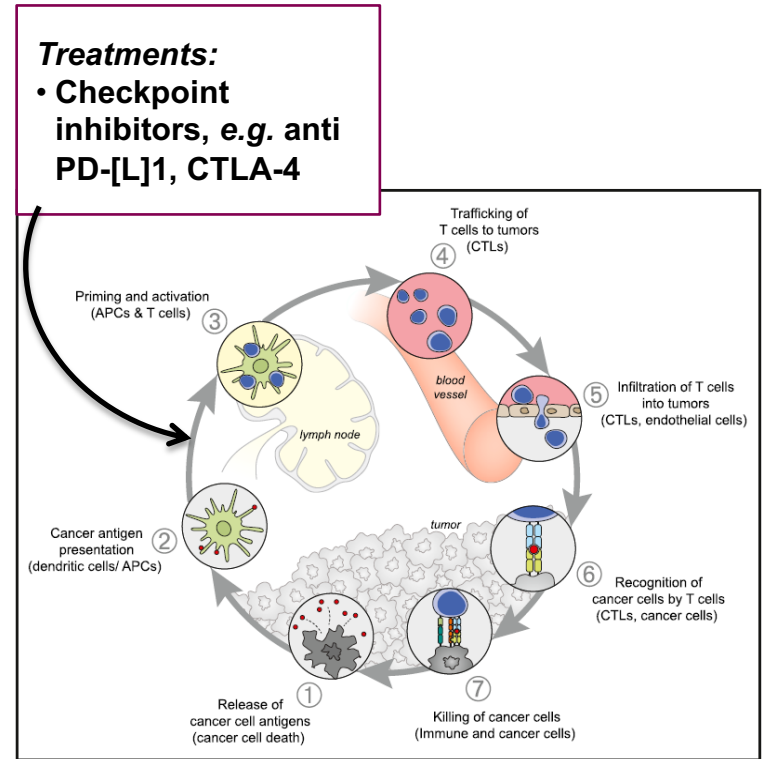
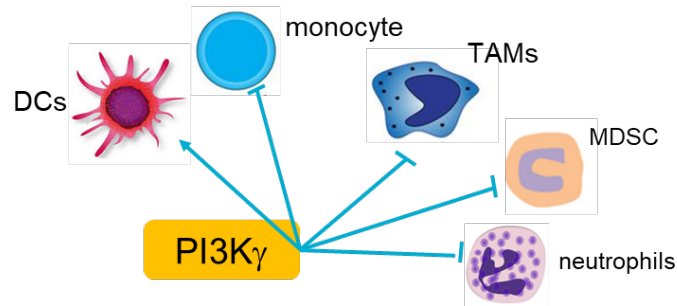


Figure 1. The Cancer-Immunity Cycle

The generation of immunity to cancer is a cyclic process that can be self-propagating, leading to an accumulation of immune-stimulatory factors that in principle should amplify and broaden T cell responses. The cycle is also characterized by inhibitory factors that lead to immune regulatory feedback mechanisms, which can halt the development or limit the immunity. This cycle can be divided into seven major steps, starting with the release of antigens from the cancer cell and ending with the killing of cancer cells. Each step is described above, with the primary cell types involved and the anatomic location of the activity listed. Abbreviations are as follows: APCs, antigen presenting cells; CTLs, cytotoxic T lymphocytes.

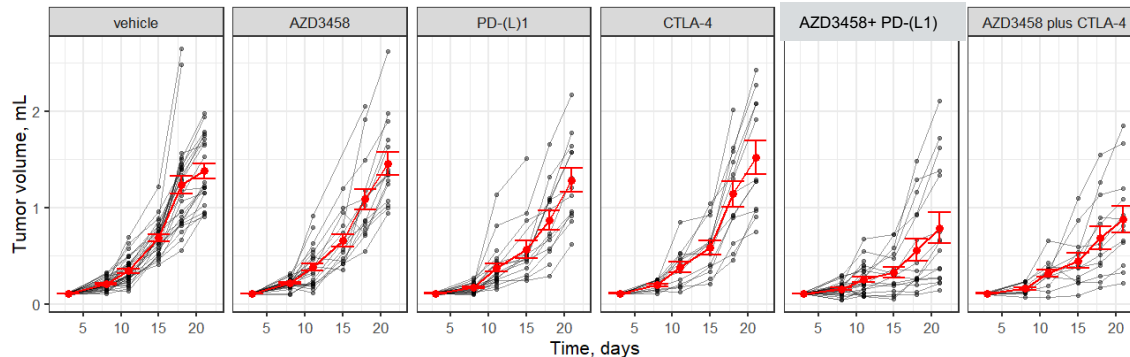
The PI3K γ isoform is a key regulator of immune cell proliferation, survival, migration and activation

- In the context of cancer, PI3K γ inhibition may re-program myeloid cells, re-polarizes macrophages to an immuno-stimulatory state, and promote cytotoxic CD8 T cell activity, thereby eliciting a tumor immune response
- AZD3458 is a highly selective PI3K γ clinical candidate with potential as monotherapy and in combination with checkpoint inhibition, to overcome resistance and to enhance efficacy of checkpoint inhibitors

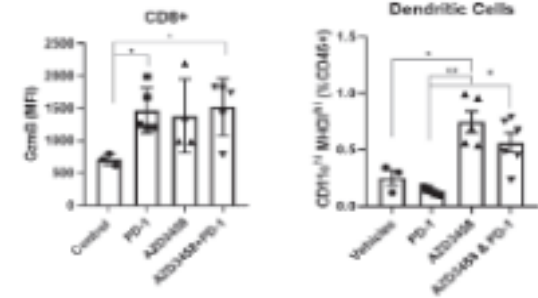


A QSP model was needed to understand the mechanism of action of AZD3458 (PI3K γ) and dose sequence optimization

- AZD3458 (PI3K γ) mono or anti-PD-(L1)/CTLA-4 is shown to have only limited efficacy in MC38. However, synergy is achieved with combination
- **Complex immune modulation: AZD3458(PI3K γ) was shown to increase CD8 and dendritic cells; however, the key contribution mechanism in anti-tumor response is not clear**
- **How dose sequencing affects anti-tumor responses is not clear**
- Could we use the model to identify responders vs. non-responders?



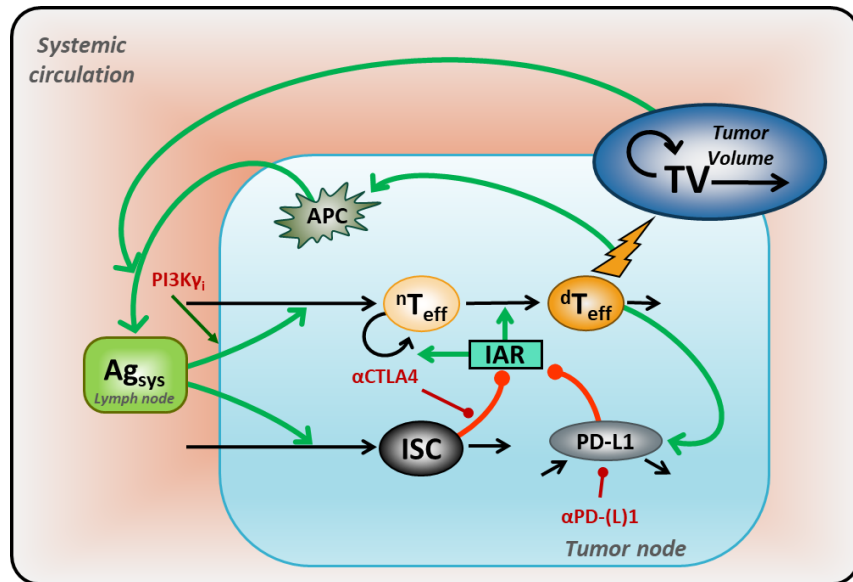
Observed increased dendritic cell infiltration and T-cell activation in MC-38 TME assessed by flow cytometry and aligned with model predictions



An immuno-oncology QSP model was adapted to incorporate PI3K γ inh MoA in a MC38 syngeneic model

- PI3K γ : PI3K γ = $k_{eff} \cdot \frac{1}{\left(1 + \frac{Cc2}{K_{dAZD}}\right)}$;**
- $TN_{inf} = k_{Ln} \cdot \frac{A_g}{A_g + S_L} \cdot PI3K\gamma$
- Infiltrated T cell proliferation and differentiation in TME:**

$$\frac{dT_{eff}^n}{dt} = \frac{k_{Ln} \cdot A_{g_{sys}}}{A_{g_{sys}} + S_L} + k_{pro} \cdot T_{eff}^n \cdot IAR - k_{el} \cdot T_{eff}^n - k_{dif} \cdot IAR \cdot T_{eff}^n; \frac{dT_{eff}^d}{dt} = k_{dif} \cdot IAR \cdot T_{eff}^n - k_{apo} \cdot T_{eff}^d$$
- IAR (Immune activation rate function):** $IAR = (1 - PDL1_{free}) \cdot (1 - ISC^*)$, where $ISC^* = \frac{A_{g_{sys}}}{A_{g_{sys}} + S_R} \cdot (1 - I_{max,ctla4} \cdot CTLA4_{occ})$
- Nonlinear mixed-effects (NLME) model analysis was performed. Parameter and relative standard error estimation were based on the stochastic approximation expectation maximization (SAEM) algorithm

