

Boston Society for Cell & Gene Therapy

**Making Drugs from T Cells:  
Model-Informed Design and Deployment of T Cell Therapies**

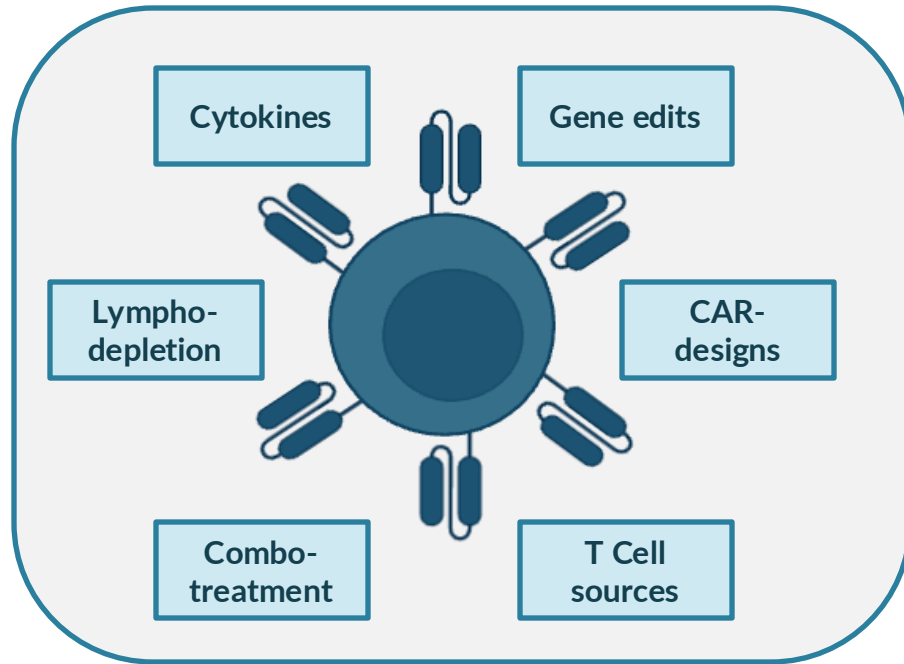
**Daniel Kirouac**  
Quantitative Systems Pharmacology  
Metrum Research Group  
Mar 20, 2025

# What is the value of mathematical modeling?

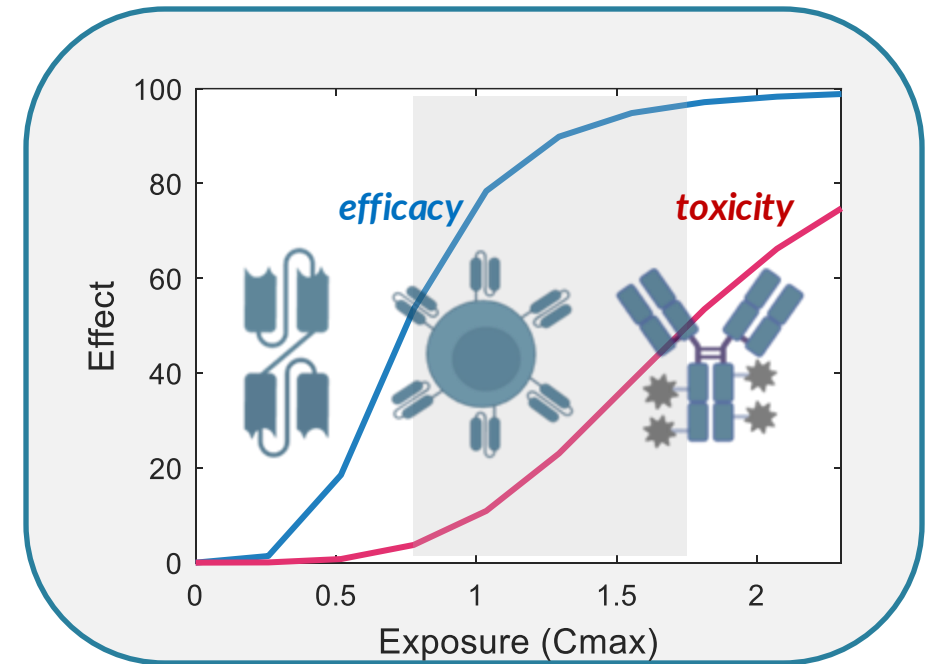
In the context of cell therapy engineering and clinical development

The number of possible experiments to conduct is infinite

## Cell therapy engineering



## Clinical strategy



## Prediction

$$\frac{dT_1}{dt} = 2\mu_{T_1} \cdot \text{Imm} \cdot \left(1 - \frac{P_A^{T_1}}{B_{50}^{T_1} + P_A^{T_1}}\right) \cdot T_1 - r_{T_1} \left(1 - \frac{P_A^{T_1}}{B_{50}^{T_1} + P_A^{T_1}}\right) \cdot T_1 - d_{T_1} \cdot T_1$$

$$\frac{dT_2}{dt} = 2\mu_{T_2} \cdot \left(1 - \text{Imm} \cdot \left(1 - \frac{P_A^{T_2}}{B_{50}^{T_2} + P_A^{T_2}}\right)\right) \cdot T_2 - \mu_{T_2} \cdot \left(\frac{P_A^{T_2}}{B_{50}^{T_2} + P_A^{T_2}}\right) \cdot T_2 - d_{T_2} \cdot T_2$$

$$\frac{dT_3}{dt} = \mu_{T_3} \cdot 2^{N_3} \cdot \left(\frac{P_A^{T_3}}{B_{50}^{T_3} + P_A^{T_3}}\right) \cdot T_3 - \mu_{T_3} \cdot \left(\frac{P_A^{T_3}}{B_{50}^{T_3} + P_A^{T_3}}\right) \cdot T_3 - r_{T_3} \left(1 - \frac{P_A^{T_3}}{B_{50}^{T_3} + P_A^{T_3}}\right) \cdot T_3 - d_{T_3} \cdot T_3$$

$$\frac{dT_X}{dt} = r_{T_X} \cdot \left(\frac{P_A^{T_X}}{B_{50}^{T_X} + P_A^{T_X}}\right) \cdot T_X - d_{T_X} \cdot T_X$$

$$\frac{dB}{dt} = \lambda_B \cdot \left(1 - \frac{B}{B_{max}}\right) - r_{Bd} \cdot \left(\frac{T_{C2}^{Bd}}{T_{K_{50}^{Bd}} + T_{C2}^{Bd}}\right) \cdot B$$

$$\frac{dP_A}{dt} = \lambda_{P_A} \cdot B - \lambda_{P_A} \cdot P_A$$

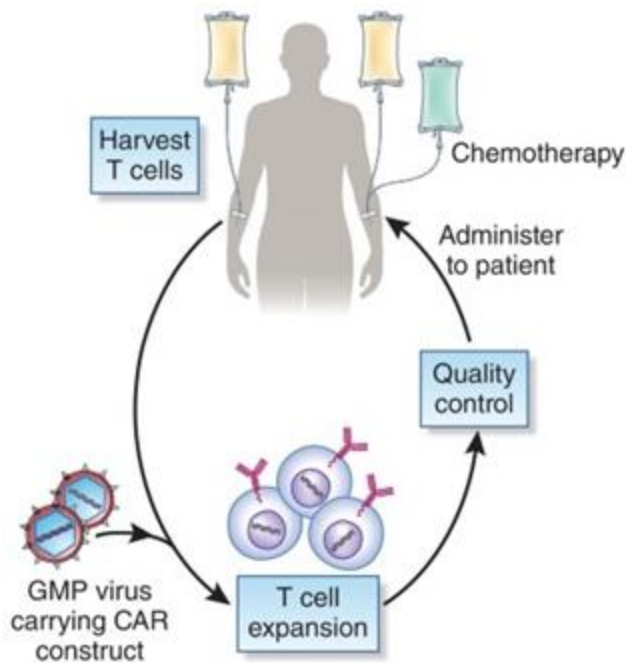
## Inference

The biological mechanisms underlying experimental data are often complex and non-intuitive

# Pharmacology of Autologous T cell therapies is highly variable

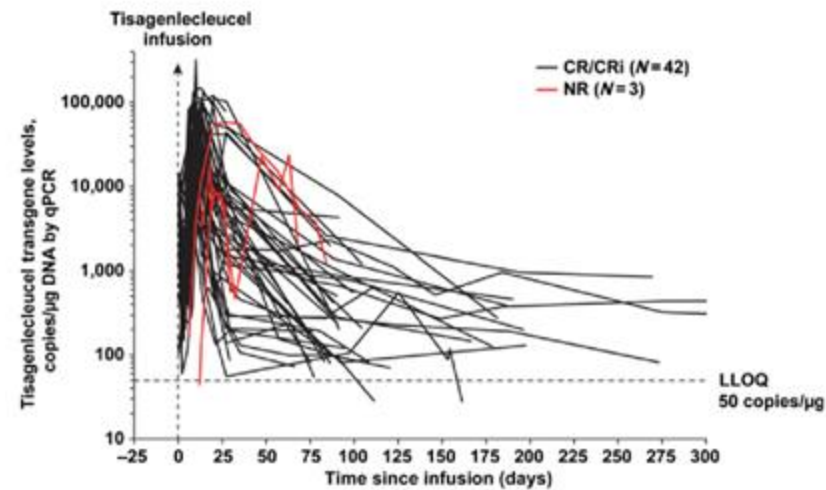
...This is problematic for drug development

