



ARTICLE

Modeling the subcutaneous pharmacokinetics of antibodies co-administered with rHuPH20

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Abstract

Predicting the subcutaneous (SC) pharmacokinetics (PK) of antibodies in humans is challenging, with clinical data currently being the only reliable data source for modeling SC absorption and bioavailability. Recombinant human hyaluronidase PH20 (rHuPH20) is an enzyme that facilitates SC delivery of high-dose, high-volume therapeutics. Numerous monoclonal antibodies have been co-administered SC with rHuPH20 in a clinical setting, establishing an extensive PK database. The goal of this work is to demonstrate how aggregated clinical data can be leveraged in a universal modeling framework for characterizing SC antibody PK, resulting in parameterization that can be used in predictive simulations of new antibodies. Data for 10 individual antibodies co-administered SC with rHuPH20 were obtained from publicly available sources. PK modeling of each antibody was conducted using the same model structure, but uniquely parameterized. The model structure consisted of a two-compartment model to capture linear kinetics, plus a target-binding mechanism to accommodate nonlinear kinetics driven by antibody-target complex formation and elimination. The clinical PK profiles for all antibodies were accurately described using the universal modeling framework. The SC PK parameters of absorption and bioavailability were consistent across the range of antibody and target properties evaluated. SC administration with rHuPH20 yielded a 30% increase in absorption rate on average and similar or better bioavailability. These parameter values can serve as initial conditions for model-based PK predictions for new antibodies co-administered SC with rHuPH20 to enable evaluation of optimal SC dose and schedule regimens prior to and during clinical development.

Study highlights**WHAT IS THE CURRENT KNOWLEDGE ON THE TOPIC?**

Preclinical animal models are of limited utility for translating/predicting SC PK of antibodies in human. Meta-analysis of clinical SC PK data/parameters across multiple antibodies has been limited to a few antibodies¹; none of the analyses have evaluated the impact of hyaluronidase.

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WHAT QUESTION DID THIS STUDY ADDRESS?

Are the SC-relevant PK parameters of clinical antibodies sufficiently generalizable across various antibody and target systems, and is there a difference when co-administered with rHuPH20?

WHAT DOES THIS STUDY ADD TO OUR KNOWLEDGE?

This study demonstrates that when rHuPH20 is used to facilitate SC administration of antibodies in human, the SC PK parameters of absorption rate and bioavailability are generally consistent. Additionally, a target-engagement kinetic model can potentially serve as a universally applicable framework to accurately model the PK of most antibody-target systems. Finally, this is the first time all clinical antibodies co-administered with rHuPH20 have been collated and comprehensively analyzed.

HOW MIGHT THIS CHANGE CLINICAL PHARMACOLOGY OR TRANSLATIONAL SCIENCE?

This work suggests that aggregated clinical data can sufficiently and reliably be used to establish initial PK model parameterization for a new antibody delivered SC in human, and that investment in the development of animal models for PK translation/prediction may not be needed. The data presented herein can be used as a foundational reference for study design and more refined modeling of SC antibodies, particularly those administered with hyaluronidase.

INTRODUCTION

The number of antibodies approved for SC delivery has increased in recent years.^{2,3} The increase in SC adoption can be attributed to its numerous potential benefits over IV administration. For the healthcare system, these include simplified administration, lower costs, reduced provider time, and potentially increased patient throughput associated with shorter patient chair time. For patients, it can reduce the treatment burden and improve quality of life via reduced administration time, and it may allow for self- or caregiver administration in a setting other than an infusion center, for example, home or local doctor's office.^{4,5} Also, SC administration may confer PK/PD benefits compared to IV; the flatter PK profile can be advantageous for reducing C_{\max} -driven toxicities and/or improving C_{\min} -driven efficacy.

In this light, characterization of the SC PK profile of new antibodies is often desirable before clinical development to evaluate various dosing regimens. However, predicting SC PK of antibodies in human from preclinical data is challenging, as there are significant differences in the SC absorption rate and bioavailability across species, none of which reproducibly correlate with what is observed in humans.^{6,7} This lack of fundamental and mechanistic understanding of SC absorption has been broadly recognized by the industry as a necessary problem to solve to enable more efficient clinical development of SC therapies.²

Such limitations extend to modeling SC PK. Modeling the IV PK of antibodies is straightforward, with well-established frameworks for capturing linear and nonlinear PK profiles.⁸ While expressions for SC absorption and bioavailability can be incorporated into these modeling frameworks, the utility is limited to a posteriori characterization rather than a priori prediction.