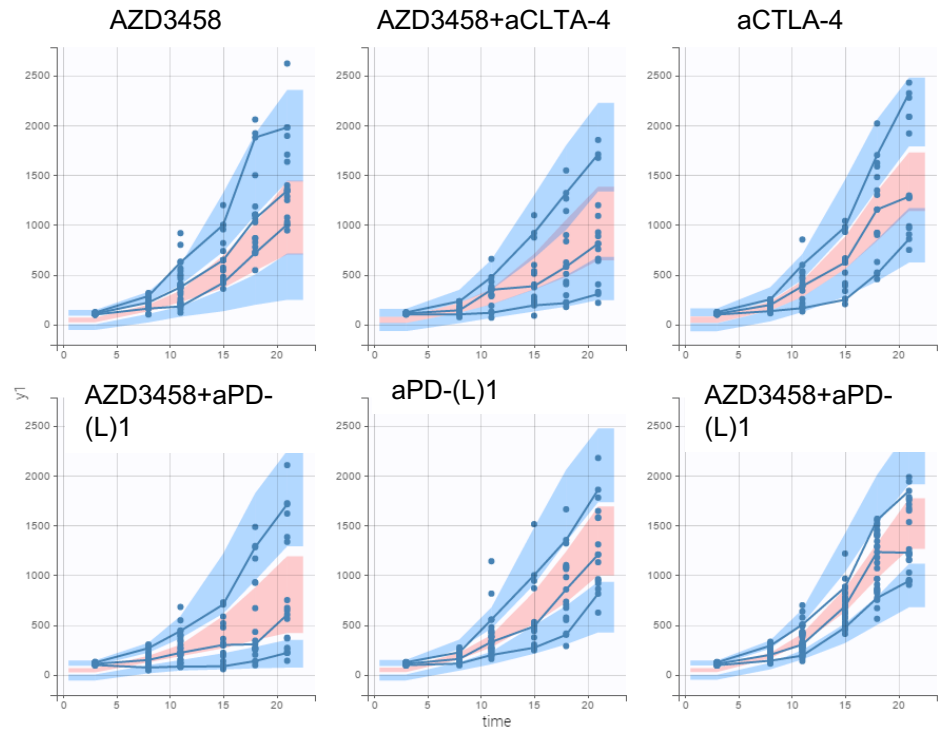
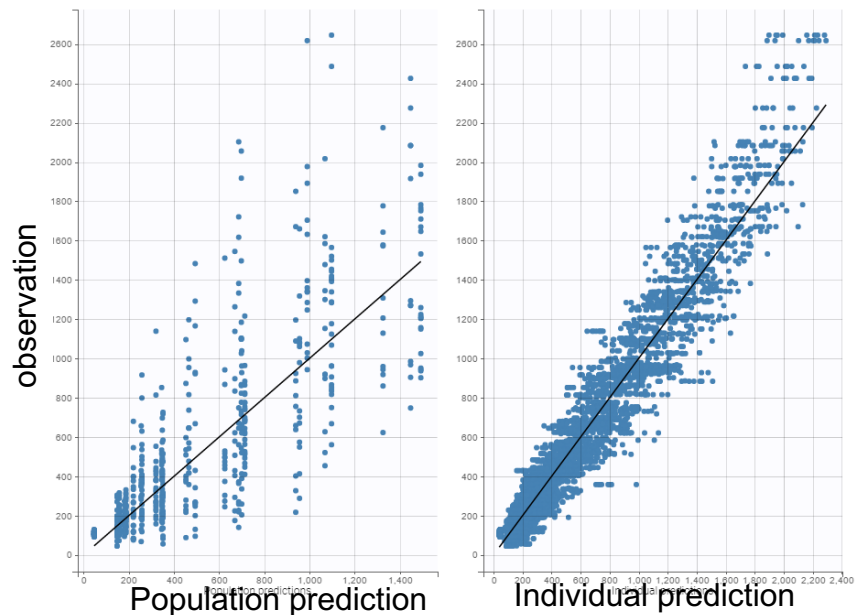


PI3K γ inh
PK module
(BID)

aPD-(L)-1
PK module
(Q3D)

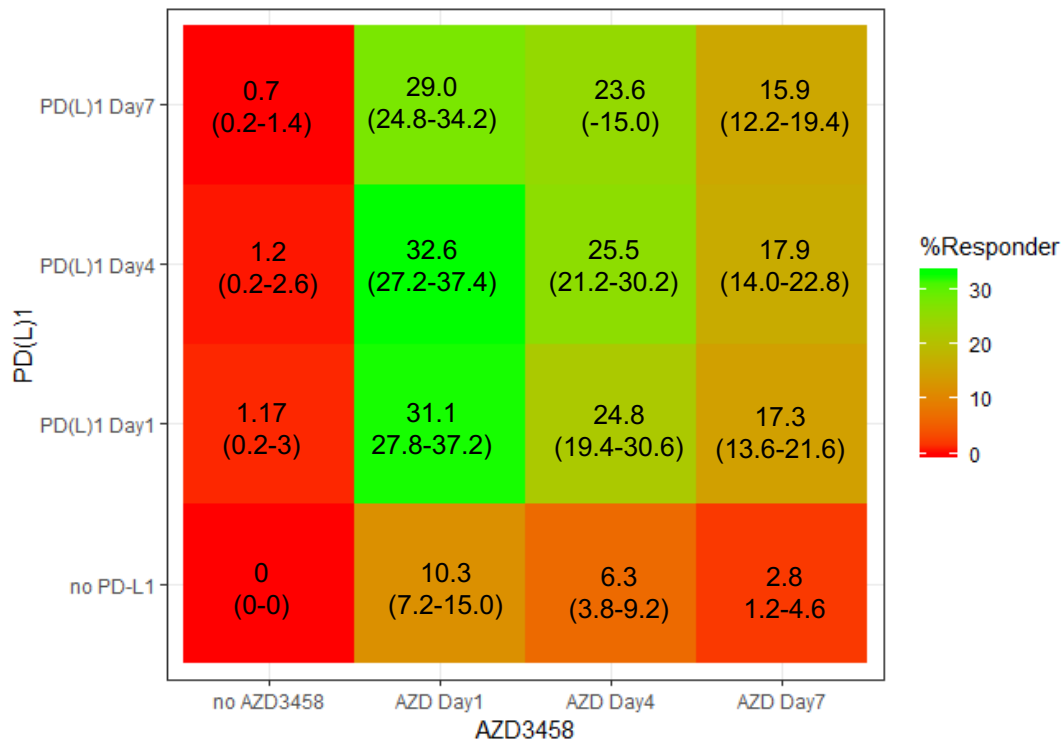
aCTLA4
PK module
(Q3D)

Model adequately described cohort- and individual-level tumor size dynamics patterns, for 4 compounds anti-PD-1, anti-PD-L1, anti-CTLA4, PI3Ky inhibitor treatment regimens in MC38



Application: Model suggests that dosing AZD3458 prior to anti-PD(L)1 leads to better efficacy

- A 50 trial-simulation of a virtual population (N=500) was used. Anti PD(L)-1 and AZD3458 were started either on days 1, 4 or 7 after animal inoculation
- Percentage of responders was calculated. Simulated percentage aligns well with observed data (e.g. ~30%)
- Simulations indicate that treating with AZD3458 prior to PD(L)-1 is preferred



Summary

- **Utility of QSP in preclinical space for immuno-oncology**
 - Quantitative framing of target and biomarkers in disease and efficacy context
 - MoA elucidation
 - Understanding of responders vs. non-responders
 - Combination selection
 - Dose scheduling/sequencing
 - TME condition vs. response predictions
 - PKPD predictions
- **Soft skill development (communication with DMPK modelers, *in vivo* pharmacologist, cell biologist)**
- **Impact: Provide rationale for dose sequencing in combination study for team**

**Case 2: Quantitative Systems Pharmacology Model of
Potassium Homeostasis for Sodium Zirconium
Cyclosilicate (SZC, LOKELMA[®])**

Hyperkalemia – What is it? Why is it bad?

Causes:

- Kidney disease
- Heart failure
- Alcohol/drug abuse
- Type 1 diabetes
- ACE, ARB side effects

Hyperkalemia is High $[K^+]_{\text{serum}}$

Normal: 3.6 - 5.0 mM, High: > 5.0 mM

Altered K^+ Nernst potential ~ gradient from intracellular (140 mM) to extracellular fluid

Altered transmembrane potential on excitable cells like myocytes (muscle) and neurons (nerve)

Raises risk of **life threatening cardiac arrhythmias and death**

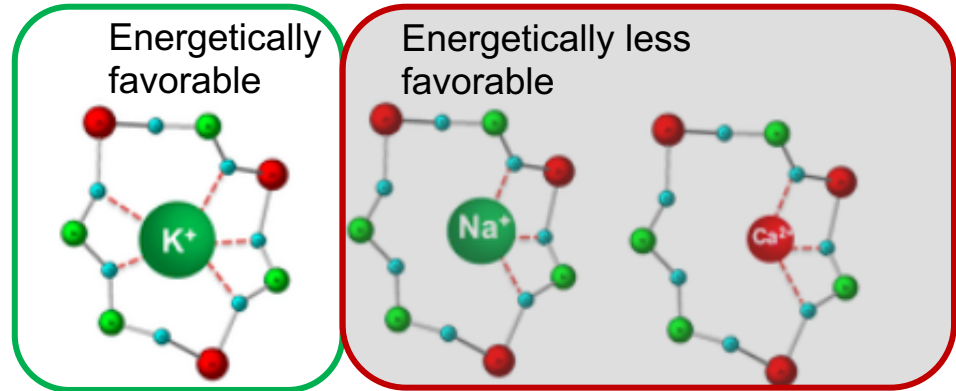
The K^+ gradient is a critical “recovery” or “resetting” force in excitable cells

Treatments:

- Low K^+ diet
- K^+ binders
- Dialysis

Sodium zirconium cyclosilicate (SZC, *LOKELMA*®)

- Orally-administered suspension
- **Insoluble** crystal that binds K^+ ions from solution
- **Not absorbed** from the gastrointestinal tract into the circulation
- Highly specific for K^+ ions
- Binds intestinal K^+ , prevents absorption, creates a gradient to pull K^+ out of circulation
- Applications for acute and chronic potassium reduction

