Mathematical Biosciences

## Research article

## An elementary mathematical modeling of drug resistance in cancer

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## Supplementary

The biological descriptions underlying the parameters along with their symbols, units, estimated values and reference sources are listed as follows.

Table S.1. Values of parameters used in simulations.

| Symbol | Value | Unit | Description | Reference |
| :---: | :---: | :---: | :---: | :---: |
| $l$ | 0.2 | day $^{-1}$ | Birth rate constant of cancer cells | Estimated from [3] |
| $d$ | 0.1 | day $^{-1}$ | Death rate constant of cancer cells | Estimated from [3] |
| $\mu$ | $1.0 * 10^{-4}$ | - | Mutation rate of sensitive cells to resistant cells | $[8]$ |
| $M$ | 1.0 | $10^{9}$ | Critical size of total cancer cells can be detected | $[2]$ |
| $N_{0}$ | 1.0 | $10^{2}$ | Initial size of drug-sensitive cells | $[3]$ |
| $h$ | 0.366 | day $^{-1}$ | Maximal death of drug on drug-sensitive cells | Assumed |
| $D_{1}$ | none | - | Concentration of drug-1 | none |
| $D_{2}$ | none | - | Concentration of drug-2 | none |
| $K_{1}$ | 0.5 | - | Michaelis constant of drug-1 | Assumed |
| $K_{2}$ | 0.1 | - | Michaelis constant of drug-2 | Assumed |

[^0]
## S.1. The case of three drugs

Consider the case of treatment with three drugs are being simultaneously used. The system that describes the dynamics of pre-treatment phase $\left(t \leq t^{*}\right)$ is given by

$$
\left\{\begin{array}{l}
N^{\prime}(t)=(l-d) N(t), \\
R_{1}^{\prime}(t)=(l-d) R_{1}(t)+\mu N(t), \\
R_{2}^{\prime}(t)=(l-d) R_{2}(t)+\mu N(t), \\
R_{3}^{\prime}(t)=(l-d) R_{3}(t)+\mu N(t), \\
R_{1,2}^{\prime}(t)=(l-d) R_{1,2}(t)+\mu R_{1}(t)+\mu R_{2}(t), \\
R_{1,3}^{\prime}(t)=(l-d) R_{1,3}(t)+\mu R_{1}(t)+\mu R_{3}(t), \\
R_{2,3}^{\prime}(t)=(l-d) R_{2,3}(t)+\mu R_{2}(t)+\mu R_{3}(t), \\
R^{\prime}(t)=(l-d) R(t)+\mu R_{1,2}(t)+\mu R_{1,3}(t)+\mu R_{2,3}(t) .
\end{array} \quad t \leq t^{*} .\right.
$$

The system that describes the dynamics after the treatment starts $\left(t>t^{*}\right)$ is given by

$$
\left\{\begin{array}{l}
N^{\prime}(t)=\left(l-d-\sum_{i=1}^{3} \frac{h \cdot D_{i}}{K_{i}+D_{i}}\right) N(t), \\
R_{1}^{\prime}(t)=\left(l-d-h\left(\frac{D_{2}}{K_{2}+D_{2}}+\frac{D_{3}}{K_{3}+D_{3}}\right)\right) R_{1}(t)+\mu N(t), \\
R_{2}^{\prime}(t)=\left(l-d-h\left(\frac{D_{1}}{K_{1}+D_{1}}+\frac{D_{3}}{K_{3}+D_{3}}\right)\right) R_{2}(t)+\mu N(t), \\
R_{3}^{\prime}(t)=\left(l-d-h\left(\frac{D_{1}}{K_{1}+D_{1}}+\frac{D_{2}}{K_{2}+D_{2}}\right)\right) R_{3}(t)+\mu N(t), \quad t>t^{*} . \\
R_{1,2}^{\prime}(t)=\left(l-d-\frac{h \cdot D_{3}}{K_{3}+D_{3}}\right) R_{1,2}(t)+\mu R_{1}(t)+\mu R_{2}(t), \\
R_{1,3}^{\prime}(t)=\left(l-d-\frac{h \cdot D_{2}}{K_{2}+D_{2}}\right) R_{1,3}(t)+\mu R_{1}(t)+\mu R_{3}(t), \\
R_{2,3}^{\prime}(t)=\left(l-d-\frac{h \cdot D_{1}}{K_{1}+D_{1}}\right) R_{2,3}(t)+\mu R_{2}(t)+\mu R_{3}(t), \\
R^{\prime}(t)=(l-d) R(t)+\mu R_{1,2}(t)+\mu R_{1,3}(t)+\mu R_{2,3}(t)
\end{array}\right.
$$

