

Modeling Drug-Carrier Interaction in the Drug Release from Nanocarriers

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Supporting Information

1. A detailed procedure to obtain the solution to the linear system

A detailed procedure to obtain the solution to the linear system of the first-order differential equations (2) and (3) is provided.

$$\frac{dc_F}{dt} = -(k_S + k_{on})c_F + k_{off}c_A, \quad (2)$$

$$\frac{dc_A}{dt} = k_{on}c_F - k_{off}c_A, \quad (3)$$

First, we normalize c_F and c_A using c_0 , and rewrite equations (2) and (3) in the matrix form,

$$\frac{dC}{dt} = -A \cdot C \quad (S1)$$

where

$$\mathbf{C} = \begin{pmatrix} c_F/c_0 \\ c_A/c_0 \end{pmatrix}, \quad \mathbf{A} = \begin{pmatrix} k_S + k_{\text{on}} & -k_{\text{off}} \\ -k_{\text{on}} & k_{\text{off}} \end{pmatrix}$$

Assuming $\mathbf{C} = \mathbf{b}e^{-\lambda t}$, where \mathbf{b} is the coefficient vector, equation (S1) is reduced to

$$(\mathbf{A} - \lambda \mathbf{I}) \cdot \mathbf{b} = 0 \quad (\text{S2})$$

Clearly, λ and \mathbf{b} are the eigenvalue and eigenvector of \mathbf{A} , respectively. The characteristic equation of the eigenvalue problem is

$$(k_S + k_{\text{on}} - \lambda)(k_{\text{off}} - \lambda) - k_{\text{on}}k_{\text{off}} = 0, \quad (\text{S3})$$

or

$$\lambda^2 - (k_S + k_{\text{on}} + k_{\text{off}})\lambda + k_S k_{\text{off}} = 0. \quad (\text{S4})$$

By solving equation (S4), one may obtain λ_1 and λ_2 as follows

$$\lambda_{1,2} = [k_S + k_{\text{on}} + k_{\text{off}} \pm \sqrt{(k_S + k_{\text{on}} + k_{\text{off}})^2 - 4k_S k_{\text{off}}}] / 2. \quad (\text{S5})$$

Using equation (S2), the eigenvectors corresponding to λ_1 and λ_2 can be obtained

$$\mathbf{b}_1 = \alpha_1 \begin{pmatrix} k_{\text{off}} \\ k_S + k_{\text{on}} - \lambda_1 \end{pmatrix}, \quad \mathbf{b}_2 = \alpha_2 \begin{pmatrix} k_{\text{off}} \\ k_S + k_{\text{on}} - \lambda_2 \end{pmatrix} \quad (\text{S6})$$

Here α_1 and α_2 are two non-zero coefficients. Accordingly, the solution to equation (S1) is

$$\mathbf{C} = \mathbf{b}_1 e^{-\lambda_1 t} + \mathbf{b}_2 e^{-\lambda_2 t} \quad (\text{S7})$$

Equivalently,

$$c_F/c_0 = \alpha_1 k_{\text{off}} e^{-\lambda_1 t} + \alpha_2 k_{\text{off}} e^{-\lambda_2 t} \quad (\text{S8})$$

$$c_A/c_0 = \alpha_1 (k_S + k_{\text{on}} - \lambda_1) e^{-\lambda_1 t} + \alpha_2 (k_S + k_{\text{on}} - \lambda_2) e^{-\lambda_2 t} \quad (\text{S9})$$

Using the initial conditions $c_F(0)/c_0 = k_{\text{off}}/(k_{\text{on}} + k_{\text{off}})$ and $c_A(0)/c_0 = k_{\text{on}}/(k_{\text{on}} + k_{\text{off}})$, α_1 and α_2 are determined to be

$$\alpha_1 = \frac{k_S - \lambda_2}{(k_{\text{on}} + k_{\text{off}})(\lambda_1 - \lambda_2)} \quad (\text{S10})$$

$$\alpha_2 = \frac{\lambda_1 - k_S}{(k_{\text{on}} + k_{\text{off}})(\lambda_1 - \lambda_2)} \quad (\text{S11})$$

Bringing α_1 and α_2 into equations (S8) and (S9), one may obtain the solution provided in the main text

$$\frac{c_F(t)}{c_0} = \frac{k_{\text{off}}}{k_{\text{on}} + k_{\text{off}}} \left[\frac{k_S - \lambda_2}{\lambda_1 - \lambda_2} e^{-\lambda_1 t} + \frac{\lambda_1 - k_S}{\lambda_1 - \lambda_2} e^{-\lambda_2 t} \right] \quad (4)$$

$$\frac{c_A(t)}{c_0} = \frac{k_{\text{on}}}{k_{\text{on}} + k_{\text{off}}} \left[\frac{-\lambda_2}{\lambda_1 - \lambda_2} e^{-\lambda_1 t} + \frac{\lambda_1}{\lambda_1 - \lambda_2} e^{-\lambda_2 t} \right]. \quad (5)$$

2. A detailed procedure to determine model parameters

We have developed a simple procedure to estimate each model parameter, which is further optimized using a customer code in Matlab 7.0. Fig. 3A is used as an example to illustrate this process.

2.1. Estimate ΔG

We begin with estimating ΔG . The magnitude of the initial release can be approximated using Eq. 8,

$$\frac{k_{\text{off}}}{k_{\text{on}} + k_{\text{off}}} = \frac{1}{1 + \exp(-\Delta G/k_{BT})} \quad (\text{S12})$$

In Fig. 3A, The magnitude of the initial release is estimated to be 0.45, leading to

$$\frac{1}{1 + \exp(-\Delta G/k_{BT})} = 0.45 \quad (\text{S13})$$

As a result, $\Delta G = -0.83$.