## Bespoke manufacturing



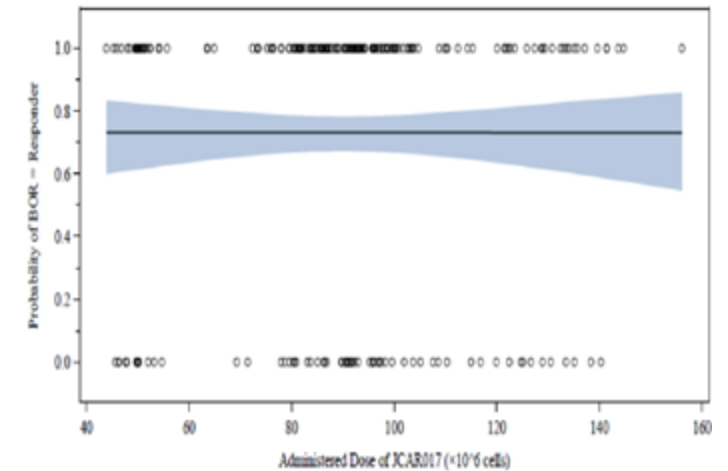
Olewus (2017) Nat Biotech 35. 520-521

## Variable pharmacokinetics



Kymriah (Tisagenlecleucel) in B-ALL  
PMID: 30190371

## Little dose-response

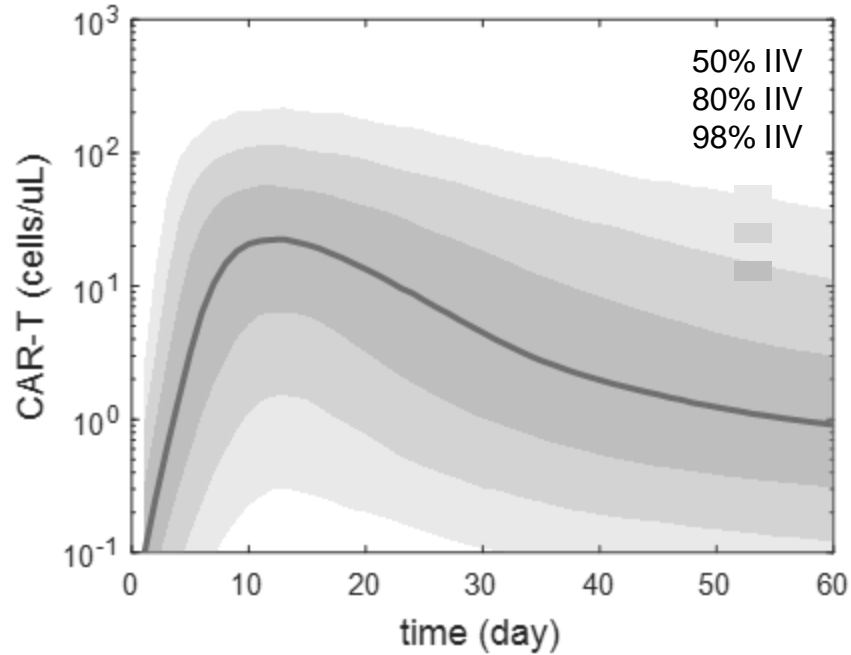


Breyanzi (JCAR017) in NHL.  
BLA: Clinical Pharmacology Review

# Pharmacology of Autologous T cell therapies is highly variable

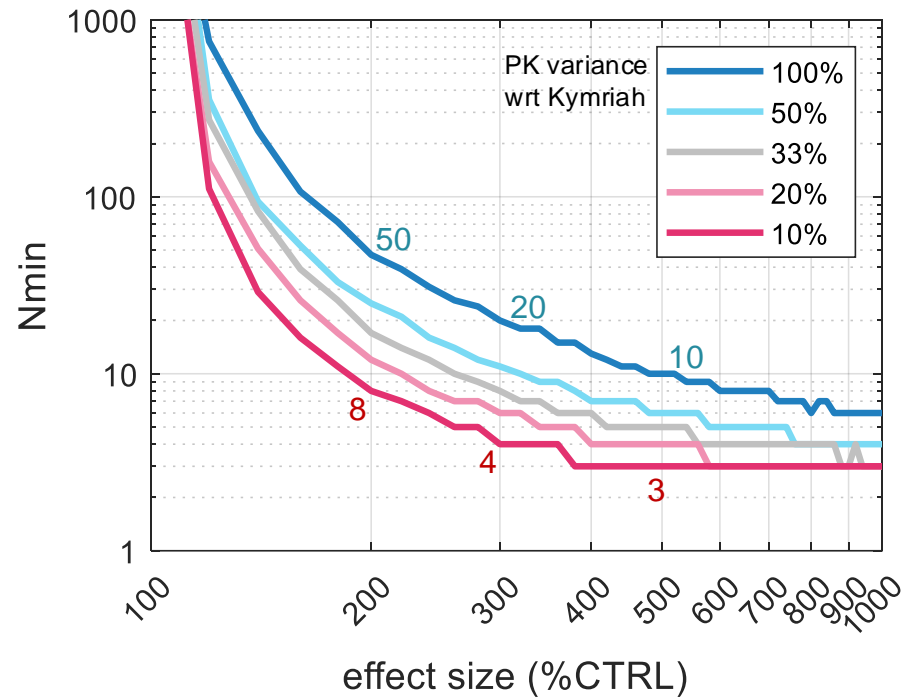
...This is problematic for drug development. E.g. Multi-arm (umbrella) trials

### Kymriah popPK simulations



Population PK model of Kymriah  
Stein et al (2019); PMID: 30848084

### Minimum N vs. Effect-size & variance

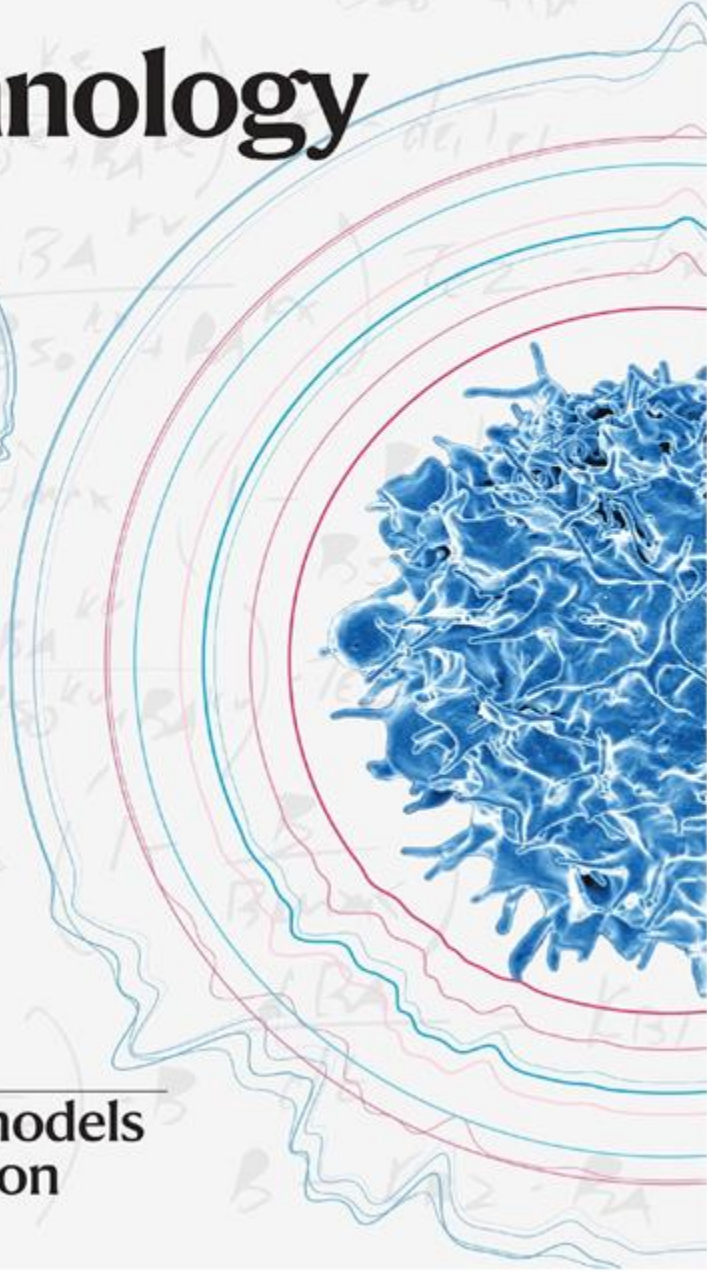
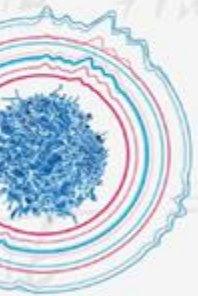
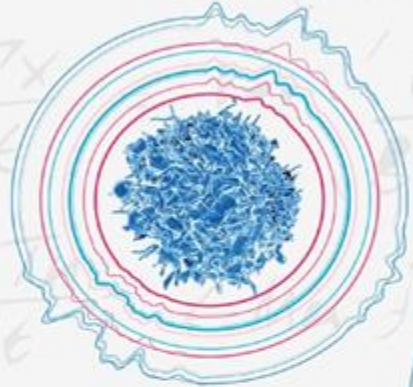


**Studying Multiple Versions of a Cellular or Gene Therapy Product in an Early-Phase Clinical Trial**

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Guidance for Industry FDA, 2022

# nature biotechnology



**Mathematical models  
of T cell regulation**

## CAR-T Pharmacology

**Model based inference:** How do dynamic interactions between CAR-Ts, tumor and patient lymphocytes drive exposure, response, and patient variability?

Kirouac, Zmurchok et al. (2023). Deconvolution of clinical variance in CAR-T pharmacology and response. *Nature Biotechnology* **41**:1606–1617.

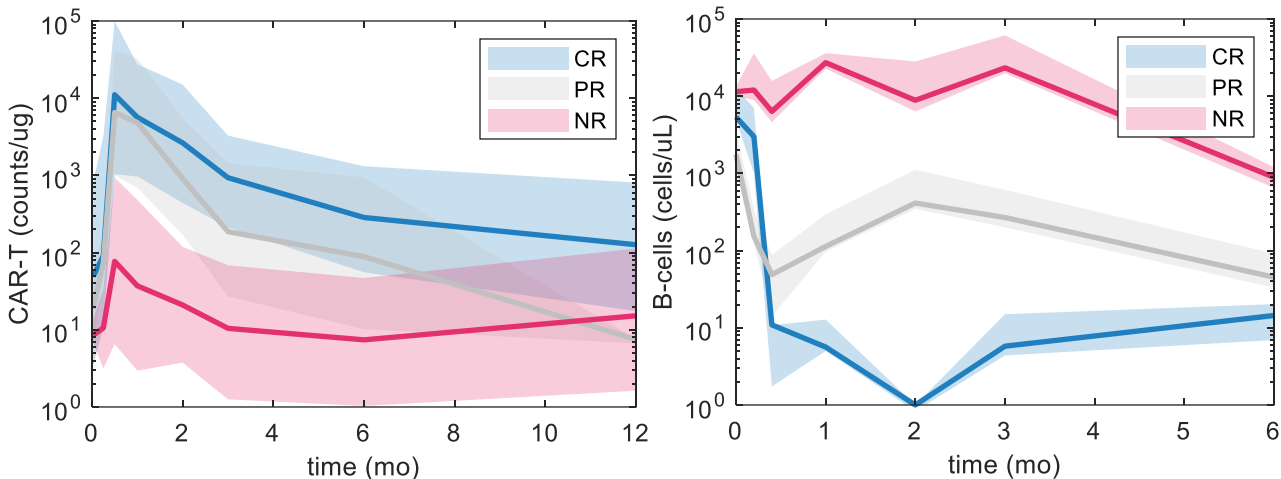
Kirouac, Zmurchok & Morris (2024). Making drugs from T cells. *npg Systems Biology & Applications* **10**: 31.



