SUPPLEMENTARY MATERIALS

Supplementary Methods

Theory

The terms in the equations are summarized in Table 1.

Creatine in a single erythrocyte

The rate of diffusion of creatine would be proportional to the concentration of creatine in the cell. We assume that the transporter activity obeys an exponential function; *i.e.* transporters diminish randomly (proportional to the number of the transporter) and are not renewed due to lack of nucleus.

$$\frac{dCr}{dt} = -\lambda_1 Cr(t) + Be^{-\lambda_2 t}$$
(1)

where Cr(t) denotes creatine concentration in a *t*-dayold erythrocyte, λ_1 denotes rate constant of creatine diffusion. $Be^{-\lambda_2 t}$ (B > 0) is creatine transporter activity. This differential equation can be solved analytically if $\lambda_1 \neq \lambda_2$.

$$\frac{d(Cr - \frac{B}{\lambda_1 - \lambda_2}e^{-\lambda_2 t})}{dt} = -\lambda_1 Cr(t) + \frac{B\lambda_1}{\lambda_1 - \lambda_2}e^{-\lambda_2 t}$$
(2)

$$Cr(t) = Ae^{-\lambda_1 t} + \frac{B}{\lambda_1 - \lambda_2}e^{-\lambda_2 t}$$
(3)

where A denotes an integral constant.

Cr(t) is a sum of two exponential functions, it can be treated as a bi-exponential function or a monoexponential function within a range of interest.

The fate of Cr(t) is dependent on whether $\lambda_1 - \lambda_2 > 0$ or not. If $\lambda_1 - \lambda_2 < 0$ and C'(0) > 0, Cr(t) has a peak when t > 0 (Figure 1 *orange line*). It can be treated as a mono-exponential function after the second term is negligible. However, as EC monotonically and rapidly decreases after birth of the erythrocytes [1], we can exclude this condition.

If $\lambda_1 - \lambda_2 < 0$ and $C'(0) \le 0$, Cr(t) decreases monotonically. As $\lambda_2 > \lambda_1$, $e^{-\lambda_2 t}$ decreases more rapidly. Moreover, the second negative term must be small, considering that C'(0) < 0 yields $\frac{B}{\lambda_2 - \lambda_1} < \frac{\lambda_1}{\lambda_2} A$. Therefore,

Table 1. Terms used in the text.

Term	Definition	Representative value
Cr(t)	creatine concentration in a	
	t-day-old erythrocyte	
λ_1	rate constant for creatine diffusion	
λ_2	rate constant for decline in creatine transporter	
λ	substitute for λ_1 or λ_2	
A, B, C, D	constants	
EC	mean erythrocyte creatine concentration	1.4 µmol/g Hb
α	a parameter of gamma distribution	25.59
β	a parameter of gamma distribution	5.59
p(t)	probability density function of RBC death	
R_0	erythrocyte production rate	-/day
R(t)	the number of erythrocytes at <i>t</i> days after birth	
$M_{\rm RBC}$	mean red blood cell age	60 days
RBC	number of erythrocytes	_

the second negative term can be considered negligible comparing the first term.

If $\lambda_1 - \lambda_2 > 0$, Cr(t) is a bi-exponential function. The logarithm of a bi-exponential function can be expressed by a bent line (Figure 1B *blue line*), because a large *t* makes one term negligible, while small $t (\rightarrow -\infty)$ makes the other term negligible.



Figure 1. Examples of Cr(t)**.** (A) normal scale; (B) logarithmic scale. *Orange line:* $\lambda_1 - \lambda_2 < 0$ and C'(0) > 0, the function has a peak. *Blue line:* $\lambda_1 - \lambda_2 > 0$, Cr(t) is a bi-exponential function. The logarithm of a bi-exponential function can be expressed by a bent line.