ENHANZE[®] drug delivery technology is based on the proprietary recombinant human hyaluronidase PH20 enzyme (rHuPH20; Halozyme Therapeutics, Inc.) that facilitates the SC delivery of co-administered therapeutics. rHuPH20 degrades the glycosaminoglycan hyaluronan, which plays a role in resistance to bulk fluid flow in the SC space, limiting large-volume SC drug delivery, dispersion, and absorption. Co-administration with rHuPH20 can overcome administration time and volume barriers associated with SC formulations, especially for high-dose, high-volume therapeutics.⁹

rHuPH20 is approved in several commercial SC products, five of which are monoclonal antibodies: daratumumab (Darzalex FASPRO[®]/DARZALEX[®] SC), trastuzumab (Herceptin Hylecta[™]/Herceptin[®] SC), pertuzumab/trastuzumab (Phesgo[®]), and rituximab (Rituxan Hycela[®]/MabThera[®]) in the United States, and atezolizumab (Tecentriq[®] SC) in Great Britain. SC PK data from these programs, as well as several other antibodies in clinical development with rHuPH20, were leveraged in this work.^{10–18} Other drug modalities commercialized with rHuPH20, that is, polyclonal IgG

(HyQvia[®]) and efgartigimod alfa (antibody fragment) (VYVGART[®] HYTRULO), were not included in the analysis.

The objective of this work is to characterize the human SC PK of antibodies, specifically when co-administered with rHuPH20. Specifically, we demonstrate how an aggregated set of clinical data for various antibodies can be leveraged in a straightforward, universal framework for modeling SC antibody PK. The parameterization from such an approach can provide a reliable starting point for predicting the human SC PK of new antibodies and enable more streamlined clinical development of such therapies.

METHODS

Data collection

The data used in this analysis consisted of 10 monoclonal antibodies that were co-administered SC with rHuPH20 in human clinical studies. All 10 antibodies have data from IV administration and SC administration with rHuPH20, and five antibodies additionally have data from SC administration without rHuPH20. All data are publicly available and are referenced in [Table 1](#) and depicted in [Figure 2](#).

Model definition

PK models have been previously developed and published for most of the antibodies in this analysis (see data references in [Table 1](#)). These models leveraged a two-compartment model structure for linear PK, and, when appropriate, accounted for nonlinear kinetics by introducing saturable clearance expressions, for example, Michaelis-Menten, to the central compartment. While the same linear compartmental structure was used in this analysis, the nonlinear approach was replaced with a more mechanistic, streamlined format.¹⁹ Specifically, PK nonlinearity, often characterized as target-mediated drug disposition, was accounted for via incorporation of antibody-target binding kinetics and an antibody-target complex elimination pathway; such an approach allows for the antibody affinity (i.e., equilibrium binding constant, K_d), intrinsic target (steady-state) concentration, and target kinetics (i.e., target half-life) to drive the total (free and bound) antibody PK with one universally applicable framework. If the target is not expressed at significant levels or its turnover is of long enough duration, its impact on the antibody PK will be negligible, with the kinetics following that of a two-compartment model. If the target is highly expressed and/or has a short half-life, this will introduce an

additional saturable, nonlinear clearance pathway for the antibody that provides a similar kinetic influence as introducing a Michaelis-Menten expression. The additional benefit of leveraging such a target-binding mechanism is that the associated parameters can be defined a priori based on literature data, as described below. A schematic of the model structure, differential equations, and description of the parameters are in [Figure 1](#).

Initial parameter estimates

Initial estimates for the two-compartment parameters, that is, V_1 and V_2 (volumes of the central (plasma) and peripheral (tissue interstitial) compartments, respectively), k_{el} (central compartment elimination rate), k_{12} and k_{21} (distribution rates for central→peripheral and peripheral→central compartments, respectively), and F (bioavailability for SC administration without rHuPH20), were defined from the previously published PK models for each of the respective antibodies. For antibodies where published data were not available, generalized population PK parameters derived from Dirks et al.²⁰ were used.

