### **Combination Therapy with T cell engager and PD-L1 Blockade Enhances the Antitumor Potency of T Cells as Predicted by a QSP Model**

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## **Supplementary Information**

### **1.1 Updated Immune checkpoint blockade dynamics for atezolizumab**

Immune checkpoint blockade dynamics has been described by Jafarnejad et al. (1) and was expanded here for anti-PD-L1 blockades. PD-1 expressed in Teff interacts with PD-L1 and PD-L2 on cancer cell in the immunological synapse (Eqs. 1 and 2). The formation of PD1\_PDLX (PD1\_PDL1 + PD1\_PDL2) (Eq. 6) will cause reduced cancer killing by Teff. The expression of PD-1 on T cells and PD-L1/PD-L2 on cancer cells or APCs were estimated based on measurements using quantitative flow cytometry using calibrated fluorescent beads (2, 3). Atezolizumab binding to the PD-L1 was modeled using a bivalent model of antibody receptor interaction on cell surface by introducing a cross-arm binding efficiency  $X$  (Eqs. 3,4 and 5). Relevant governing equations for the dynamics of the checkpoint molecules in the immune synapse are summarized below based on previous publications (1, 4, 5):

$$
\frac{dPD1_PDL1}{dt} = k_{on,PD1_PDL1} \cdot PD1 \cdot PDL1 - k_{off,PD1_PDL1} \cdot PD1_PDL1 \tag{1}
$$

$$
\frac{dPD1\_PDL2}{dt} = k_{on,PD1\_PDL2} \cdot PD1 \cdot PDL2 - k_{off,PD1\_PDL2} \cdot PD1\_PDL2 \tag{2}
$$

$$
\frac{dPDL1\_Atezo}{dt} = 2k_{on,PDL1\_Atezo} \cdot PDL1 \cdot Atezo/f_{tum} - k_{off,PDL1\_Atezo} \cdot PDL1\_Atezo
$$
 (3)

$$
\frac{dPDL1\_Atezo\_PDL1}{dt}=X\left(\frac{k_{on,PDL1\_Atezo}}{A_{syn}d_{syn}N_{A}}\right)\cdot PDL1_{Atezo}\cdot PDL1-2*k_{off,PDL1\_Atezo}\cdot PDL1\_Atezo\_PDL1
$$

$$
\frac{dPDL1}{dt} = -k_{on,PD1_PDL1} \cdot PD1 \cdot PDL1 + k_{off,PD1_PDL1} \cdot PD1_{PDL1} - 2k_{on,PD1_Nivo} \cdot PD1 \cdot \frac{Nivo}{f_{tum}} +
$$
\n
$$
k_{off,PDL1_Atezo} \cdot PDL1_Atezo - X\left(\frac{k_{on,PDL1_Atezo}}{A_{syn}d_{syn}N_A}\right) \cdot PDL1_Atezo \cdot PDL1 + 2 * k_{off,PDL1_Atezo} \cdot
$$
\n
$$
PDL1_Atezo_PDL1
$$
\n
$$
(5)
$$

(4)

where  $k_{on,PD1_X}$  and  $k_{off,PD1_X}$  are the on and off rates for interactions between PD-1 and X (PD-L1, PD-L2, and atezolizumab),  $f_{tum}$  is the porosity in the tumor,  $X$  is the intrinsic antibody cross-arm binding efficiency,  $A_{syn}$  is surface area of the synapse,  $d_{syn}$  is the thickness of the confinement space between the two cells, and  $N_A$  is Avogadro's number.  $k_{on, PDL1 \text{ Atezo}}$  was converted to units of 1/(molecule.s) using the synapse sizes ( $A_{syn} * d_{syn}$ ) and Avogadro's number N<sub>A</sub>. The number of bound PD-L1 molecules on cancer cells or APCs was translated to Teff exhaustion using a Hill equation.