# Model Training Data

PKPD profiles, CAR-T product transcriptomes and immuno-phenotypes vs. response

## Population mean PKPD: *Kymriah* in Chronic Lymphoblastic Leukemia (CLL)

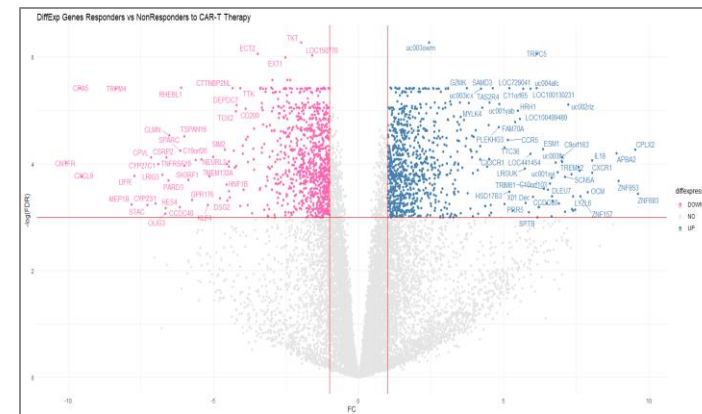


\*mean ± std, digitized from publication  
CR=8, PR =5, NR=25

**CR** = Complete Response  
**PR** = Partial Response  
**NR** = Non-Response

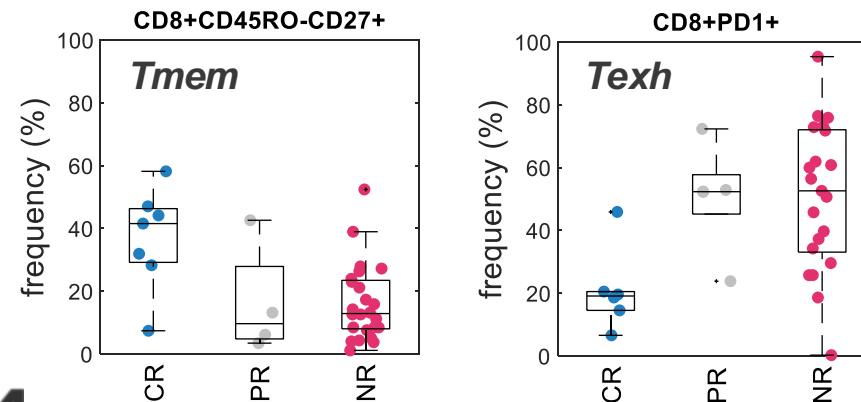
- Can we recapitulate the pharmacokinetics & tumor dynamics (PKPD) based on T cell biology?
- What kinetic parameters / molecular features distinguish robust vs. poor responding patients?

## Pre-infusion CAR-T transcriptomes



CR=5, PR =5, NR=21

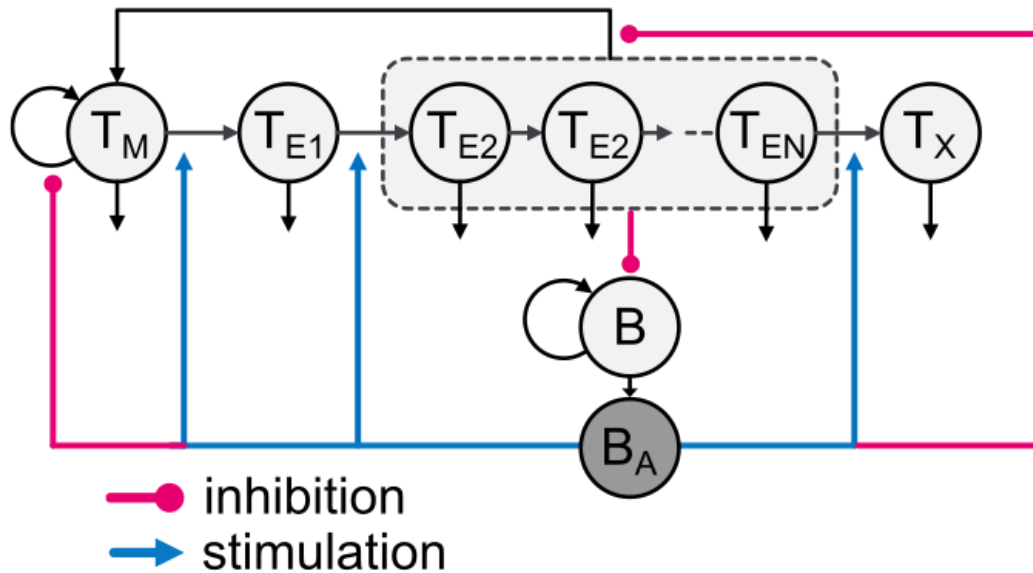
## Pre-infusion CAR-T immunophenotypes



Fraietta JA, Lacey SF, Orlando EJ, et al (2018) Determinants of response and resistance to CD19 chimeric antigen receptor (CAR) T cell therapy of chronic lymphocytic leukemia. Nat Med 24:563–571.



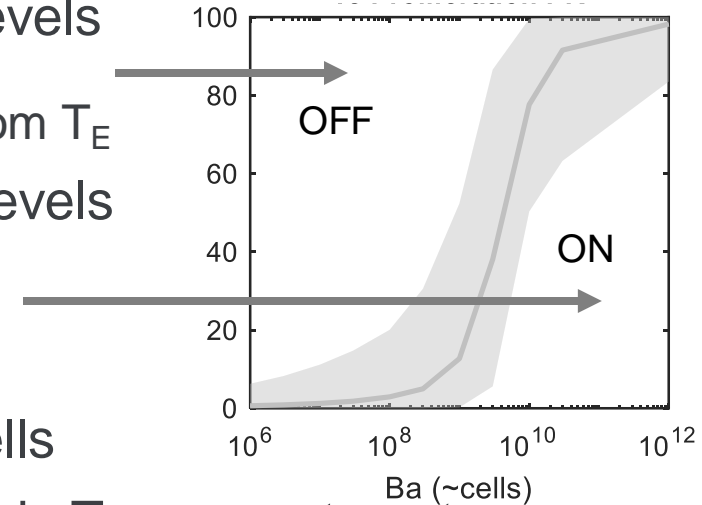
# 'Toggle switch' model structure and assumptions



- $T_M$ : memory T cells
- $T_E$ : effector T cells
- $T_X$ : exhausted T cells
- B: B cells (tumor)
- $B_A$ : B cell antigen

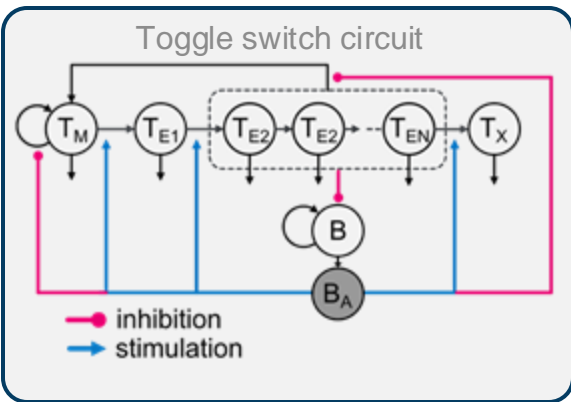
## T cell differentiation *toggle switch*

- Low antigen ( $B_A$ ) levels
  - $T_M$  self-renewal
  - $T_M$  regeneration from  $T_E$
- High antigen ( $B_A$ ) levels
  - $T_M$  differentiation
  - $T_E$  proliferation
  - $T_E$  exhaustion ( $T_X$ )
- T effectors kill B-cells
- N cell divisions within  $T_E$  compartment



# Model development and validation workflow

## Conceptual model of T cell biology



## Mechanism-based dynamical model

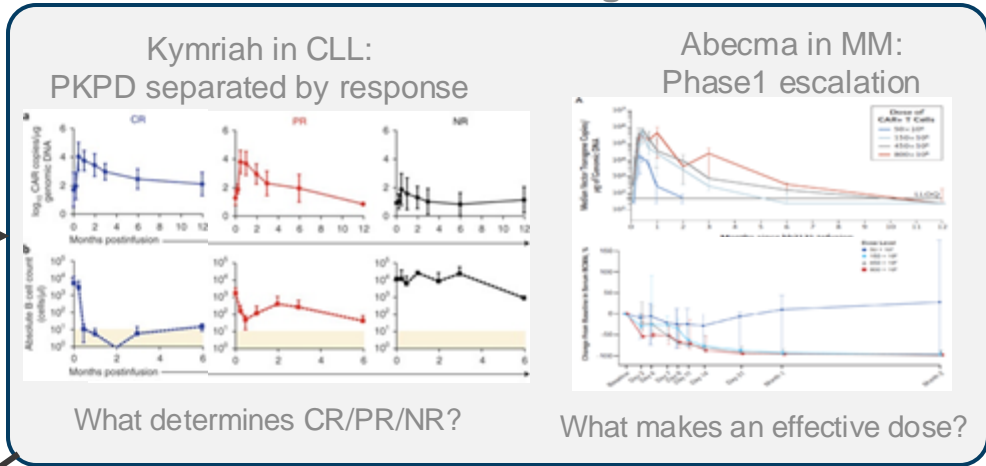
### Math

### Executable code

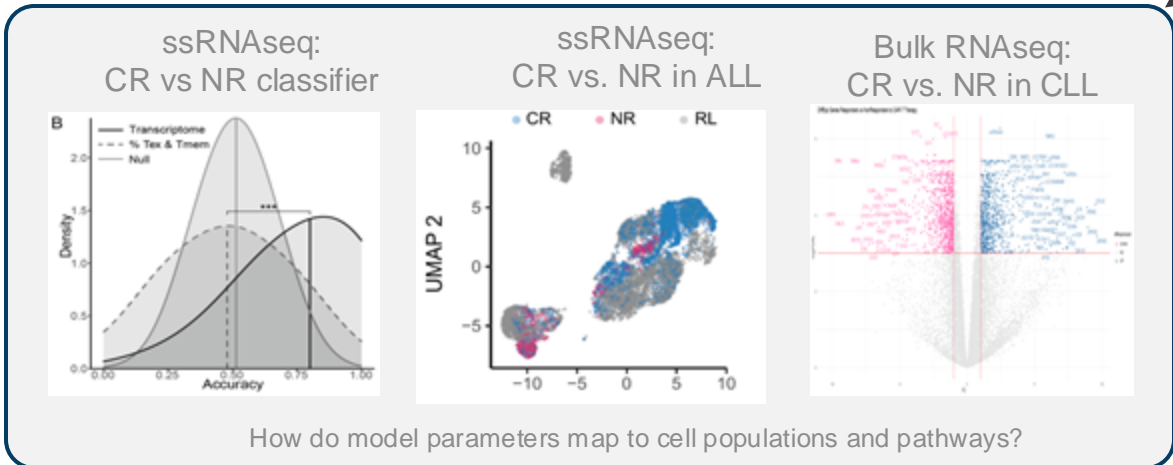
```

// ...
model = ABM()
// ...
// ...
    
```