For target and antibody-target kinetic parameters, that is, T_{ss} (target steady-state concentration), k_{syn} and k_{deg} (target synthesis and degradation rates, respectively), k_{on} and k_{off} (antibody association and dissociation rates, respectively), and k_{deg2} (antibody-target complex degradation/elimination rate), data from published literature were used. Once defined, these parameters were fixed and not considered for subsequent optimization.

Since two types of targets exist, soluble or cell-surface, slight differences in the parameterization were considered for each. For soluble targets, steady-state concentrations were readily measurable, and clearance for the antibody-target complex was assumed to follow that of the antibody, that is, $k_{deg2} = k_{el}$. For cell-surface targets, steady-state concentrations (in units of nM) were derived from receptor (target) density (X_R , in units of receptors/cell) and cell density (X_C , in units of cells/L) according to: $T_{ss} = X_R \cdot X_C / N_A \cdot 10^9$, where N_A is Avogadro's Number (i.e., $6.02e23$), and clearance for the antibody-target complex was assumed to follow that of the target, that is, $k_{deg2} = k_{deg}$. Once defined, target concentrations were fixed and not considered for subsequent optimization. All data sources and derivations are described in [Table 1](#).

Parameter fitting

For each antibody, parameter fitting was conducted for the collective datasets (i.e., all dose levels and routes of administration) as a Population PK (PopPK) assessment.

TABLE 1 Data sources, parameter values, and derivations for each antibody.

Parameter	Description	Units	Trastuzumab	Daratumumab	Rituximab
—	Target	—	HER2	CD38	CD20
—	Target Location	—	Cell	Cell	Cell
—	IV PK Data	—	[16]	[24]	[16,25]
—	SC PK Data with rHuPH20	—	[16]	[17]	[16]
—	SC PK Data without rHuPH20	—	—	—	—
V_1	Central Volume	L	2.20 [F]	3.10 [20]	2.50 [F]
V_2	Peripheral Volume	L	3.06 [34]	2.80 [20]	3.64 [35]
CL	Central Clearance	L/day	0.11 [34]	0.31 [20]	0.10 [F]
Q	Distributional Clearance	L/day	0.45 [34]	0.79 [20]	0.66 [35]
k_{el}	Central Elimination Rate	1/day	0.05 [C]	0.10 [C]	0.04 [C]
k_{12}	Central-Peripheral Rate	1/day	0.20 [C]	0.25 [C]	0.26 [C]
k_{21}	Peripheral-Central Rate	1/day	0.15 [C]	0.28 [C]	0.18 [C]
k_a	SC Absorption Rate	1/day	—	—	—
k_{aPH20}	SC Absorption Rate with PH20	1/day	0.40 [34]	0.28 [38]	0.37 [39]
F	SC Bioavailability	Fraction	—	—	—
F_{PH20}	SC with PH20 Bioavailability	Fraction	0.77 [34]	0.69 [38]	0.63 [39]
T_{ss}	Target Concentration	nM	7.58 [C]	0.17 [C]	1.67 [C]
t_{half}	Target Half-Life	day	0.42 [44]	0.13 [45]	0.17 [46]
K_d	Antibody Affinity	nM	5.00 [52]	4.40 [53]	8.00 [54]
k_{syn}	Target Synthesis Rate	nM/day	12.5 [C]	0.89 [C]	6.80 [C]
k_{deg}	Target Degradation Rate	1/day	1.65 [C]	5.33 [C]	4.08 [C]
k_{on}	Antibody-Target On Rate	1/nM·day	86.4 ^c	86.4 ^c	86.4 ^c
k_{off}	Antibody-Target Off Rate	1/day	432 [C]	380 [C]	691 [C]
k_{deg2}	Antibody-Target Degradation Rate	1/day	1.65 [C]	5.33 [C]	4.08 [C]
X_R	Receptor Density	Receptors/cell	1.00e6 [61]	1.00e5 [62]	1.00e5 [63]
X_C	Cell Density	Cells/L	4.55e9 [65]	1.00e9 ^d	1.00e10 [66]

Abbreviations: —, not applicable/available; [C], calculated, according to the following: k_{el} , CL/V_1 ; k_{12} , Q/V_1 ; k_{21} , Q/V_2 ; $T_{ss} = X_R \cdot X_C / N_A \cdot 10^9$; k_{syn} , $T \cdot k_{deg}$; k_{deg} , $\ln(2)/t_{1/2}$; k_{off} , $K_d \cdot k_{on}$; k_{deg2} , k_{deg} (cell target) or k_{el} (soluble target); [F]=fit from data; a, treated as one target; b, treated as soluble; c, assumed diffusion-limited rate of 10^6 1/M·s; d, assumed value.

