

Treatment Outcome Prediction for Cancer Patients

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Dynamics and Data Assimilation, Physiology and Bioinformatics: Mathematics at the Interface of Theory and Clinical Application

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Outline

- 1 Approaches for Prediction
- 2 CAR T-cell Therapy in DLBCL Patients
- 3 Pilot Study
- 4 Conclusion and Future Directions

Approaches for Prediction

- Physiology-Based Mechanistic Modeling
- Machine Learning Modeling

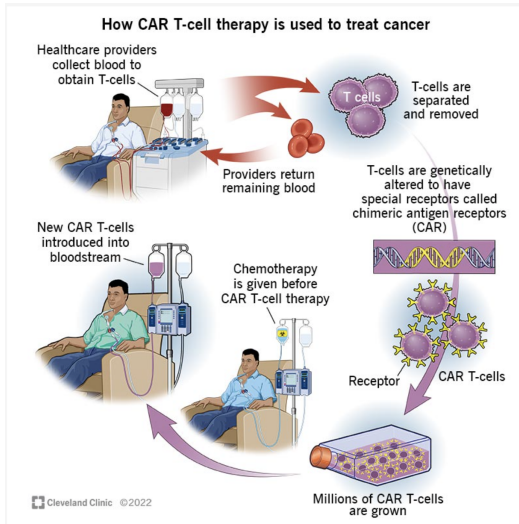
Approaches for Prediction

- Physiology-Based Mechanistic Modeling
- Machine Learning Modeling

This is very extensive field. We will focus on

- one treatment → chimeric antigen receptor (CAR) T-cell therapy
- one disease → diffuse large B-cell lymphoma (DLBCL)
- one modeling approach → mechanistic modeling

CAR T-cell Therapy¹



¹URL: <https://my.clevelandclinic.org/health/treatments/17726-car-t-cell-therapy>.

Diffuse Large B-cell Lymphoma

- An aggressive and the most common type of Non-Hodgkin lymphoma (NHL)
- 81,560 estimated new cases and 20,270 deaths in 2021 due to NHL
- DLBCL accounts for 30-35% of the newly diagnosed cases
- The number of new cases is projected to increase in the next 5 years
- CAR T-cell therapy is FDA-approved as the third-line of treatment option (for leukemia, lymphoma, myeloma)

Why do we want to predict the therapy outcome?

- Even though it has shown improvement, CAR T-cell therapy does not work for everyone
- It may cause severe adverse events: toxicity
- It is a costly treatment

When can we predict?

Modeling CAR T-cell in Glioma Setting²

$$\begin{aligned} \text{cancer cell rate of change} \quad \frac{dX}{dt} &= \overbrace{\rho X \left(1 - \frac{X}{K}\right)}^{\text{logistic growth of cancer cells}} - \overbrace{\kappa_1 XY}^{\text{CAR T-cell induced cancer cell death}} \\ \text{CAR T-cell rate of change} \quad \frac{dY}{dt} &= \underbrace{\kappa_2 XY}_{\text{cancer cell stimulated proliferation or exhaustion of CAR T-cells}} - \underbrace{\theta Y}_{\text{CAR T-cell death}} \end{aligned}$$

Model Parameters:

ρ : Cancer cell net growth rate (day^{-1})

K : carrying capacity (cell)

κ_1 : CAR T-cell killing rate ($\text{day}^{-1} \text{ cell}^{-1}$)

κ_2 : Net rate of proliferation and exhaustion of T-cells when stimulated by cancer cells ($\text{day}^{-1} \text{ cell}^{-1}$)

θ : CAR T-cell net death rate (persistence) (day^{-1})

States:

X: cancer cell count

Y: CAR T-cell count

²Prativa Sahoo et al. "Mathematical deconvolution of CAR T-cell proliferation and exhaustion from real-time killing assay data". In: *Journal of the Royal Society Interface* 17.162 (2020), p. 20190734.

Modeling CAR T-cell in Glioma Setting

By re-scaling time and state variables,

$$\tau = t\rho, \quad y = \frac{\kappa_1}{\rho} Y, \quad x = \frac{1}{K} X,$$

we obtain an equivalent dimensionless system

$$\begin{aligned}\frac{dx}{d\tau} &= x(1-x) - xy \\ \frac{dy}{d\tau} &= Bxy - Ay\end{aligned}$$

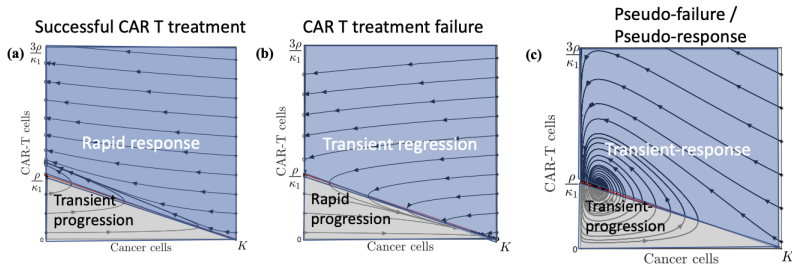
with the following dimensionless parameters

$$A = \frac{\theta}{\rho}, \quad B = \frac{\kappa_2 K}{\rho}.$$

Modeling CAR T-cell in Glioma Setting

Possible Dynamics of the CARRGO Model:

