1 Supplementary Material

2 Analytical solution for the compartmental model

- 3 When time (t) \leq time to maximal expansion (T_{max}), the compartmental model in **Figure 2** is described by
- 4 the equation $dE/dt = \rho t$, which has the analytical solution $E(t) = R_0 exp(\rho t)$, where as described earlier, R0
- 5 = Cmax/foldx. This gives the exponential growth equation (Eq 1).
- 6 When t > T_{max}, the analysis that links the equations for the compartmental model to the analytical expression
- 7 are provided below, where the rate at which effector cells decline is (α -k), the rate at which effector cells
- 8 become memory cells is k, and the rate at which memory cells decline is β .
- 9 $dE/dt = -(\alpha-k)E kE = -\alpha E$
- $10 dM/dt = \alpha E \beta M$
- 11 Rewritten in matrix notation and grouping terms gives:

$$\frac{d}{dt} \left(\begin{array}{c} E \\ M \end{array} \right) = \left(\begin{array}{c} -\alpha & 0 \\ k & -\beta \end{array} \right) \left(\begin{array}{c} E \\ M \end{array} \right)$$

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14 The eigenvalues of the above system are clearly $\lambda_1 = -\alpha$ and $\lambda_2 = -\beta$, and the associated eigenvectors are 15

$$v_1 = \begin{pmatrix} 1 \\ -k/(\alpha - \beta) \end{pmatrix}, \qquad v_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

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18 Eigenvector v_2 is easy to check, and a check of v_1 is shown below.

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$$\begin{pmatrix} -\alpha & 0 \\ k & -\beta \end{pmatrix} \begin{pmatrix} 1 \\ -k/(\alpha - \beta) \end{pmatrix} = \begin{pmatrix} -\alpha \\ k + \beta k/(\alpha - \beta) \end{pmatrix} = \begin{pmatrix} -\alpha \\ \alpha k/(\alpha - \beta) \end{pmatrix} = -\alpha \begin{pmatrix} 1 \\ -k/(\alpha - \beta) \end{pmatrix}$$

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22 This results in the solution below, where C_1 and C_2 are 2 constants to be defined further below.

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$$\begin{pmatrix} E(t) \\ M(t) \end{pmatrix} = C_1 \begin{pmatrix} 1 \\ -k/(\alpha - \beta) \end{pmatrix} e^{-\alpha(t - \mathsf{T}_{\max})} + C_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix} e^{-\beta(t - \mathsf{T}_{\max})}$$

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26 The measurement variable is:

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$$f(t - T_{\max}) = E(t - T_{\max}) + M(t - T_{\max}) = C_1 \left(1 - \frac{k}{\alpha - \beta}\right) e^{-\alpha t} + C_2 e^{-\beta t}$$

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Also, at time T_{max} , it is assumed that all cells are effector cells; thus, $E(T_{max}) = maximal concentration (C_{max})$ (giving $C_1 = C_{max}$) and $M(T_{max}) = 0$ [giving $C_2 = C_1 k / (\alpha - \beta) = C_{max} k / (\alpha - \beta)$]. Note that in the analytical formula, $C_2 = C_{max} \cdot F_B$. Thus, the differential equation model is mapped to the analytical model:

$$C_2 = C_{\max} \cdot F_B = C_{\max} k / (\alpha - \beta)$$
$$k = F_B \cdot (\alpha - \beta)$$

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34 35 **S**

5 Semi-mechanistic Model Selection

36 The semi-mechanistic model presented here is not the only model that gives rise to exponential 37 growth followed by biexponential decay. If memory cells were produced during the expansion phase as 38 well or if cells could transition back and forth from the effector to memory state, then the analytic solution 39 to such a model would have the same functional form. It is also possible that terminal phase is not due to 40 formation of memory cells, but rather the low level of constant proliferation in response to the continual 41 production of CD19 antigen; the kinetics of this process would look similar. Furthermore, DeBoer and 42 Perelson⁴ also showed in equations 8-10 of their paper that rather than have expansion stop at Tmax, 43 one could use a "cascade" model of cell division, modeling each generation of expansion complex model 44 and realistic model with successive generations of expansion. The cascade model could be approximated 45 by the simpler model used here; the advantage of the model used here is it is simpler to implement and to 46 present.