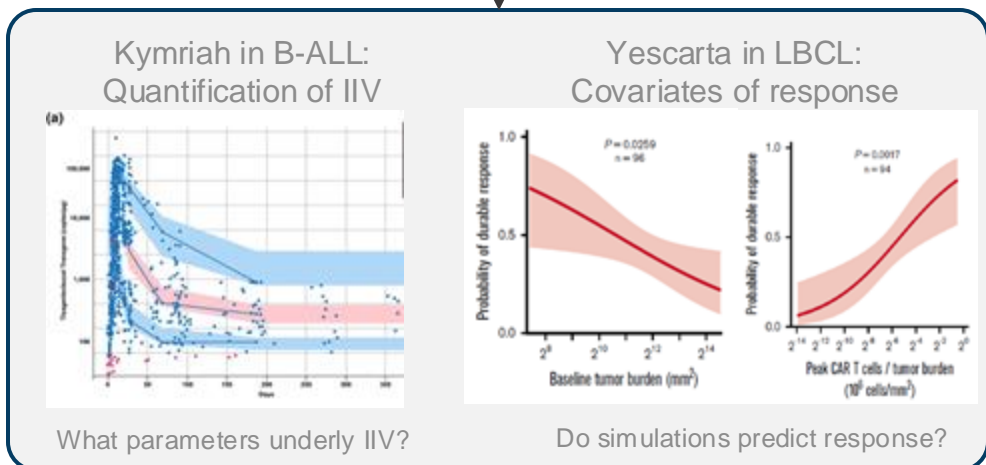
## Clinical Training Data



## Genomic "Validation" Data



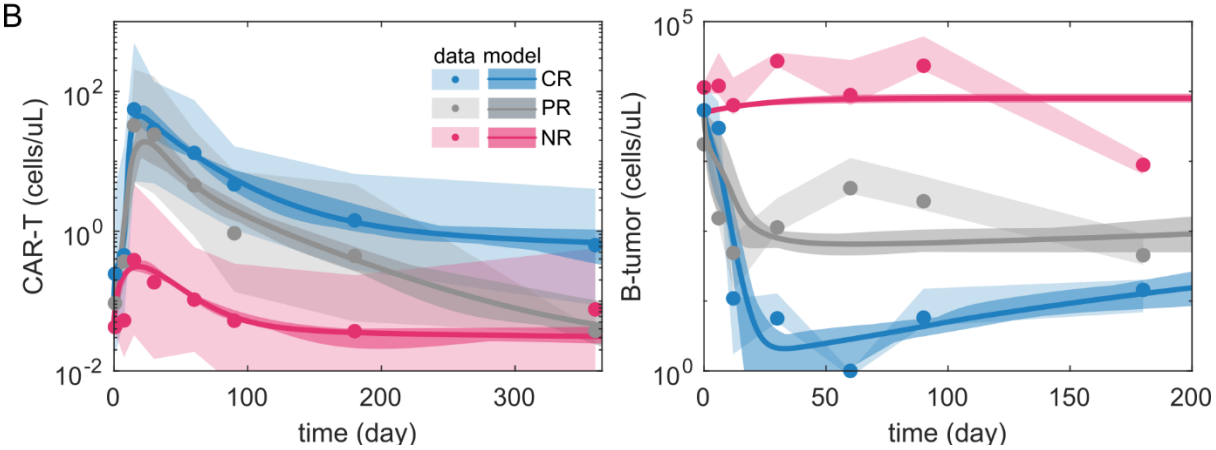
## Clinical Validation Data





# What features (parameters) separate clinical outcomes?

## Model calibration

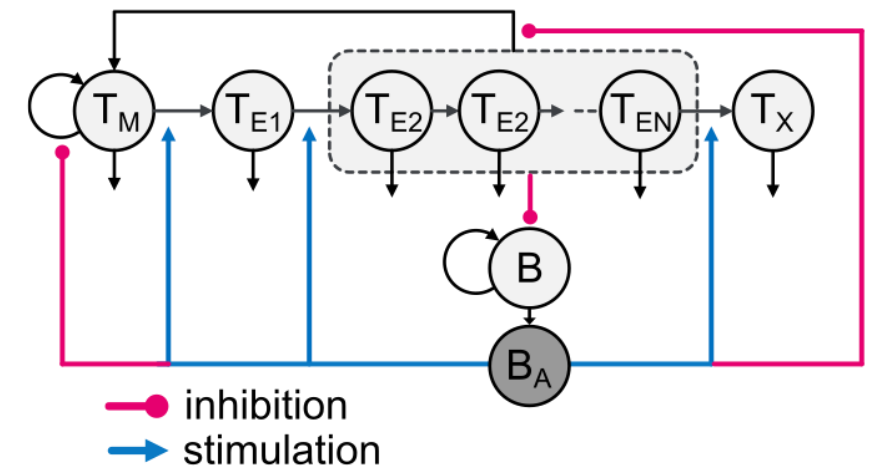
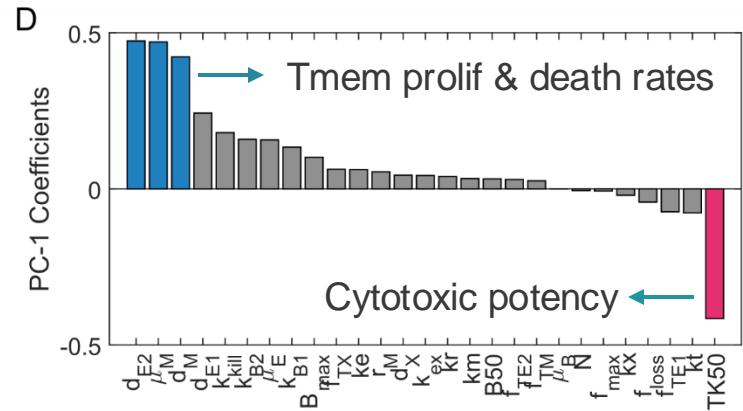
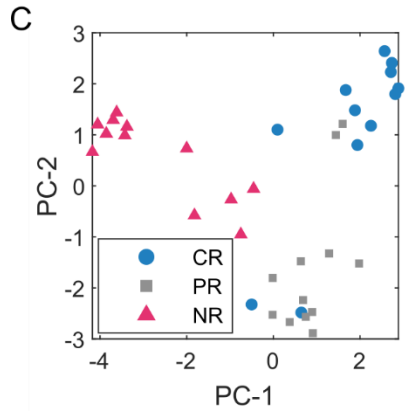


What differentiates responders (CR) vs. non-responders (NR)?

CAR-T products in CR vs. NR show:

1. Heightened memory cell turnover ( $\mu_M, d_M$ )
2. Heightened cytotoxic potency ( $TK50$ )
3. Little difference in Tmem/Texh frequency

## Parameter Analysis



\*Assume Dose =  $10^8$  cells, Tumor burden =  $10^{10}$  cells (median reported); Estimate parameters using PSO: simulations represent 90% confidence intervals

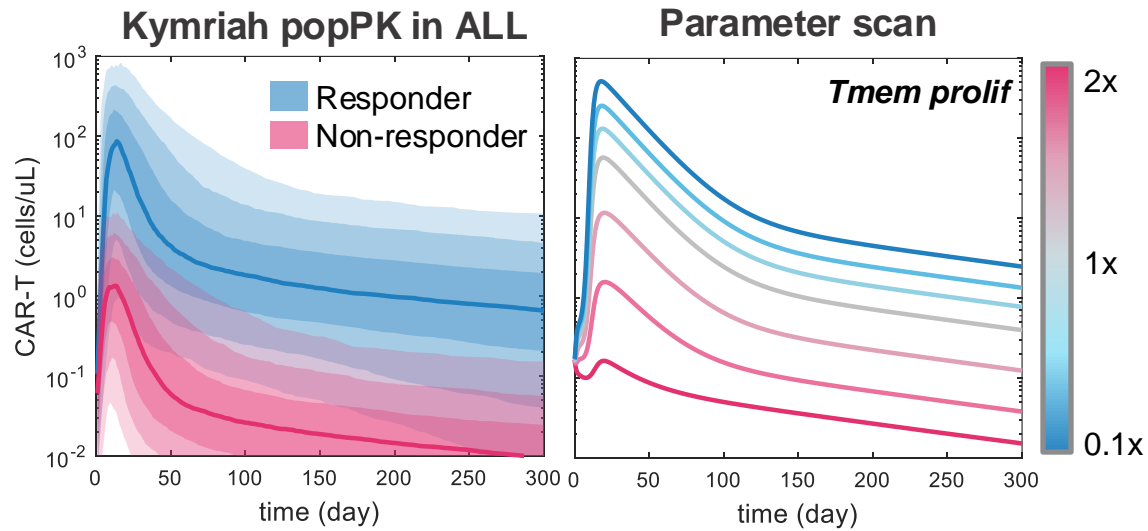
Scale counts/ug to cell/uL using data from: Kalos, M. et al. T Cells with Chimeric Antigen Receptors Have Potent Antitumor Effects and Can Establish Memory in Patients with Advanced Leukemia. Sci Transl Med 3, 95ra73-95ra73 (2011).