Covariates were not incorporated, as the average representation of the respective population is sufficient for this application. Initial simulations were conducted using the initial parameter estimates for the respective antibody and target to assess goodness of fit across the collective datasets, that is, all concentration versus time profiles. The default preference was to keep the previously defined, published parameters (i.e., clearances and volumes) unchanged, but for some antibodies re-fitting was necessary. If the simulated curves deviated from the observed data by more than 10% for a given timepoint, then parameter optimization was conducted on the necessary parameters only. For example, if the initial parameters accurately captured the IV curves but not the SC curves, then only the SC relevant parameters (e.g., F and k_{abs}) were re-defined using a least-sum-of-squares optimization routine. In Table 1, the source of each parameter for each antibody is specified. All simulations

were performed using MATLAB® (Mathworks, Natick, MA).

RESULTS

Model fitting

Figure 2 shows the ability of the model-simulated PK profiles to reproduce the measured clinical data for each antibody and route of administration. Model parameterization of IV data derived from previously published models proved an appropriate starting point for most antibodies, with no further modification to these two-compartment parameters (i.e., V_1 , V_2 , CL, Q, k_a , F) upon introduction of SC data with rHuPH20. For some antibodies, the new data did require slight changes to one or more of these parameters to enable cohesive fitting of curves from all routes

Atezolizumab	Pertuzumab	Crenezumab	Tocilizumab	Bococizumab	Adalimumab	Amivantamab
PD-L1	HER2	A β 40	(s)IL6R	PCSK9	TNF α	EGFR-MET ^a
Soluble	Cell	Soluble	Cell, Soluble ^b	Soluble	Soluble	Cell
[26]	[18]	[27]	[28]	[29]	[30]	[31]
[13]	[18]	[11]	[15]	[10]	[12]	[14]
—	—	[11]	[15,28]	[10]	[12,32]	[14]
3.28 [26]	2.77 [18]	2.90 [11]	3.50 [F]	2.75 [33]	3.10 [20]	2.00 [F]
3.63 [26]	2.49 [18]	1.60 [11]	2.80 [20]	3.02 [33]	2.80 [20]	2.80 [20]
0.20 [26]	0.16 [18]	0.18 [11]	0.40 [F]	0.30 [F]	0.31 [20]	0.20 [F]
0.55 [26]	0.62 [18]	0.15 [11]	0.79 [20]	0.28 [33]	0.79 [20]	0.79 [20]
0.06 [C]	0.06 [C]	0.06 [C]	0.11 [C]	0.11 [C]	0.10 [C]	0.10 [C]
0.17 [C]	0.22 [C]	0.05 [C]	0.23 [C]	0.10 [C]	0.25 [C]	0.40 [C]
0.15 [C]	0.25 [C]	0.09 [C]	0.28 [C]	0.09 [C]	0.28 [C]	0.28 [C]
—	—	0.35 [F]	0.23 [36]	0.25 [33]	0.26 [37]	0.25 [F]
0.27 [13]	0.35 [18]	0.35 [F]	0.40 [F]	0.40 [F]	0.30 [F]	0.40 [F]
—	—	0.80 [F]	0.80 [36]	0.30 [F]	0.70 [F]	0.62 [F]
0.77 [13]	0.71 [18]	0.80 [F]	0.90 [F]	0.33 [F]	0.85 [F]	0.82 [F]
0.002 [40]	6.02 [C]	0.06 [C]	0.87 [41]	6.76 [42]	0.0003 [43]	0.83 [C]
0.67 [47]	0.42 [44]	0.13 [48]	0.08 [49]	0.10 [33]	0.02 [50]	0.17 [51]
0.30 [55]	0.80 [56]	4.00 [57]	1.34 [58]	0.01 [33]	0.10 [59]	0.70 [60]
0.00 [C]	9.93 [C]	0.35 [C]	7.20 [C]	46.9 [C]	0.01 [C]	3.47 [C]
1.04 [C]	1.65 [C]	5.55 [C]	8.32 [C]	6.93 [C]	33.3 [C]	4.16 [C]
86.4 ^c	86.4 ^c	86.4 ^c	86.4 ^c	86.4 ^c	86.4 ^c	86.4 ^c
25.9 [C]	69.1 [C]	346 [C]	116 [C]	0.86 [C]	8.64 [C]	60.5 [C]
0.06 [C]	1.65 [C]	0.06 [C]	8.32 [C]	0.11 [C]	0.10 [C]	4.16 [C]
—	1.00e6 [61]	—	—	—	—	5.00e5 [64]
—	3.61e9 [65]	—	—	—	—	1.00e9 [67,68]