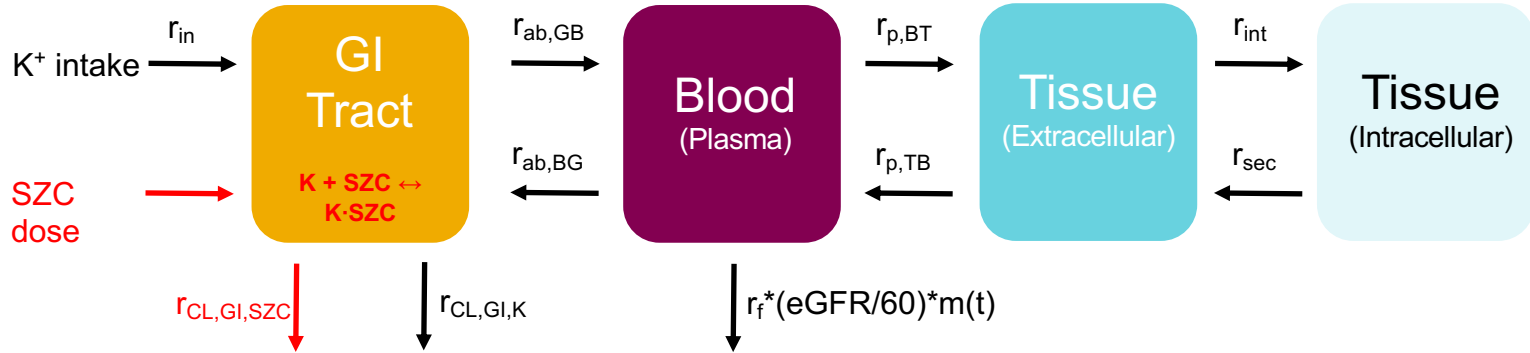


A QSP model was needed to understand SZC mechanism of action and to address clinical questions in late-stage development and clinical trial design

- The redistribution and excretion of potassium in patients treated with SZC are not fully understood
- Many clinical questions regarding clinical trial design, dose regimen confirmation and safety/manufacturing have been raised

Base case: SZC was developed for chronic use to lower serum K^+ in hyperkalemic subjects

A QSP model was developed to describe SZC-K⁺ binding within a framework of physiological K⁺ homeostasis (absorption, transport, disposition and excretion) in healthy and hyperkalemic subjects



- K⁺ intake: 3 meals/day (8am, noon, 6pm), 100mEq total, 25% breakfast, 33% lunch, 42% dinner
- A wide range of doses/regimens are taken into consideration: Dosing taken directly from patient data → accounts for deviations from planned treatment
- Model estimation (patient specific parameter sets):
 - In-vitro SZC-K⁺ binding data
 - QSP model to individual patient plasma K⁺ data from 3 SZC clinical trials (N=1101 Patients)
 - Literature data/Physiological constraints

Application: Safety assessment of a child resistant package

Background:

- Through a standardized test of the Lokelma sachets, it was concluded that a child may be able to open one sachet during a reasonable unattended time
- This study is designed to explore the immediate decrease in plasma $[K^+]$ upon consumption of 1 dose of 10g by a 2-year-old child

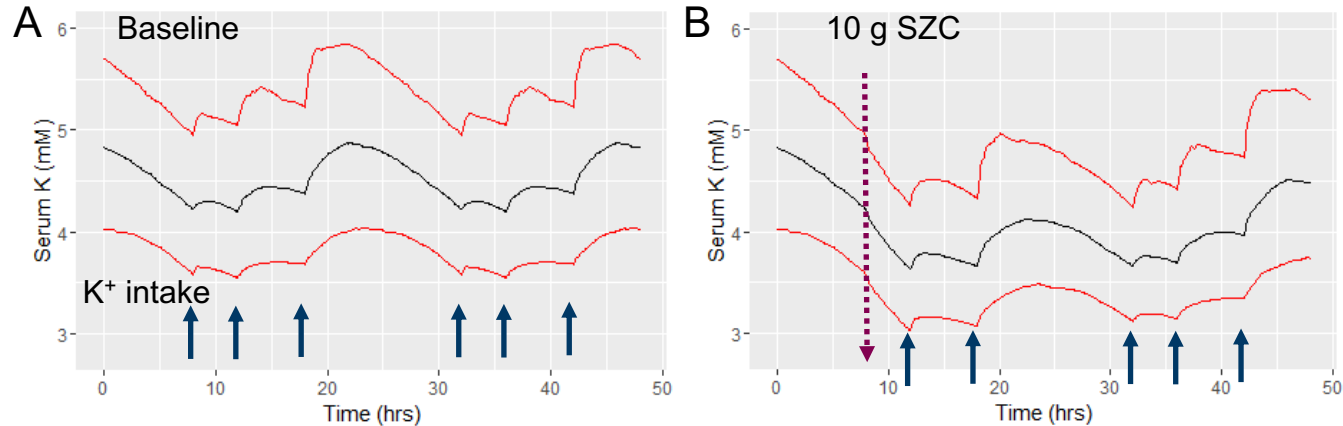
Question: Whether additional child-resistant packaging is necessary for SZC?

Method: The QSP model was modified to represent the physiology in a 2-3 year-old child (extrapolated from the adult patient parameters)



Application: Safety assessment of a child resistant package

- *Question:* Whether additional child-resistant packaging is necessary for SZC?



- Model showed that though there is an immediate reduction in plasma [K⁺] following SZC ingestion, only 2.8% of the virtual population became hypokalemic (<3.0mM, trough ~ 2.95mM, duration ~0.3hr)

Summary

- **Utility of QSP model in late stage development**
 - Within-class differentiation vs. competition (on efficacy and safety grounds)
 - MoA rationale in context of quantitative disease understanding
 - Quantitative translation across indications, across patient populations
- **Soft skill development: communication with project team to convince them of our modeling approaches/assumption/learnings**
- **Impact:** Modeling results in conjunction with clinical safety data and non-clinical toxicology data were used to show the commercial packaging met the requirements and avoided a minimum 6-month delay in US launch

Acknowledgement

- **Quantitative clinical pharmacology team at AstraZeneca**
 - Gabriel Helmlinger, Don Stanski and team
- **Collaborators at AstraZeneca**
 - M&S-Decisions, a Moscow-based modeling consultancy
 - Biosciences (preclinical)/DMPK
 - Translational science/Biomarker group
 - Bioinformatics
 - Oncology Translational Medicine Unit (clinicians)
- **PKS sciences, NIBR, Novartis**
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- **Johns Hopkins University, Winslow lab**
 - Raimond L. Winslow, Ph.D.
 - Joseph L. Greenstein, Ph.D.


ASCPT Webinar Series
**Systems Pharmacology (SP)
and Early Career (EC) Communities**

**Next Generation Pharmacometricians – Examples
from Early Career Modelers on MID3 Implementation**

Jan. 30th, 2020



**Model Informed Drug
Discovery and
Development (MID3)**



Characterizing Colon TNF α Suppression and Treatment Effects of an Anti-TNF α Monoclonal Antibody in a Mouse Inflammatory Bowel Disease Model: *Questions and Solutions for a Next Generation Pharmacometrician*

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ASCPT Webinar Series (Systems Pharmacology and Early Career Communities)
Jan. 30th, 2020

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Inflammatory Bowel Disease (IBD) Pathology: an Overview

- The **inflammatory bowel diseases** (Crohn's disease; ulcerative colitis) are chronic inflammatory disorders of the intestine and/or colon.
- Intestinal cells that come under stress (eg, **bacterial overgrowth, endotoxin, reactive oxygen species, allergen**) are damaged.
- This **dynamic inflammation** involves the release of **proinflammatory cytokines** and **chemokines** that perpetuate the inflammation.
- **Biologic therapies** target the proinflammatory cytokines (TNF- α and IL-12/23), or the integrins/block the adhesion of leukocytes.

Biologicals approved for the treatment of IBDs in the EU and/or the US

Biological	Target	Disease	Route of administration	Dosing schedule
Infliximab	TNF- α	CD/UC	i.v.	5 mg/kg at week 0, week 2, week 6, then every 8 weeks
Adalimumab	TNF- α	CD/UC	s.c.	CD: 80 or 160 mg at week 0, 40, or 80 mg at week 2, then 20 or 40 mg every 2 weeks UC: 160 mg at week 0, 80 mg at week 2, then 40 mg every 2 weeks
Golimumab	TNF- α	UC	s.c.	200 mg at week 0, 100 mg at week 2, then 50 mg every 4 weeks if weight < 80 kg (EU) or 100 mg every 4 weeks if weight \geq 80 kg (EU and US)
Certolizumab pegol	TNF- α	CD	s.c.	400 mg at week 0, week 2, week 4, then 400 mg every 4 weeks
Natalizumab ^a	α 4 integrin	CD	i.v.	300 mg at week 0, then every 4 weeks
Vedolizumab	α 4 β 7 integrin	CD/UC	i.v.	300 mg at week 0, week 2, week 6, then every 8 weeks
Ustekinumab	IL-12/23	CD	First i.v., then s.c.	i.v. infusion of 260 mg if weight \leq 55 kg, 390 mg if weight 55–85 kg, 520 mg if weight > 85 kg at week 0, then 90 mg s.c. every 8 weeks

Loading dose or induction phase is included in all cases.

Objectives of the Study

*Approved biologics data in IBD suggest that **once the inflammatory burden is reduced (eg, with loading doses during disease flare)**, the drug exposures required to maintain efficacy may be lower than that with initial induction.*

Key Questions: What is the impact of IBD on:

- anti-TNF α mAb disposition
- the ability of anti-TNF α mAb to neutralize TNF α at the colon
- the potential mechanism behind utilization of loading dose for treatment of IBD?