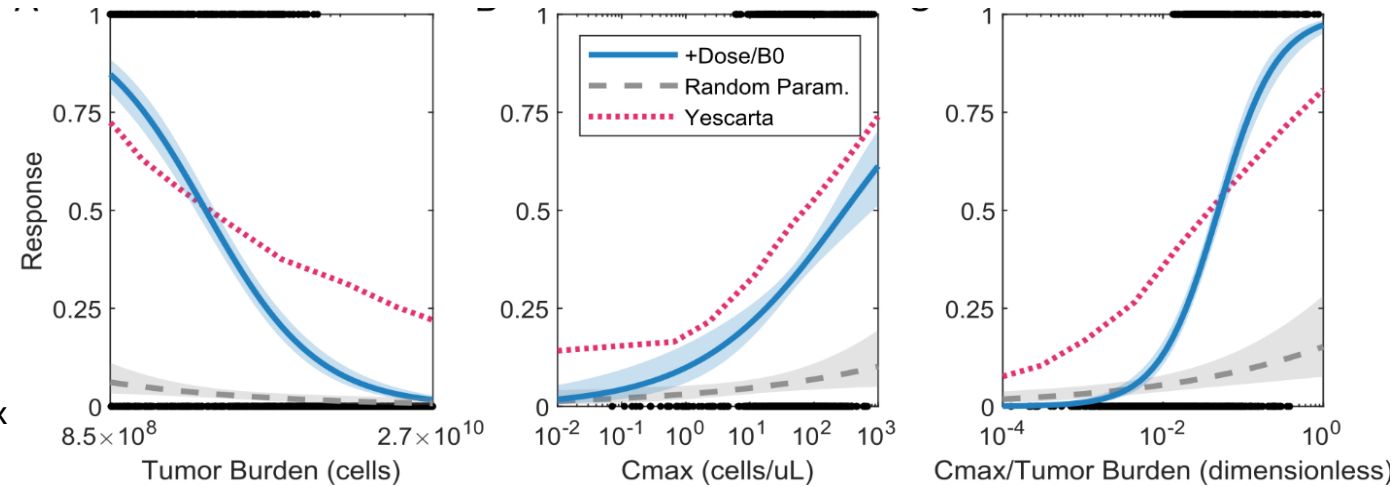
# Validation of model inferences via Clinical PKPD Data

Prediction: Tmem from CR-CART products have heightened intrinsic proliferative capacity

*Tmem* proliferation is explanatory of PK/response:  
R vs. NR, Kymriah in ALL\*



Predicted covariates of response (Cmax & tumor burden):  
*Virtual Population vs. Yescarta in LCBCL (ZUMA-1)\**

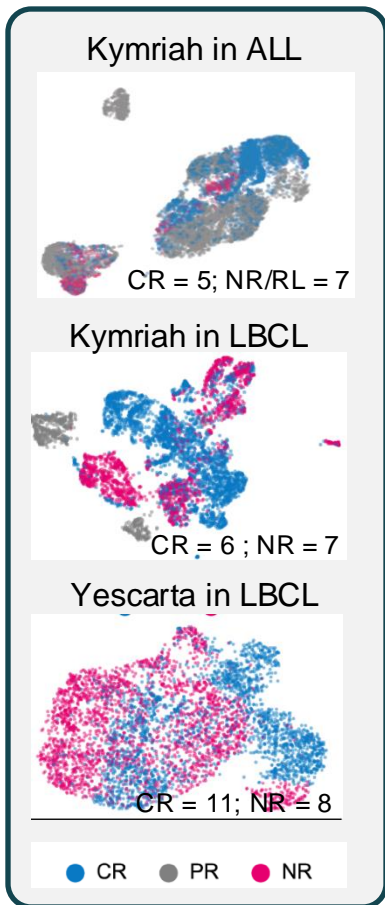


\*Liu C, Ayyar VS, Zheng X, et al (2020) Model-based Cellular Kinetic Analysis of Chimeric Antigen Receptor-T Cells in Humans. Clin Pharmacol Ther. 109(3):716-727

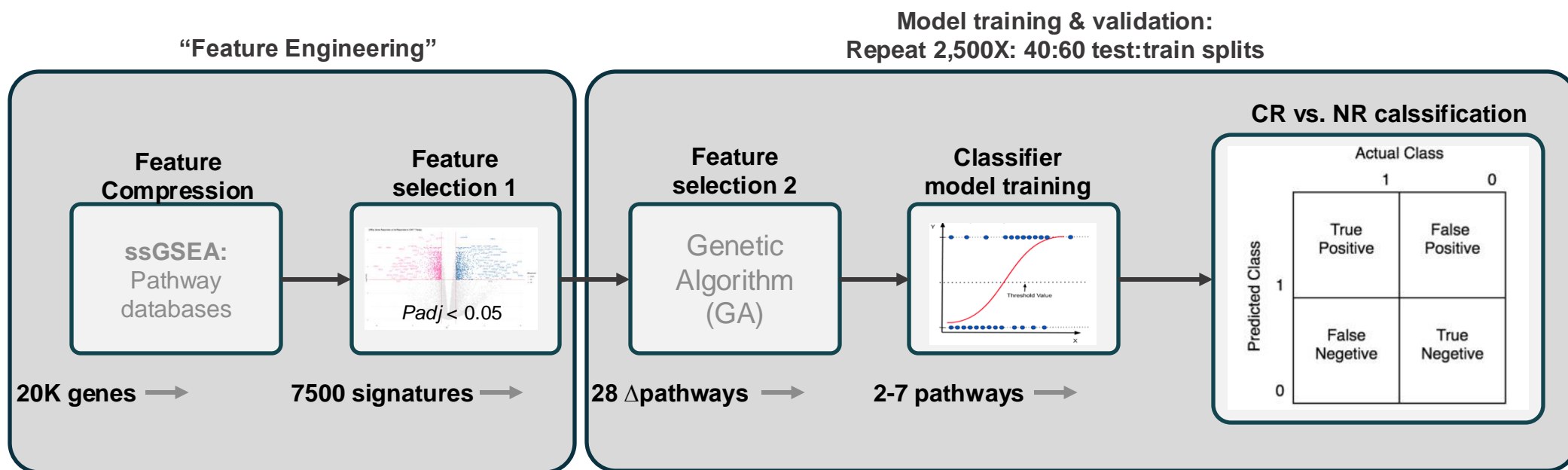
\*Locke FL, Rossi JM, Neelapu SS, et al (2020) Tumor burden, inflammation, and product attributes determine outcomes of axicabtagene ciloleucel in large B-cell lymphoma. Blood Adv 4:4898-4911.

# Can we predict response based on pre-infusion transcriptomes?

scRNAseq pre-infusion CAR-Ts  
CR/NR/PR classes



## Machine learning workflow



Large P, small N problem: the central challenge in biomedical genomics

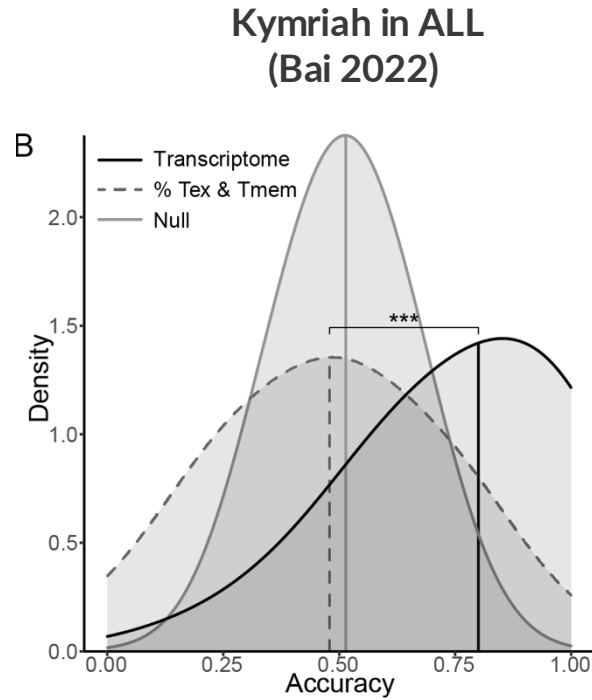
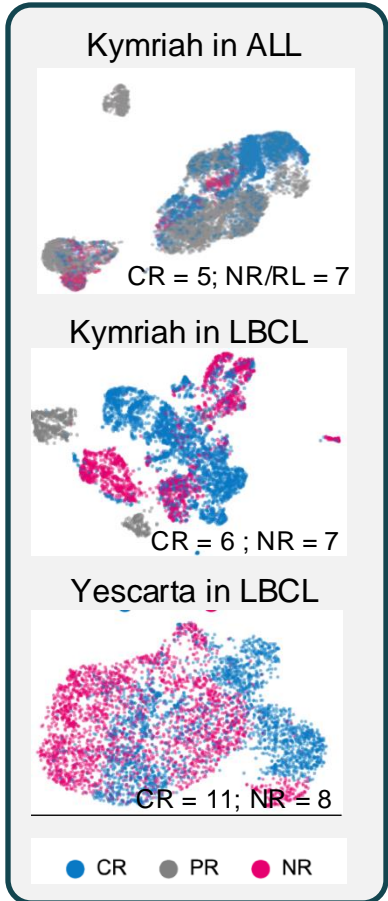
$$Accuracy = \frac{TP + TN}{TP + TN + FP + FN}$$