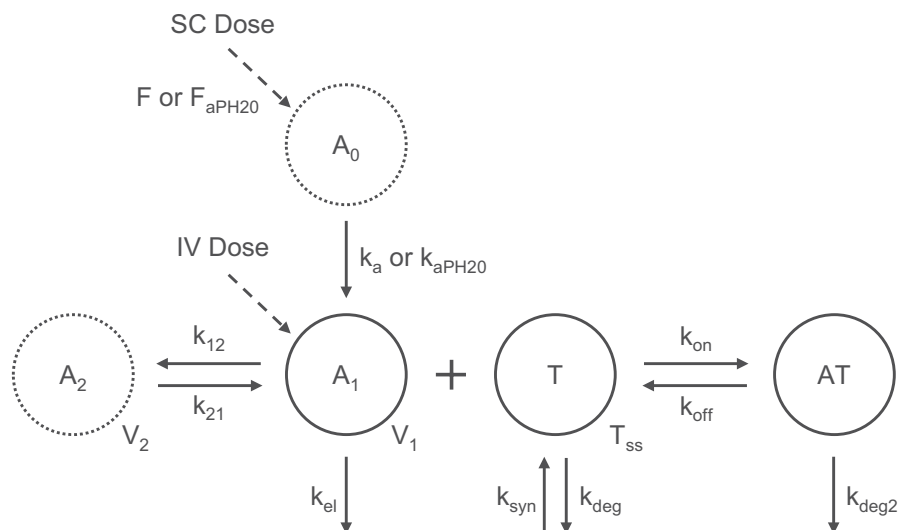
of administration (see Table 1). The results highlight that the universal, pre-defined framework presented herein can capture linear and nonlinear kinetics across various antibodies and targets. Furthermore, the extent of target-mediated drug disposition can be defined a priori, that is, independent of the PK data, based on target expression and half-life, and antibody affinity and dose.

Parameter values

Figure 3 and Table 1 outline the specific parameter values for each antibody. The volumes of distribution (V_1 , V_2) demonstrate high degrees of consistency across antibodies (~20% coefficient of variation (CV)), while central and distributional clearances (CL, Q) have higher degrees of variability across antibodies (~40% CV). Absorption rates without and with rHuPH20 are highly consistent across

antibodies (<20% CV). The absorption rate with rHuPH20 is approximately 30% higher than without, which is congruent with the mechanism of rHuPH20, wherein its degradation of hyaluronan reduces the SC tissue back-pressure and accommodates larger volumes at the injection site.⁹ Bioavailability without and with rHuPH20 is consistent across antibodies (~12% CV), except for the one outlier which was not included in the average calculations. Bioavailability with rHuPH20 demonstrated improvements ranging from 0% to 20% (absolute, i.e., bioavailability % as it relates to IV) relative to values without rHuPH20, for the five antibodies for which data were available for both administrations.