$$
PD1\_PDLX\_CC = PD1\_PDL1\_CC + PD1\_PDL2\_CC
$$
\n
$$
(6)
$$

$$
PD1\_PDLX\_APC = PD1\_PDL1\_APC + PD1\_PDL2\_APC
$$
\n
$$
(7)
$$

Here PD-L1/PD-L2 can be either expressed on cancer cells or APCs. PD-1/PD-L1/PD-L2 dynamics is assumed to be similar in T cell - cancer cell and T cell - APC, thus only a single checkpoint module was used in this study to reduce the size of the model. However, to differentiate the differences of PD-L1/PD-L2 expression in cancer cell and APC, Eqs. 1-5 are used simultaneously for cancer cell and APC when the simulations are running, resulting in two species  $PD1\_PDLX\_CC$ and  $PDI\_PDLX\_APC$ , which represent the number of bound PD-1 – PD-L1/PDL2 in T cell - cancer

cell and T cell – APC synapse. Then, the number was translated to Teff exhaustion using a Hill equation.

$$
H\_PDI\_PDLX\_CC = \left(1 - \frac{PD1\_PDLX\_CC^2}{PD1\_PDLX\_CC^2 + K_{PD1\_PDLX\_CC}^2}\right)
$$
\n
$$
(8)
$$

$$
H\_PDI\_PDLX\_APC = \left(1 - \frac{PD1\_PDLX\_APC^2}{PD1\_PDLX\_APC^2 + K_{PD1\_PDLX\_APC}^2}\right)
$$
\n
$$
(9)
$$

As we mentioned in the main paper in section 3.6, PD-L2 expression showed ambiguous results. To study the impact of PD-L2, we introduced another parameter δ in the Hill functions (6) and (7) and added it into the parameter sensitivity analysis and assigned its range between 0 and 1.

$$
PD1\_PDLX\_CC = PD1\_PDL1\_CC + \delta * PD1\_PDL2\_CC
$$
\n
$$
(10)
$$

$$
PD1\_PDLX\_APC = PD1\_PDL1\_APC + \delta * PD1\_PDL2\_APC \tag{11}
$$

### **1.2 Updated T cell activation and proliferation**

T cells activation in TdLN is based on a two-step priming model described by Jafarnejad et al [REF]. In the main paper, the function of PD-L1/PD-L2 expression in APCs has been discussed. After naïve T cells being activated by mAPCs in the first step, the PD-L1/PD-L2 expression in APCs is assumed to limit the proliferation of activated T cells into functional effector T cell by introducing an inverse Hill function (Eq. 9).

$$
TCPR = \frac{k_{\text{aTCDB,prolif}}}{n_{\text{prolif}}} \cdot 2^{n_{\text{prolif}}} \cdot T_{\text{activated,CD8}} * H\_PD1\_PDLX\_APC
$$
 (12)

where *TCPR* is T cell proliferation rate,  $k_{\text{aTCD8,prolif}}$  is doubling rate of activated T cells,  $n_{\text{prolif}}$  is the number of generations T cells proliferate, and T<sub>activated,CD8</sub> is the number of activated T cells in the TdLN. The number of generations that activated T cells proliferate (division destiny) before leaving the TdLN depends on TCR engagement, co-stimulation signal through CD28, and IL-2 receptor stimulation.

## **1.3 Updated Tumor growth**

Teff killing rate (TKR) is expressed as an inverse Hill equation of immune checkpoint inhibitors and T cell engagers as follows, which remains the same as reported by Ma et al. (6).

$$
H_TCE = \frac{CEACEA_TCE_TeffCD3^3}{CEACEA_TCE_TeffCD3^3 + K_{CEACEA_TCE_TeffCD3}^3}
$$
(13)

$$
TKR = k_{C,death, TCET_{eff}} \frac{c \cdot T_{eff}}{c + T_{tot}} * H\_TCE + k_{C,death, T_{eff}} \frac{c \cdot T_{eff}}{c + T_{tot}} * H\_PD1\_PDLX\_CC
$$
 (14)

The number of bound CEACEA\_TCE\_TeffCD3 was translated to cancer cell killing rate by Teff cells using a Hill equation. The immune checkpoint blockade dynamics elaborated by Jafarnejad et al. (1). was combined with T cell engager dynamics in the Teff cell killing rate (TKR).