# Can we predict response based on pre-infusion transcriptomes?

Much better than expected by chance, and better than immunophenotyping

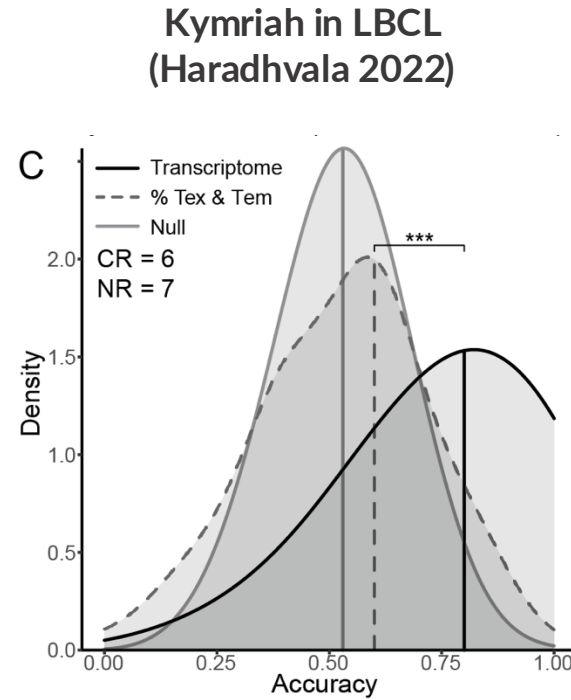
Predictive accuracy of response classification using 60:40 train:test splits

scRNAseq pre-infusion CAR-Ts  
CR/NR/PR classes

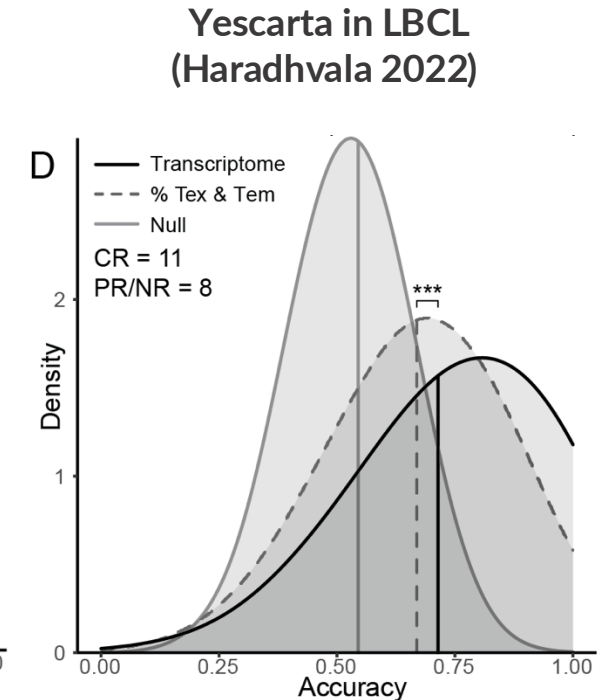


**Accuracy = 80%**  
Tmem, Tex: CITESeq data  
CR = 5; NR/RL = 7

\*\*\*  $P < 10^{-8}$  (rank-sum test)



**Accuracy = 80%**  
Tmem, Tex: ProjectTILS\*  
CR = 6; NR = 7



**Accuracy = 71%**  
Tmem, Tex: ProjectTILS  
CR = 11; NR/PR = 8

Functional attributes predictive of clinical outcomes are CART-cell-intrinsic & indication-agnostic  
Transcriptome > 'gold standard' immunophenotyping

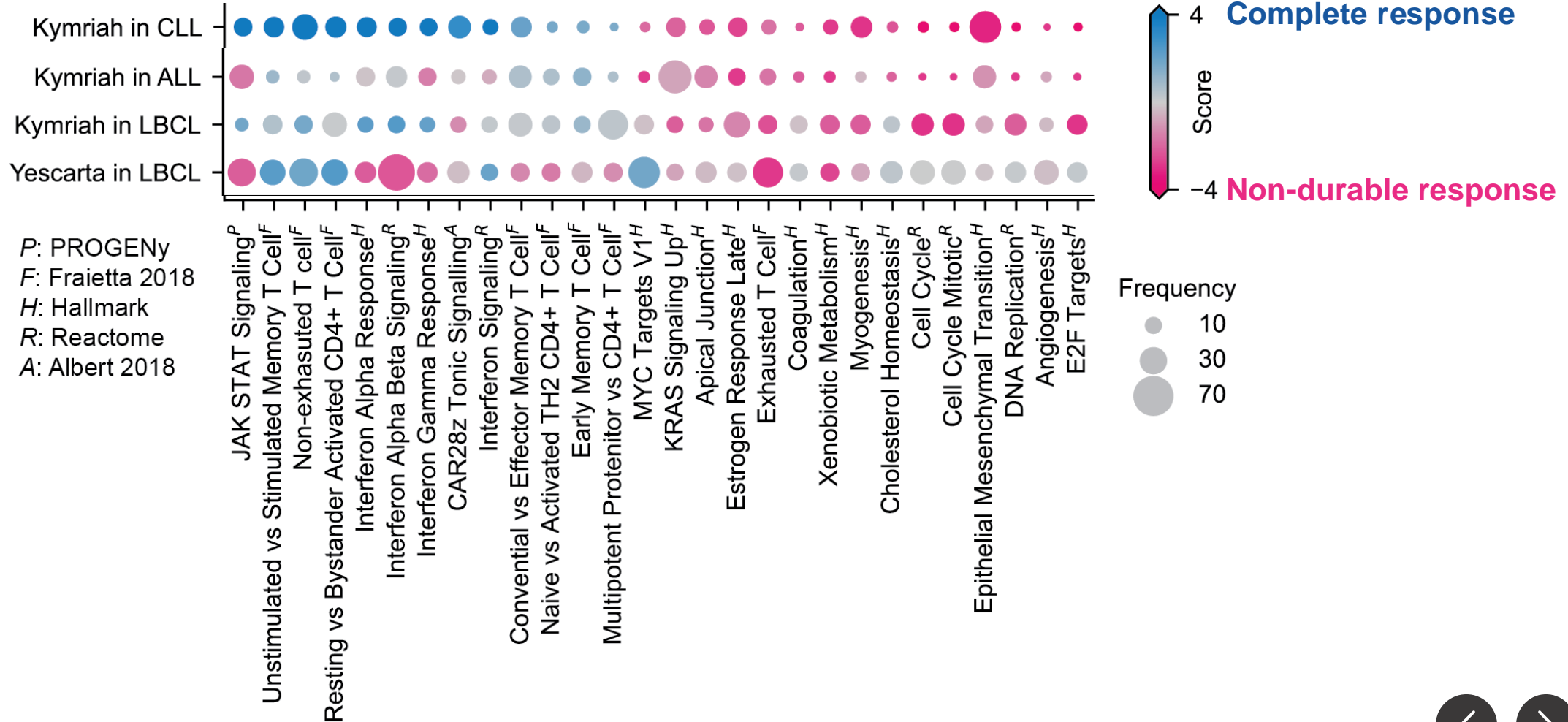
# What transcriptional features predict response?

Multivariate predictive biomarkers

CAR-T Response Score-card

## Accuracy

90%  
80%  
80%  
71%



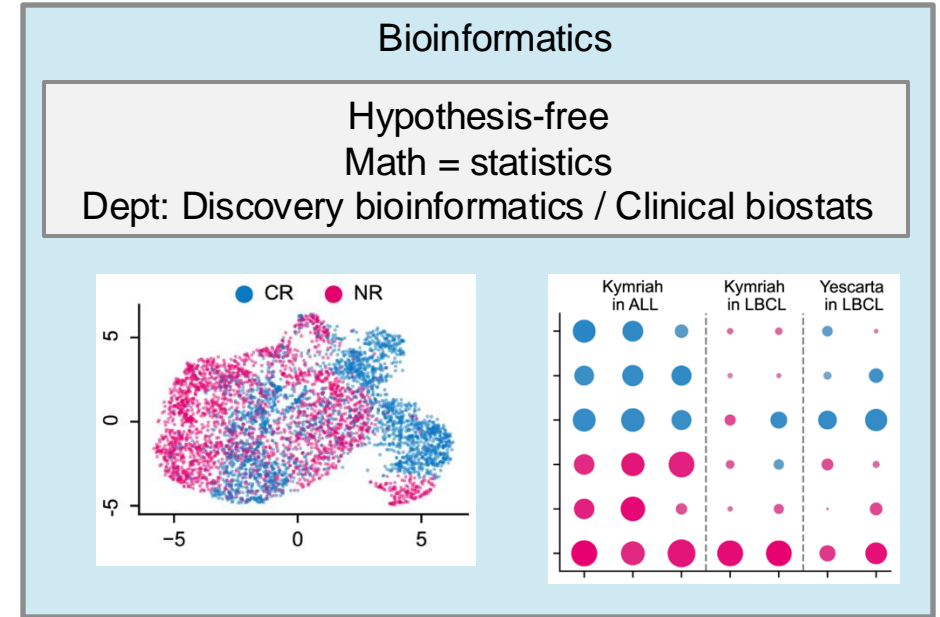
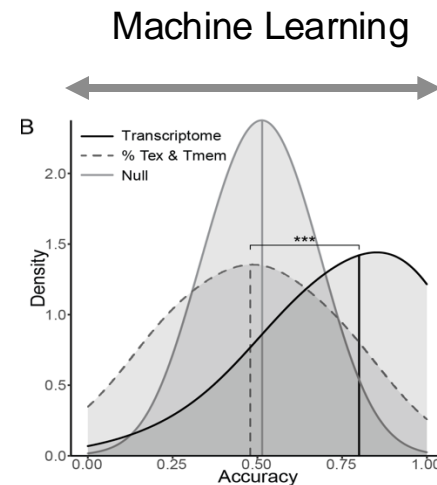
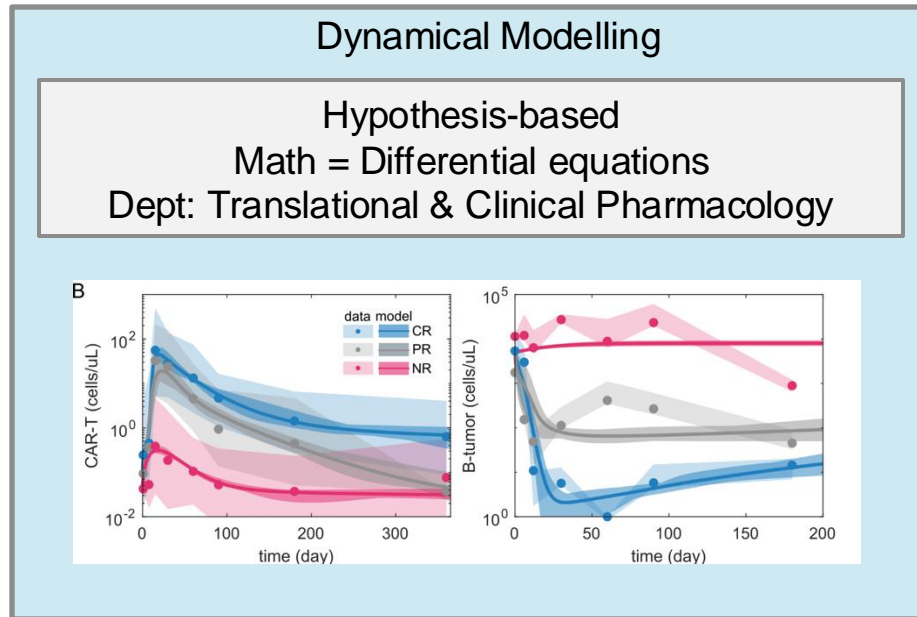
# Method-agnostic modelling

Dynamical systems modelling, bioinformatics and machine learning

## Motivation

If we can identify functional attributes of CAR-Ts which result in robust exposure & clinical response,  
Then we can design these attributes into products