The notations $D_{i}, K_{i}(i=1 \ldots 3)$ represent concentration and Michaelis constant of $i$ th drug, respectively.

## S.2. Parameter calibration

The biological meaning and values underlying the parameters are listed in Table S.1. Most of parameters used in simulations are collected or estimated from literatures, while others are assumed for numerical illustration. The detailed descriptions of the parameter values are given as follows.
(1) Parameters involved in cancer cells load

Previous studies [1] have reported that the total tumor load $M$ should be less than $10^{13}$ cells, which comes from white blood cell count measurements that range from $10^{5}$ to $10^{6}$ cells $/ \mathrm{ml}$ of blood in advanced cancers. Meanwhile, it should be more than $10^{9}$ cells or a $1-\mathrm{cm}$ mass, which is approximately the lower limit of clinical detection [2]. Moreover, the initial number of cancer cells $N_{0}$ should not be too large, so the resistance before $N_{0}$ is reached can be ignored. Thus, the initial number of cancer cells $N_{0}$ is set to $10^{2}$ as in ref. [3].

## (2) Parameter involved in mutation of cancer cells

The experimental and clinical data in Refs. [4-6] have indicated that the resistance is mainly caused by point mutation and gene amplification. The point mutation leads to the failure of the drug to bind to the target protein, which has been estimated to occur at a rate of $10^{-9}$ per base per cell division [7]. However, due to genetic instability, the rate of gene duplication is much higher than that of point mutation, which has been measured to be $10^{-4}$ per cell division [8]. Therefore, we set the mutation rate $\mu$ within a range of $10^{-9}$ to $10^{-4}$.

## (3) Parameters involved in birth and death of cancer cells

Although the experimental data in Ref. [3] suggested that the relative death rate of cancer cells $d$ along with the turnover rate $d / l$ are estimated to range from 0.1 to 0.5 , there is no study on direct measurement of turnover rate $d / l$. In order to include all possibilities, we use the similar way as in ref. [3] to set the turnover rate $d / l$ from 0 to 1 . The scenario where $d / l \ll 1$ represents very lowturnover, low death cancer, while the scenario where $d / l \approx 1$ corresponds to extremely high-turnover, high death cancer.

## S.3. The secondary mutation occurs before or after treatment

Consider the generation of resistance in the case where the secondary mutation occurs before or after treatment. First, we set the secondary mutation rate to zero after time of the beginning of the treatment, then $R_{2}^{\uparrow}(t)$ is the solution of $R(t)$ in system (2.4) with the initial condition

$$
R\left(t^{*}\right)=M\left[\frac{\mu \ln \frac{M}{N_{0}}}{l-d}\right]^{2},
$$

we have

$$
R_{2}^{\uparrow}(t)=M\left[\frac{\mu \ln \frac{M}{N_{0}}}{l-d}\right]^{2} e^{(l-d)\left(t-t^{*}\right)} .
$$