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48 Derivation of AUC to approximate C_{max}

In the scenario where no comedications are given, the area under the curve (AUC) from time 0 to time T_{end} is given by the equation below, where $R = C_{max}/fold_x$, $A = (1 - F_B) \cdot C_{max}$, and $B = F_B \cdot C_{max}$. The final equation below shows that in this model formulation, AUC is proportional to C_{max} ; thus, covariates that affect C_{max} also impact the AUC. In the derivation below, we switch from an equality to an approximation because 53 with the parameters fit here, $exp(\rho \cdot Tmax) = foldx = 3900 >> 1$, such that $[exp(\rho \cdot Tmax)-1]/exp(\rho \cdot Tmax) \approx$

$$\begin{aligned} AUC_{0-T_{\text{end}}} &= \int_{0}^{T_{\text{max}}} R \exp(\rho t) dt + \int_{T_{\text{max}}}^{T_{\text{end}}} A \exp(-\alpha (t - T_{\text{max}})) + B \exp(-\beta (t - T_{\text{end}})) dt \\ &= \left[\frac{R}{\rho} \exp(\rho t)\right]_{0}^{T_{\text{max}}} + \left[-\frac{A}{\alpha} \exp(-\alpha (t - T_{\text{max}})) - \frac{B}{\beta} \exp(-\beta (t - T_{\text{max}}))\right]_{T_{\text{max}}}^{T_{\text{end}}} \\ &= \left[\frac{R}{\rho} \left(\exp(\rho \cdot T_{\text{max}}) - 1\right)\right] + \left[\frac{A}{\alpha} \left(1 - \exp(-\alpha (T_{\text{end}} - T_{\text{max}}))\right) + \frac{B}{\beta} \left(1 - \exp(-\beta (T_{\text{end}} - T_{\text{max}}))\right)\right] \end{aligned}$$

substituting in the equations for R, A, and B gives

 $AUC_{0-\infty} \approx C_{\max} \left(\frac{1}{\rho} + \frac{1-F_B}{\alpha} + \frac{F_B}{\beta}\right)$

$$\begin{split} &= \left[\frac{C_{\max}}{\rho}\frac{\left(\exp(\rho \cdot T_{\max}) - 1\right)}{\exp(\rho \cdot T_{\max})}\right] \\ &+ \left[\frac{\left(1 - F_B\right) \cdot C_{\max}}{\alpha}\left(1 - \exp(-\alpha(T_{end} - T_{\max}))\right) + \frac{F_B \cdot C_{\max}}{\beta}\left(1 - \exp(-\beta(T_{end} - T_{\max}))\right)\right] \\ &= C_{\max} \cdot \left[\frac{1}{\rho}\frac{\left(\exp(\rho \cdot T_{\max}) - 1\right)}{\exp(\rho \cdot T_{\max})} \\ &+ \frac{\left(1 - F_B\right)}{\alpha}\left(1 - \exp(-\alpha(T_{end} - T_{\max}))\right) + \frac{F_B}{\beta}\left(1 - \exp(-\beta(T_{end} - T_{\max}))\right)\right] \\ &\approx C_{\max}\left[\frac{1}{\rho} + \frac{\left(1 - F_B\right)}{\alpha}\left(1 - \exp(-\alpha(T_{end} - T_{\max}))\right) + \frac{F_B}{\beta}\left(1 - \exp(-\beta(T_{end} - T_{\max}))\right)\right] \\ &\text{as } T_{end} \to \infty \end{split}$$