Problem applying single cell model to erythrocyte population

Equation (3) itself would not be suitable to obtain mean erythrocyte age, because EC is not measured from a single cell.

$$EC = \left(\sum_{i}^{n} Cr(t_{i})\right) / n$$

$$= A\left[\sum_{i}^{n} e^{-\lambda_{1}t_{i}}\right] / n + \frac{B}{\lambda_{1} - \lambda_{2}}\left[\sum_{i}^{n} e^{-\lambda_{2}t_{i}}\right] / n$$
(4)

 $y = e^{-\lambda x}$ is downward convex. The centroid of n polygonal, $(t_i, e^{-\lambda t_i})$ is over the curve of $y = e^{-\lambda x}$.

$$\left(\sum_{i}^{n} e^{-\lambda t_{i}}\right) / n \ge \exp\left(-\lambda \frac{\sum_{i}^{n} t_{i}}{n}\right)$$
(5)

Therefore, when $\lambda_1 - \lambda_2 > 0$,

$$EC \ge Ae^{-\lambda_1 M_{RBC}} + \frac{B}{\lambda_1 - \lambda_2} e^{-\lambda_2 M_{RBC}}$$
(6)

Erythrocyte lifespan

Kameyama et al. [2] have recently calculated RBC lifespan based on the probability density function p(t) of RBC death proposed by Shrestha et al. [3].

$$p(t) = \frac{1}{\Gamma(\alpha)\beta^{\alpha}} t^{\alpha - 1} e^{-t/\beta}$$
(7)

 Γ denotes the Euler gamma function.

$$\Gamma(\alpha) = \int_0^\infty x^{\alpha - 1} e^{-x} dx \tag{8}$$

The number of erythrocytes (*RBC*) and mean erythrocyte age (M_{RBC}) was calculated. (See Kameyama et al. [2] for details.)

$$RBC = R_0 \int_0^\infty tp(t)dt = R_0 \alpha \beta \tag{9}$$

$$M_{RBC} = \frac{(\alpha + 1)\beta}{2} \tag{10}$$

Creatine model

The number of *t*-day-old erythrocyte is R(t). Each RBC has $Ae^{-\lambda_1 t} + \frac{B}{\lambda_1 - \lambda_2}e^{-\lambda_2 t}$ creatine. Therefore, mean creatine concentration, *EC* can be described as follows:

$$EC = \int_{0}^{\infty} R(t)$$

$$\times \left[Ae^{-\lambda_{1}t} + \frac{B}{\lambda_{1} - \lambda_{2}} e^{-\lambda_{2}t} \right] dt / RBC$$

$$(11)$$

$$\int_{0}^{\infty} R(t) \times e^{-\lambda t} dt$$

$$= \left[R(t) \frac{e^{-\lambda t}}{-\lambda} \right]_{0}^{\infty} - \int_{0}^{\infty} R'(t) \frac{e^{-\lambda t}}{-\lambda} dt$$

$$(12)$$

$$= \frac{R_{0}}{\lambda} - \frac{R_{0}}{\lambda} \int_{0}^{\infty} p(t) e^{-\lambda t} dt$$

$$\int_{0}^{\infty} p(t) e^{-\lambda t} dt$$

$$= \frac{1}{\Gamma(\alpha)\beta^{\alpha}} \int_{0}^{\infty} t^{\alpha - 1} e^{-(1/\beta + \lambda)t} dt$$

$$(13)$$

Hence, EC can be expressed as follows:

$$EC = \frac{A}{\lambda_{1}\alpha\beta} \left(1 - \frac{1}{(1+\beta\lambda_{1})^{\alpha}} \right) + \frac{B}{\lambda_{1} - \lambda_{2}} \frac{1}{\lambda_{2}\alpha\beta} \left(1 - \frac{1}{(1+\beta\lambda_{2})^{\alpha}} \right)$$
(14)

Approximation of the derived relationship

The Taylor expansion provides the following equation:

$$(1+\beta\lambda)^{-\alpha} \approx 1-\alpha\beta\lambda + \frac{\alpha(\alpha+1)}{2}(\beta\lambda)^{2} - \frac{\alpha(\alpha+1)(\alpha+2)}{6}(\beta\lambda)^{3} + \dots$$
(15)

$$\frac{1}{\lambda\alpha\beta} \left(1 - \frac{1}{(1+\beta\lambda)^{\alpha}} \right)$$

$$\approx \frac{1}{\lambda\alpha\beta} \left(\alpha\beta\lambda - \frac{\alpha(\alpha+1)}{2} (\beta\lambda)^{2} + \frac{\alpha(\alpha+1)(\alpha+2)}{6} (\beta\lambda)^{3} - \ldots \right)$$

$$= 1 - \frac{\beta(\alpha+1)}{2} \lambda + \frac{(\alpha+1)(\alpha+2)}{6} (\beta\lambda)^{2} - \ldots$$
(16)