## Approach



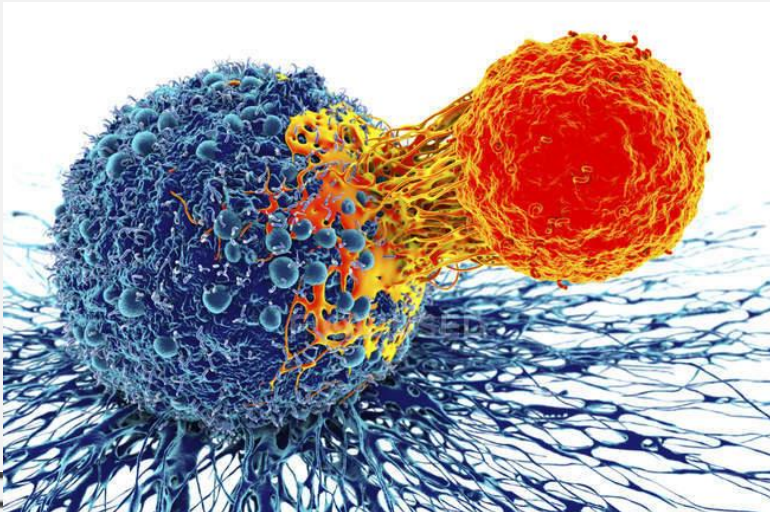
## Key finding

Not all memory cells are created equal: Tmem associated with non-durable response display functional defects characteristic of exhaustion – reduced proliferative and functional capacity



# Translational (T cell) pharmacology

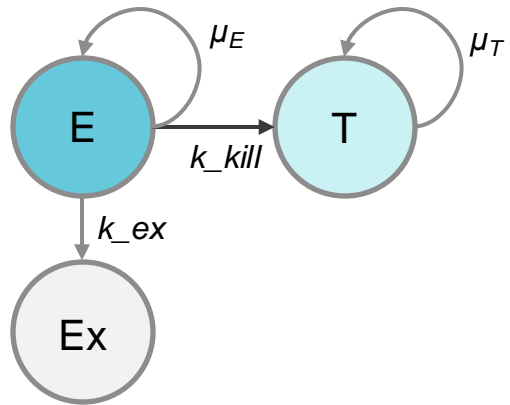
Quantifying & predicting T cell potency



# Quantifying T cell potency from co-culture assays

Data quantification & compression

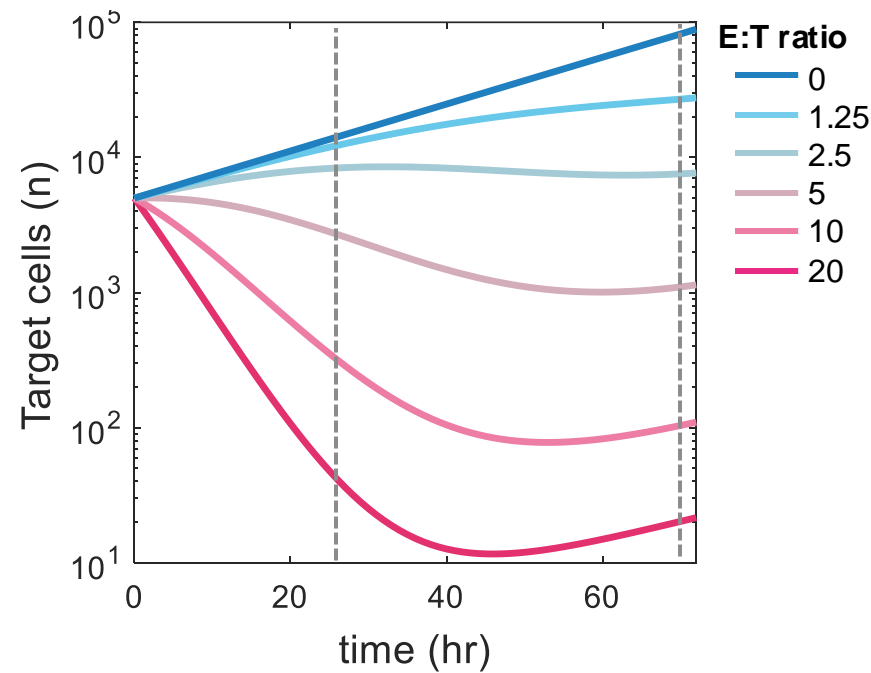
Model structure



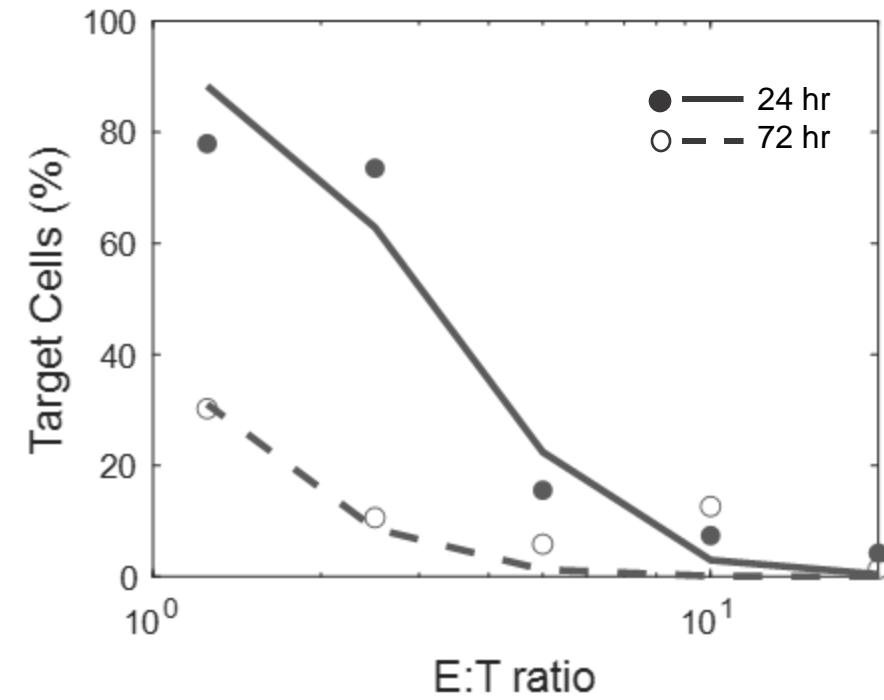
Species	Description
T	Target cells (n)
E	Effector cells (n)
Ex	Exhausted effector cells (n)

Parameter	Description
$u_E$	effector proliferation rate ( $hr^{-1}$ )
T50	EC50 limiting cell proliferation (cells)
$u_T$	target cell proliferation rate ( $hr^{-1}$ )
$k_{kill}$	killing rate ( $hr^{-1}$ )
E50	EC50 effector cell killing (cells)
$K_e$	Hill coefficient of cell killing
$k_{ex}$	Exhaustion rate ( $hr^{-1}$ )

Model Simulations:  
Target (cancer) cell dynamics

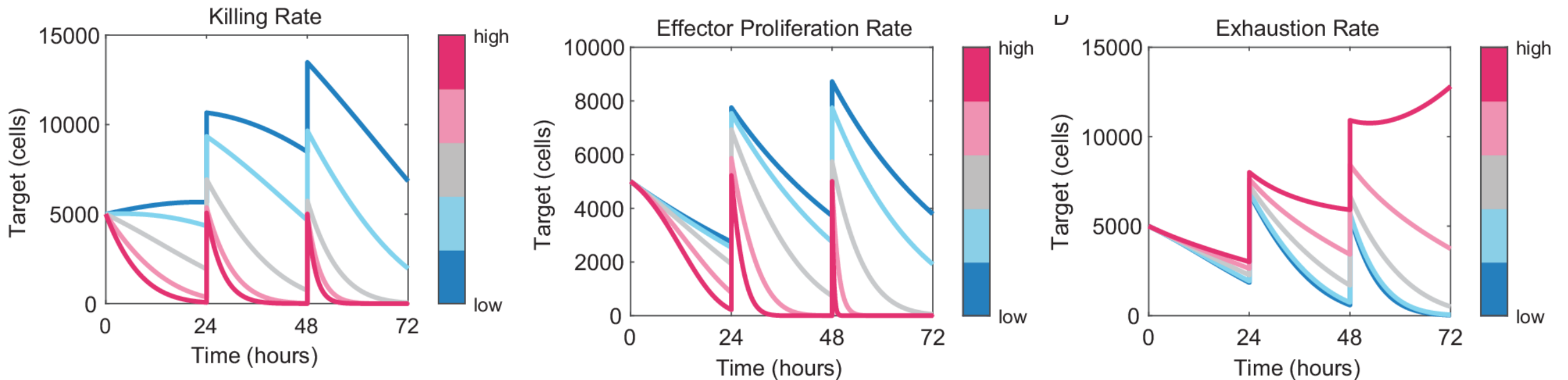


Model Fitting:  
Target (cancer) cell viability

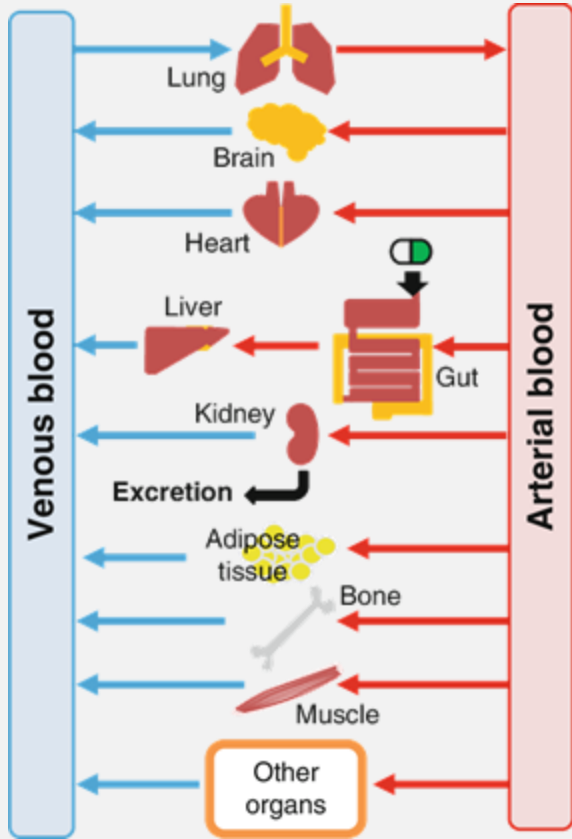


# Model-based inference from serial-killing assay data

Cytotoxic potency, Proliferation & exhaustion can be inferred using 'simple' models