Key Challenges and Opportunities:

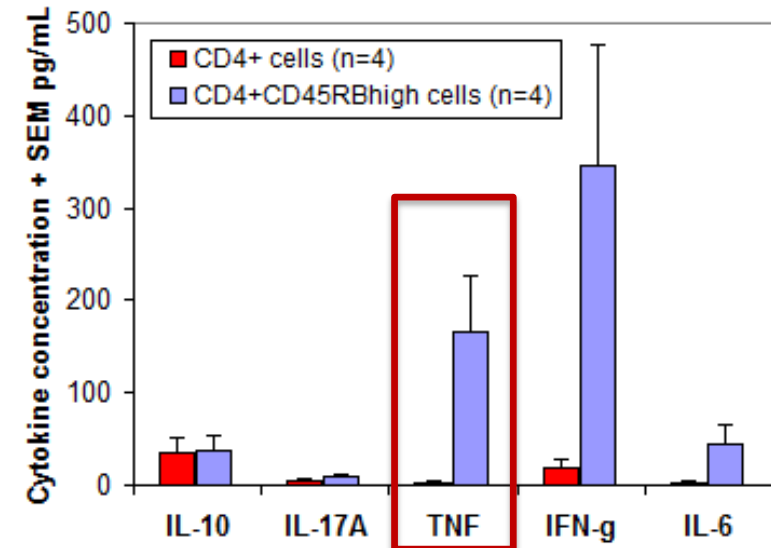
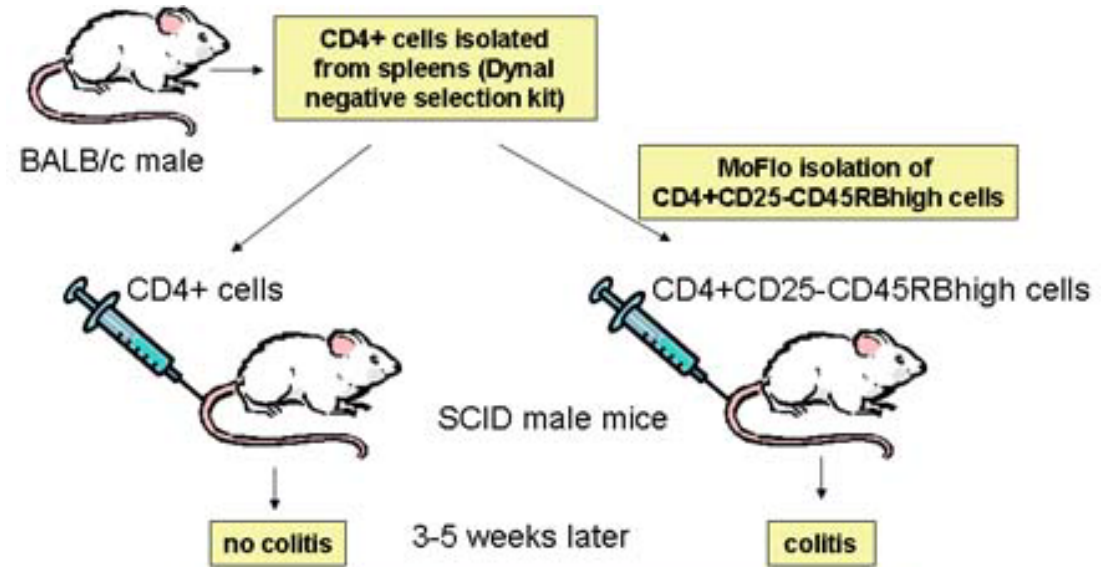
Questions for a Next Generation Pharmacometrician

- What **animal model** is relevant to human IBD?
- What **doses/dosing regimen** are needed?
- What experimental **endpoints and sampling** should be included?
- What type of bioanalytical **assays** are needed?
- What kind of **modeling approach** (including software) is fit-for-purpose and can provide robust parameter estimates to interpret the study findings?
- **What can be learned** from both experimental data and modeling results?



CD45RB^{high} Adoptive T Cell Transfer Model in Mice

- Adaptive transfer of **CD4⁺CD45RB^{high} T cells** (**naive T cells**) from healthy wild-type (WT) mice into syngeneic recipients that lack T and B cells
 - induces a **pancolitis** and **small bowel inflammation** at 5–8 wk following T cell transfer



Am J Physiol Gastrointest Liver Physiol 305: G763–G785, 2013.

Powrie F, Leach MW, Mauze S, Caddle LB, Coffman RL. *Int Immunol* 5: 1461–1471, 1993.

T cell transfer model of chronic colitis: concepts, considerations, and tricks of the trade. *Am J Physiol Gastrointest Liver Physiol*. 2009;296:G135–G146.

https://hookelabs.com/services/cro/ibd/colitis_scid/

Study Design

Disease model

- Mouse T-cell transfer IBD model

Study compounds

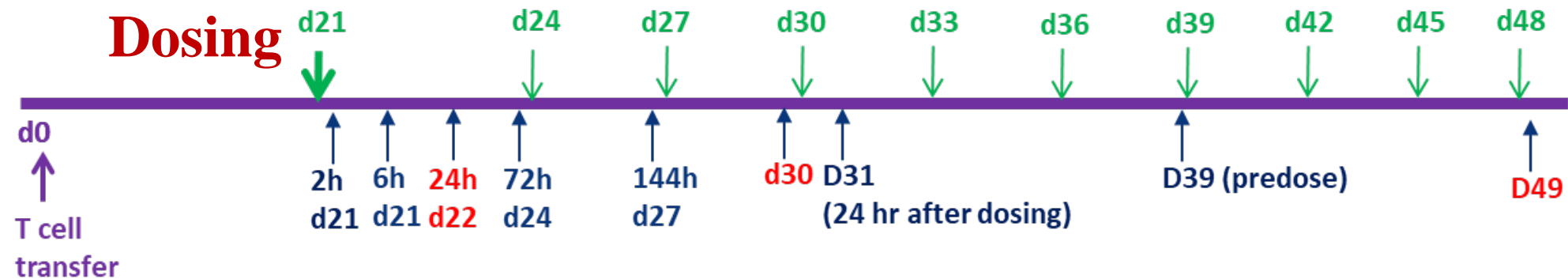
- CNTO 5048: **an anti-murine TNF α surrogate mAb of golimumab (IgG1)**
- CNTO 1322: isotype control

Readouts

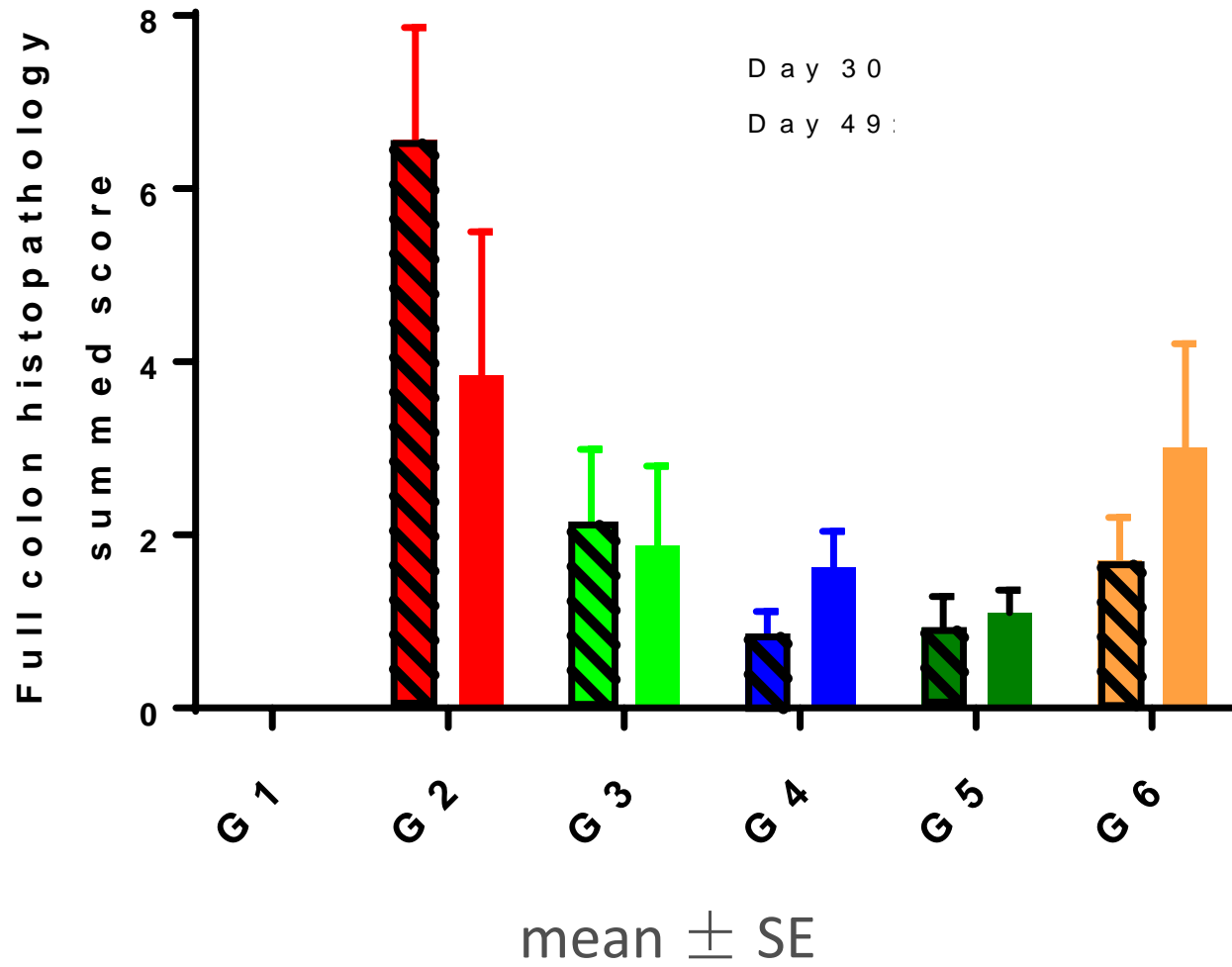
- **mAb PK:** systemic and at the tissue site (colon)
- **Target engagement:** soluble TNF α , systemic and at the tissue site (colon)
- **Disease score** including colon weight/length
- **Histopathology**