As $\alpha >> 1$,

$$\frac{1}{\lambda\alpha\beta} \left(1 - \frac{1}{\left(1 + \beta\lambda\right)^{\alpha}} \right) \approx 1 - \lambda M_{RBC} + \frac{2}{3} \lambda^2 M_{RBC}^2 - \dots (17)$$

Thus, $\frac{1}{\lambda\alpha\beta} \left(1 - \frac{1}{(1+\beta\lambda)^{\alpha}}\right)$ can be described approximately as a function of M_{RBC} . Although how α and β vary when M_{RBC} decreases or increases cannot be determined, this implies that the function $\frac{1}{\lambda\alpha\beta} \left(1 - \frac{1}{(1+\beta\lambda)^{\alpha}}\right)$ would not be greatly affected by α and β if $M_{RBC} = (\alpha + 1)\beta/2$ is satisfied. This can be confirmed numerically (Figure 2A). Calculations based on the assumption that β was constant and β was proportionate to α showed similar results. Therefore, β can be considered a constant.

Below, $\frac{1-e^{-x}}{x}$ was approximated to be De^{-Cx} , as the shape of the graphs were similar (Figure 2B).

$$C = \frac{-x_0 e^{-x_0} + (1 - e^{-x_0})}{x_0 (1 - e^{-x_0})}, \qquad D = \frac{1 - e^{-x_0}}{x_0 e^{-Cx_0}}$$
(18)

provides the same value of the two function and the differential function at $x = x_0$. Figure 2B visualizes the approximation.

As $\frac{1-e^{-x}}{x}$ can be treated as an exponential function, it follows that $\frac{1}{\lambda\beta x}\left(1-\frac{1}{(1+\beta\lambda)^x}\right)$ can also be treated as an exponential function, because

$$\frac{1}{\lambda\beta\alpha} \left(1 - \frac{1}{(1+\beta\lambda)^{\alpha}} \right)$$
$$= \frac{\log_{e}(1+\beta\lambda)}{\lambda\beta} \frac{1 - \exp(-\log_{e}(1+\beta\lambda)\alpha)}{\log_{e}(1+\beta\lambda)\alpha}$$
(19)
$$\approx \frac{\log_{e}(1+\beta\lambda)}{\lambda\beta} D \exp\left(-C \log_{e}(1+\beta\lambda) \left(\frac{2M_{RBC}}{\beta} - 1 \right) \right)$$

C, D can be estimated by equation (18). To obtain an approximation around M_{RBC0} , x_0 should be as following:

$$x_0 = \log_e (1 + \beta \lambda) \left(\frac{2M_{RBC0}}{\beta} - 1 \right)$$
(20)

In conclusion, EC can be expressed approximately as a (bi-)exponential function of $M_{\rm \tiny RBC}$.



Figure 2. (A) Relationship between M_{RBC} and $\frac{1}{\lambda\alpha\beta} \left(1 - \frac{1}{(1+\beta\lambda)^{\alpha}} \right)$. The two condition of $\frac{1}{\lambda\alpha\beta} \left(1 - \frac{1}{(1+\beta\lambda)^{\alpha}} \right)$ (constant β and $\beta \propto \alpha$) showed similar results. Note that $\frac{1}{\lambda\alpha\beta} \left(1 - \frac{1}{(1+\beta\lambda)^{\alpha}} \right)$ is consistently larger than $e^{-\lambda M_{RBC}}$. $D'e^{-C'M_{RBC}}$ is an approximation at $M_{RBC} = 36.33$ with exponential function by equation (19). (B) $\frac{1-e^{-x}}{x}$ and its approximation, De^{-Cx} (equation (18)) when $x_0 = 2$. (C, D) Semi-log scales of (A, B) show that $\frac{1-e^{-x}}{x}$ and $\frac{1}{\lambda\alpha\beta} \left(1 - \frac{1}{(1+\beta\lambda)^{\alpha}} \right)$ can be treated as an exponential function.

Supplementary References

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