Next, we only set the secondary mutation rate to zero in the pre-treatment phase, and turn it back after the treatment starts. In this case, $R_{2}^{\downarrow}(t)$ is the solution of system (2.4) with the initial conditions

$$
\left\{\begin{array}{l}
N\left(t^{*}\right)=M, \\
R_{1}\left(t^{*}\right)=\frac{M \mu \ln \frac{M}{N_{0}}}{l-d}, \\
R_{2}\left(t^{*}\right)=\frac{M \mu \ln \frac{M}{N_{0}}}{l-d}, \\
R\left(t^{*}\right)=0,
\end{array}\right.
$$

and therefore

$$
\begin{aligned}
R_{2}^{\downarrow}(t)= & \frac{M \mu^{2}\left(K_{1}+D_{1}\right)\left(K_{2}+D_{2}\right)}{h^{2} \cdot D_{1} D_{2}}\left[e^{(l-d)\left(t-t^{*}\right)}-e^{\left(l-d-\frac{h \cdot D_{1}}{K_{1}+D_{1}}\right)\left(t-t^{*}\right)}-e^{\left(l-d-\frac{h \cdot D_{2}}{K_{2}+J_{2}}\right)\left(t-t^{*}\right)}\right] \\
& +\frac{M \mu^{2}\left(K_{1}+D_{1}\right) \ln \frac{M}{N_{0}}}{h \cdot D_{1}(l-d)}\left[e^{(l-d)\left(t-t^{*}\right)}-e^{\left(l-d-\frac{h \cdot D_{1}}{K_{1}+D_{1}}\right)\left(t-t^{*}\right)}\right] \\
& +\frac{M \mu^{2}\left(K_{2}+D_{2}\right) \ln \frac{M}{N_{0}}}{h \cdot D_{2}(l-d)}\left[e^{(l-d)\left(t-t^{*}\right)}-e^{\left(l-d-\frac{h \cdot D_{2}}{K_{2}+D_{2}}\right)\left(t-t^{*}\right)}\right] \\
& +\frac{M \mu^{2}\left(K_{1}+D_{1}\right)\left(K_{2}+D_{2}\right)}{h^{2} \cdot D_{1} D_{2}} e^{\left(l-d-\frac{h \cdot D_{1}}{K_{1}+D_{1}}-\frac{h \cdot D_{2}}{K_{2}+D_{2}}\right)\left(t-t^{*}\right)} .
\end{aligned}
$$

For any $t>t^{*}, R_{2}^{\uparrow}(t)>R_{2}^{\downarrow}(t)$ is satisfied under the condition

$$
M\left[\frac{\mu \ln \frac{M}{N_{0}}}{l-d}\right]^{2} \geq \frac{2 M \mu^{2}\left(K_{1}+D_{1}\right)\left(K_{2}+D_{2}\right)}{h^{2} \cdot D_{1} D_{2}}+\frac{M \mu^{2}\left(K_{1}+D_{1}\right) \ln \frac{M}{N_{0}}}{h \cdot D_{1}(l-d)}+\frac{M \mu^{2}\left(K_{2}+D_{2}\right) \ln \frac{M}{N_{0}}}{h \cdot D_{2}(l-d)},
$$

which implies that

$$
\left[\frac{h \cdot D_{1} \ln \frac{M}{N_{0}}}{K_{1}+D_{1}}-(l-d)\right]\left[\frac{h \cdot D_{2} \ln \frac{M}{N_{0}}}{K_{2}+D_{2}}-(l-d)\right] \geq 3(l-d)^{2} .
$$

Thus, it is clear that for the treatment intensity

$$
D_{1} \geq \frac{K_{1}(l-d)}{h+d-l} \quad \text { and } \quad D_{2} \geq \frac{K_{2}(l-d)}{h+d-l},
$$

we have $R_{2}^{\uparrow}(t)>R_{2}^{\downarrow}(t)$. This result holds under the assumption that $\frac{M}{N_{0}} \geq e^{1+\sqrt{3}}$. Now, we can see that the pre-treatment phase always plays a more important role.

## References

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[^0]:    Remark: " - " in the above Table S. 1 denotes dimensionless unit.