Study Design based on Prior Studies and PK Simulations

Study Group	N	Treatment	Dose
1	30	Non-IBD mice–CNTO 5048	Single i.v. 10 mg/kg
2	30	IBD mice –CNTO 5048	Single i.v. 10 mg/kg
3	18	IBD mice – Isotype control	10mg/kg i.v. loading dose +9* 0.3mg/kg Q3D i.p. maintenance dose
4	22	IBD mice – CNTO 5048	10mg/kg i.v. loading dose +9* 0.3mg/kg Q3D i.p. maintenance dose
5	22	IBD mice – CNTO 5048	1.4mg/kg i.v. dose +9* 1.4mg/kg Q3D i.p. maintenance dose
6	22	IBD mice – CNTO 5048	0.3mg/kg i.v. dose +9* 0.3mg/kg Q3D i.p. maintenance dose

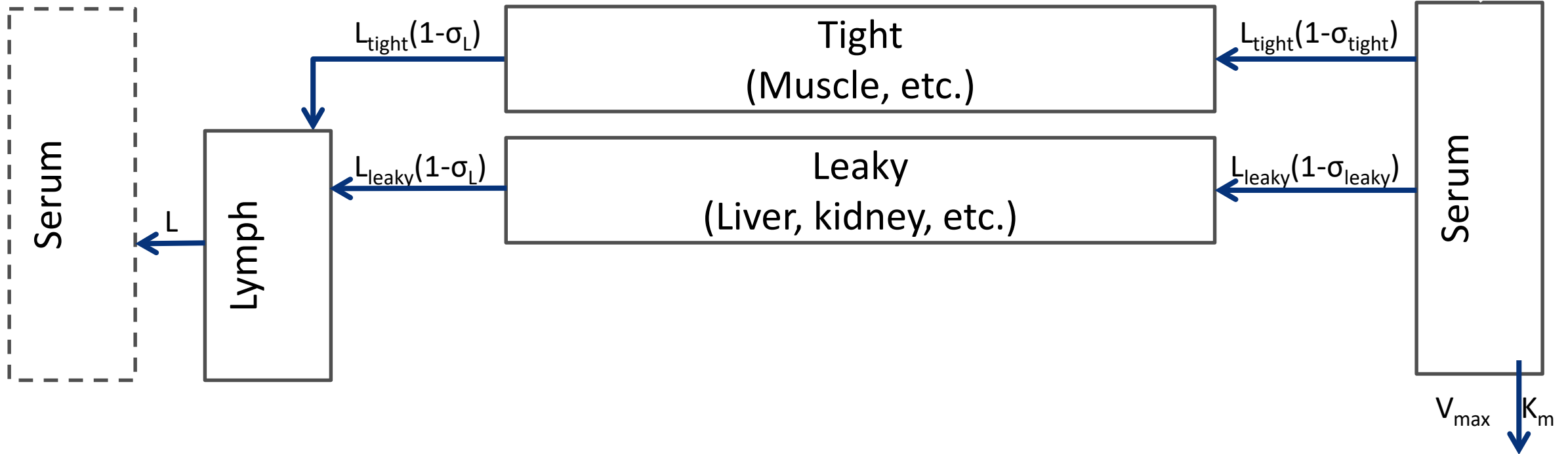


Summary of Representative Experimental Observations



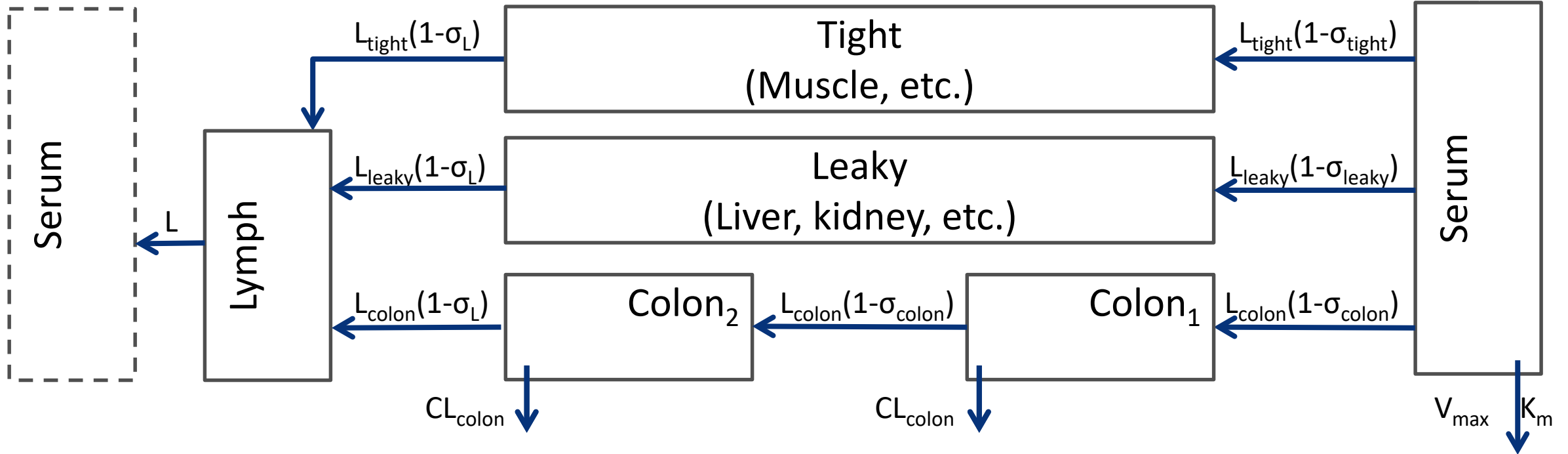
- G1: **Non-IBD mice** (without CD45RB^{high} T cell transfer), CNTO 5048 (10 mg/kg)
- G2: IBD mice (+ CD45RB^{high} T cell transfer), **isotype control** (1x 10 + 9x 0.3 mg/kg)
- G3: IBD mice, CNTO 5048 (1x 10 mg/kg)
- G4: IBD mice, CNTO 5048 (1x10 + 9x0.3 mg/kg)
- G5: IBD mice, CNTO 5048 (10x 1.4 mg/kg)
- G6: IBD mice, CNTO 5048 (10x 0.3mg/kg)

Step-wise Model Building



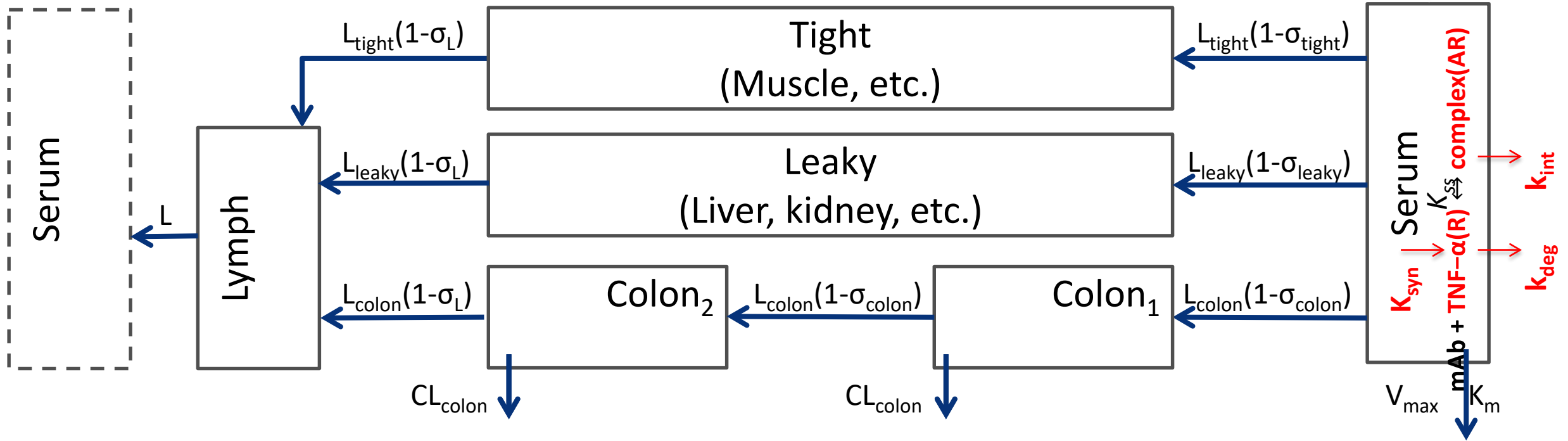
mPBPK/PD model was developed to characterize CNTO 5048 PK and disposition, colon soluble TNF α target engagement (TE) and PD effect