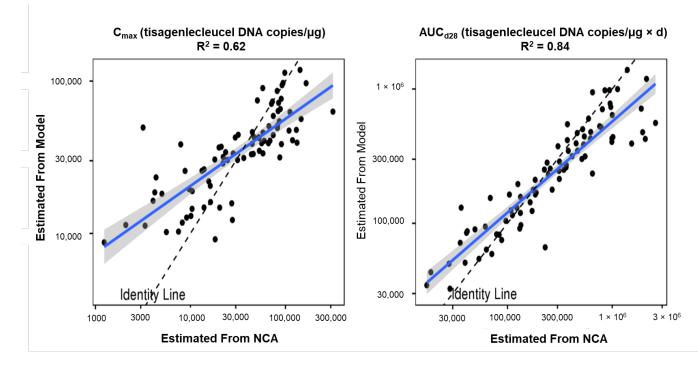
59 Covariate Analysis – Further Details

60 Sex, race, and weight were included because they were demographic parameters that are typically 61 included in a model-based analysis of exposure. Weight-adjusted dose, too, is a factor that for standard 62 drugs correlates with exposure. Down's Syndrome was included because ALL is more prevalent in 63 Down's syndrome patients. Prior lymphodepleting therapy and prior stem cell transplant were included 64 because it's possible that both could have affected the CAR-T cell ability to expand. Per protocol patients 65 received fixed, prespecified weight or surface area-adjusted lymphodepleting (LD) fludarabine and 66 cyclophosphamide chemotherapy dosing. A small percent (listed in ELIANA publication) of patients 67 received non-FC LD chemotherapy or non LD chemotherapy if blood counts were very low. In this 68 subset, 85 patients received fludarabine and cyclophosphamide, 2 received only cyclophosphamide, and 69 3 received neither. 70 For practical reasons T cell numbers were not included as a covariate in this analysis. While the inclusion 71 of T cell numbers as a covariate could alter Cmax, it is unlikely that it would impact the primary 72 conclusions of the effect of tocilizumab and corticosteroids on the rate of expansion. However, this 73 question warrants study in future analyses. If it is possible to collect data on tumor burden or CD19 74 antigen density, this data should be included in assessing the impact on Cmax and if the follow-up is long 75 enough, then β as well. The impact of missing this information in this study is that we may have larger 76 portion of unexplained variability in the exposure. 77 The relationship between C_{max} (C_{max} for patient i), the random effect η_i with variance ω^2 , and the covariate 78 j is given below. Here, x_{ii} denotes the covariate j value for patient i, z_{ij} is a transformation of x_{ij} , and θ_i 79 denotes the size of the covariate effect. 80 81 (4) $C_{maxi} = C_{max} \cdot exp(\eta_i) \cdot \prod_j z_{ij}$ 82 83 For continuous covariates (eg, dose), 84 $z_{ij} = (x_{ij}/x_{median,j})^{\theta}_{ij}$ 85 (5) 86 87 For dichotomous covariates, x_i is assigned either 0 or 1 (eg, $x_{ij} = 0$ for no tocilizumab; $x_{ij} = 1$ for 88 tocilizumab). Then z_{ij} is given by 89 90 (6) $z_{ij} = (\theta_j)^{x_{ij}}$ 91 92 For categorical covariates with n > 2 categories (eq. race) n - 1 dichotomous variables (y_{iik}) are created for each patient i, with k going from 1 to n - 1. Then z_{ij} is given by 93 94 95 (7) $z_{ij} = \prod_k (\theta_{ik})^{y_{ijk}}$ 96

98 99 100 Model MLXTRAN file 101 102 INPUT: 103 parameter = {foldx, Tmax, Cmax, Ftoci, Fster, alpha, FB, beta} 104 regressor = {TOCI1T, STERSTT} 105 106 EQUATION: 107 108 ; Figure out whether tocilizumab or steroids come first 109 ; assign first comedication to t1,F1 and second to t2,F2 110 111 ; If patient does not receive TOCILIZUMAB, then TOCI1T=99999 in dataset 112 ; If patient does not receive STEROIDS, then STERSTT=99999 in dataset 113 if TOCI1T <= STERSTT 114 t1 = min(TOCI1T,Tmax) 115 F1 = Ftoci t2 = min(STERSTT,Tmax) 116 F2 = Fster 117 118 else 119 t1 = min(STERSTT,Tmax) 120 F1 = Fster 121 t2 = min(TOCI1T,Tmax)122 F2 = Ftoci 123 end 124 125 ; Compute constants for solving system 126 ; Note that: foldx = $exp(rho^{tau})$ 127 rho = log(foldx)/Tmax 128 R0 = Cmax/exp(rho*Tmax)129 $R1 = R0^* exp($ rho* t1) 130 $R2 = R1^{*}exp(F1 * rho^{*}(t2 - t1))$ 131 CmaxAdjust = R2*exp(F1*F2*rho*(Tmax -t2)) 132 133 AA = (1-FB)*CmaxAdjust 134 BB = FB*CmaxAdjust 135 136 ; Analytical solution for system 137 if t<0 ylin = R0 138 139 elseif t < t1 140 $ylin = R0^*exp(rho^*t)$ 141 elseif t < t2 142 $ylin = R1^*exp(F1^*rho^*(t-t1))$ 143 elseif t < tmax 144 ylin = R2*exp(F1*F2*rho*(t-t2))145 else 146 ylin = AA*exp(-alpha*(t-Tmax)) + BB*exp(-beta*(t-Tmax)) 147 end 148 149 $\log y = \log(y \ln y)$ 150 151 OUTPUT: 152 output = $\{\log y\}$ 153