**Utility:** We can map the effect of molecular perturbations to functional kinetic parameters

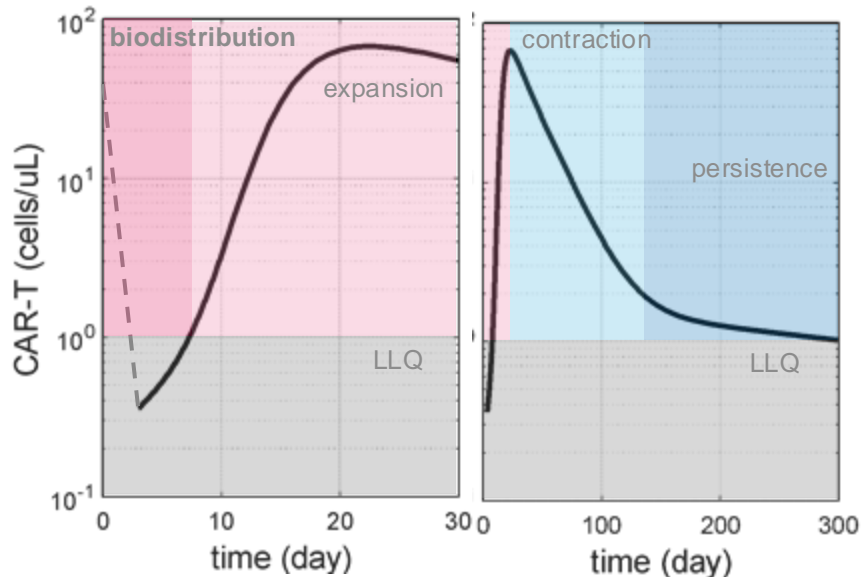


**T Cell Biodistribution:**  
the elephant in the (CART pharmacology) room

# We can *model* tissue distribution

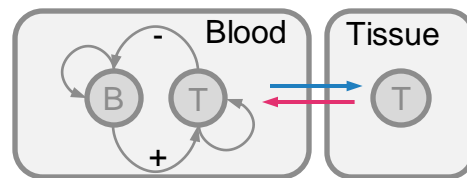
...but we don't have the data to constrain the models or make predictions

## CAR-T pharmacokinetics: 4 phases

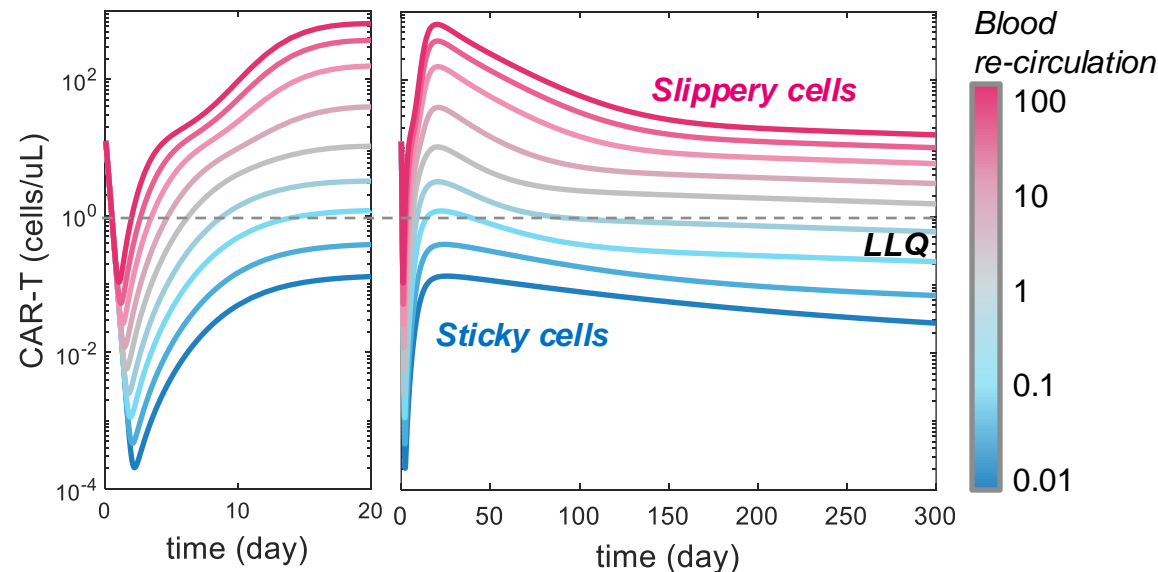


\* *biodistribution* phase is poorly characterized

## Tissue distribution can dominate T cell pharmacokinetics



Blood re-circulation rate:  
 $K_{blood} / K_{tissue}$



PKPD model of Kymriah-Responders with tissue-distribution incorporated

# Model *remixing* for the next phase of CAR-T clinical development

How to position cell therapies for autoimmunity & beyond?

## Model of TCE PKPD in Lymphoma

npj | Systems Biology and Applications [www.nature.com/npjbsa](http://www.nature.com/npjbsa)

ARTICLE OPEN [Check for updates](#)

Mitigating the risk of cytokine release syndrome in a Phase I trial of CD20/CD3 bispecific antibody mosunetuzumab in NHL: impact of translational system modeling

Iraj Hosseini<sup>1</sup>, Kapil Gadkar<sup>1</sup>, Eric Stefanich<sup>1</sup>, Chi-Chung Li<sup>1</sup>, Liping L. Sun<sup>1</sup>, Yu-Way Chu<sup>1</sup> and Saroja Ramanujan<sup>1,2,3</sup>

## Model of Lupus disease progression

ARTICLE

Disease trajectory of SLE clinical endpoints and covariates affecting disease severity and probability of response: Analysis of pooled patient-level placebo (Standard-of-Care) data to enable model-informed drug development

Kosalaram Goteti<sup>1</sup> | Jonathan French<sup>2</sup> | Ramon Garcia<sup>2</sup> | Ying Li<sup>1</sup> | Florence Casset-Semanaz<sup>1</sup> | Aida Aydemir<sup>1</sup> | Robert Townsend<sup>1</sup> | Cristina Vazquez Mateo<sup>1</sup> | Matthew Studham<sup>1</sup> | Oliver Guenther<sup>3</sup> | Amy Kao<sup>1</sup> | Marc Gastonguay<sup>2</sup> | Pascal Girard<sup>4</sup> | Lisa Benincosa<sup>1</sup> | Karthik Venkatakrishnan<sup>1</sup>

## Model of CAR-T PKPD in Lupus

Clinical and Translational Science

ARTICLE OPEN ACCESS

Mechanistic Evaluation of Anti-CD19 CAR-T Cell Therapy Repurposed in Systemic Lupus Erythematosus Using a Quantitative Systems Pharmacology Model

Hyunseo Park<sup>1,2</sup> | Ganesh M. Mugundu<sup>1</sup> | Aman P. Singh<sup>1</sup>

## Model of CAR-T PKPD in lymphoma

nature biotechnology

Article <https://doi.org/10.1038/s41587-023-01687-x>

Deconvolution of clinical variance in CAR-T cell pharmacology and response

Received: 14 March 2022 Accepted: 20 January 2023

Daniel C. Kirouac<sup>1,2,3</sup>, Cole Zmurchok<sup>1,3</sup>, Avisek Deyati<sup>1,3</sup>, Jordan Sicherman<sup>1</sup>, Chris Bond<sup>1</sup> & Peter W. Zandstra<sup>1,2</sup>



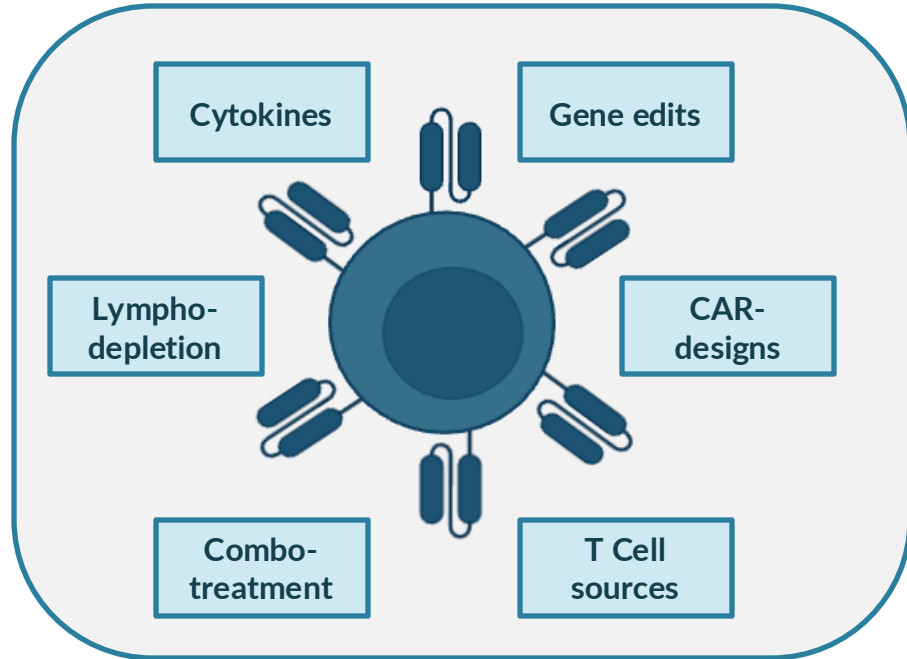
*In Silico* clinical trials for head-to-head comparisons



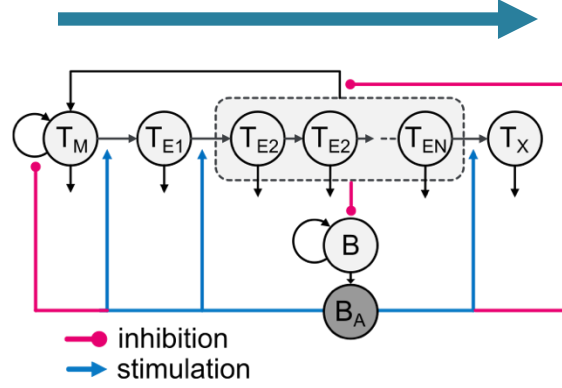
# What is the value of mathematical modeling?

In the context of cell therapy engineering and clinical development

## Cell therapy engineering



## Prediction



## Inference

## Clinical strategy