Here  $k_{C, death, T_{eff}}$  is basal cancer killing rate by Teff and  $k_{C, death, TCET_{eff}}$  is additional cancer killing rate by Teff activated by TCE, *C* is the total number of cancer cells in the tumor compartment, *Teff* is total number of Teff in the tumor and  $T_{tot}$  is total number of T cells in the tumor,  $CEACEA\_TCE\_TeffCD3$  is the total number of engaged CEA CD3 molecules bridged by TCE in the synapse, and *K<sub>CEACEA TCE TeffCD3</sub>* is sensitivity of TKR to CEACEA\_TCE\_TeffCD3. Formation of CEACEA\_TCE\_TeffCD3 will increase *TKR* according to the Hill equation (*H\_TCE*), and formation of PD1\_PDLX\_CC will slow down TKR according to the inverse Hill equation. Details of TCE dynamics were provided by Ma et al. (6).

# **Supplementary Figures**

## **Pharmacokinetics**

Pharmacokinetic of atezolizumab was modelled following the same physiologically-based pharmacokinetic model as described by Jafarnejad et al. (1). The plasma concentration of atezolizumab in our model was fitted to standard pharmacokinetic two-compartment model (Fig. S1). PK parameters were fitted to the data reported and the simulated plasma concentration of atezolizumab together with the clinical measurements at dose levels of  $1, 3, 10, 15$  mg/kg and  $1200$ mg in the central compartment (7).

 $Atezo<sub>P</sub>$ , Atezo<sub>c</sub>, Atezo<sub>LN</sub>, Atezo<sub>T</sub> indicate atezolizumab concentration in peripheral, central, TdLN and tumor compartment, respectively.

$$
V_C \frac{a_{Atezo_C}}{dt} = q_P(Atezo_P - Atezo_C) + q_{LN}(Atezo_{LN} - Atezo_C) + q_T(Atezo_T - Atezo_C) + q_{LD}Atezo_{LN}
$$
\n
$$
CL * Atezo_C
$$
\n
$$
(13)
$$

$$
V_P \frac{a \text{At} \exp}{dt} = q_P (A \text{tezo}_C - A \text{tezo}_P) \tag{14}
$$

$$
V_T \frac{a \text{A} t e^{20} T}{dt} = q_T (A t e^{20} C - A t e^{20} T) - q_{LD} A t e^{20} T \tag{15}
$$

$$
V_{LN}\frac{a_{Atezo_{LN}}}{dt} = q_{LN}(Atezo_C - Atezo_{LN}) + q_{LD}Atezo_T - q_{LD}Atezo_{LN}
$$
\n(16)



Figure S1. Simulated (solid lines) and measured (dots) atezolizumab plasma concentration.

A.

B.



Figure S2. Percent change in tumor size represented using RECIST criteria (a "spider" plot). A. Atezolizumab monotherapy (1312 virtual patients). B. Combination therapy (1299 virtual patients).



Figure S3. Bootstrapping results for atezolizumab monotherapy, cibisatamab monotherapy and combination therapy (10,000 bootstrap samples).



Figure S4. Waterfall plots for atezolizumab monotherapy while varying A. TMB; B. PD-1 expression; C. PD-L1 expression in cancer cell; D. PD-L1 expression in APCs.



Figure S5. Distributions of potential biomarker in NR and R in atezolizumab monotherapy. A. PD-L2 expression in cancer cells; B. PD-L2 expression in APCs.



Figure S6. Distribution of δ in NR and R of atezolizumab monotherapy.

# **Supplementary Tables**

## **Table. S1 Abbreviations**



 $\overline{a}$ 





# **Table. S2 Atezolizumab-related Variables and Terms Used in Equations**

# **Table. S3 Parameter Values and Ranges Used in the Sensitivity Analysis**



### **Table S4. Overall Response Rate**





#### **Table S5. Distribution of Overall Response Rate in Bootstrapping samples**

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