Step-wise Model Building



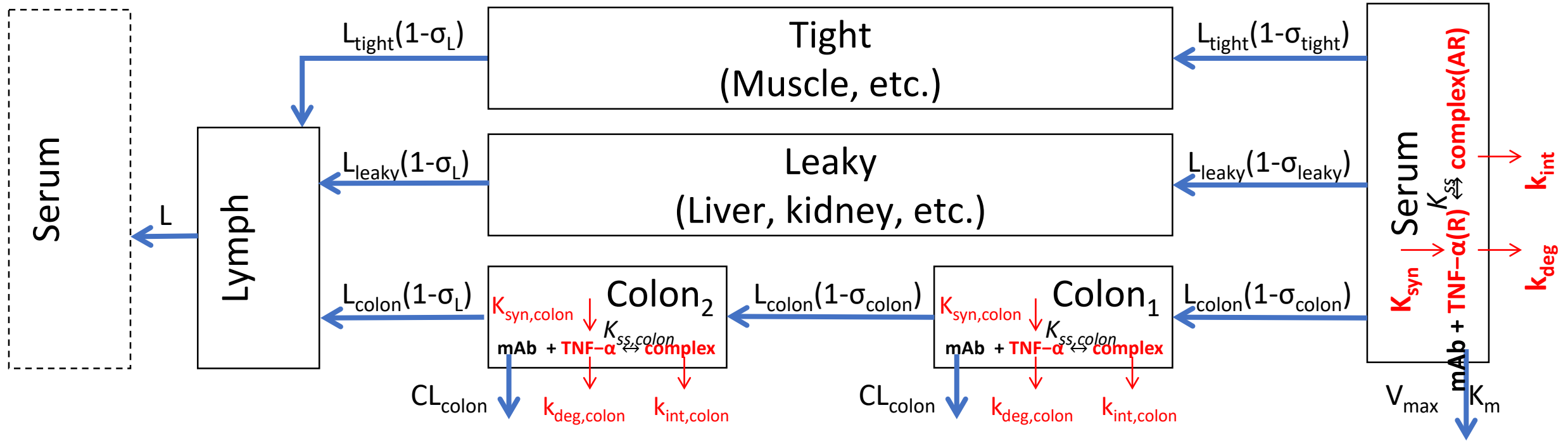
mPBPK/PD model was developed to characterize CNTO 5048 PK and disposition, colon soluble TNF α target engagement (TE) and PD effect

Step-wise Model Building



mPBPK/PD model was developed to characterize CNTO 5048 PK and disposition, colon soluble TNF α target engagement (TE) and PD effect

Step-wise Model Building



mPBPK/PD model was developed to characterize CNTO 5048 PK and disposition, colon soluble TNF α target engagement (TE) and PD effect

Modeling Software and Methods

- Model fitting was performed with **Monolix 2019R1** using **naive pooling** of data.
- **Importance sampling algorithm** was used to calculate likelihood and Stochastic Approximation was used to derive Fisher information matrix.
- The **proportional residual error model** was used for serum CNTO 5048 concentrations; **constant residual error models** were used for colon concentrations of CNTO 5048, serum and colon TNF α concentrations.
- **Model performance** was evaluated by goodness-of-fit plots and -2 log likelihood value (-2LL).

Key Model Parameters

Parameter	Definition
V_s^a	Serum volume
ISF^a	Volume of total interstitial space fluid
V_{lymph}^a	Lymph volume
L^a	Total Lymph flow rate
σ_L^a	Reflection coefficient for lymph
$V_{max,nonIBD}$	Antibody elimination rate capacity in serum for non-IBD mice
$V_{max,IBD}$	Antibody elimination rate capacity in serum for IBD mice
K_m	Apparent affinity for antibody elimination
σ_{tight}	Reflection coefficient for tight tissues
σ_{leaky}	Reflection coefficient for leaky tissues
k_a	First-order rate constant for i.p. absorption
$\sigma_{colon,nonIBD}$	Reflection coefficient for colon in non-IBD mice
$\sigma_{colon,IBD}$	Reflection coefficient for colon in IBD mice
$CL_{colon,nonIBDhy}$	Clearance of antibody from colon in non-IBD mice
$CL_{colon,IBD}$	Clearance of antibody from colon in IBD mice
$V_{colon,nonIBD}^b$	ISF volume of colon in non-IBD mice
$V_{colon,IBD}^b$	ISF volume of colon in IBD mice
$L_{colon,nonIBD}^b$	Lymph flow rate of colon in non-IBD mice
$L_{colon,IBD}^b$	Lymph flow rate of colon in IBD mice
k_{deg}	Degradation/Turnover rate constant for TNF α in serum
k_{int}	Internalization (elimination) rate constant for CNTO5048–TNF α complex in serum
R_0^c	Baseline concentration of TNF α in serum in IBD mice
K_{ss}	Quasi-equilibrium binding constant for CNTO5048 and TNF α in serum
$k_{deg,colon}$	Degradation rate constant for TNF α in colon ISF
$k_{int,colon}$	Internalization rate constant for CNTO5048–TNF α complex in colon ISF
$R_{0,colon}^{c,d}$	Baseline concentration of TNF α in colon ISF in IBD mice
$K_{ss,colon}$	Quasi-equilibrium binding constant for CNTO5048 and TNF α in colon ISF

35 equations (including derived equations) were used for the final model.

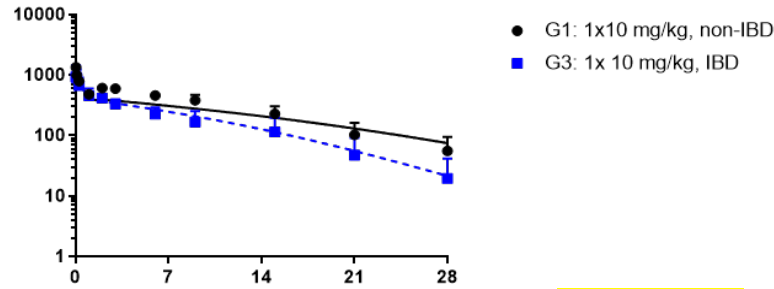
References for parameters

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Model fitting for CNTO 5048 in Serum or in Colon Homogenate

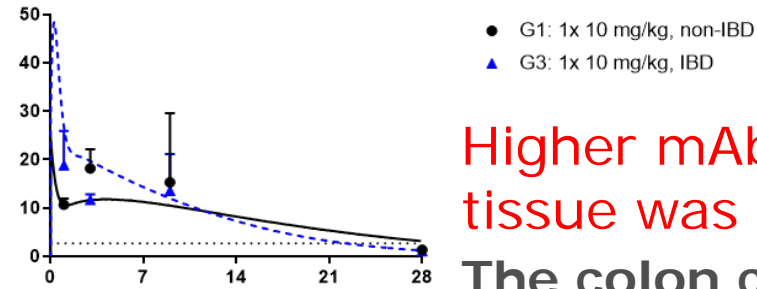
(A)

Serum PK



(B)

Colon PK

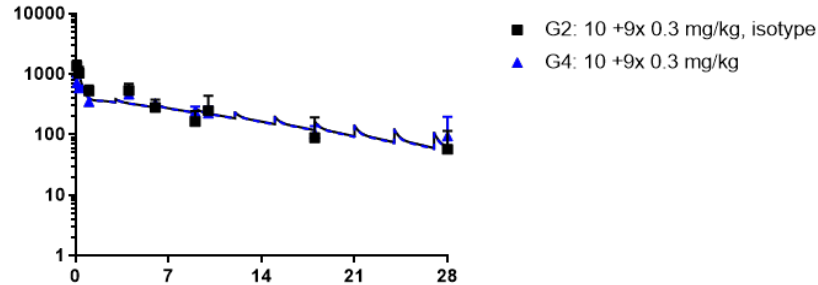


Higher mAb distribution to inflamed tissue was demonstrated.

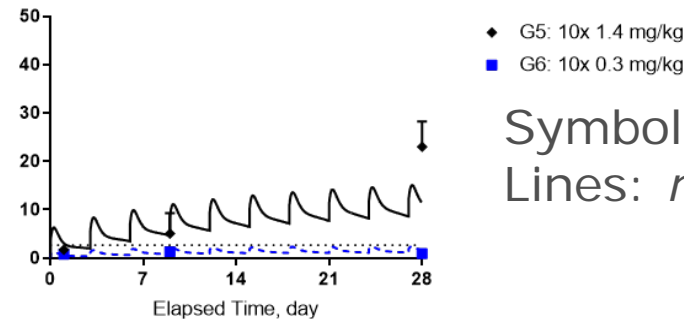
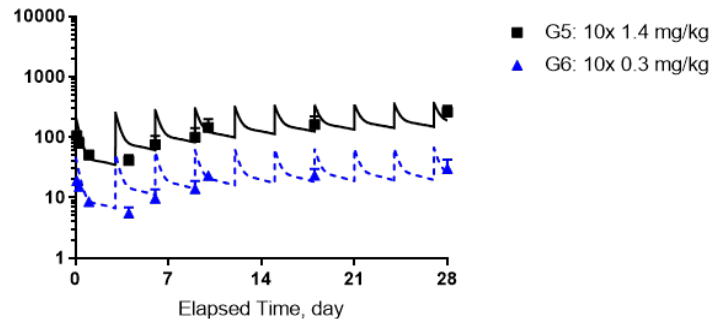
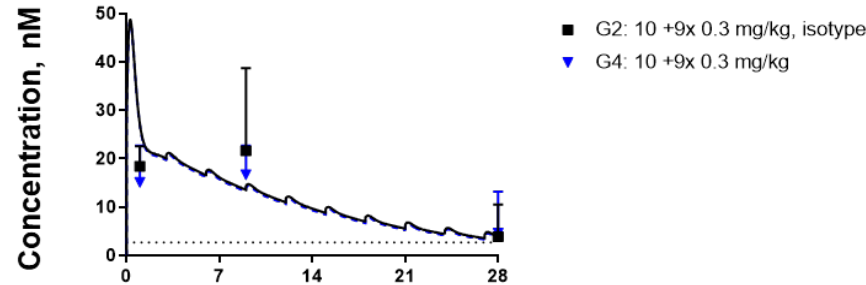
The colon concentrations of CNTO 5048 at 24 hr were **~2-fold higher** in the IBD mice

CNTO 5048 exhibited **~33%** higher systemic clearance in the IBD mice.