2.2. Estimate k_S

When t approaches to 0, Eq.6 in the main text can be expanded using Taylor expansion. Truncating the high-order terms, Eq.6 is reduced to,

$$\frac{M_t}{M_0} = \frac{\lambda_1 \lambda_2}{k_{\text{on}} + k_{\text{off}}} t = \frac{k_S k_{\text{off}}}{k_{\text{on}} + k_{\text{off}}} t \quad (\text{S14})$$

Substituting Eq. S12 into Eq. S14 yields:

$$\frac{M_t}{M_0} = \frac{k_S k_{\text{off}}}{k_{\text{on}} + k_{\text{off}}} t = \frac{k_S}{1 + \exp(-\Delta G/k_B T)} t \quad (\text{S15})$$

Eq. S16 suggests that the initial release can be approximated by a linear function. Moreover, the release rate that is equivalent to the slope of the linear function is $k_S / (1 + \exp(-\Delta G/k_B T))$. Noting Eq. S16 is valid at small t . In Fig. 3A, the release rate is estimated to be 0.2 at $t \approx 0$, resulting in,

$$k_S = 0.2(1 + \exp(-\Delta G/k_B T)) \quad (\text{S16})$$

We thus obtain $k_S = 0.44$.

2.3. Estimate k_{off}

We estimate k_{off} by rearranging Eq.6

$$\ln \left(1 - \frac{M_t}{M} \right) = -\ln \left(\frac{\lambda_2(k_S - \lambda_2)}{(k_{\text{on}} + k_{\text{off}})(\lambda_1 - \lambda_2)} e^{-\lambda_1 t} + \frac{\lambda_1(\lambda_1 - k_S)}{(k_{\text{on}} + k_{\text{off}})(\lambda_1 - \lambda_2)} e^{-\lambda_2 t} \right) \quad (\text{S17})$$

The first term on the right-hand side of Eq. S17 represents the residual of the initial burst release, and the second term represents the residual of the sustained release. When t is large enough, the residual of the initial burst release can be neglected. Then, taking natural logarithm on the both sides of Eq.S17, one may have

$$\ln \left(1 - \frac{M_t}{M} \right) = \ln \left(\frac{\lambda_1(\lambda_1 - k_S)}{(k_{\text{on}} + k_{\text{off}})(\lambda_1 - \lambda_2)} \right) - \lambda_2 t \quad (\text{S18})$$

Eq. S18 provides a mean to estimate λ_2 using the slope of $\ln(1 - M_t/M_0)$. In Fig. 3A, the slope of $\ln(1 - M_t/M_0)$ that equals to λ_2 is estimated to be 0.0094. Bringing it into Eq. S5, we obtained $k_{\text{off}} = 0.01$.

Therefore, the three model parameters are estimated to be $k_S = 0.44$, $k_{\text{off}} = 0.01$, and $\Delta G = -0.83$, respectively. Using these estimations, a Matlab optimization returns $k_S = 0.494$, $k_{\text{off}} = 0.0136$, and $\Delta G = -0.968$.

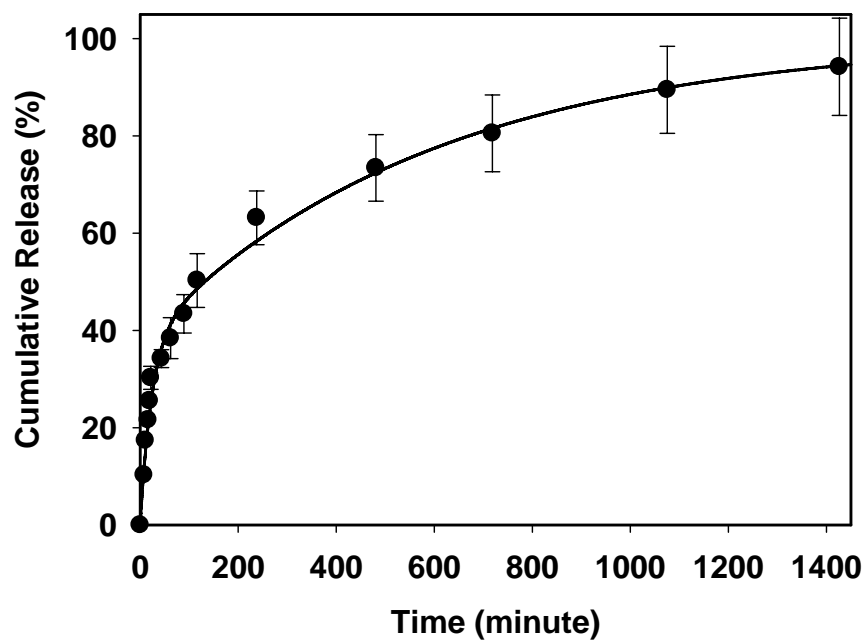


Fig. S1. The mode fit to telmisartan (TEL) release from mesoporous silica nanoparticle (MSNP). Solid line represents the model prediction. Parameters used for the simulation are: $k_S = 0.04 \text{ minute}^{-1}$, $k_{\text{off}} = 0.018 \text{ minute}^{-1}$, and $\Delta G = -1.2 \times 10^{-21} \text{ J}$.