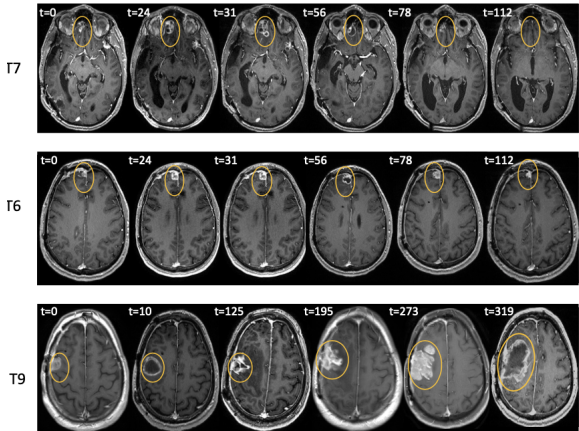
$$A = \frac{\theta}{\rho} \quad \text{and} \quad B = \frac{\kappa_2 K}{\rho}$$



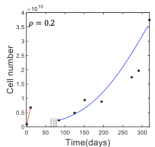
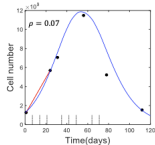
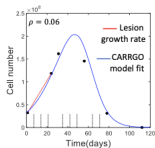
- 1 Successful CAR T-cell treatment: $A = 0, B > 0$
- 2 CAR T-cell treatment failure: $A = 0, B < 0$
- 3 Pseudo-failure or pseudo-response: $A > 0, B > 0$

Modeling CAR T-cell in Glioma Setting

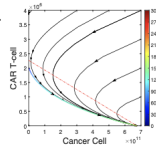
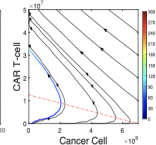
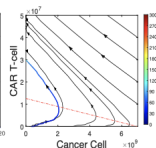
T1-Gadolinium enhanced MRI



Tumor/lesion
Volume dynamics



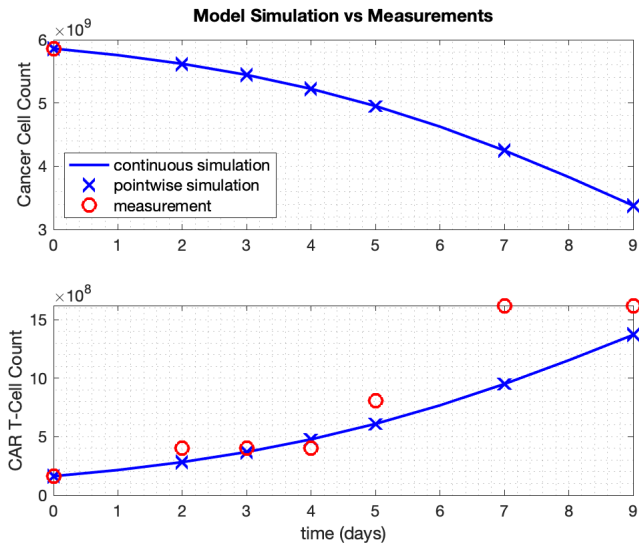
CARRGO model
dynamics



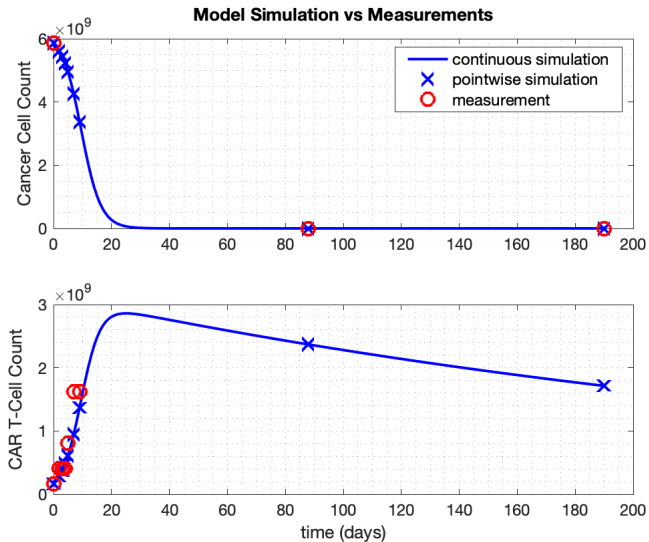
Dataset:

- TMTV measurements → cancer cell count
1 measurement used as the initial condition
- ALC measurements → CAR T-cell count
at most 10 measurements collected until after 15 days of CAR T-cell infusion

Model Estimation



Prediction



Conclusion

Mechanistic modeling could

- provide insight into cancer cell-CAR T-cell dynamics
- handle the data sparsity limitations
- provide more information than binary outcome prediction, e.g., timing of relapse

Future Directions

- More accurate estimation of the TMTV (cancer cell count) at the time of infusion
- Investigating use of other biomarkers for more accurate and direct estimation of cancer and CAR T-cell counts
- Identifying CAR T-cell proliferation and exhaustion time windows accurately
- Investigating optimal dosing strategies to enhance the treatment outcome

Thank you!

Questions?