Concentration, nM



Concentration, nM



Symbols: *observed mean (SD)*
Lines: *model prediction*

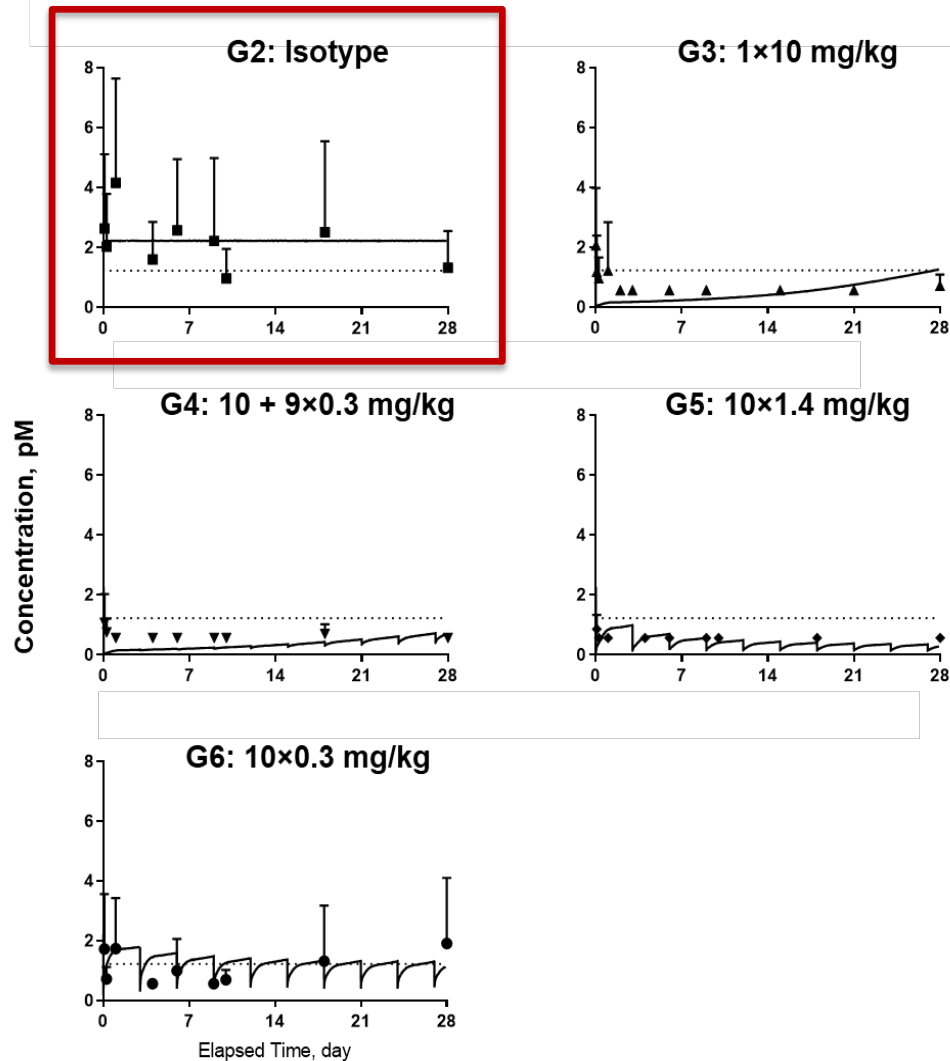
Key Learnings from PK Modeling

- Inflammatory diseases such as IBD can increase **vascular permeability** and change mAb disposition.
- Previous mPBPK modelling work had suggested that **convection and lymph drainage** are the dominant pathways for biologics' uptake and removal from tissues.
- Consistently, the **model-estimated clearance of CNTO 5048 from colon ISF** compartment in the IBD mice was similar to the reported **lymph flow rate** in IBD mice.

Model fitting for Free TNF α in Serum and in Colon Homogenate

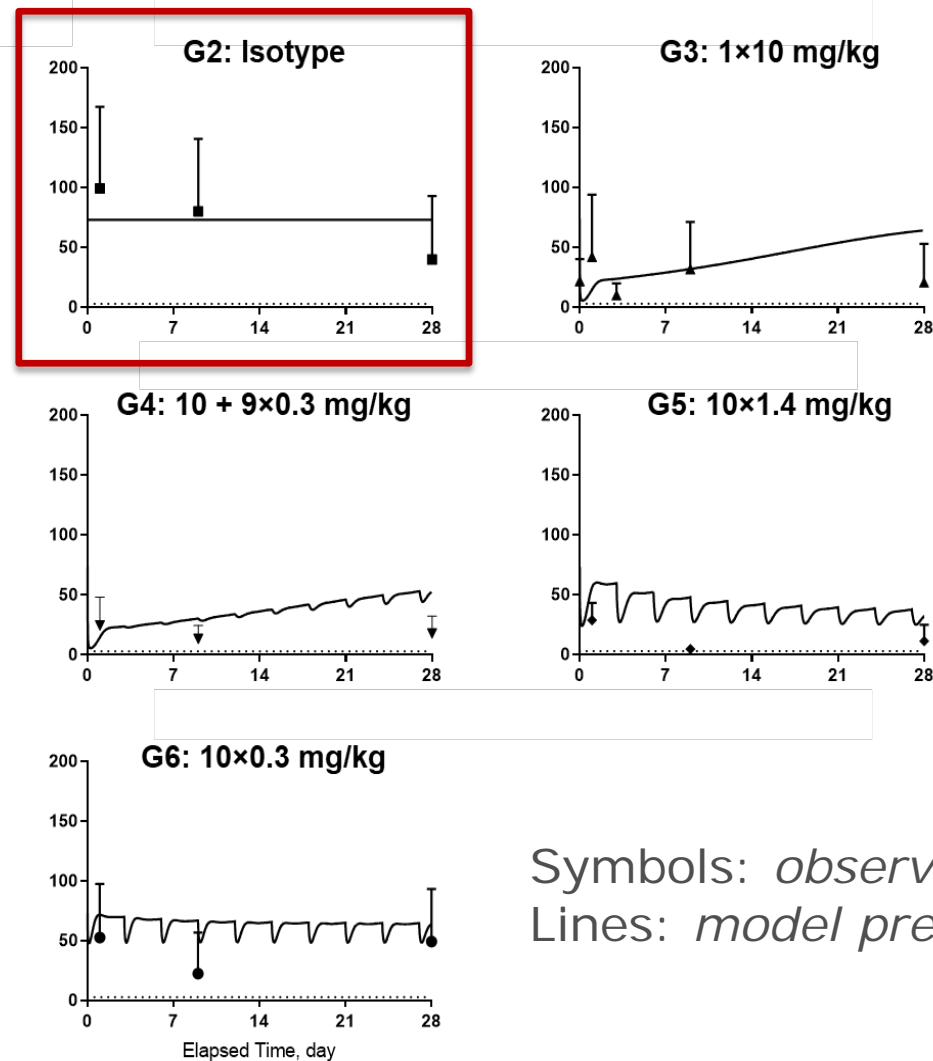
(A)

Serum free TNF α



(B)

Colon free TNF α



- Calculated baseline TNF α in **colon ISF** was \approx **200-fold higher** than serum, indicating that **colon TNF α mainly comes from local secretion.**

Symbols: *observed mean (SD)*
Lines: *model prediction*

Key Learnings from PK/TE Modeling

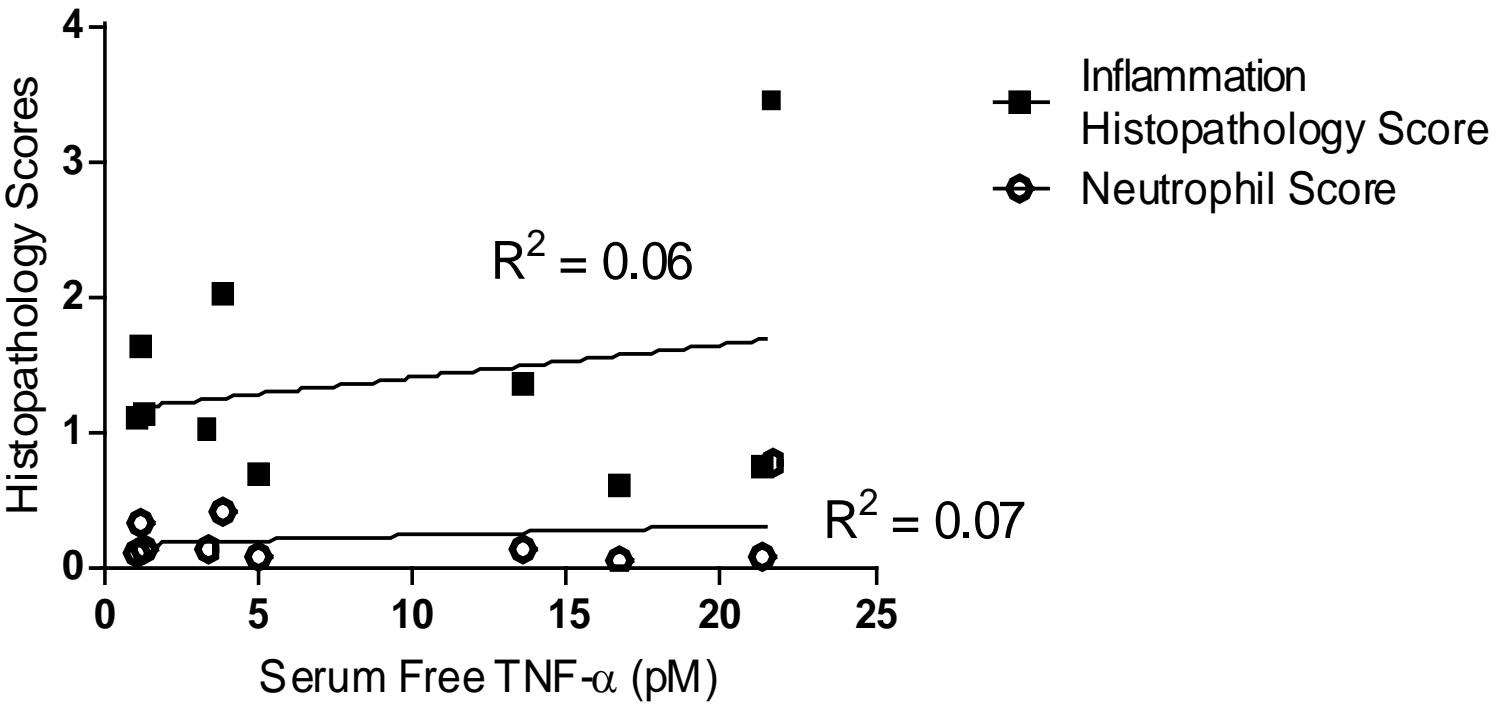
- The **K_{syn}** of TNF α in colon was estimated to be **~17-fold higher** than that in serum.
- CNTO 5048 bound to TNF α in both serum and colon ISF and form mAb/TNF α complexes.
 - ✓ **Serum K_{el}** for **CNTO 5048/TNF α complex** was substantially lower than **K_{deg}** of free TNF α , but considerably higher than the free antibody.
 - ✓ Possibly related to **larger complex formed between trimeric TNF α and the bivalent mAbs**

