# Modeling Cyclophosphamide's Effect on Leukocytes

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## Agenda

- Motivation For Project
- Background Biology and Chemistry

- Model Construction
- Differential Equations
- Results
- Further Study

## Background on Cyclophosphamide

- Pro-drug typically used in immune suppression and chemotherapy
- Can assist in oncolic virotherapy, the use of engineered viruses to combat cancerous cells
- Used to suppress the immune system enough so that the viruses can infect and kill the cancerous cells in the body
- Necessary to control the timing and amount of doses due to its toxicity

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## Pharmacodynamics of Cyclophosphamide (CY)

- Inactive until it reaches the liver and is metabolized to hydroxycyclophosphamide(HCY)
- About 70% of CY is metabolized to HCY, the rest is primarily excreted unchanged in urine

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- HCY lives in equilibrium with its tautomer aldophosphamide(AP)
- AP is what kills cells at tissue level
- HCY is primarily eliminated through AP

#### Pharmacodynamics of Cyclophosphamide



Credit: Cyclophosphamide metabolism, liver toxicity, and mortality following hematopoieitic stem cell transplantation

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## Model Building for CY

- Focused on CY concentrations in the liver and blood and the AP concentration in the liver and tissue
- Didn't include HCY in the model, AP is what interacts chemically in tissues and degrades in the liver
- Assumed that a third of the amount of HCY being activated in the liver directly converted to AP
- Blood is simply a means of transport between the liver and tissues

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Pharmacodynamics of Cyclophosphamide



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## Differential Equations For CY Pharmacodynamics

$$\frac{dC_B}{dt} = k_1(C_L - C_B) - C_B k_{EC} + D(t) \tag{1}$$

$$\frac{dC_L}{dt} = -k_1(C_L - C_B) - C_L k_{AH}$$
(2)

$$\frac{dA_L}{dt} = k_{AA}C_L - k_2(A_L - A_T) - k_{EA}A_L \tag{3}$$

$$\frac{dA_T}{dt} = -k_2(A_T - A_L) - \mu A_T \tag{4}$$

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 $\mathsf{D}(\mathsf{t})$  is the controlled dose given every 24 hours, denoted as a piecewise function

#### Some Results



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## Population Dynamics of Leukocytes

 Stem cells (S), multipotent progenitor cells(MPP), common progenitor cells(CM), lymphocytes (L) and granulocytes(G)

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- $\blacktriangleright \ \mathsf{S} \to \mathsf{MPP} \to \mathsf{CM} \to \mathsf{L\&G}$
- S and MPP also regenerate depending on L and G cell concentration
- Together L and G make up all the Leukocytes

#### Leukocyte Model

- $\phi$  -feedback functions (L and G)
- $\mu$  death rates
- ▶  $\lambda$ ,  $r_d$ ,  $r_{p'}$ ,  $r_{cm}$  rates
- θ fixed proportion L is made from CM



#### Leukocyte Model Differential Equations

$$\frac{dS}{dt} = S \ln(\frac{K}{S})(r_s - r_{p'}\phi_{p'}[L,G])\phi_s[L,G] - \mu_s S$$
(5)

$$\frac{dMPP}{dt} = S \ln(\frac{K}{S})(r_s + 2r_{p'}\phi_{p'}[L,G])\phi_s[L,G] +$$

$$MPP((\lambda - r_d)\phi_p[L, G] - \mu_p - \alpha_1 A_t(t))$$
(6)

$$\frac{dCM}{dt} = MPPr_d\phi_p[L,G]\Omega_N - CM(\mu_{cm} + r_{cm} + \alpha_2A_t(t))$$
(7)

$$\frac{dL}{dt} = CMr_{cm}\theta \frac{10}{3003} + L(r_l - \mu_l - \mu_{l*}I_{vt>vth} - \alpha_3A_t(t))$$
(8)

$$\frac{dG}{dt} = CMr_{cm}(1-\theta)\frac{10}{3003} + G\mu_g$$
(9)

### Some Issues With Interpreting this Model

- Could not find the  $\alpha_3$  value
- My model doesnt work with only  $\alpha_2$  &  $\alpha_1$
- Because of the feedback functions, effects of the doses are extremely complicated

## 10 days 5 mg/kg a day







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My solution of differential equations as a function of a loading dose, maintenance dose, and time in Mathematica

 A loading dose is a high dose given to achieve a drastic effect quickly

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- First three days
- A maintenance dose is a lower dose to maintain that effect
  - Following seven days

# Examples of Optimal Loading and Maintenance Doses depending on $\alpha_{3}$

$lpha_{3}$ value	LD	MD	$\Delta$ L 3 days	$\Delta$ L 10 days
0	200	10	17.3%	50.2%
.1	60	1	30.3%	73.6%
.2	30	1.5	29.6%	73.8%
.25	25	1.5	29.9%	74.6%
.3	22	0.5	30.4%	74.9%
.35	19	0	30.3%	74.5%

## The end product



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### Future Work

- Taking the viral response into account
- Finding a more definitive α<sub>3</sub> parameter or maybe α<sub>1</sub> & α<sub>2</sub> are functions rather than rates

- Fit data to my model
- Using elements of Control theory to find optimal dosage

## Thanks for listening!

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