A large range was observed across parameters related to target engagement. Target concentrations ranged from sub-pM to nM, target half-lives ranged from minutes to several hours, and antibody affinities ranged from tens of pM to nM.



$$dA_0/dt = -k_{abs} \cdot A_0$$

$$dA_1/dt = k_{abs} \cdot A_0/V_1 - k_{12} \cdot A_1 + k_{21} \cdot A_2 \cdot (V_2/V_1) - k_{el} \cdot A_1 - k_{on} \cdot A_1 \cdot T + k_{off} \cdot AT$$

$$dA_2/dt = k_{12} \cdot A_1 \cdot (V_1/V_2) - k_{21} \cdot A_2$$

$$dT/dt = -k_{on} \cdot A_1 \cdot T + k_{off} \cdot AT + k_{syn} - k_{deg} \cdot T$$

$$dAT/dt = k_{on} \cdot A_1 \cdot T - k_{off} \cdot AT - k_{deg2} \cdot AT$$

FIGURE 1 PK model schematic and differential equations. Model variables: A_0 , A_1 , and A_2 = antibody in the subcutaneous, central (plasma), and peripheral (tissue interstitial) compartments, respectively; T = target; AT = antibody-target complex. Model parameters: V_1 and V_2 = volumes of the central and peripheral compartments, respectively; k_{el} = central compartment elimination rate; k_{12} and k_{21} = distribution rates for central \rightarrow peripheral and peripheral \rightarrow central compartments, respectively; k_a and k_{aPH20} = absorption rates for subcutaneous \rightarrow central compartments, without and with rHuPH20, respectively; F and F_{PH20} = subcutaneous bioavailability, without and with rHuPH20, respectively; T_{ss} = target steady-state concentration; k_{syn} and k_{deg} = target synthesis and degradation rates, respectively; k_{on} and k_{off} = antibody association and dissociation rates, respectively. k_{deg2} = antibody-target complex degradation/elimination rate. Further details, including derivations and units, are in Table 1.

DISCUSSION

Preclinical data have been of limited utility in predicting SC PK of antibodies in human to date. In this work, a different approach was taken by focusing on a universally applicable model to characterize the human PK profiles of 10 clinical antibodies co-administered SC with rHuPH20. The linear PK aspects driven by two-compartment model kinetics were consistent across antibodies in this analysis and aligned with previous works.^{20,21} The nonlinear PK contributions were captured via a mechanistic target engagement framework and a priori-defined values for target and antibody parameters. For antibodies co-administered with rHuPH20, the human SC PK parameters are quite consistent across a range of antibody and target properties. Consequently, the average values in this work can be applied with a high degree of confidence in model simulations for a new antibody prior to entering the clinic to

evaluate potential dosing and schedule options for SC co-administration with rHuPH20.

The SC PK parameterization can be leveraged via a few angles in Phase I as well. If an adaptive design is used, the parameter values can serve as priors. Additionally, when considering the ability of rHuPH20 to accommodate any dose level or volume, Phase I dose escalation could occur exclusively SC, that is, without IV data. As an example, HyQvia[®], a commercial product comprised polyclonal IgG and rHuPH20, is dosed up to 600 mL in a single SC infusion²²; when considering dose escalation of an antibody up to 100 mg/kg, even with a modest formulation of 100 mg/mL, such a volume would be less than 100 mL and easily accommodated. Starting SC clinical trials with a low-concentration, high-volume formulation with rHuPH20 is also an option. Such a path was taken by Darzalex FASPRO[®], wherein the initial SC Phase I studies utilized a 20 mg/mL formulation administered at volumes