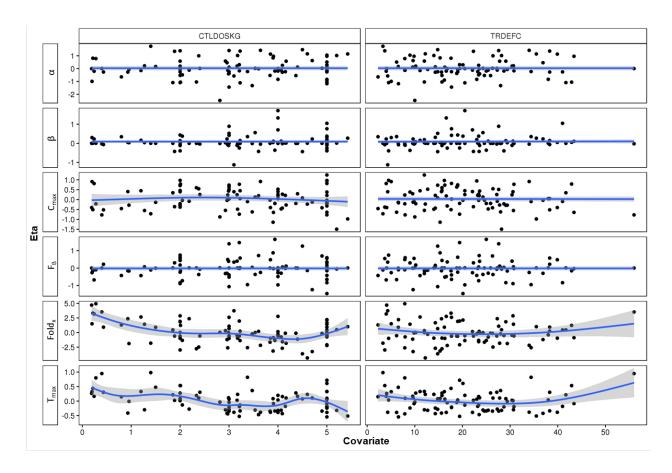
154 Figure S1 Correlation between direct and model-based estimates of maximal concentration (Cmax) and 155 area under the curve at day 28 (AUC_{d28}). The data-driven estimates of C_{max} and AUC_{d28} were calculated 156 in patients who had \geq 21 days of follow-up data. C_{max} was the maximum quantitative polymerase chain 157 reaction value in the first 28 days, and AUC_{d28} was calculated using the trapezoidal rule. The model-158 based estimate for C_{max} was derived directly from the post hoc estimate of the C_{max} parameter. The 159 model-based AUC_{d28} was calculated using the post hoc estimates of all the cellular kinetics and the 160 formula above for AUC_{0-Tend} (where T_{end} = 28). Significant correlation between the noncompartmental 161 analysis (NCA) and model-based estimates was observed. Tend, time of final measurement of transgene 162 copy number. The blue line and shaded area correspond to the linear regression with 90% confidence 163 intervals. The dotted line is the identity line.





167 Figure S2 Exploratory plots for covariate analysis. (a) Continuous covariates plotted against post hoc 168 random-effects estimates (Eta). CTLDOSKG is defined as the dose of tisagenlecleucel per kg. TRDEFC 169 is defined as the transduction efficiency. (b) Categorical covariates plotted against post hoc random-170 effects estimates. Downs denotes Down syndrome, LDFLUD denotes that the patient received 171 fludarabine lymphodepleting therapy, Ster denotes that the patient received corticosteroids, Stud denotes 172 the study (ELIANA, B2202; ENSIGN, B2205J), and Toci denotes that the patient received tocilizumab. 173 C_{max}, maximal concentration; F_B, fraction of transgene copies present during the decline at the gradual 174 rate β , starting from T_{max}; fold_x fold expansion; HSCT, hematopoietic stem cell transplant; T_{max}, time to 175 maximal expansion. The blue line and shaded area shows the loess smoothed line; no obvious trends are 176 observed.

- 177
- 178 (a)
- 179



(b)

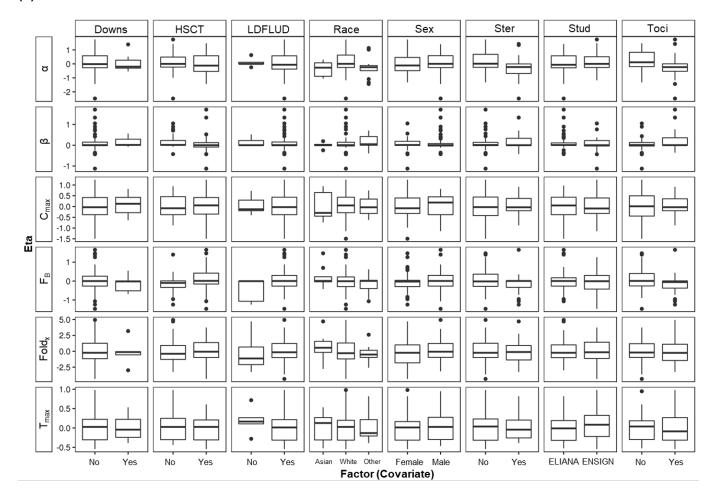


Figure S3 Predictions vs observations. (a) Blue dots are quantifiable values, and red asterisks signify
simulated residuals below the limit of quantification (BLQ). (b) Blue dots are quantifiable values, and red
asterisks signify simulated residuals BLQ. The green solid lines in the pdf plot demonstrate the true
residual distribution, and the black lines that nearly overlie the green lines denote the theoretical
distribution. IWRES, individual weighted residual; NPDE, normalized distribution prediction errors; pdf,
probability distribution function.

189 (a)