Key Learnings from PK/TE Modeling

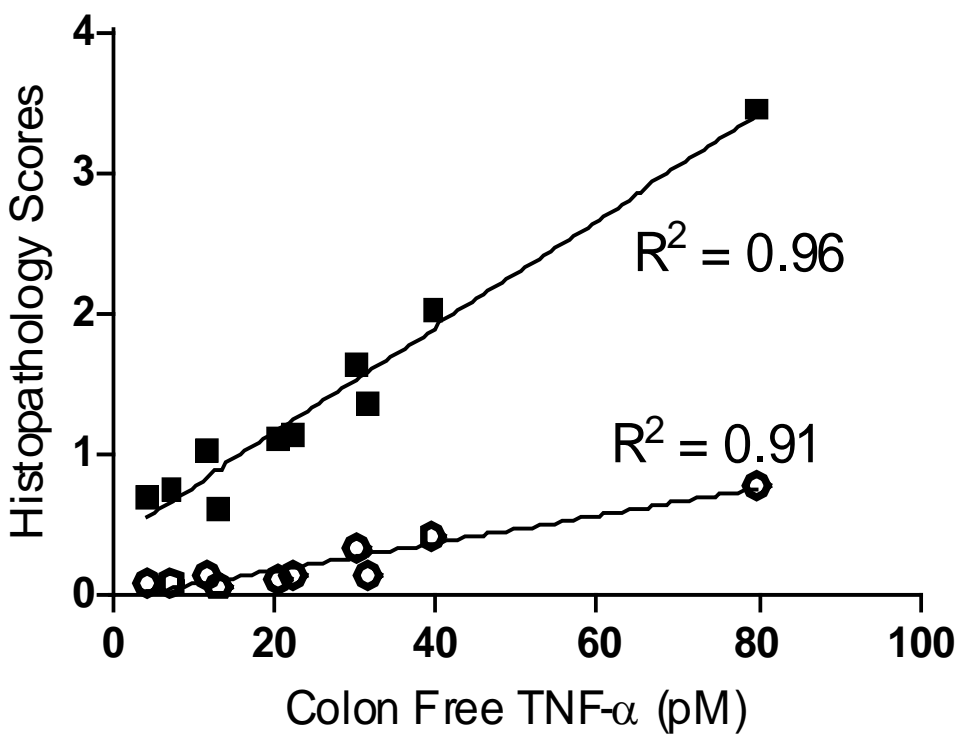
- The estimated *in vivo* quasi-equilibrium constant between CNTO 5048 and TNF α in serum and in colon were both similar to the *in vitro* value.
- The **Kdeg for free TNF α in serum** was estimated and corresponded to a **half-life ($t_{1/2}$) of 3.3 min** which was within 2-fold of the reported ~6 to 7 min $t_{1/2}$ of radiolabelled mouse TNF α in mice serum.
- The **Kdeg for free TNF α in colon** was estimated to be **~10-fold lower** than that in serum.
- The elimination of **CNTO 5048/TNF α complex in colon** was also likely mediated by lymph drainage based on parameter estimates.

Correlations between Free TNF α in Colon or Serum versus Inflammation Histopathology Scores or Neutrophil Scores

Free TNF- α in serum vs Scores



Free TNF- α in colon vs Scores



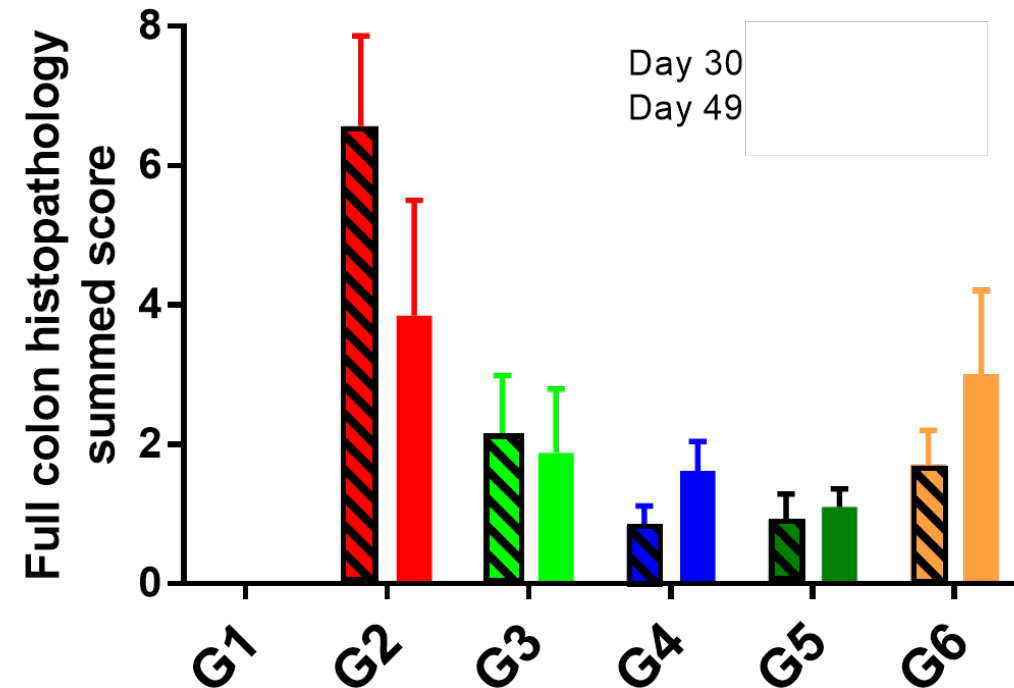
Free TNF α concentrations in colon, but not in serum were shown to correlate well with the colon histopath results.

Key Learnings related to Loading Doses

- The PK/TE assessments provided evidence to support the potential benefit of loading dose in IBD treatment.
- **First, higher mAb distribution to inflamed tissue was demonstrated.**
 - ✓ Previous studies demonstrated *mucosal healing in IBD patients treated with biological agents*
 - ✓ The *reflection coefficient* for colon in IBD mice estimated to be significantly lower than in non-IBD mice
- **Second, the importance of neutralizing TNF α at colon for IBD and higher TNF α production in inflamed colon was demonstrated.**
 - ✓ The baseline TNF α **in IBD colon** shown to be *>26 times higher*

Key Learnings related to Loading Doses

- Results from the **isotype control treated IBD** showed a trend of reduction in colon TNFa with **natural healing of disease**.
- **G6 (10 × 0.3 mg/kg)** with continuous suboptimal colon TNFa suppression showed a **trend of increased inflammation**.
- Suggested **the importance of rapid and complete TNFa suppression in IBD treatment**:
 - ✓ **insufficient suppression of TNFa may allow continuous exacerbation of inflammation**



Conclusions and Limitations:

- ✓ The *disposition of an anti-TNF α mAb* and *TNF α TE* were characterized in a mouse model, and a mPBPK/PD model was developed.
 - ✓ insights on the *rationale of using loading doses for treating IBD*
- ✓ The value of assessing *TNF α neutralization at the site of action* as a biomarker was demonstrated.

Sparse sampling for colon PD and **the temporal change of inflammation after colon TNF α suppression cannot be evaluated fully.**

Impact:

- ✓ Advancement towards ***understanding the MOAs of anti-TNF α mAbs in IBD and clinical dosing strategies for internal programs***
- ✓ Enhanced study designs and analyte selections towards ***drug design/discovery for IBD***

Key Challenges and Opportunities: *Solutions* for a Next Generation Pharmacometrician

- What **animal model** is relevant to human IBD?
 - ✓ Extensive literature reading followed by talking to experts
- What **doses/dosing regimen** are needed?
 - ✓ Consulting with internal experts who had used this animal model; PK simulations
- What experimental **endpoints and sampling** should be included?
 - ✓ Extensive communication with the CRO
 - ✓ Have the “end-goals” in mind; “fit-for-purpose” experimental design
- What type of bioanalytical **assays** are needed?
 - ✓ Discussions with PK and TE assay experts
 - ✓ Consider what is a “must-have” for modeling and data interpretation
- What kind of **modeling approach** (including software) is fit-for-purpose and can provide robust parameter estimates to interpret the study findings?
 - ✓ Extensive survey of literature
 - ✓ Applied mathematics/mechanism-based modeling: evaluate the necessary complexities
 - ✓ In-depth evaluation of modeling tools and algorithms
 - ✓ Cross-check parameter estimates; model-fitting diagnostics
- **What can be learned** from both experimental data and modeling results, and how to apply the learnings?
 - ✓ Reality check with human data and clinical experiences
 - ✓ Communications with clinical pharmacologists and clinicians
 - ✓ Brainstorm with antibody engineers for future molecules



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