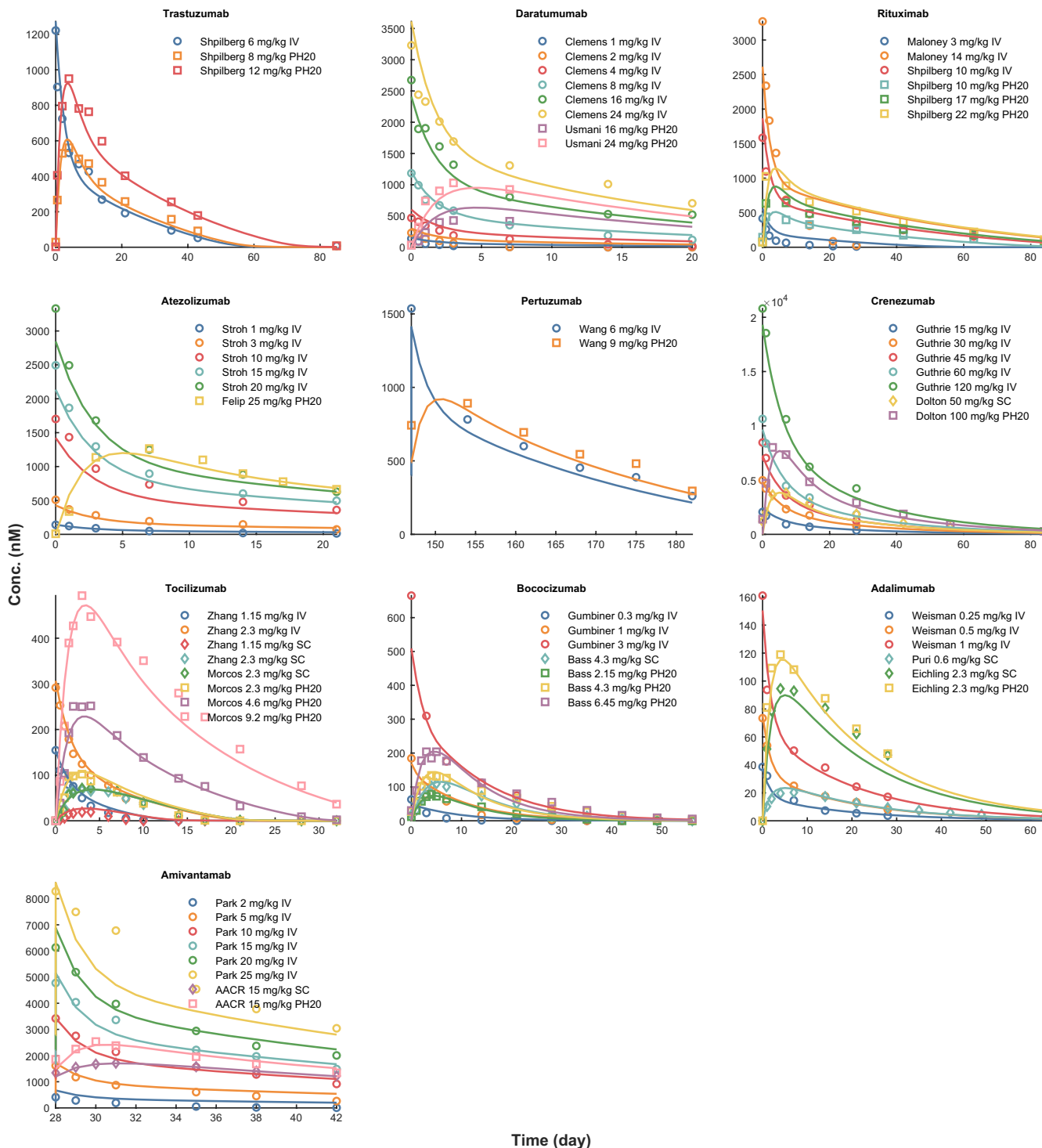


FIGURE 2 Model simulated PK profiles versus measured clinical data. Symbols represent measured mean clinical data, digitized from the respective publications; curves represent model simulations. Routes of administration are distinguished based on the following symbol shapes: ○ = intravenous (IV); ◇ = subcutaneous without rHuPH20 (SC); □ = subcutaneous with rHuPH20 (PH20).

up to 90 mL with rHuPH20.¹⁷ While there has been a precedent to date for IV studies preceding SC, this is not an explicit requirement. In fact, for oral drug delivery, IV PK studies are not always required in humans and for many molecules are never executed.²³ Based on the analysis presented herein, and the new PK data that continues to be

generated for antibodies co-administered with rHuPH20, such an analogous SC clinical development path is now possible.

Despite the high degree of consistency in the PK parameters across antibodies, and the ability to capture/predict PK nonlinearities, there will inevitably be instances where

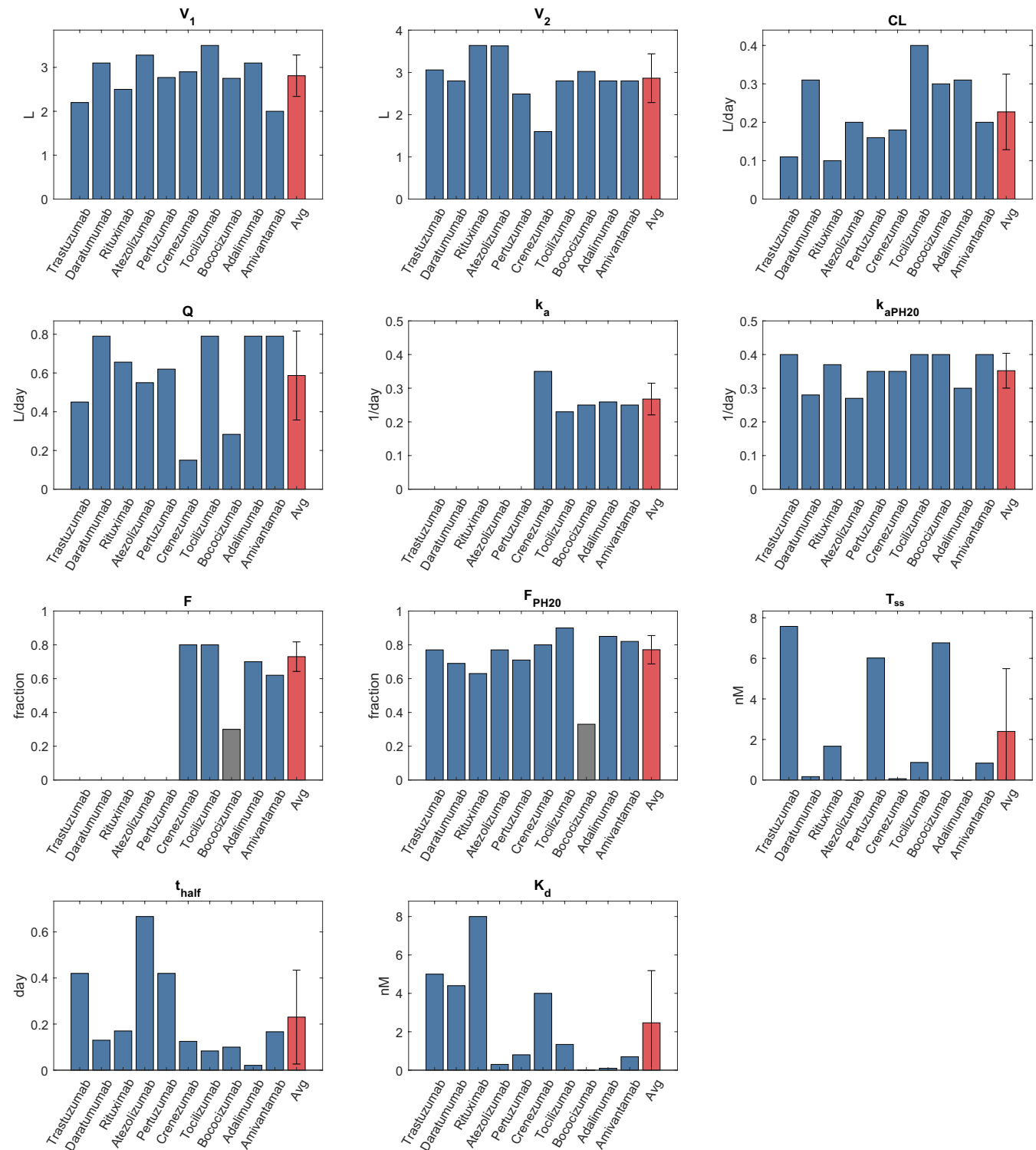


FIGURE 3 Model parameters. Parameter values are listed for the population fit of data for each respective antibody. Red bars designate the average across all antibodies shown, with error bars representing standard deviation. Boccocizumab was excluded from the calculation of the average values for F and F_{PH20} , as it was a significant outlier across the antibodies for these specific parameters.

this methodology does not accurately predict the SC PK, specifically bioavailability, as was observed with the one outlier in this analysis. Therefore, there will continue to be a need to establish correlations with the variables involved

in SC absorption and clinical PK. With more antibodies being delivered SC and the broad adoption of data analytics and artificial intelligence, the probability of accurately predicting bioavailability anomalies will increase over time.

AUTHOR CONTRIBUTIONS

R.P.N. wrote the manuscript. R.P.N. analyzed the data. R.P.N. performed the modeling. R.P.N. and M.A.P. designed the research.

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CONFLICT OF INTEREST STATEMENT

R.P.N. and M.A.P. are employees of Halozyme Therapeutics, Inc. R.P.N. and M.A.P. are shareholders of Halozyme Therapeutics, Inc.

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